

# Data Analysis-I

## Interpolation

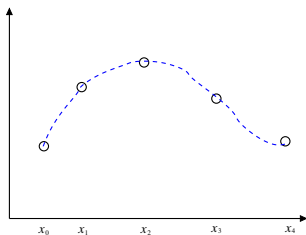
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# Interpolation

Estimate the value between data



- Polynomial interpolation
  - for a set of data  $\{(x_i, y_i)\}$ 
    - no knowledge of the relationship between  $y$  and  $x$
    - e.g.: estimate  $(x_{i+\delta}, y_{i+\delta})$  from the given dataset.
  - **Weierstrass theorem:** In general a continuous function  $f(x)$  in a finite interval  $x \in [a, b]$  can be fitted by a polynomial  $P(x)$ .
  - Find a polynomial approximation for  $f(x)$  from  $N$  pairs of numbers  $\{(x_i, f(x_i))\}$ , for  $i = 0, 1, \dots, N - 1$ .

# Lagrange Interpolation

- Construct a polynomial of the form

$$P(x) = \sum_{k=0}^{N-1} p_k(x) f(x_k) \quad (1)$$

## Lagrange Polynomial

$$p_k(x) = \frac{(x - x_0)(x - x_1) \cdots (x - x_{k-1})(x - x_{k+1}) \cdots (x - x_{N-1})}{(x_k - x_0)(x_k - x_1) \cdots (x_k - x_{k-1})(x_k - x_{k+1}) \cdots (x_k - x_{N-1})} \quad (2)$$

- If we use the Lagrange polynomial then it is clear that

$$P(x_i) = f(x_i) \quad (3)$$

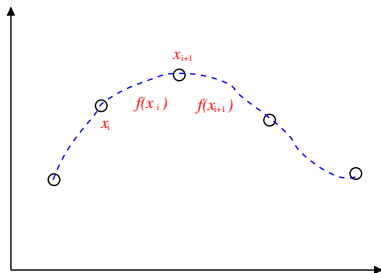
# Cubic-Spline

- spline (in dictionary): a piece of “flexible” wood or plastic that can be bent into arbitrary smooth shapes
  - in the days before computers, it was used to trace a smooth curve between points on a sheet of graph paper.
- Idea: use a simple function to approximate the relation between the dependent and independent variables
  - use the **third-degree** polynomial : “**cubic**”-spline

$$f(x) = a + bx + cx^2 + dx^3 \quad (4)$$

- To find the coefficients in each interval
  - use the condition for the continuity across each boundary

# Cubic-Spline



continuity condition:

$$f_i^{(n)}(x_i) = f_{i+1}^{(n)}(x_{i+1}) \quad \text{for } n\text{th derivative}$$

# Cubic-Spline

## Conditions for cubic spline

Given a dataset  $\{(x_i, y_i)\}$  in the interval  $[a, b]$  ( $a = x_0 < x_1 < \dots < x_n = b$ ), a cubic spline interpolant,  $f$ , for  $\{y_i\}$  is a function that satisfies the following condition:

- (a)  $f$  is a cubic polynomial, denoted by  $f_i$ , on the subinterval  $[x_i, x_{i+1}]$ .
- (b)  $f(x_i) = y_i$  for each  $i$ .
- (c)  $f_{i+1}(x_{i+1}) = f_i(x_{i+1})$  for each  $i$ .
- (d)  $f'_{i+1}(x_{i+1}) = f'_i(x_{i+1})$  for each  $i$ .
- (e)  $f''_{i+1}(x_{i+1}) = f''_i(x_{i+1})$  for each  $i$ .
- (f) One of the following sets of boundary condition is frequently used:
  - (i)  $f''(x_0) = f''(x_n)$  for natural or free boundary
  - (ii)  $f'(x_0) = y'(x_0)$  and  $f'(x_n) = y'(x_n)$  for clamped boundary

\* Other boundary conditions are also possible.

# Cubic-Spline

- To construct the cubic-spline interpolant for a given set of data  $\{y_i\}$ , we assume a third-order polynomial

$$f_i(x) = a_i + b_i(x - x_i) + c_i(x - x_i)^2 + d_i(x - x_i)^3 \quad (5)$$

for all  $i$ .

- from condition (b) and Eq. (5):

$$f_i(x_i) = a_i = y_i \quad (6)$$

- from condition (c):  $a_{i+1} = f_{i+1}(x_{i+1}) = f_i(x_{i+1})$ .
  - and if we let  $h_i \equiv x_{i+1} - x_i$  then we obtain

$$a_{i+1} = a_i + b_i h_i + c_i h_i^2 + d_i h_i^3 \quad (7)$$

- From Eq. (5)

$$b_i = f'_i(x_i) \quad (8)$$



# Cubic-Spline

- From condition (d) and  $f'_i(x) = b_i + 2c_i(x - x_i) + 3d_i(x - x_i)^2$ :

$$b_{i+1} = b_i + 2c_i h_i + 3d_i h_i^2 \quad (9)$$

- From condition (e):

$$c_i = f''(x_i)/2 \quad (10)$$

and

$$c_{i+1} = c_i + 3d_j h_j \quad (11)$$

# Cubic-Spline

Solving for  $d_i$  in Eq. (11) and substituting  $d_i$  into Eq. (7) and Eq. (9)

$$a_{i+i} = a_i + b_i h_i + \frac{h_i^3}{3} (2c_i + c_{j+1}) \quad (12)$$

and

$$b_{i+i} = b_i + h_i (c_i + c_{i+1}) \quad (13)$$

- Note: Eq. (8) and Eq. (10) do not give us any numerical information to determine  $\{b_i\}$  and  $\{c_i\}$ .
- We cannot determine  $\{b_i\}$  and  $\{c_i\}$  from Eq. (8) and Eq. (10).
- Therefore, we have to do more things!

# Cubic-Spline

The final relationship between the coefficients is obtained by first solving the appropriate equation in the form of Eq. (12) for  $b_i$ . From Eq. (12)

$$b_i = \frac{1}{h}(a_{i+1} - a_i) - \frac{h_i}{3}(2c_i + c_{i+1}) \quad (14)$$

and then, with reduction of the index ( $i \rightarrow i - 1$ ):

$$b_{i-1} = \frac{1}{h}(a_i - a_{i-1}) - \frac{h_{i-1}}{3}(2c_{i-1} + c_i) \quad (15)$$

Substituting Eq. (14) and Eq. (15) into Eq. (13), when the index is reduced by 1 ( $i \rightarrow i - 1$ ), we obtain the linear system of equations:

$$h_{i-1}c_{i-1} + 2(h_{i-1} + h_i)c_i + h_ic_{i+1} = \frac{3}{h_i}(a_{i+1} - a_i) - \frac{3}{h_{i-1}}(a_i - a_{i-1}), \quad (16)$$

for each  $i = 1, 2, \dots, n - 1$ .

# Cubic-Spline

- In Eq. (16), only  $\{c_i\}$  is a set of **unknown** parameters.
- In Eq. (16),  $\{h_i\}$  and  $\{a_i\}$  are given by the spacing between each data point and the value of  $\{y_i\}$ .
- Once the values of  $\{c_i\}$  is obtained, we can easily find  $\{b_i\}$  from Eq. (14).
- Then we can obtain  $\{d_i\}$  from Eq. (11).
- **Eq. (16) is an  $n$ -coupled linear equation.**
  - solve Eq. (16) using matrix.