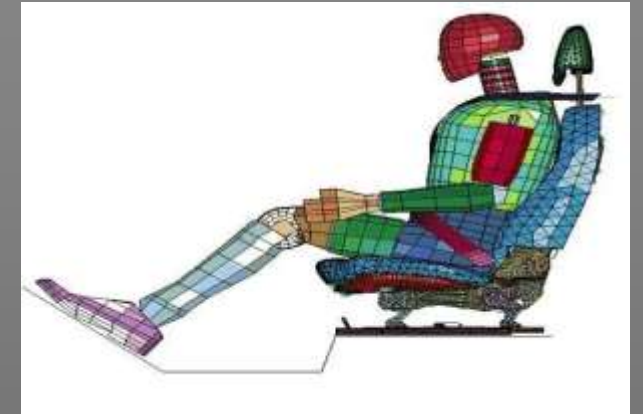
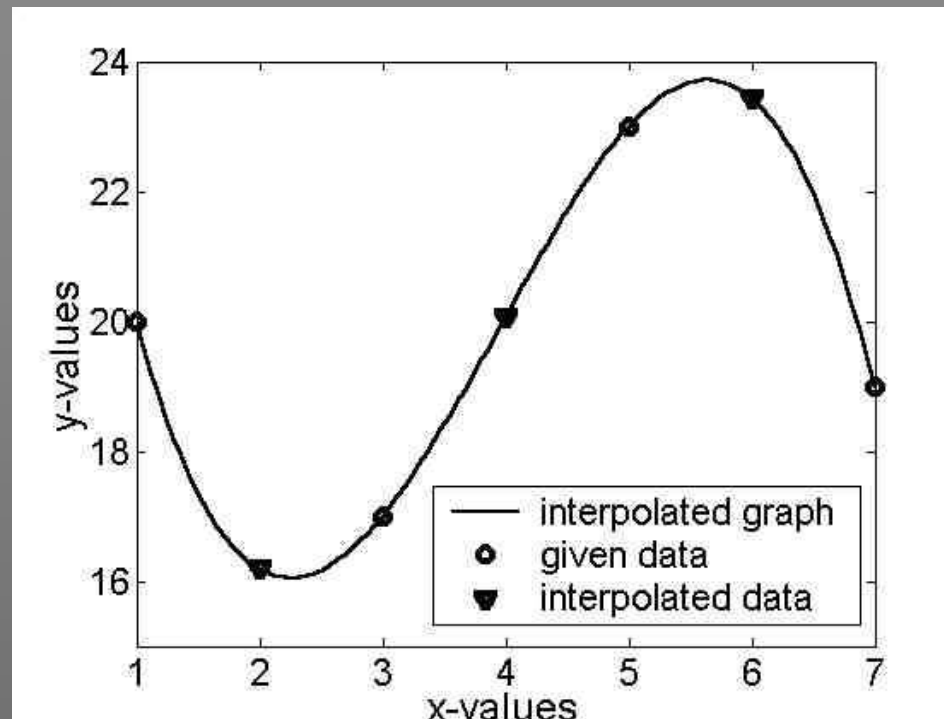
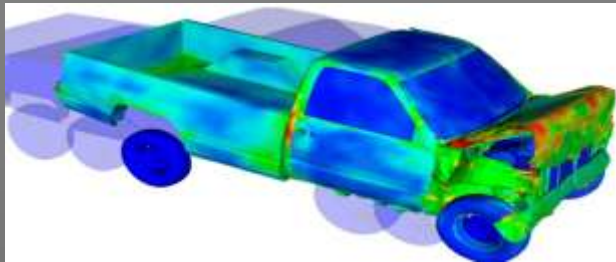
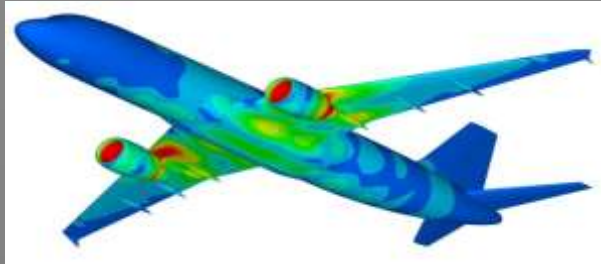


Finite Difference and Interpolation



Dr. Himanshu Pathak
himanshu@iitmandi.ac.in

Taylor Series

- ◆ Taylor series are expansions of a function $f(x)$ by some finite distance dx to $f(x+dx)$.
- ◆ In essence, the Taylor series provides a means to predict a function value at one point in terms of the function value and its derivatives at another point.
- ◆ In particular, the theorem states that any smooth function can be approximated as a polynomial.

$$f(x_{i+1}) \cong f(x_i)$$

0th order Approximation

$$f(x_{i+1}) \cong f(x_i) + f'(x_i)(x_{i+1} - x_i)$$

1st order Approximation

$$f(x_{i+1}) \cong f(x_i) + f'(x_i)(x_{i+1} - x_i) + \frac{f''(x_i)}{2!} (x_{i+1} - x_i)^2$$

2nd order Approximation

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2!} h^2 + \frac{f^{(3)}(x_i)}{3!} h^3 + \dots + \frac{f^{(n)}(x_i)}{n!} h^n + R_n$$

Nth order Approximation

$$R_n = \frac{f^{(n+1)}(\xi)}{(n+1)!} h^{n+1}$$

Remainder

Taylor Series

Example: Use zero- through fourth-order Taylor series expansions to approximate the function.

$$f(x) = -0.1x^4 - 0.15x^3 - 0.5x^2 - 0.25x + 1.2$$

$x_i = 0$ with $h = 1$. That is, predict the function's value at $x_{i+1} = 1$

By Taylor Series

0th order Approximation

$$f(x_{i+1}) \cong f(x_i)$$

$$f(x_{i+1}) \approx 1.2$$

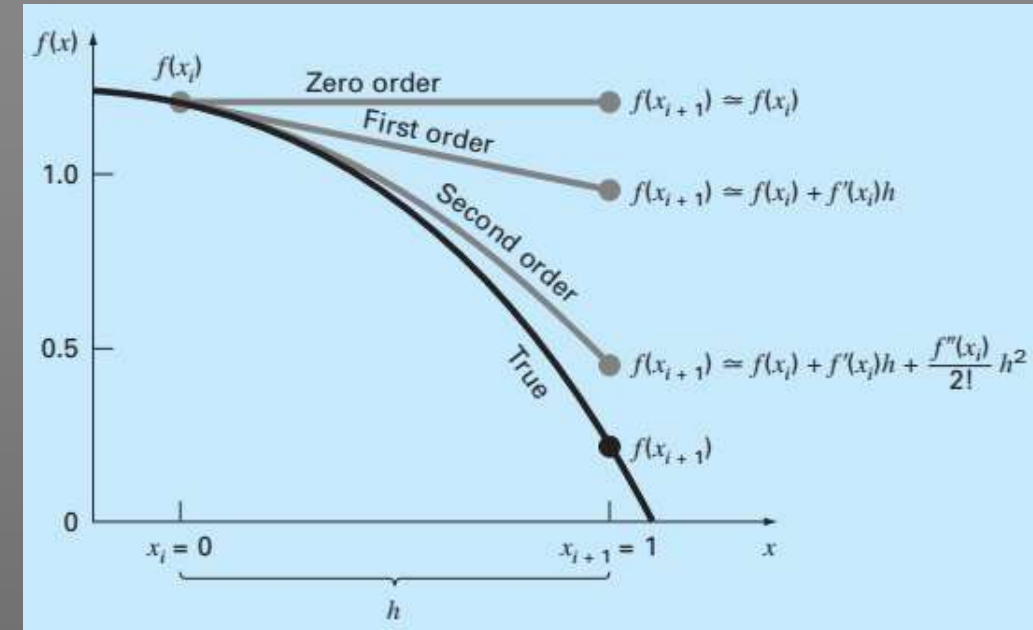
1st order Approximation

$$f(x_{i+1}) \cong f(x_i) + f'(x_i)(x_{i+1} - x_i) \quad f(1) = 0.95$$

2nd order Approximation

$$f(x_{i+1}) \cong f(x_i) + f'(x_i)(x_{i+1} - x_i) + \frac{f''(x_i)}{2!}(x_{i+1} - x_i)^2$$

$$f(1) = 0.45.$$

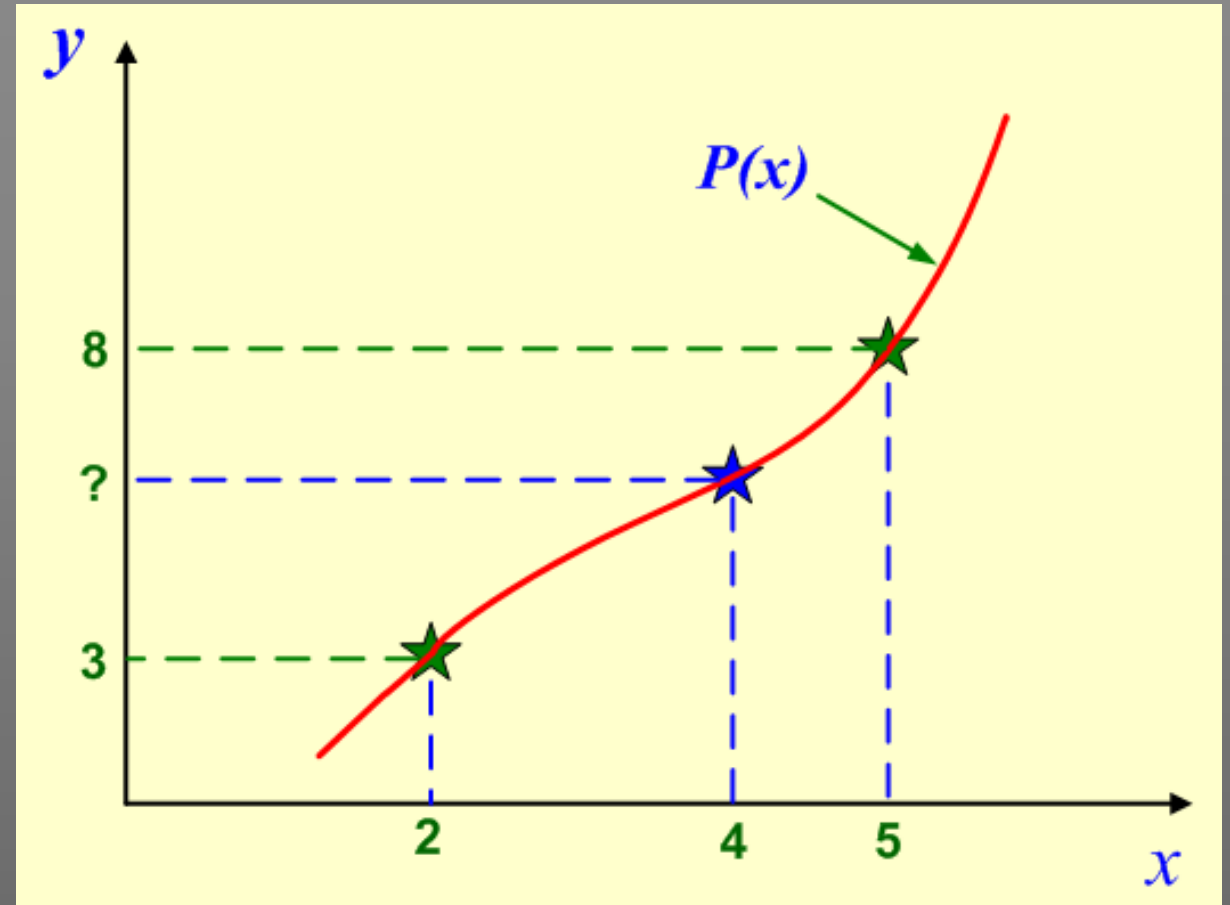


Interpolation

- ◆ Interpolation produces a function that matches the given data exactly.
- ◆ The function then can be utilized to approximate the data values at intermediate points.

Given data points: at $x_0 = 2$, $y_0 = 3$ and at $x_1 = 5$, $y_1 = 8$

Find the following: at $x = 4$, $y = ?$



Introduction

- ◆ If interpolation method provides a polynomial function with appropriate degree to exactly match a given set of data, polynomial interpolation is employed.
 - ✦ **Lagrange Interpolation**
 - ✦ **Newton Interpolation**
- ◆ Hermite interpolation method interpolate function as well as first derivative of function values.

Lagrange Interpolation

Given data points: at $x_0 = 2$, $y_0 = 3$ and at $x_1 = 5$, $y_1 = 8$

Find the following: at $x = 4$, $y = ?$ Using Lagrange interpolation technique.

$P(x)$ should satisfy the following conditions:

$P(x = 2) = 3$ and $P(x = 5) = 8$.

Assume Function

$$P(x) = 3L_0(x) + 8L_1(x)$$

$P(x)$ can satisfy the above conditions if

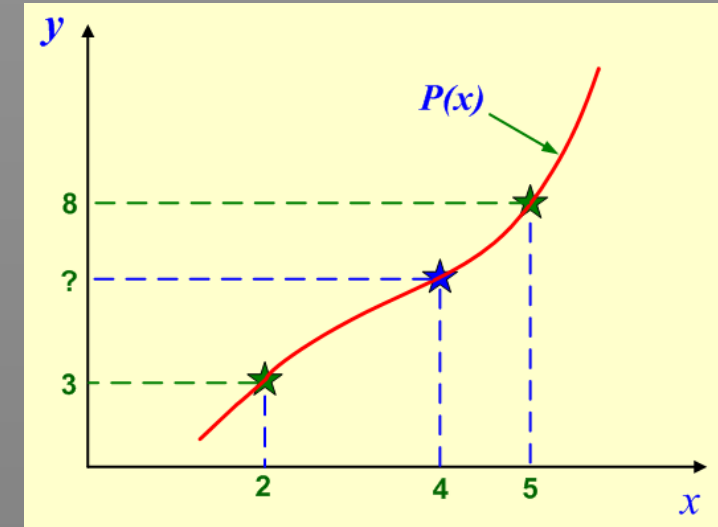
at $x = x_0 = 2$, $L_0(x) = 1$ and $L_1(x) = 0$ and

at $x = x_1 = 5$, $L_0(x) = 0$ and $L_1(x) = 1$

The conditions can be satisfied if $L_0(x)$ and $L_1(x)$ are defined in the following way:

$$L_0(x) = \frac{x-5}{2-5} \quad \text{and} \quad L_1(x) = \frac{x-2}{5-2}$$

$$L_0(x) = \frac{x-x_1}{x_0-x_1} \quad \text{and} \quad L_1(x) = \frac{x-x_0}{x_1-x_0}$$



Lagrange Interpolation Contd...

$$P(x) = 3L_0(x) + 8L_1(x) \iff L_0(x) = \frac{x - x_1}{x_0 - x_1} \quad \text{and} \quad L_1(x) = \frac{x - x_0}{x_1 - x_0}$$



$$P(x) = L_0(x)y_0 + L_1(x)y_1$$

Lagrange Interpolating Polynomial

$$P(x) = L_0(x)f(x_0) + L_1(x)f(x_1)$$

$$P(x) = \left(\frac{x - x_1}{x_0 - x_1} \right)(y_0) + \left(\frac{x - x_0}{x_1 - x_0} \right)(y_1)$$

P(x = 4) can be obtained by:

$$P(x = 4) = \left(\frac{4 - 5}{2 - 5} \right)(3) + \left(\frac{4 - 2}{5 - 2} \right)(8) \implies P(x = 4) = 6.333$$

Lagrange Interpolation Contd...

Lagrange Interpolating Polynomial for three points:

$$P(x) = \frac{(x-x_2)(x-x_3)}{(x_1-x_2)(x_1-x_3)} y_1 + \frac{(x-x_1)(x-x_3)}{(x_2-x_1)(x_2-x_3)} y_2 + \frac{(x-x_1)(x-x_2)}{(x_3-x_1)(x_3-x_2)} y_3$$

$$P(x) = L_1 y_1 + L_2 y_2 + \dots + L_n y_n$$

$L_k(x)$ can be defined by:

$$L_k(x) = \frac{(x-x_1) \dots (x-x_{k-1})(x-x_{k+1}) \dots (x-x_n)}{(x_k-x_1) \dots (x_k-x_{k-1})(x_k-x_{k+1}) \dots (x_k-x_n)}$$

Numerator $N_k(x) = (x-x_1) \dots (x-x_{k-1})(x-x_{k+1}) \dots (x-x_n)$

Denominator $D_k(x) = (x_k-x_1) \dots (x_k-x_{k-1})(x_k-x_{k+1}) \dots (x_k-x_n)$

MATLAB[®] Script

Simplest MATLAB Program for Coefficient of Lagrange Polynomial

clear all

clc

x = input('Define x vector = ');

y = input('Define y vector = ');

n=length(x);

for k = 1:n

d(k) = 1;

for i = 1: n

if i ~= k

*d(k) = d(k) * (x(k) - x(i));*

end

c(k) = y (k) / d(k);

end

end

$$P(x) = c_1 N_1 + c_2 N_2 + \dots + c_n N_n$$

$$c_k = \frac{y_k}{D_k} = \frac{y_k}{(x_k - x_1) \dots (x_k - x_{k-1})(x_k - x_{k+1}) \dots (x_k - x_n)}$$

$$N_k(x) = (x - x_1) \dots (x - x_{k-1})(x - x_{k+1}) \dots (x - x_n)$$

MATLAB[®] Script

Simplest MATLAB Program for Lagrange Polynomial

t=input('Enter point x at which polynomial get evaluated = ');

for i = 1:length(t)

p(i) = 0;

for j = 1:n

N(j) = 1;

for k = 1:n

if j ~= k

*N(j) = N(j) * (t(i) - x(k));*

end

end

*p(i) = p(i) + N(j) * c(j);*

end

end

$$P(x) = c_1 N_1 + c_2 N_2 + \dots + c_n N_n$$

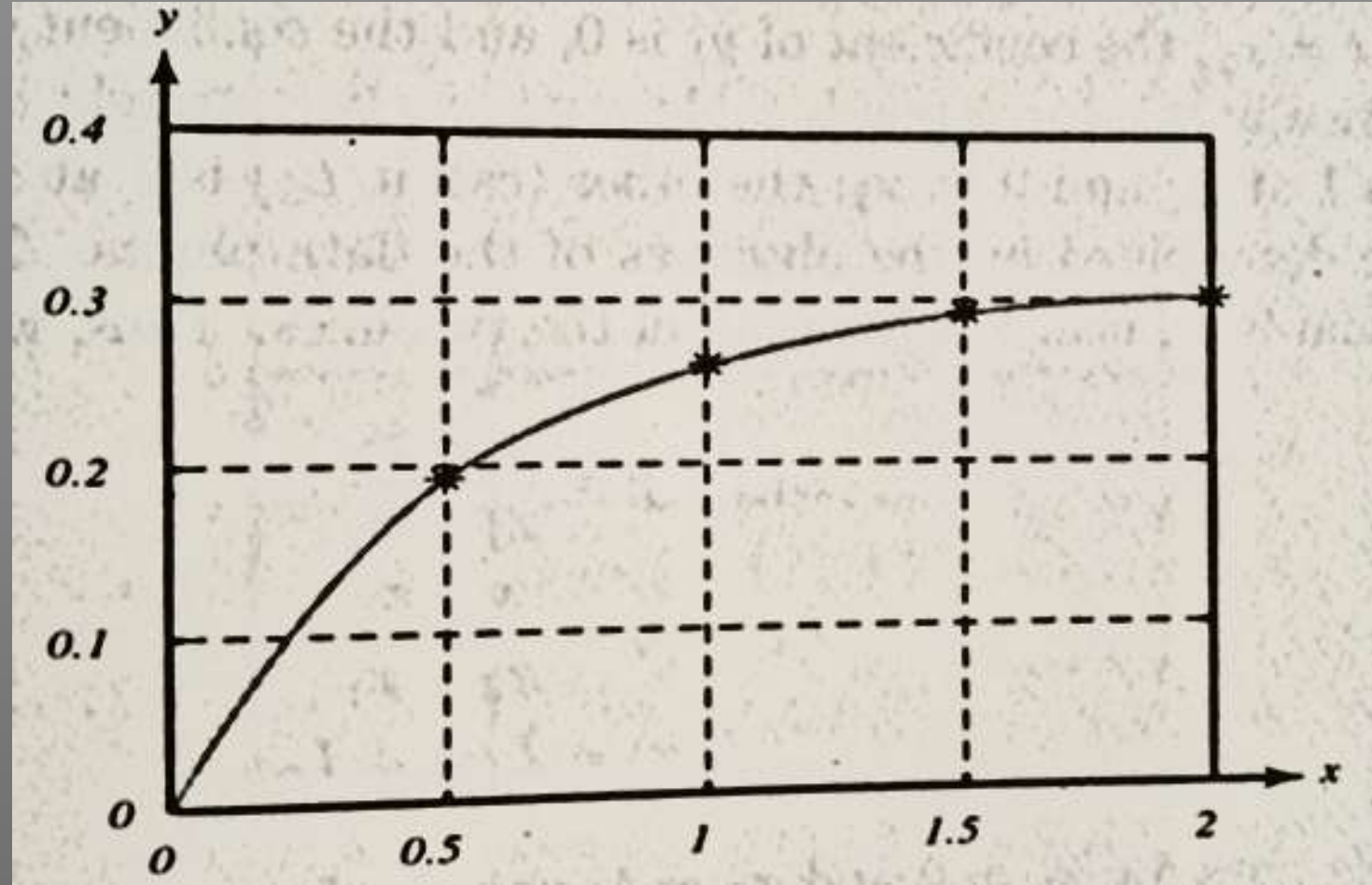
$$c_k = \frac{y_k}{D_k} = \frac{y_k}{(x_k - x_1) \dots (x_k - x_{k-1})(x_k - x_{k+1}) \dots (x_k - x_n)}$$

$$N_k(x) = (x - x_1) \dots (x - x_{k-1})(x - x_{k+1}) \dots (x - x_n)$$

p=interp1([x], [y], [t])

MATLAB® Script

Comparision of result obtained by developed and in-built function



$p=\text{interp1}([x], [y], [t])$

$p=\text{interp1}([0 \ 0.5 \ 1 \ 1.5 \ 2],[0 \ 0.19 \ 0.26 \ 0.29 \ 0.31],[0.75 \ 1.25])$

Why Newton Interpolation

- ✌ The Lagrange formula is popular because it is well known and is easy to code.
- ✌ Also, the data are not required to be specified with x in ascending or descending order.
- ✌ **If we decide to add a point to the set of nodes, we have to completely re-compute all of the polynomial functions.**
- 😊 Here we introduce an alternative form of the polynomial: the Newton form (Divided-Difference Interpolating Polynomials).

Introduction

- ◆ Newton interpolation assume a complete polynomial starts from lower degree to higher degree :

- ✦ **For two point Interpolation** (x_1, y_1) and (x_2, y_2)

$$P(x) = a_1 + a_2(x - x_1)$$

- ✦ **For three point Interpolation** (x_1, y_1) , (x_2, y_2) and (x_3, y_3)

$$P(x) = a_1 + a_2(x - x_1) + a_3(x - x_1)(x - x_2)$$

- ✦ **For n-point Interpolation** (x_1, y_1) , (x_2, y_2) , ..., (x_n, y_n)

$$P(x) = a_1 + a_2(x - x_1) + a_3(x - x_1)(x - x_2) + \dots + a_n(x - x_1) \dots (x - x_{n-1})$$

Newton Interpolation

Given data points: $(x_1, y_1), (x_2, y_2)$ and (x_3, y_3)

Find the coefficient of polynomial.

$P(x)$ for 3 data set: $P(x) = a_1 + a_2(x - x_1) + a_3(x - x_1)(x - x_2)$

At $x = x_1$ and $y = y_1$ $P(x) = y_1 = a_1 + a_2(x_1 - x_1) + a_3(x_1 - x_1)(x_1 - x_2)$

$$y_1 = a_1$$

At $x = x_2$ and $y = y_2$ $P(x) = y_2 = a_1 + a_2(x_2 - x_1) + a_3(x_2 - x_1)(x_2 - x_2)$

$$y_2 = y_1 + a_2(x_2 - x_1) \implies a_2 = \frac{y_2 - y_1}{x_2 - x_1}$$

At $x = x_3$ and $y = y_3$

$$a_3 = \frac{\frac{y_3 - y_2}{x_3 - x_2} - \frac{y_2 - y_1}{x_2 - x_1}}{x_3 - x_1}$$

Newton Interpolation Example

For given data points: $(x_1, y_1) = (-2, 4); (x_2, y_2) = (0, 2)$ and $(x_3, y_3) = (2, 8)$

Find the Newton interpolation polynomial and value at $x = 1$.

For 3-data set: $P(x) = a_1 + a_2(x - x_1) + a_3(x - x_1)(x - x_2)$

Coefficients of polynomial: $a_1 = y_1 = 4$

$$a_2 = \frac{y_2 - y_1}{x_2 - x_1} = \frac{2 - 4}{0 - (-2)} = -1$$

$$a_3 = \frac{\frac{y_3 - y_2}{x_3 - x_2} - \frac{y_2 - y_1}{x_2 - x_1}}{x_3 - x_1} = \frac{\frac{8 - 2}{2 - 0} - \frac{2 - 4}{0 - (-2)}}{2 - (-2)} = 1$$

Put these value in polynomial $P(x) = 4 - (x - x_1) + (x - x_1)(x - x_2)$

Impose co-ordinate in polynomial $P(x) = 4 - (x - (-2)) + (x - (-2))(x - 0)$

$$P(x) = 4 - (x + 2) + (x + 2)x \implies P(x) = x^2 + x + 2 \implies P(1) = 4$$

MATLAB[®] Script

MATLAB Program for Coefficient of Newton Polynomial

```
clear all  
clc  
x = input('Define x vector = ');  
y = input('Define y vector = ');  
n=length(x);  
a(1) = y(1);  
for k = 1 : n - 1  
    d(k, 1) = (y(k+1) - y(k))/(x(k+1) - x(k));  
end  
for j = 2 : n - 1  
    for k = 1 : n - j  
        d(k, j) = (d(k+1, j - 1) - d(k, j - 1))/(x(k+j) - x(k));  
    end  
end
```

```
for j = 2 : n  
    a(j) = d(1, j-1);  
end
```

$$a_1 = y_1$$

$$a_2 = \frac{y_2 - y_1}{x_2 - x_1}$$

$$a_3 = \frac{\frac{y_3 - y_2}{x_3 - x_2} - \frac{y_2 - y_1}{x_2 - x_1}}{x_3 - x_1}$$

MATLAB[©] Script

MATLAB Program for Newton Interpolation

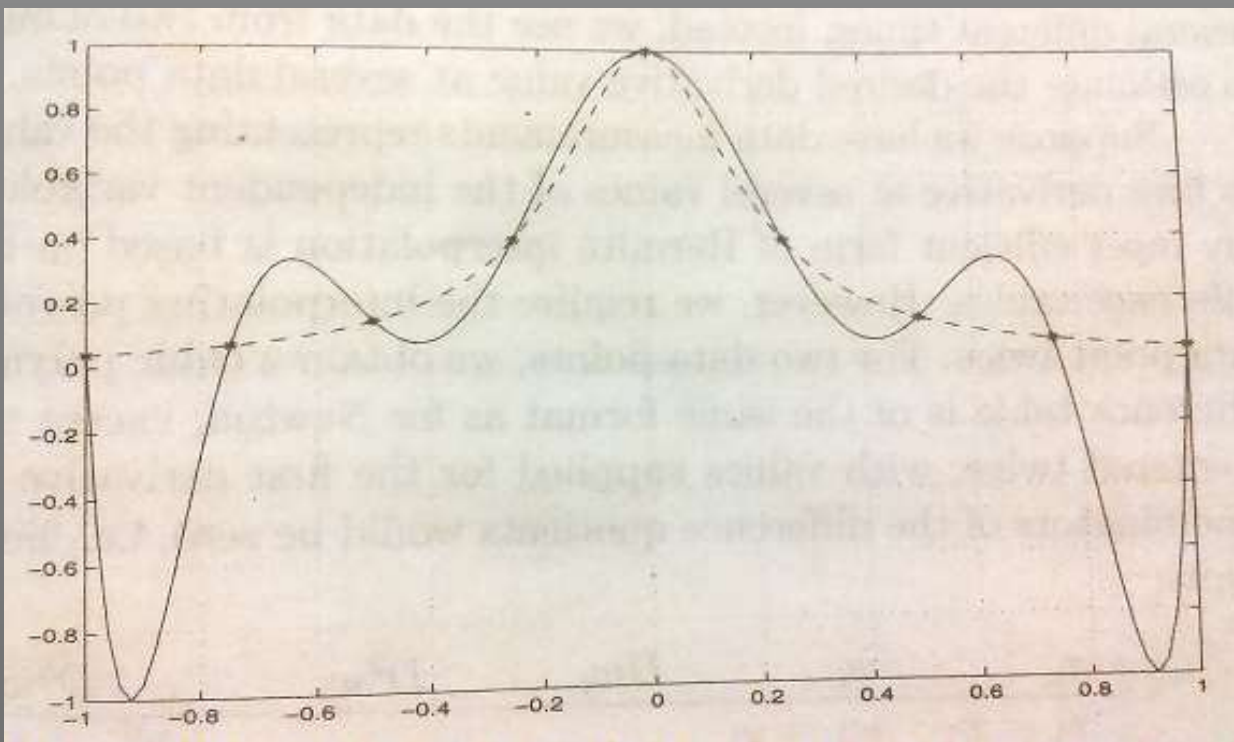
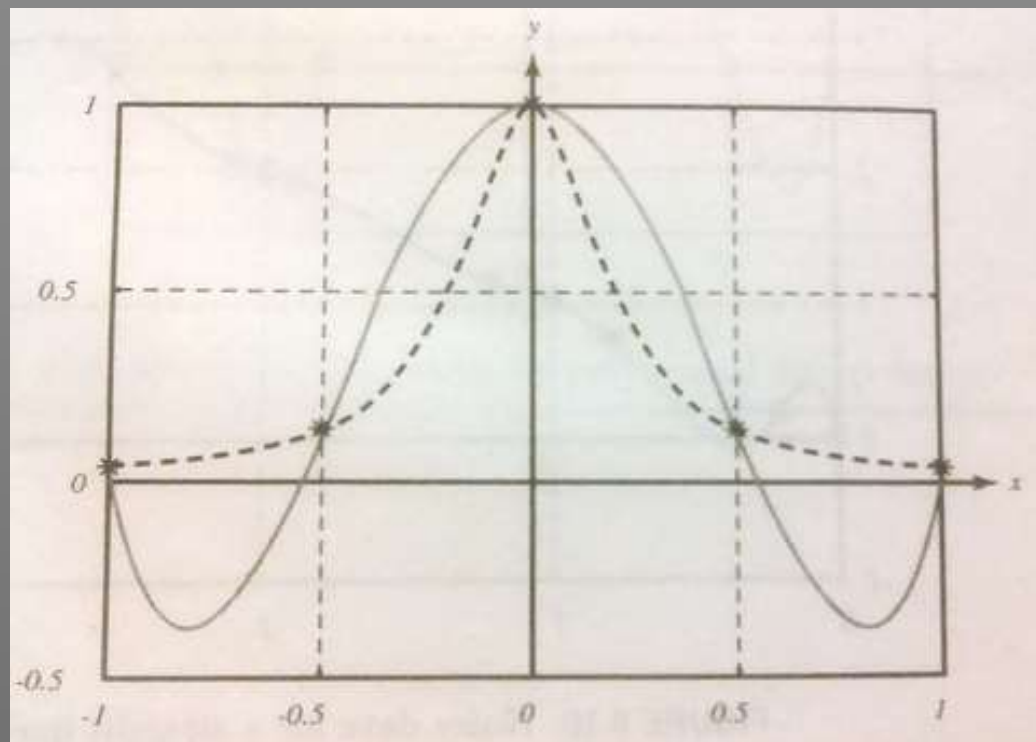
```
t=input('Enter point x at which polynomial get evaluated = ');  
for i = 1:length(t)  
    d(1) = 1;  
    N = a(1);  
    for j = 2:n  
        d(j) = (t(i) - x(j-1)) * d(j-1);  
        N(i) = N(i) + a(j) * d(j);  
    end  
end
```

$$P(x) = a_1 + a_2(x - x_1) + a_3(x - x_1)(x - x_2)$$

Piecewise Interpolation

Runge Function

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



Piecewise Interpolation

*Picewise linear interpolation
for four data points:*

$$(x_1, y_1), (x_2, y_2), (x_3, y_3) \text{ and } (x_4, y_4)$$

Provided

$$x_1 < x_2 < x_3 < x_4$$

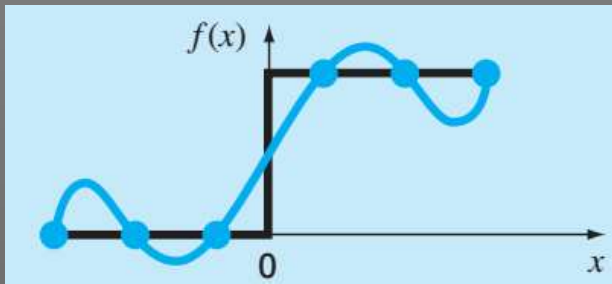
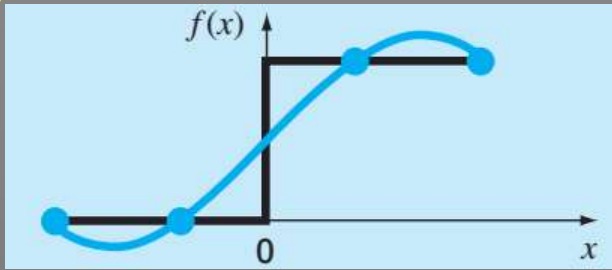
Sub-domain

$$I_1 = [x_1, x_2], \quad I_2 = [x_2, x_3], \quad I_3 = [x_3, x_4]$$

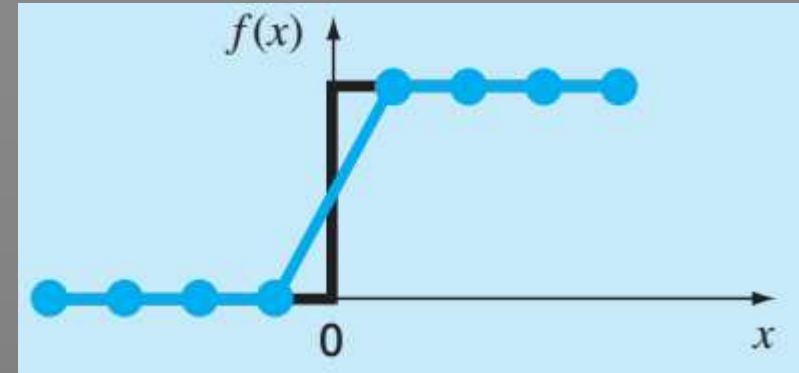
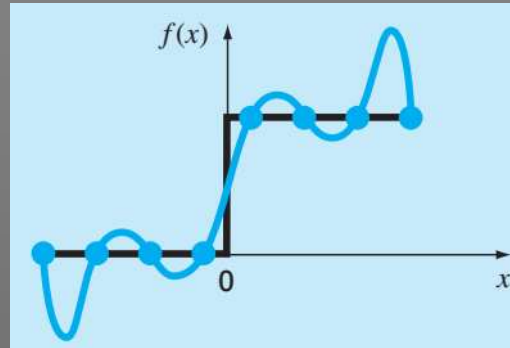
$$P(x) = \begin{cases} \left(\frac{x - x_2}{x_1 - x_2} \right) (y_1) + \left(\frac{x - x_1}{x_2 - x_1} \right) (y_2); & x_1 \leq x < x_2 \\ \left(\frac{x - x_3}{x_2 - x_3} \right) (y_2) + \left(\frac{x - x_2}{x_3 - x_2} \right) (y_3); & x_2 \leq x < x_3 \\ \left(\frac{x - x_4}{x_3 - x_4} \right) (y_3) + \left(\frac{x - x_3}{x_4 - x_3} \right) (y_4); & x_3 \leq x \leq x_4 \end{cases}$$

Spline Interpolation

- ◆ A spline is a special function defined piecewise by polynomials.
- ◆ In essence, spline interpolation is a special type of polynomial interpolation where each section/range is interpolated by different polynomial functions.



Polynomial Interpolation



Linear Spline Interpolation

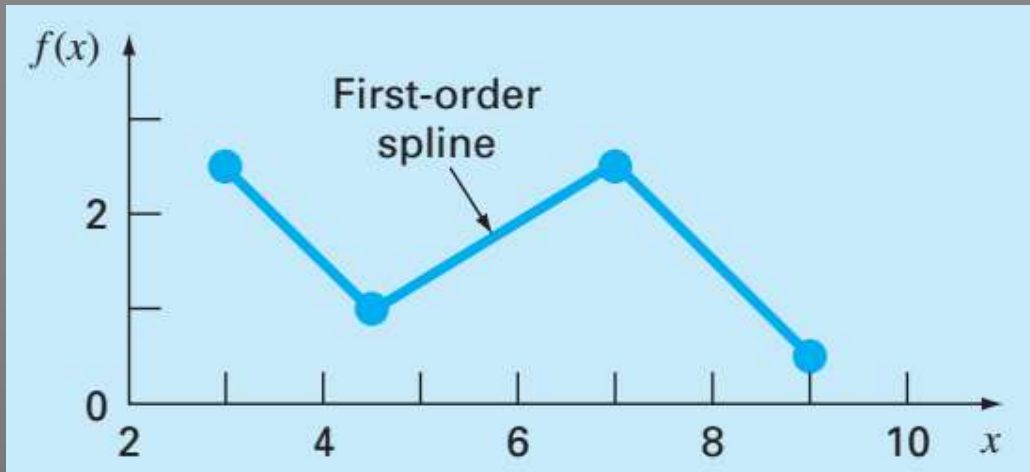
$$\begin{aligned} f(x) &= f(x_0) + m_0(x - x_0) & x_0 \leq x \leq x_1 \\ f(x) &= f(x_1) + m_1(x - x_1) & x_1 \leq x \leq x_2 \\ &\vdots \\ &\vdots \\ &\vdots \\ f(x) &= f(x_{n-1}) + m_{n-1}(x - x_{n-1}) & x_{n-1} \leq x \leq x_n \end{aligned}$$

$$m_i = \frac{f(x_{i+1}) - f(x_i)}{x_{i+1} - x_i}$$

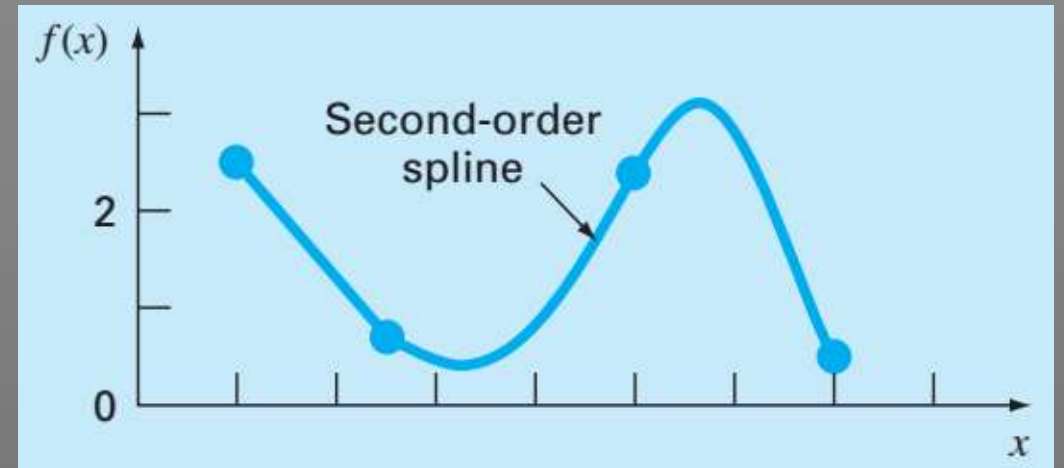
Quadratic Spline Interpolation

- ◆ A spline interpolation using second-order polynomials.
- ◆ In essence, quadratic splines have continuous first derivatives at the knots.

$$f_i(x) = a_i x^2 + b_i x + c_i$$



Linear Spline Interpolation



Quadratic Spline Interpolation

Quadratic Spline Interpolation

Quadratic Spline interpolation for four data points:

Polynomial for each interval: $f_i(x) = a_i x^2 + b_i x + c_i$

For $n+1$ data points, there are n intervals and, $3n$ unknown constants to be found for interpolation functions.

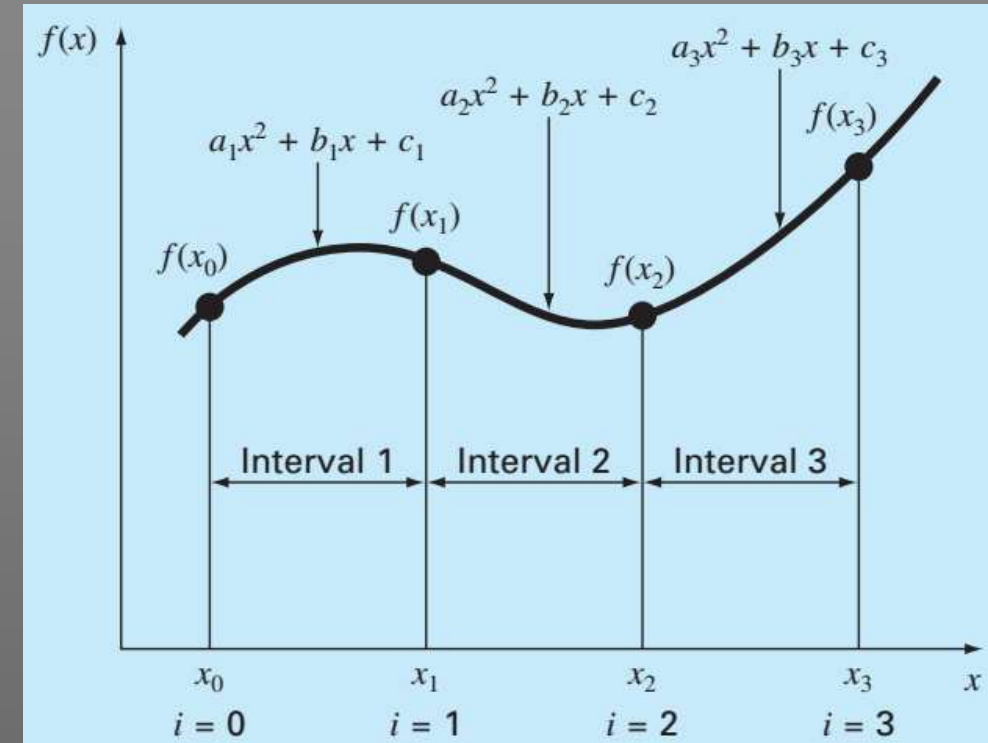
Interpolation conditions ($3n$):

01. The function values of adjacent polynomials must be equal at the interior knots.

$$\begin{aligned} a_{i-1}x_{i-1}^2 + b_{i-1}x_{i-1} + c_{i-1} &= f(x_{i-1}) \\ a_i x_{i-1}^2 + b_i x_{i-1} + c_i &= f(x_{i-1}) \end{aligned}$$

02. The first and last functions must pass through the end points.

$$\begin{aligned} a_1 x_0^2 + b_1 x_0 + c_1 &= f(x_0) \\ a_n x_n^2 + b_n x_n + c_n &= f(x_n) \end{aligned}$$



Quadratic Spline Interpolation

Quadratic Spline Interpolation

For $n+1$ data points, there are n intervals and, $3n$ unknown constants to be found for interpolation functions.

Interpolation conditions continue...

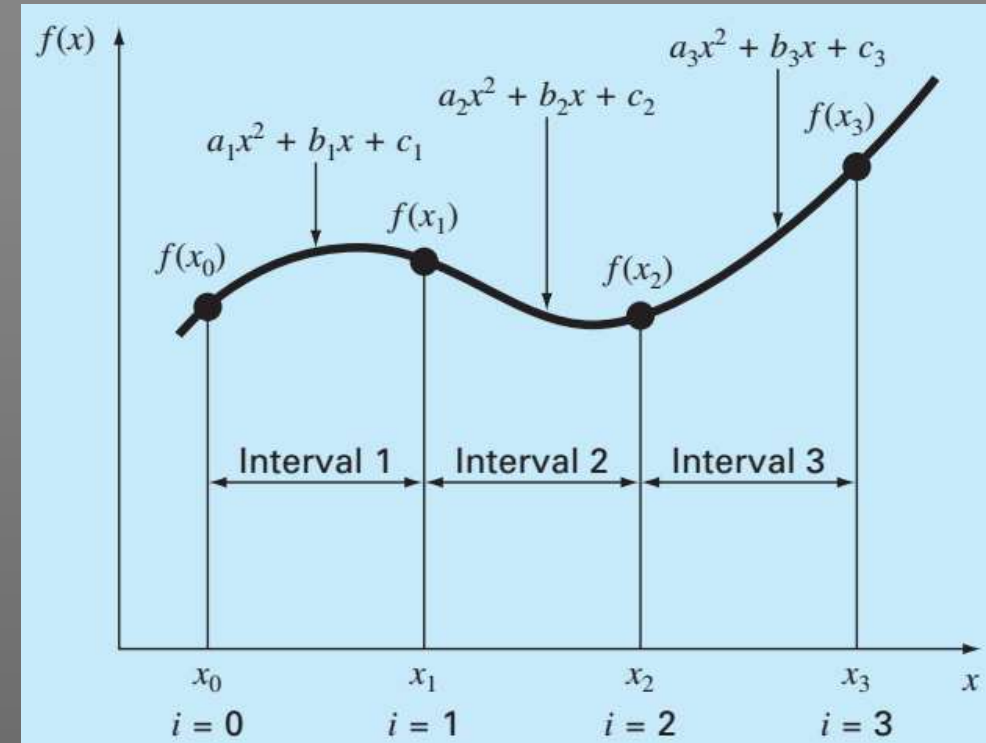
03. The first derivatives at the interior knots must be equal.

$$f'(x) = 2ax + b$$

$$2a_{i-1}x_{i-1} + b_{i-1} = 2a_ix_{i-1} + b_i$$

04. Assume that the second derivative is zero at the first point.

$$a_1 = 0$$



Quadratic Spline Interpolation

Quadratic Spline Interpolation

Given data points:

Data to be fit with spline functions.

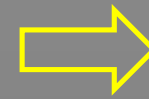
x	$f(x)$
3.0	2.5
4.5	1.0
7.0	2.5
9.0	0.5

We have four data points and $n = 3$ intervals.

Therefore, $3 \times 3 = 9$ unknowns must be determined.

$$a_{i-1}x_{i-1}^2 + b_{i-1}x_{i-1} + c_{i-1} = f(x_{i-1})$$

$$a_i x_{i-1}^2 + b_i x_{i-1} + c_i = f(x_{i-1})$$



$$20.25a_1 + 4.5b_1 + c_1 = 1.0$$

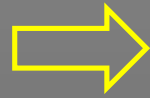
$$20.25a_2 + 4.5b_2 + c_2 = 1.0$$

$$49a_2 + 7b_2 + c_2 = 2.5$$

$$49a_3 + 7b_3 + c_3 = 2.5$$

$$a_1 x_0^2 + b_1 x_0 + c_1 = f(x_0)$$

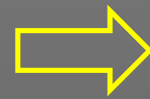
$$a_n x_n^2 + b_n x_n + c_n = f(x_n)$$



$$9a_1 + 3b_1 + c_1 = 2.5$$

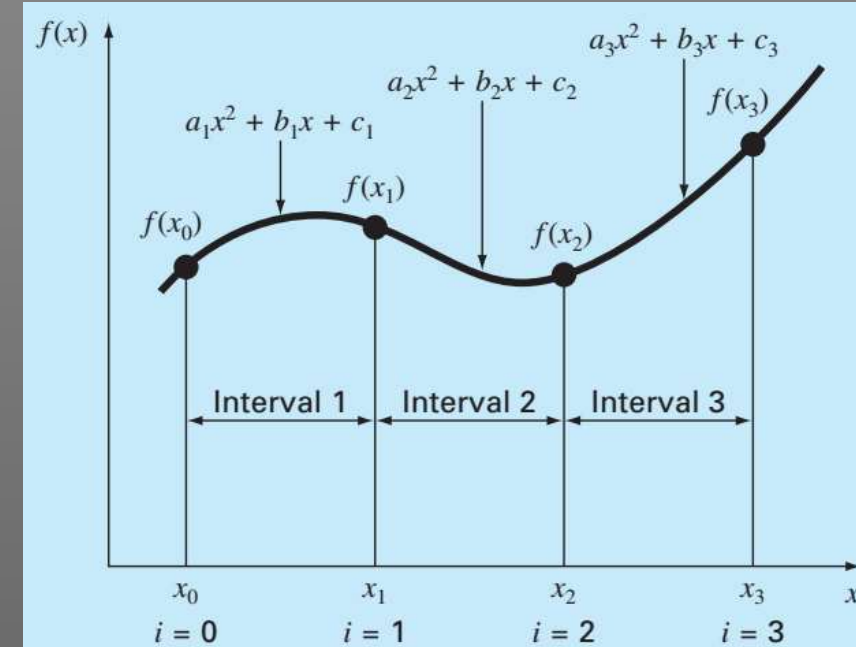
$$81a_3 + 9b_3 + c_3 = 0.5$$

$$2a_{i-1}x_{i-1} + b_{i-1} = 2a_i x_{i-1} + b_i$$



$$9a_1 + b_1 = 9a_2 + b_2$$

$$14a_2 + b_2 = 14a_3 + b_3$$



Quadratic Spline Interpolation

Quadratic Spline Interpolation

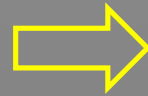
Obtained Equations:

$$20.25a_1 + 4.5b_1 + c_1 = 1.0$$

$$20.25a_2 + 4.5b_2 + c_2 = 1.0$$

$$49a_2 + 7b_2 + c_2 = 2.5$$

$$49a_3 + 7b_3 + c_3 = 2.5$$



$$\begin{bmatrix} 4.5 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 20.25 & 4.5 & 1 & 0 & 0 & 0 \\ 0 & 0 & 49 & 7 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 49 & 7 & 1 \\ 3 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 81 & 9 & 1 \\ 1 & 0 & -9 & -1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 14 & 1 & 0 & -14 & -1 & 0 \end{bmatrix} \begin{Bmatrix} b_1 \\ c_1 \\ a_2 \\ b_2 \\ c_2 \\ a_3 \\ b_3 \\ c_3 \end{Bmatrix} = \begin{Bmatrix} 1 \\ 1 \\ 2.5 \\ 2.5 \\ 2.5 \\ 0.5 \\ 0 \\ 0 \end{Bmatrix}$$

$$9a_1 + 3b_1 + c_1 = 2.5$$

$$81a_3 + 9b_3 + c_3 = 0.5$$

$$9a_1 + b_1 = 9a_2 + b_2$$

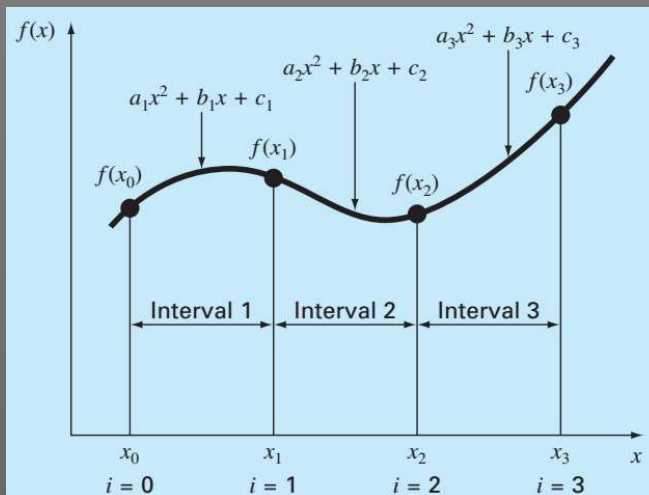
$$14a_2 + b_2 = 14a_3 + b_3$$



$$\begin{array}{lll} a_1 = 0 & b_1 = -1 & c_1 = 5.5 \\ a_2 = 0.64 & b_2 = -6.76 & c_2 = 18.46 \\ a_3 = -1.6 & b_3 = 24.6 & c_3 = -91.3 \end{array}$$



$$\begin{array}{ll} f_1(x) = -x + 5.5 & 3.0 \leq x \leq 4.5 \\ f_2(x) = 0.64x^2 - 6.76x + 18.46 & 4.5 \leq x \leq 7.0 \\ f_3(x) = -1.6x^2 + 24.6x - 91.3 & 7.0 \leq x \leq 9.0 \end{array}$$



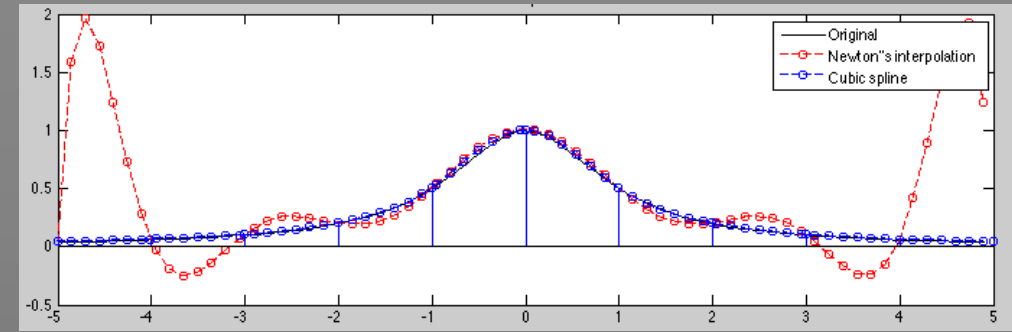
Quadratic Spline
Interpolation

Cubic Spline Interpolation

To define third-order polynomial for each interval between knots:

Polynomial for each interval: $f_i(x) = a_i x^3 + b_i x^2 + c_i x + d_i$

For $n+1$ data points, there are n intervals and, $4n$ unknown constants to be find for interpolation functions.



Cubic Spline Interpolation

Interpolation conditions (4n):

- 1. The function values of adjacent polynomials must be equal at the interior knots.*
- 2. The first and last functions must pass through the end points.*
- 3. The first derivatives at the interior knots must be equal.*
- 4. The second derivatives at the interior knots must be equal.*
- 5. The second derivatives at the end knots are zero (natural spline).*

Cubic Spline Interpolation

Interpolation equations:

$$f_i''(x) = f_i''(x_{i-1}) \frac{x - x_i}{x_{i-1} - x_i} + f_i''(x_i) \frac{x - x_{i-1}}{x_i - x_{i-1}}$$

$$\begin{aligned} f_i(x) = & \frac{f_i''(x_{i-1})}{6(x_i - x_{i-1})} (x_i - x)^3 + \frac{f_i''(x_i)}{6(x_i - x_{i-1})} (x - x_{i-1})^3 \\ & + \left[\frac{f(x_{i-1})}{x_i - x_{i-1}} - \frac{f_i''(x_{i-1})(x_i - x_{i-1})}{6} \right] (x_i - x)^2 \\ & + \left[\frac{f(x_i)}{x_i - x_{i-1}} - \frac{f_i''(x_i)(x_i - x_{i-1})}{6} \right] (x - x_{i-1})^2 \end{aligned}$$

To find $f''(x)$

$$\begin{aligned} & (x_i - x_{i-1})f''(x_{i-1}) + 2(x_{i+1} - x_{i-1})f''(x_i) + (x_{i+1} - x_i)f''(x_{i+1}) \\ & = \frac{6}{x_{i+1} - x_i} [f(x_{i+1}) - f(x_i)] + \frac{6}{x_i - x_{i+1}} [f(x_{i+1}) - f(x_i)] \end{aligned}$$

Cubic Spline Interpolation

Given data points:

Data to be fit with spline functions.

x	$f(x)$
3.0	2.5
4.5	1.0
7.0	2.5
9.0	0.5

We have four data points and $n = 3$ intervals. Therefore, $4 \times 3 = 12$ unknowns must be determined.

$$\begin{array}{ll} x_0 = 3 & f(x_0) = 2.5 \\ x_1 = 4.5 & f(x_1) = 1 \\ x_2 = 7 & f(x_2) = 2.5 \end{array}$$

To find $f''(x)$

$$\begin{aligned} & (x_i - x_{i-1})f''(x_{i-1}) + 2(x_{i+1} - x_{i-1})f''(x_i) + (x_{i+1} - x_i)f''(x_{i+1}) \\ &= \frac{6}{x_{i+1} - x_i}[f(x_{i+1}) - f(x_i)] + \frac{6}{x_i - x_{i+1}}[f(x_{i+1}) - f(x_i)] \end{aligned}$$

$$\begin{aligned} & (4.5 - 3)f''(3) + 2(7 - 3)f''(4.5) + (7 - 4.5)f''(7) \\ &= \frac{6}{7 - 4.5}(2.5 - 1) + \frac{6}{4.5 - 3}(2.5 - 1) \end{aligned}$$

$$f''(3) = 0$$

$$\begin{aligned} 8f''(4.5) + 2.5f''(7) &= 9.6 \\ 2.5f''(4.5) + 9f''(7) &= -9.6 \end{aligned}$$

$$\begin{aligned} f''(4.5) &= 1.67909 \\ f''(7) &= -1.53308 \end{aligned}$$

Cubic Spline Interpolation

To find $f_i(x)$

$$\begin{aligned} f_i(x) = & \frac{f_i''(x_{i-1})}{6(x_i - x_{i-1})}(x_i - x)^3 + \frac{f_i''(x_i)}{6(x_i - x_{i-1})}(x - x_{i-1})^3 \\ & + \left[\frac{f(x_{i-1})}{x_i - x_{i-1}} - \frac{f''(x_{i-1})(x_i - x_{i-1})}{6} \right](x_i - x) \\ & + \left[\frac{f(x_i)}{x_i - x_{i-1}} - \frac{f''(x_i)(x_i - x_{i-1})}{6} \right](x - x_{i-1}) \end{aligned}$$


$$f_1(x) = 0.186566(x - 3)^3 + 1.666667(4.5 - x) + 0.246894(x - 3)$$

$$\begin{aligned} f_2(x) = & 0.111939(7 - x)^3 - 0.102205(x - 4.5)^3 - 0.299621(7 - x) \\ & + 1.638783(x - 4.5) \end{aligned}$$

$$f_3(x) = -0.127757(9 - x)^3 + 1.761027(9 - x) + 0.25(x - 7)$$

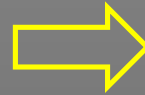
Finite Divided Difference Method

- ◆ Divided differences is a recursive division process. The method can be used to calculate the coefficients in the interpolation polynomial in the Newton form.
- ◆ The Taylor series used to approximate divided differences.

Taylor Series

$$f(x_{i+1}) = f(x_i) + f'(x_i)h + \frac{f''(x_i)}{2!}h^2 + \frac{f^{(3)}(x_i)}{3!}h^3 + \dots + \frac{f^{(n)}(x_i)}{n!}h^n + R_n$$

$$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{x_{i+1} - x_i} + O(x_{i+1} - x_i)$$



$$f'(x_i) = \frac{\Delta f_i}{h} + O(h)$$

First Forward Difference

Δf_i is referred to as the first forward difference and h is called the step size

Finite Divided Difference Method

$$\partial_x f^+ \approx \frac{f(x+dx) - f(x)}{dx} \quad \text{forward difference}$$

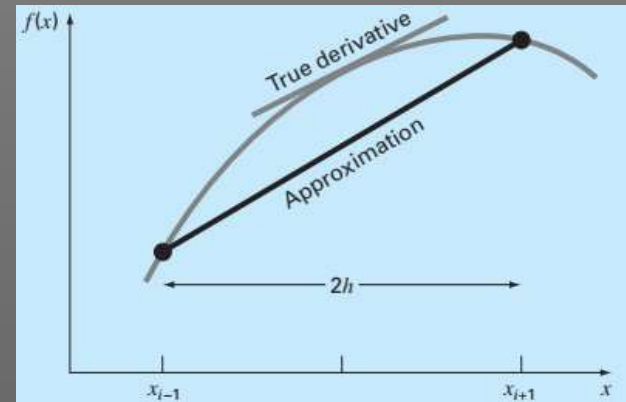
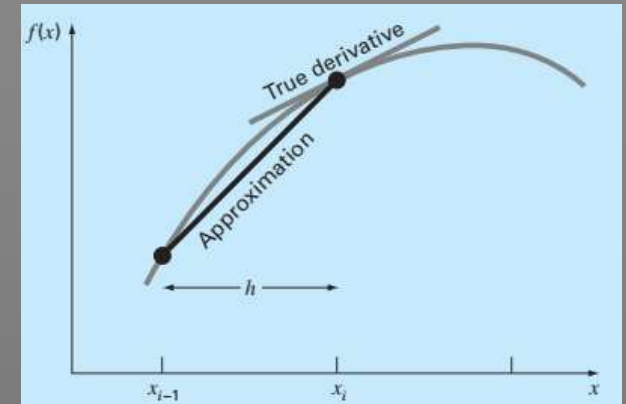
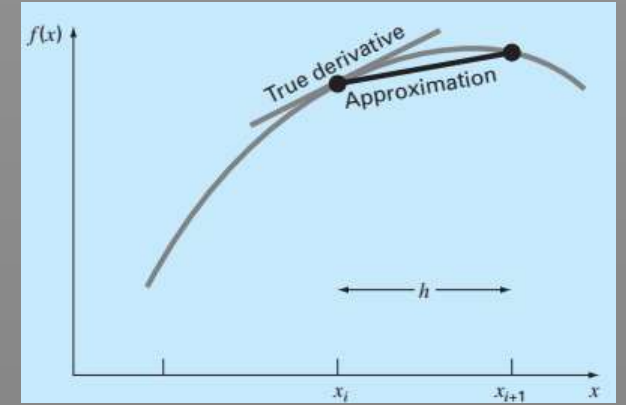
It utilizes data at i and i_{+1} to estimate the derivative.

$$\partial_x f^- \approx \frac{f(x) - f(x-dx)}{dx} \quad \text{backward difference}$$























To calculate a previous value on the basis of a present value

$$\partial_x f \approx \frac{f(x+dx) - f(x-dx)}{2dx} \quad \text{centered difference}$$

It utilizes data at i_{-1} and i_{+1} to estimate the derivative.



Finite Divided Difference Table

i	x_i	$f[x_i]$		First order differences		Second order differences		Third order differences		Fourth order differences		Fifth order differences
0	x_0	$f[x_0]$		$f[x_0, x_1]$		$f[x_0, x_1, x_2]$		$f[x_0, x_1, x_2, x_3]$		$f[x_0, x_1, x_2, x_3, x_4]$		$f[x_0, x_1, x_2, x_3, x_4, x_5]$
1	x_1	$f[x_1]$		$f[x_1, x_2]$		$f[x_1, x_2, x_3]$						
2	x_2	$f[x_2]$		$f[x_2, x_3]$				$f[x_1, x_2, x_3, x_4]$		$f[x_1, x_2, x_3, x_4, x_5]$		
3	x_3	$f[x_3]$		$f[x_3, x_4]$		$f[x_2, x_3, x_4]$						
4	x_4	$f[x_4]$				$f[x_3, x_4, x_5]$		$f[x_2, x_3, x_4, x_5]$				
5	x_5	$f[x_5]$		$f[x_4, x_5]$								



i	x_i	$f[x_i]$	1 st order differences	2 nd order differences	3 rd order differences	4 th order differences
0	0	0	$\frac{1-0}{1-0} = 1$	$\frac{7-1}{2-0} = 3$	$\frac{6-3}{3-0} = 1$	$\frac{1-1}{4-0} = 0$
1	1	1				
2	2	8	$\frac{8-1}{2-1} = 7$	$\frac{19-7}{3-1} = 6$	$\frac{9-6}{4-1} = 1$	
3	3	27	$\frac{27-8}{3-2} = 19$			
4	4	64	$\frac{64-27}{4-3} = 37$	$\frac{37-19}{4-2} = 9$		

Finite Divided Difference Table

Example: Compute $f(0.3)$ for the data using Newton's divided difference formula.

x	0	1	3	4	7
f	1	3	49	129	813

Divided difference table

x_i	f_i			
0	1			
		2		
1	3		7	
		23		3
3	49		19	
		80		3
4	129		37	
		228		
7	813			

$$f(x) = f[x_0] + (x - x_0) f[x_0, x_1] + (x - x_0)(x - x_1) f[x_0, x_1, x_2] + (x - x_0)(x - x_1)(x - x_2) f[x_0, x_1, x_2, x_3]$$

$$f(0.3) = 1 + (0.3 - 0) 2 + (0.3)(0.3 - 1) 7 + (0.3)(0.3 - 1)(0.3 - 3) 3 = 1.831$$

THANK YOU



Questions??