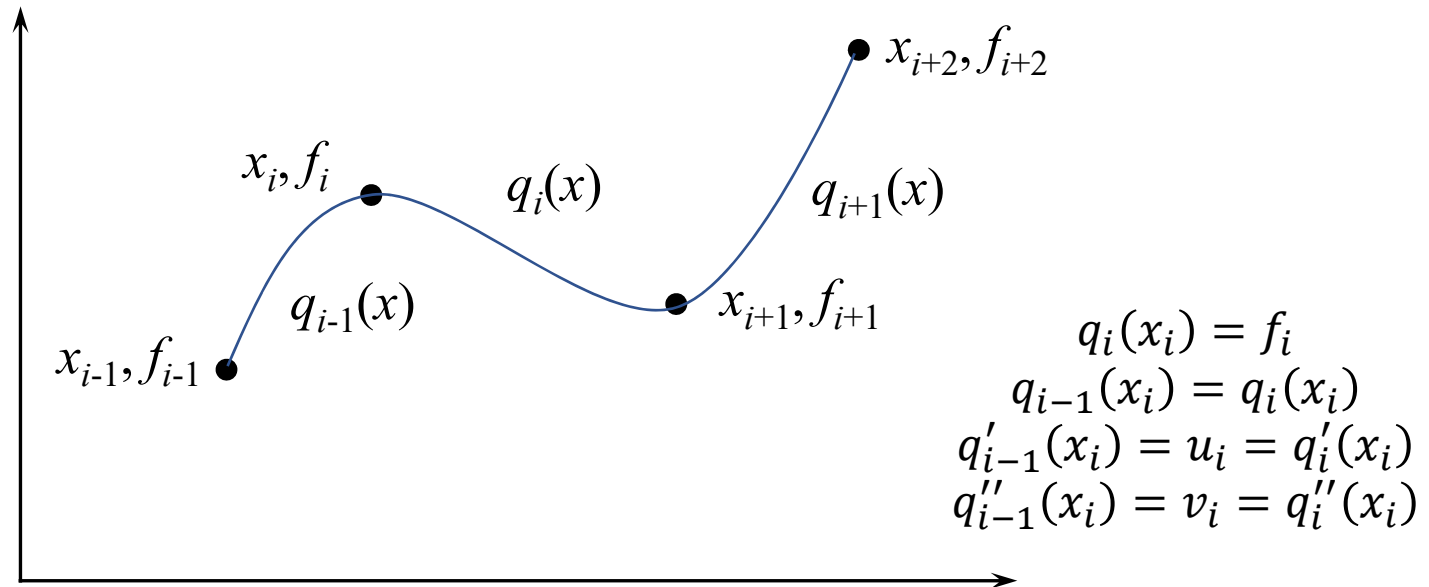


# Spline Interpolation

- ✓ **Given:**  $(n + 1)$  observations or data pairs  $[(x_0, f_0), (x_1, f_1), (x_2, f_2), \dots, (x_n, f_n)]$
- ✓ This gives a mesh of *nodes*  $\{x_0, x_1, x_2, \dots, x_n\}$  on the independent variable and the corresponding function values as  $\{f_0, f_1, f_2, \dots, f_n\}$
- ✓ **Goal:** fit an independent polynomial in each interval (between two points) with certain continuity requirements at the nodes.
  - ✓ *Linear spline*: continuity in function values,  $C^0$  continuity
  - ✓ *Quadratic spline*: continuity in function values and 1<sup>st</sup> derivatives,  $C^1$  continuity
  - ✓ *Cubic spline*: continuity in function values, 1<sup>st</sup> and 2<sup>nd</sup> derivatives,  $C^2$  continuity
- ✓ **Denote for node  $i$  or at  $x_i$ :** functional value  $f_i$ , first derivative  $u_i$ , second derivative  $v_i$

# Spline Interpolation: Cubic



- ✓ A cubic polynomial in each interval:  $(n+1)$  points,  $n$  cubic polynomials,  $4n$  unknowns
- ✓ Available conditions:  $(n + 1)$  function values,  $(n - 1)$  function continuity,  $(n - 1)$  1<sup>st</sup> derivative continuity conditions and  $(n - 1)$  2<sup>nd</sup> derivative continuity conditions, total  $4n - 2$  conditions.
- ✓ **2 free conditions to be chosen by the user!**

# Spline Interpolation: Cubic

- ✓ Cubic Spline in the interval  $\{x_i, x_{i+1}\}$ :  $q_i(x)$
- ✓  $q_i''(x)$  is a set of linear splines.
- ✓ Let us denote the 2<sup>nd</sup> derivative ( $v$ ) of the function at the  $i^{\text{th}}$  node as
$$f''(x_i) = v(x_i) = v_i$$
- ✓ Therefore:  $q_i''(x_i) = v(x_i) = v_i$  and  $q_i''(x_{i+1}) = v(x_{i+1}) = v_{i+1}$
- ✓ We may write:

$$q_i''(x) = \frac{v_{i+1}}{h_i}(x - x_i) - \frac{v_i}{h_i}(x - x_{i+1})$$
$$q_i(x) = \frac{v_{i+1}}{6h_i}(x - x_i)^3 - \frac{v_i}{6h_i}(x - x_{i+1})^3 + c_1x + c_2$$
$$q_i(x_i) = f_i = \frac{v_i h_i^2}{6} + c_1 x_i + c_2$$
$$q_i(x_{i+1}) = f_{i+1} = \frac{v_{i+1} h_i^2}{6} + c_1 x_{i+1} + c_2$$

# Spline Interpolation: Cubic

$$q_i(x) = \frac{v_{i+1}}{6h_i}(x - x_i)^3 - \frac{v_i}{6h_i}(x - x_{i+1})^3 + c_1x + c_2$$

$$q_i(x_i) = f_i = \frac{v_i h_i^2}{6} + c_1 x_i + c_2; \quad q_i(x_{i+1}) = f_{i+1} = \frac{v_{i+1} h_i^2}{6} + c_1 x_{i+1} + c_2$$

$$c_1 = f[x_{i+1}, x_i] - \frac{h_i}{6}(v_{i+1} - v_i) = \frac{f_{i+1}}{h_i} - \frac{f_i}{h_i} + \frac{h_i}{6}v_i - \frac{h_i}{6}v_{i+1}$$

$$c_2 = -\frac{f_{i+1}}{h_i}x_i + \frac{f_i}{h_i}x_{i+1} - \frac{h_i}{6}v_i x_{i+1} + \frac{h_i}{6}v_{i+1}x_i$$

$$\begin{aligned} q_i(x) &= \frac{v_{i+1}}{6} \left[ \frac{(x - x_i)^3}{h_i} - h_i(x - x_i) \right] + \frac{v_i}{6} \left[ -\frac{(x - x_{i+1})^3}{h_i} + h_i(x - x_{i+1}) \right] \\ &\quad + \frac{f_{i+1}}{h_i}(x - x_i) - \frac{f_i}{h_i}(x - x_{i+1}) \end{aligned}$$

We need to estimate  $(n + 1)$  unknown  $v_i$ . We have  $(n - 1)$  conditions from the continuity of the first derivative.

# Spline Interpolation: Cubic

$$q'_i(x_i) = q'_{i-1}(x_i); \quad i = 1, 2, 3, \dots, n-1 \quad \dots (eq. 1)$$

$$\begin{aligned} q_i(x) &= \frac{v_{i+1}}{6} \left[ \frac{(x - x_i)^3}{h_i} - h_i(x - x_i) \right] + \frac{v_i}{6} \left[ -\frac{(x - x_{i+1})^3}{h_i} + h_i(x - x_{i+1}) \right] + \frac{f_{i+1}}{h_i}(x - x_i) - \frac{f_i}{h_i}(x - x_{i+1}) \\ q'_i(x) &= \frac{v_{i+1}}{6} \left[ \frac{3(x - x_i)^2}{h_i} - h_i \right] + \frac{v_i}{6} \left[ -\frac{3(x - x_{i+1})^2}{h_i} + h_i \right] + f[x_{i+1}, x_i] \\ q'_{i-1}(x) &= \frac{v_i}{6} \left[ \frac{3(x - x_{i-1})^2}{h_{i-1}} - h_{i-1} \right] + \frac{v_{i-1}}{6} \left[ -\frac{3(x - x_i)^2}{h_{i-1}} + h_{i-1} \right] + f[x_i, x_{i-1}] \end{aligned}$$

In eq. (1),

$$\begin{aligned} -\frac{h_i}{6}v_{i+1} - \frac{h_i}{3}v_i + f[x_{i+1}, x_i] &= \frac{h_{i-1}}{3}v_i + \frac{h_{i-1}}{6}v_{i-1} + f[x_i, x_{i-1}] \\ -\frac{h_{i-1}}{6}v_{i-1} - \left(\frac{h_i + h_{i-1}}{3}\right)v_i - \frac{h_i}{6}v_{i+1} &= f[x_i, x_{i-1}] - f[x_{i+1}, x_i] \\ \frac{h_{i-1}}{6}v_{i-1} + \left(\frac{h_i + h_{i-1}}{3}\right)v_i + \frac{h_i}{6}v_{i+1} &= f[x_{i+1}, x_i] - f[x_i, x_{i-1}] \end{aligned}$$

$$\checkmark \quad \mathbf{h_{i-1}v_{i-1} + 2(h_i + h_{i-1})v_i + h_iv_{i+1} = 6f[x_{i+1}, x_i] - 6f[x_i, x_{i-1}]}$$

$$\sigma_{i-1} h_i + 2\sigma_i (h_i + h_{i+1}) + \sigma_{i+1} h_{i+1} = 6 (g_{i+1} - g_i) \rightarrow i=1,2,\dots,n$$

Natural spline

$$\left[ \begin{array}{c|c} \text{BCs} & \\ \hline h_1 & 2(h_1+h_2) & h_2 & - & - & - \\ & h_2 & 2(h_2+h_3) & h_3 & - & - \\ & & \ddots & \ddots & \ddots & \\ & & & h_m & 2(h_m+h_{m+1}) & h_{m+1} \\ \hline & \text{BCs} & & & & \end{array} \right] \begin{bmatrix} \sigma_0 \\ \sigma_1 \\ \vdots \\ \sigma_n \end{bmatrix} = 6 \begin{bmatrix} g_2 - g_1 \\ g_3 - g_2 \\ \vdots \\ g_n - g_{n-1} \end{bmatrix}$$

$n-1$  equations

# Spline Interpolation: Cubic

$$h_{i-1}v_{i-1} + 2(h_i + h_{i-1})v_i + h_iv_{i+1} = 6f[x_{i+1}, x_i] - 6f[x_i, x_{i-1}]$$

$$i = 1, 2, 3, \dots, n-1.$$

- ✓ So,  $(n-1)$  equations,  $(n+1)$  unknowns, two conditions have to be provided by the users. They decide the type of cubic splines

- ✓ *Natural Spline:*

$$v_0 = v_n = 0$$

- ✓ *Parabolic Runout:*

$$v_0 = v_1 \quad \text{and} \quad v_{n-1} = v_n$$

- ✓ *Not-a-knot:*

$$\begin{aligned} q_0(x) = q_1(x) &\Rightarrow \frac{v_1 - v_0}{h_0} = \frac{v_2 - v_1}{h_1} \\ q_{n-2}(x) = q_{n-1}(x) &\Rightarrow \frac{v_{n-1} - v_{n-2}}{h_{n-2}} = \frac{v_n - v_{n-1}}{h_{n-1}} \end{aligned}$$

- ✓ *Periodic:*

$$f_0 = f_n; \quad u_0 = u_n \quad \text{and} \quad v_0 = v_n$$

First one comes from the data (if not satisfied, the periodic spline is not appropriate); the next two give the other two equations.

# Spline Interpolation: Cubic

$$h_{i-1}v_{i-1} + 2(h_i + h_{i-1})v_i + h_iv_{i+1} = 6f[x_{i+1}, x_i] - 6f[x_i, x_{i-1}]$$

✓  $i = 1, 2, 3, \dots, n-1$ . So,  $(n-1)$  equations,  $(n+1)$  unknowns, two conditions have to be provided by the users. They decide the type of cubic splines

✓ *Clamped Spline:*

$$u_0 = q'_0(x_0) = \alpha \quad \text{and} \quad u_n = q'_{n-1}(x_n) = \beta$$

$$q'_0(x) = \frac{v_1}{6} \left[ \frac{3(x-x_0)^2}{h_0} - h_0 \right] + \frac{v_0}{6} \left[ -\frac{3(x-x_1)^2}{h_0} + h_0 \right] + f[x_1, x_0]$$

$$q'_0(x_0) = -\frac{v_1 h_0}{6} - \frac{v_0 h_0}{3} + f[x_1, x_0] = \alpha \quad \Rightarrow \quad 2v_0 + v_1 = \frac{6}{h_0} (f[x_1, x_0] - \alpha)$$

$$q'_{n-1}(x) = \frac{v_n}{6} \left[ \frac{3(x-x_{n-1})^2}{h_{n-1}} - h_{n-1} \right] + \frac{v_{n-1}}{6} \left[ -\frac{3(x-x_n)^2}{h_{n-1}} + h_{n-1} \right] + f[x_n, x_{n-1}]$$

$$q'_{n-1}(x_n) = \frac{v_n h_{n-1}}{3} + \frac{v_{n-1} h_{n-1}}{6} + f[x_n, x_{n-1}] = \beta$$

$$\Rightarrow \quad 2v_n + v_{n-1} = \frac{6}{h_{n-1}} (\beta - f[x_n, x_{n-1}])$$



## Example Problem: Q4 of Tutorial 9

Consider the function  $\exp(x)$  sampled at points

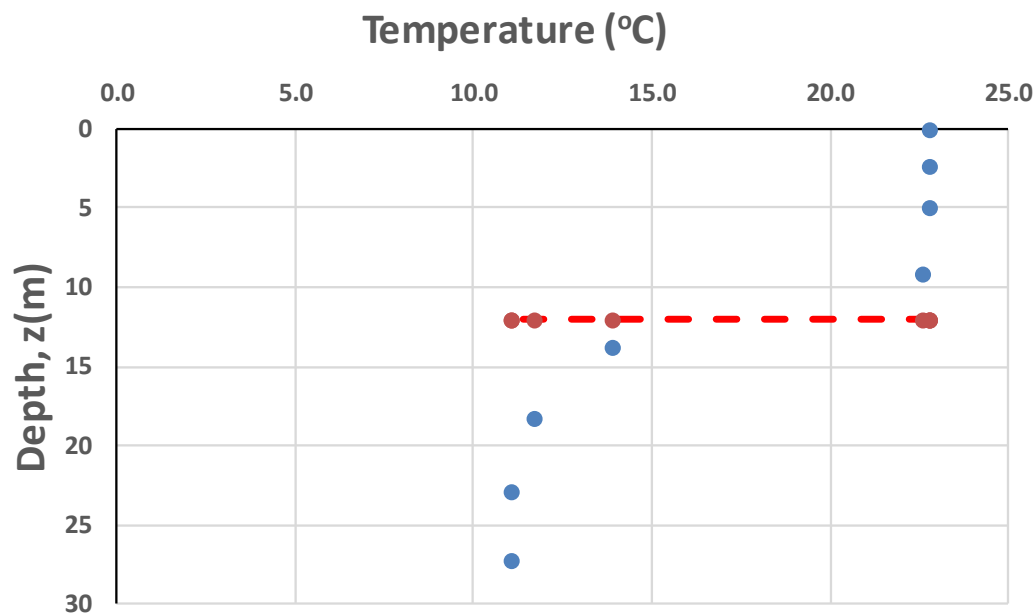
$$x = 0, 0.5, 1.0, 1.5 \text{ and } 2$$

Estimate the function value at  $x = 1.80$  by interpolating the function using - (a) natural cubic spline and (b) not-a-knot cubic spline. Calculate the true percentage error for both the splines. Which is the better spline for this problem and why?

Solution: Posted online along with Tutorial 9 solutions

# Example Problem: Heat Transfer in Lake

Lakes in temperate zone can become thermally stratified during the summer. As depicted below, warm, buoyant water near the surface overlies colder, denser bottom water. Such stratification effectively divides the lake into two layers: the *epilimnion* and the *hypolimnion* separated by a plane called the *thermocline*.



$z$ (m)	$T(^{\circ}\text{C})$
0	22.8
2.3	22.8
4.9	22.8
9.1	22.6
13.7	13.9
18.3	11.7
22.9	11.1
27.2	11.1

## Example Problem: Heat Transfer in Lake

The location of the thermocline can be defined as the inflection point of the T-z curve; i.e. where  $\frac{d^2T}{dz^2} = 0$ . It is also the point at which the absolute value of the first derivative or gradient is maximum. Use cubic splines to determine the thermocline depth of this lake. Also use splines to determine the value of the gradient at the thermocline.

# Spline Interpolation: Using Local Coordinate

$$\begin{array}{ccc} x \in [x_i, x_{i+1}] & \rightarrow & t \in [0, 1] \\ q_i(x) \text{ in } [x_i, x_{i+1}] & \rightarrow & p_i(t) \text{ in } [0, 1] \end{array}$$

✓ At each *node*  $i$ , we denote the following:

- ✓ Location:  $x_i$
- ✓ Functional value:  $f_i$
- ✓ Intervals:  $h_i = x_{i+1} - x_i$  and  $h_{i-1} = x_i - x_{i-1}$
- ✓ Derivatives: First derivative  $u_i$  and the 2<sup>nd</sup> derivative  $v_i$

✓ Transformations:

$$\begin{aligned} t = \frac{x - x_i}{x_{i+1} - x_i} = \frac{1}{h_i} (x - x_i) & \Rightarrow \frac{dt}{dx} = \frac{1}{h_i} \\ q'_i(x) = p'_i(t) \frac{dt}{dx} = \frac{1}{h_i} p'_i(t) & \quad q''_i(x) = \frac{1}{h_i^2} p''_i(t) \end{aligned}$$

# Spline Interpolation: Using Local Coordinate

✓  $C^0$  – Continuity:

$$p_{i-1}(1) = q_{i-1}(x_i) = q_i(x_i) = p_i(0) = f_i$$

✓  $C^1$  – Continuity:

$$\frac{1}{h_{i-1}} p'_{i-1}(1) = q'_{i-1}(x_i) = q'_i(x_i) = \frac{1}{h_i} p'_i(0) = u_i$$

✓  $C^2$  – Continuity:

$$\frac{1}{h_{i-1}^2} p''_{i-1}(1) = q''_{i-1}(x_i) = q''_i(x_i) = \frac{1}{h_i^2} p''_i(0) = v_i$$

# Linear and Quadratic Splines: Local Coordinate

## ✓ Linear Spline: $C^0$ – Continuous

$$\begin{aligned} p_i(t) = a_i t + b_i &\Rightarrow p_i(0) = b_i = f_i, & p_i(1) = a_i + b_i = f_{i+1} \\ p_i(t) = (f_{i+1} - f_i)t + f_i &\Rightarrow q_i(x) = f[x_{i+1}, x_i](x - x_i) + f_i \end{aligned}$$

## ✓ Quadratic Spline: $C^1$ – Continuous

$$p_i(t) = a_i t^2 + b_i t + c_i \Rightarrow p_i(0) = c_i = f_i, \quad p_i(1) = a_i + b_i + c_i = f_{i+1}$$

Using the definition of  $u_i$ :

$$\begin{aligned} \frac{1}{h_i} p'_i(0) = \frac{b_i}{h_i} = u_i &\Rightarrow a_i = h_i(f[x_{i+1}, x_i] - u_i) \\ p_i(t) &= h_i(f[x_{i+1}, x_i] - u_i)t^2 + h_i u_i t + f_i \end{aligned}$$

Using  $C^1$  – Continuity:

$$\frac{1}{h_{i-1}} p'_{i-1}(1) = \frac{1}{h_i} p'_i(0) \Rightarrow u_i = 2f[x_i, x_{i-1}] - u_{i-1}$$

# Cubic Spline: Using Local Coordinate

✓ Cubic Spline:  $C^2$  – Continuous

$$p_i(t) = a_i t^3 + b_i t^2 + c_i t + d_i$$

Using  $C^0$  – Continuity:

$$p_i(0) = d_i = f_i, \quad p_i(1) = a_i + b_i + c_i + d_i = f_{i+1}$$

Now we have two options:

- ✓ Option 1: Using the 1<sup>st</sup> derivative  $u_i$  as unknown and  $C^2$  – Continuity to estimate them
- ✓ Option 2: Using the 2<sup>nd</sup> derivative  $v_i$  as unknown and  $C^1$  – Continuity to estimate them

# Cubic Spline: Using Local Coordinates

Option 1: Using the 1<sup>st</sup> derivative  $u_i$  as unknown and  $C^2$  – Continuity to estimate them

$$\begin{aligned}p_i(t) &= a_i t^3 + b_i t^2 + c_i t + d_i \\d_i &= f_i, \quad a_i + b_i + c_i + d_i = f_{i+1} \\ \frac{1}{h_i} p'_i(0) &= \frac{c_i}{h_i} = u_i; \quad \frac{1}{h_i} p'_i(1) = \frac{3a_i + 2b_i + c_i}{h_i} = u_{i+1} \\ a_i &= h_i(u_{i+1} + u_i - 2f[x_{i+1}, x_i]) \\ b_i &= h_i(3f[x_{i+1}, x_i] - u_{i+1} - 2u_i)\end{aligned}$$

Using  $C^2$  – Continuity:

$$\begin{aligned}\frac{1}{h_{i-1}^2} p''_{i-1}(1) &= \frac{1}{h_i^2} p''_i(0) \Rightarrow \frac{6a_{i-1} + 2b_{i-1}}{h_{i-1}^2} = \frac{2b_i}{h_i^2} \\ h_i u_{i-1} + 2(h_{i-1} + h_i) u_i + h_{i-1} u_{i+1} &= 3h_{i-1} f[x_{i+1}, x_i] + 3h_i f[x_i, x_{i-1}] \\ i &= 1, 2, 3, \dots, n-1\end{aligned}$$

Using the two other conditions, one may obtain similar splines of different types!



# Cubic Spline: Using Local Coordinates

✓ *Natural Spline:*

$$v_0 = v_n = 0$$

$$v_0 = \frac{p_0''(0)}{h_0^2} = \frac{2b_0}{h_0^2} = 0; \quad b_0 = h_0(3f[x_1, x_0] - u_1 - 2u_0) = 0$$

$$v_n = \frac{p_{n-1}''(1)}{h_{n-1}^2} = \frac{6a_{n-1} + 2b_{n-1}}{h_{n-1}^2} = 0$$

$$6h_{n-1}(u_n + u_{n-1} - 2f[x_n, x_{n-1}]) + 2h_{n-1}(3f[x_n, x_{n-1}] - u_n - 2u_{n-1}) = 0$$

$$2u_0 + u_1 = 3f[x_1, x_0] \qquad 2u_n + u_{n-1} = 3f[x_n, x_{n-1}]$$

✓ *Clamped Spline:*

$$u_0 = \alpha \quad \text{and} \quad u_n = \beta$$

✓ *Parabolic Runout:*

$$\frac{p_0''(0)}{h_0^2} = \frac{p_0''(1)}{h_0^2} \Rightarrow \frac{2b_0}{h_0^2} = \frac{6a_0 + 2b_0}{h_0^2} \Rightarrow u_0 + u_1 = 2f[x_1, x_0]$$

$$\frac{p_{n-1}''(0)}{h_{n-1}^2} = \frac{p_{n-1}''(1)}{h_{n-1}^2} \Rightarrow \frac{2b_{n-1}}{h_{n-1}^2} = \frac{6a_{n-1} + 2b_{n-1}}{h_{n-1}^2} \Rightarrow u_{n-1} + u_n = 2f[x_n, x_{n-1}]$$

# Cubic Spline: Using Local Coordinates

✓ *Not-a-knot:*

$$\begin{aligned} q_0(x) = q_1(x) &\Rightarrow \frac{v_1 - v_0}{h_0} = \frac{v_2 - v_1}{h_1} \\ q_{n-2}(x) = q_{n-1}(x) &\Rightarrow \frac{v_{n-1} - v_{n-2}}{h_{n-2}} = \frac{v_n - v_{n-1}}{h_{n-1}} \end{aligned}$$

✓ *Periodic:*

$$f_0 = f_n; \quad u_0 = u_n \quad \text{and} \quad v_0 = v_n$$

First one comes from the data (if not satisfied, the periodic spline is not appropriate); the next two give the other two equations.

Formulation of these two is left as homework!

# Cubic Spline: Using Local Coordinates

Option 2: Using the 2<sup>nd</sup> derivative  $v_i$  as unknown and  $C^1$  – Continuity to estimate them

$$\begin{aligned}
 p_i(t) &= a_i t^3 + b_i t^2 + c_i t + d_i \\
 d_i &= f_i, & a_i + b_i + c_i + d_i &= f_{i+1} \\
 \frac{1}{h_i^2} p_i''(0) &= \frac{2b_i}{h_i^2} = v_i; & \frac{1}{h_i^2} p_i''(1) &= \frac{6a_i + 2b_i}{h_i^2} = v_{i+1} \\
 a_i &= \frac{h_i^2}{6} (v_{i+1} - v_i); & c_i &= h_i f[x_{i+1}, x_i] - \frac{h_i^2}{6} (v_{i+1} + 2v_i)
 \end{aligned}$$

Using  $C^1$  – Continuity:

$$\begin{aligned}
 \frac{1}{h_{i-1}} p'_{i-1}(1) &= \frac{1}{h_i} p'_i(0) \Rightarrow \frac{3a_{i-1} + 2b_{i-1} + c_{i-1}}{h_{i-1}} = \frac{c_i}{h_i} \\
 h_{i-1}v_{i-1} + 2(h_{i-1} + h_i)v_i + h_i v_{i+1} &= 6f[x_{i+1}, x_i] - 6f[x_i, x_{i-1}] \\
 i &= 1, 2, 3, \dots, n-1
 \end{aligned}$$

This is the same equation that was obtained using Lagrange polynomials!  
Boundary conditions are also the same!

# ESO 208A: Computational Methods in Engineering

## Numerical Differentiation

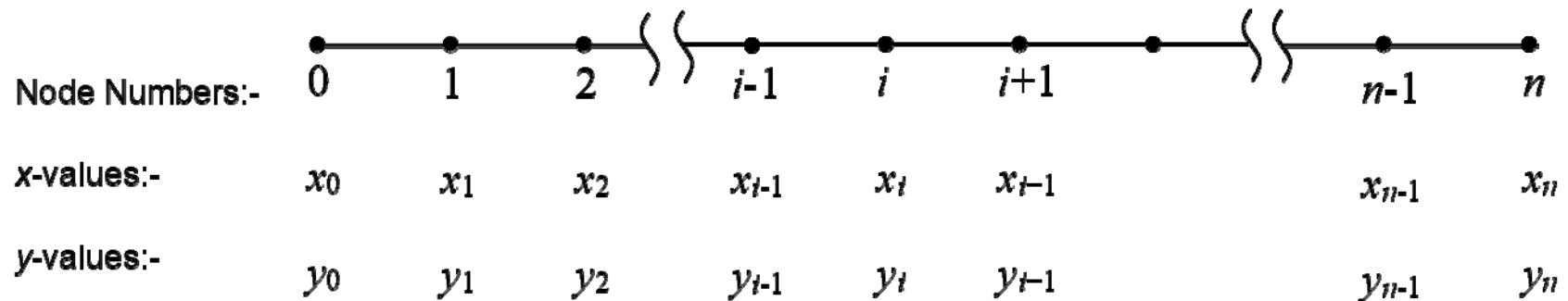
**Abhas Singh**

Department of Civil Engineering  
IIT Kanpur



**Acknowledgements: Profs. Saumyen Guha and Shivam Tripathi (CE)**

# Numerical Differentiation



Let us compute  $dy/dx$  or  $df/dx$  at node  $i$

Denote the difference operators:

$$\Delta x = x_{i+1} - x_i \quad \nabla x = x_i - x_{i-1} \quad \delta x = x_{i+1/2} - x_{i-1/2}$$

# Numerical Differentiation: Finite Difference

Approximate the function between  $\{x_i, x_{i+1}\}$  as:

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

Forward Difference:

$$\frac{df}{dx} = \frac{f_{i+1} - f_i}{\Delta x} = \frac{\Delta f}{\Delta x}$$

Approximate the function between  $\{x_{i-1}, x_i\}$  as:

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

Backward Difference:

$$\frac{df}{dx} = \frac{f_i - f_{i-1}}{\nabla x} = \frac{\nabla f}{\nabla x}$$

# Numerical Differentiation: Finite Difference

Approximate the function between three points:  $\{x_{i-1}, x_i, x_{i+1}\}$

$$\begin{aligned} f(x) &= \frac{(x - x_i)(x - x_{i+1})}{(x_{i-1} - x_i)(x_{i-1} - x_{i+1})} f_{i-1} + \frac{(x - x_{i-1})(x - x_{i+1})}{(x_i - x_{i-1})(x_i - x_{i+1})} f_i \\ &\quad + \frac{(x - x_{i-1})(x - x_i)}{(x_{i+1} - x_{i-1})(x_{i+1} - x_i)} f_{i+1} \\ &= \frac{f_{i-1}}{\nabla x (\Delta x + \nabla x)} (x - x_i)(x - x_{i+1}) - \frac{f_i}{\nabla x \Delta x} (x - x_{i-1})(x - x_{i+1}) \\ &\quad + \frac{f_{i+1}}{(\Delta x + \nabla x) \Delta x} (x - x_{i-1})(x - x_i) \end{aligned}$$

Now, evaluate  $df/dx$  at  $x = x_i$ :

$$\left. \frac{df}{dx} \right|_{x_i} = \frac{f_{i-1}}{\nabla x (\Delta x + \nabla x)} (-\Delta x) - \frac{f_i}{\nabla x \Delta x} (\nabla x - \Delta x) + \frac{f_{i+1}}{(\Delta x + \nabla x) \Delta x} (\nabla x)$$

# Numerical Differentiation: Finite Difference

## Central Difference:

$$\left. \frac{df}{dx} \right|_{x_i} = \frac{f_{i-1}}{\nabla x (\Delta x + \nabla x)} (-\Delta x) - \frac{f_i}{\nabla x \Delta x} (\nabla x - \Delta x) + \frac{f_{i+1}}{(\Delta x + \nabla x) \Delta x} (\nabla x)$$

For regular or uniform grid:  $\Delta x = \nabla x = h$

$$\left. \frac{df}{dx} \right|_{x_i} = \frac{f_{i+1} - f_{i-1}}{2h} = \frac{\delta f}{\delta x}$$

Let us assume regular grid with a mesh size of  $h$



# Numerical Differentiation: Finite Difference

Approximate the function between three points:  $\{x_{i-1}, x_i, x_{i+1}\}$

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

$$f(x) = \frac{f_{i-1}}{2h^2} (x - x_i)(x - x_{i+1}) - \frac{f_i}{h^2} (x - x_{i-1})(x - x_{i+1}) + \frac{f_{i+1}}{2h^2} (x - x_{i-1})(x - x_i)$$

Now, evaluate central difference approximations of  $df/dx$  and  $d^2f/dx^2$  at  $x = x_i$ :

$$\frac{df}{dx} = \frac{f_{i-1}}{2h^2} [(x - x_i) + (x - x_{i+1})] - \frac{f_i}{h^2} [(x - x_{i-1}) + (x - x_{i+1})] + \frac{f_{i+1}}{2h^2} [(x - x_{i-1}) + (x - x_i)]$$

$$\left. \frac{df}{dx} \right|_{x_i} = \frac{f_{i+1} - f_{i-1}}{2h}$$

$$\left. \frac{d^2f}{dx^2} \right|_{x_i} = \frac{f_{i-1} - 2f_i + f_{i+1}}{h^2}$$

## Numerical Differentiation: Finite Difference

- ✓ Similarly, one can approximate the function between three points  $\{x_i, x_{i+1}, x_{i+2}\}$  and obtain the *forward difference* expressions of the first and second derivatives at  $x = x_i$  as follows:

$$\left. \frac{df}{dx} \right|_{x_i} = \frac{-3f_i + 4f_{i+1} - f_{i+2}}{2h}$$

$$\left. \frac{d^2f}{dx^2} \right|_{x_i} = \frac{f_i - 2f_{i+1} + f_{i+2}}{h^2}$$

This is left for homework practice!

## Numerical Differentiation: Finite Difference

- ✓ Similarly, one can approximate the function between three points  $\{x_{i-2}, x_{i-1}, x_i\}$  and obtain the *backward difference* expressions of the first and second derivatives at  $x = x_i$  as follows:

$$\left. \frac{df}{dx} \right|_{x_i} = \frac{3f_i - 4f_{i-1} + f_{i-2}}{2h}$$

$$\left. \frac{d^2f}{dx^2} \right|_{x_i} = \frac{f_{i-2} - 2f_{i-1} + f_i}{h^2}$$

This is left for homework practice!

# Numerical Differentiation: Finite Difference

- ✓ *Accuracy*: How accurate is the numerical differentiation scheme with respect to the TRUE differentiation?
  - ✓ **Truncation Error** analysis
  - ✓ **Modified Wave Number, Amplitude Error and Phase Error** analysis for periodic functions
- ✓ *Recall*: True Value ( $a$ ) = Approximate Value ( $\tilde{a}$ ) + Error ( $\varepsilon$ )
- ✓ *Consistency*: A numerical expression for differentiation or a numerical differentiation scheme is consistent if it converges to the TRUE differentiation as  $h \rightarrow 0$ .