

# Runge- Phenomena

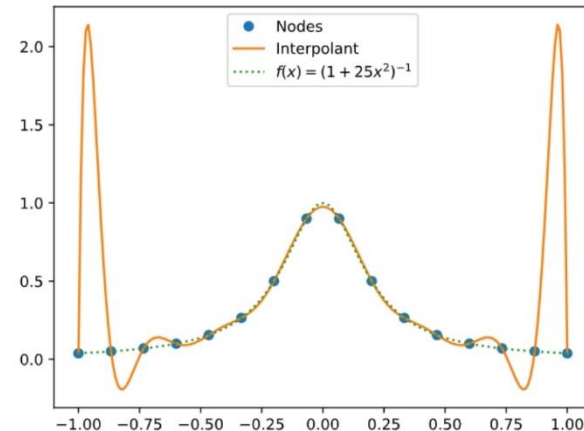
## Important Topics on Curve-Fitting Techniques



## Runge Phenomena in the interpolation of a smooth function

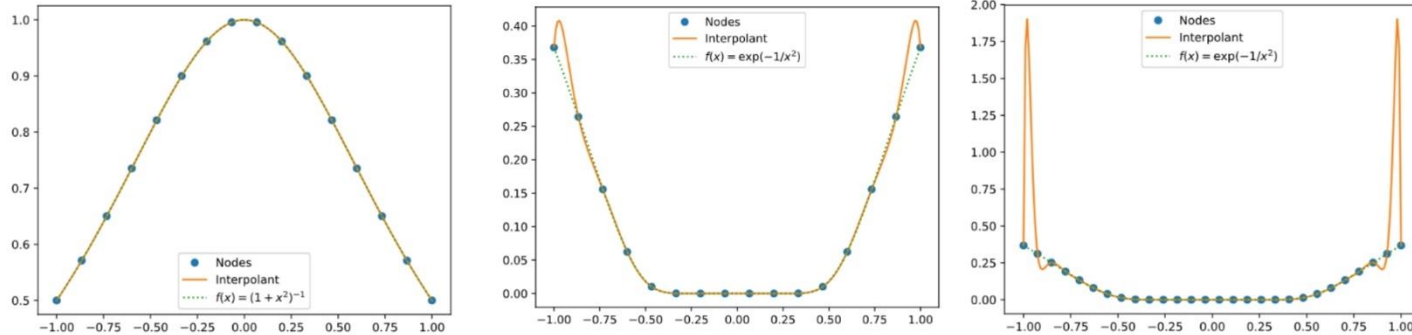
### 1. Introduction to Runge Phenomena

First of all, the “Runge” here is Carl David Tolmé Runge, better known for the Runge-Kutta algorithm for numerically solving differential equations. His name rhymes with cowabunga, not with sponge. Runge showed that polynomial interpolation at evenly-spaced points can fail spectacularly to converge. His example is the function  $f(x) = 1/(1 + x^2)$  on the interval  $[-5, 5]$ , or equivalently, and more convenient here, the function  $f(x) = 1/(1 + 25x^2)$  on the interval  $[-1, 1]$ . Here’s an example with 16 interpolation nodes.



Runge found that in order for interpolation at evenly spaced nodes in  $[-1, 1]$  to converge, the function being interpolated needs to be analytic inside a football-shaped region of the complex plane with major axis  $[-1, 1]$  on the real axis and minor axis approximately  $[-0.5255, 0.5255]$  on the imaginary axis. The function in Runge’s example has a singularity at  $0.2i$ , which is inside the football. Linear interpolation at evenly spaced points would converge for the function  $f(x) = 1/(1 + x^2)$  since the singularity at  $i$  is outside the football.

or another example, consider the function  $f(x) = \exp(-1/x^2)$ , defined to be 0 at 0. This function is infinitely differentiable but it is not analytic at the origin. With only 16 interpolation points as above, there's a small indication of trouble at the ends.



With 28 interpolation points in the plot below, the lack of convergence is clear.

## 2. Example of Runge Phenomena with Runge Function

### (2-1) Runge Function

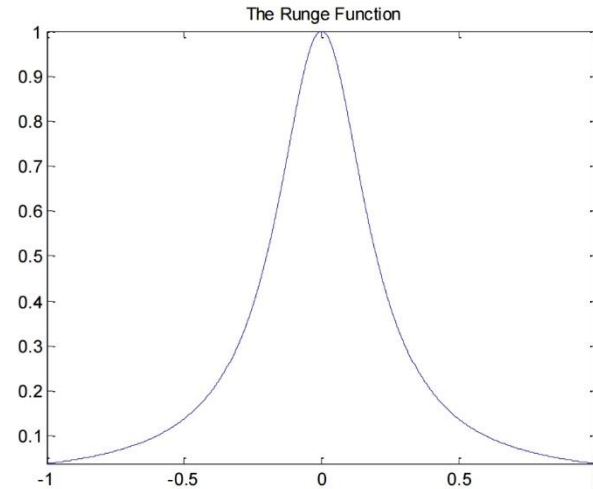
#### Definition of Runge Function

$$f(x) = \frac{1}{1 + 25x^2}$$

$$\frac{df(x)}{dx} = -\frac{50x}{(1 + 25x^2)^2}$$

#### Matlab program to plot

```
clear
clf
z = linspace(-1,1,1001);
f = @(x) 1 ./ (1 + 25.*x.^2);
plot(z,f(z))
title('The Runge Function')
axis tight
```



## (2-2) Lagrange Interpolation of Runge Function to identify the Runge Phenomena

### Program

```
clear
clf
for N = 0:2:20
    x = linspace(-1,1,N+1);
    w = lagrange_weights(x);
    pn = lagrange_eval_naive(z, x, f(x), w);
    plot(z,f(z), z, pn, x, f(x), '*k')
    title(sprintf('%d equally spaced points', N+1))
    h = legend('$f(x)$', sprintf('$p_{%d}(x)$',N));
    set(h,'Interpreter','latex')
    axis tight
    xlim([-1 1])
    snapnow
end

% LAGRANGE_EVAL_NAIVE
% A naive implementation of the Lagrange interpolation function
%
% INPUTS:
% z    evaluation points [array of size m]
% x    abscissae [array of size n]
% y    function values at x(j) [array of size n]
% w    weights as computed using the Lagrange_weights functions [array of size n]
%
% OUTPUTS:
% f    interpolation to the points [array of size m]
function Pn = Lagrange_eval_naive(z,x,y,w)

m = length(x);
```

```
% compute the sum in the interpolation
Pn = 0;
for k = 1:m

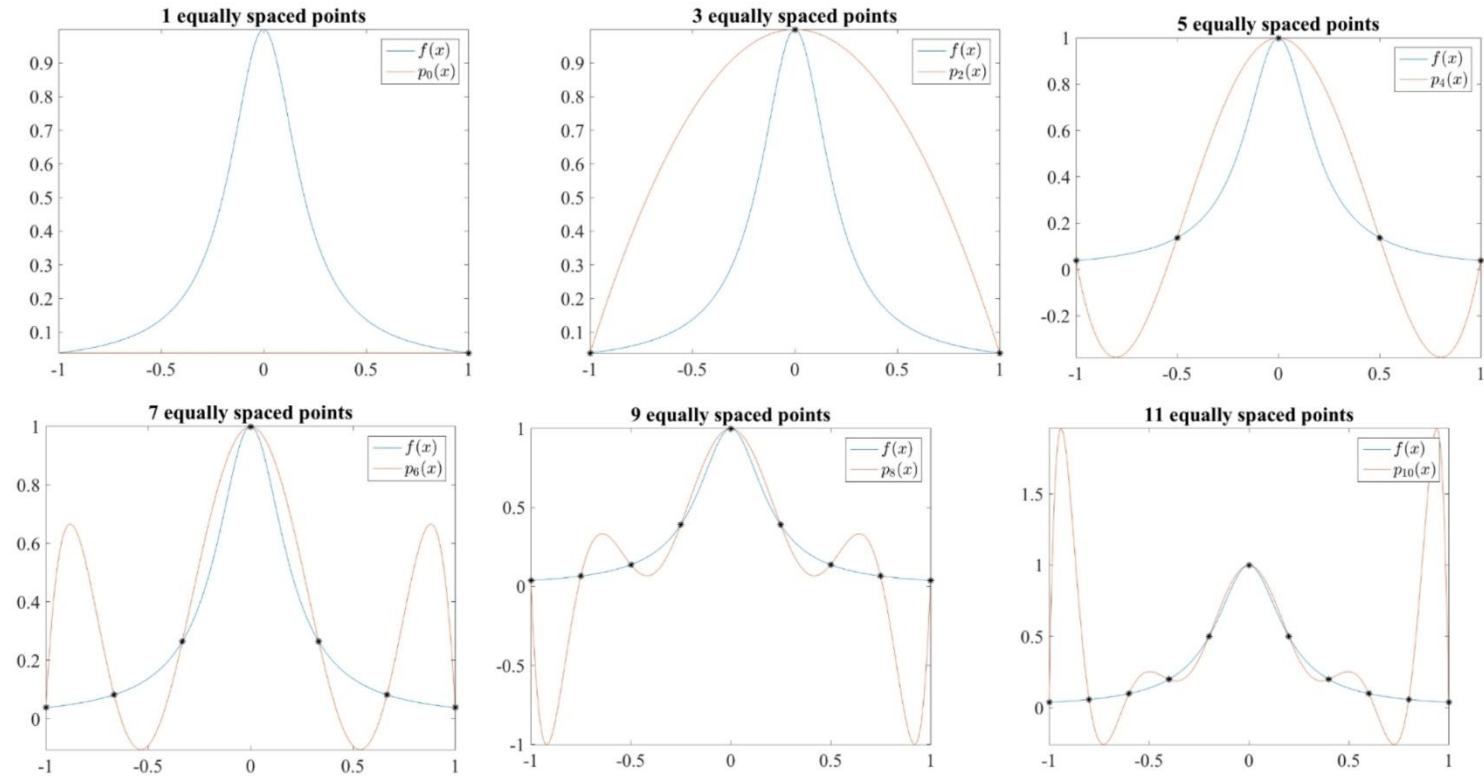
    % computes the Lk for this point
    Lk = 1;
    % computes the given w(k)
    for j=1:m
        if j ~= k
            Lk = Lk .* (z-x(j));
        end
    end
    Lk = Lk * w(k);

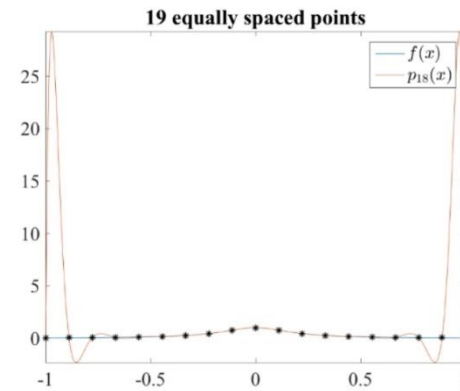
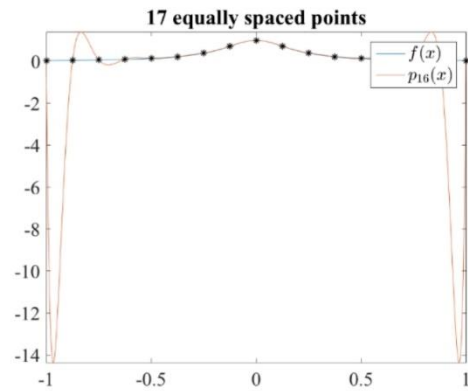
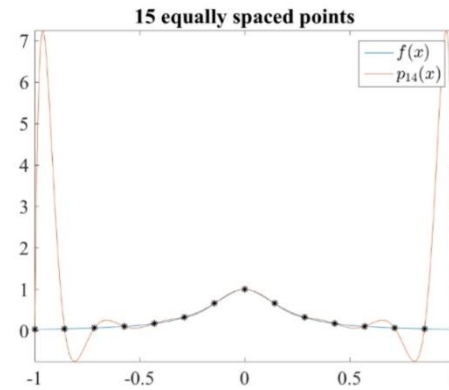
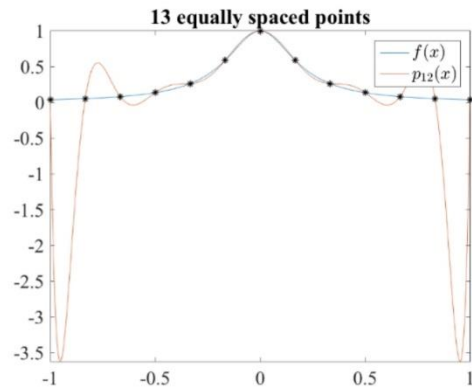
    Pn = Pn + y(k)*Lk;
End
```

```
function w = Lagrange_weights(x)

m = length(x); % really m = n+1 in our formulas
for k = 1:m
    w(k) = 1;          % chooses the w(k) we are working on

    for j=1:m          % computes the given denominator of w(k)
        if j ~= k
            w(k) = w(k) * (x(k)-x(j));
        end
    end
    w(k) = 1/w(k);     % store the actual w(k)
end
```





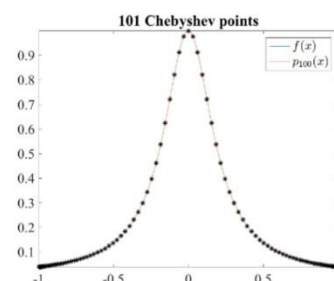
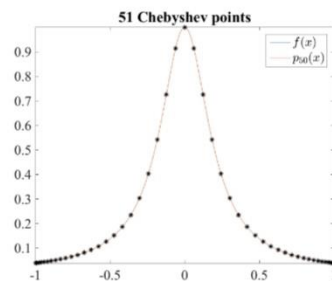
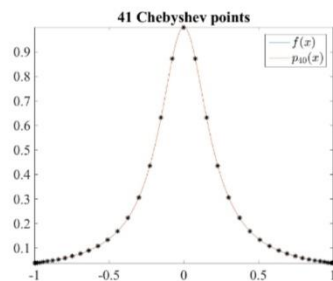
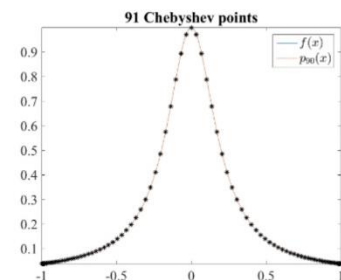
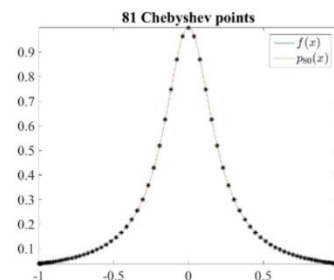
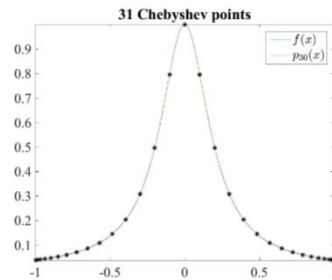
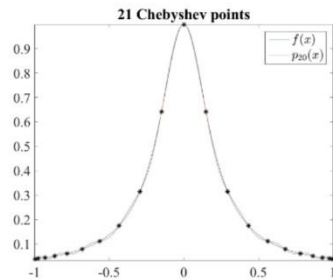
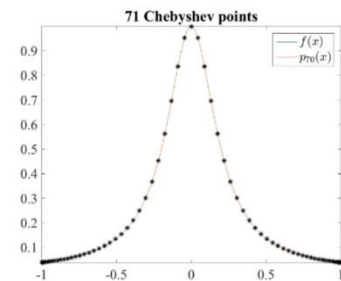
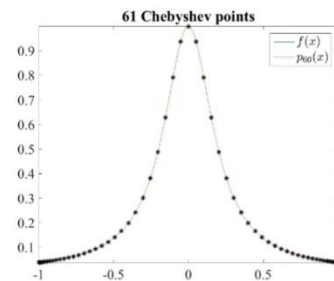
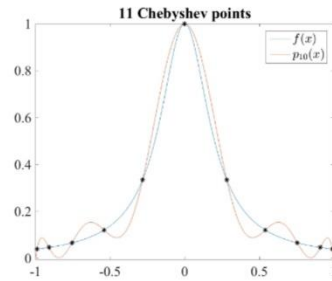
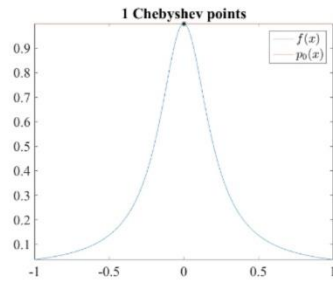


## (2-3) Prevention of Runge Phenomena using Unequal-Spaced nodes

**Example: Lagrange Interpolation Using Chebyshev nodes**

### Program

```
for N = 0:10:100
    x = chebyspace(-1,1,N+1);
    w = lagrange_weights(x);
    pn = lagrange_eval_naive(z, x, f(x), w);
    plot(z,f(z), z, pn, x, f(x), '*k')
    title(sprintf('%d Chebyshev points', N+1))
    h = legend('$f(x)$', sprintf('$p_{%d}(x)$',N));
    set(h,'Interpreter','latex')
    axis tight
    snapnow
end
```



## (2-4) Hermite Spline Interpolation: Interpolation with the minimum Runge Phenomena

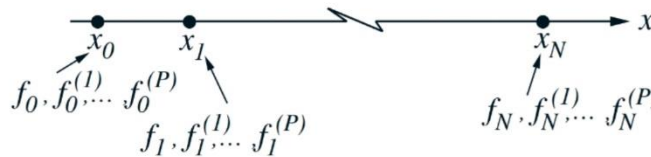
### (2-4a) Reference

(Ref: A. SPITZBART, "A GENERALIZATION OF HERMITE'S INTERPOLATION FORMULA," University of Wisconsin-Milwaukee, The American Mathematical Monthly, Vol. 67, No. 1 (Jan., 1960), pp. 42-46)

### (2-4b) Derivation in a Standard node $x \in [0, 1]$

Hermite's interpolation formula provides an expression for a polynomial which passes through given points with given slopes.

For the given data set  $\{(x_j, f_j, f_j^{(1)}, f_j^{(2)}, \dots, f_j^{(p)})\}_{j=0}^{j=n}$ , the interpolating function can be expressed as



$$g(x) = \sum_{k=0} a_k x^k \quad \text{where} \quad f_j = f(x_j), \quad f_j^{(p)} = \frac{d^p f(x_j)}{dx^p}$$

The interpolating function should satisfy the following conditions

$$\begin{aligned} g(x_j) &= f_j \quad (j = 0, \dots, n) \\ g^{(1)}(x_j) &= f_j^{(1)} \quad (j = 0, \dots, n) \\ &\vdots \\ g^{(p)}(x_j) &= f_j^{(p)} \quad (j = 0, \dots, n) \end{aligned} \quad \text{total } (p+1)(n+1) \text{ constraints with } g(x) = \sum_{k=0}^{k=(p+1)(n+1)-1} a_k x^k.$$

**[Example] Cubic Hermite interpolation with**  $p = n = 1 \rightarrow k = (p+1)(n+1) - 1 = 3$

$$g(x) = a_0 + a_1x + a_2x^2 + a_3x^3 \quad \text{with } x \in [0, 1]$$

$$\begin{pmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 1 & 2 & 3 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{pmatrix} = \begin{pmatrix} f_0 \\ f_1 \\ f_0^{(1)} \\ f_1^{(1)} \end{pmatrix} \rightarrow \begin{pmatrix} a_0 \\ a_1 \\ a_2 \\ a_3 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ -3 & 3 & -2 & -1 \\ 2 & -2 & 1 & 1 \end{pmatrix} \begin{pmatrix} f_0 \\ f_1 \\ f_0^{(1)} \\ f_1^{(1)} \end{pmatrix}$$

Therefore, the interpolating polynomial can be represented by the Hermite basis functions.

$$\begin{aligned} g(x) &= f_0 + f_0^{(1)}x + (-3f_0 + 3f_1 - 2f_0^{(1)} - f_1^{(1)})x^2 + (2f_0 - 2f_1 + f_0^{(1)} + f_1^{(1)})x^3 \\ &= (1 - 3x^2 + 2x^3)f_0 + (3x^2 - 2x^3)f_1 + (x - 2x^2 + x^3)f_0^{(1)} + (-x^2 + x^3)f_1^{(1)} \\ &= h_1^{00}(x)f_0 + h_1^{01}(x)f_1 + h_1^{10}(x)f_0^{(1)} + h_1^{11}(x)f_1^{(1)} \end{aligned}$$

$$h_1^{00}(x) = 1 - 3x^2 + 2x^3$$

$$h_1^{01}(x) = 3x^2 - 2x^3$$

$$h_1^{10}(x) = x - 2x^2 + x^3$$

$$h_1^{11}(x) = -x^2 + x^3$$

(2-4c) Derivation in General node  $x \in [x_0, x_f]$  using nondimensional independent variable

$$\tau = \frac{x - x_0}{x_f - x_0} \in [0, 1] \quad \begin{aligned} dx &= (x_f - x_0) d\tau \\ \frac{df(x)}{dx} &= \frac{df(x(\tau))}{d\tau} \frac{d\tau}{dx} = \frac{1}{(x_f - x_0)} \frac{dg(\tau)}{d\tau} = \frac{g'(\tau)}{(x_f - x_0)} \leftarrow g(\tau) = f\{x(\tau)\} \end{aligned}$$

Interpolating function

$$g(\tau) = (1 - 3\tau^2 + 2\tau^3)g_0 + (3\tau^2 - 2\tau^3)g_f + (\tau - 2\tau^2 + \tau^3)g'_0 + (-\tau^2 + \tau^3)g'_f$$

$$g'(\tau) = (-6\tau + 6\tau^2)g_0 + (6\tau - 6\tau^2)g_f + (1 - 4\tau + 3\tau^2)g'_0 + (-2\tau + 3\tau^2)g'_f$$

$$f(x_0) = g_0$$

$$g(0) = g_0$$

$$f(x_f) = g_f$$

$$g(1) = g_f$$

$$g'(0) = g'_0$$

$$g'(1) = g'_f$$

$$\begin{aligned} \Rightarrow \frac{df(x_0)}{dx} &= \frac{g'_0}{x_f - x_0} \\ \Rightarrow \frac{df(x_f)}{dx} &= \frac{g'_f}{x_f - x_0} \end{aligned}$$

$$\begin{aligned} f(\tau) &= (1 - 3\tau^2 + 2\tau^3)f_0 + (3\tau^2 - 2\tau^3)f_f \\ &\quad + (x_f - x_0) \left\{ \frac{df(x_0)}{dx} (\tau - 2\tau^2 + \tau^3) + \frac{df(x_f)}{dx} (-\tau^2 + \tau^3) \right\} \end{aligned}$$

If we have three-point data of  $(x_0, f_0)$ ,  $(x_1, f_1)$ ,  $(x_f, f_f)$  with  $x_0 < x_1 < x_f$ ,

$$\begin{aligned}\frac{df(x_f)}{dx}(-\tau^2 + \tau^3) &= \frac{f(\tau) - (1 - 3\tau^2 + 2\tau^3)f_0 - (3\tau^2 - 2\tau^3)f_f}{(x_f - x_0)} - \frac{df(x_0)}{dx}(\tau - 2\tau^2 + \tau^3) \\ \frac{df(x_f)}{dx} &= \frac{f(\tau) - (1 - 3\tau^2 + 2\tau^3)f_0 - (3\tau^2 - 2\tau^3)f_f}{(x_f - x_0)(-\tau^2 + \tau^3)} - \frac{df(x_0)}{dx} \frac{(\tau - 2\tau^2 + \tau^3)}{(-\tau^2 + \tau^3)} \\ &= \frac{f(\tau) - (1 - 3\tau^2 + 2\tau^3)f_0 - (3\tau^2 - 2\tau^3)f_f}{(x_f - x_0)(-\tau^2 + \tau^3)} + \frac{df(x_0)}{dx} \frac{(1 - \tau)}{\tau}\end{aligned}$$

Therefore,

$$\frac{df(x_f)}{dx} = \frac{f_1 - (1 - 3\tau_1^2 + 2\tau_1^3)f_0 - (3\tau_1^2 - 2\tau_1^3)f_f}{(x_f - x_0)(-\tau_1^2 + \tau_1^3)} + \frac{df(x_0)}{dx} \frac{(1 - \tau_1)}{\tau_1}$$

The data at the mid-point can be used to estimate the final gradients. With the exact mid point of  $\tau_1 = \frac{1}{2}$

$$\begin{aligned}\frac{df(x_f)}{dx} &= \frac{-8(f_1 - \frac{1}{2}f_0 - \frac{1}{2}f_f)}{(x_f - x_0)} + \frac{df(x_0)}{dx} \\ &= \frac{df(x_0)}{dx} + \frac{4(f_0 - 2f_1 + f_f)}{(x_f - x_0)} \leftarrow \frac{d^2 f(x_1)}{dx^2} \approx \frac{(f_0 - 2f_1 + f_f)}{\frac{1}{4}(x_f - x_0)^2} \\ &\approx \frac{df(x_0)}{dx} + \frac{d^2 f(x_1)}{dx^2}(x_f - x_0)\end{aligned}$$

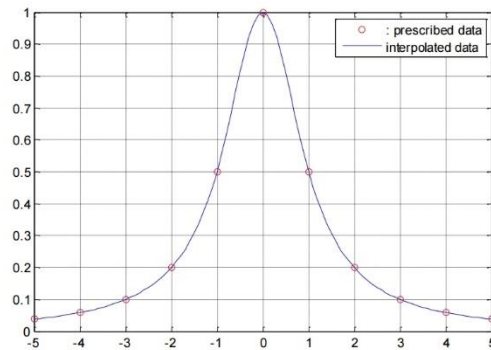
The gradient at the mid-point can be computed

$$\begin{aligned}
 f(\tau) &= (1 - 3\tau^2 + 2\tau^3)f_0 + (3\tau^2 - 2\tau^3)f_f + (x_f - x_0) \left\{ \frac{df(x_0)}{dx}(\tau - 2\tau^2 + \tau^3) + \frac{df(x_f)}{dx}(-\tau^2 + \tau^3) \right\} \\
 \frac{df(\tau)}{d\tau} &= (-6\tau + 6\tau^2)f_0 + (6\tau - 6\tau^2)f_f + (x_f - x_0) \left\{ \frac{df(x_0)}{dx}(1 - 4\tau + 3\tau^2) + \frac{df(x_f)}{dx}(-2\tau + 3\tau^2) \right\} \\
 \frac{df(t)}{dt} &= \frac{1}{(x_f - x_0)} \frac{df(\tau)}{d\tau} = \frac{(-6\tau + 6\tau^2)f_0 + (6\tau - 6\tau^2)f_f}{(x_f - x_0)} + \frac{df(x_0)}{dx}(1 - 4\tau + 3\tau^2) + \frac{df(x_f)}{dx}(-2\tau + 3\tau^2) \\
 \Rightarrow \frac{df(t_1)}{dt} &= \frac{(-6\tau_1 + 6\tau_1^2)f_0 + (6\tau_1 - 6\tau_1^2)f_f}{(x_f - x_0)} + \frac{df(x_0)}{dx}(1 - 4\tau_1 + 3\tau_1^2) + \frac{df(x_f)}{dx}(-2\tau_1 + 3\tau_1^2)
 \end{aligned}$$

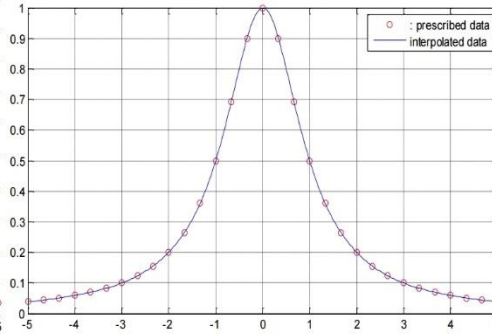
[Example#1] Application to the Modified Runge function

$$f(x) = \frac{1}{1+x^2}, \quad \frac{df(x)}{dx} = -\frac{2x}{(1+x^2)^2}$$

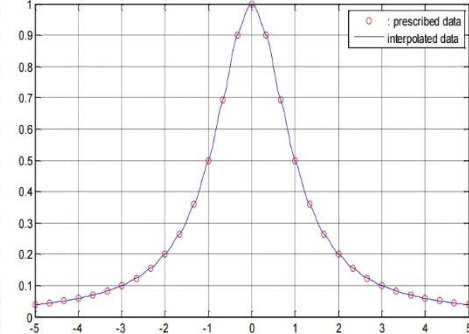
- (a) With prescribed gradients at each node : node=11
- (b) With gradients computed using Hermite algorithm at each node : node=31
- (c) With gradients computed using Central difference formula at each node : node=31



(a)



(b)

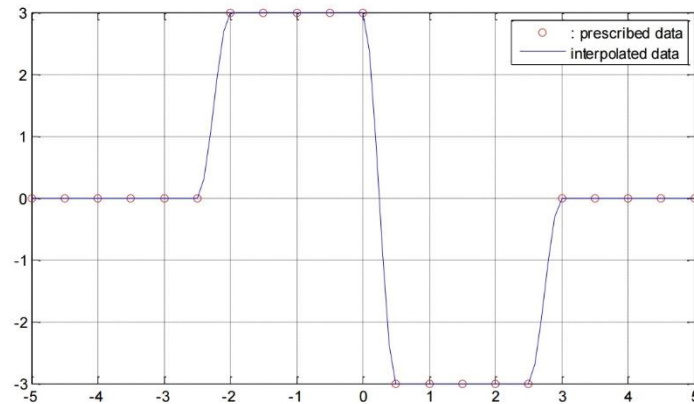


(c)

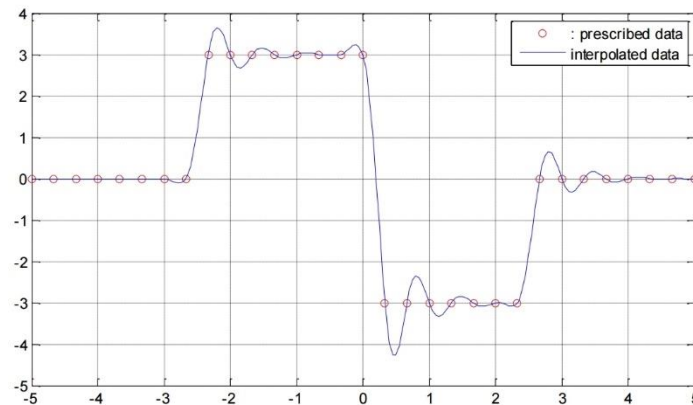


## [Example#2] Application to the Doublet function

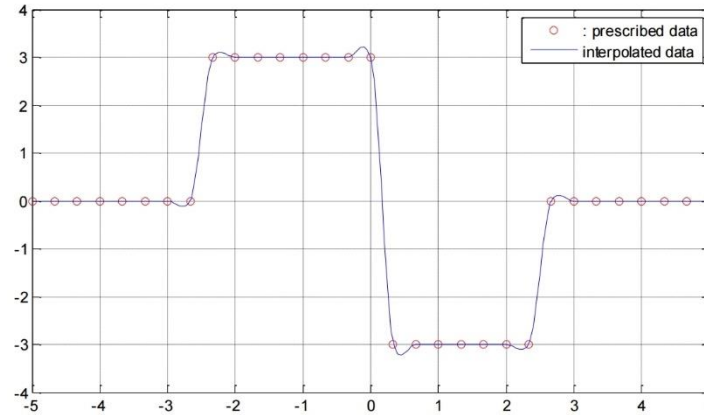
(a) With prescribed gradients at each node : node=21



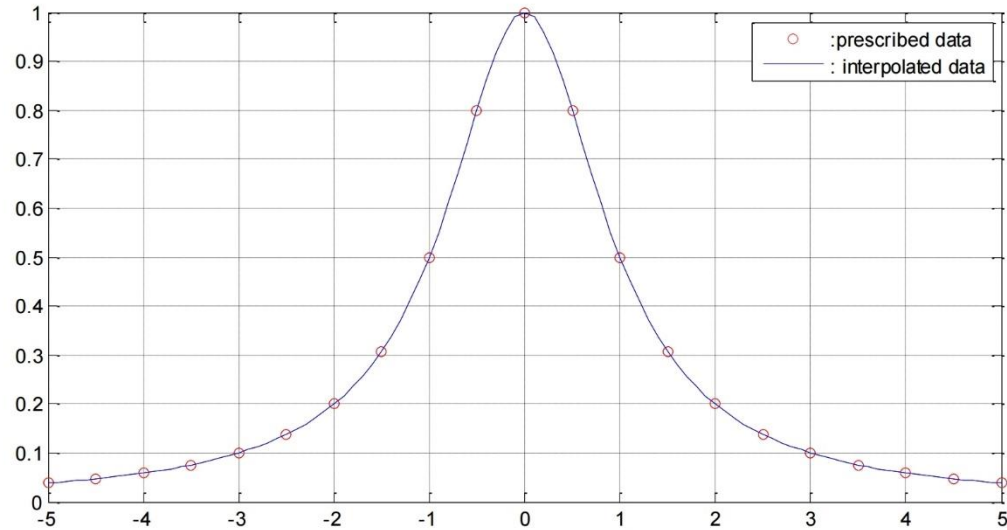
(b) With gradients computed using Hermite algorithm at each node : node=31



(c) With gradients computed using Central difference formula at each node : node=31



[Example #3] Test for modified Runge function : Quintic Hermit Polynomials



## [Example #4] Comparison with difference interpolation methods

### Piecewise Hermite Interpolation

Figure 2.5 shows some PCHIP interpolants to  $f(x) = 1/(1+x^2)$  on the interval  $[-5, 5]$ . Compare with Figures 2.1 and 2.3

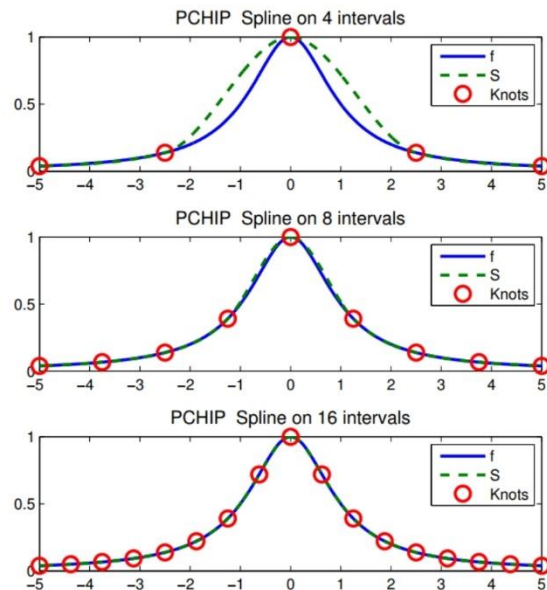
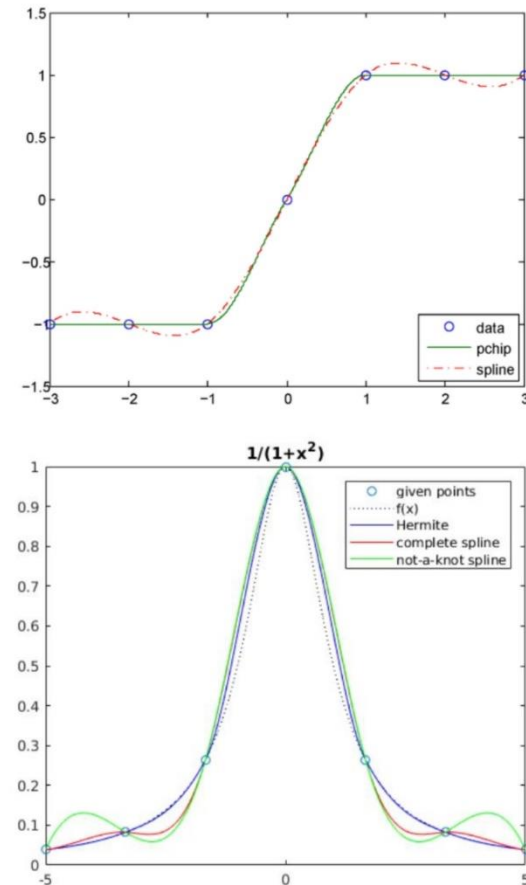
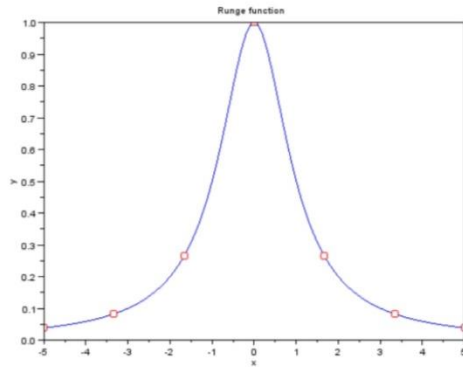


Fig. 2.5: PCHIP interpolants to  $1/(1+x^2)$ . Compare with Figure 2.3

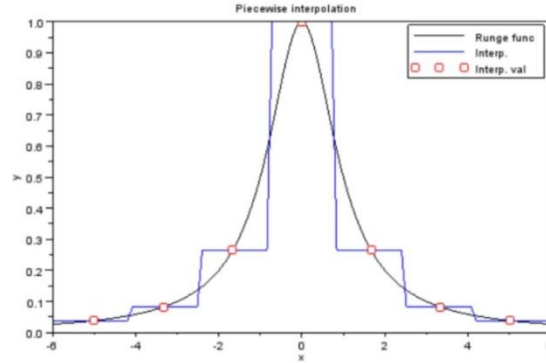
### Natural and Not-A-Knot Spline Interpolation



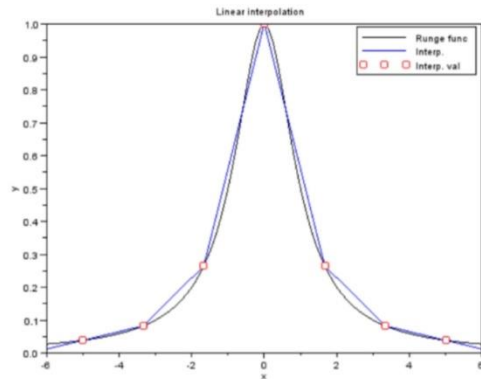
## [Example #5] Polynomial Interpolation for Runge Function



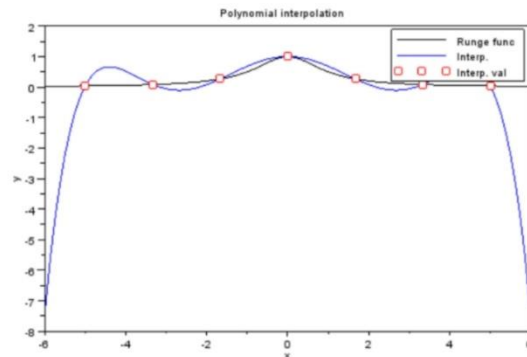
Interpolation function



Piecewise constant interpolation



Piecewise linear interpolation



Polynomial interpolation

End of Lecture