

Pade-Shammas Polynomial Approximations

Part IB

By

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INTRODUCTION

This study looks at the collaboration of two mathematical frameworks—the Pade polynomials and the Shammas Polynomials. The study combines both frameworks to approximate several common functions. This study uses x and y data (calculated from the functions being approximated) without any data transformation. Please do not be intimidated by the page count of this study. Most of the pages contain

calculation results. You can quickly browse through this document. You are probably best served by focusing on the tables that summarize the results of fitting various functions with Pade-Shammas polynomials.

The Pade Polynomials

Pade polynomials are polynomial ratios defined as:

$$y = P_{m,n}(x) = \frac{Q_m(x)}{D_n(x)} \quad (1)$$

Where $Q_m(x)$ is defined as:

$$Q_m(x) = a_0 + \sum_{i=1}^m a_i x^i \quad (2)$$

And $D_n(x)$ is defined as:

$$D_n(x) = 1 + \sum_{i=1}^n b_i x^i \quad (3)$$

The Pade polynomials have more flexibility than ordinary legacy polynomials. The trick is to find the optimum values for m and n which are the orders of the polynomials $Q_m(x)$ and $D_n(x)$. This study uses swarm optimization to determine these optimum values within the range of (2, 7). The general multiple regression model used to fit Pade polynomials is:

$$y = a_0 + \sum_{i=1}^m a_i x^i - \sum_{i=1}^n b_i y x^i \quad (4)$$

Or,

$$y = a_0 + \sum_{i=1}^m a_i x^i + \sum_{i=1}^n c_i y x^i \quad (5)$$

Where $c(i) = -b(i)$. This study uses the model in equation 5.

The Shammas Polynomials

In the HHC 2008 conference in Corvallis, Oregon, I introduced the Shammas polynomials as polynomials with non-integer powers. I explained that the powers of such polynomial change using some math expressions that involve the polynomial term sequence number. Moreover, the study explained that such expressions must never produce the same powers of two or more terms. In the 2008 presentation I showed that the Shammas polynomials can have multiple terms with a smaller power range compared to regular polynomials. This feature allows better curve fitting with Shammas polynomials that avoids the instability of raising large or small numbers to high powers. If you are not familiar with Shammas

polynomials, then you can download my 2008 presentation using the following link:

<http://www.namirshammas.com/NEW/ShammasPolynomials.pdf>.

A Shammas polynomial has the following general form:

$$y = a_0 + \sum_{i=1}^n a_i x^{gx(i,A,B,C)} \quad (6)$$

Where A, B, and C are fixed constants. Using C is optional. The function $gx(i, A, B, C)$ supplies the math expressions used to calculate the powers of the Shammas polynomial. The powers can be integers. In most cases, the powers are non-integers.

The Pade-Shammas Polynomials

The Pade-Shammas polynomials combine Pade and Shammas polynomials concepts. They have the following multiple regression general form:

$$y = a_0 + \sum_{i=1}^m a_i x^{gx1(i,A,B,C)} - \sum_{i=1}^n b_i y x^{gx2(i,A,B,C)} \quad (7)$$

Notice that the integral powers in normal Pade polynomials are replaced with power functions $gx1()$ and $gx2()$ that can be identical or different. The latter type gives the Pade-Shammas polynomials, in general, more flexibility in fitting a given set of (x, y) data. This statement is truer when the values of A, B, and C (when used) are different for the $gx1()$ and $gx2()$ functions. Part II of this study uses this scheme to give more flexibility to Pade-Shammas polynomials. Also note that Part IB conducts the Pade-Shammas polynomial fitting on the same functions but without normalizing the x and y data.

When using the Pade-Shammas polynomials to calculate new values employee the following equation:

$$y = [a_0 + \sum_{i=1}^m a_i x^{gx1(i,A,B,C)}] / [1 + \sum_{i=1}^n b_i x^{gx2(i,A,B,C)}] \quad (8)$$

 I am appending my last name to the name Pade in an effort to make sure it is a unique combined name. Using a more general descriptive name risks conflicts with other mathematicians that may use the same descriptive name for different types of Pade polynomials. I certainly would like to avoid unpleasant communications with these fine mathematicians!

This study looks at Pade-Shammas polynomials used to approximate common functions that include:

- Trigonometric functions and their inverses.
- Hyperbolic functions.
- Logarithmic functions.
- Exponential functions.
- Bessel functions $J_0(x)$ to $J_5(x)$.
- The sine and cosine integrals.
- The Fresnel sine and cosine.
- Inverse student-t functions.
- The common logarithm of the gamma function.
- The digamma function.
- The trigamma function.

The digamma function is defined as:

$$\psi(x) = \Gamma'(x) / \Gamma(x) = \frac{d \ln (\Gamma(x))}{dx} \quad (9)$$

The following Matlab function implements the code for the digamma function:

```
function y = digamma(x)
%DIGAMMA Summary of this function goes here
% Detailed explanation goes here
h = 0.001;
fp = gammaln(x+h);
fm = gammaln(x-h);
y = (fp - fm)/2/h;
end
```

The above implementation of the digamma function was suggested by Albert Chan, a member of the hp museum web site, in a post he wrote on that site. The above code gives slightly more accurate results than the expression $(\text{gamma}(x+h) - \text{gamma}(x-h))/(2*h)/\text{gamma}(x)$.

The trigamma function is defined as:

$$\psi_1(x) = \frac{d^2}{dx^2} \ln(\Gamma(x)) \quad (10)$$

The following Matlab function implements the code for the trigamma function:

```
function y = trigamma(x)
%DIGAMMA Summary of this function goes here
```

```
% Detailed explanation goes here
h = 0.001;
fp = gamma1n(x+h);
fm = gamma1n(x-h);
f0 = gamma1n(x);
y = (fp -2*f0 + fm)/h/h;
end
```

The study works with Pade-Shammas Polynomials that use the following ten equations to calculate the powers of Shammas polynomial terms:

$$\text{Power} = A + B*i \quad (11)$$

$$\text{Power} = A + B/i \quad (12)$$

$$\text{Power} = A+B*\text{sqrt}(i) \quad (13)$$

$$\text{Power} = A+B*\text{sqrt}(i+C) \quad (14)$$

$$\text{Power} = A+B*i+C*(i-1) \quad (15)$$

$$\text{Power} = A+B*i-C*(i-1) \quad (16)$$

$$\text{Power} = A+B*i+C*\text{sqrt}(i-1) \quad (17)$$

$$\text{Power} = A+B*\text{sqrt}(i)+C*(i-1) \quad (18)$$

$$\text{Power} = A+B*\text{sqrt}(i)+C*\text{sqrt}(i-1) \quad (19)$$

$$\text{Power} = A + B * [\log_{10}(i)]^4 \quad (20)$$

Where A, B, and C are constants, and i a variable that represents the term number in a Pade-Shammas polynomial. The last equation introduces a new power expression not used in the Shammas Polynomial approximation study.

There are several choices that we can use to determine the best Pade-Shammas polynomial orders and the best values for parameters A, B, and C. The best one is to use an optimization function to determine the best Pade-Shammas polynomial order and the best values for parameters A, B, and C. The function would use ranges for the polynomial orders as well as the values for A, B, C.

The study uses the Matlab particle swarm optimization function to select the best Shammas polynomial orders (for $Q_n(x)$ and $D_m(x)$) along with the best values for A, B, and C.

 Please note that I use the following model to perform the multiple regression:

$$y = a_0 + \sum_{i=1}^m a_i x^{gx1(i,A,B,C)} + \sum_{i=1}^n b_i y x^{gx2(i,A,B,C)} \quad (20)$$

I added the second set of summations instead of subtracting them (as in equation 6). Therefore, when you use the regression coefficients, use the following equation:

$$y = [a_0 + \sum_{i=1}^m a_i x^{gx1(i,A,B,C)}] / [1 - \sum_{i=1}^n b_i x^{gx2(i,A,B,C)}] \quad (21)$$

Notice the minus sign after the constant 1 in equation 21. I use equation 21 in the Matlab functions PadeShamPoly2() and PadeShamPoly3() that I present in the next section.

 The approximations that I obtain are typically for a defined and suitable interval. It is your responsibility to implement expanded versions of the approximation functions that take wider ranges of arguments and map them onto the interval used. For example, given that my approximation for $\log_{10}(x)$ uses the range (1, 10), to calculate $\log_{10}(235)$ use:

$$\log_{10}(235) =$$

$$\log_{10}(2.35 * 100) =$$

$$\log_{10}(2.35) + \log_{10}(100) =$$

$$\log_{10}(2.35) + 2$$

The argument of the $\log_{10}(x)$ function in the last line falls in the interval (1, 10).

MATLAB CODE

The algorithm in this study uses particle swarm optimization to obtain the best values for parameters A and B in a prespecified range of Pade-Shammas polynomial orders for $Q_n(x)$ and $D_m(x)$.

This study uses the function PadeShamPoly2 () to perform various Pade-Shammas polynomial curve fitting:

```
function PadeShamPoly2(fx,gx,xRange,Lb,Ub,runNum,sFxName,diaryFilename)
%PADESHAMPOLY2 Summary of this function goes here
% Detailed explanation goes here
clc
global bDeleteIfExists
global bUseDiary
global xdata
global ydata
global orderA
```

```

global orderB
global ggx

warning('off','all')
if isempty(sFxName)
    sFxName = getFuncName(fx);
end
xdata = xRange';
ydata = xdata;
for i=1:length(xdata)
    ydata(i)=fx(xdata(i));
end
ggx = gx;

xmin = min(xdata);
xmax = max(xdata);
fprintf('Fitting %s in range (%f, %f)\n', sFxName, xmin ,xmax);

options = optimoptions('particleswarm', 'Display', 'iter');
[x,psAICc] = particleswarm(@optimFunc,4,Lb,Ub,options);
A = x(1);
B = x(2);
orderA = round(x(3));
orderB = round(x(4));
if bUseDiary
    diaryFilename = strrep(diaryFilename, ".txt", strcat("_",
num2str(orderA),"_", num2str(orderB),"_run", num2str(runNum),".txt"));
    if exist(diaryFilename, 'file')==2
        if bDeleteIfExists
            delete(diaryFilename);
        else
            return;
        end
    end
end
X = [];
for i=1:orderA
    xs = xdata.^gx(i,A,B);
    X = [X;xs'];
end
for i=1:orderB
    xs = ydata.*(xdata.^gx(i,A,B));
    X = [X;xs'];
end
X = X';
lm = fitlm(X,ydata);
if bUseDiary
    diary(diaryFilename)
end
fprintf('Fitting %s in range (%f, %f)\n', sFxName, xmin ,xmax);
sp = getFuncName(gx);
fprintf('Pade-Shammas polynomial power is %s\n', sp);
format long
disp(lm);
anova = anova(lm,'summary');
disp(anova);
format short

```

```

fprintf('A = %f, B = %f\n', A, B);
fprintf('order Q(x) = %f, order D(x) = %f\n', orderA, orderB);
fprintf('Xmin = %f and Xmax = %f\n', xmin, xmax);
fprintf('Ymin = %f and Ymax = %f\n', ymin, ymax);
lstPwrA = zeros(orderA,1);
for i=1:orderA
    lstPwrA(i) = gx(i,A,B);
end
lstPwrB = zeros(orderB,1);
for i=1:orderB
    lstPwrB(i) = gx(i,A,B);
end
fprintf('List of powers for Q(x): [');
for i=1:orderA-1
    fprintf('%f, ', lstPwrA(i));
end
fprintf('%f]\n', lstPwrA(orderA));
fprintf('List of powers for D(x): [');
for i=1:orderB-1
    fprintf('%f, ', lstPwrB(i));
end
fprintf('%f]\n', lstPwrB(orderB));
fprintf('Fitting %s in range (%f, %f)\n', sFxName, xmin ,xmax);
n = length(xdata);
sumsqr = 0;
for i=1:n
    qx = lm.Coefficients{1,1};
    for j=2:orderA+1
        qx = qx + lm.Coefficients{j,1} * xdata(i)^lstPwrA(j-1);
    end
    dx = 1;
    k = 1;
    for j=orderA+2:orderA+1+orderB
        dx = dx - lm.Coefficients{j,1} * xdata(i)^lstPwrB(k);
        k = k + 1;
    end
    yc = qx/dx;
    sumsqr = sumsqr + (ydata(i) - yc)^2;
end
k = orderA + orderB + 1;
fprintf('MSS of errors squared = %e\n', sqrt(sumsqr)/n);
fprintf('Corrected MSS of errors squared = %e\n', sqrt(sumsqr)/n*sqrt(2));
fprintf('R-Squared = %12.8f\n', lm.Rsquared.Ordinary);
fprintf('R-Squared Adjusted = %12.8f\n', lm.Rsquared.Adjusted);
AIC = lm.ModelCriterion.AIC;
AICc = AIC + 2*k*(1 + (k+1)/(n-k-1));
fprintf('Particle swarm AICc = %e\n', psAICc);
fprintf('AIC = %e\n', AIC);
fprintf('AICc = %e\n', AICc);

if bUseDiary
    diary off
end
end

function AICc = optimFunc(x)
    global xdata

```

```

global ydata
global orderA
global orderB
global ggx

A = x(1);
B = x(2);
orderA = round(x(3));
orderB = round(x(4));
X = [];
for i=1:orderA
    xs = xdata.^ggx(i,A,B);
    X = [X;xs'];
end
for i=1:orderB
    xs = ydata.*(xdata.^ggx(i,A,B));
    X = [X;xs'];
end
X = X';
lm = fitlm(X,ydata);

n = length(xdata);
k = orderA + orderB + 1;
AIC = lm.ModelCriterion.AIC;
AICc = AIC + 2*k*(1 + (k+1)/(n-k-1));
if isinf(AICc), AICc = -1e+99; end
end

function sFx = getFuncName(fx)
sFx = func2str(fx);
if sFx(1:2)=="@"
    i = strfind(sFx,"");
    sFx = sFx(i(1)+1:end);
elseif sFx(1)=="@"
    sFx = strcat(sFx(2:end),".m");
else
    % return sFx as is
end
end
end

```

The parameters of function PadeShamPoly2 () are:

- The parameter fx is the handle (or inline function) for the function being approximated. An example is @(x)cos(x) which also shows the **recommended format** for the argument of parameter fx.
- The parameter gx is the handle (or inline function) for the function that calculates the powers of the Shammas polynomial. An example is @(i,A,B)A+B*sqrt(i) which also shows the **recommended format** for the argument of parameter gx.
- The parameter xRange is the array that specifies the minimum value, increment value, and maximum value for the range of approximation.

- The parameter Lb is is the array of lower limits for the parameters A and B, and the Shammas polynomial order. An example is [0.1 1 3].
- The parameter Ub is is the array of upper limits for the parameters A and B, and the Shammas polynomial order. An example is [1 4 5].
- The parameter runNum specifies the run number. The arguments for this parameter have nothing to do with the calculations and serve in fine tuning the name of the diary files, when used.
- The optional parameter sFxName is the name of the approximated function. An example is “cos(x)”.
- The parameter diaryFilename is the name of the diary file. An example is “cos_1.txt”.

The above listing performs the following tasks:

1. Initialize the data for the curve fitting. The function uses the global variables xdata and ydata to store the data for the polynomial fitting.
2. Store the handle of function gx in the global handle ggx.
3. Set the optimization options and then call the Matlab function particleswarm(). The function call returns the optimized values of A and B and the optimum value for *corrected* Akaike information criterion (AICc). The arguments for this function call are:
 - a. The handle to the local function optimFunc() that calculates the root mean sum of errors squared.
 - b. The number of optimized variables which is 2.
 - c. The lower and upper bounds arrays, Lb and Ub, respectively,
 - d. The optimization parameters for function particleswarm().
4. Retrieve the optimum values and perform a Shammas polynomial fit for the best values of A and B.
5. Display the results of the regression and its associated ANOVA table.
6. Display the list of Shammas polynomial powers.
7. Display the range of the approximated function.
8. Calculate and display the value of the mean square root of the sum of squared errors. This statistic serves as a check that the Shammas polynomial performs well in checking the training data.
9. Calculate and display the value of the *corrected* mean square root of the sum of squared errors. This value equals the mean square root of the sum of squared errors multiplied by the square root of the number of Shammas

polynomial parameters (i.e. A, B, and C, if used). In the case of PadeShamPoly2(), the function uses square root of 2.

10. Calculate and display the *corrected* Akaike information criterion. This statistic is calculated using:

$$AIC = n * \ln(SSE/n) + 2*k \quad (8)$$

$$AICc = AIC + 2*k*(k+1)/(n-k-1) \quad (9)$$

Where n is the number of observations, k is the total number of regression coefficients (including the intercept), and SSE is the sum of squared errors. The program obtains the value of AIC using `lm.ModelCriterion.AIC`. The program uses equation (3) to calculate the value for AICc.

11. Close the diary file, if one is used.

The function `optimFunc()` obtains the array x containing the current values of A and B, and the best Shammas polynomial order. The function calculates the transformed variables needed to perform a curve fit for a Shammas polynomial. This task calls the Matlab function `fitlm()`. The `optimFunc()` function returns the AICc as its result. I am using this statistic since the optimization is dealing with different Shammas polynomial orders and thus a varying number of polynomial coefficients.

The function `getFuncName()` returns a string-type function name given a handle of a function. The best way to take advantage of this function is to supply arguments like `@(x)cos(x)` and `@(x,A,B)A+B*sqrt(x)`. Such arguments allow the function to discard the part that declares the variable(s) and return the part that comes after the first closed parenthesis (e.g. `cos(x)` and `A+B*sqrt(x)` for the above examples). If you supply an argument like `@fx1` which refers to the file `fx1.m` that defines the function `fx1()` then the function `getFuncName()` returns `fx1.m`. This string value indicates that you are referencing a separate Matlab file that implements the code for `fx1`.

It is worth going over the code that estimates the values of y to calculate the sum of error squared:

```
n = length(xdata);
sumsq = 0;
for i=1:n
    qx = lm.Coefficients{1,1};
    for j=2:orderA+1
        qx = qx + lm.Coefficients{j,1} * xdata(i)^1stPwrA(j-1);
    end
```

```

dx = 1;
k = 1;
for j=orderA+2:orderA+1+orderB
    dx = dx - lm.Coefficients{j,1} * xdata(i)^lstPwrB(k);
    k = k + 1;
end
yc = qx/dx;
sumsqr = sumsqr + (ydata(i) - yc)^2;
end

```

The above code used the arrays `lstPwrA` and `lstPwrB` to store the powers used in the Shammas polynomials $Q_n(x)$ and $D_m(x)$, respectively. The first for loop calculate the value for Shammas polynomials $Q_n(x)$ using variable `qx`. Notice that the code initializes the value of `qx` using the intercept of the regression model. The loop accesses the powers of $Q_n(x)$ using `lstPwrA(j-1)`. The second for loop calculates the value for Shammas polynomials $D_m(x)$ using variable `dx`. Notice that the code initializes the value of `dx` using the constant 1. The loop accesses the powers of $Q_n(x)$ using `lstPwrB(k)`. Here `k` is initialized with the value 1 and is incremented in the for loop. Also noticed that the value of variable `dx` is decremented with the expression `lm.Coefficients{j,1} * xdata(i)^lstPwrB(k)`. I have explained the above code because you may need similar code to estimate other values of y given an array (or scalar) of x , as shown below:

```

n = length(your_xdata);
for i=1:n
    qx = lm.Coefficients{1,1};
    for j=2:orderA+1
        qx = qx + lm.Coefficients{j,1} * your_xdata(i)^lstPwrA(j-1);
    end
    dx = 1;
    k = 1;
    for j=orderA+2:orderA+1+orderB
        dx = dx - lm.Coefficients{j,1} * your_xdata(i)^lstPwrB(k);
        k = k + 1;
    end
    your_y(i) = qx/dx;
end

```

Under the current calculation scheme, note that, the function `PadeShamPoly2()` does not explicitly iterate over different values of A and B . It delegates that task to the Matlab function `particleswarm()`.

The study also uses the function `PadeShamPoly3()` to work with Shammas polynomial powers that use parameters A , B , and C :

```

function PadeShamPoly3(fx,gx,xRange,Lb,Ub,runNum,sFxName,diaryFilename)
%PADESHAMPOLY3 Summary of this function goes here
% Detailed explanation goes here
clc

```

```

global bDeleteIfExists
global bUseDiary
global xdata
global ydata
global orderA
global orderB
global ggx

warning('off','all')
if isempty(sFxName)
    sFxName = getFuncName(fx);
end
xdata = xRange';
ydata = xdata;
for i=1:length(xdata)
    ydata(i)=fx(xdata(i));
end
ggx = gx;

xmin = min(xdata);
xmax = max(xdata);
fprintf('Fitting %s in range (%f, %f)\n', sFxName, xmin ,xmax);

options = optimoptions('particleswarm', 'Display', 'iter');
[x,psAICc] = particleswarm(@optimFunc,5,Lb,Ub,options);
A = x(1);
B = x(2);
C = x(3);
orderA = round(x(4));
orderB = round(x(5));
if bUseDiary
    diaryFilename = strrep(diaryFilename, ".txt", strcat("_",
num2str(orderA), "_", num2str(orderB), "_run", num2str(runNum), ".txt"));
    if exist(diaryFilename, 'file')==2
        if bDeleteIfExists
            delete(diaryFilename);
        else
            return;
        end
    end
end
X = [];
for i=1:orderA
    xs = xdata.^gx(i,A,B,C);
    X = [X;xs'];
end
for i=1:orderB
    xs = ydata.*(xdata.^gx(i,A,B,C));
    X = [X;xs'];
end
X = X';
lm = fitlm(X,ydata);

if bUseDiary
    diary(diaryFilename)
end
fprintf('Fitting %s in range (%f, %f)\n', sFxName, xmin ,xmax);

```

```

sp = getFuncName(gx);
fprintf('Pade-Shammas polynomial power is %s\n', sp);
format long
disp(lm);
anova = anova(lm, 'summary');
disp(anova);
format short
fprintf('A = %f, B = %f, C = %f\n', A, B, C);
fprintf('order Q(x) = %f, order D(x) = %f\n', orderA, orderB);
fprintf('Xmin = %f and Xmax = %f\n', xmin, xmax);
fprintf('Ymin = %f and Ymax = %f\n', ymin, ymax);
lstPwrA = zeros(orderA,1);
for i=1:orderA
    lstPwrA(i) = gx(i,A,B,C);
end
lstPwrB = zeros(orderB,1);
for i=1:orderB
    lstPwrB(i) = gx(i,A,B,C);
end
fprintf('List of powers for Q(x): [');
for i=1:orderA-1
    fprintf('%f, ', lstPwrA(i));
end
fprintf('%f]\n', lstPwrA(orderA));
fprintf('List of powers for D(x): [');
for i=1:orderB-1
    fprintf('%f, ', lstPwrB(i));
end
fprintf('%f]\n', lstPwrB(orderB));
fprintf('Fitting %s in range (%f, %f)\n', sFxName, xmin, xmax);
n = length(xdata);
sumsq = 0;
for i=1:n
    qx = lm.Coefficients{1,1};
    for j=2:orderA+1
        qx = qx + lm.Coefficients{j,1} * xdata(i)^lstPwrA(j-1);
    end
    dx = 1;
    k = 1;
    for j=orderA+2:orderA+1+orderB
        dx = dx - lm.Coefficients{j,1} * xdata(i)^lstPwrB(k);
        k = k + 1;
    end
    yc = qx/dx;
    sumsq = sumsq + (ydata(i) - yc)^2;
end
k = orderA + orderB + 1;
fprintf('MSS of errors squared = %e\n', sqrt(sumsq)/n);
fprintf('Corrected MSS of errors squared = %e\n', sqrt(sumsq)/(n*sqrt(3)));
fprintf('R-Squared = %12.8f\n', lm.Rsquared.Ordinary);
fprintf('R-Squared Adjusted = %12.8f\n', lm.Rsquared.Adjusted);
AIC = lm.ModelCriterion.AIC;
AICc = AIC + 2*k*(1 + (k+1)/(n-k-1));
fprintf('Particle swarm AICc = %e\n', psAICc);
fprintf('AIC = %e\n', AIC);
fprintf('AICc = %e\n', AICc);

```

```

if bUseDiary
    diary off
end
end

function AICc = optimFunc(x)
    global xdata
    global ydata
    global orderA
    global orderB
    global ggx

    A = x(1);
    B = x(2);
    C = x(3);
    orderA = round(x(4));
    orderB = round(x(5));
    X = [];
    for i=1:orderA
        xs = xdata.^ggx(i,A,B,C);
        X = [X;xs'];
    end
    for i=1:orderB
        xs = ydata.*(xdata.^ggx(i,A,B,C));
        X = [X;xs'];
    end
    X = X';
    lm = fitlm(X,ydata);

    n = length(xdata);
    k = orderA + orderB + 1;
    AIC = lm.ModelCriterion.AIC;
    AICc = AIC + 2*k*(1 + (k+1)/(n-k-1));
    if isinf(AICc), AICc = -1e+99; end
end

function sFx = getFuncName(fx)
    sFx = func2str(fx);
    if sFx(1:2)=='@('
        i = strfind(sFx,')');
        sFx = sFx(i(1)+1:end);
    elseif sFx(1)=='@'
        sFx = strcat(sFx(2:end),'.m');
    else
        % return sFx as is
    end
end
end

```

The parameters of function PadeShamPoly3() are:

- The parameter fx is the handle (or inline function) for the function being approximated. An example is @(x)cos(x) which also shows the **recommended format** for the argument of parameter fx.

- The parameter gx is the handle (or inline function) for the function that calculates the powers of the Shammas polynomial. An example is `@(i,A,B,c)A+B*sqrt(i+C)` which also shows the **recommended format** for the argument of parameter gx.
- The parameter xRange is the array that specifies the minimum value, increment value, and maximum value for the range of approximation.
- The parameter Lb is is the array of lower limits for the parameters A, B, C and the Shammas polynomial order. An example is `[0.1 0 0 3]`.
- The parameter Ub is is the array of upper limits for the parameters A, B and C, and the Shammas polynomial order. An example is `[1 4 3 7]`.
- The parameter runNum specifies the run number. The arguments for this parameter have nothing to do with the calculations and serve in fine tuning the name of the diary files, when used.
- The optional parameter sFxName is the name of the approximated function. An example is “cos(x)”.
- The parameter diaryFilename is the name of the diary file. An example is “cos_1.txt”.

The above listing performs basically the same tasks as function PadeShamPoly2 () with the exception that it optimizes the parameter C and manages values related to that parameter. The function optimFunc() is the function to be optimized to get the best values for the Shammas polynomial power parameters A, B, and C, and the best Shammas polynomial order. The function getFuncName() returns a string-type function name given a handle of a function.

Using the calculation scheme, note that function PadeShamPoly3() does not explicitly iterate over different values of A, B and C. It delegates that task to the Matlab function particleswarm().

The following Matlab script goAll (which is identical to goAll in Part I) performs the various Pade-Shammas polynomial fittings for the various tested functions:

```
% Version 1.0.0 8/9/2020
global bUseDiary
global bDeleteIfExists

bUseDiary = true;
bDeleteIfExists = true; % or false
selIdx = 0; % note a zero value will execute all the models
runNum = 1;
bShutdown = false;
```

```

tic;

if selIdx==0 || selIdx==1
    lstA = [0 0.5];
    lstB = [0.1 0.5];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B)A+B*i;

PadeShamPoly2(@(x)acos(x),gx,[0:.01:1],Lb,Ub,runNum,"arccos(x)","arccos_1.txt")

PadeShamPoly2(@(x)asin(x),gx,[0:.01:1],Lb,Ub,runNum,"arcsin(x)","arcsin_1.txt")

PadeShamPoly2(@(x)atan(x),gx,[0:.01:1],Lb,Ub,runNum,"arctan(x)","arctan_1.txt")

PadeShamPoly2(@(x)sin(x),gx,[0:pi/1000:pi/2],Lb,Ub,runNum,"sin(x)","sin_1.txt")

PadeShamPoly2(@(x)cos(x),gx,[0:pi/1000:pi/2],Lb,Ub,runNum,"cos(x)","cos_1.txt")

PadeShamPoly2(@(x)tan(x),gx,[0:.01:1],Lb,Ub,runNum,"tan(x)","tan_1.txt")
PadeShamPoly2(@(x)sinh(x),gx,[0:.01:5],Lb,Ub,runNum,"sinh(x)","sinh_1.txt")
PadeShamPoly2(@(x)cosh(x),gx,[0:.01:5],Lb,Ub,runNum,"cosh(x)","cosh_1.txt")
PadeShamPoly2(@(x)tanh(x),gx,[0:.01:3],Lb,Ub,runNum,"tanh(x)","tanh_1.txt")
PadeShamPoly2(@(x)erf(x),gx,[0:.01:2.1],Lb,Ub,runNum,"erf(x)","erf_1.txt")
PadeShamPoly2(@(x)exp(x),gx,[0:.01:2],Lb,Ub,runNum,"exp(x)","exp_1.txt")
PadeShamPoly2(@(x)log(x),gx,[1:.01:10],Lb,Ub,runNum,"ln(x)","ln_1.txt")
PadeShamPoly2(@(x)log10(x),gx,[1:.01:10],Lb,Ub,runNum,"log(x)","log_1.txt")
PadeShamPoly2(@(x)10.^x,gx,[0:.01:1],Lb,Ub,runNum,"10^x","pwr10_1.txt")

PadeShamPoly2(@(x)tinva(0.95,x),gx,[2:0.1:100],Lb,Ub,runNum,"tinva(0.95,x)","tinva_1.txt")

PadeShamPoly2(@(x)tinva(0.975,x),gx,[2:0.1:100],Lb,Ub,runNum,"tinva(0.975,x)","tinva_1.txt")

PadeShamPoly2(@(x)log10(gamma(x)),gx,[2:0.1:100],Lb,Ub,runNum,"log10Gamma(x)","log10Gamma_1.txt")

PadeShamPoly2(@(x)digamma(x),gx,[2:0.1:100],Lb,Ub,runNum,"digamma(x)","digamma_1.txt")

PadeShamPoly2(@(x)trigamma(x),gx,[1:0.1:100],Lb,Ub,runNum,"trigamma(x)","trigamma_1.txt")

PadeShamPoly2(@(x)besselj(0,x),gx,[2:0.1:30],Lb,Ub,runNum,"J0(x)","J0x_1.txt")

PadeShamPoly2(@(x)besselj(1,x),gx,[0:0.1:30],Lb,Ub,runNum,"J1(x)","J1x_1.txt")

PadeShamPoly2(@(x)besselj(2,x),gx,[0:0.1:30],Lb,Ub,runNum,"J2(x)","J2x_1.txt")

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PadeShamPoly2 (@ (x) besselj (3,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J3 (x) ", "J3x_1.txt"
)

PadeShamPoly2 (@ (x) besselj (4,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J4 (x) ", "J4x_1.txt"
)

PadeShamPoly2 (@ (x) besselj (5,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J5 (x) ", "J5x_1.txt"
)

PadeShamPoly2 (@ (x) integral (@ (z) sin (z.^2) ,0,x) ,gx, [1:0.01:5] ,Lb,Ub,runNum,"Fre
snelSine (x) ", "FresnelSine_1.txt")

PadeShamPoly2 (@ (x) integral (@ (z) cos (z.^2) ,0,x) ,gx, [0.5:0.01:5] ,Lb,Ub,runNum,"F
resnelCosine (x) ", "FresnelCosine_1.txt")

PadeShamPoly2 (@ (x) integral (@ (z) sin (z) ./z,0,x) ,gx, [1.3:0.1:20] ,Lb,Ub,runNum,"S
i (x) ", "Si_1.txt")
    PadeShamPoly2 (@ (x) 0.57721566+log (x) -integral (@ (z) (1-
cos (z) ) ./z,0,x) ,gx, [0.5:0.1:20] ,Lb,Ub,runNum,"Ci (x) ", "CI_1.txt")

PadeShamPoly2 (@ (x) asinh (x) ,gx, [0:0.1:100] ,Lb,Ub,runNum,"asinh (x) ", "asinh_1.tx
t")

PadeShamPoly2 (@ (x) acosh (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"acosh (x) ", "acosh_1.tx
t")

PadeShamPoly2 (@ (x) atanh (x) ,gx, [0:0.001:0.999] ,Lb,Ub,runNum,"atanh (x) ", "atanh_
1.txt")
end

if selIdx==0 || selIdx==2
    lstA = [0 2];
    lstB = [0.1 2];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B) A+B/i;

PadeShamPoly2 (@ (x) acos (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arccos (x) ", "arccos_2.txt
")

PadeShamPoly2 (@ (x) asin (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arcsin (x) ", "arcsin_2.txt
")

PadeShamPoly2 (@ (x) atan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arctan (x) ", "arctan_2.txt
")

PadeShamPoly2 (@ (x) sin (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"sin (x) ", "sin_2.txt
")

PadeShamPoly2 (@ (x) cos (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"cos (x) ", "cos_2.txt
")
    PadeShamPoly2 (@ (x) tan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"tan (x) ", "tan_2.txt")
    PadeShamPoly2 (@ (x) sinh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"sinh (x) ", "sinh_2.txt")
    PadeShamPoly2 (@ (x) cosh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"cosh (x) ", "cosh_2.txt")
    PadeShamPoly2 (@ (x) tanh (x) ,gx, [0:.01:3] ,Lb,Ub,runNum,"tanh (x) ", "tanh_2.txt")

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PadeShamPoly2 (@ (x) erf (x) ,gx, [0:.01:2.1] ,Lb,Ub,runNum,"erf (x) ","erf_2.txt")
PadeShamPoly2 (@ (x) exp (x) ,gx, [0:.01:2] ,Lb,Ub,runNum,"exp (x) ","exp_2.txt")
PadeShamPoly2 (@ (x) log (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"ln (x) ","ln_2.txt")
PadeShamPoly2 (@ (x) log10 (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"log (x) ","log_2.txt")
PadeShamPoly2 (@ (x) 10.^x,gx, [0:.01:1] ,Lb,Ub,runNum,"10^x","pwr10_2.txt")

PadeShamPoly2 (@ (x) tinv (0.95,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.95,x) ","ti
nv1_2.txt")

PadeShamPoly2 (@ (x) tinv (0.975,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.975,x) ","
tinv2_2.txt")

PadeShamPoly2 (@ (x) log10 (gamma (x) ) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"log10Gamma (x) "
,"log10Gamma_2.txt")

PadeShamPoly2 (@ (x) digamma (x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"digamma (x) ","digamm
a_2.txt")

PadeShamPoly2 (@ (x) trigamma (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"trigamma (x) ","trig
amma_2.txt")

PadeShamPoly2 (@ (x) besselj (0,x) ,gx, [2:0.1:30] ,Lb,Ub,runNum,"J0 (x) ","J0x_2.txt"
)

PadeShamPoly2 (@ (x) besselj (1,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J1 (x) ","J1x_2.txt"
)

PadeShamPoly2 (@ (x) besselj (2,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J2 (x) ","J2x_2.txt"
)

PadeShamPoly2 (@ (x) besselj (3,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J3 (x) ","J3x_2.txt"
)

PadeShamPoly2 (@ (x) besselj (4,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J4 (x) ","J4x_2.txt"
)

PadeShamPoly2 (@ (x) besselj (5,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J5 (x) ","J5x_2.txt"
)

PadeShamPoly2 (@ (x) integral (@ (z) sin (z.^2) ,0,x) ,gx, [1:0.01:5] ,Lb,Ub,runNum,"Fre
snelSine (x) ","FresnelSine_2.txt")

PadeShamPoly2 (@ (x) integral (@ (z) cos (z.^2) ,0,x) ,gx, [0.5:0.01:5] ,Lb,Ub,runNum,"F
resnelCosine (x) ","FresnelCosine_2.txt")

PadeShamPoly2 (@ (x) integral (@ (z) sin (z) ./z,0,x) ,gx, [1.3:0.1:20] ,Lb,Ub,runNum,"S
i (x) ","Si_2.txt")
PadeShamPoly2 (@ (x) 0.57721566+log (x) -integral (@ (z) (1-
cos (z) ) ./z,0,x) ,gx, [0.5:0.1:20] ,Lb,Ub,runNum,"Ci (x) ","Ci_2.txt")

PadeShamPoly2 (@ (x) asinh (x) ,gx, [0:0.1:100] ,Lb,Ub,runNum,"asinh (x) ","asinh_2.tx
t")

PadeShamPoly2 (@ (x) acosh (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"acosh (x) ","acosh_2.tx
t")

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PadeShamPoly2 (@ (x) atanh (x) ,gx, [0:0.001:0.999] ,Lb,Ub,runNum,"atanh (x) ","atanh_
2.txt")
end

if selIdx==0 || selIdx==3
    lstA = [0 1.4];
    lstB = [0.1 1];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B)A+B*sqrt(i);

PadeShamPoly2 (@ (x) acos (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arccos (x) ","arccos_3.txt
")

PadeShamPoly2 (@ (x) asin (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arcsin (x) ","arcsin_3.txt
")

PadeShamPoly2 (@ (x) atan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arctan (x) ","arctan_3.txt
")

PadeShamPoly2 (@ (x) sin (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"sin (x) ","sin_3.txt
")

PadeShamPoly2 (@ (x) cos (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"cos (x) ","cos_3.txt
")
    PadeShamPoly2 (@ (x) tan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"tan (x) ","tan_3.txt")
    PadeShamPoly2 (@ (x) sinh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"sinh (x) ","sinh_3.txt")
    PadeShamPoly2 (@ (x) cosh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"cosh (x) ","cosh_3.txt")
    PadeShamPoly2 (@ (x) tanh (x) ,gx, [0:.01:3] ,Lb,Ub,runNum,"tanh (x) ","tanh_3.txt")
    PadeShamPoly2 (@ (x) erf (x) ,gx, [0:.01:2.1] ,Lb,Ub,runNum,"erf (x) ","erf_3.txt")
    PadeShamPoly2 (@ (x) exp (x) ,gx, [0:.01:2] ,Lb,Ub,runNum,"exp (x) ","exp_3.txt")
    PadeShamPoly2 (@ (x) log (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"ln (x) ","ln_3.txt")
    PadeShamPoly2 (@ (x) log10 (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"log (x) ","log_3.txt")
    PadeShamPoly2 (@ (x) 10.^x,gx, [0:.01:1] ,Lb,Ub,runNum,"10^x","pwr10_3.txt")

PadeShamPoly2 (@ (x) tinv (0.95,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.95,x) ","ti
nv1_3.txt")

PadeShamPoly2 (@ (x) tinv (0.975,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.975,x) ","
tinv2_3.txt")

PadeShamPoly2 (@ (x) log10 (gamma (x)) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"log10Gamma (x) "
,"log10Gamma_3.txt")

PadeShamPoly2 (@ (x) digamma (x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"digamma (x) ","digamm
a_3.txt")

PadeShamPoly2 (@ (x) trigamma (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"trigamma (x) ","trig
amma_3.txt")

PadeShamPoly2 (@ (x) besselj (0,x) ,gx, [2:0.1:30] ,Lb,Ub,runNum,"J0 (x) ","J0x_3.txt"
)

PadeShamPoly2 (@ (x) besselj (1,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J1 (x) ","J1x_3.txt"
)

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PadeShamPoly2 (@ (x) besselj (2, x) , gx, [0:0.1:30] , Lb, Ub, runNum, "J2 (x) " , "J2x_3.txt"
)

PadeShamPoly2 (@ (x) besselj (3, x) , gx, [0:0.1:30] , Lb, Ub, runNum, "J3 (x) " , "J3x_3.txt"
)

PadeShamPoly2 (@ (x) besselj (4, x) , gx, [0:0.1:30] , Lb, Ub, runNum, "J4 (x) " , "J4x_3.txt"
)

PadeShamPoly2 (@ (x) besselj (5, x) , gx, [0:0.1:30] , Lb, Ub, runNum, "J5 (x) " , "J5x_3.txt"
)

PadeShamPoly2 (@ (x) integral (@ (z) sin (z.^2) , 0, x) , gx, [1:0.01:5] , Lb, Ub, runNum, "Fre
snelSine (x) " , "FresnelSine_3.txt")

PadeShamPoly2 (@ (x) integral (@ (z) cos (z.^2) , 0, x) , gx, [0.5:0.01:5] , Lb, Ub, runNum, "F
resnelCosine (x) " , "FresnelCosine_3.txt")

PadeShamPoly2 (@ (x) integral (@ (z) sin (z) ./z, 0, x) , gx, [1.3:0.1:20] , Lb, Ub, runNum, "S
i (x) " , "Si_3.txt")
    PadeShamPoly2 (@ (x) 0.57721566+log (x) -integral (@ (z) (1-
cos (z) ) ./z, 0, x) , gx, [0.5:0.1:20] , Lb, Ub, runNum, "Ci (x) " , "Ci_3.txt")

PadeShamPoly2 (@ (x) asinh (x) , gx, [0:0.1:100] , Lb, Ub, runNum, "asinh (x) " , "asinh_3.tx
t")

PadeShamPoly2 (@ (x) acosh (x) , gx, [1:0.1:100] , Lb, Ub, runNum, "acosh (x) " , "acosh_3.tx
t")

PadeShamPoly2 (@ (x) atanh (x) , gx, [0:0.001:0.999] , Lb, Ub, runNum, "atanh (x) " , "atanh_
3.txt")
end

if selIdx==0 || selIdx==4
    lstA = [0 1];
    lstB = [0.1 1];
    lstC = [0 2];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstC(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstC(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B,C) A+B*sqrt(i+C);

PadeShamPoly3 (@ (x) acos (x) , gx, [0:.01:1] , Lb, Ub, runNum, "arccos (x) " , "arccos_4.txt
")

PadeShamPoly3 (@ (x) asin (x) , gx, [0:.01:1] , Lb, Ub, runNum, "arcsin (x) " , "arcsin_4.txt
")

PadeShamPoly3 (@ (x) atan (x) , gx, [0:.01:1] , Lb, Ub, runNum, "arctan (x) " , "arctan_4.txt
")

PadeShamPoly3 (@ (x) sin (x) , gx, [0:pi/1000:pi/2] , Lb, Ub, runNum, "sin (x) " , "sin_4.txt
")

PadeShamPoly3 (@ (x) cos (x) , gx, [0:pi/1000:pi/2] , Lb, Ub, runNum, "cos (x) " , "cos_4.txt
")

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PadeShamPoly3(@ (x) tan(x),gx,[0:.01:1],Lb,Ub,runNum,"tan(x)","tan_4.txt")
PadeShamPoly3(@ (x) sinh(x),gx,[0:.01:5],Lb,Ub,runNum,"sinh(x)","sinh_4.txt")
PadeShamPoly3(@ (x) cosh(x),gx,[0:.01:5],Lb,Ub,runNum,"cosh(x)","cosh_4.txt")
PadeShamPoly3(@ (x) tanh(x),gx,[0:.01:3],Lb,Ub,runNum,"tanh(x)","tanh_4.txt")
PadeShamPoly3(@ (x) erf(x),gx,[0:.01:2.1],Lb,Ub,runNum,"erf(x)","erf_4.txt")
PadeShamPoly3(@ (x) exp(x),gx,[0:.01:2],Lb,Ub,runNum,"exp(x)","exp_4.txt")
PadeShamPoly3(@ (x) log(x),gx,[1:.01:10],Lb,Ub,runNum,"ln(x)","ln_4.txt")
PadeShamPoly3(@ (x) log10(x),gx,[1:.01:10],Lb,Ub,runNum,"log(x)","log_4.txt")
PadeShamPoly3(@ (x) 10.^x,gx,[0:.01:1],Lb,Ub,runNum,"10^x","pwr10_4.txt")

PadeShamPoly3(@ (x) tinv(0.95,x),gx,[2:100],Lb,Ub,runNum,"tinv(0.95,x)","tinv1_4.txt")

PadeShamPoly3(@ (x) tinv(0.975,x),gx,[2:100],Lb,Ub,runNum,"tinv(0.975,x)","tinv2_4.txt")

PadeShamPoly3(@ (x) log10(gamma(x)),gx,[2:100],Lb,Ub,runNum,"log10Gamma(x)","log10Gamma_4.txt")

PadeShamPoly3(@ (x) digamma(x),gx,[2:100],Lb,Ub,runNum,"digamma(x)","digamma_4.txt")

PadeShamPoly3(@ (x) trigamma(x),gx,[1:100],Lb,Ub,runNum,"trigamma(x)","trigamma_4.txt")

PadeShamPoly3(@ (x) besselj(0,x),gx,[2:0.1:30],Lb,Ub,runNum,"J0(x)","J0x_4.txt")
)

PadeShamPoly3(@ (x) besselj(1,x),gx,[0:0.1:30],Lb,Ub,runNum,"J1(x)","J1x_4.txt")
)

PadeShamPoly3(@ (x) besselj(2,x),gx,[0:0.1:30],Lb,Ub,runNum,"J2(x)","J2x_4.txt")
)

PadeShamPoly3(@ (x) besselj(3,x),gx,[0:0.1:30],Lb,Ub,runNum,"J3(x)","J3x_4.txt")
)

PadeShamPoly3(@ (x) besselj(4,x),gx,[0:0.1:30],Lb,Ub,runNum,"J4(x)","J4x_4.txt")
)

PadeShamPoly3(@ (x) besselj(5,x),gx,[0:0.1:30],Lb,Ub,runNum,"J5(x)","J5x_4.txt")
)

PadeShamPoly3(@ (x) integral(@(z) sin(z.^2),0,x),gx,[1:0.01:5],Lb,Ub,runNum,"FresnelSine(x)","FresnelSine_4.txt")

PadeShamPoly3(@ (x) integral(@(z) cos(z.^2),0,x),gx,[0.5:0.01:5],Lb,Ub,runNum,"FresnelCosine(x)","FresnelCosine_4.txt")

PadeShamPoly3(@ (x) integral(@(z) sin(z)./z,0,x),gx,[1.3:0.1:20],Lb,Ub,runNum,"Si(x)","Si_4.txt")
PadeShamPoly3(@ (x) 0.57721566+log(x)-integral(@(z) (1-cos(z))./z,0,x),gx,[0.5:0.1:20],Lb,Ub,runNum,"Ci(x)","Ci_4.txt")

PadeShamPoly3(@ (x) asinh(x),gx,[0:0.1:100],Lb,Ub,runNum,"asinh(x)","asinh_4.txt")
)

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PadeShamPoly3(@ (x) acosh(x),gx,[1:0.1:100],Lb,Ub,runNum,"acosh(x)","acosh_4.txt")

PadeShamPoly3(@ (x) atanh(x),gx,[0:0.001:0.999],Lb,Ub,runNum,"atanh(x)","atanh_4.txt")
end

if selIdx==0 || selIdx==5
    lstA = [0 0.5];
    lstB = [0.1 0.27];
    lstC = [0 0.27];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstC(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstC(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B,C)A+B*i+C*(i-1);

PadeShamPoly3(@ (x) acos(x),gx,[0:.01:1],Lb,Ub,runNum,"arccos(x)","arccos_5.txt")

PadeShamPoly3(@ (x) asin(x),gx,[0:.01:1],Lb,Ub,runNum,"arcsin(x)","arcsin_5.txt")

PadeShamPoly3(@ (x) atan(x),gx,[0:.01:1],Lb,Ub,runNum,"arctan(x)","arctan_5.txt")

PadeShamPoly3(@ (x) sin(x),gx,[0:pi/1000:pi/2],Lb,Ub,runNum,"sin(x)","sin_5.txt")

PadeShamPoly3(@ (x) cos(x),gx,[0:pi/1000:pi/2],Lb,Ub,runNum,"cos(x)","cos_5.txt")

    PadeShamPoly3(@ (x) tan(x),gx,[0:.01:1],Lb,Ub,runNum,"tan(x)","tan_5.txt")
    PadeShamPoly3(@ (x) sinh(x),gx,[0:.01:5],Lb,Ub,runNum,"sinh(x)","sinh_5.txt")
    PadeShamPoly3(@ (x) cosh(x),gx,[0:.01:5],Lb,Ub,runNum,"cosh(x)","cosh_5.txt")
    PadeShamPoly3(@ (x) tanh(x),gx,[0:.01:3],Lb,Ub,runNum,"tanh(x)","tanh_5.txt")
    PadeShamPoly3(@ (x) erf(x),gx,[0:.01:2.1],Lb,Ub,runNum,"erf(x)","erf_5.txt")
    PadeShamPoly3(@ (x) exp(x),gx,[0:.01:2],Lb,Ub,runNum,"exp(x)","exp_5.txt")
    PadeShamPoly3(@ (x) log(x),gx,[1:.01:10],Lb,Ub,runNum,"ln(x)","ln_5.txt")
    PadeShamPoly3(@ (x) log10(x),gx,[1:.01:10],Lb,Ub,runNum,"log(x)","log_5.txt")
    PadeShamPoly3(@ (x) 10.^x,gx,[0:.01:1],Lb,Ub,runNum,"10^x","pwr10_5.txt")

PadeShamPoly3(@ (x) tinv(0.95,x),gx,[2:0.1:100],Lb,Ub,runNum,"tinv(0.95,x)","tinv1_5.txt")

PadeShamPoly3(@ (x) tinv(0.975,x),gx,[2:0.1:100],Lb,Ub,runNum,"tinv(0.975,x)","tinv2_5.txt")

PadeShamPoly3(@ (x) log10(gamma(x)),gx,[2:0.1:100],Lb,Ub,runNum,"log10Gamma(x)","log10Gamma_5.txt")

PadeShamPoly3(@ (x) digamma(x),gx,[2:0.1:100],Lb,Ub,runNum,"digamma(x)","digamma_5.txt")

PadeShamPoly3(@ (x) trigamma(x),gx,[1:0.1:100],Lb,Ub,runNum,"trigamma(x)","trigamma_5.txt")

```

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PadeShamPoly3(@ (x)besselj(0,x),gx,[2:0.1:30],Lb,Ub,runNum,"J0(x)","J0x_5.txt"
)

PadeShamPoly3(@ (x)besselj(1,x),gx,[0:0.1:30],Lb,Ub,runNum,"J1(x)","J1x_5.txt"
)

PadeShamPoly3(@ (x)besselj(2,x),gx,[0:0.1:30],Lb,Ub,runNum,"J2(x)","J2x_5.txt"
)

PadeShamPoly3(@ (x)besselj(3,x),gx,[0:0.1:30],Lb,Ub,runNum,"J3(x)","J3x_5.txt"
)

PadeShamPoly3(@ (x)besselj(4,x),gx,[0:0.1:30],Lb,Ub,runNum,"J4(x)","J4x_5.txt"
)

PadeShamPoly3(@ (x)besselj(5,x),gx,[0:0.1:30],Lb,Ub,runNum,"J5(x)","J5x_5.txt"
)

PadeShamPoly3(@ (x)integral(@(z)sin(z.^2),0,x),gx,[1:0.01:5],Lb,Ub,runNum,"FresnelSine(x)","FresnelSine_5.txt")

PadeShamPoly3(@ (x)integral(@(z)cos(z.^2),0,x),gx,[0.5:0.01:5],Lb,Ub,runNum,"FresnelCosine(x)","FresnelCosine_5.txt")

PadeShamPoly3(@ (x)integral(@(z)sin(z)./z,0,x),gx,[1.3:0.1:20],Lb,Ub,runNum,"Si(x)","Si_5.txt")
    PadeShamPoly3(@ (x)0.57721566+log(x)-integral(@(z)(1-cos(z))./z,0,x),gx,[0.5:0.1:20],Lb,Ub,runNum,"Ci(x)","Ci_5.txt")

PadeShamPoly3(@ (x)asinh(x),gx,[0:0.1:100],Lb,Ub,runNum,"asinh(x)","asinh_5.txt")

PadeShamPoly3(@ (x)acosh(x),gx,[1:0.1:100],Lb,Ub,runNum,"acosh(x)","acosh_5.txt")

PadeShamPoly3(@ (x)atanh(x),gx,[0:0.001:0.999],Lb,Ub,runNum,"atanh(x)","atanh_5.txt")
end

if selIdx==0 || selIdx==6
    1stA = [0 0.75];
    1stB = [0.1 0.25];
    1stC = [0 0.25];
    1stOrder = [3 7];
    Lb = [1stA(1) 1stB(1) 1stC(1) 1stOrder(1) 1stOrder(1)];
    Ub = [1stA(2) 1stB(2) 1stC(2) 1stOrder(2) 1stOrder(2)];
    gx = @(i,A,B,C)A+B*i-C*(i-1);

PadeShamPoly3(@ (x)acos(x),gx,[0:.01:1],Lb,Ub,runNum,"arccos(x)","arccos_6.txt")

PadeShamPoly3(@ (x)asin(x),gx,[0:.01:1],Lb,Ub,runNum,"arcsin(x)","arcsin_6.txt")

PadeShamPoly3(@ (x)atan(x),gx,[0:.01:1],Lb,Ub,runNum,"arctan(x)","arctan_6.txt")

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PadeShamPoly3(@ (x) sin(x),gx,[0:pi/1000:pi/2],Lb,Ub,runNum,"sin(x)","sin_6.txt")

PadeShamPoly3(@ (x) cos(x),gx,[0:pi/1000:pi/2],Lb,Ub,runNum,"cos(x)","cos_6.txt")

PadeShamPoly3(@ (x) tan(x),gx,[0:.01:1],Lb,Ub,runNum,"tan(x)","tan_6.txt")
PadeShamPoly3(@ (x) sinh(x),gx,[0:.01:5],Lb,Ub,runNum,"sinh(x)","sinh_6.txt")
PadeShamPoly3(@ (x) cosh(x),gx,[0:.01:5],Lb,Ub,runNum,"cosh(x)","cosh_6.txt")
PadeShamPoly3(@ (x) tanh(x),gx,[0:.01:3],Lb,Ub,runNum,"tanh(x)","tanh_6.txt")
PadeShamPoly3(@ (x) erf(x),gx,[0:.01:2.1],Lb,Ub,runNum,"erf(x)","erf_6.txt")
PadeShamPoly3(@ (x) exp(x),gx,[0:.01:2],Lb,Ub,runNum,"exp(x)","exp_6.txt")
PadeShamPoly3(@ (x) log(x),gx,[1:.01:10],Lb,Ub,runNum,"ln(x)","ln_6.txt")
PadeShamPoly3(@ (x) log10(x),gx,[1:.01:10],Lb,Ub,runNum,"log(x)","log_6.txt")
PadeShamPoly3(@ (x) 10.^x,gx,[0:.01:1],Lb,Ub,runNum,"10^x","pwr10_6.txt")

PadeShamPoly3(@ (x) tinv(0.95,x),gx,[2:0.1:100],Lb,Ub,runNum,"tinv(0.95,x)","tinv1_6.txt")

PadeShamPoly3(@ (x) tinv(0.975,x),gx,[2:0.1:100],Lb,Ub,runNum,"tinv(0.975,x)","tinv2_6.txt")

PadeShamPoly3(@ (x) log10(gamma(x)),gx,[2:0.1:100],Lb,Ub,runNum,"log10Gamma(x)","log10Gamma_6.txt")

PadeShamPoly3(@ (x) digamma(x),gx,[2:0.1:100],Lb,Ub,runNum,"digamma(x)","digamma_6.txt")

PadeShamPoly3(@ (x) trigamma(x),gx,[1:0.1:100],Lb,Ub,runNum,"trigamma(x)","trigamma_6.txt")

PadeShamPoly3(@ (x) besselj(0,x),gx,[2:0.1:30],Lb,Ub,runNum,"J0(x)","J0x_6.txt")

PadeShamPoly3(@ (x) besselj(1,x),gx,[0:0.1:30],Lb,Ub,runNum,"J1(x)","J1x_6.txt")

PadeShamPoly3(@ (x) besselj(2,x),gx,[0:0.1:30],Lb,Ub,runNum,"J2(x)","J2x_6.txt")

PadeShamPoly3(@ (x) besselj(3,x),gx,[0:0.1:30],Lb,Ub,runNum,"J3(x)","J3x_6.txt")

PadeShamPoly3(@ (x) besselj(4,x),gx,[0:0.1:30],Lb,Ub,runNum,"J4(x)","J4x_6.txt")

PadeShamPoly3(@ (x) besselj(5,x),gx,[0:0.1:30],Lb,Ub,runNum,"J5(x)","J5x_6.txt")

PadeShamPoly3(@ (x) integral(@(z) sin(z.^2),0,x),gx,[1:0.01:5],Lb,Ub,runNum,"FresnelSine(x)","FresnelSine_6.txt")

PadeShamPoly3(@ (x) integral(@(z) cos(z.^2),0,x),gx,[0.5:0.01:5],Lb,Ub,runNum,"FresnelCosine(x)","FresnelCosine_6.txt")

PadeShamPoly3(@ (x) integral(@(z) sin(z)./z,0,x),gx,[1.3:0.1:20],Lb,Ub,runNum,"Si(x)","Si_6.txt")

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PadeShamPoly3(@ (x) 0.57721566+log(x)-integral(@(z) (1-
cos(z))./z,0,x),gx,[0.5:0.1:20],Lb,Ub,runNum,"Ci(x)","Ci_6.txt")

PadeShamPoly3(@ (x) asinh(x),gx,[0:0.1:100],Lb,Ub,runNum,"asinh(x)","asinh_6.txt")

PadeShamPoly3(@ (x) acosh(x),gx,[1:0.1:100],Lb,Ub,runNum,"acosh(x)","acosh_6.txt")

PadeShamPoly3(@ (x) atanh(x),gx,[0:0.001:0.999],Lb,Ub,runNum,"atanh(x)","atanh_6.txt")
end

if selIdx==0 || selIdx==7
    lstA = [0 0.34];
    lstB = [0.1 0.39];
    lstC = [0 0.36];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstC(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstC(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B,C)A+B*i+C*sqrt(i-1);

PadeShamPoly3(@ (x) acos(x),gx,[0:.01:1],Lb,Ub,runNum,"arccos(x)","arccos_7.txt")

PadeShamPoly3(@ (x) asin(x),gx,[0:.01:1],Lb,Ub,runNum,"arcsin(x)","arcsin_7.txt")

PadeShamPoly3(@ (x) atan(x),gx,[0:.01:1],Lb,Ub,runNum,"arctan(x)","arctan_7.txt")

PadeShamPoly3(@ (x) sin(x),gx,[0:pi/1000:pi/2],Lb,Ub,runNum,"sin(x)","sin_7.txt")

PadeShamPoly3(@ (x) cos(x),gx,[0:pi/1000:pi/2],Lb,Ub,runNum,"cos(x)","cos_7.txt")

    PadeShamPoly3(@ (x) tan(x),gx,[0:.01:1],Lb,Ub,runNum,"tan(x)","tan_7.txt")
    PadeShamPoly3(@ (x) sinh(x),gx,[0:.01:5],Lb,Ub,runNum,"sinh(x)","sinh_7.txt")
    PadeShamPoly3(@ (x) cosh(x),gx,[0:.01:5],Lb,Ub,runNum,"cosh(x)","cosh_7.txt")
    PadeShamPoly3(@ (x) tanh(x),gx,[0:.01:3],Lb,Ub,runNum,"tanh(x)","tanh_7.txt")
    PadeShamPoly3(@ (x) erf(x),gx,[0:.01:2.1],Lb,Ub,runNum,"erf(x)","erf_7.txt")
    PadeShamPoly3(@ (x) exp(x),gx,[0:.01:2],Lb,Ub,runNum,"exp(x)","exp_7.txt")
    PadeShamPoly3(@ (x) log(x),gx,[1:.01:10],Lb,Ub,runNum,"ln(x)","ln_7.txt")
    PadeShamPoly3(@ (x) log10(x),gx,[1:.01:10],Lb,Ub,runNum,"log(x)","log_7.txt")
    PadeShamPoly3(@ (x) 10.^x,gx,[0:.01:1],Lb,Ub,runNum,"10^x","pwr10_7.txt")

PadeShamPoly3(@ (x) tinv(0.95,x),gx,[2:0.1:100],Lb,Ub,runNum,"tinv(0.95,x)","tinv1_7.txt")

PadeShamPoly3(@ (x) tinv(0.975,x),gx,[2:0.1:100],Lb,Ub,runNum,"tinv(0.975,x)","tinv2_7.txt")

PadeShamPoly3(@ (x) log10(gamma(x)),gx,[2:0.1:100],Lb,Ub,runNum,"log10Gamma(x)","log10Gamma_7.txt")

PadeShamPoly3(@ (x) digamma(x),gx,[2:0.1:100],Lb,Ub,runNum,"digamma(x)","digamma_7.txt")

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PadeShamPoly3(@ (x) trigamma (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"trigamma (x) ","trigamma_7.txt")

PadeShamPoly3(@ (x) besselj (0,x) ,gx, [2:0.1:30] ,Lb,Ub,runNum,"J0 (x) ","J0x_7.txt"
)

PadeShamPoly3(@ (x) besselj (1,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J1 (x) ","J1x_7.txt"
)

PadeShamPoly3(@ (x) besselj (2,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J2 (x) ","J2x_7.txt"
)

PadeShamPoly3(@ (x) besselj (3,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J3 (x) ","J3x_7.txt"
)

PadeShamPoly3(@ (x) besselj (4,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J4 (x) ","J4x_7.txt"
)

PadeShamPoly3(@ (x) besselj (5,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J5 (x) ","J5x_7.txt"
)

PadeShamPoly3(@ (x) integral (@ (z) sin (z.^2) ,0,x) ,gx, [1:0.01:5] ,Lb,Ub,runNum,"FresnelSine (x) ","FresnelSine_7.txt")

PadeShamPoly3(@ (x) integral (@ (z) cos (z.^2) ,0,x) ,gx, [0.5:0.01:5] ,Lb,Ub,runNum,"FresnelCosine (x) ","FresnelCosine_7.txt")

PadeShamPoly3(@ (x) integral (@ (z) sin (z) ./z,0,x) ,gx, [1.3:0.1:20] ,Lb,Ub,runNum,"Si (x) ","Si_7.txt")
    PadeShamPoly3(@ (x) 0.57721566+log (x) -integral (@ (z) (1-cos (z)) ./z,0,x) ,gx, [0.5:0.1:20] ,Lb,Ub,runNum,"Ci (x) ","Ci_7.txt")

PadeShamPoly3(@ (x) asinh (x) ,gx, [0:0.1:100] ,Lb,Ub,runNum,"asinh (x) ","asinh_7.txt")

PadeShamPoly3(@ (x) acosh (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"acosh (x) ","acosh_7.txt")

PadeShamPoly3(@ (x) atanh (x) ,gx, [0:0.001:0.999] ,Lb,Ub,runNum,"atanh (x) ","atanh_7.txt")
end

if selIdx==0 || selIdx==8
    lstA = [0 0.55];
    lstB = [0.1 0.4];
    lstC = [0 0.4];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstC(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstC(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B,C) A+B*sqrt(i)+C*(i-1);

PadeShamPoly3(@ (x) acos (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arccos (x) ","arccos_8.txt")

PadeShamPoly3(@ (x) asin (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arcsin (x) ","arcsin_8.txt")

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PadeShamPoly3 (@ (x) atan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arctan (x) ","arctan_8.txt
")

PadeShamPoly3 (@ (x) sin (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"sin (x) ","sin_8.txt
")

PadeShamPoly3 (@ (x) cos (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"cos (x) ","cos_8.txt
")
  PadeShamPoly3 (@ (x) tan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"tan (x) ","tan_8.txt")
  PadeShamPoly3 (@ (x) sinh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"sinh (x) ","sinh_8.txt")
  PadeShamPoly3 (@ (x) cosh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"cosh (x) ","cosh_8.txt")
  PadeShamPoly3 (@ (x) tanh (x) ,gx, [0:.01:3] ,Lb,Ub,runNum,"tanh (x) ","tanh_8.txt")
  PadeShamPoly3 (@ (x) erf (x) ,gx, [0:.01:2.1] ,Lb,Ub,runNum,"erf (x) ","erf_8.txt")
  PadeShamPoly3 (@ (x) exp (x) ,gx, [0:.01:2] ,Lb,Ub,runNum,"exp (x) ","exp_8.txt")
  PadeShamPoly3 (@ (x) log (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"ln (x) ","ln_8.txt")
  PadeShamPoly3 (@ (x) log10 (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"log (x) ","log_8.txt")
  PadeShamPoly3 (@ (x) 10.^x,gx, [0:.01:1] ,Lb,Ub,runNum,"10^x","pwr10_8.txt")

PadeShamPoly3 (@ (x) tinv (0.95,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.95,x) ","ti
nv1_8.txt")

PadeShamPoly3 (@ (x) tinv (0.975,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.975,x) ","
tinv2_8.txt")

PadeShamPoly3 (@ (x) log10 (gamma (x) ) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"log10Gamma (x) "
,"log10Gamma_8.txt")

PadeShamPoly3 (@ (x) digamma (x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"digamma (x) ","digamm
a_8.txt")

PadeShamPoly3 (@ (x) trigamma (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"trigamma (x) ","trig
amma_8.txt")

PadeShamPoly3 (@ (x) besselj (0,x) ,gx, [2:0.1:30] ,Lb,Ub,runNum,"J0 (x) ","J0x_8.txt"
)

PadeShamPoly3 (@ (x) besselj (1,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J1 (x) ","J1x_8.txt"
)

PadeShamPoly3 (@ (x) besselj (2,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J2 (x) ","J2x_8.txt"
)

PadeShamPoly3 (@ (x) besselj (3,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J3 (x) ","J3x_8.txt"
)

PadeShamPoly3 (@ (x) besselj (4,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J4 (x) ","J4x_8.txt"
)

PadeShamPoly3 (@ (x) besselj (5,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J5 (x) ","J5x_8.txt"
)

PadeShamPoly3 (@ (x) integral (@ (z) sin (z.^2) ,0,x) ,gx, [1:0.01:5] ,Lb,Ub,runNum,"Fre
snelSine (x) ","FresnelSine_8.txt")

PadeShamPoly3 (@ (x) integral (@ (z) cos (z.^2) ,0,x) ,gx, [0.5:0.01:5] ,Lb,Ub,runNum,"F
resnelCosine (x) ","FresnelCosine_8.txt")

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PadeShamPoly3(@ (x) integral (@ (z) sin (z) ./z,0,x) ,gx, [1.3:0.1:20] ,Lb,Ub,runNum,"S
i (x) ", "Si_8.txt")
    PadeShamPoly3 (@ (x) 0.57721566+log (x) -integral (@ (z) (1-
cos (z)) ./z,0,x) ,gx, [0.5:0.1:20] ,Lb,Ub,runNum,"Ci (x) ", "Ci_8.txt")

PadeShamPoly3 (@ (x) asinh (x) ,gx, [0:0.1:100] ,Lb,Ub,runNum,"asinh (x) ", "asinh_8.tx
t")

PadeShamPoly3 (@ (x) acosh (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"acosh (x) ", "acosh_8.tx
t")

PadeShamPoly3 (@ (x) atanh (x) ,gx, [0:0.001:0.999] ,Lb,Ub,runNum,"atanh (x) ", "atanh_
8.txt")
end

if selIdx==0 || selIdx==9
    lstA = [0 0.5];
    lstB = [0.1 0.7];
    lstC = [0 0.7];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstC(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstC(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B,C) A+B*sqrt(i)+C*sqrt(i-1);

PadeShamPoly3 (@ (x) acos (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arccos (x) ", "arccos_9.txt
")

PadeShamPoly3 (@ (x) asin (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arcsin (x) ", "arcsin_9.txt
")

PadeShamPoly3 (@ (x) atan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arctan (x) ", "arctan_9.txt
")

PadeShamPoly3 (@ (x) sin (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"sin (x) ", "sin_9.txt
")

PadeShamPoly3 (@ (x) cos (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"cos (x) ", "cos_9.txt
")
    PadeShamPoly3 (@ (x) tan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"tan (x) ", "tan_9.txt")
    PadeShamPoly3 (@ (x) sinh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"sinh (x) ", "sinh_9.txt")
    PadeShamPoly3 (@ (x) cosh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"cosh (x) ", "cosh_9.txt")
    PadeShamPoly3 (@ (x) tanh (x) ,gx, [0:.01:3] ,Lb,Ub,runNum,"tanh (x) ", "tanh_9.txt")
    PadeShamPoly3 (@ (x) erf (x) ,gx, [0:.01:2.1] ,Lb,Ub,runNum,"erf (x) ", "erf_9.txt")
    PadeShamPoly3 (@ (x) exp (x) ,gx, [0:.01:2] ,Lb,Ub,runNum,"exp (x) ", "exp_9.txt")
    PadeShamPoly3 (@ (x) log (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"ln (x) ", "ln_9.txt")
    PadeShamPoly3 (@ (x) log10 (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"log (x) ", "log_9.txt")
    PadeShamPoly3 (@ (x) 10.^x,gx, [0:.01:1] ,Lb,Ub,runNum,"10^x", "pwr10_9.txt")

PadeShamPoly3 (@ (x) tinv (0.95,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.95,x) ", "ti
nv1_9.txt")

PadeShamPoly3 (@ (x) tinv (0.975,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.975,x) ", "
tinv2_9.txt")

PadeShamPoly3 (@ (x) log10 (gamma (x)) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"log10Gamma (x) "
,"log10Gamma_9.txt")

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PadeShamPoly3(@ (x) digamma (x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"digamma (x) ","digamm
a_9.txt")

PadeShamPoly3(@ (x) trigamma (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"trigamma (x) ","trig
amma_9.txt")

PadeShamPoly3(@ (x) besselj (0,x) ,gx, [2:0.1:30] ,Lb,Ub,runNum,"J0 (x) ","J0x_9.txt"
)

PadeShamPoly3(@ (x) besselj (1,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J1 (x) ","J1x_9.txt"
)

PadeShamPoly3(@ (x) besselj (2,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J2 (x) ","J2x_9.txt"
)

PadeShamPoly3(@ (x) besselj (3,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J3 (x) ","J3x_9.txt"
)

PadeShamPoly3(@ (x) besselj (4,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J4 (x) ","J4x_9.txt"
)

PadeShamPoly3(@ (x) besselj (5,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J5 (x) ","J5x_9.txt"
)

PadeShamPoly3(@ (x) integral (@ (z) sin (z.^2) ,0,x) ,gx, [1:0.01:5] ,Lb,Ub,runNum,"Fre
snelSine (x) ","FresnelSine_9.txt")

PadeShamPoly3(@ (x) integral (@ (z) cos (z.^2) ,0,x) ,gx, [0.5:0.01:5] ,Lb,Ub,runNum,"F
resnelCosine (x) ","FresnelCosine_9.txt")

PadeShamPoly3(@ (x) integral (@ (z) sin (z) ./z,0,x) ,gx, [1.3:0.1:20] ,Lb,Ub,runNum,"S
i (x) ","Si_9.txt")
    PadeShamPoly3(@ (x) 0.57721566+log (x) -integral (@ (z) (1-
cos (z)) ./z,0,x) ,gx, [0.5:0.1:20] ,Lb,Ub,runNum,"Ci (x) ","Ci_9.txt")

PadeShamPoly3(@ (x) asinh (x) ,gx, [0:0.1:100] ,Lb,Ub,runNum,"asinh (x) ","asinh_9.tx
t")

PadeShamPoly3(@ (x) acosh (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"acosh (x) ","acosh_9.tx
t")

PadeShamPoly3(@ (x) atanh (x) ,gx, [0:0.001:0.999] ,Lb,Ub,runNum,"atanh (x) ","atanh_
9.txt")
end

if selIdx==0 || selIdx==10
    lstA = [0 2.2];
    lstB = [0.1 3.5];
    lstOrder = [3 7];
    Lb = [lstA(1) lstB(1) lstOrder(1) lstOrder(1)];
    Ub = [lstA(2) lstB(2) lstOrder(2) lstOrder(2)];
    gx = @(i,A,B) A+B*log10(i)^4;

PadeShamPoly2(@ (x) acos (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arccos (x) ","arccos_10.tx
t")

```

```

PadeShamPoly2 (@ (x) asin (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arcsin (x) ","arcsin_10.txt")

PadeShamPoly2 (@ (x) atan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"arctan (x) ","arctan_10.txt")

PadeShamPoly2 (@ (x) sin (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"sin (x) ","sin_10.txt")

PadeShamPoly2 (@ (x) cos (x) ,gx, [0:pi/1000:pi/2] ,Lb,Ub,runNum,"cos (x) ","cos_10.txt")
    PadeShamPoly2 (@ (x) tan (x) ,gx, [0:.01:1] ,Lb,Ub,runNum,"tan (x) ","tan_10.txt")

PadeShamPoly2 (@ (x) sinh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"sinh (x) ","sinh_10.txt")

PadeShamPoly2 (@ (x) cosh (x) ,gx, [0:.01:5] ,Lb,Ub,runNum,"cosh (x) ","cosh_10.txt")

PadeShamPoly2 (@ (x) tanh (x) ,gx, [0:.01:3] ,Lb,Ub,runNum,"tanh (x) ","tanh_10.txt")
    PadeShamPoly2 (@ (x) erf (x) ,gx, [0:.01:2.1] ,Lb,Ub,runNum,"erf (x) ","erf_10.txt")
    PadeShamPoly2 (@ (x) exp (x) ,gx, [0:.01:2] ,Lb,Ub,runNum,"exp (x) ","exp_10.txt")
    PadeShamPoly2 (@ (x) log (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"ln (x) ","ln_10.txt")

PadeShamPoly2 (@ (x) log10 (x) ,gx, [1:.01:10] ,Lb,Ub,runNum,"log (x) ","log_10.txt")
    PadeShamPoly2 (@ (x) 10.^x,gx, [0:.01:1] ,Lb,Ub,runNum,"10^x","pwr10_10.txt")

PadeShamPoly2 (@ (x) tinv (0.95,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.95,x) ","tinv1_10.txt")

PadeShamPoly2 (@ (x) tinv (0.975,x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"tinv (0.975,x) ","tinv2_10.txt")

PadeShamPoly2 (@ (x) log10 (gamma (x)) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"log10Gamma (x) ","log10Gamma_10.txt")

PadeShamPoly2 (@ (x) digamma (x) ,gx, [2:0.1:100] ,Lb,Ub,runNum,"digamma (x) ","digamma_10.txt")

PadeShamPoly2 (@ (x) trigamma (x) ,gx, [1:0.1:100] ,Lb,Ub,runNum,"trigamma (x) ","trigamma_10.txt")

PadeShamPoly2 (@ (x) besselj (0,x) ,gx, [2:0.1:30] ,Lb,Ub,runNum,"J0 (x) ","J0x_10.txt")

PadeShamPoly2 (@ (x) besselj (1,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J1 (x) ","J1x_10.txt")

PadeShamPoly2 (@ (x) besselj (2,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J2 (x) ","J2x_10.txt")

PadeShamPoly2 (@ (x) besselj (3,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J3 (x) ","J3x_10.txt")

PadeShamPoly2 (@ (x) besselj (4,x) ,gx, [0:0.1:30] ,Lb,Ub,runNum,"J4 (x) ","J4x_10.txt")

```

```

PadeShamPoly2(@ (x)besselj(5,x),gx,[0:0.1:30],Lb,Ub,runNum,"J5(x)","J5x_10.txt")

PadeShamPoly2(@ (x)integral(@(z)sin(z.^2),0,x),gx,[1:0.01:5],Lb,Ub,runNum,"FresnelSine(x)","FresnelSine_10.txt")

PadeShamPoly2(@ (x)integral(@(z)cos(z.^2),0,x),gx,[0.5:0.01:5],Lb,Ub,runNum,"FresnelCosine(x)","FresnelCosine_10.txt")

PadeShamPoly2(@ (x)integral(@(z)sin(z)./z,0,x),gx,[1.3:0.1:20],Lb,Ub,runNum,"Si(x)","Si_10.txt")
    PadeShamPoly2(@ (x)0.57721566+log(x)-integral(@(z)(1-cos(z))./z,0,x),gx,[0.5:0.1:20],Lb,Ub,runNum,"Ci(x)","Ci_10.txt")

PadeShamPoly2(@ (x)asinh(x),gx,[0:0.1:100],Lb,Ub,runNum,"asinh(x)","asinh_10.txt")

PadeShamPoly2(@ (x)acosh(x),gx,[1:0.1:100],Lb,Ub,runNum,"acosh(x)","acosh_10.txt")

PadeShamPoly2(@ (x)atanh(x),gx,[0:0.001:0.999],Lb,Ub,runNum,"atanh(x)","atanh_10.txt")
end

toc;

% make some sound
playmynotes()

if bShutdown
    system('shutdown -s');
else
    fprintf("\n\nDone!\n\n");
end

```

The above listing has the following global and operational variables:

- The global variable `bUseDiary` is a Boolean flag used to tell the functions `PadeShamPoly2()` and `PadeShamPoly3()` whether you want to copy the screen output to diary text files.
- The global variable `bDeleteIfExists` is a Boolean flag used to tell the functions `PadeShamPoly2()` and `PadeShamPoly3()` whether you want to delete diary files if they exist.
- The variable `selIdx` allows you to select calculations for one of the seven groups (when set to the targeted group number) or all of the groups (when set to 0). I am using this scheme to reduce calculation time which can be done by working with a specific set of approximations.
- The Boolean variable `bShutdown` tells the Matlab script whether to shut down the computer when done.

When the script goAll is done with the calculation it executes the script playmynotes.m to play ten random sinusoidal notes. You can comment out the call to the audible script to keep the script quiet. I wrote the script for playmynotes.m based on the code I found in the fine book *Programming Fundamentals Using MATLAB* written by M. Weeks and published in 2020 by Mercury Learning. The book comes with a CD that contains all the Matlab source code and figures in that book. The source code for playmynotes.m is:

```
n = 10;
frq = 8192;
notes = 220 + fix(220*rand(n,1));

for i = 1:n
    znotes(i,1:frq) = 0.5*sin(2*pi*notes(i)*(0:frq-1)/frq);
end

for i = 1:n
    sound(znotes(i,1:frq));
    pause(1)
end
```

You can edit the script in goAll.m to perform Pade-Shammas polynomial fitting for other functions you are interested in or for other Shammas polynomial orders.

GENERAL COMMENTS ON RESULTS

The next sections show the results of fitting various common functions each with a variety of Pade-Shammas polynomials. There is a summary table for each fitted function showing the Shammas polynomial parameters, the F statistic, the AICc statistic, the root mean sum of errors squared, and orders of the fitted Pade-Shammas polynomials. I use the the root mean sum of errors squared to select the best Shammas polynomial powers. I have deliberately not included the coefficient of determination and its adjusted value in the table. Why? These values are very close to 1 and we are better served to look at other statistics. I have included the values for the F and AICc statistics, even though they are infinity and minus infinity.

 The next sections display the results of the Pade-Shammas polynomials. The regression coefficients for the $Q_n(x)$ and $D_m(x)$ Shammas polynomials appear as one long list labeled x1 and up. The results include the orders for the $Q_n(x)$ and $D_m(x)$ Shammas polynomials which should easily indicate with regression coefficients belong to $Q_n(x)$ Shammas polynomials and which one belong to $D_m(x)$ Shammas polynomials. The results also include the list of powers associated with

$Q_n(x)$ and $D_m(x)$ Shammas polynomials. Here is an example. Looking at the results for the inverse hyperbolic cosine:

Fitting $\text{acosh}(x)$ in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-7.65714007212127	0	-Inf	0
x1	33.3385052174594	0	Inf	0
x2	-57.1709436312584	0	-Inf	0
x3	49.1740680668373	0	Inf	0
x4	-22.7142392781885	0	-Inf	0
x5	5.89936499013805	0	Inf	0
x6	-0.936296960435223	0	-Inf	0
x7	0.0666816376160401	0	Inf	0
x8	2.76468166667437	0	Inf	0
x9	-2.5074894781991	0	-Inf	0
x10	0.742729123722548	0	Inf	0

Number of observations: 991, Error degrees of freedom: 980

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957758438	990	0.812268644200442		
Model	804.145957758438	10	80.4145957758438	Inf	0
Residual	0	980	0		

$A = 0.000000$, $B = 0.152812$

order $Q(x) = 7.000000$, order $D(x) = 3.000000$

List of powers for $Q(x)$: [0.152812, 0.305624, 0.458436, 0.611248, 0.764060, 0.916872, 1.069684]

List of powers for $D(x)$: [0.152812, 0.305624, 0.458436]

Fitting $\text{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 9.351430e-07

Corrected MSS of errors squared = 1.322492e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

First look at the orders of the $Q_n(x)$ and $D_m(x)$ Shammas polynomials. They are 7 and 3, respectively. Then look at the list of regression coefficients labeled x1

through x_{10} . Since $Q_n(x)$ has the order of seven, its regression coefficients are the ones labeled x_1 through x_7 . The remaining three regression coefficients are labeled x_8 through x_{10} . You can find the list of the $Q_n(x)$ and $D_m(x)$ Shammas polynomials right below the line showing the polynomial orders.

RESULTS FOR THE INVERSE HYPERBOLIC COSINE

Using Power $A+B*i$

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-7.65714007212127	0	-Inf	0
x1	33.3385052174594	0	Inf	0
x2	-57.1709436312584	0	-Inf	0
x3	49.1740680668373	0	Inf	0
x4	-22.7142392781885	0	-Inf	0
x5	5.89936499013805	0	Inf	0
x6	-0.936296960435223	0	-Inf	0
x7	0.0666816376160401	0	Inf	0
x8	2.76468166667437	0	Inf	0
x9	-2.5074894781991	0	-Inf	0
x10	0.742729123722548	0	Inf	0

Number of observations: 991, Error degrees of freedom: 980

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957758438	990	0.812268644200442		
Model	804.145957758438	10	80.4145957758438	Inf	0
Residual	0	980	0		

$A = 0.000000$, $B = 0.152812$

order $Q(x) = 7.000000$, order $D(x) = 3.000000$

List of powers for $Q(x)$: [0.152812, 0.305624, 0.458436, 0.611248, 0.764060, 0.916872, 1.069684]

List of powers for $D(x)$: [0.152812, 0.305624, 0.458436]

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = $9.351430e-07$

Corrected MSS of errors squared = $1.322492e-06$

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = $-1.000000e+99$

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting $\text{acosh}(x)$ in range (1.000000, 100.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.231597869387483	0	Inf	0
x1	-0.0996254818328427	0	-Inf	0
x2	3.15921408159802	0	Inf	0
x3	-15.0525332981518	0	-Inf	0
x4	18.9287330553568	0	Inf	0
x5	0	0	NaN	NaN
x6	-7.16738622611794	0	-Inf	0
x7	0.00553988904295155	0	Inf	0
x8	0.593896562343958	0	Inf	0
x9	-10.5430509644483	0	-Inf	0
x10	42.5858817192991	0	Inf	0
x11	-51.0039089534056	0	-Inf	0
x12	0	0	NaN	NaN
x13	19.3616369653409	0	Inf	0

Number of observations: 991, Error degrees of freedom: 979

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957758685	990	0.812268644200692		
Model	804.145957758685	11	73.1041779780622	Inf	0
Residual	0	979	0		

A = 0.000000, B = 0.694529

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.694529, 0.347265, 0.231510, 0.173632, 0.138906, 0.115755]

List of powers for D(x): [0.694529, 0.347265, 0.231510, 0.173632, 0.138906, 0.115755, 0.099218]

Fitting $\text{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 9.444387e-07

Corrected MSS of errors squared = 1.335638e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}$

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.607068943091106	0	-Inf	0
x1	6.08742152279404	0	Inf	0
x2	-8.67635920106154	0	-Inf	0
x3	0	0	NaN	NaN
x4	0	0	NaN	NaN
x5	5.54807755286479	0	Inf	0
x6	0	0	NaN	NaN
x7	-2.35207093204665	0	-Inf	0
x8	53.1300786375054	0	Inf	0
x9	-354.890621972371	0	-Inf	0
x10	1007.97937579804	0	Inf	0
x11	-1510.59156519966	0	-Inf	0
x12	1257.16525789763	0	Inf	0
x13	-551.795117408575	0	-Inf	0
x14	100.002591640815	0	Inf	0

Number of observations: 991, Error degrees of freedom: 979

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957753158	990	0.812268644195109		
Model	804.145957753158	11	73.1041779775598	Inf	0
Residual	0	979	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 1.039883e-06

Corrected MSS of errors squared = 1.470616e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)
Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-20.5711606525456	0	-Inf	0
x1	684.804235361524	0	Inf	0
x2	-2988.58926517397	0	-Inf	0
x3	4970.82413045046	0	Inf	0
x4	-3256.71758171713	0	-Inf	0
x5	0	0	NaN	NaN
x6	864.8317433432	0	Inf	0
x7	-254.582101610785	0	-Inf	0
x8	16.6651672231044	0	Inf	0
x9	-46.8023871335487	0	-Inf	0
x10	47.7017367008104	0	Inf	0
x11	-16.5645243526515	0	-Inf	0

Number of observations: 991, Error degrees of freedom: 980

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.14595775528	990	0.812268644197253		
Model	804.14595775528	10	80.414595775528	Inf	0
Residual	0	980	0		

A = 0.002999, B = 0.106413, C = 0.000000

order Q(x) = 7.000000, order D(x) = 4.000000

List of powers for Q(x): [0.109412, 0.153490, 0.187312, 0.215825, 0.240946, 0.263656, 0.284541]

List of powers for D(x): [0.109412, 0.153490, 0.187312, 0.215825]

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 6.013680e-07

Corrected MSS of errors squared = 1.041600e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.09556452004188	0	-Inf	0
x1	3.60785831589973	0	Inf	0
x2	-1.7791951229907	0	-Inf	0
x3	0	0	NaN	NaN
x4	-0.920144144697184	0	-Inf	0
x5	2.10152193433366	0	Inf	0
x6	-0.850246205258021	0	-Inf	0
x7	-0.064230254475905	0	-Inf	0
x8	1.47356102391029	0	Inf	0
x9	-0.659740259855103	0	-Inf	0
x10	0.611420382836852	0	Inf	0
x11	-0.648199169457652	0	-Inf	0
x12	0.0986372241645598	0	Inf	0
x13	0.120358013372217	0	Inf	0
x14	0.00394012458783382	0	Inf	0

Number of observations: 991, Error degrees of freedom: 977

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957758919	990	0.812268644200928		
Model	804.145957758919	13	61.8573813660707	Inf	0
Residual	0	977	0		

A = 0.000000, B = 0.117525, C = 0.202333

order $Q(x)$ = 7.000000, order $D(x)$ = 7.000000

List of powers for $Q(x)$: [0.117525, 0.437384, 0.757243, 1.077102, 1.396960, 1.716819, 2.036678]

List of powers for $D(x)$: [0.117525, 0.437384, 0.757243, 1.077102, 1.396960, 1.716819, 2.036678]

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 3.140402e-07

Corrected MSS of errors squared = 5.439335e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting $\text{acosh}(x)$ in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.275356235819997	0	Inf	0
x1	-0.126570823822444	0	-Inf	0
x2	0.395272677752026	0	Inf	0
x3	-0.544058089263608	0	-Inf	0
x4	0.00739665652561899	0	Inf	0
x5	-0.0253692173058622	0	-Inf	0
x6	0.340191283343167	0	Inf	0
x7	0.942540619884829	0	Inf	0
x8	-0.384900077436039	0	-Inf	0
x9	0.147466913036586	0	Inf	0
x10	-0.0273230096580313	0	-Inf	0

Number of observations: 991, Error degrees of freedom: 980

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.14595775857	990	0.812268644200576		
Model	804.14595775857	10	80.4145957758571	Inf	0
Residual	0	980	0		

A = 0.259057, B = 0.100000, C = 0.242635

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.359057, 0.216422, 0.073787]

List of powers for D(x): [0.359057, 0.216422, 0.073787, -0.068847, -0.211482, -0.354117, -0.496752]

Fitting $\text{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 3.337864e-07

Corrected MSS of errors squared = 5.781351e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)
Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-6.54897489320197	0	-Inf	0
x1	16.342927014663	0	Inf	0
x2	-36.8062381819599	0	-Inf	0
x3	31.6616161304753	0	Inf	0
x4	25.9511249025635	0	Inf	0
x5	-49.1379437435576	0	-Inf	0
x6	18.0964037761085	0	Inf	0
x7	0.441085003257686	0	Inf	0
x8	1.5416459144726	0	Inf	0
x9	2.09159169895671	0	Inf	0
x10	-9.00371841449092	0	-Inf	0
x11	8.59601062646321	0	Inf	0
x12	-0.910368492982139	0	-Inf	0
x13	-1.31572962012511	0	-Inf	0

Number of observations: 991, Error degrees of freedom: 977

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957762918	990	0.812268644204967		
Model	804.145957762918	13	61.8573813663783	Inf	0
Residual	0	977	0		

A = 0.092614, B = 0.103277, C = 0.360000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.195891, 0.659168, 0.911561, 1.129260, 1.328998, 1.517259, 1.697368]

List of powers for D(x): [0.195891, 0.659168, 0.911561, 1.129260, 1.328998, 1.517259]

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 1.443199e-07

Corrected MSS of errors squared = 2.499693e-07

R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)
 Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.31403543248461	0	-Inf	0
x1	16.0774547253833	0	Inf	0
x2	-23.8127816751693	0	-Inf	0
x3	0	0	NaN	NaN
x4	18.8896833972755	0	Inf	0
x5	-8.84032101445198	0	-Inf	0
x6	14.4171435946962	0	Inf	0
x7	-59.529967650162	0	-Inf	0
x8	119.653891805069	0	Inf	0
x9	-134.281132174544	0	-Inf	0
x10	85.9896074153046	0	Inf	0
x11	-29.5014650868546	0	-Inf	0
x12	4.25190952858039	0	Inf	0

Number of observations: 991, Error degrees of freedom: 979

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957758762	990	0.81226864420077		
Model	804.145957758762	11	73.1041779780693	Inf	0
Residual	0	979	0		

A = 0.000000, B = 0.131595, C = 0.021625

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.131595, 0.207728, 0.271178, 0.328063, 0.380753]

List of powers for D(x): [0.131595, 0.207728, 0.271178, 0.328063, 0.380753, 0.430463, 0.477915]

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 1.244453e-05

Corrected MSS of errors squared = 2.155455e-05

```

R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

```

Fitting acosh(x) in range (1.000000, 100.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i}+C\sqrt{i-1}$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.21440131813467	0	-Inf	0
x1	8.44989853382429	0	Inf	0
x2	-19.7207171795824	0	-Inf	0
x3	30.4139749047284	0	Inf	0
x4	-19.1645195476864	0	-Inf	0
x5	4.23576462518046	0	Inf	0
x6	1.40604225858568	0	Inf	0
x7	0.272780520197806	0	Inf	0
x8	-2.46763455369908	0	-Inf	0
x9	2.80221201859251	0	Inf	0
x10	-1.23473872947692	0	-Inf	0
x11	0.247530179334794	0	Inf	0
x12	-0.0266405726177113	0	-Inf	0

Number of observations: 991, Error degrees of freedom: 978

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957758374	990	0.812268644200377		
Model	804.145957758374	12	67.0121631465311	Inf	0
Residual	0	978	0		

A = 0.061638, B = 0.100000, C = 0.483502

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.161638, 0.686561, 0.918618, 1.099088, 1.252249]

List of powers for D(x): [0.161638, 0.686561, 0.918618, 1.099088, 1.252249, 1.387730, 1.510547]

Fitting acosh(x) in range (1.000000, 100.000000)

MSS of errors squared = 4.964112e-07

Corrected MSS of errors squared = 8.598094e-07

R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)
 Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.1836711556097	0	-Inf	0
x1	6.49506796220532	0	Inf	0
x2	0	0	NaN	NaN
x3	-6.27076797669487	0	-Inf	0
x4	2.25765939612071	0	Inf	0
x5	-0.298288225429545	0	-Inf	0
x6	15.5892519011767	0	Inf	0
x7	-16.7848874769485	0	-Inf	0
x8	2.5360366711197	0	Inf	0
x9	-0.392852832656103	0	-Inf	0
x10	0.0548718801276958	0	Inf	0
x11	-0.00250109164412567	0	-Inf	0
x12	7.07847148139394e-05	0	Inf	0

Number of observations: 991, Error degrees of freedom: 979

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	804.145957758614	990	0.81226864420062		
Model	804.145957758614	11	73.1041779780558	Inf	0
Residual	0	979	0		

A = 0.080462, B = 1.763951

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.080462, 0.094947, 0.171873, 0.312225, 0.501499]

List of powers for D(x): [0.080462, 0.094947, 0.171873, 0.312225, 0.501499, 0.727221, 0.980197]

Fitting $\operatorname{acosh}(x)$ in range (1.000000, 100.000000)

MSS of errors squared = 5.406037e-07

Corrected MSS of errors squared = 7.645291e-07

R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the inverse hyperbolic cosine function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AIC</i> <i>c</i>	<i>MSSE</i>	<i>Q/D</i> <i>Orders</i>
$A+B*i+C*\sqrt{i-1}$	0.092	0.103	0.36	Inf	-Inf	1.44319900E-07	7, 6
$A+B*i+C*(i-1)$	0	0.117	0.202	Inf	-Inf	3.14040200E-07	7, 7
$A+B*i-C*(i-1)$	0.259	0.1	0.242	Inf	-Inf	3.33786400E-07	3, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.061	0.1	0.483	Inf	-Inf	4.96411200E-07	5, 7
$A+B*\log_{10}(i)^4$	0.08	1.763		Inf	-Inf	5.40603700E-07	5, 7
$A+B*\sqrt{i+C}$	0.002	0.106	0	Inf	-Inf	6.01368000E-07	7, 4
$A+B*i$	0	0.152		Inf	-Inf	9.35143000E-07	7, 3
$A+B/i$	0	0.694		Inf	-Inf	9.44438700E-07	6, 7
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	1.03988300E-06	7, 7
$A+B*\sqrt{i}+C*(i-1)$	0	0.131	0.021	Inf	-Inf	1.24445300E-05	5, 7

RESULTS FOR THE ARC COSINE

Using Power $A+B*i$ Fitting $\arccos(x)$ in range (0.000000, 1.000000)Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679126	0	Inf	0
x1	-43.456724171806	0	-Inf	0
x2	-90.6908551599143	0	-Inf	0
x3	761.486649383669	0	Inf	0
x4	-1490.18831588691	0	-Inf	0
x5	1380.95075672938	0	Inf	0
x6	-639.703606198682	0	-Inf	0
x7	120.031298977697	0	Inf	0
x8	28.91069463147	0	Inf	0
x9	46.2094276835702	0	Inf	0
x10	-440.593949332277	0	-Inf	0
x11	859.550127319194	0	Inf	0
x12	-780.442097674202	0	-Inf	0
x13	350.674361491544	0	Inf	0
x14	-63.3085648227558	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 86

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812058	100	0.148796738812058		
Model	14.8796738812058	14	1.06283384865756	Inf	0
Residual	0	86	0		

A = 0.357763, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.457763, 0.557763, 0.657763, 0.757763, 0.857763, 0.957763, 1.057763]

List of powers for D(x): [0.457763, 0.557763, 0.657763, 0.757763, 0.857763, 0.957763, 1.057763]

Fitting $\arccos(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 1.131438e-05

Corrected MSS of errors squared = 1.600096e-05

```
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Using Power A+B/i

```
Fitting arccos(x) in range (0.000000, 1.000000)
Pade-Shammas polynomial power is A+B/i
```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679476	0	Inf	0
x1	-0.00341315302811853	0	-Inf	0
x2	-0.521909287043382	0	-Inf	0
x3	12.6914812888227	0	Inf	0
x4	-57.6689726203883	0	-Inf	0
x5	73.0645069547046	0	Inf	0
x6	0	0	NaN	NaN
x7	-29.1324895098621	0	-Inf	0
x8	0.00362027529560987	0	Inf	0
x9	0.108918902499885	0	Inf	0
x10	-4.18917781987594	0	-Inf	0
x11	15.8191491088084	0	Inf	0
x12	0	0	NaN	NaN
x13	-45.4368590516594	0	-Inf	0
x14	34.6943485848124	0	Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812107	100	0.148796738812107		
Model	14.8796738812107	12	1.23997282343423	Inf	0
Residual	0	88	0		

A = 0.000000, B = 0.229127

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.229127, 0.114564, 0.076376, 0.057282, 0.045825, 0.038188, 0.032732]

List of powers for D(x): [0.229127, 0.114564, 0.076376, 0.057282, 0.045825, 0.038188, 0.032732]

Fitting $\arccos(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 4.514345e-05

Corrected MSS of errors squared = 6.384248e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}$

Fitting $\arccos(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679329	0	Inf	0
x1	-815.730583980435	0	-Inf	0
x2	5019.38300557651	0	Inf	0
x3	-10714.6281925562	0	-Inf	0
x4	8772.40840730565	0	Inf	0
x5	0	0	NaN	NaN
x6	-3669.41578495688	0	-Inf	0
x7	1406.4123522846	0	Inf	0
x8	537.876186938449	0	Inf	0
x9	-3435.92987718983	0	-Inf	0
x10	7883.82331234217	0	Inf	0
x11	-7831.32359060907	0	-Inf	0
x12	2491.26500427837	0	Inf	0
x13	933.743599271783	0	Inf	0
x14	-578.454635032125	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812314	100	0.148796738812314		
Model	14.8796738812314	13	1.14459029855626	Inf	0
Residual	0	87	0		

A = 0.134094, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.234094, 0.275515, 0.307299, 0.334094, 0.357700, 0.379042, 0.398669]

List of powers for D(x): [0.234094, 0.275515, 0.307299, 0.334094, 0.357700, 0.379042, 0.398669]

Fitting arccos(x) in range (0.000000, 1.000000)

MSS of errors squared = 8.918658e-05

Corrected MSS of errors squared = 1.261289e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting arccos(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632686646	0	Inf	0
x1	-2469.68948023539	0	-Inf	0
x2	14842.6884644499	0	Inf	0
x3	-31401.2759957678	0	-Inf	0
x4	25711.5074530851	0	Inf	0
x5	0	0	NaN	NaN
x6	-10925.6555420795	0	-Inf	0
x7	4240.85430422263	0	Inf	0
x8	1593.61470050916	0	Inf	0
x9	-9819.19300680217	0	-Inf	0
x10	22061.371290224	0	Inf	0
x11	-21727.9402450845	0	-Inf	0
x12	7074.93331314213	0	Inf	0
x13	2323.75939078687	0	Inf	0
x14	-1505.54544827077	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812256	100	0.148796738812256		
Model	14.8796738812256	13	1.14459029855582	Inf	0
Residual	0	87	0		

```

A = 0.645185, B = 0.247036, C = 0.236532
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.919888, 1.014629, 1.089612, 1.153656, 1.210490, 1.262110,
1.309733]
List of powers for D(x): [0.919888, 1.014629, 1.089612, 1.153656, 1.210490, 1.262110,
1.309733]
Fitting arccos(x) in range (0.000000, 1.000000)
MSS of errors squared = 1.174722e-05
Corrected MSS of errors squared = 2.034678e-05
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting arccos(x) in range (0.000000, 1.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679508	0	Inf	0
x1	-9.69232271284722	0	-Inf	0
x2	25.226325070692	0	Inf	0
x3	-37.7970473857247	0	-Inf	0
x4	37.2962303498598	0	Inf	0
x5	-24.9910836889691	0	-Inf	0
x6	10.3083064185215	0	Inf	0
x7	-1.92120437841514	0	-Inf	0
x8	6.16809710507429	0	Inf	0
x9	-16.0133742957663	0	-Inf	0
x10	23.5921508689123	0	Inf	0
x11	-22.0836757929774	0	-Inf	0
x12	13.3564032572472	0	Inf	0
x13	-4.77140691190244	0	-Inf	0
x14	0.751806500002245	0	Inf	0

Number of observations: 101, Error degrees of freedom: 86

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	14.8796738812114	100	0.148796738812114		
Model	14.8796738812114	14	1.06283384865796	Inf	0
Residual	0	86	0		

A = 0.000000, B = 0.270000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.270000, 0.540000, 0.810000, 1.080000, 1.350000, 1.620000, 1.890000]

List of powers for D(x): [0.270000, 0.540000, 0.810000, 1.080000, 1.350000, 1.620000, 1.890000]

Fitting arccos(x) in range (0.000000, 1.000000)

MSS of errors squared = 6.357835e-06

Corrected MSS of errors squared = 1.101209e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting arccos(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679406	0	Inf	0
x1	-2202.29288484209	0	-Inf	0
x2	5899.7624467663	0	Inf	0
x3	0	0	NaN	NaN
x4	-16350.1688653628	0	-Inf	0
x5	22816.573075345	0	Inf	0
x6	-12897.7451044397	0	-Inf	0
x7	2732.30053620819	0	Inf	0
x8	1453.53393850745	0	Inf	0
x9	-4195.03491075626	0	-Inf	0
x10	1547.56555802709	0	Inf	0
x11	7520.52216900646	0	Inf	0
x12	-11530.0820415162	0	-Inf	0
x13	6581.42159526681	0	Inf	0
x14	-1376.92631393491	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738813592	100	0.148796738813592		
Model	14.8796738813592	13	1.1445902985661	Inf	0
Residual	0	87	0		

A = 0.707567, B = 0.209068, C = 0.139269

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.916635, 0.986434, 1.056233, 1.126032, 1.195832, 1.265631, 1.335430]

List of powers for D(x): [0.916635, 0.986434, 1.056233, 1.126032, 1.195832, 1.265631, 1.335430]

Fitting arccos(x) in range (0.000000, 1.000000)

MSS of errors squared = 1.142964e-05

Corrected MSS of errors squared = 1.979672e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting arccos(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679578	0	Inf	0
x1	-3.03417517678718	0	-Inf	0
x2	5.21144570913736	0	Inf	0
x3	-8.76621233366959	0	-Inf	0
x4	8.9559820546881	0	Inf	0
x5	-5.22160063961711	0	-Inf	0
x6	1.28376405983404	0	Inf	0
x7	1.93150670648911	0	Inf	0
x8	-3.30138349646443	0	-Inf	0
x9	5.0218038804782	0	Inf	0
x10	-3.96606625478548	0	-Inf	0
x11	1.34958012375382	0	Inf	0
x12	0.089634059895191	0	Inf	0
x13	-0.125076261160044	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812115	100	0.148796738812115		
Model	14.8796738812115	13	1.14459029855473	Inf	0
Residual	0	87	0		

A = 0.046510, B = 0.132984, C = 0.360000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.179493, 0.672477, 0.954577, 1.201982, 1.431427, 1.649396]

List of powers for D(x): [0.179493, 0.672477, 0.954577, 1.201982, 1.431427, 1.649396, 1.859211]

Fitting arccos(x) in range (0.000000, 1.000000)

MSS of errors squared = 1.150549e-05

Corrected MSS of errors squared = 1.992809e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting arccos(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679466	0	Inf	0
x1	-29.3088653274174	0	-Inf	0
x2	126.530904668925	0	Inf	0
x3	-253.090128538486	0	-Inf	0
x4	268.517863976157	0	Inf	0
x5	-146.876360615805	0	-Inf	0
x6	32.655789510409	0	Inf	0
x7	18.6624486821928	0	Inf	0
x8	-80.7493867261917	0	-Inf	0
x9	163.246976520823	0	Inf	0
x10	-180.001898281916	0	-Inf	0
x11	110.068534608595	0	Inf	0
x12	-34.2593518249202	0	-Inf	0
x13	4.03267572214619	0	Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812114	100	0.148796738812114		
Model	14.8796738812114	13	1.14459029855472	Inf	0
Residual	0	87	0		

A = 0.000000, B = 0.400000, C = 0.031418

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.400000, 0.597104, 0.755657, 0.894255, 1.020100, 1.136887]

List of powers for D(x): [0.400000, 0.597104, 0.755657, 0.894255, 1.020100, 1.136887, 1.246810]

Fitting arccos(x) in range (0.000000, 1.000000)

MSS of errors squared = 1.542243e-05

Corrected MSS of errors squared = 2.671243e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

Fitting arccos(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679402	0	Inf	0
x1	-44.3734955639346	0	-Inf	0
x2	207.988045656669	0	Inf	0
x3	-130.706372369087	0	-Inf	0
x4	-567.378467880046	0	-Inf	0
x5	1084.89287948406	0	Inf	0
x6	-721.797769962765	0	-Inf	0
x7	169.804384309032	0	Inf	0
x8	30.0366506908317	0	Inf	0
x9	-181.784773741438	0	-Inf	0
x10	355.352350367167	0	Inf	0
x11	-285.553792926247	0	-Inf	0
x12	82.9495652290495	0	Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812183	100	0.148796738812183		
Model	14.8796738812183	12	1.23997282343486	Inf	0
Residual	0	88	0		

A = 0.113229, B = 0.100000, C = 0.049748

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.213229, 0.304398, 0.356788, 0.399394, 0.436331, 0.469417, 0.499660]

List of powers for D(x): [0.213229, 0.304398, 0.356788, 0.399394, 0.436331]

Fitting arccos(x) in range (0.000000, 1.000000)

MSS of errors squared = 5.058315e-05

Corrected MSS of errors squared = 8.761259e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting arccos(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57079632679486	0	Inf	0
x1	-28.5348974072509	0	-Inf	0
x2	28.5466780947089	0	Inf	0
x3	-1.16019860110646	0	-Inf	0
x4	-0.422378413545505	0	-Inf	0
x5	18.6659921378593	0	Inf	0
x6	-18.9159549861201	0	-Inf	0
x7	1.17905116831604	0	Inf	0
x8	-0.134343486460277	0	-Inf	0
x9	0.340071672411207	0	Inf	0
x10	-0.166991378804352	0	-Inf	0
x11	0.0321732082647812	0	Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812115	100	0.148796738812115		
Model	14.8796738812115	11	1.35269762556469	Inf	0
Residual	0	89	0		

A = 0.199118, B = 2.632339

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.199118, 0.220734, 0.335531, 0.544978]

List of powers for D(x): [0.199118, 0.220734, 0.335531, 0.544978, 0.827430, 1.164274, 1.541790]

Fitting arccos(x) in range (0.000000, 1.000000)

MSS of errors squared = 1.789853e-05

Corrected MSS of errors squared = 2.531234e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the arc cosine function. The results are sorted in ascending order using the MSSE values.

Power	A	B	C	F	AIC c	MSSE	Q/D Orders
A+B*i+C*(i-1)	0	0.27	0	Inf	-Inf	6.35783500E-06	7, 7
A+B*i	0.357	0.1		Inf	-Inf	1.13143800E-05	7, 7
A+B*i-C*(i-1)	0.707	0.209	0.139	Inf	-Inf	1.14296400E-05	7, 7
A+B*i+C*sqrt(i-1)	0.046	0.132	0.36	Inf	-Inf	1.15054900E-05	6, 7
A+B*sqrt(i+C)	0.645	0.247	0.236	Inf	-Inf	1.17472200E-05	7, 7
A+B*sqrt(i)+C*(i-1)	0	0.4	0.031	Inf	-Inf	1.54224300E-05	6, 7
A+B*log10(i)^4	0.199	2.632		Inf	-Inf	1.78985300E-05	4, 7
A+B/i	0	0.229		Inf	-Inf	4.51434500E-05	7, 7
A+B*sqrt(i)+C*sqrt(i-1)	0.113	0.1	0.049	Inf	-Inf	5.05831500E-05	7, 5
A+B*sqrt(i)	0.134	0.1		Inf	-Inf	8.91865800E-05	7, 7

RESULTS FOR THE ARC SINE

Using Power $A+B*i$

Fitting $\arcsin(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.61596085506494e-14	0	Inf	0
x1	0.00271322209040669	0	Inf	0
x2	-0.0212366191302404	0	-Inf	0
x3	0.0634094226772156	0	Inf	0
x4	-0.0877803816537527	0	-Inf	0
x5	0.0567467471111257	0	Inf	0
x6	-0.0138523636686379	0	-Inf	0
x7	6.52409971196353	0	Inf	0
x8	-18.5789264227971	0	-Inf	0
x9	29.9078447561642	0	Inf	0
x10	-29.3538693093213	0	-Inf	0
x11	17.5401285350343	0	Inf	0
x12	-5.90001864657175	0	-Inf	0
x13	0.860741358059224	0	Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812115	100	0.148796738812115		
Model	14.8796738812115	13	1.14459029855473	Inf	0
Residual	0	87	0		

$A = 0.000000$, $B = 0.117349$

order $Q(x) = 6.000000$, order $D(x) = 7.000000$

List of powers for $Q(x)$: [0.117349, 0.234697, 0.352046, 0.469394, 0.586743, 0.704091]

List of powers for $D(x)$: [0.117349, 0.234697, 0.352046, 0.469394, 0.586743, 0.704091, 0.821440]

Fitting $\arcsin(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 2.649912e-05

Corrected MSS of errors squared = 3.747541e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting arcsin(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-5.97981254580512e-14	0	-Inf	0
x1	-0.00619394340410247	0	-Inf	0
x2	1.03365133798345	0	Inf	0
x3	-16.8768890269518	0	-Inf	0
x4	86.8192161829786	0	Inf	0
x5	-187.812426503391	0	-Inf	0
x6	179.796020068434	0	Inf	0
x7	-62.9533780420367	0	-Inf	0
x8	0.0529008435447592	0	Inf	0
x9	-3.48547976089073	0	-Inf	0
x10	26.55974564912	0	Inf	0
x11	-55.4127993336421	0	-Inf	0
x12	33.2856325549378	0	Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812122	100	0.148796738812122		
Model	14.8796738812122	12	1.23997282343435	Inf	0
Residual	0	88	0		

A = 0.015819, B = 0.484141

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.499960, 0.257890, 0.177200, 0.136855, 0.112648, 0.096510, 0.084982]

List of powers for D(x): [0.499960, 0.257890, 0.177200, 0.136855, 0.112648]

Fitting arcsin(x) in range (0.000000, 1.000000)

MSS of errors squared = 5.836815e-05

Corrected MSS of errors squared = 8.254502e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}$

Fitting $\arcsin(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	6.84391921000774e-13	0	Inf	0
x1	-0.496274642089966	0	-Inf	0
x2	7.99341760128281	0	Inf	0
x3	-41.5026712263319	0	-Inf	0
x4	99.9326393878796	0	Inf	0
x5	-123.58340338145	0	-Inf	0
x6	76.3878928249908	0	Inf	0
x7	-18.731600544699	0	-Inf	0
x8	41.0987188690704	0	Inf	0
x9	-226.010525777887	0	-Inf	0
x10	514.869427737801	0	Inf	0
x11	-584.651444815693	0	-Inf	0
x12	329.245108538796	0	Inf	0
x13	-73.5512845645559	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812096	100	0.148796738812096		
Model	14.8796738812096	13	1.14459029855459	Inf	0
Residual	0	87	0		

A = 0.000000, B = 0.222740

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.222740, 0.315003, 0.385798, 0.445481, 0.498063, 0.545600, 0.589316]

List of powers for D(x): [0.222740, 0.315003, 0.385798, 0.445481, 0.498063, 0.545600]

Fitting $\arcsin(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 8.047159e-06

Corrected MSS of errors squared = 1.138040e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting $\arcsin(x)$ in range (0.000000, 1.000000)
 Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.91041906504263e-10	0	Inf	0
x1	0.00138593782211927	0	Inf	0
x2	-0.00321106261770551	0	-Inf	0
x3	0.0018250289133776	0	Inf	0
x4	6988.9438620978	0	Inf	0
x5	-54152.0888449538	0	-Inf	0
x6	169205.910717085	0	Inf	0
x7	-275131.455865906	0	-Inf	0
x8	246874.299036357	0	Inf	0
x9	-116343.076687597	0	-Inf	0
x10	22558.4677829779	0	Inf	0

Number of observations: 101, Error degrees of freedom: 90

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738784001	100	0.148796738784001		
Model	14.8796738784001	10	1.48796738784001	Inf	0
Residual	0	90	0		

A = 0.071412, B = 0.100000, C = 1.526897

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.230374, 0.259212, 0.284177]

List of powers for D(x): [0.230374, 0.259212, 0.284177, 0.306505, 0.326890, 0.345764, 0.363420]

Fitting $\arcsin(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 1.346224e-04

Corrected MSS of errors squared = 2.331728e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting arcsin(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-9.75594320238577e-13	0	-Inf	0
x1	0.000434366159101534	0	Inf	0
x2	-0.0137837116367913	0	-Inf	0
x3	0.44204459598833	0	Inf	0
x4	-1.16948463964352	0	-Inf	0
x5	1.06785711484264	0	Inf	0
x6	-0.327064228188532	0	-Inf	0
x7	2.00595464225188	0	Inf	0
x8	-1.42501572911665	0	-Inf	0
x9	0.270423271645076	0	Inf	0
x10	0.512608076134365	0	Inf	0
x11	-0.663070570206695	0	-Inf	0
x12	0.382593044855972	0	Inf	0
x13	-0.0834949625865144	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812115	100	0.148796738812115		
Model	14.8796738812115	13	1.14459029855473	Inf	0
Residual	0	87	0		

A = 0.000000, B = 0.201107, C = 0.140871

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.201107, 0.543085, 0.885063, 1.227041, 1.569020, 1.910998]

List of powers for D(x): [0.201107, 0.543085, 0.885063, 1.227041, 1.569020, 1.910998, 2.252976]

Fitting arcsin(x) in range (0.000000, 1.000000)

MSS of errors squared = 1.129378e-05

Corrected MSS of errors squared = 1.956141e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting arcsin(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.10705041628771e-11	0	-Inf	0
x1	0.0113057193815671	0	Inf	0
x2	-0.18943732794813	0	-Inf	0
x3	1.65387820050886	0	Inf	0
x4	-3.80510692180437	0	-Inf	0
x5	3.35185536084946	0	Inf	0
x6	-1.0224932006763	0	-Inf	0
x7	8.30662744927207	0	Inf	0
x8	-19.8123424968093	0	-Inf	0
x9	23.817737606036	0	Inf	0
x10	-16.9525035190402	0	-Inf	0
x11	6.82001273647953	0	Inf	0
x12	-1.17953294079936	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812112	100	0.148796738812112		
Model	14.8796738812112	12	1.23997282343427	Inf	0
Residual	0	88	0		

A = 0.162267, B = 0.248338, C = 0.000000

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.410605, 0.658943, 0.907281, 1.155620, 1.403958, 1.652296]

List of powers for D(x): [0.410605, 0.658943, 0.907281, 1.155620, 1.403958, 1.652296]

Fitting arcsin(x) in range (0.000000, 1.000000)

MSS of errors squared = 6.338664e-06

Corrected MSS of errors squared = 1.097889e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting arcsin(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.37557526297789e-13	0	Inf	0
x1	-0.000279306061204712	0	-Inf	0
x2	0.0237706221485688	0	Inf	0
x3	1.59007330008951	0	Inf	0
x4	-5.64235191662946	0	-Inf	0
x5	7.23881894183045	0	Inf	0
x6	-4.02678503191755	0	-Inf	0
x7	0.816754752186117	0	Inf	0
x8	5.32265639803189	0	Inf	0
x9	-8.94833848057112	0	-Inf	0
x10	6.60586771142996	0	Inf	0
x11	-1.91982127102818	0	-Inf	0
x12	-0.627946113303209	0	-Inf	0
x13	0.79600113384975	0	Inf	0
x14	-0.228420245293392	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 86

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812116	100	0.148796738812116		
Model	14.8796738812116	14	1.06283384865797	Inf	0
Residual	0	86	0		

A = 0.124825, B = 0.312658, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.437483, 0.750141, 1.062799, 1.375458, 1.688116, 2.000774, 2.313432]

List of powers for D(x): [0.437483, 0.750141, 1.062799, 1.375458, 1.688116, 2.000774, 2.313432]

Fitting arcsin(x) in range (0.000000, 1.000000)

MSS of errors squared = 2.937046e-06

Corrected MSS of errors squared = 5.087114e-06

R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B\sqrt{i}+C*(i-1)$

Fitting arcsin(x) in range (0.000000, 1.000000)
 Pade-Shammas polynomial power is $A+B\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.6738614390179e-13	0	Inf	0
x1	-5.19900829738961e-05	0	-Inf	0
x2	0.00267682144178296	0	Inf	0
x3	0.46019520385434	0	Inf	0
x4	-1.43530993085193	0	-Inf	0
x5	1.57237323986969	0	Inf	0
x6	-0.692425818095225	0	-Inf	0
x7	0.0925447778355341	0	Inf	0
x8	0.935974488819787	0	Inf	0
x9	1.03743874527483	0	Inf	0
x10	-2.16108549216508	0	-Inf	0
x11	2.03289432223419	0	Inf	0
x12	-1.30341804422162	0	-Inf	0
x13	0.576023074291449	0	Inf	0
x14	-0.117828561075799	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 86

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812115	100	0.148796738812115		
Model	14.8796738812115	14	1.06283384865797	Inf	0
Residual	0	86	0		

A = 0.000000, B = 0.100000, C = 0.400000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.541421, 0.973205, 1.400000, 1.823607, 2.244949, 2.664575]

List of powers for D(x): [0.100000, 0.541421, 0.973205, 1.400000, 1.823607, 2.244949, 2.664575]

Fitting arcsin(x) in range (0.000000, 1.000000)

MSS of errors squared = 4.306815e-06

Corrected MSS of errors squared = 7.459622e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

Fitting arcsin(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.34842989448312e-12	0	-Inf	0
x1	-32.486681030906	0	-Inf	0
x2	440.986916480103	0	Inf	0
x3	-2017.68223626807	0	-Inf	0
x4	4388.63699660571	0	Inf	0
x5	-4985.9013334182	0	-Inf	0
x6	2866.59718079558	0	Inf	0
x7	-660.150843149063	0	-Inf	0
x8	467.525307473131	0	Inf	0
x9	-2512.0352390757	0	-Inf	0
x10	3968.97565445065	0	Inf	0
x11	0	0	NaN	NaN
x12	-5700.05837843928	0	-Inf	0
x13	5284.18255797108	0	Inf	0
x14	-1507.58990238953	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812052	100	0.148796738812052		
Model	14.8796738812052	13	1.14459029855425	Inf	0
Residual	0	87	0		

A = 0.182794, B = 0.127247, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.310040, 0.362747, 0.403191, 0.437287, 0.467326, 0.494483, 0.519457]

List of powers for D(x): [0.310040, 0.362747, 0.403191, 0.437287, 0.467326, 0.494483, 0.519457]

Fitting arcsin(x) in range (0.000000, 1.000000)

MSS of errors squared = 6.357235e-06

Corrected MSS of errors squared = 1.101105e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting arcsin(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.81306747358741e-16	0	Inf	0
x1	-1.95224407940193e-07	0	-Inf	0
x2	2.49800407665756e-07	0	Inf	0
x3	-6.44074727363659e-08	0	-Inf	0
x4	9.8314371124856e-09	0	Inf	0
x5	1.0326056626056	0	Inf	0
x6	-0.0341432476201254	0	-Inf	0
x7	0.00169196043763387	0	Inf	0
x8	-0.000166172286825789	0	-Inf	0
x9	1.17968637370506e-05	0	Inf	0

Number of observations: 101, Error degrees of freedom: 91

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	14.8796738812115	100	0.148796738812115		
Model	14.8796738812115	9	1.65329709791239	Inf	0
Residual	0	91	0		

A = 0.000069, B = 0.326167

order Q(x) = 4.000000, order D(x) = 5.000000

List of powers for Q(x): [0.000069, 0.002747, 0.016972, 0.042924]

List of powers for D(x): [0.000069, 0.002747, 0.016972, 0.042924, 0.077922]

Fitting $\arcsin(x)$ in range (0.000000, 1.000000)
 MSS of errors squared = 3.547909e-04
 Corrected MSS of errors squared = 5.017500e-04
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the arc sine function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*i+C*\sqrt{i-1}$	0.124	0.312	0	Inf	-Inf	2.93704600E-06	7, 7
$A+B*\sqrt{i}+C*(i-1)$	0	0.1	0.4	Inf	-Inf	4.30681500E-06	7, 7
$A+B*i-C*(i-1)$	0.162	0.248	0	Inf	-Inf	6.33866400E-06	6, 6
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.182	0.127	0	Inf	-Inf	6.35723500E-06	7, 7
$A+B*\sqrt{i}$	0	0.222		Inf	-Inf	8.04715900E-06	7, 6
$A+B*i+C*(i-1)$	0	0.201	0.14	Inf	-Inf	1.12937800E-05	6, 7
$A+B*i$	0	0.117		Inf	-Inf	2.64991200E-05	6, 7
$A+B/i$	0.015	0.484		Inf	-Inf	5.83681500E-05	7, 5
$A+B*\sqrt{i+C}$	0.071	0.1	1.526	Inf	-Inf	1.34622400E-04	3, 7
$A+B*\log_{10}(i)^4$	0	0.326		Inf	-Inf	3.54790900E-04	4, 5

RESULTS FOR THE ARC TANGENT

Using Power $A+B*i$

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.68269914460934e-13	0	-Inf	0
x1	0.0036629077505218	0	Inf	0
x2	0.972197283088851	0	Inf	0
x3	0.537561656818993	0	Inf	0
x4	-4.71941377000884	0	-Inf	0
x5	7.14991382023113	0	Inf	0
x6	-4.89798883206892	0	-Inf	0
x7	1.69662339015496	0	Inf	0
x8	7.13114580525943	0	Inf	0
x9	-20.0572981937859	0	-Inf	0
x10	28.8296848542177	0	Inf	0
x11	-26.3810101870686	0	-Inf	0
x12	14.3325054064096	0	Inf	0
x13	-3.80047992909191	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862162701	100	0.0542875862162701		
Model	5.42875862162701	13	0.417596817048231	Inf	0
Residual	0	87	0		

$A = 0.445341$, $B = 0.279331$

order $Q(x) = 7.000000$, order $D(x) = 6.000000$

List of powers for $Q(x)$: [0.724672, 1.004004, 1.283335, 1.562667, 1.841998, 2.121330, 2.400661]

List of powers for $D(x)$: [0.724672, 1.004004, 1.283335, 1.562667, 1.841998, 2.121330]

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 1.324055e-11

Corrected MSS of errors squared = 1.872497e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power A+B/i

Fitting $\arctan(x)$ in range (0.000000, 1.000000)
 Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.88852141565674e-14	0	-Inf	0
x1	0.0123876016021156	0	Inf	0
x2	-0.0467335156031249	0	-Inf	0
x3	0.0944996293794637	0	Inf	0
x4	-0.064166406656884	0	-Inf	0
x5	0.0190549027225154	0	Inf	0
x6	-0.0153456510108964	0	-Inf	0
x7	0.919776399402454	0	Inf	0
x8	-13.1230701090025	0	-Inf	0
x9	56.5767445066933	0	Inf	0
x10	-89.6289177342893	0	-Inf	0
x11	46.2516602487727	0	Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862162683	100	0.0542875862162683		
Model	5.42875862162683	11	0.493523511056985	Inf	0
Residual	0	89	0		

A = 0.000000, B = 1.572490

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [1.572490, 0.786245, 0.524163, 0.393122, 0.314498]

List of powers for D(x): [1.572490, 0.786245, 0.524163, 0.393122, 0.314498, 0.262082]

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 1.190903e-09

Corrected MSS of errors squared = 1.684191e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}$

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	4.29865999967428e-13	0	Inf	0
x1	0.00818726963755553	0	Inf	0
x2	-0.0566144530755467	0	-Inf	0
x3	0.127766666249993	0	Inf	0
x4	-0.118426501253949	0	-Inf	0
x5	0.0390897517958973	0	Inf	0
x6	47.8956674632241	0	Inf	0
x7	-278.130313080059	0	-Inf	0
x8	657.96915747878	0	Inf	0
x9	-769.138380443966	0	-Inf	0
x10	443.41287157103	0	Inf	0
x11	-101.00900646923	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862162629	100	0.0542875862162629		
Model	5.42875862162629	11	0.493523511056935	Inf	0
Residual	0	89	0		

A = 0.000000, B = 0.100000

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949]

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 2.734315e-08

Corrected MSS of errors squared = 3.866905e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-5.09051842066691e-14	0	-Inf	0
x1	0.234803892058013	0	Inf	0
x2	-1.4962308782923	0	-Inf	0
x3	3.24491520141193	0	Inf	0
x4	-2.9473323785139	0	-Inf	0
x5	0.964129208829856	0	Inf	0
x6	81.5760825135422	0	Inf	0
x7	-354.364902764779	0	-Inf	0
x8	585.288627225634	0	Inf	0
x9	-428.444811403738	0	-Inf	0
x10	116.944641498878	0	Inf	0

Number of observations: 101, Error degrees of freedom: 90

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862163027	100	0.0542875862163027		
Model	5.42875862163027	10	0.542875862163027	Inf	0
Residual	0	90	0		

A = 0.023622, B = 0.100000, C = 0.379636

order Q(x) = 5.000000, order D(x) = 5.000000

List of powers for Q(x): [0.141080, 0.177882, 0.207459, 0.232897, 0.255562]

List of powers for D(x): [0.141080, 0.177882, 0.207459, 0.232897, 0.255562]

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 4.264933e-08

Corrected MSS of errors squared = 7.387081e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	5.77473358755586e-12	0	Inf	0
x1	-0.00325463602591543	0	-Inf	0
x2	0.181625254562198	0	Inf	0
x3	1.66196730718277	0	Inf	0
x4	-3.32943604674479	0	-Inf	0
x5	2.82281237189652	0	Inf	0
x6	-0.844904227155495	0	-Inf	0
x7	0.155418581937233	0	Inf	0
x8	0.705264835510118	0	Inf	0
x9	1.51189667213782	0	Inf	0
x10	-3.86589584185699	0	-Inf	0
x11	2.88096217799884	0	Inf	0
x12	-1.05248518337807	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862162703	100	0.0542875862162703		
Model	5.42875862162703	12	0.452396551802252	Inf	0
Residual	0	88	0		

A = 0.211802, B = 0.260733, C = 0.084322

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.472536, 0.817591, 1.162646, 1.507702, 1.852757, 2.197813, 2.542868]

List of powers for D(x): [0.472536, 0.817591, 1.162646, 1.507702, 1.852757]

Fitting arctan(x) in range (0.000000, 1.000000)

MSS of errors squared = 2.374420e-10

Corrected MSS of errors squared = 4.112616e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting arctan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.046118100496827	0	Inf	0
x1	0.00140036977491076	0	Inf	0
x2	-0.00749496397954672	0	-Inf	0
x3	0.0188902726142371	0	Inf	0
x4	-0.0405229909465196	0	-Inf	0
x5	-0.0193001824572527	0	-Inf	0
x6	0.000909614455708207	0	Inf	0
x7	-0.0111195612936673	0	-Inf	0
x8	0.0773592386915951	0	Inf	0
x9	-0.257627159690966	0	-Inf	0
x10	0.741685607031552	0	Inf	0
x11	0.480984232868382	0	Inf	0
x12	-0.0312826376665883	0	-Inf	0

Number of observations: 100, Error degrees of freedom: 87

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.23467420800659	99	0.0528754970505716		
Model	5.23467420800659	12	0.436222850667216	Inf	0
Residual	0	87	0		

A = 0.000000, B = 0.195111, C = 0.250000

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.195111, 0.140223, 0.085334, 0.030446, -0.024443, -0.079331]

List of powers for D(x): [0.195111, 0.140223, 0.085334, 0.030446, -0.024443, -0.079331]

Fitting arctan(x) in range (0.000000, 1.000000)

MSS of errors squared = NaN

Corrected MSS of errors squared = NaN

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*sqrt(i-1)

Fitting arctan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	3.7471871707687e-14	0	Inf	0
x1	0.0868340450242604	0	Inf	0
x2	-0.881368913458402	0	-Inf	0
x3	3.77348656198704	0	Inf	0
x4	-8.76512976411322	0	-Inf	0
x5	11.694187533678	0	Inf	0
x6	-8.28546029648666	0	-Inf	0
x7	2.40708900780335	0	Inf	0
x8	20.7632382038392	0	Inf	0
x9	-72.3762002203612	0	-Inf	0
x10	118.366721342762	0	Inf	0
x11	-107.094738052899	0	-Inf	0
x12	51.8886459327458	0	Inf	0
x13	-10.5854037020872	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862162693	100	0.0542875862162693		
Model	5.42875862162693	13	0.417596817048225	Inf	0
Residual	0	87	0		

A = 0.153138, B = 0.118140, C = 0.000000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.271278, 0.389418, 0.507558, 0.625698, 0.743837, 0.861977, 0.980117]

List of powers for D(x): [0.271278, 0.389418, 0.507558, 0.625698, 0.743837, 0.861977]

Fitting arctan(x) in range (0.000000, 1.000000)

MSS of errors squared = 1.643939e-10

Corrected MSS of errors squared = 2.847386e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	4.45420361808355e-12	0	Inf	0
x1	0.307984839789493	0	Inf	0
x2	-4.53607521341613	0	-Inf	0
x3	27.4626167325061	0	Inf	0
x4	-53.0650323036789	0	-Inf	0
x5	40.3265802102785	0	Inf	0
x6	-10.2488487380259	0	-Inf	0
x7	215.057761743011	0	Inf	0
x8	-1785.76430673101	0	-Inf	0
x9	5942.18187949407	0	Inf	0
x10	-10205.0632967501	0	-Inf	0
x11	9606.18784966117	0	Inf	0
x12	-4721.91774297849	0	-Inf	0
x13	950.003078243233	0	Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862166878	100	0.0542875862166878		
Model	5.42875862166878	13	0.417596817051445	Inf	0
Residual	0	87	0		

A = 0.432901, B = 0.400000, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.832901, 0.998586, 1.125721, 1.232901, 1.327328, 1.412697]

List of powers for D(x): [0.832901, 0.998586, 1.125721, 1.232901, 1.327328, 1.412697, 1.491201]

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 1.950609e-11

Corrected MSS of errors squared = 3.378553e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.02603039450755e-13	0	-Inf	0
x1	-0.206121301791514	0	-Inf	0
x2	2.46202611581532	0	Inf	0
x3	-9.44605232462911	0	-Inf	0
x4	16.0536427153765	0	Inf	0
x5	-12.6005747296089	0	-Inf	0
x6	3.73796126186351	0	Inf	0
x7	47.1101551574384	0	Inf	0
x8	-238.667419519637	0	-Inf	0
x9	459.113356830472	0	Inf	0
x10	-351.366555644646	0	-Inf	0
x11	0	0	NaN	NaN
x12	135.333048239902	0	Inf	0
x13	-50.5237077260214	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862162473	100	0.0542875862162473		
Model	5.42875862162473	12	0.452396551802061	Inf	0
Residual	0	88	0		

A = 0.020934, B = 0.175344, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.196277, 0.268907, 0.324638, 0.371621, 0.413014, 0.450436]

List of powers for D(x): [0.196277, 0.268907, 0.324638, 0.371621, 0.413014, 0.450436, 0.484849]

Fitting $\arctan(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 7.460719e-10

Corrected MSS of errors squared = 1.292235e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\log_{10}(i)^4$

Fitting arctan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.62226714089074e-16	0	Inf	0
x1	0	0	NaN	NaN
x2	2.25480607721411e-12	0	Inf	0
x3	-6.05372307898265e-12	0	-Inf	0
x4	6.82084565609973e-12	0	Inf	0
x5	-2.80709757748797e-12	0	-Inf	0
x6	0.999999998650833	0	Inf	0
x7	1.69224495998002e-09	0	Inf	0
x8	-3.94665407882824e-10	0	-Inf	0
x9	5.13143697398584e-11	0	Inf	0

Number of observations: 101, Error degrees of freedom: 92

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	5.42875862162705	100	0.0542875862162705		
Model	5.42875862162705	8	0.678594827703381	Inf	0
Residual	0	92	0		

A = 0.000000, B = 1.774843

order Q(x) = 5.000000, order D(x) = 4.000000

List of powers for Q(x): [0.000000, 0.014575, 0.091976, 0.233195, 0.423637]

List of powers for D(x): [0.000000, 0.014575, 0.091976, 0.233195]

Fitting arctan(x) in range (0.000000, 1.000000)

MSS of errors squared = 2.269805e-05

Corrected MSS of errors squared = 3.209989e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the arc tangent function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*i	0.445	0.279		Inf	-Inf	1.32405500E-11	7, 6
A+B*sqrt(i)+C*(i-1)	0.432	0.4	0	Inf	-Inf	1.95060900E-11	6, 7
A+B*i+C*sqrt(i-1)	0.153	0.118	0	Inf	-Inf	1.64393900E-10	7, 6
A+B*i+C*(i-1)	0.211	0.26	0.084	Inf	-Inf	2.37442000E-10	7, 5
A+B*sqrt(i)+C*sqrt(i-1)	0.02	0.175	0	Inf	-Inf	7.46071900E-10	6, 7
A+B/i	0	1.572		Inf	-Inf	1.19090300E-09	5, 6
A+B*sqrt(i)	0	0.1		Inf	-Inf	2.73431500E-08	5, 6
A+B*sqrt(i+C)	0.023	0.1	0.379	Inf	-Inf	4.26493300E-08	5, 5
A+B*log10(i)^4	0	1.774		Inf	-Inf	2.26980500E-05	5, 4
A+B*i-C*(i-1)	0	0.195	0.25	Inf	-Inf	NaN	6, 6

RESULTS FOR INEVRS E HYPERBOLIC SINE

Using Power A+B* i

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.82862398226334e-12	0	-Inf	0
x1	-3.13536390549341	0	-Inf	0
x2	18.8336524896147	0	Inf	0
x3	-45.3937629411297	0	-Inf	0
x4	54.9017735809945	0	Inf	0
x5	-33.3357117688985	0	-Inf	0
x6	8.13736171793895	0	Inf	0
x7	4.40150149685205	0	Inf	0
x8	-8.37516814620709	0	-Inf	0
x9	9.28168092860719	0	Inf	0
x10	-6.96509212368284	0	-Inf	0
x11	3.64994714861973	0	Inf	0
x12	-1.08273678573532	0	-Inf	0
x13	0.0808483797271489	0	Inf	0

Number of observations: 1001, Error degrees of freedom: 987

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	923.674148234117	1000	0.923674148234117		
Model	923.674148234117	13	71.0518575564706	Inf	0
Residual	0	987	0		

A = 0.000000, B = 0.100000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 2.985649e-08

Corrected MSS of errors squared = 4.222345e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.53778357502033e-13	0	Inf	0
x1	0.144613489559498	0	Inf	0
x2	-3.358469778396	0	-Inf	0
x3	10.2896598196742	0	Inf	0
x4	0	0	NaN	NaN
x5	-15.4766054369795	0	-Inf	0
x6	0	0	NaN	NaN
x7	8.40080794728198	0	Inf	0
x8	0.0259277832529297	0	Inf	0
x9	-2.18213730899047	0	-Inf	0
x10	17.8038689421109	0	Inf	0
x11	-38.7992275285957	0	-Inf	0
x12	24.1515612583752	0	Inf	0

Number of observations: 1001, Error degrees of freedom: 990

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	923.674148234163	1000	0.923674148234163		
Model	923.674148234163	10	92.3674148234163	Inf	0
Residual	0	990	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.100000, 0.050000, 0.033333, 0.025000, 0.020000, 0.016667, 0.014286]

List of powers for D(x): [0.100000, 0.050000, 0.033333, 0.025000, 0.020000]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 6.406461e-07

Corrected MSS of errors squared = 9.060104e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	5.48538068119977e-10	0	Inf	0
x1	-0.000718799208452402	0	-Inf	0
x2	0.00196043823681871	0	Inf	0
x3	-0.00122295275821207	0	-Inf	0
x4	93.0524914494886	0	Inf	0
x5	-666.999878337955	0	-Inf	0
x6	1968.50150976774	0	Inf	0
x7	-3031.38888494404	0	-Inf	0
x8	2578.01984908563	0	Inf	0
x9	-1152.22888015617	0	-Inf	0
x10	212.043771917614	0	Inf	0

Number of observations: 1001, Error degrees of freedom: 990

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	923.674148238661	1000	0.923674148238661		
Model	923.674148238661	10	92.3674148238661	Inf	0

Residual 0 990 0

A = 0.019139, B = 0.100062

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.119201, 0.160648, 0.192452]

List of powers for D(x): [0.119201, 0.160648, 0.192452, 0.219263, 0.242885, 0.264240, 0.283879]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 8.588162e-05

Corrected MSS of errors squared = 1.214549e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.98317484895983e-10	0	Inf	0
x1	34.6429184524044	0	Inf	0
x2	-150.36893967632	0	-Inf	0
x3	182.464282089494	0	Inf	0
x4	0	0	NaN	NaN
x5	-93.1882440961996	0	-Inf	0
x6	0	0	NaN	NaN
x7	26.4566581315686	0	Inf	0
x8	151.304718308149	0	Inf	0
x9	-731.744377748127	0	-Inf	0
x10	1449.94683882091	0	Inf	0
x11	-1452.30657053255	0	-Inf	0
x12	732.576764460442	0	Inf	0
x13	-148.784946288615	0	-Inf	0

Number of observations: 1001, Error degrees of freedom: 989

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	923.674148239864	1000	0.923674148239864		

Model	923.674148239864	11	83.9703771127149	Inf	0
Residual	0	989	0		

A = 0.000000, B = 0.124618, C = 1.150341

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.182740, 0.221187, 0.253877, 0.282813, 0.309051, 0.333230, 0.355770]

List of powers for D(x): [0.182740, 0.221187, 0.253877, 0.282813, 0.309051, 0.333230]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 2.683964e-07

Corrected MSS of errors squared = 4.648762e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*(i-1)

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is A+B*i+C*(i-1)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.67782427860001e-12	0	-Inf	0
x1	6.99507027163459	0	Inf	0
x2	-41.8562688586649	0	-Inf	0
x3	101.164905975435	0	Inf	0
x4	-124.428837810732	0	-Inf	0
x5	79.5615289320447	0	Inf	0
x6	-23.1257847690067	0	-Inf	0
x7	1.68489962354545	0	Inf	0
x8	5.53030500887262	0	Inf	0
x9	-12.7772345026042	0	-Inf	0
x10	15.3765837080249	0	Inf	0
x11	-9.56341182130823	0	-Inf	0
x12	2.43884851235093	0	Inf	0

Number of observations: 1001, Error degrees of freedom: 988

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	923.674148233829	1000	0.923674148233829		

Model	923.674148233829	12	76.9728456861524	Inf	0
Residual	0	988	0		

A = 0.001607, B = 0.100000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.101607, 0.201607, 0.301607, 0.401607, 0.501607, 0.601607, 0.701607]

List of powers for D(x): [0.101607, 0.201607, 0.301607, 0.401607, 0.501607]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 2.005677e-06

Corrected MSS of errors squared = 3.473934e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i-C*(i-1)

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is A+B*i-C*(i-1)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.04142816200501e-10	0	Inf	0
x1	198.96391292703	0	Inf	0
x2	-817.975874877945	0	-Inf	0
x3	1261.16729554166	0	Inf	0
x4	-864.289411997328	0	-Inf	0
x5	222.134569493261	0	Inf	0
x6	549.272534429659	0	Inf	0
x7	-1405.15556476168	0	-Inf	0
x8	0	0	NaN	NaN
x9	3506.13744564056	0	Inf	0
x10	-4602.78854602989	0	-Inf	0
x11	2432.22421788679	0	Inf	0
x12	-478.690644318773	0	-Inf	0

Number of observations: 1001, Error degrees of freedom: 989

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	923.674148242552	1000	0.923674148242552		

Model	923.674148242552	11	83.9703771129593	Inf	0
Residual	0	989	0		

A = 0.000000, B = 0.121424, C = 0.107812

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.121424, 0.135036, 0.148648, 0.162260, 0.175872]

List of powers for D(x): [0.121424, 0.135036, 0.148648, 0.162260, 0.175872, 0.189484, 0.203096]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 2.613156e-07

Corrected MSS of errors squared = 4.526119e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*sqrt(i-1)

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is A+B*i+C*sqrt(i-1)

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.21354904164089e-11	0	-Inf	0
x1	-6.26373485524914	0	-Inf	0
x2	45.4310721264682	0	Inf	0
x3	-119.839322775812	0	-Inf	0
x4	152.920631915371	0	Inf	0
x5	-96.5152115541371	0	-Inf	0
x6	24.2969644458831	0	Inf	0
x7	5.25314166953493	0	Inf	0
x8	-11.4049918894635	0	-Inf	0
x9	13.9435358545079	0	Inf	0
x10	-11.9584056412844	0	-Inf	0
x11	7.50401773279905	0	Inf	0
x12	-2.54429561142129	0	-Inf	0
x13	0.17250690903212	0	Inf	0

Number of observations: 1001, Error degrees of freedom: 987

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	923.674148233379	1000	0.923674148233379		
Model	923.674148233379	13	71.0518575564138	Inf	0
Residual	0	987	0		

A = 0.046272, B = 0.100000, C = 0.020685

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.146272, 0.266958, 0.375526, 0.482100, 0.587643, 0.692526]

List of powers for D(x): [0.146272, 0.266958, 0.375526, 0.482100, 0.587643, 0.692526, 0.796941]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 3.221321e-08

Corrected MSS of errors squared = 5.579491e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	6.7667786882989e-09	0	Inf	0
x1	80.2771794047143	0	Inf	0
x2	-1185.47071671643	0	-Inf	0
x3	5826.97536634251	0	Inf	0
x4	-13539.499660355	0	-Inf	0
x5	16384.6501316253	0	Inf	0
x6	-10018.5185773153	0	-Inf	0
x7	2451.90842166659	0	Inf	0
x8	29.9707843465489	0	Inf	0
x9	-186.771060489881	0	-Inf	0
x10	482.995284465815	0	Inf	0
x11	-567.718716458842	0	-Inf	0
x12	222.219918377583	0	Inf	0
x13	89.3953596694193	0	Inf	0
x14	-69.4570732994412	0	-Inf	0

Number of observations: 1001, Error degrees of freedom: 986

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	923.674148457221	1000	0.923674148457221		
Model	923.674148457221	14	65.9767248898015	Inf	0
Residual	0	986	0		

A = 0.000000, B = 0.400000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.400000, 0.565685, 0.692820, 0.800000, 0.894427, 0.979796, 1.058301]

List of powers for D(x): [0.400000, 0.565685, 0.692820, 0.800000, 0.894427, 0.979796, 1.058301]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 3.263043e-09

Corrected MSS of errors squared = 5.651757e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-9.45731234311857e-11	0	-Inf	0
x1	0.494361877127175	0	Inf	0
x2	-4.87826739746734	0	-Inf	0
x3	12.5233063402343	0	Inf	0
x4	-12.4961892328691	0	-Inf	0
x5	4.35882395973314	0	Inf	0
x6	14.4543199176728	0	Inf	0
x7	-101.272695355672	0	-Inf	0
x8	287.612607592072	0	Inf	0
x9	-418.590767194185	0	-Inf	0
x10	336.143742500617	0	Inf	0
x11	-142.156496590626	0	-Inf	0
x12	24.806979685054	0	Inf	0

Number of observations: 1001, Error degrees of freedom: 988

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	923.674148232775	1000	0.923674148232775		
Model	923.674148232775	12	76.9728456860646	Inf	0
Residual	0	988	0		

A = 0.000000, B = 0.137965, C = 0.040288

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.137965, 0.235399, 0.295937, 0.345710, 0.389074]

List of powers for D(x): [0.137965, 0.235399, 0.295937, 0.345710, 0.389074, 0.428029, 0.463705]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 1.103858e-07

Corrected MSS of errors squared = 1.911938e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting asinh(x) in range (0.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	-4.40706200358708e-14	0	-Inf	0
x2	-9.84190114206013e-14	0	-Inf	0
x3	1.5385728729706e-13	0	Inf	0
x4	1.000000000000628	0	Inf	0
x5	-7.77254918072793e-12	0	-Inf	0
x6	1.70613720680692e-12	0	Inf	0
x7	-2.47172482762075e-13	0	-Inf	0
x8	1.82543065395049e-14	0	Inf	0
x9	-4.90525540281574e-16	0	-Inf	0

Number of observations: 1001, Error degrees of freedom: 992

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
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Total	923.674148234165	1000	0.923674148234165		
Model	923.674148234165	8	115.459268529271	Inf	0
Residual	0	992	0		

A = 0.000000, B = 3.500000

order Q(x) = 3.000000, order D(x) = 6.000000

List of powers for Q(x): [0.000000, 0.028741, 0.181377]

List of powers for D(x): [0.000000, 0.028741, 0.181377, 0.459862, 0.835415, 1.283288]

Fitting asinh(x) in range (0.000000, 100.000000)

MSS of errors squared = 5.220030e-03

Corrected MSS of errors squared = 7.382237e-03

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the inverse hyperbolic sine function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AIC</i> <i>c</i>	<i>MSSE</i>	<i>Q/D</i> <i>Orders</i>
A+B*sqrt(i)+C*(i-1)	0	0.4	0	Inf	-Inf	3.26304300E-09	7, 7
A+B*i	0	0.1		Inf	-Inf	2.98564900E-08	6, 7
A+B*i+C*sqrt(i-1)	0.046	0.1	0.02	Inf	-Inf	3.22132100E-08	6, 7
A+B*sqrt(i)+C*sqrt(i-1)	0	0.137	0.04	Inf	-Inf	1.10385800E-07	5, 7
A+B*i-C*(i-1)	0	0.121	0.107	Inf	-Inf	2.61315600E-07	5, 7
A+B*sqrt(i+C)	0	0.124	1.15	Inf	-Inf	2.68396400E-07	7, 6
A+B/i	0	0.1		Inf	-Inf	6.40646100E-07	7, 5
A+B*i+C*(i-1)	0.001	0.1	0	Inf	-Inf	2.00567700E-06	7, 5
A+B*sqrt(i)	0.019	0.1		Inf	-Inf	8.58816200E-05	3, 7
A+B*log10(i)^4	0	3.5		Inf	-Inf	5.22003000E-03	3, 6

RESULTS FOR INVERSE HYPERBOLIC TANGENT

Using Power $A+B*i$

Fitting $\operatorname{atanh}(x)$ in range (0.000000, 0.999000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.30085982196643e-12	0	-Inf	0
x1	-0.158448303603618	0	-Inf	0
x2	1.65846676105782	0	Inf	0
x3	-7.12988623084183	0	-Inf	0
x4	15.9150254770242	0	Inf	0
x5	-19.0832209340183	0	-Inf	0
x6	11.6002042565199	0	Inf	0
x7	-2.80214037028888	0	-Inf	0
x8	79.6604003704102	0	Inf	0
x9	-324.702376031031	0	-Inf	0
x10	585.896150615975	0	Inf	0
x11	-589.013135755234	0	-Inf	0
x12	344.503436327721	0	Inf	0
x13	-110.456617980199	0	-Inf	0
x14	15.1121422911412	0	Inf	0

Number of observations: 1000, Error degrees of freedom: 985

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	334.064284144156	999	0.334398682826983		
Model	334.064284144156	14	23.8617345817255	Inf	0
Residual	0	985	0		

$A = 0.261003$, $B = 0.102500$

order $Q(x) = 7.000000$, order $D(x) = 7.000000$

List of powers for $Q(x)$: [0.363503, 0.466003, 0.568503, 0.671003, 0.773503, 0.876003, 0.978503]

List of powers for $D(x)$: [0.363503, 0.466003, 0.568503, 0.671003, 0.773503, 0.876003, 0.978503]

Fitting $\operatorname{atanh}(x)$ in range (0.000000, 0.999000)

MSS of errors squared = $1.051883e-04$

Corrected MSS of errors squared = $1.487587e-04$

```

R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power A+B/i

```

Fitting atanh(x) in range (0.000000, 0.999000)
Pade-Shammas polynomial power is A+B/i

```

Linear regression model:

```

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14

```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.3318675031609e-12	0	Inf	0
x1	-0.00195305300367667	0	-Inf	0
x2	0.0547478667275775	0	Inf	0
x3	-0.30925690712084	0	-Inf	0
x4	0.782553737840804	0	Inf	0
x5	-1.0554462906156	0	-Inf	0
x6	0.733510208032959	0	Inf	0
x7	-0.204155189766783	0	-Inf	0
x8	0.000647063893415588	0	Inf	0
x9	-0.16243198315025	0	-Inf	0
x10	6.16207849660325	0	Inf	0
x11	-55.1832074491972	0	-Inf	0
x12	178.259096794142	0	Inf	0
x13	-232.380321157842	0	-Inf	0
x14	104.304138143935	0	Inf	0

Number of observations: 1000, Error degrees of freedom: 985

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	334.064284144336	999	0.334398682827163		
Model	334.064284144336	14	23.8617345817383	Inf	0
Residual	0	985	0		

A = 0.000000, B = 1.837778

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [1.837778, 0.918889, 0.612593, 0.459445, 0.367556, 0.306296, 0.262540]

List of powers for D(x): [1.837778, 0.918889, 0.612593, 0.459445, 0.367556, 0.306296, 0.262540]

Fitting $\tanh(x)$ in range (0.000000, 0.999000)
MSS of errors squared = 1.078602e-04
Corrected MSS of errors squared = 1.525374e-04
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

Using Power $A+B\sqrt{i}$

Fitting $\tanh(x)$ in range (0.000000, 0.999000)
Pade-Shammas polynomial power is $A+B\sqrt{i}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-5.75214955491607e-13	0	-Inf	0
x1	0.00729482009831316	0	Inf	0
x2	-0.0522577148085721	0	-Inf	0
x3	0.121219654230536	0	Inf	0
x4	-0.115028231756669	0	-Inf	0
x5	0.0387714968859214	0	Inf	0
x6	39.7703905804479	0	Inf	0
x7	-183.029317071131	0	-Inf	0
x8	269.383071450245	0	Inf	0
x9	0	0	NaN	NaN
x10	-361.640752700482	0	-Inf	0
x11	329.273975484538	0	Inf	0
x12	-92.7573677503239	0	-Inf	0

Number of observations: 1000, Error degrees of freedom: 988

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	334.064284144416	999	0.334398682827243		
Model	334.064284144416	11	30.3694803767651	Inf	0
Residual	0	988	0		

A = 0.000000, B = 0.100000

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting $\tanh(x)$ in range (0.000000, 0.999000)
MSS of errors squared = 3.549299e-04
Corrected MSS of errors squared = 5.019467e-04
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting $\tanh(x)$ in range (0.000000, 0.999000)
Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.03193563015194e-10	0	Inf	0
x1	-21.5660655122623	0	-Inf	0
x2	197.860664960711	0	Inf	0
x3	-713.133668555098	0	-Inf	0
x4	1313.73061610416	0	Inf	0
x5	-1317.84796223936	0	-Inf	0
x6	687.1329360586	0	Inf	0
x7	-146.176520834498	0	-Inf	0
x8	793.378797003805	0	Inf	0
x9	-3965.77490982119	0	-Inf	0
x10	7327.17710113582	0	Inf	0
x11	-5390.15018012404	0	-Inf	0
x12	0	0	NaN	NaN
x13	1939.33899955769	0	Inf	0
x14	-702.969807747584	0	-Inf	0

Number of observations: 1000, Error degrees of freedom: 986

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	334.064284142942	999	0.334398682825768		
Model	334.064284142942	13	25.6972526263801	Inf	0
Residual	0	986	0		

A = 0.005932, B = 0.100000, C = 2.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.179137, 0.205932, 0.229538, 0.250881, 0.270507, 0.288774, 0.305932]

List of powers for D(x): [0.179137, 0.205932, 0.229538, 0.250881, 0.270507, 0.288774, 0.305932]

Fitting atanh(x) in range (0.000000, 0.999000)

MSS of errors squared = 1.777659e-04

Corrected MSS of errors squared = 3.078996e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting atanh(x) in range (0.000000, 0.999000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.15986018691166e-12	0	-Inf	0
x1	-0.000617436852522798	0	-Inf	0
x2	0.0123573877325102	0	Inf	0
x3	-0.107149888690631	0	-Inf	0
x4	0.523698003956056	0	Inf	0
x5	-1.00548988223505	0	-Inf	0
x6	0.81112097460273	0	Inf	0
x7	-0.233918185916186	0	-Inf	0
x8	2.70005951278169	0	Inf	0
x9	-2.96843077881497	0	-Inf	0
x10	0.961089506332603	0	Inf	0
x11	1.54989962578125	0	Inf	0
x12	-2.08813207803059	0	-Inf	0
x13	1.03621508422275	0	Inf	0
x14	-0.190701110736395	0	-Inf	0

Number of observations: 1000, Error degrees of freedom: 985

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	334.064284144385	999	0.334398682827212		
Model	334.064284144385	14	23.8617345817418	Inf	0
Residual	0	985	0		

```

A = 0.018324, B = 0.128119, C = 0.077411
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.146443, 0.351973, 0.557504, 0.763034, 0.968564, 1.174095,
1.379625]
List of powers for D(x): [0.146443, 0.351973, 0.557504, 0.763034, 0.968564, 1.174095,
1.379625]
Fitting atanh(x) in range (0.000000, 0.999000)
MSS of errors squared = 1.009742e-04
Corrected MSS of errors squared = 1.748925e-04
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i-C*(i-1)$

```

Fitting atanh(x) in range (0.000000, 0.999000)
Pade-Shammas polynomial power is A+B*i-C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.36930150305595e-07	0	Inf	0
x1	5.47781089276638e-08	0	Inf	0
x2	-2.39143412402834e-08	0	-Inf	0
x3	-2.67847214407635e-07	0	-Inf	0
x4	0.00916208643192161	0	Inf	0
x5	-0.0803475588474016	0	-Inf	0
x6	0.884355763923178	0	Inf	0
x7	0.229696198828427	0	Inf	0
x8	-0.0489560248359355	0	-Inf	0
x9	0.00608953451483071	0	Inf	0

Number of observations: 999, Error degrees of freedom: 989

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	333.586619215435	998	0.334255129474384		
Model	333.586619215435	9	37.0651799128261	Inf	0
Residual	0	989	0		

A = 0.000000, B = 0.100000, C = 0.145206

```

order Q(x) = 3.000000, order D(x) = 6.000000
List of powers for Q(x): [0.100000, 0.054794, 0.009587]
List of powers for D(x): [0.100000, 0.054794, 0.009587, -0.035619, -0.080826, -
0.126032]
Fitting atanh(x) in range (0.000000, 0.999000)
MSS of errors squared = NaN
Corrected MSS of errors squared = NaN
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*\sqrt{i-1}$

```

Fitting atanh(x) in range (0.000000, 0.999000)
Pade-Shammas polynomial power is A+B*i+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.42618025198238e-12	0	-Inf	0
x1	-0.000791215035396002	0	-Inf	0
x2	0.0239370651601764	0	Inf	0
x3	-0.212657092348578	0	-Inf	0
x4	0.932298358897832	0	Inf	0
x5	-1.70787673662436	0	-Inf	0
x6	1.35435356039914	0	Inf	0
x7	-0.389262984129981	0	-Inf	0
x8	3.09137521413423	0	Inf	0
x9	-4.86637937308446	0	-Inf	0
x10	3.58138210176815	0	Inf	0
x11	0.512225640681218	0	Inf	0
x12	-2.59315120960918	0	-Inf	0
x13	1.60643590951	0	Inf	0
x14	-0.331888517783217	0	-Inf	0

Number of observations: 1000, Error degrees of freedom: 985

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	334.064284144385	999	0.334398682827212		
Model	334.064284144385	14	23.8617345817418	Inf	0
Residual	0	985	0		

```

A = 0.000000, B = 0.165127, C = 0.062594
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.165127, 0.392848, 0.583902, 0.768923, 0.950822, 1.130725,
1.309211]
List of powers for D(x): [0.165127, 0.392848, 0.583902, 0.768923, 0.950822, 1.130725,
1.309211]
Fitting atanh(x) in range (0.000000, 0.999000)
MSS of errors squared = 1.003655e-04
Corrected MSS of errors squared = 1.738381e-04
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}+C*(i-1)$

```

Fitting atanh(x) in range (0.000000, 0.999000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.48370624139838e-10	0	-Inf	0
x1	6.45047396941165e-05	0	Inf	0
x2	-0.00439467635564524	0	-Inf	0
x3	0.35629682713656	0	Inf	0
x4	-1.29208397831106	0	-Inf	0
x5	1.77352771691334	0	Inf	0
x6	-1.0752109561327	0	-Inf	0
x7	0.241799974219377	0	Inf	0
x8	2.013399585833	0	Inf	0
x9	-1.7567086431427	0	-Inf	0
x10	0.889926478617581	0	Inf	0
x11	-0.000587182922377199	0	-Inf	0
x12	-0.324789617706907	0	-Inf	0
x13	0.253108616033955	0	Inf	0
x14	-0.074349093186511	0	-Inf	0

Number of observations: 1000, Error degrees of freedom: 985

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	334.064284144386	999	0.334398682827214		
Model	334.064284144386	14	23.8617345817419	Inf	0
Residual	0	985	0		

A = 0.053503, B = 0.120172, C = 0.317733

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.173676, 0.541186, 0.897114, 1.247048, 1.593150, 1.936531, 2.277849]

List of powers for D(x): [0.173676, 0.541186, 0.897114, 1.247048, 1.593150, 1.936531, 2.277849]

Fitting atanh(x) in range (0.000000, 0.999000)

MSS of errors squared = 1.099334e-04

Corrected MSS of errors squared = 1.904103e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

Fitting atanh(x) in range (0.000000, 0.999000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.92697010089688e-12	0	-Inf	0
x1	-0.00197496244066215	0	-Inf	0
x2	0.238405487547916	0	Inf	0
x3	-2.26502268126409	0	-Inf	0
x4	7.80649359618891	0	Inf	0
x5	-12.320185978307	0	-Inf	0
x6	9.09693851931381	0	Inf	0
x7	-2.55465383616748	0	-Inf	0
x8	5.53434637938925	0	Inf	0
x9	-57.5555507986433	0	-Inf	0
x10	217.427927895504	0	Inf	0
x11	-391.976206973507	0	-Inf	0
x12	376.241978153025	0	Inf	0
x13	-186.156132676164	0	-Inf	0
x14	37.4836379846232	0	Inf	0

Number of observations: 1000, Error degrees of freedom: 985

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
	-----	---	-----	---	-----
Total	334.06428414434	999	0.334398682827167		
Model	334.06428414434	14	23.8617345817386	Inf	0
Residual	0	985	0		

A = 0.057592, B = 0.100000, C = 0.194943

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.157592, 0.393956, 0.506488, 0.595243, 0.671084, 0.738446, 0.799677]

List of powers for D(x): [0.157592, 0.393956, 0.506488, 0.595243, 0.671084, 0.738446, 0.799677]

Fitting atanh(x) in range (0.000000, 0.999000)

MSS of errors squared = 1.128601e-04

Corrected MSS of errors squared = 1.954795e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting atanh(x) in range (0.000000, 0.999000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
	-----	---	---	-----
(Intercept)	-1.59082886128147e-13	0	-Inf	0
x1	1.29931662335091	0	Inf	0
x2	-1.65292484246198	0	-Inf	0
x3	0.413390338127791	0	Inf	0
x4	-0.0597819990200343	0	-Inf	0
x5	37.8422277926793	0	Inf	0
x6	-41.5101060008679	0	-Inf	0
x7	4.66787816804407	0	Inf	0

Number of observations: 1000, Error degrees of freedom: 992

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
	-----	---	-----	---	-----
Total	334.064284144382	999	0.334398682827209		

Model	334.064284144382	7	47.7234691634831	Inf	0
Residual	0	992	0		

A = 0.009898, B = 0.100000

order Q(x) = 4.000000, order D(x) = 3.000000

List of powers for Q(x): [0.009898, 0.010719, 0.015080, 0.023037]

List of powers for D(x): [0.009898, 0.010719, 0.015080]

Fitting atanh(x) in range (0.000000, 0.999000)

MSS of errors squared = 1.337882e-03

Corrected MSS of errors squared = 1.892052e-03

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the inverse hyperbolic tangent function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AIC</i> <i>c</i>	<i>MSSE</i>	<i>Q/D</i> <i>Orders</i>
$A+B*i+C*\sqrt{i-1}$	0	0.165	0.062	Inf	-Inf	1.00365500E-04	7, 7
$A+B*i+C*(i-1)$	0.018	0.128	0.077	Inf	-Inf	1.00974200E-04	7, 7
$A+B*i$	0.261	0.102		Inf	-Inf	1.05188300E-04	7, 7
$A+B/i$	0	1.837		Inf	-Inf	1.07860200E-04	7, 7
$A+B*\sqrt{i}+C*(i-1)$	0.053	0.12	0.317	Inf	-Inf	1.09933400E-04	7, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.057	0.1	0.194	Inf	-Inf	1.12860100E-04	7, 7
$A+B*\sqrt{i+C}$	0.005	0.1	2	Inf	-Inf	1.77765900E-04	7, 7
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	3.54929900E-04	5, 7
$A+B*\log_{10}(i)^4$	0.009	0.1		Inf	-Inf	1.33788200E-03	4, 3
$A+B*i-C*(i-1)$	0	0.1	0.145	Inf	-Inf	NaN	3, 6

RESULTS FOR THE COSINE INTEGRAL

Using Power $A+B*i$

Fitting $Ci(x)$ in range (0.500000, 20.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-10.8577019026503	0.0612016753022261	-177.40857336066	7.90297216898543e-206
x1	63.4357244937007	0.357555142153841	177.415220800843	7.84956843026664e-206
x2	-158.47178042512	0.894592092266134	-177.144177547655	1.03514366427269e-205
x3	219.45792298996	1.24283474955091	176.578521858405	1.84648231534831e-205
x4	-181.97230502112	1.03565418895738	-175.707593288755	4.51747221879484e-205
x5	90.3564313900483	0.517720414368087	174.527464790691	1.52910864381384e-204
x6	-24.8788235098404	0.143774074323662	-173.041096782397	7.18544088357097e-204
x7	2.93056120725084	0.0171119522247877	171.258145695717	4.67868169216433e-203
x8	4.79886732028317	0.00343954653253112	1395.20348827837	0
x9	-9.60524751916249	0.0142407213778302	-674.49164015777	3.5346529440251e-311
x10	10.2634539299239	0.0235953365261139	434.978069440329	1.58107865262715e-276
x11	-6.17437862666663	0.0195534444175407	-315.76935985395	3.02543023918294e-251
x12	1.98271938786574	0.00810324242733895	244.682225127115	3.91612321661173e-231
x13	-0.265501097378095	0.00134327212672726	-197.652502494011	2.51704974507315e-214

Number of observations: 196, Error degrees of freedom: 182

Root Mean Squared Error: 4.42e-09

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: 1.58e+16, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843003334	195	0.020540556051453		
Model	4.00540843003334	13	0.308108340771795	1.57839114238442e+16	0

Residual 3.5527136788005e-15 182 1.95204048285742e-17

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000]

Fitting Ci(x) in range (0.500000, 20.000000)

MSS of errors squared = 3.225425e-04

Corrected MSS of errors squared = 4.561440e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -7.092657e+03

AIC = -7.122978e+03

AICc = -7.092657e+03

Using Power A+B/i

Fitting Ci(x) in range (0.500000, 20.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.23382784992146e-06	0	Inf	0
x1	2.47031196232775e-07	0	Inf	0
x2	-6.15422578337748e-06	0	-Inf	0
x3	2.05309199985766e-05	0	Inf	0
x4	-1.58575551694842e-05	0	-Inf	0
x5	-0.0132201710694911	0	-Inf	0
x6	2.07409241633583	0	Inf	0
x7	-29.0393393176227	0	-Inf	0
x8	106.058556479983	0	Inf	0
x9	-118.024163302899	0	-Inf	0
x10	0	0	NaN	NaN
x11	39.9440738952802	0	Inf	0

Number of observations: 196, Error degrees of freedom: 185

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843003133	195	0.0205405560514427		
Model	4.00540843003133	10	0.400540843003133	Inf	0
Residual	0	185	0		

```

A = 0.003985, B = 0.100000
order Q(x) = 4.000000, order D(x) = 7.000000
List of powers for Q(x): [0.103985, 0.053985, 0.037318, 0.028985]
List of powers for D(x): [0.103985, 0.053985, 0.037318, 0.028985, 0.023985, 0.020652,
0.018271]
Fitting Ci(x) in range (0.500000, 20.000000)
MSS of errors squared = 1.660945e-02
Corrected MSS of errors squared = 2.348930e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}$

```

Fitting Ci(x) in range (0.500000, 20.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.7103486713701e-06	0	-Inf	0
x1	1.77030402569127e-05	0	Inf	0
x2	-2.63783763917053e-05	0	-Inf	0
x3	1.13790240153827e-05	0	Inf	0
x4	43.8829282335994	0	Inf	0
x5	-244.431393483013	0	-Inf	0
x6	558.939689168165	0	Inf	0
x7	-634.117578984367	0	-Inf	0
x8	355.727020087637	0	Inf	0
x9	-79.000664982219	0	-Inf	0

Number of observations: 196, Error degrees of freedom: 186

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843002808	195	0.020540556051426		
Model	4.00540843002808	9	0.445045381114231	Inf	0
Residual	0	186	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 3.000000, order D(x) = 6.000000

```

List of powers for Q(x): [0.100000, 0.141421, 0.173205]
 List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949]
 Fitting Ci(x) in range (0.500000, 20.000000)
 MSS of errors squared = 2.036249e-02
 Corrected MSS of errors squared = 2.879691e-02
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting Ci(x) in range (0.500000, 20.000000)
 Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.00137927629005035	0	Inf	0
x1	-0.0468383738428659	0	-Inf	0
x2	0.146569252001989	0	Inf	0
x3	0	0	NaN	NaN
x4	-0.575825600078233	0	-Inf	0
x5	0.934546865939261	0	Inf	0
x6	-0.603261837771303	0	-Inf	0
x7	0.143430414992839	0	Inf	0
x8	92.5645373110536	0	Inf	0
x9	-682.83327761725	0	-Inf	0
x10	2082.8040777113	0	Inf	0
x11	-3316.1855656124	0	-Inf	0
x12	2913.41444331197	0	Inf	0
x13	-1343.58404063428	0	-Inf	0
x14	254.819825537146	0	Inf	0

Number of observations: 196, Error degrees of freedom: 182

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843000466	195	0.0205405560513059		
Model	4.00540843000466	13	0.308108340769589	Inf	0
Residual	0	182	0		

A = 0.000000, B = 0.103386, C = 0.125301

```

order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.109672, 0.150721, 0.182772, 0.209986, 0.234058, 0.255875,
0.275972]
List of powers for D(x): [0.109672, 0.150721, 0.182772, 0.209986, 0.234058, 0.255875,
0.275972]
Fitting Ci(x) in range (0.500000, 20.000000)
MSS of errors squared = 7.856315e-02
Corrected MSS of errors squared = 1.360754e-01
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting Ci(x) in range (0.500000, 20.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-13.7973978706768	0	-Inf	0
x1	73.7116064845174	0	Inf	0
x2	-176.970375958908	0	-Inf	0
x3	240.132121081904	0	Inf	0
x4	-195.987953936838	0	-Inf	0
x5	95.9474188004369	0	Inf	0
x6	-26.0658152093312	0	-Inf	0
x7	3.03044323293563	0	Inf	0
x8	4.00717178045524	0	Inf	0
x9	-6.48416094649987	0	-Inf	0
x10	4.84043235292065	0	Inf	0
x11	-0.919061904788914	0	-Inf	0
x12	-0.955107471469705	0	-Inf	0
x13	0.623902991411966	0	Inf	0
x14	-0.11331498759221	0	-Inf	0

Number of observations: 196, Error degrees of freedom: 181

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843002877	195	0.0205405560514296		
Model	4.00540843002877	14	0.286100602144912	Inf	0


```

A = 0.054527, B = 0.104682, C = 0.185142
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.159208, 0.078748, -0.001712]
List of powers for D(x): [0.159208, 0.078748, -0.001712, -0.082173, -0.162633, -
0.243093, -0.323554]
Fitting Ci(x) in range (0.500000, 20.000000)
MSS of errors squared = 6.473171e-02
Corrected MSS of errors squared = 1.121186e-01
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*\sqrt{i-1}$

```

Fitting Ci(x) in range (0.500000, 20.000000)
Pade-Shammas polynomial power is A+B*i+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-14.1504781266895	0	-Inf	0
x1	82.6334177110934	0	Inf	0
x2	-206.327187853089	0	-Inf	0
x3	285.583497718018	0	Inf	0
x4	-236.678890941742	0	-Inf	0
x5	117.457607597174	0	Inf	0
x6	-32.3232929126401	0	-Inf	0
x7	3.80536557983531	0	Inf	0
x8	4.35340733716146	0	Inf	0
x9	-7.43221872602512	0	-Inf	0
x10	5.84236686089748	0	Inf	0
x11	-1.3730542404655	0	-Inf	0
x12	-0.95248560546198	0	-Inf	0
x13	0.692116968856975	0	Inf	0
x14	-0.130247508454478	0	-Inf	0

Number of observations: 196, Error degrees of freedom: 181

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843003184	195	0.0205405560514454		

Model	4.00540843003184	14	0.286100602145132	Inf	0
Residual	0	181	0		

A = 0.000000, B = 0.100279, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100279, 0.200557, 0.300836, 0.401115, 0.501394, 0.601672, 0.701951]

List of powers for D(x): [0.100279, 0.200557, 0.300836, 0.401115, 0.501394, 0.601672, 0.701951]

Fitting Ci(x) in range (0.500000, 20.000000)

MSS of errors squared = 3.532208e-04

Corrected MSS of errors squared = 6.117964e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)+C*(i-1)

Fitting Ci(x) in range (0.500000, 20.000000)

Pade-Shammas polynomial power is A+B*sqrt(i)+C*(i-1)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.09107858528485e-05	0	-Inf	0
x1	0.000287953909014347	0	Inf	0
x2	-0.000825463113678138	0	-Inf	0
x3	0.00082961710330531	0	Inf	0
x4	-0.000281196801712115	0	-Inf	0
x5	223.616091101931	0	Inf	0
x6	-1824.11461606163	0	-Inf	0
x7	5865.99981680031	0	Inf	0
x8	-9660.72866530842	0	-Inf	0
x9	8692.77363432352	0	Inf	0
x10	-4081.68635672251	0	-Inf	0
x11	785.140095867716	0	Inf	0

Number of observations: 196, Error degrees of freedom: 184

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843001664	195	0.0205405560513674		

Model	4.00540843001664	11	0.364128039092422	Inf	0
Residual	0	184	0		

A = 0.048949, B = 0.108820, C = 0.000000

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.157769, 0.202844, 0.237431, 0.266590]

List of powers for D(x): [0.157769, 0.202844, 0.237431, 0.266590, 0.292279, 0.315503, 0.336861]

Fitting Ci(x) in range (0.500000, 20.000000)

MSS of errors squared = 3.648642e-02

Corrected MSS of errors squared = 6.319634e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting Ci(x) in range (0.500000, 20.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.7103486713701e-06	0	-Inf	0
x1	1.77030402569127e-05	0	Inf	0
x2	-2.63783763917053e-05	0	-Inf	0
x3	1.13790240153827e-05	0	Inf	0
x4	43.8829282335994	0	Inf	0
x5	-244.431393483013	0	-Inf	0
x6	558.939689168165	0	Inf	0
x7	-634.117578984367	0	-Inf	0
x8	355.727020087637	0	Inf	0
x9	-79.000664982219	0	-Inf	0

Number of observations: 196, Error degrees of freedom: 186

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843002808	195	0.020540556051426		
Model	4.00540843002808	9	0.445045381114231	Inf	0
Residual	0	186	0		

```

A = 0.000000, B = 0.100000, C = 0.000000
order Q(x) = 3.000000, order D(x) = 6.000000
List of powers for Q(x): [0.100000, 0.141421, 0.173205]
List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949]
Fitting Ci(x) in range (0.500000, 20.000000)
MSS of errors squared = 2.036249e-02
Corrected MSS of errors squared = 3.526887e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\log_{10}(i)^4$

```

Fitting Ci(x) in range (0.500000, 20.000000)
Pade-Shammas polynomial power is  $A+B*\log_{10}(i)^4$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	4.87616632614704e-12	0	Inf	0
x1	0	0	NaN	NaN
x2	-6.15049805619635e-12	0	-Inf	0
x3	1.46979545252498e-12	0	Inf	0
x4	-1.95481969286238e-13	0	-Inf	0
x5	0.999999999999203	0	Inf	0
x6	9.40315595404661e-13	0	Inf	0
x7	-1.42964704389682e-13	0	-Inf	0

Number of observations: 196, Error degrees of freedom: 189

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	4.00540843003173	195	0.0205405560514448		
Model	4.00540843003173	6	0.667568071671955	Inf	0
Residual	0	189	0		

```

A = 0.000000, B = 0.584434
order Q(x) = 4.000000, order D(x) = 3.000000
List of powers for Q(x): [0.000000, 0.004799, 0.030287, 0.076788]
List of powers for D(x): [0.000000, 0.004799, 0.030287]
Fitting Ci(x) in range (0.500000, 20.000000)
MSS of errors squared = 1.032852e-02

```

```
Corrected MSS of errors squared = 1.460674e-02  
R-Squared = 1.00000000  
R-Squared Adjusted = 1.00000000  
Particle swarm AICc = -1.000000e+99  
AIC = -Inf  
AICc = -Inf
```

Comments

The following table shows the summary results for the ten models that fit the cosine integral. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Order s</i>
A+B*i	0	0.1		1.5E16	-7.09E+03	3.22542500E-04	7, 6
A+B*i+C*sqrt(i-1)	0	0.1	0	Inf	-Inf	3.53220800E-04	7, 7
A+B*i+C*(i-1)	0	0.1	0.005	Inf	-Inf	3.57057100E-04	7, 7
A+B*log10(i)^4	0	0.584		Inf	-Inf	1.03285200E-02	4, 3
A+B/i	0.003	0.1		Inf	-Inf	1.66094500E-02	4, 7
A+B*sqrt(i)	0	0.1		Inf	-Inf	2.03624900E-02	3, 6
A+B*sqrt(i)+C*sqrt(i-1)	0	0.1	0	Inf	-Inf	2.03624900E-02	3, 6
A+B*sqrt(i)+C*(i-1)	0.048	0.108	0	Inf	-Inf	3.64864200E-02	4, 7
A+B*i-C*(i-1)	0.054	0.104	0.185	Inf	-Inf	6.47317100E-02	3, 7
A+B*sqrt(i+C)	0	0.103	0.125	Inf	-Inf	7.85631500E-02	7, 7

RESULTS FOR THE COSINE FUNCTION

Using Power A+B* i

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000000006398	0	Inf	0
x1	-2.83764391147102	0	-Inf	0
x2	6.21321791127825	0	Inf	0
x3	-7.60467155968501	0	-Inf	0
x4	5.84734252676208	0	Inf	0
x5	-3.21599309055961	0	-Inf	0
x6	0.886715117833606	0	Inf	0
x7	2.83754959217052	0	Inf	0
x8	-6.21117223686066	0	-Inf	0
x9	7.58544205410103	0	Inf	0
x10	-5.74235593483609	0	-Inf	0
x11	2.48657843151995	0	Inf	0
x12	-0.516864607214192	0	-Inf	0
x13	0.0259980465678655	0	Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125824	500	0.0952531872251649		
Model	47.6265936125824	13	3.6635841240448	Inf	0
Residual	0	487	0		

A = 0.500000, B = 0.293947

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.793947, 1.087894, 1.381841, 1.675789, 1.969736, 2.263683]

List of powers for D(x): [0.793947, 1.087894, 1.381841, 1.675789, 1.969736, 2.263683, 2.557630]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.021816e-11

Corrected MSS of errors squared = 1.445066e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.99999999530184	0	Inf	0
x1	-0.0507538956493821	0	-Inf	0
x2	1.15895533982233	0	Inf	0
x3	11.8358423564934	0	Inf	0
x4	-79.3512137125645	0	-Inf	0
x5	106.525269162186	0	Inf	0
x6	0	0	NaN	NaN
x7	-40.6288894615876	0	-Inf	0
x8	-0.0108001555506238	0	-Inf	0
x9	2.69333374248133	0	Inf	0
x10	-55.8576987720605	0	-Inf	0
x11	290.111985192389	0	Inf	0
x12	-568.777049602484	0	-Inf	0
x13	462.984690952333	0	Inf	0
x14	-131.049898522441	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125819	500	0.0952531872251639		
Model	47.6265936125819	13	3.66358412404476	Inf	0
Residual	0	487	0		

A = 2.000000, B = 1.569551

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [3.569551, 2.784776, 2.523184, 2.392388, 2.313910, 2.261592, 2.224222]

List of powers for D(x): [3.569551, 2.784776, 2.523184, 2.392388, 2.313910, 2.261592, 2.224222]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 2.309491e-12

Corrected MSS of errors squared = 3.266113e-12

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000000012039	0	Inf	0
x1	-1.06857662734986	0	-Inf	0
x2	3.20922821916932	0	Inf	0
x3	-6.48154228209613	0	-Inf	0
x4	6.89986969018347	0	Inf	0
x5	-4.52307712354533	0	-Inf	0
x6	1.42181971489968	0	Inf	0
x7	1.0684853051289	0	Inf	0
x8	-3.20437651731659	0	-Inf	0
x9	5.97768900600931	0	Inf	0
x10	-6.79448434870351	0	-Inf	0
x11	4.20835921283277	0	Inf	0
x12	-1.16682372343738	0	-Inf	0
x13	0.0639927498961305	0	Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125825	500	0.095253187225165		
Model	47.6265936125825	13	3.66358412404481	Inf	0
Residual	0	487	0		

A = 0.404976, B = 0.915218

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [1.320194, 1.699290, 1.990181, 2.235413, 2.451466, 2.646794]

List of powers for D(x): [1.320194, 1.699290, 1.990181, 2.235413, 2.451466, 2.646794, 2.826416]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 2.308064e-12

Corrected MSS of errors squared = 3.264095e-12

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999996254675	0	Inf	0
x1	2.91666183874307	0	Inf	0
x2	-13.0717242476773	0	-Inf	0
x3	17.984803022493	0	Inf	0
x4	-7.72388652629697	0	-Inf	0
x5	-3.16781488309747	0	-Inf	0
x6	2.64595455667386	0	Inf	0
x7	-3.95995499368001	0	-Inf	0
x8	16.0265980001766	0	Inf	0
x9	-26.7411568358824	0	-Inf	0
x10	22.2010350161842	0	Inf	0
x11	-8.97694782504876	0	-Inf	0
x12	1.36956181393598	0	Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125804	500	0.0952531872251608		
Model	47.6265936125804	12	3.96888280104837	Inf	0
Residual	0	488	0		

A = 1.000000, B = 0.751445, C = 0.981473

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [2.057770, 2.297516, 2.499405, 2.677166, 2.837813, 2.985504]

List of powers for D(x): [2.057770, 2.297516, 2.499405, 2.677166, 2.837813, 2.985504]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.450776e-11

Corrected MSS of errors squared = 2.512818e-11

R-Squared = 1.000000000

R-Squared Adjusted = 1.000000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999999535	0	Inf	0
x1	-10.1954796025179	0	-Inf	0
x2	26.7688352505184	0	Inf	0
x3	-34.1292122809161	0	-Inf	0
x4	24.995224558908	0	Inf	0
x5	-10.2959490365398	0	-Inf	0
x6	1.87702532168987	0	Inf	0
x7	10.1962443230062	0	Inf	0
x8	-26.7803439814613	0	-Inf	0
x9	34.2026703428384	0	Inf	0
x10	-25.2500544889046	0	-Inf	0
x11	10.8002581186732	0	Inf	0
x12	-2.40256154349193	0	-Inf	0
x13	0.195948762048893	0	Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125822	500	0.0952531872251643		
Model	47.6265936125822	13	3.66358412404478	Inf	0
Residual	0	487	0		

A = 0.195822, B = 0.172239, C = 0.011933

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.368061, 0.552233, 0.736406, 0.920578, 1.104750, 1.288922]

List of powers for D(x): [0.368061, 0.552233, 0.736406, 0.920578, 1.104750, 1.288922, 1.473094]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 3.999109e-11

Corrected MSS of errors squared = 6.926660e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.99999999998236	0	Inf	0
x1	-16.120310345245	0	-Inf	0
x2	38.6135901777788	0	Inf	0
x3	-36.4368825838819	0	-Inf	0
x4	12.8594226126403	0	Inf	0
x5	3.8558042167698	0	Inf	0
x6	-5.30766163657254	0	-Inf	0
x7	1.57066249631245	0	Inf	0
x8	16.1590681061509	0	Inf	0
x9	-39.0479456535516	0	-Inf	0
x10	38.4713183031115	0	Inf	0
x11	-17.9525104323899	0	-Inf	0
x12	3.30598529376893	0	Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125825	500	0.095253187225165		
Model	47.6265936125825	12	3.96888280104854	Inf	0
Residual	0	488	0		

A = 0.328389, B = 0.141152, C = 0.000000

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.469541, 0.610693, 0.751845, 0.892997, 1.034149, 1.175301, 1.316453]

List of powers for D(x): [0.469541, 0.610693, 0.751845, 0.892997, 1.034149]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.668555e-09

Corrected MSS of errors squared = 2.890022e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999992061	0	Inf	0
x1	-74.0205874614701	0	-Inf	0
x2	297.852403373871	0	Inf	0
x3	-531.499282643359	0	-Inf	0
x4	528.759665604945	0	Inf	0
x5	-305.708915901567	0	-Inf	0
x6	96.5054749106032	0	Inf	0
x7	-12.884318589707	0	-Inf	0
x8	74.0324749800789	0	Inf	0
x9	-297.979079953762	0	-Inf	0
x10	532.063451450809	0	Inf	0
x11	-530.10113941903	0	-Inf	0
x12	307.49756943738	0	Inf	0
x13	-97.7623639008035	0	-Inf	0

x14 13.2408710915133 0 Inf 0

Number of observations: 501, Error degrees of freedom: 486

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125532	500	0.0952531872251064		
Model	47.6265936125532	14	3.4018995437538	Inf	0
Residual	0	486	0		

A = 0.294525, B = 0.111659, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.406184, 0.517842, 0.629501, 0.741160, 0.852819, 0.964477, 1.076136]

List of powers for D(x): [0.406184, 0.517842, 0.629501, 0.741160, 0.852819, 0.964477, 1.076136]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 8.178935e-11

Corrected MSS of errors squared = 1.416633e-10

R-Squared = 1.000000000

R-Squared Adjusted = 1.000000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000000028507	0	Inf	0
x1	-3.00910448115923	0	-Inf	0
x2	1.56610768860418	0	Inf	0
x3	-0.692510753646427	0	-Inf	0
x4	0.0326755151426206	0	Inf	0
x5	1.29072476143831	0	Inf	0
x6	-0.585453557502998	0	-Inf	0
x7	3.00909910633481	0	Inf	0
x8	-1.56564794502628	0	-Inf	0
x9	0.210248260695334	0	Inf	0
x10	-0.0951779016728096	0	-Inf	0

x11	0.341568738517691	0	Inf	0
x12	-0.211928623435554	0	-Inf	0
x13	0.0476501917073728	0	Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125815	500	0.095253187225163		
Model	47.6265936125815	13	3.66358412404473	Inf	0
Residual	0	487	0		

A = 0.550000, B = 0.400000, C = 0.374918

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.950000, 1.490604, 1.992657, 2.474754, 2.944100, 3.404386]

List of powers for D(x): [0.950000, 1.490604, 1.992657, 2.474754, 2.944100, 3.404386, 3.857809]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 7.877539e-11

Corrected MSS of errors squared = 1.364430e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999965469	0	Inf	0
x1	-2.84825537040488	0	-Inf	0
x2	8.63077752837764	0	Inf	0
x3	-11.6385203301734	0	-Inf	0
x4	-1.81689029918978	0	-Inf	0
x5	20.5202829473184	0	Inf	0
x6	-20.6223074874591	0	-Inf	0
x7	6.82855918040392	0	Inf	0
x8	2.84943609348871	0	Inf	0
x9	-8.87452286402508	0	-Inf	0

x10	14.5392811930098	0	Inf	0
x11	-10.3639145176721	0	-Inf	0
x12	2.7504309194102	0	Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125794	500	0.0952531872251587		
Model	47.6265936125794	12	3.96888280104828	Inf	0
Residual	0	488	0		

A = 0.268952, B = 0.100000, C = 0.341449

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.368952, 0.751822, 0.925039, 1.060359, 1.175456, 1.277404, 1.369902]

List of powers for D(x): [0.368952, 0.751822, 0.925039, 1.060359, 1.175456]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.117161e-09

Corrected MSS of errors squared = 1.934979e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting cos(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	6.67087006615928e-09	0	Inf	0
x1	0	0	NaN	NaN
x2	-8.42207134940915e-09	0	-Inf	0
x3	2.03999871751251e-09	0	Inf	0
x4	-2.88008038769335e-10	0	-Inf	0
x5	0.999999993329123	0	Inf	0
x6	8.43055448464781e-09	0	Inf	0
x7	-2.06629067932774e-09	0	-Inf	0
x8	3.30412399304244e-10	0	Inf	0
x9	-4.71088357994288e-11	0	-Inf	0

x10	3.35524767533505e-11	0	Inf	0
x11	-1.1704745699842e-11	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 490

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125817	500	0.0952531872251634		
Model	47.6265936125817	10	4.76265936125817	Inf	0
Residual	0	490	0		

A = 0.000000, B = 1.318864

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.000000, 0.010830, 0.068346, 0.173284]

List of powers for D(x): [0.000000, 0.010830, 0.068346, 0.173284, 0.314800, 0.483566, 0.672711]

Fitting cos(x) in range (0.000000, 1.570796)

MSS of errors squared = 4.323874e-06

Corrected MSS of errors squared = 6.114881e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the cosine function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*sqrt(i)	0.404	0.915		Inf	-Inf	2.30806400E-12	6, 7
A+B/i	2	1.569		Inf	-Inf	2.30949100E-12	7, 7
A+B*i	0.5	0.293		Inf	-Inf	1.02181600E-11	6, 7
A+B*sqrt(i+C)	1	0.751	0.981	Inf	-Inf	1.45077600E-11	6, 6
A+B*i+C*(i-1)	0.195	0.172	0.011	Inf	-Inf	3.99910900E-11	6, 7
A+B*sqrt(i)+C*(i-1)	0.55	0.4	0.374	Inf	-Inf	7.87753900E-11	6, 7
A+B*i+C*sqrt(i-1)	0.294	0.111	0	Inf	-Inf	8.17893500E-11	7, 7
A+B*sqrt(i)+C*sqrt(i-1)	0.268	0.1	0.341	Inf	-Inf	1.11716100E-09	7, 5
A+B*i-C*(i-1)	0.328	0.141	0	Inf	-Inf	1.66855500E-09	7, 5
A+B*log10(i)^4	0	1.318		Inf	-Inf	4.32387400E-06	4, 7

RESULTS FOR THE HYPERBOLIC COSINE FUNCTION

Using Power A+B* i

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000003761644	0	Inf	0
x1	-5.19150343165304	0	-Inf	0
x2	9.42867299469357	0	Inf	0
x3	-9.55413864523507	0	-Inf	0
x4	6.55942077944298	0	Inf	0
x5	-3.38454151697485	0	-Inf	0
x6	1.16675676203538	0	Inf	0
x7	-0.206189159477844	0	-Inf	0
x8	5.19030041213919	0	Inf	0
x9	-9.4068440948231	0	-Inf	0
x10	9.38802661477074	0	Inf	0
x11	-5.85725495371894	0	-Inf	0
x12	2.23047638431803	0	Inf	0
x13	-0.468406730819815	0	-Inf	0
x14	0.0413385842730231	0	Inf	0

Number of observations: 501, Error degrees of freedom: 486

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	167322.933901256	500	334.645867802511		
Model	167322.933901256	14	11951.638135804	Inf	0
Residual	0	486	0		

A = 0.113375, B = 0.383545

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.496920, 0.880465, 1.264010, 1.647556, 2.031101, 2.414646, 2.798191]

List of powers for D(x): [0.496920, 0.880465, 1.264010, 1.647556, 2.031101, 2.414646, 2.798191]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 5.970871e-08

Corrected MSS of errors squared = 8.444087e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.99999999999912	0	Inf	0
x1	-0.0418649513064423	0	-Inf	0
x2	2.64428633441156	0	Inf	0
x3	-20.0853498791941	0	-Inf	0
x4	42.3650030043131	0	Inf	0
x5	-25.8820742499361	0	-Inf	0
x6	0.0169737696546466	0	Inf	0
x7	-0.953513112425266	0	-Inf	0
x8	4.6879424308728	0	Inf	0
x9	0	0	NaN	NaN
x10	-12.4215731823203	0	-Inf	0
x11	0	0	NaN	NaN
x12	9.67016992677498	0	Inf	0

Number of observations: 501, Error degrees of freedom: 490

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	167322.933901257	500	334.645867802513		
Model	167322.933901257	10	16732.2933901257	Inf	0
Residual	0	490	0		

A = 0.000000, B = 0.100000

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.050000, 0.033333, 0.025000, 0.020000]

List of powers for D(x): [0.100000, 0.050000, 0.033333, 0.025000, 0.020000, 0.016667, 0.014286]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 1.844386e-01

Corrected MSS of errors squared = 2.608355e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}$

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999542490982	0	Inf	0
x1	-5.93211653241034	0	-Inf	0
x2	27.2853233054993	0	Inf	0
x3	-52.0259185004678	0	-Inf	0
x4	53.1163480364365	0	Inf	0
x5	-28.8921295213015	0	-Inf	0
x6	6.65749230194046	0	Inf	0
x7	5.99676594836134	0	Inf	0
x8	-26.0514227213365	0	-Inf	0
x9	48.69073561645	0	Inf	0
x10	-46.3031338348608	0	-Inf	0
x11	22.102657353553	0	Inf	0
x12	-4.21909903841467	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	167322.933901201	500	334.645867802403		
Model	167322.933901201	12	13943.5778251001	Inf	0
Residual	0	488	0		

A = 1.122875, B = 0.714893

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [1.837767, 2.133885, 2.361105, 2.552660, 2.721423, 2.873997]

List of powers for D(x): [1.837767, 2.133885, 2.361105, 2.552660, 2.721423, 2.873997]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 5.110217e-09

Corrected MSS of errors squared = 7.226938e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000030148222	0	Inf	0
x1	-9.66609567874522	0	-Inf	0
x2	37.1565322453097	0	Inf	0
x3	-64.3273370123518	0	-Inf	0
x4	64.4072982310385	0	Inf	0
x5	-38.056186550206	0	-Inf	0
x6	11.8047867472123	0	Inf	0
x7	-1.28331241478765	0	-Inf	0
x8	9.56125779469517	0	Inf	0
x9	-35.5765448599645	0	-Inf	0
x10	57.99322838507	0	Inf	0
x11	-49.3028085389875	0	-Inf	0
x12	21.4258730377836	0	Inf	0
x13	-3.77218644360277	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	167322.933901276	500	334.645867802552		
Model	167322.933901276	13	12870.9949154828	Inf	0
Residual	0	487	0		

A = 0.426332, B = 0.849609, C = 0.980844

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [1.622092, 1.893192, 2.121476, 2.322472, 2.504115, 2.671107, 2.826509]

List of powers for D(x): [1.622092, 1.893192, 2.121476, 2.322472, 2.504115, 2.671107]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 3.185721e-09

Corrected MSS of errors squared = 5.517831e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*(i-1)

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is A+B*i+C*(i-1)

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999460146	0	Inf	0
x1	-139.222106533179	0	-Inf	0
x2	492.733897203201	0	Inf	0
x3	-699.967352618705	0	-Inf	0
x4	477.941770067106	0	Inf	0
x5	-132.420257965267	0	-Inf	0
x6	-8.78486346679109	0	-Inf	0
x7	8.75870410366847	0	Inf	0
x8	141.181641122106	0	Inf	0
x9	-508.212068430337	0	-Inf	0
x10	750.543796320775	0	Inf	0
x11	-565.274915996607	0	-Inf	0
x12	216.25541892029	0	Inf	0
x13	-33.5196585051332	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	167322.933901196	500	334.645867802392		
Model	167322.933901196	13	12870.9949154766	Inf	0
Residual	0	487	0		

A = 0.500000, B = 0.100000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.600000, 0.700000, 0.800000, 0.900000, 1.000000, 1.100000, 1.200000]

List of powers for D(x): [0.600000, 0.700000, 0.800000, 0.900000, 1.000000, 1.100000]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 3.449671e-07

Corrected MSS of errors squared = 5.975005e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.000000000000859	0	Inf	0
x1	274.942305444366	0	Inf	0
x2	-1505.35437490459	0	-Inf	0
x3	3436.51418869217	0	Inf	0
x4	-4194.29713842935	0	-Inf	0
x5	2892.4442936555	0	Inf	0
x6	-1070.00566379786	0	-Inf	0
x7	164.756023006919	0	Inf	0
x8	-14.3409489746016	0	-Inf	0
x9	52.5215106542517	0	Inf	0
x10	-65.9581899805477	0	-Inf	0
x11	28.7778656220461	0	Inf	0

Number of observations: 501, Error degrees of freedom: 489

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	167322.933901299	500	334.645867802599		
Model	167322.933901299	11	15211.175809209	Inf	0
Residual		0	489		0

A = 0.092200, B = 0.156876, C = 0.192786

order Q(x) = 7.000000, order D(x) = 4.000000

List of powers for Q(x): [0.249076, 0.213166, 0.177256, 0.141345, 0.105435, 0.069525, 0.033615]

List of powers for D(x): [0.249076, 0.213166, 0.177256, 0.141345]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 2.727809e-04

Corrected MSS of errors squared = 4.724703e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999988582	0	Inf	0
x1	-6.49566073919565	0	-Inf	0
x2	28.3907973853386	0	Inf	0
x3	-56.3308191035069	0	-Inf	0
x4	57.1105137143194	0	Inf	0
x5	-31.5874078915506	0	-Inf	0
x6	8.85772680523187	0	Inf	0
x7	-0.928286222628174	0	-Inf	0
x8	6.49918440816551	0	Inf	0
x9	-28.5171835569756	0	-Inf	0
x10	57.1204006694083	0	Inf	0
x11	-59.2265384818577	0	-Inf	0
x12	34.4152373433091	0	Inf	0
x13	-10.6923053618845	0	-Inf	0
x14	1.39027622646702	0	Inf	0

Number of observations: 501, Error degrees of freedom: 486

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	167322.93390126	500	334.64586780252		
Model	167322.93390126	14	11951.6381358043	Inf	0
Residual	0	486	0		

A = 0.212820, B = 0.100000, C = 0.177532

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.312820, 0.590352, 0.763888, 0.920314, 1.067884, 1.209794, 1.347683]

List of powers for D(x): [0.312820, 0.590352, 0.763888, 0.920314, 1.067884, 1.209794, 1.347683]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 1.863380e-07

Corrected MSS of errors squared = 3.227468e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000006863123	0	Inf	0
x1	-2.93484536692877	0	-Inf	0
x2	5.1384567095687	0	Inf	0
x3	-5.24437874034525	0	-Inf	0
x4	4.2839530267997	0	Inf	0
x5	-2.67414352287349	0	-Inf	0
x6	1.03134225122127	0	Inf	0
x7	-0.175653992698149	0	-Inf	0
x8	2.93406489107087	0	Inf	0
x9	-5.12157196424284	0	-Inf	0
x10	5.11265188683958	0	Inf	0
x11	-3.14359739738701	0	-Inf	0
x12	1.15423028868768	0	Inf	0
x13	-0.230066858269749	0	-Inf	0
x14	0.0190407936371533	0	Inf	0

Number of observations: 501, Error degrees of freedom: 486

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	167322.933901251	500	334.645867802503		
Model	167322.933901251	14	11951.6381358037	Inf	0
Residual	0	486	0		

A = 0.477002, B = 0.336892, C = 0.292704

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.813895, 1.246144, 1.645925, 2.028899, 2.401132, 2.765736, 3.124559]

List of powers for D(x): [0.813895, 1.246144, 1.645925, 2.028899, 2.401132, 2.765736, 3.124559]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 5.461935e-08

Corrected MSS of errors squared = 9.460349e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000021635585	0	Inf	0
x1	-2.6342789856004	0	-Inf	0
x2	9.1338022169512	0	Inf	0
x3	-19.0410566790929	0	-Inf	0
x4	24.9867486203873	0	Inf	0
x5	-20.6103046665438	0	-Inf	0
x6	9.61211923887416	0	Inf	0
x7	-1.91051059594204	0	-Inf	0
x8	2.63304721142352	0	Inf	0
x9	-9.03749967642746	0	-Inf	0
x10	18.2442815621177	0	Inf	0
x11	-20.3861112550331	0	-Inf	0
x12	13.0357211687451	0	Inf	0
x13	-4.48183678463891	0	-Inf	0

```
x14          0.644704140849949      0      Inf      0
```

Number of observations: 501, Error degrees of freedom: 486

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	167322.933901271	500	334.645867802542		
Model	167322.933901271	14	11951.6381358051	Inf	0
Residual	0	486	0		

A = 0.253753, B = 0.700000, C = 0.313988

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.953753, 1.557691, 1.910235, 2.197597, 2.446977, 2.670495, 2.874890]

List of powers for D(x): [0.953753, 1.557691, 1.910235, 2.197597, 2.446977, 2.670495, 2.874890]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 1.013756e-08

Corrected MSS of errors squared = 1.755878e-08

R-Squared = 1.000000000

R-Squared Adjusted = 1.000000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*log10(i)^4

Fitting cosh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is A+B*log10(i)^4

Linear regression model:

```
y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10
```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	-1.36168987508264e-10	0	-Inf	0
x2	1.71179308581942e-10	0	Inf	0
x3	-5.70978413327695e-11	0	-Inf	0
x4	4.31104598292336e-11	0	Inf	0
x5	-3.32259341582777e-11	0	-Inf	0
x6	1.52605910799632e-11	0	Inf	0
x7	-3.37976369583935e-12	0	-Inf	0
x8	1.00000000013624	0	Inf	0
x9	-1.58300162281077e-10	0	-Inf	0
x10	2.22646680239719e-11	0	Inf	0

Number of observations: 501, Error degrees of freedom: 491

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	167322.933901258	500	334.645867802515		
Model	167322.933901258	9	18591.4371001397	Inf	0
Residual	0	491	0		

A = 0.000000, B = 1.412746

order Q(x) = 7.000000, order D(x) = 3.000000

List of powers for Q(x): [0.000000, 0.011601, 0.073211, 0.185619, 0.337208, 0.517989, 0.720597]

List of powers for D(x): [0.000000, 0.011601, 0.073211]

Fitting cosh(x) in range (0.000000, 5.000000)

MSS of errors squared = 4.610947e-02

Corrected MSS of errors squared = 6.520864e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the hyperbolic cosine function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\sqrt{i+C}$	0.426	0.849	0.98	Inf	-Inf	3.18572100E-09	7, 6
$A+B*\sqrt{i}$	1.122	0.714		Inf	-Inf	5.11021700E-09	6, 6
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.253	0.7	0.313	Inf	-Inf	1.01375600E-08	7, 7
$A+B*\sqrt{i}+C*(i-1)$	0.477	0.336	0.292	Inf	-Inf	5.46193500E-08	7, 7
$A+B*i$	0.113	0.383		Inf	-Inf	5.97087100E-08	7, 7
$A+B*i+C*\sqrt{i-1}$	0.212	0.1	0.177	Inf	-Inf	1.86338000E-07	7, 7
$A+B*i+C*(i-1)$	0.5	0.1	0	Inf	-Inf	3.44967100E-07	7, 6
$A+B*i-C*(i-1)$	0.092	0.156	0.192	Inf	-Inf	2.72780900E-04	7, 4
$A+B*\log_{10}(i)^4$	0	1.412		Inf	-Inf	4.61094700E-02	7, 3
$A+B/i$	0	0.1		Inf	-Inf	1.84438600E-01	5, 7

RESULTS FOR THE DIGAMMA FUNCTION

Using Power $A+B*i$

Fitting digamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-5.28623963656016	0	-Inf	0
x1	42.3784834374483	0	Inf	0
x2	0	0	NaN	NaN
x3	-160.063351804008	0	-Inf	0
x4	182.9404806626	0	Inf	0
x5	-60.040896784895	0	-Inf	0
x6	7.00921976256799	0	Inf	0
x7	-11.7999302328169	0	-Inf	0
x8	14.8949347674795	0	Inf	0
x9	-17.0355259638384	0	-Inf	0
x10	9.1671589844638	0	Inf	0
x11	-1.48974684970119	0	-Inf	0
x12	0.129917375485397	0	Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.522284336722	980	0.731145188098696		

Model	716.522284336722	11	65.1383894851565	Inf	0
Residual	0	969	0		

A = 0.148594, B = 0.100000

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.248594, 0.348594, 0.448594, 0.548594, 0.648594]

List of powers for D(x): [0.248594, 0.348594, 0.448594, 0.548594, 0.648594, 0.748594, 0.848594]

Fitting digamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 6.501735e-10

Corrected MSS of errors squared = 9.194841e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting digamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.01693828382219	0	-Inf	0
x1	12.9692331781861	0	Inf	0
x2	-60.2941286298228	0	-Inf	0
x3	49.2727426135126	0	Inf	0
x4	-0.515138948084874	0	-Inf	0
x5	15.9773464098578	0	Inf	0
x6	-402.626668953048	0	-Inf	0
x7	1808.26188159657	0	Inf	0
x8	-2687.77889956352	0	-Inf	0
x9	1267.5616888822	0	Inf	0

Number of observations: 981, Error degrees of freedom: 971

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.522284330928	980	0.731145188092783		
Model	716.522284330928	9	79.6135871478808	Inf	0
Residual	0	971	0		

```

A = 0.179165, B = 0.311893
order Q(x) = 3.000000, order D(x) = 6.000000
List of powers for Q(x): [0.491058, 0.335111, 0.283129]
List of powers for D(x): [0.491058, 0.335111, 0.283129, 0.257138, 0.241544, 0.231147]
Fitting digamma(x) in range (2.000000, 100.000000)
MSS of errors squared = 1.930917e-10
Corrected MSS of errors squared = 2.730729e-10
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}$

```

Fitting digamma(x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i}$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.77142438628825	0	-Inf	0
x1	433.959029899299	0	Inf	0
x2	0	0	NaN	NaN
x3	-889.16640211061	0	-Inf	0
x4	0	0	NaN	NaN
x5	0	0	NaN	NaN
x6	455.838931016782	0	Inf	0
x7	257.328130604393	0	Inf	0
x8	-1528.99434204548	0	-Inf	0
x9	3609.98459386437	0	Inf	0
x10	-4068.46358444006	0	-Inf	0
x11	2212.11307111189	0	Inf	0
x12	-482.944721410203	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 971

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.52228423779	980	0.731145187997745		
Model	716.52228423779	9	79.6135871375322	Inf	0
Residual	0	971	0		

A = 0.595524, B = 0.100000

```

order Q(x) = 6.000000, order D(x) = 6.000000
List of powers for Q(x): [0.695524, 0.736946, 0.768730, 0.795524, 0.819131, 0.840473]
List of powers for D(x): [0.695524, 0.736946, 0.768730, 0.795524, 0.819131, 0.840473]
Fitting digamma(x) in range (2.000000, 100.000000)
MSS of errors squared = 3.340423e-10
Corrected MSS of errors squared = 4.724072e-10
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i+C}$

```

Fitting digamma(x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i+C}$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-17.3767071785233	0	-Inf	0
x1	1587.71534099751	0	Inf	0
x2	-6117.76721785554	0	-Inf	0
x3	7442.2733515424	0	Inf	0
x4	0	0	NaN	NaN
x5	-6772.72993966551	0	-Inf	0
x6	4972.48410306524	0	Inf	0
x7	-1094.60570190915	0	-Inf	0
x8	40.8085376511807	0	Inf	0
x9	-119.934156488532	0	-Inf	0
x10	123.071555660662	0	Inf	0
x11	-42.957674294736	0	-Inf	0

Number of observations: 99, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	77.1013803282087	98	0.786748778859273		
Model	77.1013803282087	10	7.71013803282087	Inf	0
Residual	0	88	0		

A = 0.000000, B = 0.100000, C = 0.825989

order Q(x) = 7.000000, order D(x) = 4.000000

List of powers for Q(x): [0.135129, 0.168107, 0.195601, 0.219681, 0.241371, 0.261266, 0.279750]

List of powers for D(x): [0.135129, 0.168107, 0.195601, 0.219681]

Fitting digamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 3.932838e-11

Corrected MSS of errors squared = 6.811876e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting digamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.27747293243368	0	-Inf	0
x1	33.7230066935861	0	Inf	0
x2	0	0	NaN	NaN
x3	-125.671003462376	0	-Inf	0
x4	131.601521445285	0	Inf	0
x5	-28.2804299693213	0	-Inf	0
x6	-7.67387381684377	0	-Inf	0
x7	0.330674990968134	0	Inf	0
x8	2.96359822807054	0	Inf	0
x9	2.37803181139254	0	Inf	0
x10	-3.14974252246048	0	-Inf	0
x11	-6.32948550686111	0	-Inf	0
x12	4.70843951786422	0	Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.52228433554	980	0.73114518809749		
Model	716.52228433554	11	65.1383894850491	Inf	0
Residual	0	969	0		

A = 0.234885, B = 0.100000, C = 0.044701

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for $Q(x)$: [0.334885, 0.479585, 0.624286, 0.768987, 0.913687, 1.058388, 1.203089]

List of powers for $D(x)$: [0.334885, 0.479585, 0.624286, 0.768987, 0.913687]

Fitting digamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 1.821049e-10

Corrected MSS of errors squared = 3.154150e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting digamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.47656358826854	0	-Inf	0
x1	100.893504251306	0	Inf	0
x2	-187.65768088061	0	-Inf	0
x3	82.332710361909	0	Inf	0
x4	6.56331139679358	0	Inf	0
x5	0.663875786979556	0	Inf	0
x6	-0.0965704081570141	0	-Inf	0
x7	-0.00228111948210221	0	-Inf	0
x8	5.66202878109956	0	Inf	0
x9	1.23303303822742	0	Inf	0
x10	-7.24676224305973	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.522284336425	980	0.731145188098393		
Model	716.522284336425	10	71.6522284336425	Inf	0
Residual	0	970	0		

A = 0.297059, B = 0.184350, C = 0.057748

order $Q(x)$ = 7.000000, order $D(x)$ = 3.000000

List of powers for $Q(x)$: [0.481409, 0.608010, 0.734611, 0.861213, 0.987814, 1.114415, 1.241016]

List of powers for D(x): [0.481409, 0.608010, 0.734611]
 Fitting digamma(x) in range (2.000000, 100.000000)
 MSS of errors squared = 1.555065e-10
 Corrected MSS of errors squared = 2.693451e-10
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting digamma(x) in range (2.000000, 100.000000)
 Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.12849202141196	0	-Inf	0
x1	12.5123132849532	0	Inf	0
x2	-26.9721031111015	0	-Inf	0
x3	0	0	NaN	NaN
x4	24.7489487559323	0	Inf	0
x5	-8.74023728588752	0	-Inf	0
x6	-1.0574338237795	0	-Inf	0
x7	2.77235815355909	0	Inf	0
x8	1.91902783629064	0	Inf	0
x9	-6.81719789552356	0	-Inf	0
x10	0.684347033950835	0	Inf	0
x11	1.28013067971888	0	Inf	0
x12	0.0563301689053562	0	Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.522284335408	980	0.731145188097355		
Model	716.522284335408	11	65.1383894850371	Inf	0
Residual	0	969	0		

A = 0.216284, B = 0.203409, C = 0.201586

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.419694, 0.824688, 1.111597, 1.379078, 1.636502, 1.887499]

List of powers for D(x): [0.419694, 0.824688, 1.111597, 1.379078, 1.636502, 1.887499]

Fitting digamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 1.554940e-09

Corrected MSS of errors squared = 2.693235e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting digamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.56067994977082	0	-Inf	0
x1	75.57553141111177	0	Inf	0
x2	-152.106665928995	0	-Inf	0
x3	0	0	NaN	NaN
x4	129.496717775274	0	Inf	0
x5	-52.7415629901084	0	-Inf	0
x6	7.00605115326526	0	Inf	0
x7	5.3963305605513	0	Inf	0
x8	-18.5908981943842	0	-Inf	0
x9	2.24698352383432	0	Inf	0
x10	2.69811936052171	0	Inf	0
x11	-0.0743906252178552	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.522284336281	980	0.731145188098246		
Model	716.522284336281	10	71.6522284336281	Inf	0
Residual	0	970	0		

A = 0.286541, B = 0.367442, C = 0.042313

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.653983, 0.848496, 1.007596, 1.148365, 1.277419]

List of powers for D(x): [0.653983, 0.848496, 1.007596, 1.148365, 1.277419, 1.398152]

Fitting digamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 2.203713e-10

Corrected MSS of errors squared = 3.816943e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting digamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.9815917484297	0	-Inf	0
x1	30.0097517682567	0	Inf	0
x2	-94.2972686538877	0	-Inf	0
x3	55.9790819528192	0	Inf	0
x4	54.2590114640021	0	Inf	0
x5	-43.2324347223404	0	-Inf	0
x6	3.93686445638749	0	Inf	0
x7	1.2071956647313	0	Inf	0
x8	-11.8702870647582	0	-Inf	0
x9	5.79917380148489	0	Inf	0
x10	1.4702413092081	0	Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.522284336169	980	0.731145188098131		
Model	716.522284336169	10	71.6522284336169	Inf	0
Residual	0	970	0		

A = 0.101148, B = 0.293316, C = 0.179718

order Q(x) = 5.000000, order D(x) = 5.000000

List of powers for Q(x): [0.394464, 0.695678, 0.863346, 0.999061, 1.116459]

List of powers for D(x): [0.394464, 0.695678, 0.863346, 0.999061, 1.116459]

Fitting digamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 1.612427e-09

Corrected MSS of errors squared = 2.792806e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\log_{10}(i)^4$

Fitting digamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	2.78559846608508e-08	0	Inf	0
x2	-3.47579425660411e-08	0	-Inf	0
x3	6.81335947668265e-09	0	Inf	0
x4	1.00000000257542	0	Inf	0
x5	-2.60764230108039e-09	0	-Inf	0
x6	-1.0771725393225e-10	0	-Inf	0
x7	-1.13440794513948e-11	0	-Inf	0
x8	3.16891928135001e-13	0	Inf	0

Number of observations: 981, Error degrees of freedom: 973

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	716.52228433692	980	0.731145188098898		
Model	716.52228433692	7	102.360326333846	Inf	0
Residual	0	973	0		

A = 0.000000, B = 3.021862

order Q(x) = 3.000000, order D(x) = 5.000000

List of powers for Q(x): [0.000000, 0.024815, 0.156599]

List of powers for D(x): [0.000000, 0.024815, 0.156599, 0.397039, 0.721288]

Fitting digamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 2.273145e-07

Corrected MSS of errors squared = 3.214713e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the digamma function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*sqrt(i+C)	0	0.1	0.825	Inf	-Inf	3.93283800E-11	7, 4
A+B*i-C*(i-1)	0.297	0.184	0.057	Inf	-Inf	1.55506500E-10	7, 3
A+B*i+C*(i-1)	0.234	0.1	0.044	Inf	-Inf	1.82104900E-10	7, 5
A+B/i	0.179	0.311		Inf	-Inf	1.93091700E-10	3, 6
A+B*sqrt(i)+C*(i-1)	0.286	0.367	0.042	Inf	-Inf	2.20371300E-10	5, 6
A+B*sqrt(i)	0.595	0.1		Inf	-Inf	3.34042300E-10	6, 6
A+B*i	0.148	0.1		Inf	-Inf	6.50173500E-10	5, 7
A+B*i+C*sqrt(i-1)	0.216	0.203	0.201	Inf	-Inf	1.55494000E-09	6, 6
A+B*sqrt(i)+C*sqrt(i-1)	0.101	0.293	0.179	Inf	-Inf	1.61242700E-09	5, 5
A+B*log10(i)^4	0	3.021		Inf	-Inf	2.27314500E-07	3, 5

RESULTS FOR THE ERROR FUNCTION

Using Power A+B* i

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.87740924935337e-11	0	-Inf	0
x1	1.03530122771787	0	Inf	0
x2	-7.50054706650226	0	-Inf	0
x3	20.1053906183722	0	Inf	0
x4	-22.2403541282385	0	-Inf	0
x5	6.36409483592401	0	Inf	0
x6	4.94384562378568	0	Inf	0
x7	-2.7062385721464	0	-Inf	0
x8	493.040927127705	0	Inf	0
x9	-2423.08000841371	0	-Inf	0
x10	5143.94142792366	0	Inf	0
x11	-5999.68144901832	0	-Inf	0
x12	4035.07405282292	0	Inf	0
x13	-1477.82055575748	0	-Inf	0
x14	229.523834177133	0	Inf	0

Number of observations: 211, Error degrees of freedom: 196

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	18.2280044424114	210	0.08680002115434		
Model	18.2280044424114	14	1.3020003173151	Inf	0
Residual	0	196	0		

A = 0.500000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.600000, 0.700000, 0.800000, 0.900000, 1.000000, 1.100000, 1.200000]

List of powers for D(x): [0.600000, 0.700000, 0.800000, 0.900000, 1.000000, 1.100000, 1.200000]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 1.058465e-06

Corrected MSS of errors squared = 1.496895e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	3.37306817848118e-14	0	Inf	0
x1	0.000307560185404361	0	Inf	0
x2	0.00928065951955667	0	Inf	0
x3	-0.173862393377827	0	-Inf	0
x4	0.837148250941183	0	Inf	0
x5	-1.69645607096286	0	-Inf	0
x6	1.54273919251598	0	Inf	0
x7	-0.519074787378861	0	-Inf	0
x8	-0.0062919644874074	0	-Inf	0
x9	0.879783094772137	0	Inf	0
x10	-12.8308047412923	0	-Inf	0
x11	49.321338376758	0	Inf	0
x12	-57.6103532108872	0	-Inf	0
x13	0	0	NaN	NaN
x14	21.2462306507044	0	Inf	0

Number of observations: 211, Error degrees of freedom: 197

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
	-----	---	-----	---	-----
Total	18.2280044423161	210	0.0868000211538862		
Model	18.2280044423161	13	1.40215418787047	Inf	0
Residual	0	197	0		

A = 0.000000, B = 0.858846

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.858846, 0.429423, 0.286282, 0.214711, 0.171769, 0.143141, 0.122692]

List of powers for D(x): [0.858846, 0.429423, 0.286282, 0.214711, 0.171769, 0.143141, 0.122692]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 1.314502e-08

Corrected MSS of errors squared = 1.858986e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
	-----	---	---	-----
(Intercept)	7.77936447933404e-12	0	Inf	0
x1	0.0826105675116671	0	Inf	0
x2	-2.0035158445681	0	-Inf	0
x3	16.6653736834764	0	Inf	0
x4	-41.9192366422803	0	-Inf	0
x5	44.9633159708813	0	Inf	0
x6	-21.1381977487601	0	-Inf	0
x7	3.37282033619422	0	Inf	0
x8	47.4082938375172	0	Inf	0
x9	-322.398413730969	0	-Inf	0
x10	974.185937230295	0	Inf	0
x11	-1586.71400974492	0	-Inf	0
x12	1449.33935505891	0	Inf	0
x13	-700.811542486742	0	-Inf	0
x14	139.962884521746	0	Inf	0

Number of observations: 211, Error degrees of freedom: 196

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.2280044423284	210	0.0868000211539448		
Model	18.2280044423284	14	1.30200031730917	Inf	0
Residual	0	196	0		

A = 0.106211, B = 0.555514

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.661725, 0.891826, 1.068389, 1.217239, 1.348378, 1.466936, 1.575962]

List of powers for D(x): [0.661725, 0.891826, 1.068389, 1.217239, 1.348378, 1.466936, 1.575962]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 3.007229e-09

Corrected MSS of errors squared = 4.252863e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.52731680046938e-11	0	Inf	0
x1	-3.17271004696044	0	-Inf	0
x2	31.8060746502786	0	Inf	0
x3	-111.249131362506	0	-Inf	0
x4	178.046729578834	0	Inf	0
x5	-133.723181337169	0	-Inf	0
x6	38.2943638192458	0	Inf	0
x7	1819.39356789207	0	Inf	0
x8	-9688.5934306253	0	-Inf	0
x9	18763.3966318403	0	Inf	0
x10	-14322.7732727706	0	-Inf	0
x11	0	0	NaN	NaN
x12	5448.5757477895	0	Inf	0

x13 -2019.0017898714 0 -Inf 0

Number of observations: 211, Error degrees of freedom: 198

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.2280044424008	210	0.0868000211542896		
Model	18.2280044424008	12	1.51900037020007	Inf	0
Residual	0	198	0		

A = 0.168138, B = 0.148292, C = 1.340222

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.394991, 0.439159, 0.477076, 0.510823, 0.541533, 0.569902]

List of powers for D(x): [0.394991, 0.439159, 0.477076, 0.510823, 0.541533, 0.569902, 0.596395]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 1.146963e-08

Corrected MSS of errors squared = 1.986599e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-9.79109539666852e-12	0	-Inf	0
x1	0.00408583982358763	0	Inf	0
x2	-0.0485267735990758	0	-Inf	0
x3	0.216981703515167	0	Inf	0
x4	-0.266793576045706	0	-Inf	0
x5	0.100877689898994	0	Inf	0
x6	4.10777431097042	0	Inf	0
x7	-8.15156677212599	0	-Inf	0
x8	9.77454355644478	0	Inf	0
x9	-7.19688196071673	0	-Inf	0
x10	3.00321572369814	0	Inf	0
x11	-0.544946354909321	0	-Inf	0

Number of observations: 211, Error degrees of freedom: 199

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.2280044423162	210	0.0868000211538868		
Model	18.2280044423162	11	1.65709131293784	Inf	0
Residual	0	199	0		

A = 0.074053, B = 0.123117, C = 0.087317

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.197170, 0.407605, 0.618039, 0.828474, 1.038909]

List of powers for D(x): [0.197170, 0.407605, 0.618039, 0.828474, 1.038909, 1.249343]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 4.595840e-08

Corrected MSS of errors squared = 7.960229e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.66824058136634e-14	0	-Inf	0
x1	-0.0612591531994926	0	-Inf	0
x2	0.514547277057595	0	Inf	0
x3	-1.78007091037132	0	-Inf	0
x4	3.22692703746542	0	Inf	0
x5	-3.21024449165791	0	-Inf	0
x6	1.6576870686594	0	Inf	0
x7	-0.347426656825618	0	-Inf	0
x8	4.35299872131725	0	Inf	0
x9	-7.63449663902528	0	-Inf	0
x10	6.74817612120377	0	Inf	0
x11	-3.0088273851366	0	-Inf	0
x12	0.541959111958549	0	Inf	0

Number of observations: 211, Error degrees of freedom: 198

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
	-----	---	-----	---	-----
Total	18.228004442316	210	0.0868000211538859		
Model	18.228004442316	12	1.519000370193	Inf	0
Residual	0	198	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 4.568448e-07

Corrected MSS of errors squared = 7.912783e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
	-----	---	---	-----
(Intercept)	-2.53538212918212e-14	0	-Inf	0
x1	-0.00157172319904631	0	-Inf	0
x2	0.0166758666191414	0	Inf	0
x3	-0.0683848874294313	0	-Inf	0
x4	0.136183896602023	0	Inf	0
x5	-0.133546164274213	0	-Inf	0
x6	0.0595358364273103	0	Inf	0
x7	-0.00876770861173876	0	-Inf	0
x8	5.52634081197987	0	Inf	0
x9	-13.0217879688067	0	-Inf	0
x10	16.7147153066417	0	Inf	0
x11	-12.2990305513788	0	-Inf	0
x12	4.90711328549641	0	Inf	0
x13	-0.827499354349039	0	-Inf	0

Number of observations: 211, Error degrees of freedom: 197

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.2280044423161	210	0.086800021153886		
Model	18.2280044423161	13	1.40215418787047	Inf	0
Residual	0	197	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 1.346369e-08

Corrected MSS of errors squared = 2.331979e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-5.16678287310899e-13	0	-Inf	0
x1	0.0121654918978234	0	Inf	0
x2	-0.282622837420879	0	-Inf	0
x3	2.10837845785753	0	Inf	0
x4	-4.47440027235066	0	-Inf	0
x5	3.74752079078277	0	Inf	0
x6	-1.10413833078495	0	-Inf	0
x7	-0.00208304747547659	0	-Inf	0
x8	12.1989597934527	0	Inf	0
x9	-51.8815306455194	0	-Inf	0
x10	115.661480507803	0	Inf	0
x11	-151.073990972946	0	-Inf	0
x12	116.533348155784	0	Inf	0

x13	-49.2579010377729	0	-Inf	0
x14	8.81391419382531	0	Inf	0

Number of observations: 211, Error degrees of freedom: 196

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.228004442315	210	0.0868000211538811		
Model	18.228004442315	14	1.30200031730822	Inf	0
Residual	0	196	0		

A = 0.000000, B = 0.400000, C = 0.082636

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.400000, 0.648321, 0.858092, 1.047908, 1.224971, 1.392975, 1.554116]

List of powers for D(x): [0.400000, 0.648321, 0.858092, 1.047908, 1.224971, 1.392975, 1.554116]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 2.245008e-06

Corrected MSS of errors squared = 3.888469e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.34311460818705e-10	0	Inf	0
x1	3.93752111582157	0	Inf	0
x2	-16.8756345715308	0	-Inf	0
x3	-18.868431417492	0	-Inf	0
x4	143.387189927751	0	Inf	0
x5	-182.656409913133	0	-Inf	0
x6	71.1043501512367	0	Inf	0
x7	2092.70667915025	0	Inf	0
x8	-12133.8720398329	0	-Inf	0
x9	20217.5798028106	0	Inf	0

x10	0	0	NaN	NaN
x11	-31409.9773089595	0	-Inf	0
x12	30062.2189470999	0	Inf	0
x13	-8827.69000131443	0	-Inf	0

Number of observations: 211, Error degrees of freedom: 198

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.2280044430375	210	0.0868000211573214		
Model	18.2280044430375	12	1.51900037025313	Inf	0
Residual	0	198	0		

A = 0.500000, B = 0.158971, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.658971, 0.724819, 0.775346, 0.817943, 0.855471, 0.889398]

List of powers for D(x): [0.658971, 0.724819, 0.775346, 0.817943, 0.855471, 0.889398, 0.920598]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 1.246308e-08

Corrected MSS of errors squared = 2.158669e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting erf(x) in range (0.000000, 2.100000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	-7.35839509515437e-16	0	-Inf	0
x2	-1.15692664222352e-14	0	-Inf	0
x3	5.20858098705128e-14	0	Inf	0
x4	-1.20426633153809e-13	0	-Inf	0
x5	8.68050067117744e-14	0	Inf	0
x6	-2.29710988088434e-14	0	-Inf	0
x7	1.000000000000431	0	Inf	0
x8	-5.05101091344547e-12	0	-Inf	0

x9 7.60337787132167e-13 0 Inf 0

Number of observations: 211, Error degrees of freedom: 202

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.2280044423161	210	0.086800021153886		
Model	18.2280044423161	8	2.27850055528951	Inf	0
Residual	0	202	0		

A = 0.000000, B = 3.183182

order Q(x) = 6.000000, order D(x) = 3.000000

List of powers for Q(x): [0.000000, 0.026140, 0.164959, 0.418235, 0.759794, 1.167125]

List of powers for D(x): [0.000000, 0.026140, 0.164959]

Fitting erf(x) in range (0.000000, 2.100000)

MSS of errors squared = 3.419280e-04

Corrected MSS of errors squared = 4.835593e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the error function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*sqrt(i)	0.106	0.555		Inf	-Inf	3.00722900E-09	7, 7
A+B*sqrt(i+C)	0.168	0.148	1.34	Inf	-Inf	1.14696300E-08	6, 7
A+B*sqrt(i)+C*sqrt(i-1)	0.5	0.158	0	Inf	-Inf	1.24630800E-08	6, 7
A+B/i	0	0.858		Inf	-Inf	1.31450200E-08	7, 7
A+B*i+C*sqrt(i-1)	0	0.1	0	Inf	-Inf	1.34636900E-08	7, 6
A+B*i+C*(i-1)	0.074	0.123	0.087	Inf	-Inf	4.59584000E-08	5, 6
A+B*i-C*(i-1)	0	0.1	0	Inf	-Inf	4.56844800E-07	7, 5
A+B*i	0.5	0.1		Inf	-Inf	1.05846500E-06	7, 7
A+B*sqrt(i)+C*(i-1)	0	0.4	0.082	Inf	-Inf	2.24500800E-06	7, 7
A+B*log10(i)^4	0	3.183		Inf	-Inf	3.41928000E-04	6, 3

RESULTS FOR THE EXPONENTIAL FUNCTION

Using Power A+B* i

Fitting exp(x) in range (0.000000, 2.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999993916318	0	Inf	0
x1	-1.20527712687302	0	-Inf	0
x2	2.68922344593565	0	Inf	0
x3	-2.4168028368735	0	-Inf	0
x4	1.80675203027805	0	Inf	0
x5	-0.728090191452736	0	-Inf	0
x6	0.19603144441784	0	Inf	0
x7	1.20861778349971	0	Inf	0
x8	-1.5042461548019	0	-Inf	0
x9	1.18429881199309	0	Inf	0
x10	-0.441935997396052	0	-Inf	0
x11	0.0596313968142522	0	Inf	0

Number of observations: 201, Error degrees of freedom: 189

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122341	200	3.25080452061171		
Model	650.160904122341	11	59.1055367383947	Inf	0

Residual 0 189 0

A = 0.163844, B = 0.431727

order Q(x) = 6.000000, order D(x) = 5.000000

List of powers for Q(x): [0.595571, 1.027299, 1.459026, 1.890753, 2.322481, 2.754208]

List of powers for D(x): [0.595571, 1.027299, 1.459026, 1.890753, 2.322481]

Fitting exp(x) in range (0.000000, 2.000000)

MSS of errors squared = 3.295222e-09

Corrected MSS of errors squared = 4.660148e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting exp(x) in range (0.000000, 2.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999946151	0	Inf	0
x1	-0.0238465805613957	0	-Inf	0
x2	1.37103804964088	0	Inf	0
x3	-11.6085810354062	0	-Inf	0
x4	26.7890616775246	0	Inf	0
x5	-17.4949912279066	0	-Inf	0
x6	0.0199083483920018	0	Inf	0
x7	-1.4209143706928	0	-Inf	0
x8	11.8211714156381	0	Inf	0
x9	-27.0090807828413	0	-Inf	0
x10	17.5768927697567	0	Inf	0

Number of observations: 201, Error degrees of freedom: 190

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122351	200	3.25080452061175		
Model	650.160904122351	10	65.0160904122351	Inf	0
Residual	0	190	0		

A = 0.000000, B = 1.275872

```

order Q(x) = 5.000000, order D(x) = 5.000000
List of powers for Q(x): [1.275872, 0.637936, 0.425291, 0.318968, 0.255174]
List of powers for D(x): [1.275872, 0.637936, 0.425291, 0.318968, 0.255174]
Fitting exp(x) in range (0.000000, 2.000000)
MSS of errors squared = 2.871380e-07
Corrected MSS of errors squared = 4.060745e-07
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}$

```

Fitting exp(x) in range (0.000000, 2.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i}$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999980808	0	Inf	0
x1	-46.962527983001	0	-Inf	0
x2	129.874740403595	0	Inf	0
x3	-95.433575178441	0	-Inf	0
x4	-15.4994012623929	0	-Inf	0
x5	27.0283267517359	0	Inf	0
x6	52.4136067516097	0	Inf	0
x7	-167.606701623701	0	-Inf	0
x8	181.296183408899	0	Inf	0
x9	-65.1058706781908	0	-Inf	0

Number of observations: 201, Error degrees of freedom: 191

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122377	200	3.25080452061189		
Model	650.160904122377	9	72.2401004580419	Inf	0
Residual	0	191	0		

A = 0.083921, B = 0.100000

```

order Q(x) = 5.000000, order D(x) = 4.000000

```

```

List of powers for Q(x): [0.183921, 0.225343, 0.257127, 0.283921, 0.307528]

```

```

List of powers for D(x): [0.183921, 0.225343, 0.257127, 0.283921]

```

```

Fitting exp(x) in range (0.000000, 2.000000)

```

MSS of errors squared = 2.745695e-06
 Corrected MSS of errors squared = 3.883000e-06
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting $\exp(x)$ in range (0.000000, 2.000000)
 Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000000002397	0	Inf	0
x1	-169.704225217909	0	-Inf	0
x2	754.903060466914	0	Inf	0
x3	-1365.28325965497	0	-Inf	0
x4	1247.29907261989	0	Inf	0
x5	-576.101386228965	0	-Inf	0
x6	108.080989606663	0	Inf	0
x7	159.055990045051	0	Inf	0
x8	-654.922265792614	0	-Inf	0
x9	1028.72624493289	0	Inf	0
x10	-722.729551809769	0	-Inf	0
x11	190.798121460866	0	Inf	0

Number of observations: 201, Error degrees of freedom: 189

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122465	200	3.25080452061233		
Model	650.160904122465	11	59.1055367384059	Inf	0
Residual	0	189	0		

A = 0.215295, B = 0.234212, C = 1.103622

order Q(x) = 6.000000, order D(x) = 5.000000

List of powers for Q(x): [0.554993, 0.627909, 0.689748, 0.744409, 0.793928, 0.839532]

List of powers for D(x): [0.554993, 0.627909, 0.689748, 0.744409, 0.793928]

Fitting $\exp(x)$ in range (0.000000, 2.000000)

MSS of errors squared = 1.195179e-08

Corrected MSS of errors squared = 2.070110e-08

```

R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting exp(x) in range (0.000000, 2.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.9999999999999946	0	Inf	0
x1	-5.26459886324561	0	-Inf	0
x2	6.62351933883474	0	Inf	0
x3	-4.39500588682177	0	-Inf	0
x4	0.805712699812568	0	Inf	0
x5	0.622091253953599	0	Inf	0
x6	-0.482691904820773	0	-Inf	0
x7	0.0960599415091427	0	Inf	0
x8	5.26495827249125	0	Inf	0
x9	-6.6363462619358	0	-Inf	0
x10	5.20763545559433	0	Inf	0
x11	-5.27417782774082	0	-Inf	0
x12	4.25591513699347	0	Inf	0
x13	-1.71114693391594	0	-Inf	0
x14	0.259170351727046	0	Inf	0

Number of observations: 201, Error degrees of freedom: 186

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122344	200	3.25080452061172		
Model	650.160904122344	14	46.4400645801674	Inf	0
Residual	0	186	0		

A = 0.123382, B = 0.176224, C = 0.152633

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.299606, 0.628462, 0.957319, 1.286176, 1.615033, 1.943889, 2.272746]

List of powers for D(x): [0.299606, 0.628462, 0.957319, 1.286176, 1.615033, 1.943889, 2.272746]

Fitting $\exp(x)$ in range (0.000000, 2.000000)
 MSS of errors squared = 8.063728e-11
 Corrected MSS of errors squared = 1.396679e-10
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting $\exp(x)$ in range (0.000000, 2.000000)
 Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999946293	0	Inf	0
x1	-2.48948739113973	0	-Inf	0
x2	2.98939601886377	0	Inf	0
x3	-1.52805748419834	0	-Inf	0
x4	0.329459751458094	0	Inf	0
x5	2.49281565283461	0	Inf	0
x6	-3.04382413747108	0	-Inf	0
x7	2.07486675231353	0	Inf	0
x8	0.0574532437423314	0	Inf	0
x9	-1.50723523291177	0	-Inf	0
x10	1.03264217794966	0	Inf	0
x11	-0.217564542021843	0	-Inf	0

Number of observations: 201, Error degrees of freedom: 189

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122341	200	3.2508045206117		
Model	650.160904122341	11	59.1055367383946	Inf	0
Residual	0	189	0		

A = 0.139485, B = 0.250000, C = 0.005583

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.389485, 0.633903, 0.878320, 1.122737]

List of powers for D(x): [0.389485, 0.633903, 0.878320, 1.122737, 1.367155, 1.611572, 1.855989]

Fitting $\exp(x)$ in range (0.000000, 2.000000)

MSS of errors squared = 4.573696e-09
 Corrected MSS of errors squared = 7.921874e-09
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting $\exp(x)$ in range (0.000000, 2.000000)
 Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000000551173	0	Inf	0
x1	1.65239522610172	0	Inf	0
x2	-1.53735764283304	0	-Inf	0
x3	2.58950753244784	0	Inf	0
x4	-1.37714138514757	0	-Inf	0
x5	0.548827656799103	0	Inf	0
x6	-1.6540339516731	0	-Inf	0
x7	2.19552220912594	0	Inf	0
x8	-0.591867350585311	0	-Inf	0
x9	0.180349098675421	0	Inf	0
x10	-0.274367883660966	0	-Inf	0
x11	0.0862914854339068	0	Inf	0

Number of observations: 201, Error degrees of freedom: 189

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.16090412234	200	3.2508045206117		
Model	650.16090412234	11	59.1055367383945	Inf	0
Residual	0	189	0		

A = 0.109739, B = 0.332559, C = 0.169398

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.442298, 0.944255, 1.346981, 1.733380, 2.111329]

List of powers for D(x): [0.442298, 0.944255, 1.346981, 1.733380, 2.111329, 2.483877]

Fitting $\exp(x)$ in range (0.000000, 2.000000)

MSS of errors squared = 3.560869e-09

Corrected MSS of errors squared = 6.167605e-09

```
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Using Power $A+B\sqrt{i}+C*(i-1)$

```
Fitting exp(x) in range (0.000000, 2.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i}+C*(i-1)$ 
```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.99999999999655	0	Inf	0
x1	-23.0479206528905	0	-Inf	0
x2	92.9861251967369	0	Inf	0
x3	-120.758654103576	0	-Inf	0
x4	0	0	NaN	NaN
x5	134.458383536135	0	Inf	0
x6	-114.701975649472	0	-Inf	0
x7	30.8101768805748	0	Inf	0
x8	23.1564467106774	0	Inf	0
x9	-85.3598908809494	0	-Inf	0
x10	79.8632460474089	0	Inf	0
x11	112.555298377728	0	Inf	0
x12	-290.430504050506	0	-Inf	0
x13	217.229951350559	0	Inf	0
x14	-56.289035358098	0	-Inf	0

Number of observations: 201, Error degrees of freedom: 187

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122271	200	3.25080452061136		
Model	650.160904122271	13	50.0123772401747	Inf	0
Residual	0	187	0		

A = 0.550000, B = 0.400000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.950000, 1.115685, 1.242820, 1.350000, 1.444427, 1.529796, 1.608301]

List of powers for D(x): [0.950000, 1.115685, 1.242820, 1.350000, 1.444427, 1.529796, 1.608301]

Fitting $\exp(x)$ in range (0.000000, 2.000000)
MSS of errors squared = 9.260218e-12
Corrected MSS of errors squared = 1.603917e-11
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting $\exp(x)$ in range (0.000000, 2.000000)
Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999992882471	0	Inf	0
x1	-0.296839509428743	0	-Inf	0
x2	-0.453694731023085	0	-Inf	0
x3	1.83608419797343	0	Inf	0
x4	-1.8925313555347	0	-Inf	0
x5	0.75485269511939	0	Inf	0
x6	0.296635778529638	0	Inf	0
x7	1.1042519666183	0	Inf	0
x8	-1.06743541785815	0	-Inf	0
x9	0.317845291308441	0	Inf	0

Number of observations: 201, Error degrees of freedom: 191

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122341	200	3.25080452061171		
Model	650.160904122341	9	72.2401004580379	Inf	0
Residual	0	191	0		

A = 0.046439, B = 0.210203, C = 0.609702

order Q(x) = 5.000000, order D(x) = 4.000000

List of powers for Q(x): [0.256642, 0.953412, 1.272769, 1.522879, 1.735870]

List of powers for D(x): [0.256642, 0.953412, 1.272769, 1.522879]

Fitting $\exp(x)$ in range (0.000000, 2.000000)

MSS of errors squared = 1.955620e-08

Corrected MSS of errors squared = 3.387233e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting $\exp(x)$ in range (0.000000, 2.000000)
 Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	-1.0077223533868e-10	0	-Inf	0
x2	1.24883667733323e-10	0	Inf	0
x3	-2.73282489415117e-11	0	-Inf	0
x4	3.14633172202497e-12	0	Inf	0
x5	1.00000000010077	0	Inf	0
x6	-1.25080069259733e-10	0	-Inf	0
x7	2.78811182790211e-11	0	Inf	0
x8	-3.76781399054828e-12	0	-Inf	0
x9	2.19763564037908e-13	0	Inf	0

Number of observations: 201, Error degrees of freedom: 192

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	650.160904122341	200	3.25080452061171		
Model	650.160904122341	8	81.2701130152927	Inf	0
Residual	0	192	0		

A = 0.000000, B = 1.595510

order Q(x) = 4.000000, order D(x) = 5.000000

List of powers for Q(x): [0.000000, 0.013102, 0.082683, 0.209633]

List of powers for D(x): [0.000000, 0.013102, 0.082683, 0.209633, 0.380832]

Fitting $\exp(x)$ in range (0.000000, 2.000000)

MSS of errors squared = 3.020454e-03

Corrected MSS of errors squared = 4.271567e-03

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the exponential function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\sqrt{i}+C*(i-1)$	0.55	0.4	0	Inf	-Inf	9.26021800E-12	7, 7
$A+B*i+C*(i-1)$	0.123	0.176	0.152	Inf	-Inf	8.06372800E-11	7, 7
$A+B*i$	0.163	0.431		Inf	-Inf	3.29522200E-09	6, 5
$A+B*i+C*\sqrt{i-1}$	0.109	0.332	0.169	Inf	-Inf	3.56086900E-09	5, 6
$A+B*i-C*(i-1)$	0.139	0.25	0.005	Inf	-Inf	4.57369600E-09	4, 7
$A+B*\sqrt{i+C}$	0.215	0.234	1.103	Inf	-Inf	1.19517900E-08	6, 5
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.046	0.21	0.609	Inf	-Inf	1.95562000E-08	5, 4
$A+B/i$	0	1.275		Inf	-Inf	2.87138000E-07	5, 5
$A+B*\sqrt{i}$	0.083	0.1		Inf	-Inf	2.74569500E-06	5, 4
$A+B*\log_{10}(i)^4$	0	1.595		Inf	-Inf	3.02045400E-03	4, 5

RESULTS FOR THE FRESNEL COSINE

Using Power $A+B*i$

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-7.82498321437591e-05	0	-Inf	0
x1	0.000210470338019939	0	Inf	0
x2	-0.000188518805767774	0	-Inf	0
x3	5.62308980613566e-05	0	Inf	0
x4	6.69092536053876	0	Inf	0
x5	-19.156653474278	0	-Inf	0
x6	30.4249166050997	0	Inf	0
x7	-28.9507976648984	0	-Inf	0
x8	16.5056104302928	0	Inf	0
x9	-5.2207759269688	0	-Inf	0
x10	0.706774743466249	0	Inf	0

Number of observations: 451, Error degrees of freedom: 440

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.8248115391142	450	0.0240551367535871		
Model	10.8248115391142	10	1.08248115391142	Inf	0
Residual	0	440	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.200000, 0.300000]
List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
Fitting FresnelCosine(x) in range (0.500000, 5.000000)
MSS of errors squared = 4.707377e-02
Corrected MSS of errors squared = 6.657236e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power A+B/i

```

Fitting FresnelCosine(x) in range (0.500000, 5.000000)
Pade-Shammas polynomial power is A+B/i

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.000567568402155187	0	Inf	0
x1	-0.000255294530847131	0	-Inf	0
x2	0.00231376701209572	0	Inf	0
x3	-0.00262530433083697	0	-Inf	0
x4	0.0650205904377489	0	Inf	0
x5	-4.51340459877061	0	-Inf	0
x6	34.0905639528398	0	Inf	0
x7	-70.0245289900059	0	-Inf	0
x8	41.3823482279123	0	Inf	0

Number of observations: 451, Error degrees of freedom: 442

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.8248115391142	450	0.0240551367535871		
Model	10.8248115391142	8	1.35310144238928	Inf	0
Residual	0	442	0		

```

A = 0.019705, B = 0.340734
order Q(x) = 3.000000, order D(x) = 5.000000
List of powers for Q(x): [0.360440, 0.190072, 0.133283]

```

List of powers for D(x): [0.360440, 0.190072, 0.133283, 0.104889, 0.087852]
 Fitting FresnelCosine(x) in range (0.500000, 5.000000)
 MSS of errors squared = 2.638361e-02
 Corrected MSS of errors squared = 3.731207e-02
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*\sqrt{i}$

Fitting FresnelCosine(x) in range (0.500000, 5.000000)
 Pade-Shammas polynomial power is $A+B*\sqrt{i}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.208085343982628	0	-Inf	0
x1	7.87274749481302	0	Inf	0
x2	-35.2987190836814	0	-Inf	0
x3	50.8218773830584	0	Inf	0
x4	0	0	NaN	NaN
x5	-65.6661772163842	0	-Inf	0
x6	58.7584687926542	0	Inf	0
x7	-16.2801119492744	0	-Inf	0
x8	35.9371932617265	0	Inf	0
x9	-158.062175730233	0	-Inf	0
x10	224.412416227318	0	Inf	0
x11	0	0	NaN	NaN
x12	-283.969911851004	0	-Inf	0
x13	251.978027240745	0	Inf	0
x14	-69.2955492339086	0	-Inf	0

Number of observations: 451, Error degrees of freedom: 438

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.8248115391145	450	0.0240551367535877		
Model	10.8248115391145	12	0.902067628259538	Inf	0
Residual	0	438	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

MSS of errors squared = 2.803037e-02

Corrected MSS of errors squared = 3.964093e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.204075294861438	0	-Inf	0
x1	7.65174744459708	0	Inf	0
x2	-34.1426071440201	0	-Inf	0
x3	48.9569948793088	0	Inf	0
x4	0	0	NaN	NaN
x5	-62.8093465102123	0	-Inf	0
x6	56.0216102614902	0	Inf	0
x7	-15.4743233746778	0	-Inf	0
x8	35.1061963800855	0	Inf	0
x9	-152.928574004723	0	-Inf	0
x10	215.53223106784	0	Inf	0
x11	0	0	NaN	NaN
x12	-269.581460534899	0	-Inf	0
x13	238.040014175124	0	Inf	0
x14	-65.168407372681	0	-Inf	0

Number of observations: 451, Error degrees of freedom: 438

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.8248115391159	450	0.0240551367535909		
Model	10.8248115391159	12	0.90206762825966	Inf	0
Residual	0	438	0		

```

A = 0.000000, B = 0.123303, C = 0.000000
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.123303, 0.174377, 0.213568, 0.246607, 0.275715, 0.302030,
0.326230]
List of powers for D(x): [0.123303, 0.174377, 0.213568, 0.246607, 0.275715, 0.302030,
0.326230]
Fitting FresnelCosine(x) in range (0.500000, 5.000000)
MSS of errors squared = 2.317256e-02
Corrected MSS of errors squared = 4.013605e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting FresnelCosine(x) in range (0.500000, 5.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.213665115150878	0	-Inf	0
x1	8.66921517463282	0	Inf	0
x2	-34.6566884089561	0	-Inf	0
x3	62.8068578845535	0	Inf	0
x4	-63.7019190330805	0	-Inf	0
x5	37.478993633528	0	Inf	0
x6	-12.0112756149984	0	-Inf	0
x7	1.62848455922045	0	Inf	0
x8	35.0237580685407	0	Inf	0
x9	-133.150926758473	0	-Inf	0
x10	230.624585565832	0	Inf	0
x11	-224.552370768981	0	-Inf	0
x12	127.333729284775	0	Inf	0
x13	-39.4720223168711	0	-Inf	0
x14	5.19324352039676	0	Inf	0

Number of observations: 451, Error degrees of freedom: 436

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	10.8248115391158	450	0.0240551367535906		
Model	10.8248115391158	14	0.773200824222555	Inf	0
Residual	0	436	0		

A = 0.146369, B = 0.102732, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.249101, 0.351832, 0.454564, 0.557295, 0.660027, 0.762758, 0.865490]

List of powers for D(x): [0.249101, 0.351832, 0.454564, 0.557295, 0.660027, 0.762758, 0.865490]

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

MSS of errors squared = 9.604513e-01

Corrected MSS of errors squared = 1.663551e+00

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.000658422130012715	0	Inf	0
x1	-1.05986786243563e-05	0	-Inf	0
x2	7.94386712018448e-05	0	Inf	0
x3	-0.000593787388773735	0	-Inf	0
x4	-0.000150101289399633	0	-Inf	0
x5	1.66265563863331e-05	0	Inf	0
x6	0.00557359681233615	0	Inf	0
x7	-0.0621809544584865	0	-Inf	0
x8	0.768789388761456	0	Inf	0
x9	0.385785700424757	0	Inf	0
x10	-0.127260825380808	0	-Inf	0
x11	0.0335471864631476	0	Inf	0
x12	-0.00425409262329064	0	-Inf	0

Number of observations: 451, Error degrees of freedom: 438

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	10.8248115391137	450	0.024055136753586		
Model	10.8248115391137	12	0.902067628259474	Inf	0
Residual	0	438	0		

A = 0.000000, B = 0.104357, C = 0.150412
 order Q(x) = 5.000000, order D(x) = 7.000000
 List of powers for Q(x): [0.104357, 0.058302, 0.012247, -0.033808, -0.079863]
 List of powers for D(x): [0.104357, 0.058302, 0.012247, -0.033808, -0.079863, -0.125918, -0.171973]
 Fitting FresnelCosine(x) in range (0.500000, 5.000000)
 MSS of errors squared = 1.077030e-01
 Corrected MSS of errors squared = 1.865471e-01
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting FresnelCosine(x) in range (0.500000, 5.000000)
 Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.179433014195043	0	-Inf	0
x1	1.74574141068874	0	Inf	0
x2	-12.5017461171825	0	-Inf	0
x3	32.9048544817159	0	Inf	0
x4	-42.4369660227377	0	-Inf	0
x5	29.6982584381246	0	Inf	0
x6	-10.866703526622	0	-Inf	0
x7	1.6360968094549	0	Inf	0
x8	7.46278160545254	0	Inf	0
x9	-44.1658844875562	0	-Inf	0
x10	106.64659659331	0	Inf	0
x11	-128.997994872584	0	-Inf	0
x12	85.7667361459313	0	Inf	0
x13	-30.0842194153693	0	-Inf	0
x14	4.37187116004808	0	Inf	0

Number of observations: 451, Error degrees of freedom: 436

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.8248115391183	450	0.0240551367535961		
Model	10.8248115391183	14	0.773200824222733	Inf	0
Residual	0	436	0		

A = 0.226051, B = 0.100000, C = 0.236449

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.326051, 0.662500, 0.860441, 1.035593, 1.198949, 1.354767, 1.505231]

List of powers for D(x): [0.326051, 0.662500, 0.860441, 1.035593, 1.198949, 1.354767, 1.505231]

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

MSS of errors squared = 5.958619e-02

Corrected MSS of errors squared = 1.032063e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.198084167834828	0	-Inf	0
x1	0.866536446675748	0	Inf	0
x2	-2.22671642713347	0	-Inf	0
x3	3.4845995766605	0	Inf	0
x4	-3.22878882037357	0	-Inf	0
x5	1.76124514288857	0	Inf	0
x6	-0.524881565845925	0	-Inf	0
x7	0.0661107401844604	0	Inf	0
x8	3.82662205872364	0	Inf	0
x9	-8.47961937872162	0	-Inf	0
x10	11.9023430810626	0	Inf	0
x11	-10.1343453824008	0	-Inf	0
x12	5.16782952216567	0	Inf	0
x13	-1.45868957749395	0	-Inf	0
x14	0.17583654329335	0	Inf	0

Number of observations: 451, Error degrees of freedom: 436

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.8248115391138	450	0.0240551367535861		
Model	10.8248115391138	14	0.773200824222412	Inf	0
Residual	0	436	0		

A = 0.001921, B = 0.173228, C = 0.171772

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.175149, 0.418674, 0.645505, 0.863693, 1.076359, 1.285102, 1.490872]

List of powers for D(x): [0.175149, 0.418674, 0.645505, 0.863693, 1.076359, 1.285102, 1.490872]

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

MSS of errors squared = 1.155420e+00

Corrected MSS of errors squared = 2.001246e+00

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.208085343982628	0	-Inf	0
x1	7.87274749481302	0	Inf	0
x2	-35.2987190836814	0	-Inf	0
x3	50.8218773830584	0	Inf	0
x4	0	0	NaN	NaN
x5	-65.6661772163842	0	-Inf	0
x6	58.7584687926542	0	Inf	0
x7	-16.2801119492744	0	-Inf	0
x8	35.9371932617265	0	Inf	0
x9	-158.062175730233	0	-Inf	0
x10	224.412416227318	0	Inf	0
x11	0	0	NaN	NaN
x12	-283.969911851004	0	-Inf	0
x13	251.978027240745	0	Inf	0

x14 -69.2955492339086 0 -Inf 0

Number of observations: 451, Error degrees of freedom: 438

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.8248115391145	450	0.0240551367535877		
Model	10.8248115391145	12	0.902067628259538	Inf	0
Residual	0	438	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

MSS of errors squared = 2.803037e-02

Corrected MSS of errors squared = 4.855002e-02

R-Squared = 1.000000000

R-Squared Adjusted = 1.000000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.00141373994447756	0	Inf	0
x1	-0.32827987358211	0	-Inf	0
x2	0.378688789786403	0	Inf	0
x3	-0.0518223892662673	0	-Inf	0
x4	1773.03586172418	0	Inf	0
x5	-2260.7922571258	0	-Inf	0
x6	592.702998148852	0	Inf	0
x7	-115.183579256126	0	-Inf	0
x8	11.2369762131893	0	Inf	0

Number of observations: 451, Error degrees of freedom: 442

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.8248115391193	450	0.0240551367535985		
Model	10.8248115391193	8	1.35310144238991	Inf	0
Residual	0	442	0		

A = 0.099312, B = 0.329071

order Q(x) = 3.000000, order D(x) = 5.000000

List of powers for Q(x): [0.099312, 0.102014, 0.116365]

List of powers for D(x): [0.099312, 0.102014, 0.116365, 0.142548, 0.177858]

Fitting FresnelCosine(x) in range (0.500000, 5.000000)

MSS of errors squared = 3.862884e-02

Corrected MSS of errors squared = 5.462944e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the Fresnel cosine. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\sqrt{i+C}$	0	0.123	0	Inf	-Inf	2.31725600E-02	7, 7
$A+B/i$	0.019	0.34		Inf	-Inf	2.63836100E-02	3, 5
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	2.80303700E-02	7, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0	0.1	0	Inf	-Inf	2.80303700E-02	7, 7
$A+B*\log_{10}(i)^4$	0.099	0.329		Inf	-Inf	3.86288400E-02	3, 5
$A+B*i$	0	0.1		Inf	-Inf	4.70737700E-02	3, 7
$A+B*i+C*\sqrt{i-1}$	0.226	0.1	0.236	Inf	-Inf	5.95861900E-02	7, 7
$A+B*i-C*(i-1)$	0	0.104	0.15	Inf	-Inf	1.07703000E-01	5, 7
$A+B*i+C*(i-1)$	0.146	0.102	0	Inf	-Inf	9.60451300E-01	7, 7
$A+B*\sqrt{i}+C*(i-1)$	0.001	0.173	0.171	Inf	-Inf	1.15542000E+00	7, 7

RESULTS FOR THE FRESNEL SINE

Using Power $A+B*i$

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.00199499373335929	0	Inf	0
x1	-0.11887993237616	0	-Inf	0
x2	0.338738006359876	0	Inf	0
x3	-0.371435327673556	0	-Inf	0
x4	0.1843718477742	0	Inf	0
x5	-0.0347841332174057	0	-Inf	0
x6	284.799701350957	0	Inf	0
x7	-1286.27379853018	0	-Inf	0
x8	2488.02563924193	0	Inf	0
x9	-2618.79346597405	0	-Inf	0
x10	1574.40838374873	0	Inf	0
x11	-510.92387957874	0	-Inf	0
x12	69.7574022630659	0	Inf	0

Number of observations: 401, Error degrees of freedom: 388

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791571622	400	0.0184114697892905		

Model	7.36458791571622	12	0.613715659643018	Inf	0
Residual	0	388	0		

A = 0.404551, B = 0.100000

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.504551, 0.604551, 0.704551, 0.804551, 0.904551]

List of powers for D(x): [0.504551, 0.604551, 0.704551, 0.804551, 0.904551, 1.004551, 1.104551]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 4.465937e-02

Corrected MSS of errors squared = 6.315789e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.00035859768394241	0	-Inf	0
x1	-4.24712401218693e-05	0	-Inf	0
x2	0.000931380537326324	0	Inf	0
x3	-0.00237869372533758	0	-Inf	0
x4	0	0	NaN	NaN
x5	0.00184838224556664	0	Inf	0
x6	0	0	NaN	NaN
x7	0	0	NaN	NaN
x8	0.0188651469481039	0	Inf	0
x9	-1.01504690196963	0	-Inf	0
x10	4.8978414587383	0	Inf	0
x11	0	0	NaN	NaN
x12	-12.7650719318947	0	-Inf	0
x13	0	0	NaN	NaN
x14	9.86341222774978	0	Inf	0

Number of observations: 401, Error degrees of freedom: 391

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	7.36458791572862	400	0.0184114697893215		
Model	7.36458791572862	9	0.818287546192068	Inf	0
Residual	0	391	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.050000, 0.033333, 0.025000, 0.020000, 0.016667, 0.014286]

List of powers for D(x): [0.100000, 0.050000, 0.033333, 0.025000, 0.020000, 0.016667, 0.014286]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 2.435858e-01

Corrected MSS of errors squared = 3.444823e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}$

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0979231148015773	0	-Inf	0
x1	22.8927899628106	0	Inf	0
x2	-130.599206117056	0	-Inf	0
x3	264.126543012223	0	Inf	0
x4	-206.907296481654	0	-Inf	0
x5	0	0	NaN	NaN
x6	80.4556317107849	0	Inf	0
x7	-29.8705385905962	0	-Inf	0
x8	215.392885400098	0	Inf	0
x9	-1097.34841700918	0	-Inf	0
x10	1673.55407131011	0	Inf	0
x11	0	0	NaN	NaN
x12	-2283.84661904429	0	-Inf	0
x13	2073.99764631814	0	Inf	0
x14	-580.749568205431	0	-Inf	0

Number of observations: 401, Error degrees of freedom: 388

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791571126	400	0.0184114697892781		
Model	7.36458791571126	12	0.613715659642605	Inf	0
Residual	0	388	0		

A = 0.101500, B = 0.109050

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.210549, 0.255719, 0.290379, 0.319599, 0.345342, 0.368616, 0.390018]

List of powers for D(x): [0.210549, 0.255719, 0.290379, 0.319599, 0.345342, 0.368616, 0.390018]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 2.893078e-02

Corrected MSS of errors squared = 4.091431e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	3.49158598554652e-05	0	Inf	0
x1	-0.000442843601534044	0	-Inf	0
x2	0.000744705506273715	0	Inf	0
x3	-0.000336745656562684	0	-Inf	0
x4	120.963900415796	0	Inf	0
x5	-595.243330566703	0	-Inf	0
x6	1097.04503792399	0	Inf	0
x7	-806.254623501242	0	-Inf	0
x8	0	0	NaN	NaN
x9	288.696101434267	0	Inf	0
x10	-104.207085809115	0	-Inf	0

Number of observations: 401, Error degrees of freedom: 391

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791572549	400	0.0184114697893137		
Model	7.36458791572549	9	0.818287546191721	Inf	0
Residual	0	391	0		

A = 0.000000, B = 0.100000, C = 0.678897

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.129572, 0.163673, 0.191805]

List of powers for D(x): [0.129572, 0.163673, 0.191805, 0.216308, 0.238304, 0.258436, 0.277108]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 2.795289e-01

Corrected MSS of errors squared = 4.841582e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*(i-1)

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is A+B*i+C*(i-1)

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0887369040121108	0	-Inf	0
x1	38.6220239122778	0	Inf	0
x2	-173.32073713485	0	-Inf	0
x3	329.727329817249	0	Inf	0
x4	-338.664334201549	0	-Inf	0
x5	197.370457816354	0	Inf	0
x6	-61.7200618646809	0	-Inf	0
x7	8.07408038368541	0	Inf	0
x8	484.31938939023	0	Inf	0
x9	-2236.54824417451	0	-Inf	0
x10	4392.97324487932	0	Inf	0
x11	-4674.81042705441	0	-Inf	0
x12	2833.15836760946	0	Inf	0
x13	-924.958002081186	0	-Inf	0
x14	126.865601069231	0	Inf	0

Number of observations: 401, Error degrees of freedom: 386

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791568046	400	0.0184114697892012		
Model	7.36458791568046	14	0.526041993977176	Inf	0
Residual	0	386	0		

A = 0.500000, B = 0.100000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.600000, 0.700000, 0.800000, 0.900000, 1.000000, 1.100000, 1.200000]

List of powers for D(x): [0.600000, 0.700000, 0.800000, 0.900000, 1.000000, 1.100000, 1.200000]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 6.288563e-02

Corrected MSS of errors squared = 1.089211e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.101851224693886	0	-Inf	0
x1	1.34808278420268	0	Inf	0
x2	-3.69902465136695	0	-Inf	0
x3	4.60062998697109	0	Inf	0
x4	-3.04757849146647	0	-Inf	0
x5	1.04676147166677	0	Inf	0
x6	-0.147010091686941	0	-Inf	0
x7	13.6963104942697	0	Inf	0
x8	-38.6304935328993	0	-Inf	0
x9	49.7082717677606	0	Inf	0
x10	-34.296831478157	0	-Inf	0
x11	12.3572036492653	0	Inf	0
x12	-1.83449243882706	0	-Inf	0

Number of observations: 401, Error degrees of freedom: 388

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791572757	400	0.0184114697893189		
Model	7.36458791572757	12	0.613715659643964	Inf	0
Residual	0	388	0		

A = 0.030000, B = 0.221616, C = 0.085551

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.251616, 0.387681, 0.523745, 0.659810, 0.795875, 0.931940]

List of powers for D(x): [0.251616, 0.387681, 0.523745, 0.659810, 0.795875, 0.931940]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 3.074406e-02

Corrected MSS of errors squared = 5.325028e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0579945361640116	0	-Inf	0
x1	0.349220781214121	0	Inf	0
x2	-0.896051497912748	0	-Inf	0
x3	1.26921688827317	0	Inf	0
x4	-1.07104137468937	0	-Inf	0
x5	0.537925730194566	0	Inf	0
x6	-0.14869883533214	0	-Inf	0
x7	0.017422886801178	0	Inf	0
x8	6.29192421506499	0	Inf	0
x9	-16.9664420006874	0	-Inf	0
x10	25.4168981136274	0	Inf	0
x11	-22.8454144377617	0	-Inf	0
x12	12.3201741456938	0	Inf	0
x13	-3.69104127501231	0	-Inf	0
x14	0.473901102439784	0	Inf	0

Number of observations: 401, Error degrees of freedom: 386

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791572831	400	0.0184114697893208		
Model	7.36458791572831	14	0.526041993980594	Inf	0
Residual	0	386	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 3.876296e-02

Corrected MSS of errors squared = 6.713941e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0124695339762302	0	-Inf	0
x1	0.98065893416895	0	Inf	0
x2	-3.71319729818831	0	-Inf	0
x3	4.19931382335418	0	Inf	0
x4	0	0	NaN	NaN
x5	-2.57993993090229	0	-Inf	0
x6	1.12563412989723	0	Inf	0
x7	152.905342948801	0	Inf	0
x8	-765.501584735289	0	-Inf	0
x9	1157.63278754769	0	Inf	0
x10	0	0	NaN	NaN
x11	-1566.95432501426	0	-Inf	0
x12	1419.94952514857	0	Inf	0

x13 -397.031746296024 0 -Inf 0

Number of observations: 401, Error degrees of freedom: 389

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791573667	400	0.0184114697893417		
Model	7.36458791573667	11	0.669507992339697	Inf	0
Residual	0	389	0		

A = 0.071045, B = 0.100000, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.171045, 0.212466, 0.244250, 0.271045, 0.294652, 0.315994]

List of powers for D(x): [0.171045, 0.212466, 0.244250, 0.271045, 0.294652, 0.315994, 0.335620]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 9.070531e-02

Corrected MSS of errors squared = 1.571062e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.080855685748405	0	-Inf	0
x1	3.53231997351224	0	Inf	0
x2	-22.4265452733347	0	-Inf	0
x3	60.4128265230683	0	Inf	0
x4	-85.8597339752778	0	-Inf	0
x5	67.8426976078292	0	Inf	0
x6	-28.292559433108	0	-Inf	0
x7	4.87185662893488	0	Inf	0
x8	47.6089032464505	0	Inf	0
x9	-318.876002406532	0	-Inf	0
x10	900.740713154109	0	Inf	0
x11	-1339.21545830862	0	-Inf	0

x12	1105.7370912458	0	Inf	0
x13	-481.592223968966	0	-Inf	0
x14	86.5969565148014	0	Inf	0

Number of observations: 401, Error degrees of freedom: 386

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791576563	400	0.0184114697894141		
Model	7.36458791576563	14	0.526041993983259	Inf	0
Residual	0	386	0		

A = 0.000000, B = 0.388152, C = 0.006714

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.388152, 0.555644, 0.681794, 0.787933, 0.881362, 0.965787, 1.043399]

List of powers for D(x): [0.388152, 0.555644, 0.681794, 0.787933, 0.881362, 0.965787, 1.043399]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 3.714196e-02

Corrected MSS of errors squared = 6.433176e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*log10(i)^4

Fitting FresnelSine(x) in range (1.000000, 5.000000)

Pade-Shammas polynomial power is A+B*log10(i)^4

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-6.14454988943739e-10	0	-Inf	0
x1	0	0	NaN	NaN
x2	7.99695443593263e-10	0	Inf	0
x3	-2.31059966682614e-10	0	-Inf	0
x4	5.15417348892737e-11	0	Inf	0
x5	-5.72046212932854e-12	0	-Inf	0
x6	1.00000000081081	0	Inf	0
x7	-1.05826565575761e-09	0	-Inf	0
x8	3.10258908095811e-10	0	Inf	0

x9	-7.09044302804523e-11	0	-Inf	0
x10	8.09768419964426e-12	0	Inf	0

Number of observations: 401, Error degrees of freedom: 391

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.36458791572851	400	0.0184114697893213		
Model	7.36458791572851	9	0.818287546192056	Inf	0
Residual	0	391	0		

A = 0.000000, B = 1.640122

order Q(x) = 5.000000, order D(x) = 5.000000

List of powers for Q(x): [0.000000, 0.013468, 0.084995, 0.215494, 0.391481]

List of powers for D(x): [0.000000, 0.013468, 0.084995, 0.215494, 0.391481]

Fitting FresnelSine(x) in range (1.000000, 5.000000)

MSS of errors squared = 5.377593e-01

Corrected MSS of errors squared = 7.605066e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the Fresnel sine function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*sqrt(i)	0.101	0.109		Inf	-Inf	2.89307800E-02	7, 7
A+B*i-C*(i-1)	0.03	0.221	0.085	Inf	-Inf	3.07440600E-02	6, 6
A+B*sqrt(i)+C*sqrt(i-1)	0	0.388	0.006	Inf	-Inf	3.71419600E-02	7, 7
A+B*i+C*sqrt(i-1)	0	0.1	0	Inf	-Inf	3.87629600E-02	7, 7
A+B*i	0.404	0.1		Inf	-Inf	4.46593700E-02	5, 7
A+B*i+C*(i-1)	0.5	0.1	0	Inf	-Inf	6.28856300E-02	7, 7
A+B*sqrt(i)+C*(i-1)	0.071	0.1	0	Inf	-Inf	9.07053100E-02	6, 7
A+B/i	0	0.1		Inf	-Inf	2.43585800E-01	7, 7
A+B*sqrt(i+C)	0	0.1	0.678	Inf	-Inf	2.79528900E-01	3, 7
A+B*log10(i)^4	0	1.64		Inf	-Inf	5.37759300E-01	5, 5

RESULTS FOR THE BESSEL JO(X)

Using Power A+B* i

Fitting JO(x) in range (2.000000, 30.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.04516654041521e-06	0	Inf	0
x1	-2.43417694456647e-06	0	-Inf	0
x2	1.87951487414986e-06	0	Inf	0
x3	-4.81277384114398e-07	0	-Inf	0
x4	5.68138955718568	0	Inf	0
x5	-13.8144505323762	0	-Inf	0
x6	18.6355902043765	0	Inf	0
x7	-15.0629146159393	0	-Inf	0
x8	7.29514011172701	0	Inf	0
x9	-1.96018043599893	0	-Inf	0
x10	0.225421378873146	0	Inf	0

Number of observations: 281, Error degrees of freedom: 270

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.94294139670076	280	0.0283676478453599		
Model	7.94294139670076	10	0.794294139670076	Inf	0
Residual	0	270	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.200000, 0.300000]
List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
Fitting J0(x) in range (2.000000, 30.000000)
MSS of errors squared = 4.357535e-02
Corrected MSS of errors squared = 6.162485e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power A+B/i

```

Fitting J0(x) in range (2.000000, 30.000000)
Pade-Shammas polynomial power is A+B/i

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.00330917382956018	0	Inf	0
x1	-0.00956141906123751	0	-Inf	0
x2	0.495001336387036	0	Inf	0
x3	-2.75485561648368	0	-Inf	0
x4	3.66170048488443	0	Inf	0
x5	0	0	NaN	NaN
x6	-1.39559363373393	0	-Inf	0
x7	-3.43369129595611	0	-Inf	0
x8	448.027661252887	0	Inf	0
x9	-5232.36208274088	0	-Inf	0
x10	16442.6919675133	0	Inf	0
x11	-16153.1478822938	0	-Inf	0
x12	0	0	NaN	NaN
x13	4499.2240240517	0	Inf	0

Number of observations: 281, Error degrees of freedom: 269

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.9429413970904	280	0.0283676478467514		
Model	7.9429413970904	11	0.722085581553673	Inf	0

Total	7.94294139668597	280	0.028367647845307		
Model	7.94294139668597	13	0.610995492052767	Inf	0
Residual	0	267	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting J0(x) in range (2.000000, 30.000000)

MSS of errors squared = 1.448260e-01

Corrected MSS of errors squared = 2.048149e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i+C)

Fitting J0(x) in range (2.000000, 30.000000)

Pade-Shammas polynomial power is A+B*sqrt(i+C)

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0591678772506568	0	-Inf	0
x1	23.8603240733894	0	Inf	0
x2	-113.051849583973	0	-Inf	0
x3	199.412650675127	0	Inf	0
x4	-140.642751927779	0	-Inf	0
x5	0	0	NaN	NaN
x6	46.8882230340302	0	Inf	0
x7	-16.4074285249095	0	-Inf	0
x8	396.048533868575	0	Inf	0
x9	-1871.24541385415	0	-Inf	0
x10	3292.76524726205	0	Inf	0
x11	-2317.3963433184	0	-Inf	0
x12	0	0	NaN	NaN
x13	769.740182804549	0	Inf	0
x14	-268.912213299745	0	-Inf	0

Number of observations: 281, Error degrees of freedom: 268

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.94294139671335	280	0.0283676478454048		
Model	7.94294139671335	12	0.661911783059446	Inf	0
Residual	0	268	0		

A = 0.006906, B = 0.100000, C = 1.675297

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.170469, 0.198617, 0.223130, 0.245135, 0.265272, 0.283949, 0.301444]

List of powers for D(x): [0.170469, 0.198617, 0.223130, 0.245135, 0.265272, 0.283949, 0.301444]

Fitting J0(x) in range (2.000000, 30.000000)

MSS of errors squared = 3.656333e-02

Corrected MSS of errors squared = 6.332954e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*(i-1)

Fitting J0(x) in range (2.000000, 30.000000)

Pade-Shammas polynomial power is A+B*i+C*(i-1)

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.00101761092894995	0	-Inf	0
x1	0.00347938367623816	0	Inf	0
x2	-0.00500986866387603	0	-Inf	0
x3	0.00374391073032639	0	Inf	0
x4	-0.0014082196586	0	-Inf	0
x5	0.000212050794586773	0	Inf	0
x6	4.65775800585941	0	Inf	0
x7	-10.1471238751472	0	-Inf	0
x8	12.7609113180759	0	Inf	0
x9	-9.70187455438361	0	-Inf	0
x10	4.43324091899871	0	Inf	0
x11	-1.12519621667696	0	-Inf	0
x12	0.122271864922209	0	Inf	0

Number of observations: 281, Error degrees of freedom: 268

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.94294139670021	280	0.0283676478453579		
Model	7.94294139670021	12	0.661911783058351	Inf	0
Residual	0	268	0		

A = 0.001816, B = 0.109502, C = 0.015475

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.111318, 0.236295, 0.361271, 0.486248, 0.611225]

List of powers for D(x): [0.111318, 0.236295, 0.361271, 0.486248, 0.611225, 0.736201, 0.861178]

Fitting J0(x) in range (2.000000, 30.000000)

MSS of errors squared = 1.163867e-01

Corrected MSS of errors squared = 2.015876e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting J0(x) in range (2.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.00159409033900336	0	Inf	0
x1	-0.411765675166239	0	-Inf	0
x2	1.85296510423389	0	Inf	0
x3	-3.15099940106463	0	-Inf	0
x4	2.40356918029008	0	Inf	0
x5	-0.69536326563267	0	-Inf	0
x6	-200.904560003592	0	-Inf	0
x7	1014.45537448284	0	Inf	0
x8	-1911.81479299676	0	-Inf	0
x9	1454.55202606522	0	Inf	0
x10	0	0	NaN	NaN
x11	-589.657993355093	0	-Inf	0
x12	234.369945449556	0	Inf	0

Number of observations: 281, Error degrees of freedom: 269

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.94294139670545	280	0.0283676478453766		
Model	7.94294139670545	11	0.722085581518678	Inf	0
Residual	0	269	0		

A = 0.082804, B = 0.100000, C = 0.114728

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.182804, 0.168076, 0.153347, 0.138619, 0.123890]

List of powers for D(x): [0.182804, 0.168076, 0.153347, 0.138619, 0.123890, 0.109162, 0.094434]

Fitting J0(x) in range (2.000000, 30.000000)

MSS of errors squared = 1.990995e-02

Corrected MSS of errors squared = 3.448504e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting J0(x) in range (2.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.09900116585693e-06	0	Inf	0
x1	-4.68977422456532e-06	0	-Inf	0
x2	3.46802347047324e-06	0	Inf	0
x3	-8.49188660111527e-07	0	-Inf	0
x4	5.48012452598476	0	Inf	0
x5	-12.8464061533139	0	-Inf	0
x6	16.6986331183066	0	Inf	0
x7	-12.9991702475114	0	-Inf	0
x8	6.06024317013542	0	Inf	0
x9	-1.56669820286688	0	-Inf	0
x10	0.173261784806812	0	Inf	0

Number of observations: 281, Error degrees of freedom: 270

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	7.94294139670023	280	0.028367647845358		
Model	7.94294139670023	10	0.794294139670023	Inf	0
Residual	0	270	0		

A = 0.000000, B = 0.117216, C = 0.000000

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.117216, 0.234433, 0.351649]

List of powers for D(x): [0.117216, 0.234433, 0.351649, 0.468866, 0.586082, 0.703298, 0.820515]

Fitting J0(x) in range (2.000000, 30.000000)

MSS of errors squared = 1.728190e-01

Corrected MSS of errors squared = 2.993314e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting J0(x) in range (2.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.000256385150614329	0	-Inf	0
x1	0.00390338492099405	0	Inf	0
x2	-0.0126936194583862	0	-Inf	0
x3	0.0172897275379227	0	Inf	0
x4	-0.0108147398516984	0	-Inf	0
x5	0.00257161183205898	0	Inf	0
x6	35.6424689382841	0	Inf	0
x7	-208.570609950533	0	-Inf	0
x8	536.834746469554	0	Inf	0
x9	-740.657748800791	0	-Inf	0
x10	572.930793970429	0	Inf	0
x11	-235.226478428876	0	-Inf	0
x12	40.0468272256238	0	Inf	0

Number of observations: 281, Error degrees of freedom: 268

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	7.94294139668666	280	0.0283676478453095		
Model	7.94294139668666	12	0.661911783057222	Inf	0
Residual	0	268	0		

A = 0.000000, B = 0.124642, C = 0.008693

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.124642, 0.184963, 0.233271, 0.275361, 0.313478]

List of powers for D(x): [0.124642, 0.184963, 0.233271, 0.275361, 0.313478, 0.348772, 0.381927]

Fitting J0(x) in range (2.000000, 30.000000)

MSS of errors squared = 5.602324e-02

Corrected MSS of errors squared = 9.703510e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

Fitting J0(x) in range (2.000000, 30.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.00662744824103848	0	Inf	0
x1	-4.52320506057333	0	-Inf	0
x2	26.6021626053376	0	Inf	0
x3	-54.4583327945364	0	-Inf	0
x4	42.8774542835066	0	Inf	0
x5	0	0	NaN	NaN
x6	-16.6911150230791	0	-Inf	0
x7	6.18640947474629	0	Inf	0
x8	2366.62223105527	0	Inf	0
x9	-21041.9152941333	0	-Inf	0
x10	70502.2846369557	0	Inf	0
x11	-118886.723800065	0	-Inf	0
x12	108560.575377353	0	Inf	0
x13	-51458.0998144956	0	-Inf	0
x14	9958.25665309183	0	Inf	0

Number of observations: 281, Error degrees of freedom: 267

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.94294139795548	280	0.028367647849841		
Model	7.94294139795548	13	0.610995492150421	Inf	0
Residual	0	267	0		

A = 0.191058, B = 0.100000, C = 0.000083

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.291058, 0.332563, 0.364381, 0.391203, 0.414832, 0.436194, 0.455838]

List of powers for D(x): [0.291058, 0.332563, 0.364381, 0.391203, 0.414832, 0.436194, 0.455838]

Fitting J0(x) in range (2.000000, 30.000000)

MSS of errors squared = 4.250955e-01

Corrected MSS of errors squared = 7.362869e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting J0(x) in range (2.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.99688531505215	0	Inf	0
x1	-11.9946573886114	0	-Inf	0
x2	9.81111162457279	0	Inf	0
x3	-0.889766423417033	0	-Inf	0
x4	0.0823558157397766	0	Inf	0
x5	-0.00617399065954897	0	-Inf	0
x6	0.000293607105718371	0	Inf	0
x7	-6.3082653463866e-06	0	-Inf	0
x8	3.88784155068136	0	Inf	0
x9	-3.12797020508908	0	-Inf	0
x10	0.258890436009218	0	Inf	0
x11	-0.0198616091910341	0	-Inf	0
x12	0.00107820945213439	0	Inf	0
x13	-2.76201222483588e-05	0	-Inf	0

Number of observations: 281, Error degrees of freedom: 267

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.94294139670038	280	0.0283676478453585		
Model	7.94294139670038	13	0.610995492053876	Inf	0
Residual	0	267	0		

A = 0.051323, B = 3.500000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.051323, 0.080065, 0.232701, 0.511185, 0.886738, 1.334611, 1.836563]

List of powers for D(x): [0.051323, 0.080065, 0.232701, 0.511185, 0.886738, 1.334611]

Fitting J₀(x) in range (2.000000, 30.000000)

MSS of errors squared = 1.844570e-01

Corrected MSS of errors squared = 2.608616e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the Bessel J₀(x) function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*i-C*(i-1)	0.082	0.1	0.114	Inf	-Inf	1.99099500E-02	5, 7
A+B/i	0.136	0.181		Inf	-Inf	2.07852300E-02	6, 7
A+B*sqrt(i+C)	0.006	0.1	1.675	Inf	-Inf	3.65633300E-02	7, 7
A+B*i	0	0.1		Inf	-Inf	4.35753500E-02	3, 7
A+B*sqrt(i)+C*(i-1)	0	0.124	0.008	Inf	-Inf	5.60232400E-02	5, 7
A+B*i+C*(i-1)	0.001	0.109	0.015	Inf	-Inf	1.16386700E-01	5, 7
A+B*sqrt(i)	0	0.1		Inf	-Inf	1.44826000E-01	7, 7
A+B*i+C*sqrt(i-1)	0	0.117	0	Inf	-Inf	1.72819000E-01	3, 7
A+B*log10(i)^4	0.051	3.5		Inf	-Inf	1.84457000E-01	7, 6
A+B*sqrt(i)+C*sqrt(i-1)	0.191	0.1	0	Inf	-Inf	4.25095500E-01	7, 7

RESULTS FOR THE BESSEL J1(X)

Using Power A+B* i

Fitting J1(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.38023825616514e-13	1.2209214126979e-08	-2.768596095547e-05	0.999977929100345
x1	-0.0209474946857677	0.00054886113840632	-38.1653814051967	3.0283871065861e-114
x2	0.115002211362598	0.00294336565162512	39.0716699772254	1.08069172934225e-116
x3	-0.257334361208171	0.00648126413255296	-39.704347168275	2.22865362329464e-118
x4	0.301887159056333	0.00752304222953422	40.1283350332893	1.69353270972961e-119
x5	-0.1964837351927	0.00486507113960385	-40.3866109157654	3.5566291743298e-120
x6	0.0674312718771317	0.00166458192369043	40.5094341813075	1.69745912969672e-120
x7	-0.00955015916125282	0.000235693949337575	-40.5193225710453	1.59943457714813e-120
x8	5.76390624995364	0.00748223452601495	770.345573894153	0
x9	-14.2429238407241	0.03717613666259	-383.120063550247	0
x10	19.5552702750191	0.0769056368772311	254.276163218521	0
x11	-16.1084154677452	0.0847732025946358	-190.017776546341	5.13198330770138e-303
x12	7.95957254353815	0.0525088755111721	151.585278984781	3.16418146694421e-275
x13	-2.1841373951198	0.0173263612441347	-126.058631950731	1.16413715865071e-252
x14	0.256716536589612	0.00237919337587932	107.900660447465	9.7029777733081e-234

Number of observations: 301, Error degrees of freedom: 286

Root Mean Squared Error: 1.22e-08

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: 5.98e+15, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	12.4852228770457	300	0.0416174095901524		
Model	12.4852228770457	14	0.89180163407469	5.98263961966381e+15	0
Residual	4.2632564145606e-14	286	1.49064909600021e-16		

A = 0.000000, B = 0.100000

```

order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
Fitting J1(x) in range (0.000000, 30.000000)
MSS of errors squared = 8.049462e-02
Corrected MSS of errors squared = 1.138366e-01
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.012045e+04
AIC = -1.015213e+04
AICc = -1.012045e+04

```

Using Power A+B/i

```

Fitting J1(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B/i

```

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	3.16161968215517e-14	0	Inf	0
x1	2.72643140450986e-05	0	Inf	0
x2	-0.00221568798820221	0	-Inf	0
x3	0.0214095171510719	0	Inf	0
x4	-0.0606715703295075	0	-Inf	0
x5	0.0558332079122937	0	Inf	0
x6	0	0	NaN	NaN
x7	-0.0143827310069892	0	-Inf	0
x8	-0.00577313312708963	0	-Inf	0
x9	0.972605705115855	0	Inf	0
x10	-14.2811248878197	0	-Inf	0
x11	54.3232798480489	0	Inf	0
x12	-62.7445887013746	0	-Inf	0
x13	0	0	NaN	NaN
x14	22.7356011690373	0	Inf	0

Number of observations: 301, Error degrees of freedom: 288

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	12.4852228770435	300	0.0416174095901451		
Model	12.4852228770435	12	1.04043523975363	Inf	0

Residual 0 288 0

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.050000, 0.033333, 0.025000, 0.020000, 0.016667, 0.014286]

List of powers for D(x): [0.100000, 0.050000, 0.033333, 0.025000, 0.020000, 0.016667, 0.014286]

Fitting J1(x) in range (0.000000, 30.000000)

MSS of errors squared = 1.928379e-01

Corrected MSS of errors squared = 2.727140e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting J1(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.7686219422259e-12	0	Inf	0
x1	8.82000514705104e-05	0	Inf	0
x2	-0.000314274026098757	0	-Inf	0
x3	0.000360350713007178	0	Inf	0
x4	-0.000134231861217747	0	-Inf	0
x5	70.7714698407228	0	Inf	0
x6	-527.628298999445	0	-Inf	0
x7	1628.34677598811	0	Inf	0
x8	-2616.90697951694	0	-Inf	0
x9	2316.34740760438	0	Inf	0
x10	-1074.84826806061	0	-Inf	0
x11	204.917893037115	0	Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	12.4852228771261	300	0.0416174095904203		
Model	12.4852228771261	11	1.13502026155692	Inf	0

Residual 0 289 0

A = 0.000000, B = 0.100000

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting J1(x) in range (0.000000, 30.000000)

MSS of errors squared = 1.004593e-01

Corrected MSS of errors squared = 1.420709e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting J1(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	3.64996081668451e-13	0	Inf	0
x1	-3.92266700022375	0	-Inf	0
x2	32.2720287036765	0	Inf	0
x3	-104.80944645705	0	-Inf	0
x4	174.894637154774	0	Inf	0
x5	-159.71059968095	0	-Inf	0
x6	76.1508702072661	0	Inf	0
x7	-14.8748226127619	0	-Inf	0
x8	688.942477175452	0	Inf	0
x9	-4668.85295775682	0	-Inf	0
x10	13118.5563572754	0	Inf	0
x11	-19497.5780617789	0	-Inf	0
x12	16166.7986962779	0	Inf	0
x13	-7096.078581312	0	-Inf	0
x14	1289.21206940439	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	12.4852228770842	300	0.0416174095902806		
Model	12.4852228770842	14	0.891801634077442	Inf	0
Residual	0	286	0		

A = 0.000000, B = 0.100000, C = 1.399721

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.154910, 0.184383, 0.209755, 0.232373, 0.252977, 0.272024, 0.289823]

List of powers for D(x): [0.154910, 0.184383, 0.209755, 0.232373, 0.252977, 0.272024, 0.289823]

Fitting J1(x) in range (0.000000, 30.000000)

MSS of errors squared = 5.623638e-02

Corrected MSS of errors squared = 9.740426e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*(i-1)

Fitting J1(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B*i+C*(i-1)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.38023825616514e-13	1.2209214126979e-08	-2.768596095547e-05	0.999977929100345
x1	-0.0209474946857677	0.00054886113840632	-38.1653814051967	3.0283871065861e-114
x2	0.115002211362598	0.00294336565162512	39.0716699772254	1.08069172934225e-116
x3	-0.257334361208171	0.00648126413255296	-39.704347168275	2.22865362329464e-118
x4	0.301887159056333	0.00752304222953422	40.1283350332893	1.69353270972961e-119
x5	-0.1964837351927	0.00486507113960385	-40.3866109157654	3.5566291743298e-120
x6	0.0674312718771317	0.00166458192369043	40.5094341813075	1.69745912969672e-120
x7	-0.00955015916125282	0.000235693949337575	-40.5193225710453	1.59943457714813e-120
x8	5.76390624995364	0.00748223452601495	770.345573894153	0
x9	-14.2429238407241	0.03717613666259	-383.120063550247	0
x10	19.5552702750191	0.0769056368772311	254.276163218521	0
x11	-16.1084154677452	0.0847732025946358	-190.017776546341	5.13198330770138e-303
x12	7.95957254353815	0.0525088755111721	151.585278984781	3.16418146694421e-275
x13	-2.1841373951198	0.0173263612441347	-126.058631950731	1.16413715865071e-252
x14	0.256716536589612	0.00237919337587932	107.900660447465	9.7029777733081e-234

Number of observations: 301, Error degrees of freedom: 286

Root Mean Squared Error: 1.22e-08

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: 5.98e+15, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	12.4852228770457	300	0.0416174095901524		
Model	12.4852228770457	14	0.89180163407469	5.98263961966381e+15	0
Residual	4.2632564145606e-14	286	1.49064909600021e-16		

```

A = 0.000000, B = 0.100000, C = 0.000000
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
Fitting J1(x) in range (0.000000, 30.000000)
MSS of errors squared = 8.049462e-02
Corrected MSS of errors squared = 1.394208e-01
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.012045e+04
AIC = -1.015213e+04
AICc = -1.012045e+04

```

Using Power $A+B*i-C*(i-1)$

```

Fitting J1(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i-C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-9.16018129644335e-06	0	-Inf	0
x1	-1.730200353928e-06	0	-Inf	0
x2	8.86855508466071e-06	0	Inf	0
x3	-1.78030834347004e-05	0	-Inf	0
x4	1.98216170682694e-05	0	Inf	0
x5	-0.00301213626380203	0	-Inf	0
x6	0.028487389676674	0	Inf	0
x7	-0.131927312133106	0	-Inf	0
x8	0.542411683088947	0	Inf	0
x9	0.640108014930607	0	Inf	0
x10	-0.085529256220343	0	-Inf	0
x11	0.00946162456165389	0	Inf	0

Number of observations: 300, Error degrees of freedom: 288

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	12.4839313141885	299	0.0417522786427708		
Model	12.4839313141885	11	1.1349028467444	Inf	0
Residual	0	288	0		

```

A = 0.142566, B = 0.164313, C = 0.250000
order Q(x) = 4.000000, order D(x) = 7.000000
List of powers for Q(x): [0.306880, 0.221193, 0.135507, 0.049820]
List of powers for D(x): [0.306880, 0.221193, 0.135507, 0.049820, -0.035866, -
0.121553, -0.207239]
Fitting J1(x) in range (0.000000, 30.000000)
MSS of errors squared = NaN
Corrected MSS of errors squared = NaN
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*\sqrt{i-1}$

```

Fitting J1(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.38023825616514e-13	1.2209214126979e-08	-2.768596095547e-05	0.999977929100345
x1	-0.0209474946857677	0.00054886113840632	-38.1653814051967	3.0283871065861e-114
x2	0.115002211362598	0.00294336565162512	39.0716699772254	1.08069172934225e-116
x3	-0.257334361208171	0.00648126413255296	-39.704347168275	2.22865362329464e-118
x4	0.301887159056333	0.00752304222953422	40.1283350332893	1.69353270972961e-119
x5	-0.1964837351927	0.00486507113960385	-40.3866109157654	3.5566291743298e-120
x6	0.0674312718771317	0.00166458192369043	40.5094341813075	1.69745912969672e-120
x7	-0.00955015916125282	0.000235693949337575	-40.5193225710453	1.59943457714813e-120
x8	5.76390624995364	0.00748223452601495	770.345573894153	0
x9	-14.2429238407241	0.03717613666259	-383.120063550247	0
x10	19.5552702750191	0.0769056368772311	254.276163218521	0
x11	-16.1084154677452	0.0847732025946358	-190.017776546341	5.13198330770138e-303
x12	7.95957254353815	0.0525088755111721	151.585278984781	3.16418146694421e-275
x13	-2.1841373951198	0.0173263612441347	-126.058631950731	1.16413715865071e-252
x14	0.256716536589612	0.00237919337587932	107.900660447465	9.7029777733081e-234

Number of observations: 301, Error degrees of freedom: 286

Root Mean Squared Error: 1.22e-08

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: 5.98e+15, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	12.4852228770457	300	0.0416174095901524		
Model	12.4852228770457	14	0.89180163407469	5.98263961966381e+15	0
Residual	4.2632564145606e-14	286	1.49064909600021e-16		

```

A = 0.000000, B = 0.100000, C = 0.000000
order Q(x) = 7.000000, order D(x) = 7.000000

```

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

Fitting J1(x) in range (0.000000, 30.000000)

MSS of errors squared = 8.049462e-02

Corrected MSS of errors squared = 1.394208e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.012045e+04

AIC = -1.015213e+04

AICc = -1.012045e+04

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting J1(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	4.16779993686073e-14	0	Inf	0
x1	-0.231122988487284	0	-Inf	0
x2	2.43275844472221	0	Inf	0
x3	-9.16838064406484	0	-Inf	0
x4	16.9549032069839	0	Inf	0
x5	-16.7159189480753	0	-Inf	0
x6	8.46357898666374	0	Inf	0
x7	-1.7358179747193	0	-Inf	0
x8	65.7731396031835	0	Inf	0
x9	-475.962740303627	0	-Inf	0
x10	1436.00629466761	0	Inf	0
x11	-2264.5103213503	0	-Inf	0
x12	1971.5038847276	0	Inf	0
x13	-901.327008725975	0	-Inf	0
x14	169.516751192986	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	12.4852228771752	300	0.041617409590584		
Model	12.4852228771752	14	0.891801634083943	Inf	0
Residual	0	286	0		

```

A = 0.000000, B = 0.100000, C = 0.000000
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949,
0.264575]
List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949,
0.264575]
Fitting J1(x) in range (0.000000, 30.000000)
MSS of errors squared = 5.611945e-02
Corrected MSS of errors squared = 9.720173e-02
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

```

Fitting J1(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	4.55437568568894e-12	0	Inf	0
x1	-1.49308157476752	0	-Inf	0
x2	15.7169877866518	0	Inf	0
x3	-59.2355520859034	0	-Inf	0
x4	109.546744817736	0	Inf	0
x5	-108.005447577344	0	-Inf	0
x6	54.6863053729183	0	Inf	0
x7	-11.2159561986887	0	-Inf	0
x8	259.116582844721	0	Inf	0
x9	-2093.56452875217	0	-Inf	0
x10	6659.07011498771	0	Inf	0
x11	-10850.857003227	0	-Inf	0
x12	9665.93931265067	0	Inf	0
x13	-4495.77634598328	0	-Inf	0
x14	857.071866252072	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	12.4852228767456	300	0.0416174095891522		
Model	12.4852228767456	14	0.891801634053261	Inf	0
Residual	0	286	0		

A = 0.055870, B = 0.100000, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.155870, 0.197291, 0.229075, 0.255870, 0.279477, 0.300819, 0.320445]

List of powers for D(x): [0.155870, 0.197291, 0.229075, 0.255870, 0.279477, 0.300819, 0.320445]

Fitting J1(x) in range (0.000000, 30.000000)

MSS of errors squared = 2.946177e-01

Corrected MSS of errors squared = 5.102928e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting J1(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.54379741130639e-13	0	-Inf	0
x1	-47.2096607996728	0	-Inf	0
x2	63.3256777451639	0	Inf	0
x3	-21.6946419687327	0	-Inf	0
x4	6.98930228570547	0	Inf	0
x5	-1.58760434322143	0	-Inf	0
x6	0.176927083013585	0	Inf	0
x7	25.8607253716613	0	Inf	0
x8	0	0	NaN	NaN
x9	-48.6625434344838	0	-Inf	0
x10	32.7013278537112	0	Inf	0
x11	-10.2481014442616	0	-Inf	0
x12	1.34859164835435	0	Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	12.4852228770439	300	0.0416174095901464		
Model	12.4852228770439	11	1.13502026154945	Inf	0
Residual		0	289	0	

A = 0.017685, B = 0.100000

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.017685, 0.018506, 0.022868, 0.030824, 0.041554, 0.054351]

List of powers for D(x): [0.017685, 0.018506, 0.022868, 0.030824, 0.041554, 0.054351]

Fitting J₁(x) in range (0.000000, 30.000000)

MSS of errors squared = 9.417841e-01

Corrected MSS of errors squared = 1.331884e+00

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the Bessel J₁(x) function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*sqrt(i)+C*(i-1)	0	0.1	0	Inf	-Inf	5.61194500E-02	7, 7
A+B*sqrt(i+C)	0	0.1	1.399	Inf	-Inf	5.62363800E-02	7, 7
A+B*i	0	0.1		5.90E+15	-1.01E+04	8.04946200E-02	7, 7
A+B*i+C*(i-1)	0	0.1	0	5.90E+15	-1.01E+04	8.04946200E-02	7, 7
A+B*i+C*sqrt(i-1)	0	0.1	0	5.90E+19	-1.01E+04	8.04946200E-02	7, 7
A+B*sqrt(i)	0	0.1		Inf	-Inf	1.00459300E-01	4, 7
A+B/i	0	0.1		Inf	-Inf	1.92837900E-01	7, 7
A+B*sqrt(i)+C*sqrt(i-1)	0.055	0.1	0	Inf	-Inf	2.94617700E-01	7, 7
A+B*log10(i)^4	0.017	0.1		Inf	-Inf	9.41784100E-01	6, 6
A+B*i-C*(i-1)	0.142	0.164	0.25	Inf	-Inf	NaN	4, 7

RESULTS FOR THE BESSEL J2(X)

Using Power A+B* i

Fitting J2(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.59306227962375e-14	6.58223188124914e-09	1.15356955157629e-05	0.999990803860698
x1	-0.00052901216428357	1.77278675124421e-05	-29.8407106163383	5.72343964667716e-90
x2	0.00234398211008186	7.93115124650395e-05	29.5541219329929	4.63649835175916e-89
x3	-0.00408371064449921	0.000140476941380456	-29.0703271609483	1.62207607173096e-87
x4	0.00350833307597933	0.000123365288542111	28.4385755299535	1.75932397388015e-85
x5	-0.00148974749723866	5.37827434797724e-05	-27.6993585832778	4.51277504719217e-83
x6	0.000250575633800697	9.31982191808002e-06	26.8863113483523	2.17949032283642e-80
x7	5.83169133638146	0.00553921409202514	1052.8012168328	0
x8	-14.5567066062165	0.027530519135932	-528.747988163346	0
x9	20.1593422016116	0.0569331332931487	354.08805093883	0
x10	-16.7274001895898	0.0627018617886895	-266.776770456394	0
x11	8.31560312856906	0.0387845952589246	214.404793270483	8.46907567475231e-319
x12	-2.29311829262665	0.0127748863803126	-179.502049909467	8.12851448553961e-297
x13	0.270584882405149	0.00175043874765303	154.581177300804	2.23479236901774e-278

Number of observations: 301, Error degrees of freedom: 287

Root Mean Squared Error: 6.58e-09

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: 1.85e+16, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.4218514727574	300	0.0347395049091915		
Model	10.4218514727574	13	0.801680882519803	1.85035548344044e+16	0
Residual	1.24344978758018e-14	287	4.33257765707378e-17		

A = 0.000000, B = 0.100000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000]
 List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

Fitting J2(x) in range (0.000000, 30.000000)

MSS of errors squared = 4.126671e-01

Corrected MSS of errors squared = 5.835994e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.039855e+04

AIC = -1.042802e+04

AICc = -1.039855e+04

Using Power A+B/i

Fitting J2(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-5.88139809891648e-13	0	-Inf	0
x1	-5.80312080424327e-09	0	-Inf	0
x2	1.19389821383093e-07	0	Inf	0
x3	-3.20172061773455e-07	0	-Inf	0
x4	2.06962046629735e-07	0	Inf	0
x5	0.000904495090255909	0	Inf	0
x6	-0.504470032749013	0	-Inf	0
x7	16.0836676494639	0	Inf	0
x8	-126.642198556212	0	-Inf	0
x9	371.45561822546	0	Inf	0
x10	-449.282077388603	0	-Inf	0
x11	189.888555605128	0	Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.4218514727639	300	0.0347395049092128		
Model	10.4218514727639	11	0.947441042978532	Inf	0
Residual	0	289	0		

A = 0.006158, B = 0.475762

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.481920, 0.244039, 0.164745, 0.125099]

List of powers for D(x): [0.481920, 0.244039, 0.164745, 0.125099, 0.101310, 0.085452, 0.074124]

Fitting J2(x) in range (0.000000, 30.000000)

MSS of errors squared = 1.083663e-01

Corrected MSS of errors squared = 1.532531e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}$

Fitting J2(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	5.59908566110094e-13	0	Inf	0
x1	-0.00150161453973932	0	-Inf	0
x2	0.0117287069866154	0	Inf	0
x3	-0.0325893534157251	0	-Inf	0
x4	0.0423727109036417	0	Inf	0
x5	-0.0263844029314398	0	-Inf	0
x6	0.00637395990816999	0	Inf	0
x7	66.4393608920963	0	Inf	0
x8	-482.60108272921	0	-Inf	0
x9	1460.03411942579	0	Inf	0
x10	-2307.50228618876	0	-Inf	0
x11	2012.70613709862	0	Inf	0
x12	-921.672591031689	0	-Inf	0
x13	173.596342475355	0	Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.4218514727971	300	0.0347395049093237		
Model	10.4218514727971	13	0.801680882522854	Inf	0
Residual	0	287	0		

A = 0.000000, B = 0.100000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949]
 List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting J2(x) in range (0.000000, 30.000000)

MSS of errors squared = 3.522285e-02

Corrected MSS of errors squared = 4.981264e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting J2(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.17879865487033e-12	0	-Inf	0
x1	-0.0054805464110776	0	-Inf	0
x2	0.041390723065588	0	Inf	0
x3	-0.112822289148261	0	-Inf	0
x4	0.144806418602205	0	Inf	0
x5	-0.0893070281790776	0	-Inf	0
x6	0.0214127473106266	0	Inf	0
x7	166.933208285091	0	Inf	0
x8	-1264.32518674895	0	-Inf	0
x9	3876.1125734658	0	Inf	0
x10	-6162.56168891324	0	-Inf	0
x11	5391.66429603207	0	Inf	0
x12	-2472.99787367996	0	-Inf	0
x13	466.174671347867	0	Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.421851473029	300	0.0347395049100968		
Model	10.421851473029	13	0.801680882540696	Inf	0
Residual	0	287	0		

A = 0.025789, B = 0.103500, C = 0.131166

```

order Q(x) = 6.000000, order D(x) = 7.000000
List of powers for Q(x): [0.135868, 0.176883, 0.208933, 0.236155, 0.260237, 0.282067]
List of powers for D(x): [0.135868, 0.176883, 0.208933, 0.236155, 0.260237, 0.282067,
0.302177]
Fitting J2(x) in range (0.000000, 30.000000)
MSS of errors squared = 4.928877e-02
Corrected MSS of errors squared = 8.537066e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting J2(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.97404915585691e-14	0	Inf	0
x1	-0.000529018011990477	0	-Inf	0
x2	0.00234400683903811	0	Inf	0
x3	-0.00408375083485452	0	-Inf	0
x4	0.00350836460105805	0	Inf	0
x5	-0.00148975944324468	0	-Inf	0
x6	0.000250577379113001	0	Inf	0
x7	5.83168528488302	0	Inf	0
x8	-14.5566762197012	0	-Inf	0
x9	20.159278688157	0	Inf	0
x10	-16.7273294645402	0	-Inf	0
x11	8.31555888204158	0	Inf	0
x12	-2.29310354856041	0	-Inf	0
x13	0.270582838125632	0	Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.4218514727575	300	0.0347395049091918		
Model	10.4218514727575	13	0.801680882519811	Inf	0
Residual	0	287	0		

```

A = 0.000000, B = 0.100001, C = 0.000000
order Q(x) = 6.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100001, 0.200001, 0.300002, 0.400002, 0.500003, 0.600003]
List of powers for D(x): [0.100001, 0.200001, 0.300002, 0.400002, 0.500003, 0.600003,
0.700004]
Fitting J2(x) in range (0.000000, 30.000000)
MSS of errors squared = 4.153023e-01
Corrected MSS of errors squared = 7.193247e-01
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i-C*(i-1)$

```

Fitting J2(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i-C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.000137570758536226	0	-Inf	0
x1	4.12348359114004e-06	0	Inf	0
x2	-2.77526581565323e-05	0	-Inf	0
x3	7.90708678110265e-05	0	Inf	0
x4	-0.000128835063798885	0	-Inf	0
x5	0.000171376483218569	0	Inf	0
x6	3.95911577228762e-05	0	Inf	0
x7	0.00299868426808512	0	Inf	0
x8	-0.0278310887150271	0	-Inf	0
x9	0.116672947074024	0	Inf	0
x10	-0.307405071773208	0	-Inf	0
x11	0.772502968200166	0	Inf	0
x12	0.471901050307489	0	Inf	0
x13	-0.0288395191562172	0	-Inf	0

Number of observations: 300, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.4204484729586	299	0.0348509982373198		
Model	10.4204484729586	13	0.801572959458355	Inf	0
Residual	0	286	0		

```

A = 0.317494, B = 0.144719, C = 0.246086
order Q(x) = 6.000000, order D(x) = 7.000000
List of powers for Q(x): [0.462213, 0.360845, 0.259478, 0.158111, 0.056743, -0.044624]
List of powers for D(x): [0.462213, 0.360845, 0.259478, 0.158111, 0.056743, -0.044624,
-0.145991]
Fitting J2(x) in range (0.000000, 30.000000)
MSS of errors squared = NaN
Corrected MSS of errors squared = NaN
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*\sqrt{i-1}$

```

Fitting J2(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.26155585124214e-13	0	Inf	0
x1	-0.0012710330186426	0	-Inf	0
x2	0.00651026297661967	0	Inf	0
x3	-0.0137140446845646	0	-Inf	0
x4	0.0152428894481205	0	Inf	0
x5	-0.009444464259071733	0	-Inf	0
x6	0.00309733476259722	0	Inf	0
x7	-0.000420476701891743	0	-Inf	0
x8	5.85426356939372	0	Inf	0
x9	-14.6673676485813	0	-Inf	0
x10	20.3833417904609	0	Inf	0
x11	-16.9682399387734	0	-Inf	0
x12	8.46109667224148	0	Inf	0
x13	-2.33996132097703	0	-Inf	0
x14	0.27686448371625	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.4218514727575	300	0.0347395049091915		

Model	10.4218514727575	14	0.744417962339818	Inf	0
Residual	0	286	0		

A = 0.000012, B = 0.100000, C = 0.000014

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100012, 0.200026, 0.300032, 0.400036, 0.500040, 0.600043, 0.700046]

List of powers for D(x): [0.100012, 0.200026, 0.300032, 0.400036, 0.500040, 0.600043, 0.700046]

Fitting J2(x) in range (0.000000, 30.000000)

MSS of errors squared = 9.200833e-02

Corrected MSS of errors squared = 1.593631e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting J2(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.60415231923659e-12	0	-Inf	0
x1	-9.20441456749919e-05	0	-Inf	0
x2	0.000525697587515834	0	Inf	0
x3	-0.00102929867301992	0	-Inf	0
x4	0.000849682053761697	0	Inf	0
x5	-0.000254030090882537	0	-Inf	0
x6	67.298365974012	0	Inf	0
x7	-491.44186071745	0	-Inf	0
x8	1492.83617303204	0	Inf	0
x9	-2367.42937439713	0	-Inf	0
x10	2071.20077039495	0	Inf	0
x11	-951.038962242184	0	-Inf	0
x12	179.574887900068	0	Inf	0

Number of observations: 301, Error degrees of freedom: 288

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	10.4218514727559	300	0.0347395049091862		
Model	10.4218514727559	12	0.868487622729656	Inf	0
Residual	0	288	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting J2(x) in range (0.000000, 30.000000)

MSS of errors squared = 3.580406e-01

Corrected MSS of errors squared = 6.201446e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting J2(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.60415231923659e-12	0	-Inf	0
x1	-9.20441456749919e-05	0	-Inf	0
x2	0.000525697587515834	0	Inf	0
x3	-0.00102929867301992	0	-Inf	0
x4	0.000849682053761697	0	Inf	0
x5	-0.000254030090882537	0	-Inf	0
x6	67.298365974012	0	Inf	0
x7	-491.44186071745	0	-Inf	0
x8	1492.83617303204	0	Inf	0
x9	-2367.42937439713	0	-Inf	0
x10	2071.20077039495	0	Inf	0
x11	-951.038962242184	0	-Inf	0
x12	179.574887900068	0	Inf	0

Number of observations: 301, Error degrees of freedom: 288

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	10.4218514727559	300	0.0347395049091862		
Model	10.4218514727559	12	0.868487622729656	Inf	0
Residual	0	288	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting J2(x) in range (0.000000, 30.000000)

MSS of errors squared = 3.580406e-01

Corrected MSS of errors squared = 6.201446e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting J2(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.01973318952329e-15	0	-Inf	0
x1	-1.07718302351161e-07	0	-Inf	0
x2	1.43110178533457e-07	0	Inf	0
x3	-3.89633054116047e-08	0	-Inf	0
x4	4.81708114815555e-09	0	Inf	0
x5	1.01466518285469	0	Inf	0
x6	-0.0152212926909397	0	-Inf	0
x7	0.000592403281997427	0	Inf	0
x8	-3.77721829472083e-05	0	-Inf	0
x9	1.46668021851484e-06	0	Inf	0

Number of observations: 301, Error degrees of freedom: 291

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	10.4218514727574	300	0.0347395049091912		
Model	10.4218514727574	9	1.15798349697304	Inf	0
Residual	0	291	0		

```
A = 0.000278, B = 2.810089
order Q(x) = 4.000000, order D(x) = 5.000000
List of powers for Q(x): [0.000278, 0.023354, 0.145903, 0.369493]
List of powers for D(x): [0.000278, 0.023354, 0.145903, 0.369493, 0.671018]
Fitting J2(x) in range (0.000000, 30.000000)
MSS of errors squared = 4.046963e-02
Corrected MSS of errors squared = 5.723270e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Comments

The following table shows the summary results for the ten models that fit the Bessel $J_2(x)$ function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*sqrt(i)	0	0.1		Inf	-Inf	3.52228500E-02	6, 7
A+B*log10(i)^4	0	2.81		Inf	-Inf	4.04696300E-02	4, 5
A+B*sqrt(i+C)	0.025	0.103	0.131	Inf	-Inf	4.92887700E-02	6, 7
A+B*i+C*sqrt(i-1)	0	0.1	0	Inf	-Inf	9.20083300E-02	7, 7
A+B/i	0.006	0.475		Inf	-Inf	1.08366300E-01	4, 7
A+B*sqrt(i)+C*(i-1)	0	0.1	0	Inf	-Inf	3.58040600E-01	5, 7
A+B*sqrt(i)+C*sqrt(i-1)	0	0.1	0	Inf	-Inf	3.58040600E-01	5, 7
A+B*i	0	0.1		1.80E+16	-1.04E+04	4.12667100E-01	6, 7
A+B*i+C*(i-1)	0	0.1	0	Inf	-Inf	4.15302300E-01	6, 7
A+B*i-C*(i-1)	0.317	0.144	0.246	Inf	-Inf	NaN	6, 7

RESULTS FOR THE BESSEL J3(X)

Using Power A+B* i

Fitting J3(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	9.02259760435892e-14	0	Inf	0
x1	0.000742741912171055	0	Inf	0
x2	-0.00410938322357813	0	-Inf	0
x3	0.00936041689539658	0	Inf	0
x4	-0.0112411054452086	0	-Inf	0
x5	0.00751193021918971	0	Inf	0
x6	-0.00265058782196768	0	-Inf	0
x7	0.000386111056671235	0	Inf	0
x8	5.80440169605711	0	Inf	0
x9	-14.2449555976457	0	-Inf	0
x10	19.3120309767446	0	Inf	0
x11	-15.6739426877951	0	-Inf	0
x12	7.62084271944232	0	Inf	0
x13	-2.05574211064639	0	-Inf	0
x14	0.237358400249723	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	9.18626714756054	300	0.0306208904918685		
Model	9.18626714756054	14	0.656161939111467	Inf	0
Residual	0	286	0		

A = 0.001412, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.101412, 0.201412, 0.301412, 0.401412, 0.501412, 0.601412, 0.701412]

List of powers for D(x): [0.101412, 0.201412, 0.301412, 0.401412, 0.501412, 0.601412, 0.701412]

Fitting J3(x) in range (0.000000, 30.000000)

MSS of errors squared = 2.176376e-01

Corrected MSS of errors squared = 3.077860e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting J3(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	9.39059030150913e-12	0	Inf	0
x1	1.20738230642461e-09	0	Inf	0
x2	-7.54796004921315e-09	0	-Inf	0
x3	6.55122228302082e-09	0	Inf	0
x4	0.0268182373742176	0	Inf	0
x5	-12.1301190864753	0	-Inf	0
x6	335.888956901618	0	Inf	0
x7	-2358.33453603815	0	-Inf	0
x8	6269.44233097965	0	Inf	0
x9	-6954.54833931053	0	-Inf	0
x10	2720.65488830629	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	9.18626714747713	300	0.0306208904915904		

Model	9.18626714747713	10	0.918626714747713	Inf	0
Residual	0	290	0		

A = 0.066579, B = 0.338257

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.404836, 0.235708, 0.179331]

List of powers for D(x): [0.404836, 0.235708, 0.179331, 0.151143, 0.134230, 0.122955, 0.114901]

Fitting J3(x) in range (0.000000, 30.000000)

MSS of errors squared = Inf

Corrected MSS of errors squared = Inf

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting J3(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.09892073556909e-12	0	-Inf	0
x1	2.91398333251731e-08	0	Inf	0
x2	-6.03545597509154e-08	0	-Inf	0
x3	3.14655141993133e-08	0	Inf	0
x4	66.2620651800218	0	Inf	0
x5	-480.70537327249	0	-Inf	0
x6	1452.80654294098	0	Inf	0
x7	-2294.00920370597	0	-Inf	0
x8	1999.29432140965	0	Inf	0
x9	-914.832688715781	0	-Inf	0
x10	172.184336152511	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	9.18626714754912	300	0.0306208904918304		
Model	9.18626714754912	10	0.918626714754912	Inf	0
Residual	0	290	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.141421, 0.173205]
List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949,
0.264575]
Fitting J3(x) in range (0.000000, 30.000000)
MSS of errors squared = 5.338281e-02
Corrected MSS of errors squared = 7.549470e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i+C}$

```

Fitting J3(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i+C}$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.65555140078255e-11	0	-Inf	0
x1	-0.000394424006953832	0	-Inf	0
x2	0.0018378678375255	0	Inf	0
x3	-0.00314403159932709	0	-Inf	0
x4	0.00235237637228296	0	Inf	0
x5	-0.000651783487046811	0	-Inf	0
x6	1432.75267911058	0	Inf	0
x7	-9532.26196802774	0	-Inf	0
x8	26313.3968481274	0	Inf	0
x9	-38501.383238511	0	-Inf	0
x10	31490.9979531363	0	Inf	0
x11	-13657.6394387907	0	-Inf	0
x12	2455.13716467259	0	Inf	0

Number of observations: 301, Error degrees of freedom: 288

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	9.18626714705185	300	0.0306208904901728		
Model	9.18626714705185	12	0.76552226225432	Inf	0
Residual	0	288	0		

```

A = 0.000000, B = 0.100000, C = 2.000000
order Q(x) = 5.000000, order D(x) = 7.000000
List of powers for Q(x): [0.173205, 0.200000, 0.223607, 0.244949, 0.264575]
List of powers for D(x): [0.173205, 0.200000, 0.223607, 0.244949, 0.264575, 0.282843,
0.300000]
Fitting J3(x) in range (0.000000, 30.000000)
MSS of errors squared = 1.711075e-01
Corrected MSS of errors squared = 2.963669e-01
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting J3(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.13262118154776e-13	0	Inf	0
x1	0.000688035625759358	0	Inf	0
x2	-0.00380164430503733	0	-Inf	0
x3	0.00864233859384267	0	Inf	0
x4	-0.0103518526771913	0	-Inf	0
x5	0.00689589292332343	0	Inf	0
x6	-0.00242435476914633	0	-Inf	0
x7	0.000351716898992408	0	Inf	0
x8	5.47095221709184	0	Inf	0
x9	-13.0957738458331	0	-Inf	0
x10	17.543751809922	0	Inf	0
x11	-14.1152548730606	0	-Inf	0
x12	6.811932110963	0	Inf	0
x13	-1.82494674357546	0	-Inf	0
x14	0.20933225479701	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	9.18626714756065	300	0.0306208904918688		

Model	9.18626714756065	14	0.656161939111475	Inf	0
Residual	0	286	0		

A = 0.000000, B = 0.100000, C = 0.002536

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.202536, 0.305072, 0.407608, 0.510144, 0.612680, 0.715216]

List of powers for D(x): [0.100000, 0.202536, 0.305072, 0.407608, 0.510144, 0.612680, 0.715216]

Fitting J3(x) in range (0.000000, 30.000000)

MSS of errors squared = 4.797372e-02

Corrected MSS of errors squared = 8.309293e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i-C*(i-1)

Fitting J3(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B*i-C*(i-1)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.15713249869236e-08	0	-Inf	0
x1	-4.70330297513831e-08	0	-Inf	0
x2	2.25411543448997e-07	0	Inf	0
x3	-3.76951064407266e-07	0	-Inf	0
x4	2.31570501221898e-07	0	Inf	0
x5	0.000714561360749354	0	Inf	0
x6	-0.00713156021681442	0	-Inf	0
x7	0.0316425545488547	0	Inf	0
x8	-0.0846807324430776	0	-Inf	0
x9	0.175289635405377	0	Inf	0
x10	0.900765662342957	0	Inf	0
x11	-0.0166001989003393	0	-Inf	0

Number of observations: 300, Error degrees of freedom: 288

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	9.18529698161764	299	0.0307200567947078		

Model	9.18529698161764	11	0.835026998328877	Inf	0
Residual	0	288	0		

A = 0.586223, B = 0.101834, C = 0.242040

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.688057, 0.547852, 0.407647, 0.267441]

List of powers for D(x): [0.688057, 0.547852, 0.407647, 0.267441, 0.127236, -0.012969, -0.153174]

Fitting J3(x) in range (0.000000, 30.000000)

MSS of errors squared = NaN

Corrected MSS of errors squared = NaN

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*sqrt(i-1)

Fitting J3(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is A+B*i+C*sqrt(i-1)

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00431027045248e-13	0	Inf	0
x1	0.000695834811079659	0	Inf	0
x2	-0.00393754735241566	0	-Inf	0
x3	0.00906590974203387	0	Inf	0
x4	-0.0109473603192749	0	-Inf	0
x5	0.0073405457977715	0	Inf	0
x6	-0.00259601364023097	0	-Inf	0
x7	0.000378758014911619	0	Inf	0
x8	5.52868515717009	0	Inf	0
x9	-13.5493121349547	0	-Inf	0
x10	18.5117787891378	0	Inf	0
x11	-15.0908601036211	0	-Inf	0
x12	7.3588871006995	0	Inf	0
x13	-1.98921426386303	0	-Inf	0
x14	0.23002866633932	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	9.18626714756082	300	0.0306208904918694		
Model	9.18626714756082	14	0.656161939111487	Inf	0
Residual	0	286	0		

A = 0.000000, B = 0.100000, C = 0.002658

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.202658, 0.303759, 0.404604, 0.505316, 0.605944, 0.706511]

List of powers for D(x): [0.100000, 0.202658, 0.303759, 0.404604, 0.505316, 0.605944, 0.706511]

Fitting J3(x) in range (0.000000, 30.000000)

MSS of errors squared = 1.418638e-01

Corrected MSS of errors squared = 2.457152e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}+C*(i-1)$

Fitting J3(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.0974597171372e-12	0	-Inf	0
x1	4.50898626594081e-08	0	Inf	0
x2	-8.77864369010372e-08	0	-Inf	0
x3	4.33194696900064e-08	0	Inf	0
x4	30.4990797990681	0	Inf	0
x5	-171.538478181491	0	-Inf	0
x6	430.559285411992	0	Inf	0
x7	-582.774403816947	0	-Inf	0
x8	443.776093757623	0	Inf	0
x9	-179.767154204931	0	-Inf	0
x10	30.2455772051891	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	9.18626714754948	300	0.0306208904918316		
Model	9.18626714754948	10	0.918626714754949	Inf	0
Residual	0	290	0		

A = 0.000000, B = 0.100000, C = 0.010666

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.152087, 0.194537]

List of powers for D(x): [0.100000, 0.152087, 0.194537, 0.231998, 0.266271, 0.298280, 0.328572]

Fitting J3(x) in range (0.000000, 30.000000)

MSS of errors squared = 1.580726e-01

Corrected MSS of errors squared = 2.737898e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting J3(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.09892073556909e-12	0	-Inf	0
x1	2.91398333251731e-08	0	Inf	0
x2	-6.03545597509154e-08	0	-Inf	0
x3	3.14655141993133e-08	0	Inf	0
x4	66.2620651800218	0	Inf	0
x5	-480.70537327249	0	-Inf	0
x6	1452.80654294098	0	Inf	0
x7	-2294.00920370597	0	-Inf	0
x8	1999.29432140965	0	Inf	0
x9	-914.832688715781	0	-Inf	0
x10	172.184336152511	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	9.18626714754912	300	0.0306208904918304		
Model	9.18626714754912	10	0.918626714754912	Inf	0

```

Residual          0      290          0

A = 0.000000, B = 0.100000, C = 0.000000
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.141421, 0.173205]
List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949,
0.264575]
Fitting J3(x) in range (0.000000, 30.000000)
MSS of errors squared = 5.338281e-02
Corrected MSS of errors squared = 9.246174e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\log_{10}(i)^4$

```

Fitting J3(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*log10(i)^4

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-8.55832891200296e-17	0	-Inf	0
x1	0	0	NaN	NaN
x2	6.3696023857332e-15	0	Inf	0
x3	-6.20919395239003e-15	0	-Inf	0
x4	0.999999975232782	0	Inf	0
x5	3.13195554867027e-08	0	Inf	0
x6	-7.58816855126443e-09	0	-Inf	0
x7	1.03582696892288e-09	0	Inf	0

Number of observations: 301, Error degrees of freedom: 294

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	9.18626714756074	300	0.0306208904918691		
Model	9.18626714756074	6	1.53104452459346	Inf	0
Residual	0	294	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 3.000000, order D(x) = 4.000000
List of powers for Q(x): [0.000000, 0.000821, 0.005182]

```

```
List of powers for D(x): [0.000000, 0.000821, 0.005182, 0.013139]
Fitting J3(x) in range (0.000000, 30.000000)
MSS of errors squared = 2.329207e-02
Corrected MSS of errors squared = 3.293996e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Comments

The following table shows the summary results for the ten models that fit the Bessel $J_3(x)$ function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\log_{10}(i)^4$	0	0.1		Inf	-Inf	2.32920700E-02	3, 4
$A+B*i+C*(i-1)$	0	0.1	0.002	Inf	-Inf	4.79737200E-02	7, 7
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	5.33828100E-02	3, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0	0.1	0	Inf	-Inf	5.33828100E-02	3, 7
$A+B*i+C*\sqrt{i-1}$	0	0.1	0.002	Inf	-Inf	1.41863800E-01	7, 7
$A+B*\sqrt{i}+C*(i-1)$	0	0.1	0.01	Inf	-Inf	1.58072600E-01	3, 7
$A+B*\sqrt{i+C}$	0	0.1	2	Inf	-Inf	1.71107500E-01	5, 7
$A+B*i$	0.001	0.1		Inf	-Inf	2.17637600E-01	7, 7
$A+B/i$	0.066	0.338		Inf	-Inf	Inf	3, 7
$A+B*i-C*(i-1)$	0.586	0.101	0.242	Inf	-Inf	NaN	4, 7

RESULTS FOR THE BESSEL J4(X)

Using Power $A+B*i$

Fitting J4(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.18417283075703e-10	0	-Inf	0
x1	6.97515402348515e-08	0	Inf	0
x2	-1.10164268024451e-07	0	-Inf	0
x3	4.3554514879354e-08	0	Inf	0
x4	5.70773953488567	0	Inf	0
x5	-13.9419483754961	0	-Inf	0
x6	18.8922997826091	0	Inf	0
x7	-15.3382162828327	0	-Inf	0
x8	7.46099486817145	0	Inf	0
x9	-2.01340066056683	0	-Inf	0
x10	0.232527728741069	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.389774640655	300	0.02796591546885		
Model	8.389774640655	10	0.8389774640655	Inf	0
Residual	0	290	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.200000, 0.300000]
List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
Fitting J4(x) in range (0.000000, 30.000000)
MSS of errors squared = 3.025319e-01
Corrected MSS of errors squared = 4.278447e-01
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power A+B/i

```

Fitting J4(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B/i

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.7556315412939e-12	0	-Inf	0
x1	6.82421265393356e-07	0	Inf	0
x2	-8.79057198029283e-06	0	-Inf	0
x3	2.02206374173058e-05	0	Inf	0
x4	-1.21119425724022e-05	0	-Inf	0
x5	-8.86615391140583	0	-Inf	0
x6	1029.24659177579	0	Inf	0
x7	-11494.3620039144	0	-Inf	0
x8	35256.1732430132	0	Inf	0
x9	-34108.8493083634	0	-Inf	0
x10	0	0	NaN	NaN
x11	9327.65763111168	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.38977464078134	300	0.0279659154692711		
Model	8.38977464078134	10	0.838977464078134	Inf	0
Residual	0	290	0		

```

A = 0.090090, B = 0.100000
order Q(x) = 4.000000, order D(x) = 7.000000
List of powers for Q(x): [0.190090, 0.140090, 0.123423, 0.115090]
List of powers for D(x): [0.190090, 0.140090, 0.123423, 0.115090, 0.110090, 0.106757,
0.104376]
Fitting J4(x) in range (0.000000, 30.000000)
MSS of errors squared = Inf
Corrected MSS of errors squared = Inf
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}$

```

Fitting J4(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-6.67154760913807e-11	0	-Inf	0
x1	5.02737068451741e-08	0	Inf	0
x2	-1.02802060933903e-07	0	-Inf	0
x3	5.30949094067685e-08	0	Inf	0
x4	112.638331024621	0	Inf	0
x5	-838.908164573384	0	-Inf	0
x6	2542.3242961491	0	Inf	0
x7	-3998.83190757983	0	-Inf	0
x8	3462.61990891223	0	Inf	0
x9	-1572.38415434953	0	-Inf	0
x10	293.541689820365	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.38977464061351	300	0.0279659154687117		
Model	8.38977464061351	10	0.838977464061351	Inf	0
Residual	0	290	0		

```

A = 0.029206, B = 0.125342
order Q(x) = 3.000000, order D(x) = 7.000000

```

```
List of powers for Q(x): [0.154548, 0.206466, 0.246304]
List of powers for D(x): [0.154548, 0.206466, 0.246304, 0.279889, 0.309479, 0.336229,
0.360829]
Fitting J4(x) in range (0.000000, 30.000000)
MSS of errors squared = 3.975816e-02
Corrected MSS of errors squared = 5.622653e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Using Power $A+B*\sqrt{i+C}$

```
Fitting J4(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is  $A+B*\sqrt{i+C}$ 
```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.00447363920038e-11	0	-Inf	0
x1	-3.7263520155494e-06	0	-Inf	0
x2	1.38137606614031e-05	0	Inf	0
x3	-1.63239997788792e-05	0	-Inf	0
x4	6.23913867544435e-06	0	Inf	0
x5	707.567873766394	0	Inf	0
x6	-5734.99632282659	0	-Inf	0
x7	18193.4840314558	0	Inf	0
x8	-29545.4993585566	0	-Inf	0
x9	26228.3486909903	0	Inf	0
x10	-12158.2499288339	0	-Inf	0
x11	2310.34501243233	0	Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.38977464092802	300	0.0279659154697601		
Model	8.38977464092802	11	0.762706785538911	Inf	0
Residual	0	289	0		

A = 0.092548, B = 0.100000, C = 0.165698

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.200516, 0.239711, 0.270472, 0.296649]

List of powers for D(x): [0.200516, 0.239711, 0.270472, 0.296649, 0.319830, 0.340857, 0.360237]

Fitting J4(x) in range (0.000000, 30.000000)

MSS of errors squared = 1.896505e-01

Corrected MSS of errors squared = 3.284843e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting J4(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.18417283075703e-10	0	-Inf	0
x1	6.97515402348515e-08	0	Inf	0
x2	-1.10164268024451e-07	0	-Inf	0
x3	4.3554514879354e-08	0	Inf	0
x4	5.70773953488567	0	Inf	0
x5	-13.9419483754961	0	-Inf	0
x6	18.8922997826091	0	Inf	0
x7	-15.3382162828327	0	-Inf	0
x8	7.46099486817145	0	Inf	0
x9	-2.01340066056683	0	-Inf	0
x10	0.232527728741069	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.389774640655	300	0.02796591546885		
Model	8.389774640655	10	0.8389774640655	Inf	0
Residual	0	290	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

Fitting J4(x) in range (0.000000, 30.000000)
 MSS of errors squared = 3.025319e-01
 Corrected MSS of errors squared = 5.240006e-01
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting J4(x) in range (0.000000, 30.000000)
 Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	3.52642642038e-14	0	Inf	0
x1	-0.020364014538238	0	-Inf	0
x2	0.123065382778226	0	Inf	0
x3	-0.309757858727234	0	-Inf	0
x4	0.415654316457038	0	Inf	0
x5	-0.313609601473785	0	-Inf	0
x6	0.126144865307587	0	Inf	0
x7	-0.021133089683167	0	-Inf	0
x8	-6.39226944276158	0	-Inf	0
x9	34.1175991028068	0	Inf	0
x10	-68.7739398315677	0	-Inf	0
x11	56.9413101981822	0	Inf	0
x12	0	0	NaN	NaN
x13	-30.1449840439306	0	-Inf	0
x14	15.2522839696052	0	Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.38977464065438	300	0.0279659154688479		
Model	8.38977464065438	13	0.645367280050337	Inf	0
Residual	0	287	0		

A = 0.000000, B = 0.149116, C = 0.167340
 order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for $Q(x)$: [0.149116, 0.130892, 0.112668, 0.094444, 0.076220, 0.057996, 0.039773]

List of powers for $D(x)$: [0.149116, 0.130892, 0.112668, 0.094444, 0.076220, 0.057996, 0.039773]

Fitting $J4(x)$ in range (0.000000, 30.000000)

MSS of errors squared = 1.431851e-01

Corrected MSS of errors squared = 2.480039e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting $J4(x)$ in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.18417283075703e-10	0	-Inf	0
x1	6.97515402348515e-08	0	Inf	0
x2	-1.10164268024451e-07	0	-Inf	0
x3	4.3554514879354e-08	0	Inf	0
x4	5.70773953488567	0	Inf	0
x5	-13.9419483754961	0	-Inf	0
x6	18.8922997826091	0	Inf	0
x7	-15.3382162828327	0	-Inf	0
x8	7.46099486817145	0	Inf	0
x9	-2.01340066056683	0	-Inf	0
x10	0.232527728741069	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.389774640655	300	0.02796591546885		
Model	8.389774640655	10	0.8389774640655	Inf	0
Residual	0	290	0		

A = 0.000000, B = 0.100000, C = 0.000000

order $Q(x)$ = 3.000000, order $D(x)$ = 7.000000

List of powers for $Q(x)$: [0.100000, 0.200000, 0.300000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

Fitting J4(x) in range (0.000000, 30.000000)

MSS of errors squared = 3.025319e-01

Corrected MSS of errors squared = 5.240006e-01

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting J4(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	8.94917773438693e-12	0	Inf	0
x1	0.0615053990326698	0	Inf	0
x2	-0.652151965738136	0	-Inf	0
x3	2.47341852727948	0	Inf	0
x4	-4.6004830943785	0	-Inf	0
x5	4.55998106149	0	Inf	0
x6	-2.32045826960268	0	-Inf	0
x7	0.47818835082762	0	Inf	0
x8	731.21812534758	0	Inf	0
x9	-6178.52816631118	0	-Inf	0
x10	20075.6579779845	0	Inf	0
x11	-33117.4407818038	0	-Inf	0
x12	29728.0874706793	0	Inf	0
x13	-13895.4393604336	0	-Inf	0
x14	2657.44473064056	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.38977464062877	300	0.0279659154687626		
Model	8.38977464062877	14	0.599269617187769	Inf	0
Residual	0	286	0		

A = 0.118904, B = 0.104654, C = 0.000000

```

order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.223558, 0.266907, 0.300170, 0.328212, 0.352917, 0.375253,
0.395792]
List of powers for D(x): [0.223558, 0.266907, 0.300170, 0.328212, 0.352917, 0.375253,
0.395792]
Fitting J4(x) in range (0.000000, 30.000000)
MSS of errors squared = 3.034207e-02
Corrected MSS of errors squared = 5.255401e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

```

Fitting J4(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-9.18201727640616e-13	0	-Inf	0
x1	-1.5600265522576e-06	0	-Inf	0
x2	8.52514747483833e-06	0	Inf	0
x3	-1.61403837134019e-05	0	-Inf	0
x4	1.29527428120118e-05	0	Inf	0
x5	-3.77734190309836e-06	0	-Inf	0
x6	64.6212322589909	0	Inf	0
x7	-464.046800247112	0	-Inf	0
x8	1391.63746035815	0	Inf	0
x9	-2183.22826781393	0	-Inf	0
x10	1891.9859459908	0	Inf	0
x11	-861.330218512147	0	-Inf	0
x12	161.360647884541	0	Inf	0

Number of observations: 301, Error degrees of freedom: 288

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.38977464063113	300	0.0279659154687704		
Model	8.38977464063113	12	0.699147886719261	Inf	0
Residual	0	288	0		

```

A = 0.000000, B = 0.100000, C = 0.000000
order Q(x) = 5.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607]
List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949,
0.264575]
Fitting J4(x) in range (0.000000, 30.000000)
MSS of errors squared = 7.049163e-02
Corrected MSS of errors squared = 1.220951e-01
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\log_{10}(i)^4$

```

Fitting J4(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is  $A+B*\log_{10}(i)^4$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.50058561819535e-16	0	Inf	0
x1	0	0	NaN	NaN
x2	-6.25662124080979e-08	0	-Inf	0
x3	1.48175725854924e-07	0	Inf	0
x4	-1.4654629527914e-07	0	-Inf	0
x5	8.44660776581081e-08	0	Inf	0
x6	-2.74025197510622e-08	0	-Inf	0
x7	3.87322383472096e-09	0	Inf	0
x8	0.999994427400859	0	Inf	0
x9	7.30244480779931e-06	0	Inf	0
x10	-2.18930972615988e-06	0	-Inf	0
x11	5.23583959823262e-07	0	Inf	0
x12	-6.41199032827459e-08	0	-Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	8.38977464065495	300	0.0279659154688498		
Model	8.38977464065495	11	0.762706785514087	Inf	0
Residual	0	289	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 7.000000, order D(x) = 5.000000
List of powers for Q(x): [0.000000, 0.000821, 0.005182, 0.013139, 0.023869, 0.036665,
0.051007]
List of powers for D(x): [0.000000, 0.000821, 0.005182, 0.013139, 0.023869]
Fitting J4(x) in range (0.000000, 30.000000)
MSS of errors squared = 1.218818e-02
Corrected MSS of errors squared = 1.723669e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Comments

The following table shows the summary results for the ten models that fit the Bessel $J_4(x)$ function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\log_{10}(i)^4$	0	0.1		Inf	-Inf	1.21881800E-02	7, 5
$A+B*\sqrt{i}+C*(i-1)$	0.118	0.104	0	Inf	-Inf	3.03420700E-02	7, 7
$A+B*\sqrt{i}$	0.029	0.125		Inf	-Inf	3.97581600E-02	3, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0	0.1	0	Inf	-Inf	7.04916300E-02	5, 7
$A+B*i-C*(i-1)$	0	0.149	0.167	Inf	-Inf	1.43185100E-01	7, 7
$A+B*\sqrt{i+C}$	0.092	0.1	0.165	Inf	-Inf	1.89650500E-01	4, 7
$A+B*i$	0	0.1		Inf	-Inf	3.02531900E-01	3, 7
$A+B*i+C*(i-1)$	0	0.1	0	Inf	-Inf	3.02531900E-01	3, 7
$A+B*i+C*\sqrt{i-1}$	0	0.1	0	Inf	-Inf	3.02531900E-01	3, 7
$A+B/i$	0.09	0.1		Inf	-Inf	Inf	4, 7

RESULTS FOR THE BESSEL J5(X)

Using Power $A+B*i$

Fitting J5(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.6172442540745e-11	0	-Inf	0
x1	-8.14637836827588e-07	0	-Inf	0
x2	2.20300672799594e-06	0	Inf	0
x3	-1.93429354004425e-06	0	-Inf	0
x4	5.54993694823536e-07	0	Inf	0
x5	12.3890709260405	0	Inf	0
x6	-36.9133408862103	0	-Inf	0
x7	53.3822090677525	0	Inf	0
x8	-44.4693328004486	0	-Inf	0
x9	21.8261741101937	0	Inf	0
x10	-5.8922648557002	0	-Inf	0
x11	0.677449225307313	0	Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538914194	300	0.0250602846304731		
Model	7.51808538914194	11	0.683462308103812	Inf	0

```

Residual          0      289          0

A = 0.065000, B = 0.100000
order Q(x) = 4.000000, order D(x) = 7.000000
List of powers for Q(x): [0.165000, 0.265000, 0.365000, 0.465000]
List of powers for D(x): [0.165000, 0.265000, 0.365000, 0.465000, 0.565000, 0.665000,
0.765000]
Fitting J5(x) in range (0.000000, 30.000000)
MSS of errors squared = 5.599545e-02
Corrected MSS of errors squared = 7.918952e-02
R-Squared =      1.00000000
R-Squared Adjusted =    1.00000000
Particle swarm AICc =  -1.000000e+99
AIC =  -Inf
AICc =  -Inf

```

Using Power $A+B/i$

Fitting J5(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-7.7659705378933e-10	0	-Inf	0
x1	1.23195487462083e-09	0	Inf	0
x2	-1.27195915979383e-08	0	-Inf	0
x3	1.33551429445358e-08	0	Inf	0
x4	0.00698350724261756	0	Inf	0
x5	-4.95009077723238	0	-Inf	0
x6	160.045583605004	0	Inf	0
x7	-1218.02148481256	0	-Inf	0
x8	3404.6940630437	0	Inf	0
x9	-3909.7537681462	0	-Inf	0
x10	1568.97870746077	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538912819	300	0.0250602846304273		
Model	7.51808538912819	10	0.751808538912819	Inf	0
Residual	0	290	0		

```

A = 0.113248, B = 0.658658
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.771905, 0.442577, 0.332800]
List of powers for D(x): [0.771905, 0.442577, 0.332800, 0.277912, 0.244979, 0.223024,
0.207342]
Fitting J5(x) in range (0.000000, 30.000000)
MSS of errors squared = 7.625698e-02
Corrected MSS of errors squared = 1.078437e-01
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}$

```

Fitting J5(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.49069500573412e-13	0	Inf	0
x1	2.49018616169899e-05	0	Inf	0
x2	-0.000198022727160473	0	-Inf	0
x3	0.000557923288232532	0	Inf	0
x4	-0.000733982976769954	0	-Inf	0
x5	0.000461785227888016	0	Inf	0
x6	-0.000112604616074125	0	-Inf	0
x7	63.3701718779798	0	Inf	0
x8	-451.438935127698	0	-Inf	0
x9	1345.60147176961	0	Inf	0
x10	-2100.24475708579	0	-Inf	0
x11	1811.93306893256	0	Inf	0
x12	-821.563933453096	0	-Inf	0
x13	153.342912888891	0	Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538918058	300	0.0250602846306019		
Model	7.51808538918058	13	0.578314260706198	Inf	0
Residual	0	287	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 6.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949]
List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949,
0.264575]
Fitting J5(x) in range (0.000000, 30.000000)
MSS of errors squared = 3.825715e-02
Corrected MSS of errors squared = 5.410378e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i+C}$

```

Fitting J5(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i+C}$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.25231762843682e-13	0	Inf	0
x1	2.84290927366553e-05	0	Inf	0
x2	-0.000225875036096159	0	-Inf	0
x3	0.000635998187690148	0	Inf	0
x4	-0.000836276641825181	0	-Inf	0
x5	0.000525921611333161	0	Inf	0
x6	-0.000128197151039333	0	-Inf	0
x7	66.750832260708	0	Inf	0
x8	-478.071378416761	0	-Inf	0
x9	1428.34854336341	0	Inf	0
x10	-2232.51435063783	0	-Inf	0
x11	1927.86943589237	0	Inf	0
x12	-874.738257293767	0	-Inf	0
x13	163.35517461805	0	Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538911648	300	0.0250602846303883		
Model	7.51808538911648	13	0.578314260701267	Inf	0

```

Residual          0      287          0

A = 0.001877, B = 0.100000, C = 0.000000
order Q(x) = 6.000000, order D(x) = 7.000000
List of powers for Q(x): [0.101877, 0.143298, 0.175082, 0.201877, 0.225483, 0.246826]
List of powers for D(x): [0.101877, 0.143298, 0.175082, 0.201877, 0.225483, 0.246826,
0.266452]
Fitting J5(x) in range (0.000000, 30.000000)
MSS of errors squared = Inf
Corrected MSS of errors squared = Inf
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting J5(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.78946261855095e-10	0	-Inf	0
x1	3.01298732223585e-08	0	Inf	0
x2	-4.4602297715871e-08	0	-Inf	0
x3	1.66754686166304e-08	0	Inf	0
x4	5.48006432898285	0	Inf	0
x5	-12.8527207888313	0	-Inf	0
x6	16.7238022088741	0	Inf	0
x7	-13.0386320434623	0	-Inf	0
x8	6.09100754963117	0	Inf	0
x9	-1.57865019562427	0	-Inf	0
x10	0.175114312147288	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538914219	300	0.0250602846304739		
Model	7.51808538914219	10	0.751808538914219	Inf	0
Residual	0	290	0		

```

A = 0.000000, B = 0.110566, C = 0.000000
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.110566, 0.221132, 0.331697]
List of powers for D(x): [0.110566, 0.221132, 0.331697, 0.442263, 0.552829, 0.663395,
0.773960]
Fitting J5(x) in range (0.000000, 30.000000)
MSS of errors squared = 1.469235e-02
Corrected MSS of errors squared = 2.544790e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i-C*(i-1)$

```

Fitting J5(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i-C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.92441465477686e-12	0	-Inf	0
x1	-4.75543604243517e-07	0	-Inf	0
x2	1.26181690556494e-06	0	Inf	0
x3	-1.07072195136675e-06	0	-Inf	0
x4	2.94124210981545e-07	0	Inf	0
x5	5.29156883057369	0	Inf	0
x6	-11.9817906129929	0	-Inf	0
x7	15.0492788203866	0	Inf	0
x8	-11.3237544763992	0	-Inf	0
x9	5.10442774415135	0	Inf	0
x10	-1.27633157857683	0	-Inf	0
x11	0.136564518889263	0	Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538914192	300	0.0250602846304731		
Model	7.51808538914192	11	0.683462308103811	Inf	0
Residual	0	289	0		

A = 0.000000, B = 0.124381, C = 0.000000

```

order Q(x) = 4.000000, order D(x) = 7.000000
List of powers for Q(x): [0.124381, 0.248763, 0.373144, 0.497526]
List of powers for D(x): [0.124381, 0.248763, 0.373144, 0.497526, 0.621907, 0.746288,
0.870670]
Fitting J5(x) in range (0.000000, 30.000000)
MSS of errors squared = 3.234307e-01
Corrected MSS of errors squared = 5.601983e-01
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*\sqrt{i-1}$

```

Fitting J5(x) in range (0.000000, 30.000000)
Pade-Shammas polynomial power is A+B*i+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.1127415441016e-09	0	-Inf	0
x1	1.43919907628225e-07	0	Inf	0
x2	-2.13776352978843e-07	0	-Inf	0
x3	8.04418443236719e-08	0	Inf	0
x4	14.7906295600666	0	Inf	0
x5	-45.4691034405816	0	-Inf	0
x6	66.5139899155777	0	Inf	0
x7	-55.636972328597	0	-Inf	0
x8	27.3215220128726	0	Inf	0
x9	-7.36438963867764	0	-Inf	0
x10	0.844273558792823	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538914298	300	0.0250602846304766		
Model	7.51808538914298	10	0.751808538914298	Inf	0
Residual	0	290	0		

A = 0.085143, B = 0.103339, C = 0.000000

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.188482, 0.291821, 0.395161]

List of powers for D(x): [0.188482, 0.291821, 0.395161, 0.498500, 0.601839, 0.705178, 0.808517]

Fitting J5(x) in range (0.000000, 30.000000)

MSS of errors squared = 1.406293e-02

Corrected MSS of errors squared = 2.435770e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting J5(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	8.76755116283119e-13	0	Inf	0
x1	-6.05396471722256e-08	0	-Inf	0
x2	2.2717198345401e-07	0	Inf	0
x3	-2.70378953275737e-07	0	-Inf	0
x4	1.03781975817447e-07	0	Inf	0
x5	63.6053896481925	0	Inf	0
x6	-453.79752004476	0	-Inf	0
x7	1354.1807521168	0	Inf	0
x8	-2115.66035797966	0	-Inf	0
x9	1826.7628432047	0	Inf	0
x10	-828.912199810541	0	-Inf	0
x11	154.821092702359	0	Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.5180853891558	300	0.0250602846305193		
Model	7.5180853891558	11	0.683462308105073	Inf	0
Residual	0	289	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting J5(x) in range (0.000000, 30.000000)

MSS of errors squared = 2.802466e-02

Corrected MSS of errors squared = 4.854013e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting J5(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.49069500573412e-13	0	Inf	0
x1	2.49018616169899e-05	0	Inf	0
x2	-0.000198022727160473	0	-Inf	0
x3	0.000557923288232532	0	Inf	0
x4	-0.000733982976769954	0	-Inf	0
x5	0.000461785227888016	0	Inf	0
x6	-0.000112604616074125	0	-Inf	0
x7	63.3701718779798	0	Inf	0
x8	-451.438935127698	0	-Inf	0
x9	1345.60147176961	0	Inf	0
x10	-2100.24475708579	0	-Inf	0
x11	1811.93306893256	0	Inf	0
x12	-821.563933453096	0	-Inf	0
x13	153.342912888891	0	Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538918058	300	0.0250602846306019		
Model	7.51808538918058	13	0.578314260706198	Inf	0
Residual	0	287	0		

A = 0.000000, B = 0.100000, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949]
 List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting J5(x) in range (0.000000, 30.000000)

MSS of errors squared = 3.825715e-02

Corrected MSS of errors squared = 6.626332e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting J5(x) in range (0.000000, 30.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	4.08082652468774e-16	0	Inf	0
x1	0	0	NaN	NaN
x2	-2.99111906829586e-07	0	-Inf	0
x3	7.08795707257603e-07	0	Inf	0
x4	-7.01689262132809e-07	0	-Inf	0
x5	4.0491590236772e-07	0	Inf	0
x6	-1.31520111841885e-07	0	-Inf	0
x7	1.86096707027938e-08	0	Inf	0
x8	0.999999988124992	0	Inf	0
x9	1.50207392258925e-08	0	Inf	0
x10	-3.64457913296705e-09	0	-Inf	0
x11	4.98846162999823e-10	0	Inf	0

Number of observations: 301, Error degrees of freedom: 290

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.51808538914207	300	0.0250602846304736		
Model	7.51808538914207	10	0.751808538914207	Inf	0
Residual	0	290	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 4.000000

```
List of powers for Q(x): [0.000000, 0.000821, 0.005182, 0.013139, 0.023869, 0.036665,
0.051007]
List of powers for D(x): [0.000000, 0.000821, 0.005182, 0.013139]
Fitting J5(x) in range (0.000000, 30.000000)
MSS of errors squared = 3.411671e-02
Corrected MSS of errors squared = 4.824832e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Comments

The following table shows the summary results for the ten models that fit the Bessel $J_5(x)$ function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*i+C*\sqrt{i-1}$	0.085	0.103	0	Inf	-Inf	1.40629300E-02	3, 7
$A+B*i+C*(i-1)$	0	0.11	0	Inf	-Inf	1.46923500E-02	3, 7
$A+B*\sqrt{i}+C*(i-1)$	0	0.1	0	Inf	-Inf	2.80246600E-02	4, 7
$A+B*\log_{10}(i)^4$	0	0.1		Inf	-Inf	3.41167100E-02	7, 4
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	3.82571500E-02	6, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0	0.1	0	Inf	-Inf	3.82571500E-02	6, 7
$A+B*i$	0.065	0.1		Inf	-Inf	5.59954500E-02	4, 7
$A+B/i$	0.113	0.658		Inf	-Inf	7.62569800E-02	3, 7
$A+B*i-C*(i-1)$	0	0.124	0	Inf	-Inf	3.23430700E-01	4, 7
$A+B*\sqrt{i+C}$	0.001	0.1	0	Inf	-Inf	Inf	6, 7

RESULTS FOR THE NATURAL LOGRITHM

Using Power $A+B*i$

Fitting $\ln(x)$ in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-8.59224786729495	0	-Inf	0
x1	0	0	NaN	NaN
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	0	0	NaN	NaN
x5	-2240.76844324728	0	-Inf	0
x6	2432.50090935225	0	Inf	0
x7	-183.140218199546	0	-Inf	0
x8	-104.605917124061	0	-Inf	0
x9	364.265950363828	0	Inf	0
x10	-594.530069535926	0	-Inf	0
x11	568.300443796264	0	Inf	0
x12	-447.463872890853	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 892

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	312.39398039607	900	0.3471044226623		
Model	312.39398039607	8	39.0492475495087	Inf	0
Residual	0	892	0		

A = 0.478954, B = 0.100000

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.578954, 0.678954, 0.778954, 0.878954, 0.978954, 1.078954, 1.178954]

List of powers for D(x): [0.578954, 0.678954, 0.778954, 0.878954, 0.978954]

Fitting ln(x) in range (1.000000, 10.000000)

MSS of errors squared = 4.527610e-13

Corrected MSS of errors squared = 6.403008e-13

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting ln(x) in range (1.000000, 10.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-17.893479692792	0	-Inf	0
x1	17.8934797233262	0	Inf	0
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	-36.1654602095231	0	-Inf	0
x5	1395.57758445162	0	Inf	0
x6	-5170.67774362836	0	-Inf	0
x7	0	0	NaN	NaN
x8	9123.98254153609	0	Inf	0
x9	0	0	NaN	NaN
x10	-5318.41151610651	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 894

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	312.393980394103	900	0.347104422660114		
Model	312.393980394103	6	52.0656633990171	Inf	0

Total	312.393980392944	900	0.347104422658826		
Model	312.393980392944	7	44.6277114847063	Inf	0
Residual	0	893	0		

A = 0.152284, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.252284, 0.293706, 0.325490, 0.352284, 0.375891, 0.397233, 0.416860]

List of powers for D(x): [0.252284, 0.293706, 0.325490, 0.352284, 0.375891, 0.397233, 0.416860]

Fitting ln(x) in range (1.000000, 10.000000)

MSS of errors squared = 4.856034e-13

Corrected MSS of errors squared = 6.867469e-13

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting ln(x) in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-8.03486778104742	0	-Inf	0
x1	-79635.8836760197	0	-Inf	0
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	69990.8308755784	0	Inf	0
x5	9653.08766856994	0	Inf	0
x6	-2279.97048510983	0	-Inf	0
x7	-7664.23270000454	0	-Inf	0
x8	-4257.98218302653	0	-Inf	0
x9	-3045.28581116142	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 893

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	312.393980452257	900	0.34710442272473		
Model	312.393980452257	7	44.6277114931795	Inf	0

```

Residual          0      893          0

A = 0.214552, B = 0.274864, C = 1.510999
order Q(x) = 5.000000, order D(x) = 4.000000
List of powers for Q(x): [0.650105, 0.729583, 0.798339, 0.859809, 0.915913]
List of powers for D(x): [0.650105, 0.729583, 0.798339, 0.859809]
Fitting ln(x) in range (1.000000, 10.000000)
MSS of errors squared = 5.739412e-14
Corrected MSS of errors squared = 9.940954e-14
R-Squared =      1.00000000
R-Squared Adjusted =      1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

Fitting $\ln(x)$ in range (1.000000, 10.000000)
Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-13.3437565394552	0	-Inf	0
x1	-15.316808310446	0	-Inf	0
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	28.6605648517079	0	Inf	0
x5	-6.67741591734955	0	-Inf	0
x6	-7.3906512384542	0	-Inf	0
x7	-9.45252420429962	0	-Inf	0
x8	-3.39130254257255	0	-Inf	0
x9	0.202707001024465	0	Inf	0
x10	-0.0212418544025209	0	-Inf	0
x11	0.00120555659006209	0	Inf	0

Number of observations: 901, Error degrees of freedom: 891

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	312.393980393313	900	0.347104422659236		
Model	312.393980393313	9	34.7104422659236	Inf	0
Residual	0	891	0		

```

A = 0.010022, B = 0.248431, C = 0.033960
order Q(x) = 4.000000, order D(x) = 7.000000
List of powers for Q(x): [0.258453, 0.540845, 0.823236, 1.105627]
List of powers for D(x): [0.258453, 0.540845, 0.823236, 1.105627, 1.388019, 1.670410, 1.952801]
Fitting ln(x) in range (1.000000, 10.000000)
MSS of errors squared = 1.846120e-13
Corrected MSS of errors squared = 3.197573e-13
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i-C*(i-1)$

```

Fitting ln(x) in range (1.000000, 10.000000)
Pade-Shammas polynomial power is A+B*i-C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-33.2310001499109	0	-Inf	0
x1	0	0	NaN	NaN
x2	0	0	NaN	NaN
x3	-1962.74288205097	0	-Inf	0
x4	0	0	NaN	NaN
x5	0	0	NaN	NaN
x6	0	0	NaN	NaN
x7	1995.97388222427	0	Inf	0
x8	-335.390947295864	0	-Inf	0
x9	738.024797225112	0	Inf	0
x10	-456.750589554951	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 895

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	312.393980394796	900	0.347104422660885		
Model	312.393980394796	5	62.4787960789593	Inf	0
Residual	0	895	0		

```

A = 0.000000, B = 0.133726, C = 0.127432
order Q(x) = 7.000000, order D(x) = 3.000000

```

List of powers for Q(x): [0.133726, 0.140021, 0.146315, 0.152610, 0.158904, 0.165199, 0.171493]

List of powers for D(x): [0.133726, 0.140021, 0.146315]

Fitting ln(x) in range (1.000000, 10.000000)

MSS of errors squared = 1.911439e-12

Corrected MSS of errors squared = 3.310709e-12

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting ln(x) in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-5.93838297481804	0	-Inf	0
x1	-15.7785682273234	0	-Inf	0
x2	6.9300941274154	0	Inf	0
x3	0	0	NaN	NaN
x4	0	0	NaN	NaN
x5	14.3178986366884	0	Inf	0
x6	0.480317221688685	0	Inf	0
x7	-0.0113586937169705	0	-Inf	0
x8	-9.30658885669531	0	-Inf	0
x9	-12.8084624706717	0	-Inf	0
x10	-3.96329092707304	0	-Inf	0
x11	-8.51285579266257	0	-Inf	0
x12	-2.4702015876063	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 890

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	312.393980393368	900	0.347104422659298		
Model	312.393980393368	10	31.2393980393368	Inf	0
Residual	0	890	0		

A = 0.265269, B = 0.352323, C = 0.299138

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.617592, 1.269052, 1.745282, 2.192682, 2.625158, 3.048098, 3.464263]

List of powers for D(x): [0.617592, 1.269052, 1.745282, 2.192682, 2.625158]

Fitting ln(x) in range (1.000000, 10.000000)

MSS of errors squared = 5.252672e-12

Corrected MSS of errors squared = 9.097895e-12

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting ln(x) in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-6.8305772097232	0	-Inf	0
x1	-265.27724719136	0	-Inf	0
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	0	0	NaN	NaN
x5	0	0	NaN	NaN
x6	241.832539573797	0	Inf	0
x7	30.2752848445648	0	Inf	0
x8	-40.1985949542806	0	-Inf	0
x9	-91.4900392951185	0	-Inf	0
x10	-84.8060075271792	0	-Inf	0
x11	-71.9231542894163	0	-Inf	0
x12	-93.3169291991063	0	-Inf	0
x13	-39.4295458654385	0	-Inf	0
x14	-1.98607640576285	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 890

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	312.39398039312	900	0.347104422659022		
Model	312.39398039312	10	31.239398039312	Inf	0
Residual	0	890	0		

```

A = 0.538391, B = 0.104487, C = 0.271731
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.642878, 0.957888, 1.262829, 1.562557, 1.858953, 2.152984,
2.445221]
List of powers for D(x): [0.642878, 0.957888, 1.262829, 1.562557, 1.858953, 2.152984,
2.445221]
Fitting ln(x) in range (1.000000, 10.000000)
MSS of errors squared = 9.158967e-14
Corrected MSS of errors squared = 1.586380e-13
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

```

Fitting ln(x) in range (1.000000, 10.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-8.14867838692351	0	-Inf	0
x1	-6.79622993055562	0	-Inf	0
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	0	0	NaN	NaN
x5	45.1012490865158	0	Inf	0
x6	-30.1563407355358	0	-Inf	0
x7	-6.70245626602969	0	-Inf	0
x8	-10.3288930287248	0	-Inf	0
x9	1.94442180000058	0	Inf	0
x10	-9.44610757268834	0	-Inf	0
x11	5.3135486850873	0	Inf	0
x12	1.71741785515815	0	Inf	0

Number of observations: 901, Error degrees of freedom: 891

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	312.393980392892	900	0.347104422658769		
Model	312.393980392892	9	34.7104422658769	Inf	0
Residual	0	891	0		

```

A = 0.073233, B = 0.356632, C = 0.470183
order Q(x) = 6.000000, order D(x) = 6.000000
List of powers for Q(x): [0.429865, 1.047770, 1.355878, 1.600879, 1.811054, 1.998162]
List of powers for D(x): [0.429865, 1.047770, 1.355878, 1.600879, 1.811054, 1.998162]
Fitting ln(x) in range (1.000000, 10.000000)
MSS of errors squared = 2.414819e-10
Corrected MSS of errors squared = 4.182590e-10
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\log_{10}(i)^4$

```

Fitting ln(x) in range (1.000000, 10.000000)
Pade-Shammas polynomial power is A+B*log10(i)^4

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	0	0	NaN	NaN
x2	0	0	NaN	NaN
x3	1.0838277808944e-15	0	Inf	0
x4	1.00001468075528	0	Inf	0
x5	-1.92541611899552e-05	0	-Inf	0
x6	5.79837424700339e-06	0	Inf	0
x7	-1.39807960696438e-06	0	-Inf	0
x8	1.73111273162223e-07	0	Inf	0

Number of observations: 901, Error degrees of freedom: 895

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	312.393980393285	900	0.347104422659206		
Model	312.393980393285	5	62.4787960786571	Inf	0
Residual	0	895	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 3.000000, order D(x) = 5.000000
List of powers for Q(x): [0.000000, 0.000821, 0.005182]
List of powers for D(x): [0.000000, 0.000821, 0.005182, 0.013139, 0.023869]

```

```
Fitting ln(x) in range (1.000000, 10.000000)
MSS of errors squared = 6.914598e-02
Corrected MSS of errors squared = 9.778718e-02
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Comments

The following table shows the summary results for the ten models that fit the natural logarithm function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\sqrt{i+C}$	0.214	0.274	1.51	Inf	-Inf	5.73941200E-14	5, 4
$A+B*\sqrt{i}+C*(i-1)$	0.538	0.104	0.271	Inf	-Inf	9.15896700E-14	7, 7
$A+B*i+C*(i-1)$	0.01	0.248	0.033	Inf	-Inf	1.84612000E-13	4, 7
$A+B*i$	0.478	0.1		Inf	-Inf	4.52761000E-13	7, 5
$A+B*\sqrt{i}$	0.152	0.1		Inf	-Inf	4.85603400E-13	7, 7
$A+B*i-C*(i-1)$	0	0.133	0.127	Inf	-Inf	1.91143900E-12	7, 3
$A+B*i+C*\sqrt{i-1}$	0.265	0.352	0.299	Inf	-Inf	5.25267200E-12	7, 5
$A+B/i$	0.274	0.1		Inf	-Inf	1.87134300E-11	3, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.073	0.356	0.47	Inf	-Inf	2.41481900E-10	6, 6
$A+B*\log_{10}(i)^4$	0	0.1		Inf	-Inf	6.91459800E-02	3, 5

RESULTS FOR THE COMMON LOGARITHM

Using Power $A+B*i$

Fitting $\log(x)$ in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.40789134198338	0	-Inf	0
x1	0	0	NaN	NaN
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	4852.31126829642	0	Inf	0
x5	-4848.90337692329	0	-Inf	0
x6	-73.5152981573804	0	-Inf	0
x7	261.700213711598	0	Inf	0
x8	-506.40157477477	0	-Inf	0
x9	1166.94387117995	0	Inf	0
x10	536.92174900924	0	Inf	0
x11	-14.2488353936858	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 892

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587779403	900	0.0654679541977114		
Model	58.9211587779403	8	7.36514484724254	Inf	0

Residual 0 892 0

A = 0.492062, B = 0.123433

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.615495, 0.738929, 0.862362, 0.985796, 1.109229]

List of powers for D(x): [0.615495, 0.738929, 0.862362, 0.985796, 1.109229, 1.232663]

Fitting log(x) in range (1.000000, 10.000000)

MSS of errors squared = 5.737639e-14

Corrected MSS of errors squared = 8.114246e-14

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting log(x) in range (1.000000, 10.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.49921879165346	0	-Inf	0
x1	3.45181735281987	0	Inf	0
x2	2371.69215657384	0	Inf	0
x3	0	0	NaN	NaN
x4	-4688.22592077077	0	-Inf	0
x5	0	0	NaN	NaN
x6	2316.58116566388	0	Inf	0
x7	-0.634761879906234	0	-Inf	0
x8	-297.304136728964	0	-Inf	0
x9	-1304.62658733281	0	-Inf	0
x10	703.187876364102	0	Inf	0

Number of observations: 901, Error degrees of freedom: 892

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587791879	900	0.0654679541990976		
Model	58.9211587791879	8	7.36514484739848	Inf	0
Residual	0	892	0		

A = 0.453947, B = 0.965543

```

order Q(x) = 6.000000, order D(x) = 4.000000
List of powers for Q(x): [1.419490, 0.936719, 0.775795, 0.695333, 0.647056, 0.614871]
List of powers for D(x): [1.419490, 0.936719, 0.775795, 0.695333]
Fitting log(x) in range (1.000000, 10.000000)
MSS of errors squared = 8.254884e-14
Corrected MSS of errors squared = 1.167417e-13
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}$

```

Fitting log(x) in range (1.000000, 10.000000)
Pade-Shammas polynomial power is  $A+B*\sqrt{i}$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-10.0373461209231	0	-Inf	0
x1	-12459.3781220954	0	-Inf	0
x2	-3748.54913519234	0	-Inf	0
x3	16217.9646034118	0	Inf	0
x4	-441.600977622148	0	-Inf	0
x5	-1559.29076341226	0	-Inf	0
x6	-379.207614579982	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 894

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587780089	900	0.0654679541977877		
Model	58.9211587780089	6	9.82019312966815	Inf	0
Residual	0	894	0		

A = 0.112296, B = 0.100000

```

order Q(x) = 3.000000, order D(x) = 3.000000
List of powers for Q(x): [0.212296, 0.253718, 0.285501]
List of powers for D(x): [0.212296, 0.253718, 0.285501]
Fitting log(x) in range (1.000000, 10.000000)
MSS of errors squared = 5.214265e-15
Corrected MSS of errors squared = 7.374084e-15
R-Squared = 1.000000000

```

R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting $\log(x)$ in range (1.000000, 10.000000)
 Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-9.9897621855948	0	-Inf	0
x1	0	0	NaN	NaN
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	0	0	NaN	NaN
x5	9.98976218579049	0	Inf	0
x6	-117.267506495935	0	-Inf	0
x7	574.622609359972	0	Inf	0
x8	-1206.38797377501	0	-Inf	0
x9	1273.73541861871	0	Inf	0
x10	-673.517618324661	0	-Inf	0
x11	141.937419816223	0	Inf	0

Number of observations: 901, Error degrees of freedom: 893

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587783804	900	0.0654679541982004		
Model	58.9211587783804	7	8.41730839691148	Inf	0
Residual	0	893	0		

A = 0.117678, B = 0.100000, C = 0.053282

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.220307, 0.260970, 0.292414, 0.319005, 0.342473]

List of powers for D(x): [0.220307, 0.260970, 0.292414, 0.319005, 0.342473, 0.363712]

Fitting $\log(x)$ in range (1.000000, 10.000000)

MSS of errors squared = 9.817733e-14

Corrected MSS of errors squared = 1.700481e-13

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting $\log(x)$ in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.27329515932367	0	-Inf	0
x1	-23.5520012854875	0	-Inf	0
x2	29.5063792342807	0	Inf	0
x3	0	0	NaN	NaN
x4	0	0	NaN	NaN
x5	-0.182206183777942	0	-Inf	0
x6	-1.48014910390745	0	-Inf	0
x7	-0.0187275011238678	0	-Inf	0
x8	-13.4532856346755	0	-Inf	0
x9	-12.3202861895383	0	-Inf	0
x10	4.28476695019386	0	Inf	0
x11	0	0	NaN	NaN
x12	1.82677386856542	0	Inf	0
x13	0.374439212365932	0	Inf	0

Number of observations: 901, Error degrees of freedom: 890

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587782005	900	0.0654679541980005		
Model	58.9211587782005	10	5.89211587782005	Inf	0
Residual	0	890	0		

A = 0.182397, B = 0.218536, C = 0.115237

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.400933, 0.734705, 1.068478, 1.402250, 1.736023, 2.069795, 2.403567]

List of powers for D(x): [0.400933, 0.734705, 1.068478, 1.402250, 1.736023, 2.069795]

Fitting $\log(x)$ in range (1.000000, 10.000000)

MSS of errors squared = 9.984474e-13

Corrected MSS of errors squared = 1.729362e-12

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting $\log(x)$ in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.58156442224547	0	-Inf	0
x1	0	0	NaN	NaN
x2	18552.5756618341	0	Inf	0
x3	0	0	NaN	NaN
x4	0	0	NaN	NaN
x5	-18548.9940973911	0	-Inf	0
x6	-219.410321902444	0	-Inf	0
x7	2317.04229820069	0	Inf	0
x8	3940.10448785256	0	Inf	0
x9	5192.75351967882	0	Inf	0
x10	1621.25068197916	0	Inf	0
x11	-43.5731035523581	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 892

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587840333	900	0.0654679542044815		
Model	58.9211587840333	8	7.36514484800416	Inf	0
Residual	0	892	0		

A = 0.530415, B = 0.100000, C = 0.000000

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.630415, 0.730415, 0.830415, 0.930415, 1.030415]

List of powers for D(x): [0.630415, 0.730415, 0.830415, 0.930415, 1.030415, 1.130415]

Fitting $\log(x)$ in range (1.000000, 10.000000)

MSS of errors squared = 4.294145e-15

Corrected MSS of errors squared = 7.437678e-15

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting $\log(x)$ in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.66030107777155	0	-Inf	0
x1	-2.49477921858088	0	-Inf	0
x2	-18.39478243502	0	-Inf	0
x3	13.6466703258212	0	Inf	0
x4	11.1271902822184	0	Inf	0
x5	0.776002123705074	0	Inf	0
x6	-6.54394141037586	0	-Inf	0
x7	-17.5991614163369	0	-Inf	0
x8	-18.4169522124534	0	-Inf	0
x9	-4.07436788545862	0	-Inf	0
x10	-0.132404765838152	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 890

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587781619	900	0.0654679541979576		
Model	58.9211587781619	10	5.89211587781619	Inf	0
Residual	0	890	0		

A = 0.052728, B = 0.278588, C = 0.301774

order Q(x) = 5.000000, order D(x) = 5.000000

List of powers for Q(x): [0.331316, 0.911678, 1.315264, 1.689767, 2.049216]

List of powers for D(x): [0.331316, 0.911678, 1.315264, 1.689767, 2.049216]

Fitting $\log(x)$ in range (1.000000, 10.000000)

MSS of errors squared = 1.887376e-14

Corrected MSS of errors squared = 3.269031e-14

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting $\log(x)$ in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.91945781420955	0	-Inf	0
x1	-29.5630676572493	0	-Inf	0
x2	-4.86017180317009	0	-Inf	0
x3	40.9709895797382	0	Inf	0
x4	0	0	NaN	NaN
x5	-2.62829230422642	0	-Inf	0
x6	-16.6393673062091	0	-Inf	0
x7	-39.722777207176	0	-Inf	0
x8	-13.7209293979567	0	-Inf	0
x9	3.93083144894266	0	Inf	0
x10	0.616438000252339	0	Inf	0
x11	-0.00545527046448447	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 890

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587781793	900	0.065467954197977		
Model	58.9211587781793	10	5.89211587781793	Inf	0
Residual	0	890	0		

A = 0.074771, B = 0.380348, C = 0.266016

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.455119, 0.878680, 1.265585, 1.633515, 1.989319]

List of powers for D(x): [0.455119, 0.878680, 1.265585, 1.633515, 1.989319, 2.336509]

Fitting log(x) in range (1.000000, 10.000000)

MSS of errors squared = 3.517210e-14

Corrected MSS of errors squared = 6.091986e-14

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

Fitting log(x) in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.46288544077442	0	-Inf	0
x1	-19.3673121249338	0	-Inf	0
x2	0	0	NaN	NaN
x3	23.830197567283	0	Inf	0
x4	-12.4614564363139	0	-Inf	0
x5	-29.2368497919062	0	-Inf	0
x6	14.6309626360739	0	Inf	0
x7	-17.7369638016	0	-Inf	0
x8	10.6018422771839	0	Inf	0
x9	-3.5242425936154	0	-Inf	0
x10	0.502693216820963	0	Inf	0

Number of observations: 901, Error degrees of freedom: 891

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587782023	900	0.0654679541980025		
Model	58.9211587782023	9	6.54679541980025	Inf	0
Residual	0	891	0		

A = 0.151262, B = 0.231038, C = 0.322361

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.382300, 0.800360, 1.007319]

List of powers for D(x): [0.382300, 0.800360, 1.007319, 1.171683, 1.312601, 1.438008, 1.552151]

Fitting log(x) in range (1.000000, 10.000000)

MSS of errors squared = 1.156091e-13

Corrected MSS of errors squared = 2.002408e-13

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting log(x) in range (1.000000, 10.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.77115127558843	0	-Inf	0
x1	0	0	NaN	NaN
x2	21741.9829195199	0	Inf	0
x3	-20832.8000344269	0	-Inf	0
x4	-906.411733697181	0	-Inf	0
x5	-3604.08062955648	0	-Inf	0
x6	7508.46853897914	0	Inf	0
x7	3465.10266160938	0	Inf	0
x8	131.538673981648	0	Inf	0
x9	-0.343665767059351	0	-Inf	0

Number of observations: 901, Error degrees of freedom: 892

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	58.9211587744444	900	0.0654679541938272		
Model	58.9211587744444	8	7.36514484680555	Inf	0
Residual	0	892	0		

A = 0.856073, B = 3.195061

order Q(x) = 4.000000, order D(x) = 5.000000

List of powers for Q(x): [0.856073, 0.882311, 1.021648, 1.275869]

List of powers for D(x): [0.856073, 0.882311, 1.021648, 1.275869, 1.618702]

Fitting log(x) in range (1.000000, 10.000000)

MSS of errors squared = 4.138004e-14

Corrected MSS of errors squared = 5.852022e-14

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the common logarithm function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*i-C*(i-1)$	0.53	0.1	0	Inf	-Inf	4.29414500E-15	5, 6
$A+B*\sqrt{i}$	0.112	0.1		Inf	-Inf	5.21426500E-15	3, 3
$A+B*i+C*\sqrt{i-1}$	0.052	0.278	0.301	Inf	-Inf	1.88737600E-14	5, 5
$A+B*\sqrt{i}+C*(i-1)$	0.074	0.38	0.266	Inf	-Inf	3.51721000E-14	5, 6
$A+B*\log_{10}(i)^4$	0.856	3.195		Inf	-Inf	4.13800400E-14	4, 5
$A+B*i$	0.492	0.123		Inf	-Inf	5.73763900E-14	5, 6
$A+B/i$	0.453	0.965		Inf	-Inf	8.25488400E-14	6, 4
$A+B*\sqrt{i+C}$	0.117	0.1	0.053	Inf	-Inf	9.81773300E-14	5, 6
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.151	0.231	0.322	Inf	-Inf	1.15609100E-13	3, 7
$A+B*i+C*(i-1)$	0.182	0.218	0.115	Inf	-Inf	9.98447400E-13	7, 6

RESULTS FOR THE COMMON LOGARITHM OF THE GAMMA FUNCTION

Using Power $A+B*i$

Fitting $\log_{10}\Gamma(x)$ in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.821466466466011	0	Inf	0
x1	-2.03164995681268	0	-Inf	0
x2	1.40981546198646	0	Inf	0
x3	0	0	NaN	NaN
x4	-0.199658392817594	0	-Inf	0
x5	3.5210710207115	0	Inf	0
x6	-5.19738896820676	0	-Inf	0
x7	4.12460359942692	0	Inf	0
x8	-1.85666095880281	0	-Inf	0
x9	0.449340566452975	0	Inf	0
x10	-0.0456516909532447	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 971

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.44232855	980	2195.17085951893		

Model	2151267.44232855	9	239029.715814283	Inf	0
Residual	0	971	0		

A = 0.000000, B = 0.100000

order Q(x) = 4.000000, order D(x) = 6.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000]

Fitting log10Gamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 1.963046e-06

Corrected MSS of errors squared = 2.776166e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting log10Gamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.088001144895251	0	Inf	0
x1	-0.112635484572341	0	-Inf	0
x2	0.803149912192791	0	Inf	0
x3	-0.778516765279196	0	-Inf	0
x4	0	0	NaN	NaN
x5	0	0	NaN	NaN
x6	0.38483623225307	0	Inf	0
x7	-26.8425659825419	0	-Inf	0
x8	188.324226586311	0	Inf	0
x9	-358.0545333214	0	-Inf	0
x10	197.187832149695	0	Inf	0

Number of observations: 981, Error degrees of freedom: 972

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.4423282	980	2195.17085951857		
Model	2151267.4423282	8	268908.430291025	Inf	0
Residual	0	972	0		

```

A = 0.038954, B = 0.114883
order Q(x) = 5.000000, order D(x) = 5.000000
List of powers for Q(x): [0.153837, 0.096396, 0.077249, 0.067675, 0.061931]
List of powers for D(x): [0.153837, 0.096396, 0.077249, 0.067675, 0.061931]
Fitting log10Gamma(x) in range (2.000000, 100.000000)
MSS of errors squared = 2.378245e-05
Corrected MSS of errors squared = 3.363347e-05
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}$

```

Fitting log10Gamma(x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i}$ 

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.90777772365962	0	Inf	0
x1	7.5174142467948	0	Inf	0
x2	-3.7536440375383	0	-Inf	0
x3	0	0	NaN	NaN
x4	-14.629082491341	0	-Inf	0
x5	0	0	NaN	NaN
x6	0	0	NaN	NaN
x7	8.95721222739222	0	Inf	0
x8	-12.7529169379462	0	-Inf	0
x9	6.24011415003546	0	Inf	0
x10	-5.02645475812461	0	-Inf	0
x11	2.82473083979006	0	Inf	0
x12	-1.03761146995661	0	-Inf	0
x13	0.223707830770391	0	Inf	0
x14	-0.0214878033110754	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.44232626	980	2195.17085951659		
Model	2151267.44232626	11	195569.767484205	Inf	0
Residual	0	969	0		

```

A = 0.196689, B = 0.676945
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.873635, 1.154035, 1.369193, 1.550580, 1.710386, 1.854860,
1.987719]
List of powers for D(x): [0.873635, 1.154035, 1.369193, 1.550580, 1.710386, 1.854860,
1.987719]
Fitting log10Gamma(x) in range (2.000000, 100.000000)
MSS of errors squared = 2.628136e-10
Corrected MSS of errors squared = 3.716746e-10
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i+C}$

```

Fitting log10Gamma(x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*sqrt(i+C)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	8.39361318412931	0	Inf	0
x1	-264.376392971618	0	-Inf	0
x2	460.673209171141	0	Inf	0
x3	0	0	NaN	NaN
x4	-263.423040530374	0	-Inf	0
x5	0	0	NaN	NaN
x6	0	0	NaN	NaN
x7	58.7325859760675	0	Inf	0
x8	22.3614782803603	0	Inf	0
x9	-64.2007084972414	0	-Inf	0
x10	74.2512689257744	0	Inf	0
x11	-39.6988197138963	0	-Inf	0
x12	8.15911583017222	0	Inf	0

Number of observations: 99, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	220712.747500381	98	2252.17089286103		
Model	220712.747500381	9	24523.6386111535	Inf	0

```

Residual          0      89          0

A = 0.000000, B = 0.126791, C = 1.160190
order Q(x) = 7.000000, order D(x) = 5.000000
List of powers for Q(x): [0.186352, 0.225395, 0.258610, 0.288019, 0.314692, 0.339274,
0.362192]
List of powers for D(x): [0.186352, 0.225395, 0.258610, 0.288019, 0.314692]
Fitting log10Gamma(x) in range (2.000000, 100.000000)
MSS of errors squared = 2.098146e-08
Corrected MSS of errors squared = 3.634096e-08
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting log10Gamma(x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.09390337872941	0	Inf	0
x1	-6.77411562844253	0	-Inf	0
x2	9.59786171889768	0	Inf	0
x3	-11.1097849433761	0	-Inf	0
x4	13.1560619095436	0	Inf	0
x5	-5.96433844546168	0	-Inf	0
x6	2.9736886005153	0	Inf	0
x7	-0.0446668092880984	0	-Inf	0
x8	0.00445138124293748	0	Inf	0
x9	-0.000188061289104892	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 971

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.44232852	980	2195.1708595189		
Model	2151267.44232852	9	239029.71581428	Inf	0
Residual	0	971	0		

A = 0.040611, B = 0.248026, C = 0.020361

```

order Q(x) = 5.000000, order D(x) = 4.000000
List of powers for Q(x): [0.288637, 0.557024, 0.825412, 1.093799, 1.362186]
List of powers for D(x): [0.288637, 0.557024, 0.825412, 1.093799]
Fitting log10Gamma(x) in range (2.000000, 100.000000)
MSS of errors squared = 1.854269e-09
Corrected MSS of errors squared = 3.211688e-09
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i-C*(i-1)$

```

Fitting log10Gamma(x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*i-C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.765803567490689	0	Inf	0
x1	-0.163795759704938	0	-Inf	0
x2	0.591700675747153	0	Inf	0
x3	-1.19372564870729	0	-Inf	0
x4	0.00391967574463522	0	Inf	0
x5	-0.0553582946318991	0	-Inf	0
x6	0.665253789790442	0	Inf	0
x7	0.483788634888977	0	Inf	0
x8	-0.0978386707961098	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 972

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.44232858	980	2195.17085951896		
Model	2151267.44232858	8	268908.430291073	Inf	0
Residual	0	972	0		

A = 0.220968, B = 0.100000, C = 0.236886

order Q(x) = 3.000000, order D(x) = 5.000000

List of powers for Q(x): [0.320968, 0.184082, 0.047196]

List of powers for D(x): [0.320968, 0.184082, 0.047196, -0.089690, -0.226576]

Fitting log10Gamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 1.103332e-06

Corrected MSS of errors squared = 1.911028e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting log10Gamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.42123148097441	0	-Inf	0
x1	-7.49462637195772	0	-Inf	0
x2	1.15076811839519	0	Inf	0
x3	0	0	NaN	NaN
x4	15.5039894753428	0	Inf	0
x5	0	0	NaN	NaN
x6	-7.73996744980882	0	-Inf	0
x7	6.61057069084886	0	Inf	0
x8	6.7527129346609	0	Inf	0
x9	0.337903244700781	0	Inf	0
x10	-0.0131996811957404	0	-Inf	0
x11	0.000520534216168253	0	Inf	0
x12	-9.68295558068155e-06	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.4423289	980	2195.17085951928		
Model	2151267.4423289	10	215126.74423289	Inf	0
Residual	0	970	0		

A = 0.044516, B = 0.289845, C = 0.010546

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.334361, 0.634753, 0.928967, 1.222164, 1.514836, 1.807171]

List of powers for D(x): [0.334361, 0.634753, 0.928967, 1.222164, 1.514836, 1.807171]

Fitting log10Gamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 4.067030e-10

Corrected MSS of errors squared = 7.044302e-10

```

R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}+C*(i-1)$

```

Fitting log10Gamma(x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is  $A+B\sqrt{i}+C*(i-1)$ 

```

Linear regression model:

```

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10

```

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.886351513869465	0	-Inf	0
x1	-0.365371469976944	0	-Inf	0
x2	-0.284426554303293	0	-Inf	0
x3	3.20208563783325	0	Inf	0
x4	-1.39358652889752	0	-Inf	0
x5	-0.266464742261947	0	-Inf	0
x6	-0.00606922735253027	0	-Inf	0
x7	2.37050194527565e-05	0	Inf	0
x8	2.54862830254448	0	Inf	0
x9	0.466584574959352	0	Inf	0
x10	0.0182325220283437	0	Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.44232859	980	2195.17085951896		
Model	2151267.44232859	10	215126.744232859	Inf	0
Residual	0	970	0		

A = 0.051926, B = 0.100000, C = 0.387237

order Q(x) = 7.000000, order D(x) = 3.000000

List of powers for Q(x): [0.151926, 0.580584, 0.999604, 1.413636, 1.824480, 2.233059, 2.639922]

List of powers for D(x): [0.151926, 0.580584, 0.999604]

Fitting log10Gamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 3.611675e-10

Corrected MSS of errors squared = 6.255604e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting log10Gamma(x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-27.2783014928659	0	-Inf	0
x1	193.366426855826	0	Inf	0
x2	-530.896323809137	0	-Inf	0
x3	0	0	NaN	NaN
x4	947.040595162078	0	Inf	0
x5	0	0	NaN	NaN
x6	-1202.26298818729	0	-Inf	0
x7	620.030374167757	0	Inf	0
x8	1.34817742157071	0	Inf	0
x9	-0.833918451998216	0	-Inf	0
x10	0.266820408592356	0	Inf	0

Number of observations: 981, Error degrees of freedom: 972

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.44232862	980	2195.170859519		
Model	2151267.44232862	8	268908.430291077	Inf	0
Residual	0	972	0		

A = 0.000000, B = 0.100000, C = 0.054362

order Q(x) = 7.000000, order D(x) = 3.000000

List of powers for Q(x): [0.100000, 0.195783, 0.250084, 0.294157, 0.332330, 0.366505, 0.397733]

List of powers for D(x): [0.100000, 0.195783, 0.250084]

Fitting log10Gamma(x) in range (2.000000, 100.000000)

MSS of errors squared = 6.822774e-08

Corrected MSS of errors squared = 1.181739e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\log_{10}(i)^4$

Fitting $\log_{10}\Gamma(x)$ in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	1.99961830648737e-08	0	Inf	0
x2	0	0	NaN	NaN
x3	-3.11758980320813e-08	0	-Inf	0
x4	1.11644532147849e-08	0	Inf	0
x5	0.999999917042169	0	Inf	0
x6	9.17996614628453e-08	0	Inf	0
x7	-1.18243145907812e-08	0	-Inf	0
x8	7.24554356943926e-10	0	Inf	0
x9	-1.92452421488811e-11	0	-Inf	0
x10	8.6457941604409e-14	0	Inf	0
x11	3.24734067495661e-15	0	Inf	0

Number of observations: 981, Error degrees of freedom: 971

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2151267.44232858	980	2195.17085951896		
Model	2151267.44232858	9	239029.715814287	Inf	0
Residual	0	971	0		

A = 0.000000, B = 3.500000

order $Q(x)$ = 4.000000, order $D(x)$ = 7.000000

List of powers for $Q(x)$: [0.000000, 0.028741, 0.181377, 0.459862]

List of powers for $D(x)$: [0.000000, 0.028741, 0.181377, 0.459862, 0.835415, 1.283288, 1.785239]

Fitting $\log_{10}\Gamma(x)$ in range (2.000000, 100.000000)

MSS of errors squared = 3.544408e-06

Corrected MSS of errors squared = 5.012549e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the common logarithm of the gamma function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*sqrt(i)	0.196	0.676		Inf	-Inf	2.62813600E-10	7, 7
A+B*sqrt(i)+C*(i-1)	0.051	0.1	0.387	Inf	-Inf	3.61167500E-10	7, 3
A+B*i+C*sqrt(i-1)	0.044	0.289	0.01	Inf	-Inf	4.06703000E-10	6, 6
A+B*i+C*(i-1)	0.04	0.248	0.02	Inf	-Inf	1.85426900E-09	5, 4
A+B*sqrt(i+C)	0	0.126	1.16	Inf	-Inf	2.09814600E-08	7, 5
A+B*sqrt(i)+C*sqrt(i-1)	0	0.1	0.054	Inf	-Inf	6.82277400E-08	7, 3
A+B*i-C*(i-1)	0.22	0.1	0.236	Inf	-Inf	1.10333200E-06	3, 5
A+B*i	0	0.1		Inf	-Inf	1.96304600E-06	4, 6
A+B*log10(i)^4	0	3.5		Inf	-Inf	3.54440800E-06	4, 7
A+B/i	0.038	0.114		Inf	-Inf	2.37824500E-05	5, 5

RESULTS FOR THE COMMON EXPONENT (10^X)

Using Power A+B* i

Fitting 10^x in range (0.000000, 1.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999998214	0	Inf	0
x1	-66.7969895360496	0	-Inf	0
x2	209.626732742231	0	Inf	0
x3	-239.038809989091	0	-Inf	0
x4	74.6592068102763	0	Inf	0
x5	71.1448288200368	0	Inf	0
x6	-67.2231626680505	0	-Inf	0
x7	16.7236643569933	0	Inf	0
x8	65.7018164998608	0	Inf	0
x9	-199.47045273076	0	-Inf	0
x10	200.576995269033	0	Inf	0
x11	0	0	NaN	NaN
x12	-146.783780103911	0	-Inf	0
x13	104.761258795901	0	Inf	0
x14	-23.7953847837527	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	645.004627337368	100	6.45004627337368		
Model	645.004627337368	13	49.6157405644129	Inf	0
Residual	0	87	0		

A = 0.367134, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.467134, 0.567134, 0.667134, 0.767134, 0.867134, 0.967134, 1.067134]

List of powers for D(x): [0.467134, 0.567134, 0.667134, 0.767134, 0.867134, 0.967134, 1.067134]

Fitting 10^x in range (0.000000, 1.000000)

MSS of errors squared = 1.161610e-10

Corrected MSS of errors squared = 1.642765e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting 10^x in range (0.000000, 1.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000000000031	0	Inf	0
x1	0.189078264632143	0	Inf	0
x2	-6.86506694988387	0	-Inf	0
x3	84.0772142233575	0	Inf	0
x4	-321.03400814076	0	-Inf	0
x5	456.634716175001	0	Inf	0
x6	-213.620610160512	0	-Inf	0
x7	-0.0286381081264091	0	-Inf	0
x8	6.1139825087904	0	Inf	0
x9	-83.7463362413515	0	-Inf	0
x10	323.415911625009	0	Inf	0
x11	-459.461618450568	0	-Inf	0
x12	214.668566325427	0	Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	645.004627337566	100	6.45004627337566		
Model	645.004627337566	12	53.7503856114639	Inf	0
Residual	0	88	0		

A = 0.304803, B = 1.521183

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [1.825986, 1.065394, 0.811864, 0.685099, 0.609039, 0.558333]

List of powers for D(x): [1.825986, 1.065394, 0.811864, 0.685099, 0.609039, 0.558333]

Fitting 10^x in range (0.000000, 1.000000)

MSS of errors squared = 2.011185e-09

Corrected MSS of errors squared = 2.844246e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}$

Fitting 10^x in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000000013259	0	Inf	0
x1	-0.205580545455225	0	-Inf	0
x2	21.1287845848849	0	Inf	0
x3	-176.359067024344	0	-Inf	0
x4	493.464088285461	0	Inf	0
x5	-614.946132847315	0	-Inf	0
x6	350.878505921102	0	Inf	0
x7	-73.1467948151202	0	-Inf	0
x8	-0.46595683757024	0	-Inf	0
x9	17.5318225772412	0	Inf	0
x10	-29.0854813821837	0	-Inf	0
x11	12.8382352926064	0	Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	645.004627337404	100	6.45004627337404		
Model	645.004627337404	11	58.6367843034003	Inf	0
Residual	0	89	0		

A = 0.543869, B = 0.420425

order Q(x) = 7.000000, order D(x) = 4.000000

List of powers for Q(x): [0.964294, 1.138440, 1.272066, 1.384719, 1.483967, 1.573695, 1.656208]

List of powers for D(x): [0.964294, 1.138440, 1.272066, 1.384719]

Fitting 10^x in range (0.000000, 1.000000)

MSS of errors squared = 7.137793e-09

Corrected MSS of errors squared = 1.009436e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting 10^x in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.99999999992621	0	Inf	0
x1	-357.298456108889	0	-Inf	0
x2	1320.35231269768	0	Inf	0
x3	-1446.47816604095	0	-Inf	0
x4	0	0	NaN	NaN
x5	839.612757497386	0	Inf	0
x6	-357.177821381107	0	-Inf	0
x7	292.54324724564	0	Inf	0
x8	-893.585957125411	0	-Inf	0
x9	648.223948327099	0	Inf	0
x10	0	0	NaN	NaN
x11	674.206217786515	0	Inf	0
x12	-1222.808927549	0	-Inf	0
x13	502.420408649047	0	Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	645.004627337235	100	6.45004627337235		
Model	645.004627337235	11	58.636784303385	Inf	0
Residual	0	89	0		

A = 0.151516, B = 0.100000, C = 0.669555

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.280727, 0.314903, 0.343076, 0.367607, 0.389624, 0.409770]

List of powers for D(x): [0.280727, 0.314903, 0.343076, 0.367607, 0.389624, 0.409770, 0.428455]

Fitting 10^x in range (0.000000, 1.000000)

MSS of errors squared = 5.093332e-08

Corrected MSS of errors squared = 8.821910e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting 10^x in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.00000000012569	0	Inf	0
x1	0.277457555648942	0	Inf	0
x2	-1.62849521097402	0	-Inf	0
x3	3.40718182264866	0	Inf	0
x4	-2.38092502859325	0	-Inf	0
x5	1.14533462158629	0	Inf	0
x6	-0.280166276855088	0	-Inf	0
x7	1.7680905567585	0	Inf	0
x8	0.284527992638486	0	Inf	0
x9	-1.64837100206719	0	-Inf	0
x10	0.788445111379962	0	Inf	0
x11	-0.09458175271528	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	645.004627337434	100	6.45004627337434		
Model	645.004627337434	11	58.6367843034031	Inf	0
Residual	0	89	0		

A = 0.082618, B = 0.235787, C = 0.159722

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.318406, 0.713915, 1.109424, 1.504933, 1.900442]

List of powers for D(x): [0.318406, 0.713915, 1.109424, 1.504933, 1.900442, 2.295951]

Fitting 10^x in range (0.000000, 1.000000)

MSS of errors squared = 6.654980e-09

Corrected MSS of errors squared = 1.152676e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i-C*(i-1)

Fitting 10^x in range (0.000000, 1.000000)

Pade-Shammas polynomial power is A+B*i-C*(i-1)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999996695	0	Inf	0
x1	1412.87098839969	0	Inf	0
x2	-9767.28620658573	0	-Inf	0
x3	28659.3287131371	0	Inf	0
x4	-45355.5303083513	0	-Inf	0
x5	40563.7923285068	0	Inf	0
x6	-19349.6229461457	0	-Inf	0
x7	3835.69255778985	0	Inf	0
x8	-100.036703604403	0	-Inf	0
x9	312.614152394213	0	Inf	0
x10	-317.122377881427	0	-Inf	0
x11	105.520416412507	0	Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	645.004627337111	100	6.45004627337111		
Model	645.004627337111	11	58.6367843033737	Inf	0

```

Residual          0      89          0

A = 0.062238, B = 0.227133, C = 0.181501
order Q(x) = 7.000000, order D(x) = 4.000000
List of powers for Q(x): [0.289372, 0.335004, 0.380637, 0.426269, 0.471901, 0.517534,
0.563166]
List of powers for D(x): [0.289372, 0.335004, 0.380637, 0.426269]
Fitting 10^x in range (0.000000, 1.000000)
MSS of errors squared = 3.306578e-08
Corrected MSS of errors squared = 5.727161e-08
R-Squared =      1.00000000
R-Squared Adjusted =      1.00000000
Particle swarm AICc =  -1.000000e+99
AIC =  -Inf
AICc =  -Inf

```

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting 10^x in range (0.000000, 1.000000)
Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.999999999999411	0	Inf	0
x1	-4.63160196447215	0	-Inf	0
x2	11.1941715276166	0	Inf	0
x3	-13.0927319893983	0	-Inf	0
x4	6.49446105943609	0	Inf	0
x5	2.88994712201787	0	Inf	0
x6	-5.16052805139882	0	-Inf	0
x7	2.06466505333297	0	Inf	0
x8	4.65027650591438	0	Inf	0
x9	-11.4876780746352	0	-Inf	0
x10	15.1920678159618	0	Inf	0
x11	-9.83547857495268	0	-Inf	0
x12	2.40497405303729	0	Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	645.00462733744	100	6.4500462733744		
Model	645.00462733744	12	53.7503856114533	Inf	0

```

Residual          0      88          0

A = 0.149684, B = 0.229834, C = 0.000000
order Q(x) = 7.000000, order D(x) = 5.000000
List of powers for Q(x): [0.379518, 0.609352, 0.839186, 1.069020, 1.298853, 1.528687,
1.758521]
List of powers for D(x): [0.379518, 0.609352, 0.839186, 1.069020, 1.298853]
Fitting 10^x in range (0.000000, 1.000000)
MSS of errors squared = 2.848532e-09
Corrected MSS of errors squared = 4.933802e-09
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}+C*(i-1)$

```

Fitting 10^x in range (0.000000, 1.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.000000000000021	0	Inf	0
x1	7.35713660096297	0	Inf	0
x2	-14.0708336097598	0	-Inf	0
x3	20.7230249571311	0	Inf	0
x4	-16.4514890446156	0	-Inf	0
x5	10.9961419963106	0	Inf	0
x6	-4.68484306807967	0	-Inf	0
x7	1.70248398005301	0	Inf	0
x8	-7.3627125413345	0	-Inf	0
x9	14.7927387876326	0	Inf	0
x10	-15.2072102018451	0	-Inf	0
x11	25.4529263133997	0	Inf	0
x12	-31.1580406652	0	-Inf	0
x13	17.3350924557674	0	Inf	0
x14	-3.50995632967308	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 86

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	645.004627337419	100	6.45004627337419		
Model	645.004627337419	14	46.07175909553	Inf	0
Residual	0	86	0		

A = 0.311348, B = 0.196554, C = 0.290161

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.507903, 0.879479, 1.232112, 1.574939, 1.911500, 2.243609, 2.572346]

List of powers for D(x): [0.507903, 0.879479, 1.232112, 1.574939, 1.911500, 2.243609, 2.572346]

Fitting 10^x in range (0.000000, 1.000000)

MSS of errors squared = 2.361292e-11

Corrected MSS of errors squared = 4.089878e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

Fitting 10^x in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.000000000000218	0	Inf	0
x1	-19.9643575173662	0	-Inf	0
x2	196.185565642624	0	Inf	0
x3	-1007.56278297722	0	-Inf	0
x4	2681.41688622001	0	Inf	0
x5	-3689.02926601697	0	-Inf	0
x6	2499.68764685113	0	Inf	0
x7	-661.980412530319	0	-Inf	0
x8	16.2280442638961	0	Inf	0
x9	-57.1423223751807	0	-Inf	0
x10	66.3919442765943	0	Inf	0
x11	-24.4529941307128	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	645.004627337548	100	6.45004627337548		
Model	645.004627337548	11	58.6367843034135	Inf	0
Residual	0	89	0		

A = 0.119023, B = 0.207702, C = 0.085167

order Q(x) = 7.000000, order D(x) = 4.000000

List of powers for Q(x): [0.326725, 0.497925, 0.599217, 0.681940, 0.753792, 0.818226, 0.877166]

List of powers for D(x): [0.326725, 0.497925, 0.599217, 0.681940]

Fitting 10^x in range (0.000000, 1.000000)

MSS of errors squared = 2.642460e-08

Corrected MSS of errors squared = 4.576875e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting 10^x in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.99999999992749	0	Inf	0
x1	-2402.57230616887	0	-Inf	0
x2	3085.40449610539	0	Inf	0
x3	-847.091735319695	0	-Inf	0
x4	189.830046664802	0	Inf	0
x5	-28.9829026960168	0	-Inf	0
x6	2.4356750399256	0	Inf	0
x7	2245.21456866834	0	Inf	0
x8	-2866.80913430612	0	-Inf	0
x9	757.013931360307	0	Inf	0
x10	-149.495569163009	0	-Inf	0
x11	15.0738760783889	0	Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	645.004627337106	100	6.45004627337106		

Model	645.004627337106	11	58.6367843033733	Inf	0
Residual	0	89	0		

A = 0.332069, B = 0.916622

order Q(x) = 6.000000, order D(x) = 5.000000

List of powers for Q(x): [0.332069, 0.339596, 0.379570, 0.452503, 0.550857, 0.668151]

List of powers for D(x): [0.332069, 0.339596, 0.379570, 0.452503, 0.550857]

Fitting 10^x in range (0.000000, 1.000000)

MSS of errors squared = 3.566797e-08

Corrected MSS of errors squared = 5.044213e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the commn exponent function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\sqrt{i}+C*(i-1)$	0.311	0.196	0.29	Inf	-Inf	2.36129200E-11	7, 7
$A+B*i$	0.367	0.1		Inf	-Inf	1.16161000E-10	7, 7
$A+B/i$	0.304	1.521		Inf	-Inf	2.01118500E-09	6, 6
$A+B*i+C*\sqrt{i-1}$	0.149	0.229	0	Inf	-Inf	2.84853200E-09	7, 5
$A+B*i+C*(i-1)$	0.082	0.235	0.159	Inf	-Inf	6.65498000E-09	5, 6
$A+B*\sqrt{i}$	0.543	0.42		Inf	-Inf	7.13779300E-09	7, 4
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.119	0.207	0.085	Inf	-Inf	2.64246000E-08	7, 4
$A+B*i-C*(i-1)$	0.062	0.227	0.181	Inf	-Inf	3.30657800E-08	7, 4
$A+B*\log_{10}(i)^4$	0.332	0.916		Inf	-Inf	3.56679700E-08	6, 5
$A+B*\sqrt{i+C}$	0.151	0.1	0.669	Inf	-Inf	5.09333200E-08	6, 7

RESULTS FOR THE SINE INTEGRAL

Using Power $A+B*i$

Fitting $\text{Si}(x)$ in range (1.300000, 20.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.425556690751173	0	-Inf	0
x1	2.47044534545918	0	Inf	0
x2	-6.09782731177053	0	-Inf	0
x3	8.30314362814464	0	Inf	0
x4	-6.7405331930106	0	-Inf	0
x5	3.26404532971158	0	Inf	0
x6	-0.873329665332348	0	-Inf	0
x7	0.0996294253655694	0	Inf	0
x8	5.60579140933404	0	Inf	0
x9	-13.4770435329002	0	-Inf	0
x10	18.0113541933338	0	Inf	0
x11	-14.4506048302733	0	-Inf	0
x12	6.95960922946009	0	Inf	0
x13	-1.86290822219788	0	-Inf	0
x14	0.213783905679764	0	Inf	0

Number of observations: 188, Error degrees of freedom: 173

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	2.24017998653672	187	0.0119795721205172		
Model	2.24017998653672	14	0.160012856181194	Inf	0
Residual	0	173	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 3.594849e-04

Corrected MSS of errors squared = 5.083884e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0362976298383173	0	-Inf	0
x1	-0.0188500025879627	0	-Inf	0
x2	4.22116045657442	0	Inf	0
x3	-61.6033608846453	0	-Inf	0
x4	235.555828136321	0	Inf	0
x5	-323.432628202178	0	-Inf	0
x6	145.315118933014	0	Inf	0
x7	0.132155986276034	0	Inf	0
x8	-84.3564073151651	0	-Inf	0
x9	2503.99480099526	0	Inf	0
x10	-17837.3575910223	0	-Inf	0
x11	47337.0829526069	0	Inf	0
x12	-52140.0586995271	0	-Inf	0
x13	20221.561764266	0	Inf	0

Number of observations: 188, Error degrees of freedom: 174

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.24017998524394	187	0.011979572113604		
Model	2.24017998524394	13	0.172321537326457	Inf	0
Residual	0	174	0		

A = 0.410005, B = 0.784603

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [1.194608, 0.802307, 0.671540, 0.606156, 0.566926, 0.540773]

List of powers for D(x): [1.194608, 0.802307, 0.671540, 0.606156, 0.566926, 0.540773, 0.522092]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 2.592084e-03

Corrected MSS of errors squared = 3.665760e-03

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0565126209889402	0	-Inf	0
x1	90.7959353724243	0	Inf	0
x2	-562.844459534331	0	-Inf	0
x3	1195.91877776555	0	Inf	0
x4	-969.91098829601	0	-Inf	0
x5	0	0	NaN	NaN
x6	395.55959345562	0	Inf	0
x7	-149.462277573786	0	-Inf	0
x8	4349.82650937993	0	Inf	0
x9	-39381.0334311569	0	-Inf	0
x10	133555.20682181	0	Inf	0
x11	-227292.266567069	0	-Inf	0
x12	209101.837648503	0	Inf	0
x13	-99740.2306687372	0	-Inf	0
x14	19407.6596148649	0	Inf	0

Number of observations: 188, Error degrees of freedom: 174

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.24017998820523	187	0.0119795721294397		
Model	2.24017998820523	13	0.172321537554248	Inf	0
Residual	0	174	0		

A = 0.240418, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.340418, 0.381839, 0.413623, 0.440418, 0.464025, 0.485367, 0.504993]

List of powers for D(x): [0.340418, 0.381839, 0.413623, 0.440418, 0.464025, 0.485367, 0.504993]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 2.737419e-03

Corrected MSS of errors squared = 3.871294e-03

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.0071018383288281	0	Inf	0
x1	-0.506294960693741	0	-Inf	0
x2	1.39048060716888	0	Inf	0
x3	-1.30680626861986	0	-Inf	0
x4	0.415534561690554	0	Inf	0
x5	479.910832161863	0	Inf	0
x6	-2246.77194714639	0	-Inf	0
x7	3920.07060127484	0	Inf	0
x8	-2738.63629927552	0	-Inf	0
x9	0	0	NaN	NaN
x10	899.028380712612	0	Inf	0
x11	-312.601584200232	0	-Inf	0

Number of observations: 188, Error degrees of freedom: 177

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.24017998652738	187	0.0119795721204673		
Model	2.24017998652738	10	0.224017998652738	Inf	0
Residual	0	177	0		

A = 0.000000, B = 0.100000, C = 2.000000

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.173205, 0.200000, 0.223607, 0.244949]

List of powers for D(x): [0.173205, 0.200000, 0.223607, 0.244949, 0.264575, 0.282843, 0.300000]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 2.510348e-02

Corrected MSS of errors squared = 4.348051e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.427930999621684	0	-Inf	0
x1	3.66170767389108	0	Inf	0
x2	-10.1512202254442	0	-Inf	0
x3	14.3756829940615	0	Inf	0
x4	-11.8830014274407	0	-Inf	0
x5	5.80740522922075	0	Inf	0
x6	-1.56103804814469	0	-Inf	0
x7	0.178432001047373	0	Inf	0
x8	8.14716502100531	0	Inf	0
x9	-21.9708640850488	0	-Inf	0
x10	30.5058782336262	0	Inf	0
x11	-24.8991551211206	0	-Inf	0
x12	12.0934654792223	0	Inf	0
x13	-3.25001178932425	0	-Inf	0
x14	0.373482902243512	0	Inf	0

Number of observations: 188, Error degrees of freedom: 173

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.24017998653683	187	0.0119795721205178		
Model	2.24017998653683	14	0.160012856181202	Inf	0
Residual	0	173	0		

A = 0.029957, B = 0.102298, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.132255, 0.234552, 0.336850, 0.439147, 0.541445, 0.643743, 0.746040]

List of powers for D(x): [0.132255, 0.234552, 0.336850, 0.439147, 0.541445, 0.643743, 0.746040]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 3.431560e-04

Corrected MSS of errors squared = 5.943636e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.428548323898187	0	-Inf	0
x1	0.271845085890291	0	Inf	0
x2	-2.36166226693914	0	-Inf	0
x3	8.71050226303403	0	Inf	0
x4	-17.6208958020763	0	-Inf	0
x5	20.9618263112148	0	Inf	0
x6	-14.3770076583104	0	-Inf	0
x7	4.84399518407974	0	Inf	0
x8	0.562128595059939	0	Inf	0
x9	-4.86058462656482	0	-Inf	0
x10	17.9401692450159	0	Inf	0
x11	-36.5296925138925	0	-Inf	0
x12	44.0162448330191	0	Inf	0

x13	-30.7901715943759	0	-Inf	0
x14	10.6618480833924	0	Inf	0

Number of observations: 188, Error degrees of freedom: 173

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.24017998653618	187	0.0119795721205143		
Model	2.24017998653618	14	0.160012856181155	Inf	0
Residual	0	173	0		

A = 0.614502, B = 0.148505, C = 0.250000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.763007, 0.661512, 0.560017, 0.458522, 0.357027, 0.255532, 0.154037]

List of powers for D(x): [0.763007, 0.661512, 0.560017, 0.458522, 0.357027, 0.255532, 0.154037]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 3.357664e-04

Corrected MSS of errors squared = 5.815645e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.139448418746337	0	-Inf	0
x1	0.286066405585294	0	Inf	0
x2	-0.559987935085542	0	-Inf	0
x3	0.852230819364153	0	Inf	0
x4	-0.645966699276785	0	-Inf	0
x5	0.244460520577484	0	Inf	0
x6	-0.0369813693253091	0	-Inf	0
x7	1.97157588980792	0	Inf	0
x8	-4.03047480316782	0	-Inf	0
x9	7.10090159761033	0	Inf	0

x10	-6.78572384439967	0	-Inf	0
x11	3.69103134261753	0	Inf	0
x12	-1.08040792141346	0	-Inf	0
x13	0.132703591852692	0	Inf	0

Number of observations: 188, Error degrees of freedom: 174

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.24017998653667	187	0.011979572120517		
Model	2.24017998653667	13	0.172321537425898	Inf	0
Residual	0	174	0		

A = 0.000000, B = 0.100000, C = 0.213990

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.413990, 0.602627, 0.770641, 0.927980, 1.078496]

List of powers for D(x): [0.100000, 0.413990, 0.602627, 0.770641, 0.927980, 1.078496, 1.224166]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 6.753705e-03

Corrected MSS of errors squared = 1.169776e-02

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.422619684349003	0	-Inf	0
x1	50.5652027705901	0	Inf	0
x2	-331.979038581483	0	-Inf	0
x3	897.039485849604	0	Inf	0
x4	-1270.19699554263	0	-Inf	0
x5	997.535939286971	0	Inf	0
x6	-413.263635576388	0	-Inf	0
x7	70.7245043752207	0	Inf	0
x8	91.433827350713	0	Inf	0

x9	-577.234276717155	0	-Inf	0
x10	1525.48583411634	0	Inf	0
x11	-2129.98602927722	0	-Inf	0
x12	1657.60655289024	0	Inf	0
x13	-682.757051931212	0	-Inf	0
x14	116.448134418487	0	Inf	0

Number of observations: 188, Error degrees of freedom: 173

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.24017998661716	187	0.0119795721209474		
Model	2.24017998661716	14	0.16001285618694	Inf	0
Residual	0	173	0		

A = 0.156337, B = 0.336640, C = 0.000000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.492976, 0.632417, 0.739414, 0.829616, 0.909086, 0.980932, 1.047001]

List of powers for D(x): [0.492976, 0.632417, 0.739414, 0.829616, 0.909086, 0.980932, 1.047001]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 2.427920e-04

Corrected MSS of errors squared = 4.205281e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0887151026355799	0	-Inf	0
x1	1.33799990465402	0	Inf	0
x2	-11.547119421802	0	-Inf	0
x3	31.18904092885	0	Inf	0
x4	-37.7878149196761	0	-Inf	0
x5	21.7582941490811	0	Inf	0

x6	-4.86136110744107	0	-Inf	0
x7	16.7450639652319	0	Inf	0
x8	-193.385179689841	0	-Inf	0
x9	678.305628519121	0	Inf	0
x10	-1127.4322360645	0	-Inf	0
x11	997.116566141211	0	Inf	0
x12	-454.88149923794	0	-Inf	0
x13	84.5313138854043	0	Inf	0

Number of observations: 188, Error degrees of freedom: 174

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.2401799865264	187	0.011979572120462		
Model	2.2401799865264	13	0.172321537425108	Inf	0
Residual	0	174	0		

A = 0.154192, B = 0.107765, C = 0.146183

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.261957, 0.452778, 0.547581, 0.622919, 0.687529, 0.745037]

List of powers for D(x): [0.261957, 0.452778, 0.547581, 0.622919, 0.687529, 0.745037, 0.797386]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 2.718253e-03

Corrected MSS of errors squared = 4.708152e-03

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting Si(x) in range (1.300000, 20.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.118678567816826	0	-Inf	0
x1	617.590575952046	0	Inf	0
x2	-799.303750976383	0	-Inf	0
x3	227.546484828179	0	Inf	0
x4	-53.1705731495321	0	-Inf	0

x5	8.01895959287142	0	Inf	0
x6	-0.560910059962875	0	-Inf	0
x7	5142.04985927783	0	Inf	0
x8	-6711.88030268056	0	-Inf	0
x9	2012.45891649623	0	Inf	0
x10	-534.004558232708	0	-Inf	0
x11	104.531262652895	0	Inf	0
x12	-12.9006909947583	0	-Inf	0
x13	0.743288692601158	0	Inf	0

Number of observations: 188, Error degrees of freedom: 174

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	2.24017998645108	187	0.0119795721200592		
Model	2.24017998645108	13	0.172321537419314	Inf	0
Residual	0	174	0		

A = 0.543952, B = 1.381114

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.543952, 0.555294, 0.615524, 0.725415, 0.873610, 1.050343]

List of powers for D(x): [0.543952, 0.555294, 0.615524, 0.725415, 0.873610, 1.050343, 1.248415]

Fitting Si(x) in range (1.300000, 20.000000)

MSS of errors squared = 5.395159e-03

Corrected MSS of errors squared = 7.629906e-03

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the sine Integral. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\sqrt{i}+C*(i-1)$	0.156	0.336	0	Inf	-Inf	2.42792000E-04	7, 7
$A+B*i-C*(i-1)$	0.614	0.148	0.25	Inf	-Inf	3.35766400E-04	7, 7
$A+B*i+C*(i-1)$	0.029	0.102	0	Inf	-Inf	3.43156000E-04	7, 7
$A+B*i$	0	0.1		Inf	-Inf	3.59484900E-04	7, 7
$A+B/i$	0.41	0.784		Inf	-Inf	2.59208400E-03	6, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.154	0.107	0.146	Inf	-Inf	2.71825300E-03	6, 7
$A+B*\sqrt{i}$	0.24	0.1		Inf	-Inf	2.73741900E-03	7, 7
$A+B*\log_{10}(i)^4$	0.543	1.381		Inf	-Inf	5.39515900E-03	6, 7
$A+B*i+C*\sqrt{i-1}$	0	0.1	0.213	Inf	-Inf	6.75370500E-03	6, 7
$A+B*\sqrt{i+C}$	0	0.1	2	Inf	-Inf	2.51034800E-02	4, 7

RESULTS FOR THE SINE FUNCTION

Using Power $A+B*i$

Fitting $\sin(x)$ in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.2068921216943e-11	0	-Inf	0
x1	-0.0786726285995106	0	-Inf	0
x2	0.918514374263889	0	Inf	0
x3	-4.22728353259993	0	-Inf	0
x4	9.43597675240397	0	Inf	0
x5	-8.94182178158167	0	-Inf	0
x6	2.93317433307807	0	Inf	0
x7	70.854773913851	0	Inf	0
x8	-259.835135548709	0	-Inf	0
x9	406.72316034698	0	Inf	0
x10	-334.554865969934	0	-Inf	0
x11	142.853535045707	0	Inf	0
x12	-25.0888699196799	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125749	500	0.0952531872251497		
Model	47.6265936125749	12	3.96888280104791	Inf	0

Residual 0 488 0

A = 0.418401, B = 0.111637

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.530037, 0.641674, 0.753311, 0.864947, 0.976584, 1.088220]

List of powers for D(x): [0.530037, 0.641674, 0.753311, 0.864947, 0.976584, 1.088220]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.810734e-09

Corrected MSS of errors squared = 2.560765e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.07284607614182e-12	0	-Inf	0
x1	0.0480955893501182	0	Inf	0
x2	-1.97611116459285	0	-Inf	0
x3	15.2301840347718	0	Inf	0
x4	-41.6917226452856	0	-Inf	0
x5	46.6029398659128	0	Inf	0
x6	-18.2129015575006	0	-Inf	0
x7	-1.48933899471062	0	-Inf	0
x8	185.77464477222	0	Inf	0
x9	-2185.14957870701	0	-Inf	0
x10	6934.27867233445	0	Inf	0
x11	-6869.07571581975	0	-Inf	0
x12	0	0	NaN	NaN
x13	1936.66074108584	0	Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936130762	500	0.0952531872261523		
Model	47.6265936130762	12	3.96888280108968	Inf	0

Residual 0 488 0

A = 0.198937, B = 0.350751

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.549688, 0.374312, 0.315854, 0.286625, 0.269087, 0.257395]

List of powers for D(x): [0.549688, 0.374312, 0.315854, 0.286625, 0.269087, 0.257395, 0.249044]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.543643e-09

Corrected MSS of errors squared = 2.183041e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.14907506432789e-12	0	-Inf	0
x1	-0.0471789918681063	0	-Inf	0
x2	0.47232572523849	0	Inf	0
x3	-1.57994056124427	0	-Inf	0
x4	2.39182240933276	0	Inf	0
x5	-1.69650576011482	0	-Inf	0
x6	0.459491044367077	0	Inf	0
x7	44.2484667018629	0	Inf	0
x8	-249.101739077195	0	-Inf	0
x9	575.804149397217	0	Inf	0
x10	-660.344659250549	0	-Inf	0
x11	374.443886847244	0	Inf	0
x12	-84.0501210965179	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936126044	500	0.0952531872252087		
Model	47.6265936126044	12	3.96888280105036	Inf	0

Residual 0 488 0

A = 0.000000, B = 0.100000

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 4.219952e-09

Corrected MSS of errors squared = 5.967913e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B\sqrt{i+C}$

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.67854229987735e-09	0	-Inf	0
x1	-259.038151201195	0	-Inf	0
x2	1737.03175181744	0	Inf	0
x3	-3323.95295008896	0	-Inf	0
x4	0	0	NaN	NaN
x5	6604.04380417211	0	Inf	0
x6	-7042.17895975531	0	-Inf	0
x7	2284.13800689623	0	Inf	0
x8	37005.8595848865	0	Inf	0
x9	-215413.958439786	0	-Inf	0
x10	444396.776377346	0	Inf	0
x11	-355802.47954155	0	-Inf	0
x12	0	0	NaN	NaN
x13	144839.335873703	0	Inf	0
x14	-55024.5855519719	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936448717	500	0.0952531872897434		

Model	47.6265936448717	12	3.96888280373931	Inf	0
Residual	0	488	0		

A = 0.589351, B = 0.100000, C = 0.974335

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.729862, 0.761813, 0.788708, 0.812383, 0.833775, 0.853440, 0.871739]

List of powers for D(x): [0.729862, 0.761813, 0.788708, 0.812383, 0.833775, 0.853440, 0.871739]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.331523e-09

Corrected MSS of errors squared = 2.306265e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*i+C*(i-1)

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is A+B*i+C*(i-1)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.5406877295714e-12	0	Inf	0
x1	-0.00456339287145751	0	-Inf	0
x2	0.066402697011615	0	Inf	0
x3	-0.410874962575317	0	-Inf	0
x4	1.37888046285497	0	Inf	0
x5	-2.36242772583948	0	-Inf	0
x6	1.96910229125777	0	Inf	0
x7	-0.622641940764061	0	-Inf	0
x8	3.13738365163391	0	Inf	0
x9	-4.2101243189536	0	-Inf	0
x10	2.74124266032682	0	Inf	0
x11	-0.684993856416449	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 489

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.6265936125817	500	0.0952531872251634		

Model	47.6265936125817	11	4.32969032841652	Inf	0
Residual	0	489	0		

A = 0.020432, B = 0.128153, C = 0.059953

order Q(x) = 7.000000, order D(x) = 4.000000

List of powers for Q(x): [0.148585, 0.336691, 0.524797, 0.712903, 0.901009, 1.089115, 1.277221]

List of powers for D(x): [0.148585, 0.336691, 0.524797, 0.712903]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.155031e-08

Corrected MSS of errors squared = 2.000573e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-6.09498399547333e-14	0	-Inf	0
x1	4.66444317192252	0	Inf	0
x2	-17.5169753003226	0	-Inf	0
x3	25.4157184629337	0	Inf	0
x4	-17.8881356933832	0	-Inf	0
x5	6.17578102427117	0	Inf	0
x6	-0.843586867472919	0	-Inf	0
x7	-97.9291676831983	0	-Inf	0
x8	549.693287612716	0	Inf	0
x9	-1248.56797079592	0	-Inf	0
x10	1431.65656467397	0	Inf	0
x11	-818.306115292348	0	-Inf	0
x12	170.168397152863	0	Inf	0
x13	14.2763946492685	0	Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	47.6265936125695	500	0.0952531872251391		
Model	47.6265936125695	13	3.66358412404381	Inf	0
Residual	0	487	0		

A = 0.552144, B = 0.183414, C = 0.250000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.735558, 0.668971, 0.602385, 0.535799, 0.469212, 0.402626]

List of powers for D(x): [0.735558, 0.668971, 0.602385, 0.535799, 0.469212, 0.402626, 0.336040]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.060105e-09

Corrected MSS of errors squared = 1.836156e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.61593454590514e-11	0	-Inf	0
x1	-0.00107618976802544	0	-Inf	0
x2	0.112562716002555	0	Inf	0
x3	-0.888145644273418	0	-Inf	0
x4	7.41892096518527	0	Inf	0
x5	-9.36552534972946	0	-Inf	0
x6	3.16934850914049	0	Inf	0
x7	-4.48442915541758	0	-Inf	0
x8	41.837965879641	0	Inf	0
x9	-124.801719170171	0	-Inf	0
x10	193.691512782958	0	Inf	0
x11	-169.243695229242	0	-Inf	0
x12	78.5258358138599	0	Inf	0
x13	-15.0555961615268	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	47.6265936125845	500	0.0952531872251691		
Model	47.6265936125845	13	3.66358412404496	Inf	0
Residual	0	487	0		

A = 0.229565, B = 0.115339, C = 0.261605

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.344904, 0.721848, 0.945547, 1.144034, 1.329470, 1.506565]

List of powers for D(x): [0.344904, 0.721848, 0.945547, 1.144034, 1.329470, 1.506565, 1.677736]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 1.906576e-10

Corrected MSS of errors squared = 3.302286e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.828495198642e-13	0	-Inf	0
x1	-0.149023965963171	0	-Inf	0
x2	1.49000326710088	0	Inf	0
x3	-4.97879577656448	0	-Inf	0
x4	7.53004738177976	0	Inf	0
x5	-5.33625035384771	0	-Inf	0
x6	1.44406395425349	0	Inf	0
x7	67.3795677506176	0	Inf	0
x8	-317.698477410629	0	-Inf	0
x9	467.210593603629	0	Inf	0
x10	0	0	NaN	NaN
x11	-616.684176599343	0	-Inf	0
x12	555.665310392174	0	Inf	0
x13	-154.872870628056	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	47.6265936125856	500	0.0952531872251713		
Model	47.6265936125856	12	3.9688828010488	Inf	0
Residual	0	488	0		

A = 0.028847, B = 0.100000, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.128847, 0.170268, 0.202052, 0.228847, 0.252453, 0.273796]

List of powers for D(x): [0.128847, 0.170268, 0.202052, 0.228847, 0.252453, 0.273796, 0.293422]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 3.971808e-09

Corrected MSS of errors squared = 6.879374e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	4.29086873281323e-11	0	Inf	0
x1	-2.90745588789456	0	-Inf	0
x2	30.8393070576553	0	Inf	0
x3	-107.798207270076	0	-Inf	0
x4	169.258530075151	0	Inf	0
x5	-123.893325869626	0	-Inf	0
x6	34.5026203532957	0	Inf	0
x7	730.115032591176	0	Inf	0
x8	-5003.25121012425	0	-Inf	0
x9	12798.7746751484	0	Inf	0
x10	-15677.9978335909	0	-Inf	0
x11	9325.59465856097	0	Inf	0
x12	-2172.23706769434	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	47.6265936135195	500	0.095253187227039		
Model	47.6265936135195	12	3.96888280112663	Inf	0
Residual	0	488	0		

A = 0.194365, B = 0.117978, C = 0.000000

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.312343, 0.361211, 0.398709, 0.430321, 0.458172, 0.483351]

List of powers for D(x): [0.312343, 0.361211, 0.398709, 0.430321, 0.458172, 0.483351]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 9.186803e-10

Corrected MSS of errors squared = 1.591201e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting sin(x) in range (0.000000, 1.570796)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.19973162008208e-15	0	Inf	0
x1	0	0	NaN	NaN
x2	-4.10789173517263e-08	0	-Inf	0
x3	1.07064958573034e-07	0	Inf	0
x4	-1.2611103132514e-07	0	-Inf	0
x5	9.19889585737795e-08	0	Inf	0
x6	-3.94688204233746e-08	0	-Inf	0
x7	7.60021450265251e-09	0	Inf	0
x8	1.00000052061521	0	Inf	0
x9	-6.54396335732349e-07	0	-Inf	0
x10	1.53543629390029e-07	0	Inf	0
x11	-1.97569975877017e-08	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 490

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	47.6265936125817	500	0.0952531872251634		
Model	47.6265936125817	10	4.76265936125817	Inf	0
Residual	0	490	0		

A = 0.000000, B = 0.413912

order Q(x) = 7.000000, order D(x) = 4.000000

List of powers for Q(x): [0.000000, 0.003399, 0.021450, 0.054383, 0.098797, 0.151762, 0.211123]

List of powers for D(x): [0.000000, 0.003399, 0.021450, 0.054383]

Fitting sin(x) in range (0.000000, 1.570796)

MSS of errors squared = 2.888362e-06

Corrected MSS of errors squared = 4.084760e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the sine function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*i+C*\sqrt{i-1}$	0.229	0.115	0.261	Inf	-Inf	1.90657600E-10	6, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.194	0.117	0	Inf	-Inf	9.18680300E-10	6, 6
$A+B*i-C*(i-1)$	0.552	0.183	0.25	Inf	-Inf	1.06010500E-09	6, 7
$A+B*\sqrt{i}+C$	0.589	0.1	0.974	Inf	-Inf	1.33152300E-09	7, 7
$A+B/i$	0.198	0.35		Inf	-Inf	1.54364300E-09	6, 7
$A+B*i$	0.418	0.111		Inf	-Inf	1.81073400E-09	6, 6
$A+B*\sqrt{i}+C*(i-1)$	0.028	0.1	0	Inf	-Inf	3.97180800E-09	6, 7
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	4.21995200E-09	6, 6
$A+B*i+C*(i-1)$	0.02	0.128	0.059	Inf	-Inf	1.15503100E-08	7, 4
$A+B*\log_{10}(i)^4$	0	0.413		Inf	-Inf	2.88836200E-06	7, 4

RESULTS FOR THE HYPERBOLIC SINE FUNCTION

Using Power $A+B*i$

Fitting $\sinh(x)$ in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-8.04819262586465e-09	0	-Inf	0
x1	0.0284259824278277	0	Inf	0
x2	0.90847364344029	0	Inf	0
x3	1.39968346705291	0	Inf	0
x4	-7.27624960196728	0	-Inf	0
x5	8.66784724901867	0	Inf	0
x6	-4.37894885464301	0	-Inf	0
x7	0.845244764048435	0	Inf	0
x8	9.12126046929743	0	Inf	0
x9	-21.651089380892	0	-Inf	0
x10	24.0352397972057	0	Inf	0
x11	-15.4626475156985	0	-Inf	0
x12	5.95696399905382	0	Inf	0
x13	-1.28556497112159	0	-Inf	0
x14	0.120353923082118	0	Inf	0

Number of observations: 501, Error degrees of freedom: 486

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	169774.232792751	500	339.548465585502		
Model	169774.232792751	14	12126.7309137679	Inf	0
Residual		0	486		0

A = 0.500000, B = 0.262131

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.762131, 1.024261, 1.286392, 1.548522, 1.810653, 2.072783, 2.334914]

List of powers for D(x): [0.762131, 1.024261, 1.286392, 1.548522, 1.810653, 2.072783, 2.334914]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 2.597451e-08

Corrected MSS of errors squared = 3.673351e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.92907107370706e-09	0	-Inf	0
x1	-0.238283703757263	0	-Inf	0
x2	18.221282662506	0	Inf	0
x3	-166.879689832552	0	-Inf	0
x4	493.68934100947	0	Inf	0
x5	-576.383585376355	0	-Inf	0
x6	231.60209218042	0	Inf	0
x7	-1.85162047695607	0	-Inf	0
x8	295.530152441026	0	Inf	0
x9	-3693.95614771385	0	-Inf	0
x10	12010.0324589673	0	Inf	0
x11	-12041.4850579571	0	-Inf	0
x12	0	0	NaN	NaN
x13	3432.72072104958	0	Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169774.23279267	500	339.54846558534		
Model	169774.23279267	12	14147.8527327225	Inf	0
Residual	0	488	0		

A = 0.420536, B = 0.502244

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.922780, 0.671658, 0.587950, 0.546097, 0.520985, 0.504243]

List of powers for D(x): [0.922780, 0.671658, 0.587950, 0.546097, 0.520985, 0.504243, 0.492285]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 2.138926e-06

Corrected MSS of errors squared = 3.024898e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	9.90037219459727e-10	0	Inf	0
x1	26.2408823790982	0	Inf	0
x2	-184.88336641205	0	-Inf	0
x3	349.749257577897	0	Inf	0
x4	0	0	NaN	NaN
x5	-638.613990360631	0	-Inf	0
x6	645.171191493701	0	Inf	0
x7	-197.653449585757	0	-Inf	0
x8	1488.59185353406	0	Inf	0
x9	-9015.49496798335	0	-Inf	0
x10	18821.1915929097	0	Inf	0
x11	-15043.5367155193	0	-Inf	0
x12	0	0	NaN	NaN
x13	5991.56920109444	0	Inf	0
x14	-2241.32992006868	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169774.232792852	500	339.548465585704		
Model	169774.232792852	12	14147.8527327377	Inf	0
Residual	0	488	0		

A = 0.362740, B = 0.125652

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.488392, 0.540439, 0.580376, 0.614045, 0.643707, 0.670524, 0.695185]

List of powers for D(x): [0.488392, 0.540439, 0.580376, 0.614045, 0.643707, 0.670524, 0.695185]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 2.372601e-06

Corrected MSS of errors squared = 3.355365e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.43301029734834e-10	0	Inf	0
x1	-0.514209690782541	0	-Inf	0
x2	3.72438497947748	0	Inf	0
x3	-7.52350460348438	0	-Inf	0
x4	0	0	NaN	NaN
x5	16.0715801517014	0	Inf	0
x6	-17.5848605327166	0	-Inf	0
x7	5.82696098236354	0	Inf	0
x8	218.443772020217	0	Inf	0
x9	-999.017075047189	0	-Inf	0
x10	1714.85602320179	0	Inf	0
x11	-1181.62595481345	0	-Inf	0
x12	0	0	NaN	NaN
x13	378.29228547282	0	Inf	0
x14	-129.949349749537	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 488

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169774.232792816	500	339.548465585632		
Model	169774.232792816	12	14147.8527327347	Inf	0
Residual	0	488	0		

A = 0.000000, B = 0.126605, C = 1.391221

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.195777, 0.233147, 0.265305, 0.293965, 0.320070, 0.344200, 0.366746]

List of powers for D(x): [0.195777, 0.233147, 0.265305, 0.293965, 0.320070, 0.344200, 0.366746]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 3.661159e-06

Corrected MSS of errors squared = 6.341313e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.35357358119317e-10	0	-Inf	0
x1	0.00166065911476189	0	Inf	0
x2	-0.0315496789069003	0	-Inf	0
x3	0.32504830068097	0	Inf	0
x4	-0.633767753187719	0	-Inf	0
x5	0.528787130217626	0	Inf	0
x6	-0.206056548138178	0	-Inf	0
x7	0.0309749826754234	0	Inf	0
x8	2.073439410235	0	Inf	0
x9	-2.03498639698789	0	-Inf	0
x10	1.46100883797489	0	Inf	0
x11	-0.665908502978266	0	-Inf	0

x12	0.172962141280727	0	Inf	0
x13	-0.0193618571810924	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169774.232792744	500	339.548465585487		
Model	169774.232792744	13	13059.5563686726	Inf	0
Residual	0	487	0		

A = 0.011295, B = 0.164589, C = 0.137780

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.175884, 0.478253, 0.780622, 1.082991, 1.385360, 1.687729, 1.990098]

List of powers for D(x): [0.175884, 0.478253, 0.780622, 1.082991, 1.385360, 1.687729]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 1.812801e-06

Corrected MSS of errors squared = 3.139863e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0712970143411907	0	-Inf	0
x1	-0.000803662258734649	0	-Inf	0
x2	0.00542547104391039	0	Inf	0
x3	-0.0180176074942059	0	-Inf	0
x4	0.0612303147052082	0	Inf	0
x5	0.0260555030768531	0	Inf	0
x6	-0.00280146477883772	0	-Inf	0
x7	0.000208571110969221	0	Inf	0
x8	-0.00341550337170804	0	-Inf	0
x9	0.0318197778277133	0	Inf	0
x10	-0.148967044866651	0	-Inf	0

x11	0.729326999581122	0	Inf	0
x12	0.456367163526859	0	Inf	0
x13	-0.0734192259578145	0	-Inf	0
x14	0.00828773875423669	0	Inf	0

Number of observations: 500, Error degrees of freedom: 485

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169558.095142868	499	339.795781849435		
Model	169558.095142868	14	12111.2925102049	Inf	0
Residual	0	485	0		

A = 0.227919, B = 0.121652, C = 0.223568

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.349571, 0.247654, 0.145738, 0.043821, -0.058096, -0.160013, -0.261929]

List of powers for D(x): [0.349571, 0.247654, 0.145738, 0.043821, -0.058096, -0.160013, -0.261929]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = NaN

Corrected MSS of errors squared = NaN

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-9.34839999611049e-12	0	-Inf	0
x1	-0.0086557594887174	0	-Inf	0
x2	0.183290890210494	0	Inf	0
x3	-1.0837724368987	0	-Inf	0
x4	2.86892609958523	0	Inf	0
x5	-3.71680281779312	0	-Inf	0
x6	2.31634717252426	0	Inf	0
x7	-0.558358844988467	0	-Inf	0

x8	14.9089562089785	0	Inf	0
x9	-41.9856693129729	0	-Inf	0
x10	53.5900110980648	0	Inf	0
x11	-36.8038144933824	0	-Inf	0
x12	13.2725457069987	0	Inf	0
x13	-1.98285824620336	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169774.232792734	500	339.548465585467		
Model	169774.232792734	13	13059.5563686718	Inf	0
Residual	0	487	0		

A = 0.125750, B = 0.118460, C = 0.000000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.244210, 0.362670, 0.481130, 0.599590, 0.718050, 0.836511, 0.954971]

List of powers for D(x): [0.244210, 0.362670, 0.481130, 0.599590, 0.718050, 0.836511]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 1.245598e-05

Corrected MSS of errors squared = 2.157439e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-4.44501828590081e-09	0	-Inf	0
x1	0.984184338042706	0	Inf	0
x2	-14.2277513878998	0	-Inf	0
x3	61.8570216062893	0	Inf	0
x4	-109.338636956053	0	-Inf	0
x5	85.4351817572556	0	Inf	0
x6	-24.6646438737787	0	-Inf	0

x7	67.4575336674504	0	Inf	0
x8	-364.828453683656	0	-Inf	0
x9	820.133859976664	0	Inf	0
x10	-950.806907744706	0	-Inf	0
x11	588.195454907141	0	Inf	0
x12	-178.347300746279	0	-Inf	0
x13	19.1572199457628	0	Inf	0

Number of observations: 501, Error degrees of freedom: 487

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169774.232793031	500	339.548465586062		
Model	169774.232793031	13	13059.5563686947	Inf	0
Residual	0	487	0		

A = 0.275328, B = 0.400000, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.675328, 0.841013, 0.968148, 1.075328, 1.169755, 1.255123]

List of powers for D(x): [0.675328, 0.841013, 0.968148, 1.075328, 1.169755, 1.255123, 1.333628]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 9.914760e-05

Corrected MSS of errors squared = 1.717287e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.46761488295846e-10	0	-Inf	0
x1	0.119995047481029	0	Inf	0
x2	-3.80435839872946	0	-Inf	0
x3	25.9162975172607	0	Inf	0
x4	-70.5925208536264	0	-Inf	0
x5	92.0300638087154	0	Inf	0

x6	-57.7631198644904	0	-Inf	0
x7	14.0785712757142	0	Inf	0
x8	22.58522328193	0	Inf	0
x9	-157.625548472424	0	-Inf	0
x10	451.254296507553	0	Inf	0
x11	-670.48067136442	0	-Inf	0
x12	551.941325483276	0	Inf	0
x13	-239.61015958184	0	-Inf	0
x14	42.9483587646335	0	Inf	0

Number of observations: 501, Error degrees of freedom: 486

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169774.232792983	500	339.548465585967		
Model	169774.232792983	14	12126.7309137845	Inf	0
Residual	0	486	0		

A = 0.000000, B = 0.435429, C = 0.070683

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.435429, 0.686473, 0.854146, 0.993285, 1.115015, 1.224632, 1.325175]

List of powers for D(x): [0.435429, 0.686473, 0.854146, 0.993285, 1.115015, 1.224632, 1.325175]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 1.624160e-06

Corrected MSS of errors squared = 2.813127e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting sinh(x) in range (0.000000, 5.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-7.30838063413437e-13	0	-Inf	0
x1	0.0086821829159018	0	Inf	0
x2	-0.0128203910903591	0	-Inf	0

x3	0.00713941145524896	0	Inf	0
x4	-0.00345192794628011	0	-Inf	0
x5	0.000766748251322857	0	Inf	0
x6	5.0456771773669	0	Inf	0
x7	-4.44501277745912	0	-Inf	0
x8	0.437489196468936	0	Inf	0
x9	-0.0413157623563869	0	-Inf	0
x10	0.00299934519047505	0	Inf	0
x11	-0.00010609139438882	0	-Inf	0

Number of observations: 501, Error degrees of freedom: 489

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	169774.232792744	500	339.548465585487		
Model	169774.232792744	11	15434.0211629767	Inf	0
Residual	0	489	0		

A = 0.065405, B = 3.500000

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.065405, 0.094147, 0.246783, 0.525267, 0.900820]

List of powers for D(x): [0.065405, 0.094147, 0.246783, 0.525267, 0.900820, 1.348693]

Fitting sinh(x) in range (0.000000, 5.000000)

MSS of errors squared = 2.687631e-06

Corrected MSS of errors squared = 3.800884e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the hyperbolic sine function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*i	0.5	0.262		Inf	-Inf	2.59745100E-08	7, 7
A+B*sqrt(i)+C*sqrt(i-1)	0	0.435	0.07	Inf	-Inf	1.62416000E-06	7, 7
A+B*i+C*(i-1)	0.011	0.164	0.137	Inf	-Inf	1.81280100E-06	7, 6
A+B/i	0.42	0.502		Inf	-Inf	2.13892600E-06	6, 7
A+B*sqrt(i)	0.362	0.125		Inf	-Inf	2.37260100E-06	7, 7
A+B*log10(i)^4	0.065	3.5		Inf	-Inf	2.68763100E-06	5, 6
A+B*sqrt(i+C)	0	0.126	1.391	Inf	-Inf	3.66115900E-06	7, 7
A+B*i+C*sqrt(i-1)	0.125	0.118	0	Inf	-Inf	1.24559800E-05	7, 6
A+B*sqrt(i)+C*(i-1)	0.275	0.4	0	Inf	-Inf	9.91476000E-05	6, 7
A+B*i-C*(i-1)	0.227	0.121	0.223	Inf	-Inf	NaN	7, 7

RESULTS FOR THE TANGENT FUNCTION

Using Power A+B* i

Fitting tan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	8.20139078885225e-14	0	Inf	0
x1	-0.00160323800075477	0	-Inf	0
x2	0.865474573582817	0	Inf	0
x3	0.515099050278576	0	Inf	0
x4	-2.93139588936604	0	-Inf	0
x5	6.22384588304203	0	Inf	0
x6	-3.83855672112883	0	-Inf	0
x7	0.479290390486426	0	Inf	0
x8	0.517216524295003	0	Inf	0
x9	1.48481617942837	0	Inf	0
x10	-10.0903358630299	0	-Inf	0
x11	16.3713197655651	0	Inf	0
x12	-14.6307704827087	0	-Inf	0
x13	8.66140517893234	0	Inf	0
x14	-2.15617572832214	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 86

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	18.4802838293816	100	0.184802838293816		
Model	18.4802838293816	14	1.32002027352726	Inf	0
Residual	0	86	0		

A = 0.236878, B = 0.372100

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.608979, 0.981079, 1.353179, 1.725280, 2.097380, 2.469481, 2.841581]

List of powers for D(x): [0.608979, 0.981079, 1.353179, 1.725280, 2.097380, 2.469481, 2.841581]

Fitting tan(x) in range (0.000000, 1.000000)

MSS of errors squared = 7.204837e-12

Corrected MSS of errors squared = 1.018918e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting tan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.1796544970202e-13	0	Inf	0
x1	-3.4771467654824e-05	0	-Inf	0
x2	7.0429316581369e-05	0	Inf	0
x3	-2.11087025795872e-05	0	-Inf	0
x4	0.00127729573237577	0	Inf	0
x5	-0.634407618152805	0	-Inf	0
x6	19.0746584436793	0	Inf	0
x7	-146.005635377334	0	-Inf	0
x8	421.130489872223	0	Inf	0
x9	-503.54501487484	0	-Inf	0
x10	210.978622916788	0	Inf	0

Number of observations: 101, Error degrees of freedom: 90

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	18.4802838293815	100	0.184802838293815		
Model	18.4802838293815	10	1.84802838293815	Inf	0
Residual	0	90	0		

A = 0.011415, B = 0.655309

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.666724, 0.339069, 0.229851]

List of powers for D(x): [0.666724, 0.339069, 0.229851, 0.175242, 0.142477, 0.120633, 0.105031]

Fitting tan(x) in range (0.000000, 1.000000)

MSS of errors squared = 1.529117e-08

Corrected MSS of errors squared = 2.162498e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}$

Fitting tan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.12672578893368e-13	0	Inf	0
x1	-0.0216947746086053	0	-Inf	0
x2	10.3095249160481	0	Inf	0
x3	-58.0062963082145	0	-Inf	0
x4	161.649512752597	0	Inf	0
x5	-227.196291364576	0	-Inf	0
x6	154.018676256358	0	Inf	0
x7	-40.3778768655806	0	-Inf	0
x8	9.31516641690371	0	Inf	0
x9	-18.5341085366465	0	-Inf	0
x10	2.34000269889559	0	Inf	0
x11	17.2804147964123	0	Inf	0
x12	-9.6426162196157	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	18.4802838293813	100	0.184802838293813		
Model	18.4802838293813	12	1.54002365244844	Inf	0
Residual	0	88	0		

A = 0.518948, B = 0.427171

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.946119, 1.123059, 1.258830, 1.373291, 1.474132, 1.565300, 1.649137]

List of powers for D(x): [0.946119, 1.123059, 1.258830, 1.373291, 1.474132]

Fitting tan(x) in range (0.000000, 1.000000)

MSS of errors squared = 1.187766e-10

Corrected MSS of errors squared = 1.679754e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting tan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	7.04895397972904e-13	0	Inf	0
x1	0.487189989402153	0	Inf	0
x2	2.23533238490827	0	Inf	0
x3	-6.42165694885142	0	-Inf	0
x4	10.213177618321	0	Inf	0
x5	-5.59117850809204	0	-Inf	0
x6	-9.85903044891283	0	-Inf	0
x7	51.9641054362468	0	Inf	0
x8	-148.02286046256	0	-Inf	0
x9	269.968913156246	0	Inf	0
x10	-302.936693314727	0	-Inf	0
x11	187.64383891795	0	Inf	0
x12	-48.3508377888153	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	18.4802838293769	100	0.184802838293769		
Model	18.4802838293769	12	1.54002365244807	Inf	0
Residual	0	88	0		

A = 0.020062, B = 0.684187, C = 0.871879

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.956144, 1.179528, 1.366344, 1.530223, 1.677982]

List of powers for D(x): [0.956144, 1.179528, 1.366344, 1.530223, 1.677982, 1.813609, 1.939677]

Fitting tan(x) in range (0.000000, 1.000000)

MSS of errors squared = 4.122283e-11

Corrected MSS of errors squared = 7.140003e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting tan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-6.19346407847808e-12	0	-Inf	0
x1	0.00217454679828553	0	Inf	0
x2	-0.0493837014153254	0	-Inf	0
x3	0.977529301718513	0	Inf	0
x4	-1.99780532215012	0	-Inf	0
x5	1.53996679466745	0	Inf	0
x6	-0.420644882542422	0	-Inf	0
x7	3.41061968534738	0	Inf	0
x8	-5.05470419571976	0	-Inf	0
x9	4.17206901773343	0	Inf	0
x10	-2.03228149164485	0	-Inf	0
x11	0.471012993877281	0	Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.4802838293816	100	0.184802838293816		

Model	18.4802838293816	11	1.68002580267105	Inf	0
Residual	0	89	0		

A = 0.161330, B = 0.188771, C = 0.104758

order Q(x) = 6.000000, order D(x) = 5.000000

List of powers for Q(x): [0.350102, 0.643631, 0.937160, 1.230689, 1.524218, 1.817748]

List of powers for D(x): [0.350102, 0.643631, 0.937160, 1.230689, 1.524218]

Fitting tan(x) in range (0.000000, 1.000000)

MSS of errors squared = 4.248200e-09

Corrected MSS of errors squared = 7.358099e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting tan(x) in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0509698314045562	0	-Inf	0
x1	-0.00276763390736128	0	-Inf	0
x2	0.0505164590206326	0	Inf	0
x3	0.00356843892344844	0	Inf	0
x4	-0.00031931455798812	0	-Inf	0
x5	-0.0279976007588578	0	-Inf	0
x6	0.924779647778547	0	Inf	0
x7	0.125481056994809	0	Inf	0
x8	-0.0222811584740714	0	-Inf	0

Number of observations: 100, Error degrees of freedom: 91

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.0954624389228	99	0.182782448878008		
Model	18.0954624389228	8	2.26193280486535	Inf	0
Residual	0	91	0		

A = 0.003729, B = 0.113453, C = 0.219083

order Q(x) = 4.000000, order D(x) = 4.000000

List of powers for Q(x): [0.117182, 0.011553, -0.094077, -0.199706]
 List of powers for D(x): [0.117182, 0.011553, -0.094077, -0.199706]
 Fitting tan(x) in range (0.000000, 1.000000)
 MSS of errors squared = NaN
 Corrected MSS of errors squared = NaN
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting tan(x) in range (0.000000, 1.000000)
 Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.12104934194455e-15	0	Inf	0
x1	9.89778910684726e-05	0	Inf	0
x2	-0.00260891245228195	0	-Inf	0
x3	-0.000935150502160888	0	-Inf	0
x4	0.0842050578845037	0	Inf	0
x5	-0.0683443027343324	0	-Inf	0
x6	1.88471853406097	0	Inf	0
x7	-4.13178239772796	0	-Inf	0
x8	8.34809574105368	0	Inf	0
x9	-8.99993968291739	0	-Inf	0
x10	5.04605864422932	0	Inf	0
x11	-1.15512284883046	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.4802838293817	100	0.184802838293817		
Model	18.4802838293817	11	1.68002580267106	Inf	0
Residual	0	89	0		

A = 0.001416, B = 0.100000, C = 0.198160

order Q(x) = 5.000000, order D(x) = 6.000000

List of powers for Q(x): [0.101416, 0.399576, 0.581657, 0.744640, 0.897737]

List of powers for D(x): [0.101416, 0.399576, 0.581657, 0.744640, 0.897737, 1.044516]

Fitting $\tan(x)$ in range (0.000000, 1.000000)
MSS of errors squared = 1.671391e-10
Corrected MSS of errors squared = 2.894934e-10
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting $\tan(x)$ in range (0.000000, 1.000000)
Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.94118842082977e-13	0	-Inf	0
x1	0.013778939707562	0	Inf	0
x2	-0.154075394479141	0	-Inf	0
x3	0.556135359452176	0	Inf	0
x4	-0.363453205539184	0	-Inf	0
x5	17.5091946037909	0	Inf	0
x6	-67.1349513822938	0	-Inf	0
x7	126.34975767652	0	Inf	0
x8	-138.195049661617	0	-Inf	0
x9	89.2147149377571	0	Inf	0
x10	-31.3504333945094	0	-Inf	0
x11	4.57313074906942	0	Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.4802838293821	100	0.184802838293821		
Model	18.4802838293821	11	1.6800258026711	Inf	0
Residual	0	89	0		

A = 0.296305, B = 0.100000, C = 0.132119

order Q(x) = 4.000000, order D(x) = 7.000000

List of powers for Q(x): [0.396305, 0.569846, 0.733749, 0.892664]

List of powers for D(x): [0.396305, 0.569846, 0.733749, 0.892664, 1.048390, 1.201851, 1.353597]

Fitting $\tan(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 6.379956e-10
 Corrected MSS of errors squared = 1.105041e-09
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting $\tan(x)$ in range (0.000000, 1.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.48473453936118e-13	0	Inf	0
x1	-0.000941288718083327	0	-Inf	0
x2	0.0321491981234499	0	Inf	0
x3	-0.142517463771996	0	-Inf	0
x4	0.210088512778042	0	Inf	0
x5	-0.0978793509057569	0	-Inf	0
x6	4.56917286198682	0	Inf	0
x7	-22.9606314864349	0	-Inf	0
x8	39.3971401239005	0	Inf	0
x9	0	0	NaN	NaN
x10	-62.367409277686	0	-Inf	0
x11	60.2214702713097	0	Inf	0
x12	-17.8603201244095	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.4802838293805	100	0.184802838293805		
Model	18.4802838293805	11	1.68002580267096	Inf	0
Residual	0	89	0		

A = 0.007299, B = 0.100000, C = 0.111769

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.107299, 0.260490, 0.338570, 0.400889, 0.454444]

List of powers for D(x): [0.107299, 0.260490, 0.338570, 0.400889, 0.454444, 0.502172, 0.545652]

Fitting $\tan(x)$ in range (0.000000, 1.000000)

MSS of errors squared = 1.981809e-10
 Corrected MSS of errors squared = 3.432594e-10
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting $\tan(x)$ in range (0.000000, 1.000000)
 Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.71092456684199e-15	0	-Inf	0
x1	-9.63985781532556e-05	0	-Inf	0
x2	0.000158945665998394	0	Inf	0
x3	-0.000168501161791305	0	-Inf	0
x4	0.00051858278670031	0	Inf	0
x5	-0.000254802326656989	0	-Inf	0
x6	-1.79876482061812e-05	0	-Inf	0
x7	4.4558671537584e-07	0	Inf	0
x8	2.25164473815963	0	Inf	0
x9	-1.34503999181382	0	-Inf	0
x10	0.105695856853265	0	Inf	0
x11	-0.0147107879773169	0	-Inf	0
x12	0.00266082896962138	0	Inf	0
x13	-0.000340719720494663	0	-Inf	0

Number of observations: 101, Error degrees of freedom: 87

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	18.4802838293815	100	0.184802838293815		
Model	18.4802838293815	13	1.42156029456781	Inf	0
Residual	0	87	0		

A = 0.024248, B = 3.438275

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.024248, 0.052482, 0.202426, 0.475999, 0.844929, 1.284904, 1.778003]

List of powers for D(x): [0.024248, 0.052482, 0.202426, 0.475999, 0.844929, 1.284904]

```
Fitting tan(x) in range (0.000000, 1.000000)
MSS of errors squared = 5.033570e-11
Corrected MSS of errors squared = 7.118543e-11
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Comments

The following table shows the summary results for the ten models that fit the tangent function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
A+B*i	0.236	0.372		Inf	-Inf	7.20483700E-12	7, 7
A+B*sqrt(i+C)	0.02	0.684	0.871	Inf	-Inf	4.12228300E-11	5, 7
A+B*log10(i)^4	0.024	3.438		Inf	-Inf	5.03357000E-11	7, 6
A+B*sqrt(i)	0.518	0.427		Inf	-Inf	1.18776600E-10	7, 5
A+B*i+C*sqrt(i-1)	0.001	0.1	0.198	Inf	-Inf	1.67139100E-10	5, 6
A+B*sqrt(i)+C*sqrt(i-1)	0.007	0.1	0.111	Inf	-Inf	1.98180900E-10	5, 7
A+B*sqrt(i)+C*(i-1)	0.296	0.1	0.132	Inf	-Inf	6.37995600E-10	4, 7
A+B*i+C*(i-1)	0.161	0.188	0.104	Inf	-Inf	4.24820000E-09	6, 5
A+B/i	0.011	0.655		Inf	-Inf	1.52911700E-08	3, 7
A+B*i-C*(i-1)	0.003	0.113	0.219	Inf	-Inf	NaN	4, 4

RESULTS FOR THE HYPERBOLIC TANGENT FUNCTION

Using Power A+B* i

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is A+B*i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.55393703836974e-10	0	-Inf	0
x1	0.0187119772864276	0	Inf	0
x2	-0.224516735181885	0	-Inf	0
x3	3.20913885974697	0	Inf	0
x4	-7.05250653519415	0	-Inf	0
x5	6.90644914143744	0	Inf	0
x6	-3.40772616665961	0	-Inf	0
x7	0.739820886101112	0	Inf	0
x8	10.7176123994735	0	Inf	0
x9	-29.731584232802	0	-Inf	0
x10	40.0229808115127	0	Inf	0
x11	-31.2676996721796	0	-Inf	0
x12	13.6352939049698	0	Inf	0
x13	-2.62525458219081	0	-Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	23.0506117213929	300	0.0768353724046429		
Model	23.0506117213929	13	1.77312397856868	Inf	0
Residual	0	287	0		

A = 0.366685, B = 0.231426

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.598111, 0.829536, 1.060962, 1.292387, 1.523813, 1.755239, 1.986664]

List of powers for D(x): [0.598111, 0.829536, 1.060962, 1.292387, 1.523813, 1.755239]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 1.217599e-09

Corrected MSS of errors squared = 1.721944e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.14033575722486e-13	0	Inf	0
x1	-0.0222415048434651	0	-Inf	0
x2	1.22746370267971	0	Inf	0
x3	-14.1383456755636	0	-Inf	0
x4	60.7510720079822	0	Inf	0
x5	-117.83331274199	0	-Inf	0
x6	104.864753327026	0	Inf	0
x7	-34.850640446927	0	-Inf	0
x8	0.0316281083250381	0	Inf	0
x9	-2.06796624357823	0	-Inf	0
x10	16.7906906682754	0	Inf	0
x11	-36.8612442552049	0	-Inf	0
x12	23.1085347662824	0	Inf	0

Number of observations: 301, Error degrees of freedom: 288

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	23.0506117213935	300	0.076835372404645		
Model	23.0506117213935	12	1.92088431011612	Inf	0
Residual	0	288	0		

A = 0.000000, B = 1.020916

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [1.020916, 0.510458, 0.340305, 0.255229, 0.204183, 0.170153, 0.145845]

List of powers for D(x): [1.020916, 0.510458, 0.340305, 0.255229, 0.204183]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 1.764994e-05

Corrected MSS of errors squared = 2.496079e-05

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.63810091736571e-12	0	-Inf	0
x1	-8.02264501203594	0	-Inf	0
x2	101.087979486882	0	Inf	0
x3	-437.47347754677	0	-Inf	0
x4	909.052075422495	0	Inf	0
x5	-993.108282840593	0	-Inf	0
x6	551.650801893118	0	Inf	0
x7	-123.186461091962	0	-Inf	0
x8	31.7671690840701	0	Inf	0
x9	-132.603565863345	0	-Inf	0
x10	217.098173122209	0	Inf	0
x11	-158.457715126175	0	-Inf	0
x12	43.1959515050533	0	Inf	0

Number of observations: 301, Error degrees of freedom: 288

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	23.0506117213848	300	0.0768353724046161		
Model	23.0506117213848	12	1.9208843101154	Inf	0
Residual	0	288	0		

A = 0.006404, B = 0.100000

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.106404, 0.147826, 0.179609, 0.206404, 0.230011, 0.251353, 0.270979]

List of powers for D(x): [0.106404, 0.147826, 0.179609, 0.206404, 0.230011]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 1.784839e-06

Corrected MSS of errors squared = 2.524144e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i+C}$

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i+C}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.17569843946465e-10	0	-Inf	0
x1	15.6491149567213	0	Inf	0
x2	-108.213269876474	0	-Inf	0
x3	244.586338875789	0	Inf	0
x4	-227.249330145755	0	-Inf	0
x5	75.2368911709601	0	Inf	0
x6	4234.85393990131	0	Inf	0
x7	-24499.4440524054	0	-Inf	0
x8	40550.2997095616	0	Inf	0
x9	0	0	NaN	NaN
x10	-61870.6794560913	0	-Inf	0
x11	58622.4518173023	0	Inf	0
x12	-17036.494753773	0	-Inf	0

Number of observations: 301, Error degrees of freedom: 289

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	23.050611719609	300	0.0768353723986967		
Model	23.050611719609	11	2.09551015632809	Inf	0
Residual	0	289	0		

A = 0.378258, B = 0.100000, C = 0.000000

order Q(x) = 5.000000, order D(x) = 7.000000

List of powers for Q(x): [0.478258, 0.519679, 0.551463, 0.578258, 0.601865]

List of powers for D(x): [0.478258, 0.519679, 0.551463, 0.578258, 0.601865, 0.623207, 0.642833]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 4.569463e-08

Corrected MSS of errors squared = 7.914543e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.51319897410502e-11	0	Inf	0
x1	-0.00252904290876793	0	-Inf	0
x2	0.120454583932219	0	Inf	0
x3	1.56751378050533	0	Inf	0
x4	-3.38312654608382	0	-Inf	0
x5	3.13485206541446	0	Inf	0
x6	-1.44609430717249	0	-Inf	0
x7	0.311149364426396	0	Inf	0
x8	1.35832949794871	0	Inf	0
x9	0.255332368643138	0	Inf	0
x10	-2.80537665423042	0	-Inf	0
x11	3.52802130965272	0	Inf	0
x12	-2.65763887993998	0	-Inf	0
x13	1.19877037860623	0	Inf	0
x14	-0.274263413382735	0	-Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
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Total	23.0506117213926	300	0.0768353724046421		
Model	23.0506117213926	14	1.64647226581376	Inf	0
Residual	0	286	0		

A = 0.321582, B = 0.128515, C = 0.209872

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.450097, 0.788485, 1.126872, 1.465260, 1.803648, 2.142036, 2.480423]

List of powers for D(x): [0.450097, 0.788485, 1.126872, 1.465260, 1.803648, 2.142036, 2.480423]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 2.387061e-10

Corrected MSS of errors squared = 4.134511e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-3.71558741749303e-12	0	-Inf	0
x1	0.00583905824111335	0	Inf	0
x2	-0.080937855442929	0	-Inf	0
x3	0.455286131404723	0	Inf	0
x4	-0.750828892936743	0	-Inf	0
x5	0.676858256435486	0	Inf	0
x6	-0.375529281463885	0	-Inf	0
x7	0.106863903391618	0	Inf	0
x8	4.8838627316033	0	Inf	0
x9	-10.6686593517311	0	-Inf	0
x10	13.3030745541947	0	Inf	0
x11	-10.028654009158	0	-Inf	0
x12	4.26346716481907	0	Inf	0
x13	-0.802397296235352	0	-Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	23.0506117213923	300	0.0768353724046411		
Model	23.0506117213923	13	1.77312397856864	Inf	0
Residual	0	287	0		

A = 0.037006, B = 0.226127, C = 0.000000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.263132, 0.489259, 0.715386, 0.941512, 1.167639, 1.393766, 1.619892]

List of powers for D(x): [0.263132, 0.489259, 0.715386, 0.941512, 1.167639, 1.393766]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 3.114889e-09

Corrected MSS of errors squared = 5.395146e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.00169947988069e-12	0	-Inf	0
x1	0.00300338939757576	0	Inf	0
x2	-0.0475045681052305	0	-Inf	0
x3	0.318818027168672	0	Inf	0
x4	-0.462291628313052	0	-Inf	0
x5	0.255025732477475	0	Inf	0
x6	-0.0430645844652006	0	-Inf	0
x7	4.08790518783641	0	Inf	0
x8	-8.37672030400205	0	-Inf	0
x9	10.866942884504	0	Inf	0
x10	-9.26966654174938	0	-Inf	0
x11	4.98467383887496	0	Inf	0
x12	-1.52201136166139	0	-Inf	0
x13	0.19738134887843	0	Inf	0

Number of observations: 301, Error degrees of freedom: 287

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	23.0506117213922	300	0.0768353724046407		
Model	23.0506117213922	13	1.77312397856863	Inf	0
Residual	0	287	0		

A = 0.000000, B = 0.242413, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.242413, 0.484826, 0.727238, 0.969651, 1.212064, 1.454477]

List of powers for D(x): [0.242413, 0.484826, 0.727238, 0.969651, 1.212064, 1.454477, 1.696890]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 1.792873e-09

Corrected MSS of errors squared = 3.105346e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*(i-1)$

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	3.52625594315292e-10	0	Inf	0
x1	-0.00283150213087954	0	-Inf	0
x2	0.401680740556144	0	Inf	0
x3	2.97596752411273	0	Inf	0
x4	-5.14158353121837	0	-Inf	0
x5	4.46680084042503	0	Inf	0
x6	-1.97537139049346	0	-Inf	0
x7	0.478616137718833	0	Inf	0
x8	-2.27516425991538	0	-Inf	0
x9	6.50978996175315	0	Inf	0
x10	-8.608955250451	0	-Inf	0
x11	6.8086348574532	0	Inf	0
x12	-4.40698514120293	0	-Inf	0
x13	1.85535209822199	0	Inf	0
x14	-0.462619821262156	0	-Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	23.0506117213934	300	0.0768353724046446		
Model	23.0506117213934	14	1.64647226581381	Inf	0
Residual	0	286	0		

A = 0.340713, B = 0.100000, C = 0.400000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.440713, 0.882134, 1.313918, 1.740713, 2.164320, 2.585662, 3.005288]

List of powers for D(x): [0.440713, 0.882134, 1.313918, 1.740713, 2.164320, 2.585662, 3.005288]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 5.073413e-10

Corrected MSS of errors squared = 8.787410e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}+C*\sqrt{i-1}$

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-2.22521640797171e-12	0	-Inf	0
x1	0.00111235923101414	0	Inf	0
x2	-0.192733543820148	0	-Inf	0
x3	2.33859003149195	0	Inf	0
x4	-5.24160456627489	0	-Inf	0
x5	4.170675306193	0	Inf	0
x6	-0.814311473600872	0	-Inf	0
x7	-0.231314668208109	0	-Inf	0
x8	3.65424623530469	0	Inf	0
x9	-27.2544320478478	0	-Inf	0
x10	97.7725681979958	0	Inf	0
x11	-174.910933088833	0	-Inf	0
x12	169.318600068975	0	Inf	0

x13	-85.0642524937779	0	-Inf	0
x14	17.4442692028141	0	Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	23.0506117213876	300	0.0768353724046252		
Model	23.0506117213876	14	1.6464722658134	Inf	0
Residual	0	286	0		

A = 0.120496, B = 0.176854, C = 0.358539

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.297350, 0.729144, 0.933866, 1.095211, 1.233031, 1.355415, 1.466644]

List of powers for D(x): [0.297350, 0.729144, 0.933866, 1.095211, 1.233031, 1.355415, 1.466644]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 9.548312e-10

Corrected MSS of errors squared = 1.653816e-09

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting tanh(x) in range (0.000000, 3.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.57042618211036e-10	0	Inf	0
x1	190.43965418143	0	Inf	0
x2	-253.48692598407	0	-Inf	0
x3	93.1770992249972	0	Inf	0
x4	-44.1493300412124	0	-Inf	0
x5	19.0873680345295	0	Inf	0
x6	-5.36923893582942	0	-Inf	0
x7	0.766586606086981	0	Inf	0
x8	1110.72344592179	0	Inf	0
x9	-1414.00850346384	0	-Inf	0

x10	369.178544466581	0	Inf	0
x11	-72.8532490481596	0	-Inf	0
x12	6.56992922873773	0	Inf	0
x13	1.21516562987998	0	Inf	0
x14	-0.436173934161315	0	-Inf	0

Number of observations: 301, Error degrees of freedom: 286

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	23.0506117214359	300	0.0768353724047863		
Model	23.0506117214359	14	1.64647226581685	Inf	0
Residual	0	286	0		

A = 1.203935, B = 2.409930

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [1.203935, 1.223725, 1.328823, 1.520573, 1.779161, 2.087545, 2.433164]

List of powers for D(x): [1.203935, 1.223725, 1.328823, 1.520573, 1.779161, 2.087545, 2.433164]

Fitting tanh(x) in range (0.000000, 3.000000)

MSS of errors squared = 1.548673e-10

Corrected MSS of errors squared = 2.190154e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the hyperbolic tangent function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*\log_{10}(i)^4$	1.203	2.409		Inf	-Inf	1.54867300E-10	7, 7
$A+B*i+C*(i-1)$	0.321	0.128	0.209	Inf	-Inf	2.38706100E-10	7, 7
$A+B*\sqrt{i}+C*(i-1)$	0.34	0.1	0.4	Inf	-Inf	5.07341300E-10	7, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.12	0.176	0.358	Inf	-Inf	9.54831200E-10	7, 7
$A+B*i$	0.366	0.231		Inf	-Inf	1.21759900E-09	7, 6
$A+B*i+C*\sqrt{i-1}$	0	0.242	0	Inf	-Inf	1.79287300E-09	6, 7
$A+B*i-C*(i-1)$	0.037	0.226	0	Inf	-Inf	3.11488900E-09	7, 6
$A+B*\sqrt{i+C}$	0.378	0.1	0	Inf	-Inf	4.56946300E-08	5, 7
$A+B*\sqrt{i}$	0.006	0.1		Inf	-Inf	1.78483900E-06	7, 5
$A+B/i$	0	1.02		Inf	-Inf	1.76499400E-05	7, 5

RESULTS FOR THE TWO-SIDED T-INVERSE DISTRIBUTION

Using Power $A+B*i$

Fitting $\text{tinv}(0.975,x)$ in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.836497504355684	0	Inf	0
x1	-9.15699341578339	0	-Inf	0
x2	25.6490761939587	0	Inf	0
x3	-33.0894707896416	0	-Inf	0
x4	21.2338053201022	0	Inf	0
x5	-5.3774556481826	0	-Inf	0
x6	-0.161226271598928	0	-Inf	0
x7	0.010956970870844	0	Inf	0
x8	7.58201937901558	0	Inf	0
x9	-18.0131164510103	0	-Inf	0
x10	20.8717895021151	0	Inf	0
x11	-12.7044528791837	0	-Inf	0
x12	3.26806595244053	0	Inf	0

Number of observations: 981, Error degrees of freedom: 968

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.9047757848054	980	0.0488824242702096		

Model	47.9047757848054	12	3.99206464873378	Inf	0
Residual	0	968	0		

A = 0.057896, B = 0.144208

order Q(x) = 7.000000, order D(x) = 5.000000

List of powers for Q(x): [0.202105, 0.346313, 0.490521, 0.634729, 0.778937, 0.923145, 1.067353]

List of powers for D(x): [0.202105, 0.346313, 0.490521, 0.634729, 0.778937]

Fitting tinv(0.975,x) in range (2.000000, 100.000000)

MSS of errors squared = 6.961561e-11

Corrected MSS of errors squared = 9.845133e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting tinv(0.975,x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.302474656620125	0	-Inf	0
x1	-0.960122584562905	0	-Inf	0
x2	5.33026275732446	0	Inf	0
x3	-18.5917972764921	0	-Inf	0
x4	20.545484991427	0	Inf	0
x5	0	0	NaN	NaN
x6	-6.69575643749424	0	-Inf	0
x7	0.489864767584984	0	Inf	0
x8	-2.6949638477266	0	-Inf	0
x9	7.0103083967293	0	Inf	0
x10	5.48902082367459	0	Inf	0
x11	-37.7509372772111	0	-Inf	0
x12	28.5096862183103	0	Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.9047757848324	980	0.0488824242702372		

Model	47.9047757848324	11	4.35497961680295	Inf	0
Residual	0	969	0		

A = 0.099802, B = 1.522363

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [1.622165, 0.860984, 0.607256, 0.480393, 0.404275, 0.353529]

List of powers for D(x): [1.622165, 0.860984, 0.607256, 0.480393, 0.404275, 0.353529]

Fitting tinv(0.975,x) in range (2.000000, 100.000000)

MSS of errors squared = 9.408634e-11

Corrected MSS of errors squared = 1.330582e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\sqrt{i}$

Fitting tinv(0.975,x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*\sqrt{i}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.198103439484202	0	Inf	0
x1	-1.48599806602108	0	-Inf	0
x2	2.34519912004544	0	Inf	0
x3	-1.05664626464927	0	-Inf	0
x4	27.541076343713	0	Inf	0
x5	-119.320340469076	0	-Inf	0
x6	204.712649275844	0	Inf	0
x7	-142.286128270504	0	-Inf	0
x8	0	0	NaN	NaN
x9	46.3865144258303	0	Inf	0
x10	-16.0338252030629	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 971

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.9047757848133	980	0.0488824242702176		
Model	47.9047757848133	9	5.32275286497925	Inf	0
Residual	0	971	0		

```

A = 0.000000, B = 0.100000
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.141421, 0.173205]
List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949,
0.264575]
Fitting tinv(0.975,x) in range (2.000000, 100.000000)
MSS of errors squared = 5.366981e-06
Corrected MSS of errors squared = 7.590058e-06
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i+C}$

```

Fitting tinv(0.975,x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*sqrt(i+C)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.207387501105041	0	Inf	0
x1	-2.63928052734731	0	-Inf	0
x2	4.42448078940577	0	Inf	0
x3	-1.99062577855022	0	-Inf	0
x4	77.4311123838739	0	Inf	0
x5	-343.978338181687	0	-Inf	0
x6	583.651950273145	0	Inf	0
x7	-399.355994449579	0	-Inf	0
x8	0	0	NaN	NaN
x9	126.397472806535	0	Inf	0
x10	-43.1463654218147	0	-Inf	0

Number of observations: 99, Error degrees of freedom: 89

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	7.69602630400818	98	0.0785308806531447		
Model	7.69602630400818	9	0.855114033778687	Inf	0
Residual	0	89	0		

```

A = 0.010881, B = 0.100000, C = 0.558449
order Q(x) = 3.000000, order D(x) = 7.000000

```

List of powers for Q(x): [0.135719, 0.170833, 0.199520]
 List of powers for D(x): [0.135719, 0.170833, 0.199520, 0.224386, 0.246645, 0.266976, 0.285807]
 Fitting $\text{tinv}(0.975, x)$ in range (2.000000, 100.000000)
 MSS of errors squared = 2.061111e-08
 Corrected MSS of errors squared = 3.569950e-08
 R-Squared = 1.00000000
 R-Squared Adjusted = 1.00000000
 Particle swarm AICc = -1.000000e+99
 AIC = -Inf
 AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting $\text{tinv}(0.975, x)$ in range (2.000000, 100.000000)
 Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.721687486166002	0	-Inf	0
x1	0.576948574231336	0	Inf	0
x2	0	0	NaN	NaN
x3	-1.23565436370282	0	-Inf	0
x4	2.57152363539223	0	Inf	0
x5	-2.1395216525292	0	-Inf	0
x6	0	0	NaN	NaN
x7	0.203502587605134	0	Inf	0
x8	1.9008101878085	0	Inf	0
x9	-0.983116464416583	0	-Inf	0
x10	0.259752246755933	0	Inf	0
x11	-1.11278084794401	0	-Inf	0
x12	1.09782947520495	0	Inf	0
x13	-0.000157584228557745	0	-Inf	0
x14	-0.103827018526969	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 968

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.9047757847729	980	0.0488824242701765		
Model	47.9047757847729	12	3.99206464873108	Inf	0
Residual	0	968	0		

```

A = 0.000000, B = 0.200395, C = 0.163420
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.200395, 0.564209, 0.928024, 1.291839, 1.655654, 2.019468,
2.383283]
List of powers for D(x): [0.200395, 0.564209, 0.928024, 1.291839, 1.655654, 2.019468,
2.383283]
Fitting tinv(0.975,x) in range (2.000000, 100.000000)
MSS of errors squared = 4.185507e-10
Corrected MSS of errors squared = 7.249511e-10
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i-C*(i-1)$

```

Fitting tinv(0.975,x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*i-C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.19092988295853	0	Inf	0
x1	-0.0555523811810384	0	-Inf	0
x2	0.317961186519862	0	Inf	0
x3	-1.18885711872794	0	-Inf	0
x4	-1.38996401943944	0	-Inf	0
x5	0.136732787270005	0	Inf	0
x6	-0.0113565353158571	0	-Inf	0
x7	0.0281306680696109	0	Inf	0
x8	-0.158864536047914	0	-Inf	0
x9	0.572129830613047	0	Inf	0
x10	0.600281866730981	0	Inf	0
x11	-0.0416694907275086	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.9047757848114	980	0.0488824242702157		
Model	47.9047757848114	11	4.35497961680104	Inf	0
Residual	0	969	0		

```

A = 0.197684, B = 0.122641, C = 0.244915
order Q(x) = 6.000000, order D(x) = 5.000000
List of powers for Q(x): [0.320325, 0.198052, 0.075778, -0.046496, -0.168770, -
0.291044]
List of powers for D(x): [0.320325, 0.198052, 0.075778, -0.046496, -0.168770]
Fitting tinv(0.975,x) in range (2.000000, 100.000000)
MSS of errors squared = 9.936866e-11
Corrected MSS of errors squared = 1.721116e-10
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*\sqrt{i-1}$

```

Fitting tinv(0.975,x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*i+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-1.43408795429111	0	-Inf	0
x1	2.18677891927342	0	Inf	0
x2	-1.69317365879945	0	-Inf	0
x3	0	0	NaN	NaN
x4	-2.07429209175379	0	-Inf	0
x5	6.74606688269603	0	Inf	0
x6	-4.84331807476932	0	-Inf	0
x7	0.000335124279612671	0	Inf	0
x8	2.368981921178	0	Inf	0
x9	-1.83935271441315	0	-Inf	0
x10	0.603243823940772	0	Inf	0
x11	0.89883702598872	0	Inf	0
x12	-3.41211217831933	0	-Inf	0
x13	2.46776508012289	0	Inf	0

Number of observations: 981, Error degrees of freedom: 968

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.9047757848422	980	0.0488824242702471		
Model	47.9047757848422	12	3.99206464873685	Inf	0
Residual	0	968	0		

```

A = 0.059398, B = 0.184463, C = 0.136013
order Q(x) = 7.000000, order D(x) = 6.000000
List of powers for Q(x): [0.243861, 0.564337, 0.805139, 1.032832, 1.253740, 1.470311,
1.683802]
List of powers for D(x): [0.243861, 0.564337, 0.805139, 1.032832, 1.253740, 1.470311]
Fitting tinv(0.975,x) in range (2.000000, 100.000000)
MSS of errors squared = 5.750159e-11
Corrected MSS of errors squared = 9.959567e-11
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}+C*(i-1)$

```

Fitting tinv(0.975,x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.15693926683385	0	Inf	0
x1	-144.834385128895	0	-Inf	0
x2	452.827848038334	0	Inf	0
x3	0	0	NaN	NaN
x4	-1314.3731963174	0	-Inf	0
x5	1536.95521592652	0	Inf	0
x6	-531.752350628014	0	-Inf	0
x7	132.413134952648	0	Inf	0
x8	-578.055268993054	0	-Inf	0
x9	849.69330624786	0	Inf	0
x10	-424.001986076546	0	-Inf	0
x11	0	0	NaN	NaN
x12	-25.447630413489	0	-Inf	0
x13	46.4000072322429	0	Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.904775785512	980	0.0488824242709306		
Model	47.904775785512	11	4.35497961686473	Inf	0

```

Residual          0      969          0

A = 0.144106, B = 0.147557, C = 0.000000
order Q(x) = 6.000000, order D(x) = 7.000000
List of powers for Q(x): [0.291663, 0.352783, 0.399682, 0.439219, 0.474053, 0.505545]
List of powers for D(x): [0.291663, 0.352783, 0.399682, 0.439219, 0.474053, 0.505545,
0.534504]
Fitting tinv(0.975,x) in range (2.000000, 100.000000)
MSS of errors squared = 2.007892e-10
Corrected MSS of errors squared = 3.477771e-10
R-Squared =      1.00000000
R-Squared Adjusted =      1.00000000
Particle swarm AICc =  -1.000000e+99
AIC =  -Inf
AICc =  -Inf

```

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

Fitting $\text{tinv}(0.975,x)$ in range (2.000000, 100.000000)
Pade-Shammas polynomial power is $A+B\sqrt{i}+C\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.793428779332711	0	-Inf	0
x1	0.704520083313841	0	Inf	0
x2	1.06768174945043	0	Inf	0
x3	0	0	NaN	NaN
x4	-15.7832109384656	0	-Inf	0
x5	29.3297803756985	0	Inf	0
x6	-15.2037451744517	0	-Inf	0
x7	0.0677444588008329	0	Inf	0
x8	2.07698876619554	0	Inf	0
x9	-3.55828476474727	0	-Inf	0
x10	3.27887406484451	0	Inf	0
x11	5.62971344506776	0	Inf	0
x12	-13.8375855973866	0	-Inf	0
x13	7.45827786928696	0	Inf	0

Number of observations: 981, Error degrees of freedom: 968

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	47.9047757847998	980	0.0488824242702039		

Model	47.9047757847998	12	3.99206464873332	Inf	0
Residual	0	968	0		

A = 0.087769, B = 0.106187, C = 0.395489

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.193956, 0.633429, 0.830996, 0.985150, 1.116188, 1.232213, 1.337460]

List of powers for D(x): [0.193956, 0.633429, 0.830996, 0.985150, 1.116188, 1.232213]

Fitting tinv(0.975,x) in range (2.000000, 100.000000)

MSS of errors squared = 6.242375e-11

Corrected MSS of errors squared = 1.081211e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*log10(i)^4

Fitting tinv(0.975,x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is A+B*log10(i)^4

Linear regression model:

y ~ 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12 + x13 + x14

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	-1.55152885716826e-06	0	-Inf	0
x2	0	0	NaN	NaN
x3	0	0	NaN	NaN
x4	3.25847018574496e-06	0	Inf	0
x5	0	0	NaN	NaN
x6	-3.08743566080988e-06	0	-Inf	0
x7	1.3804755327399e-06	0	Inf	0
x8	0.99999683143781	0	Inf	0
x9	0	0	NaN	NaN
x10	5.77507938487138e-06	0	Inf	0
x11	0	0	NaN	NaN
x12	-6.96545845073554e-06	0	-Inf	0
x13	5.99666628301916e-06	0	Inf	0
x14	-1.63772418642872e-06	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 972

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	47.9047757848119	980	0.0488824242702162		
Model	47.9047757848119	8	5.98809697310149	Inf	0
Residual		0	972		0

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.000000, 0.000821, 0.005182, 0.013139, 0.023869, 0.036665, 0.051007]

List of powers for D(x): [0.000000, 0.000821, 0.005182, 0.013139, 0.023869, 0.036665, 0.051007]

Fitting tinv(0.975,x) in range (2.000000, 100.000000)

MSS of errors squared = 1.338433e-04

Corrected MSS of errors squared = 1.892830e-04

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Comments

The following table shows the summary results for the ten models that fit the two-sided t-inverse function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B*i+C*\sqrt{i-1}$	0.059	0.184	0.136	Inf	-Inf	5.75015900E-11	7, 6
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.087	0.106	0.395	Inf	-Inf	6.24237500E-11	7, 6
$A+B*i$	0.057	0.144		Inf	-Inf	6.96156100E-11	7, 5
$A+B/i$	0.099	1.522		Inf	-Inf	9.40863400E-11	6, 6
$A+B*i-C*(i-1)$	0.197	0.122	0.244	Inf	-Inf	9.93686600E-11	6, 5
$A+B*\sqrt{i}+C*(i-1)$	0.144	0.147	0	Inf	-Inf	2.00789200E-10	6, 7
$A+B*i+C*(i-1)$	0	0.2	0.163	Inf	-Inf	4.18550700E-10	7, 7
$A+B*\sqrt{i+C}$	0.01	0.1	0.558	Inf	-Inf	2.06111100E-08	3, 7
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	5.36698100E-06	3, 7
$A+B*\log_{10}(i)^4$	0	0.1		Inf	-Inf	1.33843300E-04	7, 7

RESULTS FOR THE ONE-SIDED T-INVERSE DISTRIBUTION

Using Power $A+B*i$

Fitting `tnv(0.95,x)` in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.52877196108537	0	Inf	0
x1	-7.40882958015674	0	-Inf	0
x2	14.7726130833848	0	Inf	0
x3	-15.2009618027443	0	-Inf	0
x4	8.09315588371896	0	Inf	0
x5	-1.7854107808168	0	-Inf	0
x6	4.78783963770226	0	Inf	0
x7	-9.41494205854173	0	-Inf	0
x8	9.57180066772459	0	Inf	0
x9	-5.05724792417194	0	-Inf	0
x10	1.11537005474117	0	Inf	0
x11	-0.00271628165599271	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7800443490302	980	0.0161020860704389		
Model	15.7800443490302	11	1.43454948627547	Inf	0

```

Residual          0      969          0

A = 0.000000, B = 0.113640
order Q(x) = 5.000000, order D(x) = 6.000000
List of powers for Q(x): [0.113640, 0.227280, 0.340919, 0.454559, 0.568199]
List of powers for D(x): [0.113640, 0.227280, 0.340919, 0.454559, 0.568199, 0.681839]
Fitting tinv(0.95,x) in range (2.000000, 100.000000)
MSS of errors squared = 6.862124e-11
Corrected MSS of errors squared = 9.704509e-11
R-Squared =      1.00000000
R-Squared Adjusted =      1.00000000
Particle swarm AICc =  -1.000000e+99
AIC =  -Inf
AICc =  -Inf

```

Using Power $A+B/i$

Fitting $\text{tinv}(0.95,x)$ in range (2.000000, 100.000000)
Pade-Shammas polynomial power is $A+B/i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.59106314148774	0	Inf	0
x1	-0.638377496532929	0	-Inf	0
x2	3.14853947682664	0	Inf	0
x3	-18.3856184394215	0	-Inf	0
x4	40.3628439887868	0	Inf	0
x5	-26.3690179677418	0	-Inf	0
x6	0.38810612397954	0	Inf	0
x7	-2.27666049177107	0	-Inf	0
x8	11.3874713931377	0	Inf	0
x9	-26.1664503462198	0	-Inf	0
x10	18.6250130274469	0	Inf	0
x11	0	0	NaN	NaN
x12	-0.911521982272278	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7800443490638	980	0.0161020860704732		
Model	15.7800443490638	11	1.43454948627852	Inf	0
Residual	0	969	0		

```

A = 0.000000, B = 2.000000
order Q(x) = 5.000000, order D(x) = 7.000000
List of powers for Q(x): [2.000000, 1.000000, 0.666667, 0.500000, 0.400000]
List of powers for D(x): [2.000000, 1.000000, 0.666667, 0.500000, 0.400000, 0.333333,
0.285714]
Fitting tinv(0.95,x) in range (2.000000, 100.000000)
MSS of errors squared = 2.450928e-11
Corrected MSS of errors squared = 3.466135e-11
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}$

```

Fitting tinv(0.95,x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.62995838498931	0	Inf	0
x1	-28.5430803071815	0	-Inf	0
x2	63.3428180343543	0	Inf	0
x3	0	0	NaN	NaN
x4	-80.0117784563243	0	-Inf	0
x5	0	0	NaN	NaN
x6	83.1633501239254	0	Inf	0
x7	-39.5809956979476	0	-Inf	0
x8	17.4960348024552	0	Inf	0
x9	-38.7951255876156	0	-Inf	0
x10	0	0	NaN	NaN
x11	48.9413945182553	0	Inf	0
x12	0	0	NaN	NaN
x13	-50.8203723748492	0	-Inf	0
x14	24.1780253620709	0	Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7800443490354	980	0.0161020860704443		

Model	15.7800443490354	10	1.57800443490354	Inf	0
Residual	0	970	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 7.000000

List of powers for Q(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

List of powers for D(x): [0.100000, 0.141421, 0.173205, 0.200000, 0.223607, 0.244949, 0.264575]

Fitting tinv(0.95,x) in range (2.000000, 100.000000)

MSS of errors squared = 7.432402e-09

Corrected MSS of errors squared = 1.051100e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i+C)

Fitting tinv(0.95,x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is A+B*sqrt(i+C)

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.96935740128013	0	Inf	0
x1	-253.48092931631	0	-Inf	0
x2	870.566161009448	0	Inf	0
x3	-1075.13756811955	0	-Inf	0
x4	497.448857064756	0	Inf	0
x5	0	0	NaN	NaN
x6	-41.3605731150202	0	-Inf	0
x7	88.6844486548612	0	Inf	0
x8	-204.208455018459	0	-Inf	0
x9	0	0	NaN	NaN
x10	357.445987023511	0	Inf	0
x11	-333.468153298488	0	-Inf	0
x12	92.545328671766	0	Inf	0

Number of observations: 99, Error degrees of freedom: 88

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	2.40972607707393	98	0.0245890416027952		
Model	2.40972607707393	10	0.240972607707393	Inf	0
Residual	0	88	0		

A = 0.000000, B = 0.100000, C = 1.728105

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.165170, 0.193083, 0.217442, 0.239335, 0.259386, 0.277995]

List of powers for D(x): [0.165170, 0.193083, 0.217442, 0.239335, 0.259386, 0.277995]

Fitting tinv(0.95,x) in range (2.000000, 100.000000)

MSS of errors squared = 1.383725e-10

Corrected MSS of errors squared = 2.396682e-10

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*(i-1)$

Fitting tinv(0.95,x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i+C*(i-1)$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.23699603635656	0	Inf	0
x1	-14.0031411910915	0	-Inf	0
x2	30.3432863257649	0	Inf	0
x3	-24.0895448647869	0	-Inf	0
x4	0	0	NaN	NaN
x5	12.9772892867467	0	Inf	0
x6	-6.72754994712944	0	-Inf	0
x7	10.4263533440527	0	Inf	0
x8	-20.8351013218518	0	-Inf	0
x9	15.2957311849803	0	Inf	0
x10	-0.0209197247329545	0	-Inf	0
x11	-7.92175535973914	0	-Inf	0
x12	4.09785142920234	0	Inf	0
x13	-0.000602793084040433	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 968

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	15.7800443490858	980	0.0161020860704957		
Model	15.7800443490858	12	1.31500369575715	Inf	0
Residual	0	968	0		

A = 0.235492, B = 0.186619, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.422110, 0.608729, 0.795348, 0.981966, 1.168585, 1.355204]

List of powers for D(x): [0.422110, 0.608729, 0.795348, 0.981966, 1.168585, 1.355204, 1.541822]

Fitting tinv(0.95,x) in range (2.000000, 100.000000)

MSS of errors squared = 9.555266e-12

Corrected MSS of errors squared = 1.655021e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting tinv(0.95,x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$y \sim 1 + x1 + x2 + x3 + x4 + x5 + x6 + x7 + x8 + x9 + x10 + x11 + x12$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.31457334514128	0	Inf	0
x1	-26.7300119360709	0	-Inf	0
x2	50.549585520633	0	Inf	0
x3	0	0	NaN	NaN
x4	-75.3512589617303	0	-Inf	0
x5	72.0598572865147	0	Inf	0
x6	-21.8902932661077	0	-Inf	0
x7	19.7118110948385	0	Inf	0
x8	-38.2134122277554	0	-Inf	0
x9	6.47387437172582	0	Inf	0
x10	42.9464104862758	0	Inf	0
x11	-43.1588982762208	0	-Inf	0
x12	13.247734598187	0	Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue

Total	15.7800443490162	980	0.0161020860704247		
Model	15.7800443490162	11	1.4345494862742	Inf	0
Residual	0	969	0		

A = 0.224898, B = 0.109309, C = 0.000000

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for Q(x): [0.334207, 0.443516, 0.552825, 0.662134, 0.771443, 0.880752]

List of powers for D(x): [0.334207, 0.443516, 0.552825, 0.662134, 0.771443, 0.880752]

Fitting tinv(0.95,x) in range (2.000000, 100.000000)

MSS of errors squared = 6.391525e-12

Corrected MSS of errors squared = 1.107045e-11

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting tinv(0.95,x) in range (2.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.690308856514507	0	Inf	0
x1	-1.4396705328121	0	-Inf	0
x2	1.08568329646199	0	Inf	0
x3	-0.298148474356974	0	-Inf	0
x4	2.79700241705124	0	Inf	0
x5	-4.68213670569951	0	-Inf	0
x6	5.31575423509694	0	Inf	0
x7	-3.60837265873991	0	-Inf	0
x8	1.46568026325221	0	Inf	0
x9	-0.324477430754422	0	-Inf	0
x10	0.030462749875892	0	Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.780044349033	980	0.0161020860704418		
Model	15.780044349033	10	1.5780044349033	Inf	0
Residual	0	970	0		

```

A = 0.044396, B = 0.100000, C = 0.098251
order Q(x) = 3.000000, order D(x) = 7.000000
List of powers for Q(x): [0.144396, 0.342647, 0.483344]
List of powers for D(x): [0.144396, 0.342647, 0.483344, 0.614572, 0.740899, 0.864093,
0.985062]
Fitting tinv(0.95,x) in range (2.000000, 100.000000)
MSS of errors squared = 2.543609e-09
Corrected MSS of errors squared = 4.405659e-09
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}+C*(i-1)$

```

Fitting tinv(0.95,x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.46702060343898	0	Inf	0
x1	-38.5246449836669	0	-Inf	0
x2	123.943941894659	0	Inf	0
x3	-167.951153913096	0	-Inf	0
x4	109.888368086149	0	Inf	0
x5	-28.8159450649577	0	-Inf	0
x6	-0.0430452566153475	0	-Inf	0
x7	25.4365554912918	0	Inf	0
x8	-79.4709810144681	0	-Inf	0
x9	105.397595476231	0	Inf	0
x10	-68.169764876426	0	-Inf	0
x11	17.8121698253	0	Inf	0

Number of observations: 981, Error degrees of freedom: 969

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7800443490798	980	0.0161020860704896		
Model	15.7800443490798	11	1.43454948627998	Inf	0
Residual	0	969	0		

```

A = 0.333025, B = 0.100000, C = 0.108801
order Q(x) = 6.000000, order D(x) = 5.000000
List of powers for Q(x): [0.433025, 0.583247, 0.723832, 0.859428, 0.991836, 1.121979]
List of powers for D(x): [0.433025, 0.583247, 0.723832, 0.859428, 0.991836]
Fitting tinv(0.95,x) in range (2.000000, 100.000000)
MSS of errors squared = 1.146141e-10
Corrected MSS of errors squared = 1.985174e-10
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

```

Fitting tinv(0.95,x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	2.15966326866004	0	Inf	0
x1	-383.167698942334	0	-Inf	0
x2	436.148602916083	0	Inf	0
x3	0	0	NaN	NaN
x4	1015.81572505563	0	Inf	0
x5	0	0	NaN	NaN
x6	-2957.03601445611	0	-Inf	0
x7	1886.93525279201	0	Inf	0
x8	-305.344654296339	0	-Inf	0
x9	2521.26285863527	0	Inf	0
x10	-4351.00635517155	0	-Inf	0
x11	0	0	NaN	NaN
x12	4516.53134623129	0	Inf	0
x13	-2380.57915530456	0	-Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7800443486371	980	0.0161020860700378		
Model	15.7800443486371	10	1.57800443486371	Inf	0
Residual	0	970	0		

```

A = 0.500000, B = 0.100000, C = 0.000000
order Q(x) = 7.000000, order D(x) = 6.000000
List of powers for Q(x): [0.600000, 0.641421, 0.673205, 0.700000, 0.723607, 0.744949,
0.764575]
List of powers for D(x): [0.600000, 0.641421, 0.673205, 0.700000, 0.723607, 0.744949]
Fitting tinv(0.95,x) in range (2.000000, 100.000000)
MSS of errors squared = 8.717360e-10
Corrected MSS of errors squared = 1.509891e-09
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\log_{10}(i)^4$

```

Fitting tinv(0.95,x) in range (2.000000, 100.000000)
Pade-Shammas polynomial power is A+B*log10(i)^4

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	1.92310974629927	0	Inf	0
x1	-355.57913928288	0	-Inf	0
x2	429.309734498611	0	Inf	0
x3	-84.5605792362826	0	-Inf	0
x4	8.91655277170115	0	Inf	0
x5	178.665002123943	0	Inf	0
x6	-214.811006396839	0	-Inf	0
x7	41.0329028022751	0	Inf	0
x8	-3.72588323847895	0	-Inf	0
x9	-0.170341323503187	0	-Inf	0
x10	0.00778819981247088	0	Inf	0

Number of observations: 981, Error degrees of freedom: 970

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7800443490042	980	0.0161020860704124		
Model	15.7800443490042	10	1.57800443490042	Inf	0
Residual	0	970	0		

```

A = 0.177989, B = 1.098095
order Q(x) = 4.000000, order D(x) = 6.000000

```

```
List of powers for Q(x): [0.177989, 0.187006, 0.234894, 0.322266]
List of powers for D(x): [0.177989, 0.187006, 0.234894, 0.322266, 0.440093, 0.580609]
Fitting tinv(0.95,x) in range (2.000000, 100.000000)
MSS of errors squared = 6.204015e-10
Corrected MSS of errors squared = 8.773802e-10
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf
```

Comments

The following table shows the summary results for the ten models that fit the one-sided t-inverse function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Order S</i>
$A+B*i-C*(i-1)$	0.224	0.109	0	Inf	-Inf	6.39152500E-12	6, 6
$A+B*i+C*(i-1)$	0.235	0.186	0	Inf	-Inf	9.55526600E-12	6, 7
$A+B/i$	0	2		Inf	-Inf	2.45092800E-11	5, 7
$A+B*i$	0	0.113		Inf	-Inf	6.86212400E-11	5, 6
$A+B*\sqrt{i}+C*(i-1)$	0.333	0.1	0.108	Inf	-Inf	1.14614100E-10	6, 5
$A+B*\sqrt{i}+C$	0	0.1	1.728	Inf	-Inf	1.38372500E-10	6, 6
$A+B*\log_{10}(i)^4$	0.177	1.098		Inf	-Inf	6.20401500E-10	4, 6
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.5	0.1	0	Inf	-Inf	8.71736000E-10	7, 6
$A+B*i+C*\sqrt{i-1}$	0.044	0.1	0.098	Inf	-Inf	2.54360900E-09	3, 7
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	7.43240200E-09	7, 7

RESULTS FOR THE TRIGAMMA FUNCTION

Using Power $A+B*i$

Fitting $\text{tigamma}(x)$ in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.316267477286059	0	Inf	0
x_1	-1.47094967133996	0	-Inf	0
x_2	2.93427884850287	0	Inf	0
x_3	-3.25206906099128	0	-Inf	0
x_4	2.16117720594072	0	Inf	0
x_5	-0.860911551955789	0	-Inf	0
x_6	0.190368328830336	0	Inf	0
x_7	-0.018031780743235	0	-Inf	0
x_8	5.43947819585191	0	Inf	0
x_9	-12.3012214011848	0	-Inf	0
x_{10}	14.8163331818008	0	Inf	0
x_{11}	-10.031826848517	0	-Inf	0
x_{12}	3.62133233438471	0	Inf	0
x_{13}	-0.544174368807734	0	-Inf	0

Number of observations: 991, Error degrees of freedom: 977

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

SumSq	DF	MeanSq	F	pValue
-------	----	--------	---	--------

Total	15.7946327360629	990	0.0159541744808716		
Model	15.7946327360629	13	1.21497174892791	Inf	0
Residual	0	977	0		

A = 0.000000, B = 0.100000

order Q(x) = 7.000000, order D(x) = 6.000000

List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000, 0.700000]

List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000]

Fitting tgamma(x) in range (1.000000, 100.000000)

MSS of errors squared = 2.923851e-07

Corrected MSS of errors squared = 4.134949e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B/i

Fitting trigamma(x) in range (1.000000, 100.000000)

Pade-Shammas polynomial power is A+B/i

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	8.70556238382505e-06	0	Inf	0
x1	-2.94939118242585e-06	0	-Inf	0
x2	2.95514261885602e-05	0	Inf	0
x3	-3.54236153284696e-05	0	-Inf	0
x4	0.0225943559647527	0	Inf	0
x5	-1.1172383486968	0	-Inf	0
x6	5.24405887793829	0	Inf	0
x7	0	0	NaN	NaN
x8	-13.3562259923466	0	-Inf	0
x9	0	0	NaN	NaN
x10	10.206811177671	0	Inf	0

Number of observations: 991, Error degrees of freedom: 982

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7946327360623	990	0.015954174480871		

Model	15.7946327360623	8	1.97432909200778	Inf	0
Residual	0	982	0		

A = 0.000125, B = 0.110440

order Q(x) = 3.000000, order D(x) = 7.000000

List of powers for Q(x): [0.110566, 0.055345, 0.036939]

List of powers for D(x): [0.110566, 0.055345, 0.036939, 0.027735, 0.022213, 0.018532, 0.015903]

Fitting trigamma(x) in range (1.000000, 100.000000)

MSS of errors squared = 3.064597e-08

Corrected MSS of errors squared = 4.333995e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power A+B*sqrt(i)

Fitting trigamma(x) in range (1.000000, 100.000000)

Pade-Shammas polynomial power is A+B*sqrt(i)

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11}$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.00517203693153136	0	Inf	0
x1	-0.0767006551728714	0	-Inf	0
x2	0.237329570906909	0	Inf	0
x3	-0.307863058975598	0	-Inf	0
x4	0.183932735349704	0	Inf	0
x5	-0.0418777782158866	0	-Inf	0
x6	46.6616203194089	0	Inf	0
x7	-267.491087578643	0	-Inf	0
x8	626.153957339918	0	Inf	0
x9	-725.172268686161	0	-Inf	0
x10	414.537241400024	0	Inf	0
x11	-93.6894584483868	0	-Inf	0

Number of observations: 991, Error degrees of freedom: 979

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7946327360993	990	0.0159541744809084		
Model	15.7946327360993	11	1.43587570328176	Inf	0


```

order Q(x) = 5.000000, order D(x) = 5.000000
List of powers for Q(x): [0.100477, 0.141759, 0.173481, 0.200239, 0.223821]
List of powers for D(x): [0.100477, 0.141759, 0.173481, 0.200239, 0.223821]
Fitting trigamma(x) in range (1.000000, 100.000000)
MSS of errors squared = 2.849308e-08
Corrected MSS of errors squared = 4.935147e-08
R-Squared = 1.000000000
R-Squared Adjusted = 1.000000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*i+C*(i-1)$

```

Fitting trigamma(x) in range (1.000000, 100.000000)
Pade-Shammas polynomial power is A+B*i+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0470695718097521	0	-Inf	0
x1	0.229469381254758	0	Inf	0
x2	-0.425867485787886	0	-Inf	0
x3	0.394709123716743	0	Inf	0
x4	-0.197565222387741	0	-Inf	0
x5	0.051410964615587	0	Inf	0
x6	-0.00548881261066801	0	-Inf	0
x7	6.21811516368628	0	Inf	0
x8	-15.4017952701282	0	-Inf	0
x9	20.147636422455	0	Inf	0
x10	-14.8785913540267	0	-Inf	0
x11	5.88605300542037	0	Inf	0
x12	-0.971173810319844	0	-Inf	0

Number of observations: 991, Error degrees of freedom: 978

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.794632736061	990	0.0159541744808697		
Model	15.794632736061	12	1.31621939467175	Inf	0
Residual	0	978	0		

A = 0.005821, B = 0.100000, C = 0.000000

order Q(x) = 6.000000, order D(x) = 6.000000

List of powers for $Q(x)$: [0.105821, 0.205821, 0.305821, 0.405821, 0.505821, 0.605821]

List of powers for $D(x)$: [0.105821, 0.205821, 0.305821, 0.405821, 0.505821, 0.605821]

Fitting trigamma(x) in range (1.000000, 100.000000)

MSS of errors squared = 6.859719e-09

Corrected MSS of errors squared = 1.188138e-08

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i-C*(i-1)$

Fitting trigamma(x) in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i-C*(i-1)$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	-0.000902626282364408	0	-Inf	0
x2	0.00628263991486252	0	Inf	0
x3	-0.0182153984040929	0	-Inf	0
x4	0.0281571643744327	0	Inf	0
x5	-0.0244738658513339	0	-Inf	0
x6	0.0113409306142504	0	Inf	0
x7	-0.00218886860578883	0	-Inf	0
x8	-0.0057454577048371	0	-Inf	0
x9	0.0348719421545257	0	Inf	0
x10	-0.103419072939238	0	-Inf	0
x11	1.05351044229409	0	Inf	0
x12	0.0207821609316605	0	Inf	0

Number of observations: 991, Error degrees of freedom: 979

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7946327360624	990	0.0159541744808711		
Model	15.7946327360624	11	1.4358757032784	Inf	0
Residual	0	979	0		

A = 0.006705, B = 0.100000, C = 0.134738

order $Q(x)$ = 7.000000, order $D(x)$ = 5.000000

List of powers for Q(x): [0.106705, 0.071967, 0.037229, 0.002491, -0.032248, -0.066986, -0.101724]

List of powers for D(x): [0.106705, 0.071967, 0.037229, 0.002491, -0.032248]

Fitting trigamma(x) in range (1.000000, 100.000000)

MSS of errors squared = 6.062081e-07

Corrected MSS of errors squared = 1.049983e-06

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*i+C*\sqrt{i-1}$

Fitting trigamma(x) in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*i+C*\sqrt{i-1}$

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13} + x_{14}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.106540200225157	0	-Inf	0
x1	0.52113810672475	0	Inf	0
x2	-1.07508656235934	0	-Inf	0
x3	1.21060363068301	0	Inf	0
x4	-0.804970456254268	0	-Inf	0
x5	0.317083957684188	0	Inf	0
x6	-0.0687630114222293	0	-Inf	0
x7	0.00635323755540073	0	Inf	0
x8	6.54839004502022	0	Inf	0
x9	-18.4333540491779	0	-Inf	0
x10	28.9006073361024	0	Inf	0
x11	-27.2269268344798	0	-Inf	0
x12	15.385588435747	0	Inf	0
x13	-4.81710838880901	0	-Inf	0
x14	0.642913671274129	0	Inf	0

Number of observations: 991, Error degrees of freedom: 976

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7946327360623	990	0.015954174480871		
Model	15.7946327360623	14	1.12818805257588	Inf	0
Residual	0	976	0		

```

A = 0.000000, B = 0.100000, C = 0.000000
order Q(x) = 7.000000, order D(x) = 7.000000
List of powers for Q(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
List of powers for D(x): [0.100000, 0.200000, 0.300000, 0.400000, 0.500000, 0.600000,
0.700000]
Fitting trigamma(x) in range (1.000000, 100.000000)
MSS of errors squared = 3.913079e-08
Corrected MSS of errors squared = 6.777652e-08
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B*\sqrt{i}+C*(i-1)$

```

Fitting trigamma(x) in range (1.000000, 100.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	-0.0190963366481445	0	-Inf	0
x1	0.161580590202673	0	Inf	0
x2	-0.314387510516258	0	-Inf	0
x3	0	0	NaN	NaN
x4	0.637164836049296	0	Inf	0
x5	-0.781774809330593	0	-Inf	0
x6	0.388553800742398	0	Inf	0
x7	-0.0720949022711135	0	-Inf	0
x8	15.2285427694952	0	Inf	0
x9	-61.4288329596906	0	-Inf	0
x10	113.337256620568	0	Inf	0
x11	-108.741157180665	0	-Inf	0
x12	53.0374638265428	0	Inf	0
x13	-10.4332400465311	0	-Inf	0

Number of observations: 991, Error degrees of freedom: 978

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.794632736062	990	0.0159541744808707		
Model	15.794632736062	12	1.31621939467183	Inf	0

```

Residual          0      978          0

A = 0.000000, B = 0.100000, C = 0.021862
order Q(x) = 7.000000, order D(x) = 6.000000
List of powers for Q(x): [0.100000, 0.163283, 0.216928, 0.265585, 0.311053, 0.354257,
0.395744]
List of powers for D(x): [0.100000, 0.163283, 0.216928, 0.265585, 0.311053, 0.354257]
Fitting trigamma(x) in range (1.000000, 100.000000)
MSS of errors squared = 5.590151e-09
Corrected MSS of errors squared = 9.682426e-09
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Using Power $A+B\sqrt{i}+C\sqrt{i-1}$

```

Fitting trigamma(x) in range (1.000000, 100.000000)
Pade-Shammas polynomial power is A+B*sqrt(i)+C*sqrt(i-1)

```

Linear regression model:

$$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9 + x_{10} + x_{11} + x_{12} + x_{13}$$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0.877629955452427	0	Inf	0
x1	-35.0858759681486	0	-Inf	0
x2	176.573052345166	0	Inf	0
x3	-370.331015905447	0	-Inf	0
x4	389.476681430699	0	Inf	0
x5	-204.095571052351	0	-Inf	0
x6	42.5853516020921	0	Inf	0
x7	20.9051324057242	0	Inf	0
x8	-47.0405142838625	0	-Inf	0
x9	0	0	NaN	NaN
x10	58.9901317938976	0	Inf	0
x11	0	0	NaN	NaN
x12	-60.2071033395883	0	-Inf	0
x13	28.3521999780955	0	Inf	0

Number of observations: 991, Error degrees of freedom: 979

R-squared: 1, Adjusted R-squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7946327361047	990	0.0159541744809138		

Model	15.7946327361047	11	1.43587570328225	Inf	0
Residual	0	979	0		

A = 0.010270, B = 0.100000, C = 0.000000

order Q(x) = 6.000000, order D(x) = 7.000000

List of powers for Q(x): [0.110270, 0.151692, 0.183475, 0.210270, 0.233877, 0.255219]

List of powers for D(x): [0.110270, 0.151692, 0.183475, 0.210270, 0.233877, 0.255219, 0.274845]

Fitting trigamma(x) in range (1.000000, 100.000000)

MSS of errors squared = 9.794456e-08

Corrected MSS of errors squared = 1.696449e-07

R-Squared = 1.00000000

R-Squared Adjusted = 1.00000000

Particle swarm AICc = -1.000000e+99

AIC = -Inf

AICc = -Inf

Using Power $A+B*\log_{10}(i)^4$

Fitting trigamma(x) in range (1.000000, 100.000000)

Pade-Shammas polynomial power is $A+B*\log_{10}(i)^4$

Linear regression model:

$y \sim 1 + x_1 + x_2 + x_3 + x_4 + x_5 + x_6 + x_7 + x_8 + x_9$

Estimated Coefficients:

	Estimate	SE	tStat	pValue
(Intercept)	0	0	NaN	NaN
x1	2.7620855861624e-10	0	Inf	0
x2	-3.10047119200234e-10	0	-Inf	0
x3	4.1151964058492e-11	0	Inf	0
x4	-2.35927653300156e-12	0	-Inf	0
x5	4.7057406156676e-14	0	Inf	0
x6	0.999999999193013	0	Inf	0
x7	1.02749095568335e-09	0	Inf	0
x8	-2.52533476213327e-10	0	-Inf	0
x9	2.89889991575354e-11	0	Inf	0

Number of observations: 991, Error degrees of freedom: 982

R-squared: 1, Adjusted R-Squared: 1

F-statistic vs. constant model: Inf, p-value = 0

	SumSq	DF	MeanSq	F	pValue
Total	15.7946327360623	990	0.015954174480871		
Model	15.7946327360623	8	1.97432909200779	Inf	0
Residual	0	982	0		

```

A = 0.000000, B = 3.500000
order Q(x) = 5.000000, order D(x) = 4.000000
List of powers for Q(x): [0.000000, 0.028741, 0.181377, 0.459862, 0.835415]
List of powers for D(x): [0.000000, 0.028741, 0.181377, 0.459862]
Fitting trigamma(x) in range (1.000000, 100.000000)
MSS of errors squared = 2.192617e-05
Corrected MSS of errors squared = 3.100829e-05
R-Squared = 1.00000000
R-Squared Adjusted = 1.00000000
Particle swarm AICc = -1.000000e+99
AIC = -Inf
AICc = -Inf

```

Comments

The following table shows the summary results for the ten models that fit the trigamma function. The results are sorted in ascending order using the MSSE values.

<i>Power</i>	<i>A</i>	<i>B</i>	<i>C</i>	<i>F</i>	<i>AICc</i>	<i>MSSE</i>	<i>Q/D Orders</i>
$A+B\sqrt{i}+C*(i-1)$	0	0.1	0.021	Inf	-Inf	5.59015100E-09	7, 6
$A+B*i+C*(i-1)$	0.005	0.1	0	Inf	-Inf	6.85971900E-09	6, 6
$A+B*\sqrt{i}+C$	0	0.1	0.009	Inf	-Inf	2.84930800E-08	5, 5
$A+B*\sqrt{i}$	0	0.1		Inf	-Inf	3.04493800E-08	5, 6
$A+B/i$	0	0.11		Inf	-Inf	3.06459700E-08	3, 7
$A+B*i+C*\sqrt{i-1}$	0	0.1	0	Inf	-Inf	3.91307900E-08	7, 7
$A+B*\sqrt{i}+C*\sqrt{i-1}$	0.01	0.1	0	Inf	-Inf	9.79445600E-08	6, 7
$A+B*i$	0	0.1		Inf	-Inf	2.92385100E-07	7, 6
$A+B*i-C*(i-1)$	0.006	0.1	0.134	Inf	-Inf	6.06208100E-07	7, 5
$A+B*\log_{10}(i)^4$	0	3.5		Inf	-Inf	2.19261700E-05	5, 4

CONCLUSIONS

Overall, the Pade-Shammas polynomials did well in fitting many tested equations. The Pade-Shammas polynomials struggled in fitting the following equations:

- The Bessel functions.
- The Fresnel cosine and sine integrals.

So how do the various Shammas Polynomial powers compare with each other. The next table shows the frequency of these powers in attaining the top 3 spots. The counts colored in red indicate the top of their placement.

<i>Power</i>	<i>Top 1</i>	<i>Top 2</i>	<i>Top 3</i>
$A+B*i$	3	1	1
$A+B/i$	4	3	5

<i>Power</i>	<i>Top 1</i>	<i>Top 2</i>	<i>Top 3</i>
$A+B*\sqrt{i}$	0	3	2
$A+B*\sqrt{i+C}$	5	2	3
$A+B*i+C*(i-1)$	4	3	3
$A+B*i-C*(i-1)$	1	8	5
$A+B*i+C*\sqrt{i-1}$	3	3	4
$A+B*\sqrt{i}+C*(i-1)$	6	1	4
$A+B*\sqrt{i}+C*\sqrt{i-1}$	6	5	2
$A+B*\log_{10}(i)^4$	0	3	3

The top Shammas polynomial powers are $A+B*\sqrt{i}+C*(i-1)$ and $A+B*\sqrt{i}+C*\sqrt{i-1}$. The second top Shammas polynomial power is $A+B*i-C*(i-1)$. The third top Shammas polynomial powers are $A+B*i-C*(i-1)$ and $A+B/i$.

Overall the power expressions $A+B*\sqrt{i}+C*\sqrt{i-1}$, $A+B*\sqrt{i}+C*(i-1)$, and $A+B*i-C*(i-1)$ are the favorites.

LOOKING AT THE EFFECT OF NORMALIZATION THE DATA

In this section we look at the effect of normalizing data (in Part I). The next table shows the best normalized and non-normalized MSSE values (regardless of the Shammas polynomial power, to make things a bit simpler). The lines in blue favor data normalization while the lines in red favor non-normalization. The criterion for significant difference is a difference in the exponents by three orders.

<i>Function</i>	<i>MSSE (normalized)</i>	<i>MSSE</i>
acosh	4.72541900E-06	1.44319900E-07
acos	4.72541900E-06	6.35783500E-06
asin	3.84022700E-07	2.93704600E-06
atan	1.08003700E-12	1.32405500E-11
asinh	7.83524800E-07	3.26304300E-09
atanh	7.59681300E-06	1.00365500E-04
Ci	8.63345800E-03	3.22542500E-04
cos	1.18441600E-13	2.30806400E-12
cosh	2.26390500E-11	3.18572100E-09
digamma	1.92179900E-08	3.93283800E-11
erf	2.74329100E-09	3.00722900E-09
exp	1.13495600E-13	9.26021800E-12
Fresnel cos	3.15519000E-02	2.31725600E-02

<i>Function</i>	<i>MSSE (normalized)</i>	<i>MSSE</i>
Fresnel sin	2.47606100E-02	2.89307800E-02
J0(x)	3.96863400E-02	1.99099500E-02
J1(x)	4.22681800E-02	5.61194500E-02
J2(x)	6.91673700E-02	3.52228500E-02
J3(x)	6.91673700E-02	2.32920700E-02
J4(x)	4.23174300E-02	1.21881800E-02
J5(x)	4.72908700E-02	1.40629300E-02
ln	1.04284400E-11	5.73941200E-14
log10	1.38373400E-11	4.29414500E-15
log10gamma	2.52067800E-09	2.62813600E-10
10^x	1.21838900E-14	2.36129200E-11
Si	9.78105700E-05	2.42792000E-04
sin	1.21654700E-13	1.90657600E-10
sinh	5.14194300E-12	2.59745100E-08
tan	3.21995800E-15	7.20483700E-12
tanh	2.08310500E-11	1.54867300E-10
tinv2	8.64941700E-10	5.75015900E-11
tinv1	7.08397900E-10	6.39152500E-12
trigamma	7.08397900E-10	5.59015100E-09

The above table shows that for most functions the MSSE values are not that statistically different. There is a slight advantage for normalized data for a few fitted functions. I was expecting more advantage in normalizing data.

NEXT

The next study I recommend for you to look at is Part II. In Part II the Pade-Shammas polynomials perform curve fitting on the same set of functions using different expressions for the power functions $gx1()$ and $gx2()$ and using separate sets of values for A, B, and C for these functions.

DOCUMENT HISTORY

<i>Date</i>	<i>Version</i>	<i>Comments</i>
10/4/2020	1.00.00	Initial release.

