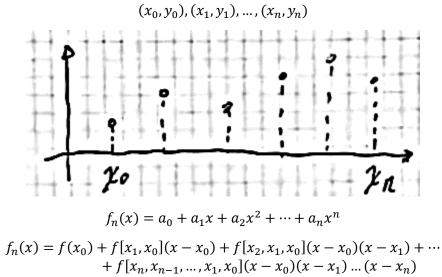
What we've looked at so far:



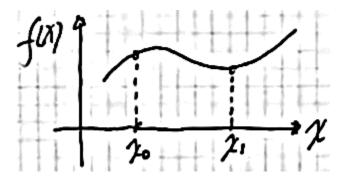
$$f[x_{n}, x_{n-1}, ..., x_{1}, x_{0}](x - x_{0})(x - x_{1})$$

$$f(x_{0}) = y_{0}$$

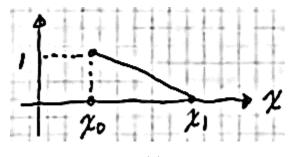
$$f[x_{1}, x_{0}] = \frac{f(x_{1}) - f(x_{0})}{x_{1} - x_{0}} = \frac{y_{1} - y_{0}}{x_{1} - x_{0}}$$

$$f[x_{2}, x_{1}, x_{0}] = \frac{f[x_{2}, x_{1}] - f[x_{1}, x_{0}]}{x_{2} - x_{0}}$$

Lagrange Interpolating Polynomial



Given $(x_0, f(x_0)), (x_1, f(x_1)), \underline{\text{fitting line}}$.



$$L_0(x):$$

$$L_0(x_0) = 1$$

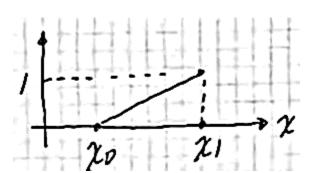
$$L_0(x_1)=0$$

$$L_0(x) = b_0(x - x_1)$$

$$L_0(x) = b_0(x_0 - x_1) = 1$$

$$b_0 = \frac{1}{x_0 - x_1}$$

$$L_0(x) = \frac{x - x_1}{x_0 - x_1}$$

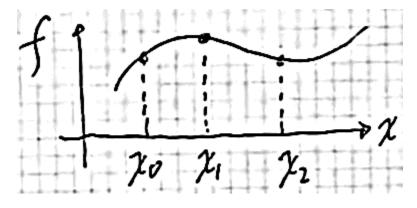


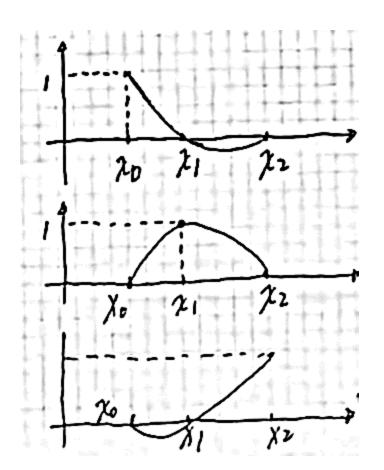
$$l_1(x)$$
:
 $l_1(x_0) = 0$
 $l_1(x_1) = 1$

$$L_1(x) = \frac{x - x_0}{x_1 - x_0}$$

$$f_1(x) = f(x_0) \cdot L_0(x) + f(x_1) \cdot L_1(x)$$

Given $(x_0, f(x_0))$, $(x_1, f(x_1))$, $(x_2, f(x_2))$ fitting polynomial of degree 2:





$$L_0(x) = \begin{cases} 1 & ; & x = x_0 \\ 0 & ; & x = x_1 \\ 0 & ; & x = x_2 \end{cases}$$

$$L_1(x) = \begin{cases} 0 & ; & x = x_0 \\ 1 & ; & x = x_1 \\ 0 & ; & x = x_2 \end{cases}$$

$$L_2(x) = \begin{cases} 0 & ; & x = x_0 \\ 0 & ; & x = x_1 \\ 1 & ; & x = x_2 \end{cases}$$

$$f_2(x) = f(x_0) \cdot L_0(x) + f(x_1) \cdot L_1(x) + f(x_2) \cdot L_2(x)$$

$$L_0(x) = b_0(x - x_1)(x - x_2)$$

$$L_0(x) = b_0(x_0 - x_1)(x_0 - x_2) = 1$$

$$b_0 = \frac{1}{(x_0 - x_1)(x_0 - x_2)}$$

$$L_0(x) = \frac{(x - x_1)(x - x_2)}{(x_0 - x_1)(x_0 - x_2)}$$

$$L_1(x) = \frac{(x - x_0)(x - x_2)}{(x_1 - x_0)(x_1 - x_2)}$$

$$L_2(x) = \frac{(x - x_0)(x - x_1)}{(x_2 - x_0)(x_2 - x_1)}$$

Example

Given the following data, use the Lagrange interpolating polynomial to fit the data.

$$x_0 = 1$$
; $f(x_0) = 0$
 $x_1 = 4$; $f(x_1) = 1.386294$
 $x_2 = 6$; $f(x_2) = 1.791760$

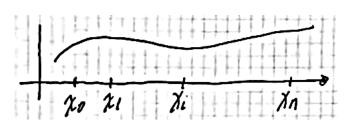
Solution

$$L_0(x) = \frac{(x-4)(x-6)}{(1-4)(1-6)} = \left(\frac{1}{15}\right)(x^2 - 10x + 24)$$

$$L_1(x) = \frac{(x-1)(x-6)}{(4-1)(4-6)} = -\left(\frac{1}{6}\right)(x^2 - 7x + 6)$$

$$L_2(x) = \frac{(x-1)(x-4)}{(6-1)(6-4)} = \left(\frac{1}{10}\right)(x^2 - 5x + 4)$$

$$\therefore f_2(x) = f(x_0) \cdot L_0(x) + f_1(x) \cdot L_1(x) + f_2(x) \cdot L_2(x)$$



$$L_0(x) = \begin{cases} 1 & ; & x = x_0 \\ 0 & ; & x = x_1, \dots, x_n \end{cases}$$

$$L_1(x) = \begin{cases} 1 & ; & x = x_1 \\ 0 & ; & other \ x \end{cases}$$

$$L_i(x) = \begin{cases} 1 & ; & x = x_i \\ 0 & ; & other \ x \end{cases}$$
 (Where $i = 0, 1, \dots n$)

$$L_{i}(x) = \frac{(x - x_{0}) \dots (x - x_{i-1})(x - x_{i+1}) \dots (x - x_{n})}{(x_{i} - x_{0}) \dots (x_{i} - x_{i-1})(x_{i} - x_{i+1}) \dots (x_{i} - x_{n})}$$

$$= \prod_{j=0}^{n} \frac{x - x_{j}}{x_{i} - x_{j}}$$

(Where
$$j = 0, 1, ...n$$
; but $j \neq i$)

Estimate error:

$$R_n = f[x_1, x_n, x_{n-1}, ..., x_0] \prod_{i=0}^{n} (x - x_i)$$

Inverse Interpolation

x	1	2	3	4	5	6	7
$f(x) = \frac{1}{x}$	1	0.5	0.3333	0.25	0.2	0.16667	0.1428

Find x such that f(x) = 0.3

- 1. Interchange $x \leftrightarrow f(x)$, construct the interpolation polynomial
- 2. Using a few point construct a polynomial then solving the equation to find x.

Using (2, 0.5), (3, 0.3333), (4, 0.25) to construct a polynomial:

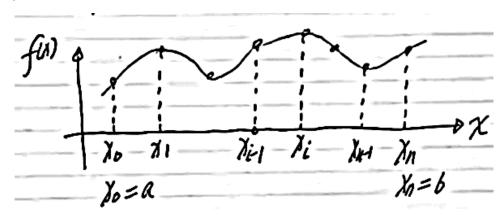
$$f_2(x) = 1.08333 - 0.375x + 0.041667x^2$$

 $0.3 = f_2(x) = 1.08333 - 0.375x + 0.041667x^2$
 $\rightarrow x = 3.295842, 5.704158$

The exact value of x is:

$$f(x) = \frac{1}{x} = 0.3 \rightarrow x = 3.333$$

Spline Interpolation



Given a set of n+1 data points (x_i, y_i) where no two x_i are the same and $a=x_0 < x_1 < \cdots x_n = b$, the spline S(x) is a piecewise function satisfying:

- 1. $S(x) \in C^2[a,b](S(x),S'(x),S''(x))$ exist and continuous
- 2. On each interval $[x_{i-1}, x_i]$, S(x) is a cubic polynomial i = 1, 2, ..., n
- 3. $S(x_i) = f(x_i) = y_i$, i = 0, 1, ..., n

Assume that

$$S(x) = \begin{cases} C_1(x) & ; & x_0 < x < x_2 \\ & \vdots \\ C_2(x) & ; & x_1 < x < x_2 \\ & \vdots \\ C_n(x) & ; & x_{n+1} < x < x_n \end{cases}$$

And

$$C_i(x) = a_{0i} + a_{1i}x + a_{2i}x^2 + a_{3i}x^3$$

 $i = 1, 2, ..., n$
 $a_{3i} \neq 0$

There are a 4n unknowns

The equations:

$$\begin{split} C_i(x)|_{x=x_{i-1}} &= C_i(x_{i-1}) = f(x-i) (=y_{i-1}) \\ C_i(x_{i-1}) &= y_{i-1} & (i=1,2,3,\ldots n-1) \\ C_i(x_i) &= y_i & (i=1,2,3,\ldots n-1) \\ (x_i) &= C'_{i+1}(x_i) & (i=1,2,3,\ldots,n-1) \\ C''_i(x_i) &= C''_{i+1}(x_i) & (i=1,2,3,\ldots,n-1) \end{split}$$

Total of 4n-2 equations – boundary conditions are needed.

Case 1: The first derivatives at the endpoints are given

Consider clamped boundary conditions

$$C'_1(x_0) = f'_0$$

$$C'_n(x_n) = f'_n$$

Case 2: The second derivatives at the endpoints are given.

$$C_1''(x_0) = f_0''$$

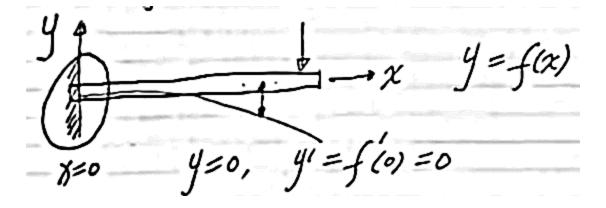
 $C_n''(x_n) = f_n''$

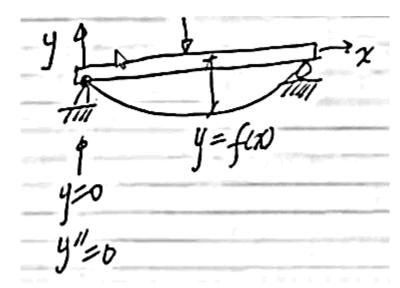
Special case $f_0^{\prime\prime}=f_n^{\ \prime\prime}$ is called natural or simple B.C.'s

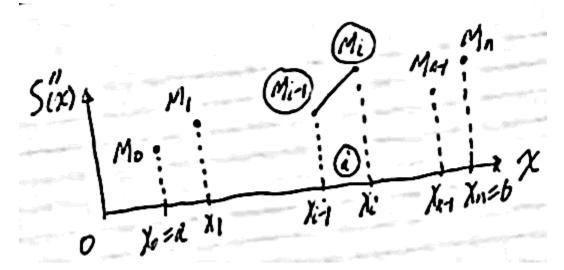
Case 3: Periodic conditions

$$C_1(x_0) = C_n(x_n)$$

 $C'_1(x_0) = C'_n(x_n)$
 $C''_1(x_0) = C''_n(x_n)$







Use the second derivatives

$$S''(x_i) = M_i \quad i = 0, 1, 2, ..., n$$

To find S(x) In the interval $x_{i-1} < x < x_i$:

$$C_i''(x) = M_{i-1} \frac{x_i - x}{x_i - x_{i-1}} + M_i \frac{x - x_{i-1}}{x_i - x_{i-1}}$$
 $i = 1, 2, ..., n$

Integrate the moment function twice:

$$C'_i(x) = -M_{i-1} \frac{(x_i - x)^2}{2h_i} + M_i \frac{(x - x_{i-1})^2}{2h_i} + \alpha$$

Here $h_i = x_i - x_{i-1}$

$$C_i(x) = M_{i-1} \frac{(x_i - x)^3}{6h_i} + M_i \frac{(x - x_{i-1})^3}{6h_i} + \alpha(x - x_{i-1}) + \beta$$

At $x = x_{i-1}$:

$$C_{i}(x) = y_{i-1} = f(x_{i-1})$$

$$\therefore C_{i}(x_{i-1}) = M_{i-1} \frac{(x_{i} - x_{i-1})^{3}}{6h_{i}} + 0 + 0 + \beta = f(x_{i-1})$$

$$\beta = f(x_{i-1}) - M_{i-1} \frac{h_{i}^{2}}{6}$$

At $x = x_i$:

$$C_{i}(x) = y_{i} = f(x_{i})$$

$$\therefore C_{i}(x_{i}) = M_{i} \frac{(x_{i} - x_{i+1})^{3}}{6h_{i}} + \alpha(x_{i} - x_{i-1}) + \beta = f(x_{i})$$

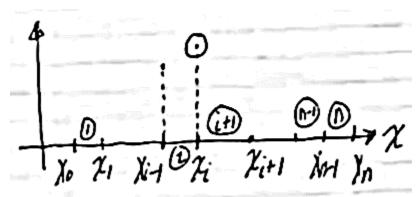
$$\alpha = (M_{i-1} - M_{i}) \frac{h_{i}}{6} + \frac{f(x_{i}) - f(x_{i-1})}{h_{i}}$$

The cubic function:

$$\begin{split} C_i(x) &= M_{i-1} \frac{(x_i - x)^3}{6h_i} + M_i \frac{(x - x_{i-1})^3}{6h_i} + \left[(M_{i-1} - M_i) \frac{h_i}{6} + \frac{f(x_i) - f(x_{i-1})}{h_i} \right] (x - x_{i-1}) + f(x_{i-1}) - M_{i-1} \frac{h_i^2}{6} \\ C_i(x) &= M_{i-1} \frac{(x_i - x)^3}{6h_i} + M_i \frac{(x - x_{i-1})^3}{6h_i} + \left(f(x_{i-1}) - M_{i-1} \frac{h_i^2}{6} \right) \frac{x_i - x}{h_i} + \left(f(x_i) - M_i \frac{h_i^2}{6} \right) \frac{x - x_{i-1}}{h_i} \end{split}$$

The first derivative of $C_i(x)$:

$$C'_{i}(x) = -M_{i-1}\frac{(x_{i}-x)^{2}}{2h_{i}} + M_{i}\frac{(x-x_{i-1})^{2}}{2h_{i}} - \left(f(x_{i-1}) - M_{i-1}\frac{h_{i}^{2}}{6}\right)\frac{1}{h_{i}} + \left(f(x_{i}) - M_{i}\frac{h_{i}^{2}}{6}\right)\frac{1}{h_{i}}$$



At $x = x_i$, we have:

$$C'_{i}(x_{i}) = 0 + M_{i} \frac{(x_{i} - x_{i-1})^{2}}{2h_{i}} + \frac{f(x_{i}) - f(x_{i-1})}{h_{i}} + M_{i-1} \frac{h_{i}}{6} - M_{i} \frac{h_{i}}{6}$$
$$= (M_{i-1} + 2M_{i}) \frac{h_{i}}{6} + f[x_{i}, x_{i-1}]$$

For interval $i+1, x_i \le x \le x_{i+1}$ $(i=1,2,\ldots,n-1)$

$$C'_{i+1}(x) = -M_i \frac{(x_{i+1} - x)^2}{2h_{i+1}} + M_i \frac{(x - x_i)^2}{2h_{i+1}} - \left(f(x_i) - M_{i-1} \frac{h_{i+1}^2}{6}\right) \frac{1}{h_{i+1}} + \left(f(x_{i+1}) - M_{i+1} \frac{h_{i+1}^2}{6}\right) \frac{1}{h_{i+1}}$$

At $x = x_i$:

$$C'_{i+1}(x_i) = -M_i \frac{(x_{i+1} - x_i)^2}{2h_{i+1}} + 0 + \frac{f(x_{i-1}) - f(x_i)}{h_{i+1}} + M_i \frac{h_{i+1}}{6} - M_{i+1} \frac{h_{i+1}}{6}$$
$$C'_{i+1}(x_i) = -(2M_i + M_{i+1}) \frac{h_{i+1}}{6} + f(x_{i+1}, x_i)$$

Since $C'_{i}(x_{i}) = C_{i+1}'(x_{i})$ (i = 1, 2, ..., n-1)

$$(M_{i-1} + 2M_i) \frac{h_i}{6} + f[x_i, x_{i-1}]$$

$$= -(2M_i + M_{i+1}) \frac{h_{i+1}}{6} + f[x_{i+1}, x_i]$$

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

$$M_{i-1} = \frac{h_i}{h_i + h_{i+1}} + 2M_i + M_{i+1} \frac{h_i}{h_i + h_{i+1}} = 6 \frac{f[x_{i+1}, x_i] - f[x_i, x_{i-1}]}{h_i + h_{i+1}}$$

Define:

$$\alpha_i = \frac{h_i}{h_i + h_{i+1}} \qquad \qquad \left(\ \ i = 1, 2, \dots, n-1 \right)$$

$$\beta_i = \frac{h_{i+1}}{h_i + h_{i+1}}$$
 And $\alpha_i + \beta_i = 1$

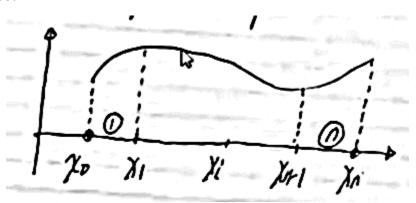
Since:

$$\begin{aligned} h_i &= x_i - x_{i-1} \\ h_{i+1} &= x_{i+1} - x_i \\ h_i + h_{i+1} &= x_{i+1} - x_{i-1} \end{aligned}$$

$$\alpha_i M_{i+1} + 2M_i + \beta_i M_{i+1} = 6f[x_{i+1}, x_i, x_{i-1}] = \gamma_i \quad ; \quad i = 1, 2, \dots, n-1$$

The boundary conditions

Case 1: The clamped



$$\begin{aligned} & \text{Given } C_1'(x_0) = f_0' \quad ; \quad C_n'(x_n) = f_n' \\ & C_1'(x_0) = -M_0 \frac{(x_1 - x_0)^2}{2h_1} + M_1 \frac{(x_0 - x_0)^2}{2h_1} - \left(f(x_0) - M_0 \frac{h_1^2}{6} \right) \frac{1}{h_1} + \left(f(x_1) - M_1 \frac{h_1^2}{6} \right) \frac{1}{h_1} \\ &= f_0' \\ & \to 2M_0 + M_1 = 6 \frac{f[x_1, x_0] - f_0'}{h_1} \overset{\text{def}}{=} \gamma_0 \\ & C_n'(x_n) = -M_{n-1} \frac{(x_n - x_n)^2}{2h_n} + M_n \frac{(x_n - x_{n+1})^2}{2h_n} - \left(f(x_{n+1}) - M_{n+1} \frac{h_n^2}{6} \right) \frac{1}{h_n} + \left(f(x_n) - M_n \frac{h_n^2}{6} \right) \frac{1}{h_n} \\ &= f_n' \\ & \to M_{n-1} + 2M_n = 6 \frac{f_n' - f[x_n, x_{n-1}]}{h_n} \overset{\triangle}{=} \gamma_n \end{aligned}$$

All the equations:

All the equations:
$$2M_0 + M_1 = \gamma_0$$

$$\alpha_1 M_0 + 2M_1 + \beta_1 M_2 = \gamma_1$$

$$\alpha_2 M_1 + 2M_2 + \beta_2 M_3 = \gamma_2$$

$$\vdots$$

$$\alpha_{n-1} M_{n-2} + 2M_{n-1} + \beta_{n-1} M_n = \gamma_{n-1}$$

$$M_{n-1} + 2M_n = \gamma_n$$

For the first row $eta_0=1$, and for the last row $lpha_n=1$ (eta_0 is added to make the equation look consistent)

Case 2, the natural boundary conditions:

Given:

$$M_0 = f_0^{\prime\prime}$$

$$M_n = f_n^{\prime\prime}$$

Let:

$$\beta_0 = \alpha_n = 0$$

$$\gamma_0 = 2M_0 = 2f_0^{"}$$

$$\gamma_n = 2M_n = 2f_n^{"}$$

Error and convergence:

Assume that $f(x) \in C^4[a,b]$, S(x) is the cubic spline interpolating function that satisfies clamped or natural boundary conditions.

Let
$$h = \max h_i$$
 (1 < i < n)
Where $h_i = x_i - x_{i-1}$

Then,

$$\left[\max_{x \in [a,b]} \right] \left| f_{(x)}^{(k)} - S_{(x)}^{(k)} \right| \le C_k \left[\max_{x \in [a,b]} \right] \left| f_{(x)}^{(k)} \right| \cdot h^{4-k}$$

For k = 0, 1, 2 with:

$$C_0 = \frac{5}{384}$$
 ; $C_1 = \frac{1}{24}$; $C_2 = \frac{3}{8}$

The interpolation is much better for the function itself, and it becomes worse for the derivatives.

As with all other functions, the accuracy of a derivative function is worse than the original function itself. Consider the coefficients as well, which get much larger as the order of the derivatives increases.

Consider k = 0, the function converges very quickly, at h^4

Consider k = 1, the derivative function converges more slowly, converging at h^3

Consider k = 2, the derivative functions converges even more slowly, converging at h^2