Indian Institute of Technology Roorkee

CHN-323 Computer Applications in Chemical Engineering

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Interpretation

 \triangleright What will be f(3.5)?

\boldsymbol{x}	f(x)
2	8
3	27
4	64
5	125
6	216

Difference operators

> Let us say we have a data set

$$\{(f_j, x_j) | j = 1, 2, ..., m\}$$

Data is equally spaced

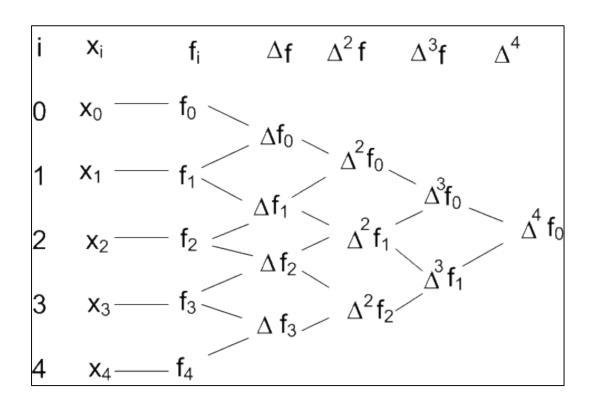
$$x_{i+1} = x_i + h$$

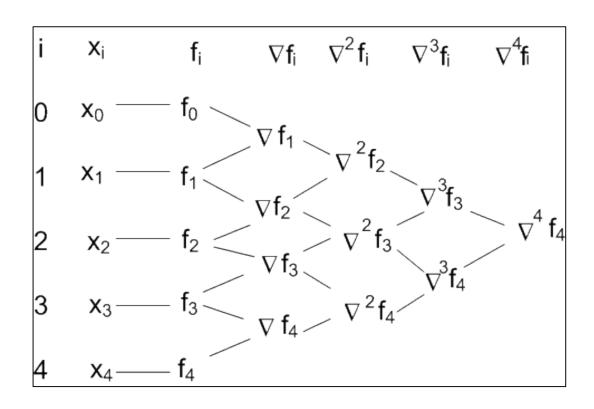
where $i = 1, 2,, m - 1$

- > h: step size, discretization interval
- $ightharpoonup \Delta f_i = f_{i+1} f_i$ Forward difference operator
- $ightharpoonup
 abla f_i = f_i f_{i-1}$ Backward difference operator
- $\succ Ef_i = f_{i+1}$ Shift operator

Difference operators: higher order

Difference tables





Shift operator

- $\succ Ef_i = f_{i+1}$
- $\triangleright E^2 f_i = f_{i+2}$

what are f_i and f_{i+1} ?

$$Ef(x_i) = f(x_i + h)$$

- \triangleright We know that $\Delta f_i = f_{i+1} f_i$
- This can be rewritten as $\Delta f_i = E f_i f_i = (E-1) f_i$ $\Delta = (E-1)$ Or $E = (1+\Delta)$
- \triangleright Generalizing this, $E^{\alpha} = (1 + \Delta)^{\alpha}$
- \triangleright Exercise: Prove that $E^2 = (1 + \Delta)^2$

Polynomial approximation of dataset

> Our data set

$$\{(f_j, x_j) | j = 1, 2, ..., m\}$$

This can be approximated as

$$f(x) = P_n(x) + R(x)$$

Where $P_n(x)$ is the polynomial approximation of f(x) and R(x) is the residual error.

 \triangleright What will be f(3.5)?

x	f(x)
2	8
3	27
4	64
5	125
6	216

➤ In other words, we need to find f(3 + 0.5) or f(2 + 1.5) or f(4 - 0.5).

Or
$$f(x_i + \alpha h)$$

$$f(x_i + \alpha h) = E^{\alpha} f(x_i) = (1 + \Delta)^{\alpha} f(x_i)$$

> Using binomial theorem, we can write

$$f(x_i + \alpha h)$$

$$= \left[1 + \alpha \Delta + \frac{\alpha(\alpha - 1)}{2!} \Delta^2 + \frac{\alpha(\alpha - 1)(\alpha - 2)}{3!} \Delta^3 + \cdots\right] f(x_i)$$

 \succ Can choose α to be 0, 1, ..., m-1 and indeed other non-integer real values to reach any x value in $[x_1, x_m]$

Numerical differentiation

- ightharpoonup We need to estimate $\frac{df}{dx}$
- > We have already written

$$f(x) = f(x_i + \alpha h) = \left[1 + \alpha \Delta + \frac{\alpha(\alpha - 1)}{2!} \Delta^2 + \frac{\alpha(\alpha - 1)(\alpha - 2)}{3!} \Delta^3 + \cdots\right] f(x_i)$$

> Thus we can write

$$f'(x) \approx \frac{1}{h} \left[\Delta + \frac{\alpha + (\alpha - 1)}{2} \Delta^2 + \frac{\{\alpha(\alpha - 1) + (\alpha - 1)(\alpha - 2) + \alpha(\alpha - 2)\}}{6} \Delta^3 + \cdots \right] f(x_i)$$

For x_1 , $\alpha = 0$

$$f'(x_1) \approx \frac{1}{h} \left[\Delta - \frac{1}{2} \Delta^2 + \frac{1}{3} \Delta^3 + \dots \pm \frac{1}{m-1} \Delta^{m-1} \right] f(x_1) + O(h^{m-1})$$

Consider only the first term

$$f'(x_1) \approx \frac{1}{h} \Delta f(x_1) + O(h)$$

 $\approx \frac{1}{h} \{f_2 - f_1\} + O(h)$

Two-points (2P)
First order accurate (FOA)
Forward difference (FDA)

Generalized form: $f'(x_i) \approx \frac{1}{h} \{f_{i+1} - f_i\} + O(h)$

Consider the first two terms

$$f'(x_1) \approx \frac{1}{h} \left[\Delta f(x_1) - \frac{1}{2} \Delta^2 f(x_1) \right] + O(h^2)$$

$$\approx \frac{1}{h} \left[(f_2 - f_1) - \frac{1}{2} (f_1 - 2f_2 + f_3) \right] + O(h^2)$$

$$\approx \frac{1}{2h} \left[-3f_1 + 4f_2 - f_3 \right] + O(h^2)$$

3P, SOA, FDA

Generalized form: $f'(x_i) \approx \frac{1}{2h} [-3f_i + 4f_{i+1} - f_{i+2}] + O(h^2)$

For x_2 , $\alpha = 1$

$$f'(x_2) \approx \frac{1}{h} \left[\Delta + \frac{1}{2} \Delta^2 - \frac{1}{6} \Delta^3 + \cdots \right] f(x_1) + O(h^{m-1})$$

> Consider the first two terms

$$f'(x_2) \approx \frac{1}{h} \left[\Delta f(x_1) + \frac{1}{2} \Delta^2 f(x_1) \right] + O(h^2)$$

$$\approx \frac{1}{h} \left[(f_2 - f_1) + \frac{1}{2} (f_1 - 2f_2 + f_3) \right] + O(h^2)$$

$$\approx \frac{1}{2h} [f_3 - f_1] + O(h^2)$$

2P, SOA, CDA

 \triangleright Generalizing this, we can write another formula for $f'(x_i)$

$$f'(x_i) \approx \frac{1}{2h} \{ f_{i+1} - f_{i-1} \} + O(h^2)$$

First Derivative		
Method	Formula	Truncation Error
Two-point forward dif- ference	$f'(x_i) = \frac{f(x_{i+1}) - f(x_i)}{h}$	O(h)
Three-point forward difference	$f'(x_i) = \frac{-3f(x_i) + 4f(x_{i+1}) - f(x_{i+2})}{2h}$	$O(h^2)$
Two-point central dif- ference	$f'(x_i) = \frac{f(x_{i+1}) - f(x_{i-1})}{2h}$	$O(h^2)$

Two-point backward difference	$f'(x_i) = \frac{f(x_i) - f(x_{i-1})}{h}$	O(h)
Three-point backward difference	$f'(x_i) = \frac{f(x_{i-2}) - 4f(x_{i-1}) + 3f(x_i)}{2h}$	$O(h^2)$

- > Second order derivatives
- > We have already written

$$f(x) = f(x_i + \alpha h) = \left[1 + \alpha \Delta + \frac{\alpha(\alpha - 1)}{2!} \Delta^2 + \frac{\alpha(\alpha - 1)(\alpha - 2)}{3!} \Delta^3 + \cdots \right] f(x_i)$$

> Thus we can write

$$f''(x) \approx \frac{1}{h^2} \left[\Delta^2 + \frac{\{\alpha + (\alpha - 1) + (\alpha - 1) + (\alpha - 2) + \alpha + (\alpha - 2)\}}{6} \Delta^3 + \cdots \right] f(x_i)$$

For x_1 , $\alpha = 0$

$$f''(x_1) \approx \frac{1}{h^2} \left[\Delta^2 - \Delta^3 + \frac{11}{12} \Delta^4 + \cdots \right] f(x_1)$$

> Consider only the first term

$$f''(x_1) \approx \frac{1}{h^2} [\Delta^2 f(x_1)] + O(h)$$

 $\approx \frac{1}{h^2} \{f_1 - 2f_2 + f_3\} + O(h)$ 3P, FOA, FDA

Generalized form:
$$f''(x_i) \approx \frac{1}{h^2} \{ f_i - 2f_{i+1} + f_{i+2} \} + O(h)$$

Consider the first two terms

$$f''(x_1) \approx \frac{1}{h^2} [\Delta^2 f(x_1) - \Delta^3 f(x_1)] + O(h^2)$$
$$\approx \frac{1}{h^2} \{2f_1 - 5f_2 + 4f_3 - f_4\} + O(h^2)$$
 4P, SOA, FDA

Generalized form:
$$f''(x_i) \approx \frac{1}{h^2} \{2f_i - 5f_{i+1} + 4f_{i+2} - f_{i+3}\} + O(h^2)$$

 \triangleright For x_2 , $\alpha = 1$

$$f''(x_2) \approx \frac{1}{h^2} [\Delta^2 + 0\Delta^3 + \cdots] f(x_1)$$

> Consider the first two terms

$$f''(x_2) \approx \frac{1}{h^2} [\Delta^2 f(x_1)] + O(h^2)$$

 $\approx \frac{1}{h^2} \{f_1 - 2f_2 + f_3\} + O(h^2)$ 3P, SOA, CDA

 \triangleright Generalizing this, we can write another formula for $f''(x_i)$

$$f''(x_i) \approx \frac{1}{h^2} \{ f_{i-1} - 2f_i + f_{i+1} \} + O(h^2)$$

Second Derivative		
Method	Formula	Truncation Error
Three-point forward difference	$f''(x_i) = \frac{f(x_i) - 2f(x_{i+1}) + f(x_{i+2})}{h^2}$	O(h)
Four-point forward difference	$f''(x_i) = \frac{2f(x_i) - 5f(x_{i+1}) + 4f(x_{i+2}) - f(x_{i+3})}{h^2}$	$O(h^2)$
Three-point central difference	$f''(x_i) = \frac{f(x_{i-1}) - 2f(x_i) + f(x_{i+1})}{h^2}$	$O(h^2)$

difference	$f''(x_i) = \frac{1}{h^2}$	O(h)
Four-point backward difference	$f''(x_i) = \frac{-f(x_{i-3}) + 4f(x_{i-2}) - 5f(x_{i-1}) + 2f(x_i)}{h^2}$	$O(h^2)$

Example

- \rightarrow What is f'(3.5)?
 - Using 2P, FOA, FDA
 - Using 2P, FOA, BDA
 - Using 2P, SOA, CDA
- \triangleright What is f''(3.5)?
 - Using 3P, SOA, CDA

X	f(x)
3.4	0.294118
3.5	0.285714
3.6	0.277778