

6. Numerical Differentiation

Given: $f_i = f(x_i)$ for $i=0,1,2,\dots,N$

Find: An approximation of $f'(x)$ or $f'(x_i)$

§ Polynomial approximation

Step 1: approximate $f(x)$ by the polynomial of degree N

$$f(x) \approx \sum_{j=0}^N f_j L_{N,j}(x) + \frac{1}{(N+1)!} f^{(N+1)}(\xi(x)) \cdot \prod_{j=0}^N (x-x_j)$$

Step 2: take derivative with respect to x :

$$f'(x) \approx \sum_{j=0}^N f_j L'_{N,j}(x) + \frac{1}{(N+1)!} \frac{d}{dx} \left\{ f^{(N+1)}(\xi(x)) \right\} \cdot \prod_{j=0}^N (x-x_j) + \frac{1}{(N+1)!} f^{(N+1)}(\xi(x)) \cdot \frac{d}{dx} \left\{ \prod_{j=0}^N (x-x_j) \right\}$$

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In particular, at $x = x_k$,

$$f'(x_k) \approx \sum_{j=0}^N f_j L'_{N,j}(x_k) + \frac{1}{(N+1)!} \frac{d}{dx} \left\{ f^{(N+1)}(\xi(x)) \right\}_{x=x_k} \cdot \prod_{j=0, j \neq k}^N (x_k - x_j) + \frac{1}{(N+1)!} f^{(N+1)}(\xi(x_k)) \cdot \frac{d}{dx} \left\{ \prod_{j=0, j \neq k}^N (x - x_j) \right\}_{x=x_k}$$

In remark, one may approximate $f'(x_k) \approx \sum_{j=0}^N f_j L'_{N,j}(x_k)$

with a **truncation error** $E_{tr} = \frac{1}{(N+1)!} f^{(N+1)}(\xi(x_k)) \cdot \prod_{j=0, j \neq k}^N (x_k - x_j)$

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§ Polynomial approximation

example: $N+1=3$, $\{x_0, x_1, x_2\} = \{x_0, x_0+h, x_0+2h\}$

$$\left. \begin{aligned} L_{N,0}(x) &= \frac{(x-x_1)(x-x_2)}{(x_0-x_1)(x_0-x_2)} \\ L_{N,1}(x) &= \frac{(x-x_0)(x-x_2)}{(x_1-x_0)(x_1-x_2)} \\ L_{N,2}(x) &= \frac{(x-x_0)(x-x_1)}{(x_2-x_0)(x_2-x_1)} \end{aligned} \right\} \begin{aligned} L'_{N,0}(x) &= \frac{(x-x_1)+(x-x_2)}{(x_0-x_1)(x_0-x_2)} = \frac{(x-x_1)+(x-x_2)}{2h^2} \\ L'_{N,1}(x) &= \frac{(x-x_0)+(x-x_2)}{(x_1-x_0)(x_1-x_2)} = \frac{(x-x_0)+(x-x_2)}{-h^2} \\ L'_{N,2}(x) &= \frac{(x-x_0)+(x-x_1)}{(x_2-x_0)(x_2-x_1)} = \frac{(x-x_0)+(x-x_1)}{2h^2} \end{aligned}$$

Therefore,

$$f'(x_0) \approx \sum_{j=0}^2 f_j L'_{N,j}(x_0) = f_0 \cdot \frac{-3h}{2h^2} + f_1 \cdot \frac{-2h}{-h^2} + f_2 \cdot \frac{-h}{2h^2} = \frac{1}{h} \left(-\frac{3}{2}f_0 + 2f_1 - \frac{1}{2}f_2 \right)$$

$$f'(x_1) \approx \sum_{j=0}^2 f_j L'_{N,j}(x_1) = f_0 \cdot \frac{-h}{2h^2} + f_1 \cdot \frac{0}{-h^2} + f_2 \cdot \frac{h}{2h^2} = \frac{1}{2h} (f_2 - f_0)$$

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truncation error $E_{tr} = \frac{1}{(N+1)!} f^{(N+1)}(\xi(x_k)) \cdot \prod_{j=0, j \neq k}^N (x_k - x_j)$

$$\begin{aligned} E_{tr}(x_0) &= \frac{1}{3!} f^{(3)}(\xi(x_0)) \cdot \prod_{j=1}^2 (x_0 - x_j) \\ &= \frac{1}{6} f^{(3)}(\xi(x_0)) \cdot (x_0 - x_1)(x_0 - x_2) \\ &= \frac{h^2}{3} f^{(3)}(\xi(x_0)) = O(h^2) \sim 2^{\text{nd}} \text{ order accuracy} \end{aligned}$$

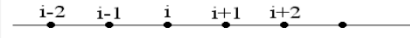
$$\begin{aligned} E_{tr}(x_1) &= \frac{1}{3!} f^{(3)}(\xi(x_1)) \cdot \prod_{j=0, j \neq 1}^2 (x_1 - x_j) \\ &= \frac{1}{6} f^{(3)}(\xi(x_1)) \cdot (x_1 - x_0)(x_1 - x_2) \\ &= -\frac{h^2}{6} f^{(3)}(\xi(x_1)) = O(h^2) \end{aligned}$$

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§ Finite Difference Approximation

Given: $f_i = f(x_i)$ for $x_i = a + i \times (b-a)/N$, $i=0,1,2,\dots,N$

Find: $f'(x_i)$



Step 1: Taylor series expansion $f(x)$ about $x = x_i$

$$f(x) = f(x_i) + (x - x_i) \cdot f'(x_i) + \frac{1}{2}(x - x_i)^2 \cdot f''(x_i) + \frac{1}{6}(x - x_i)^3 \cdot f^{(3)}(x_i) + \frac{1}{24}(x - x_i)^4 \cdot f^{(4)}(x_i) + \dots$$

$$f(x_{i+1}) = f_{i+1} = f_i + h \cdot f'(x_i) + \frac{1}{2}h^2 \cdot f''(x_i) + \frac{1}{6}h^3 \cdot f^{(3)}(x_i) + \frac{1}{24}h^4 \cdot f^{(4)}(x_i) + \dots$$

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$$(a) \quad f(x_{i+1}) = f_{i+1} = f_i + h \cdot f'(x_i) + \frac{1}{2}h^2 \cdot f''(x_i) + \frac{1}{6}h^3 \cdot f^{(3)}(x_i) + \frac{1}{24}h^4 \cdot f^{(4)}(x_i) + \dots$$

$$(b) \quad f(x_{i-1}) = f_{i-1} = f_i - h \cdot f'(x_i) + \frac{1}{2}h^2 \cdot f''(x_i) - \frac{1}{6}h^3 \cdot f^{(3)}(x_i) + \frac{1}{24}h^4 \cdot f^{(4)}(x_i) + \dots$$

From (a) or (b) or both

$$(a): \quad f_{i+1} - f_i = h \cdot f'(x_i) + \frac{1}{2}h^2 \cdot f''(x_i) + \dots$$

$$(b): \quad f_{i-1} - f_i = -h \cdot f'(x_i) + \frac{1}{2}h^2 \cdot f''(x_i) + \dots$$

$$(a) - (b): \quad f_{i+1} - f_{i-1} = 2h \cdot f'(x_i) + \frac{1}{3}h^3 \cdot f^{(3)}(x_i) + \dots$$

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§ Finite Difference Approximation

$$(i) \text{ forward difference: } f'(x_i) = \frac{f_{i+1} - f_i}{h} - \frac{1}{2}h \cdot f''(x_i) + O(h^2)$$

$$(ii) \text{ backward difference: } f'(x_i) = \frac{f_i - f_{i-1}}{h} + \frac{1}{2}h \cdot f''(x_i) + O(h^2)$$

$$(iii) \text{ central difference: } f'(x_i) = \frac{f_{i+1} - f_{i-1}}{2h} - \frac{1}{6}h^2 \cdot f^{(3)}(x_i) + O(h^4)$$

truncation error

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Using more points



$$(c): \quad f(x_{i+2}) = f_{i+2} = f_i + 2h \cdot f'(x_i) + \frac{1}{2} \cdot 4h^2 \cdot f''(x_i) + \frac{1}{6} \cdot 8h^3 \cdot f^{(3)}(x_i) + \frac{1}{24} \cdot 16h^4 \cdot f^{(4)}(x_i) + \dots$$

$$(d): \quad f(x_{i-2}) = f_{i-2} = f_i - 2h \cdot f'(x_i) + \frac{1}{2} \cdot 4h^2 \cdot f''(x_i) - \frac{1}{6} \cdot 8h^3 \cdot f^{(3)}(x_i) + \frac{1}{24} \cdot 16h^4 \cdot f^{(4)}(x_i) + \dots$$

$$(c) - (d): \quad f_{i+2} - f_{i-2} = 4h \cdot f'(x_i) + \frac{8}{3} \cdot h^3 \cdot f^{(3)}(x_i) + \dots$$

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$$(iv) \quad f'(x_i) = \frac{f_{i+1} - f_{i-2}}{4h} - \underbrace{\frac{2}{3}h^2 \cdot f^{(3)}(x_i)}_{\text{truncation error}} + O(h^4)$$

$$(iii) \quad \text{central difference: } f'(x_i) = \frac{f_{i+1} - f_{i-1}}{2h} - \underbrace{\frac{1}{6}h^2 \cdot f^{(3)}(x_i)}_{\text{truncation error}} + O(h^4)$$

$$4 \times (iii) - (iv): \quad 3f'(x_i) = 4 \cdot \frac{f_{i+1} - f_{i-1}}{2h} - \frac{f_{i+2} - f_{i-2}}{4h} + O(h^4)$$

$$\Rightarrow \quad f'(x_i) = \frac{f_{i-2} - 8f_{i-1} + 8f_{i+1} - f_{i+2}}{12h} + O(h^4)$$

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§ Finite Difference Approximation

- second derivatives:

$$f_{i+1} = f_i + h \cdot f'(x_i) + \frac{1}{2}h^2 \cdot f''(x_i) + \frac{1}{6}h^3 \cdot f^{(3)}(x_i) + \frac{1}{24}h^4 \cdot f^{(4)}(x_i) + \dots$$

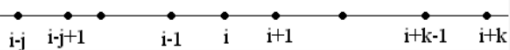
$$f_{i-1} = f_i - h \cdot f'(x_i) + \frac{1}{2}h^2 \cdot f''(x_i) - \frac{1}{6}h^3 \cdot f^{(3)}(x_i) + \frac{1}{24}h^4 \cdot f^{(4)}(x_i) + \dots$$

$$f_{i+1} + f_{i-1} = 2f_i + h^2 \cdot f''(x_i) + \frac{1}{12}h^4 \cdot f^{(4)}(x_i) + O(h^6)$$

$$f''(x_i) = \frac{f_{i+1} - 2f_i + f_{i-1}}{h^2} - \underbrace{\frac{1}{12}h^2 \cdot f^{(4)}(x_i)}_{\text{truncation error}} + O(h^6)$$

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IN GENERAL (e.g. higher order derivatives or higher order accuracy or even non-uniformly spaced grids)



- Select j points from LHS and k points from RHS to form the formula:

$$f^{(m)}(x_i) = \alpha_{-j}f_{i-j} + \alpha_{-j+1}f_{i-j+1} + \dots + \alpha_0f_i + \dots + \alpha_{k-1}f_{i+k-1} + \alpha_kf_{i+k}$$

of α 's (degrees of freedom) = $(j+1+k)$; $\alpha \propto h^{-m}$

- Taylor – series expand all f_j except f_i to obtain:

$$= \beta_0f_i + \beta_1f'(x_i) + \dots + \beta_mf^{(m)}(x_i) + \dots + \beta_{j+k}f^{(j+k)}(x_i) + O(\alpha h^{j+k+1})$$

$$\beta_i = \beta_i(\alpha)$$

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$$f^{(m)}(x_i) \approx \alpha_{-j}f_{i-j} + \alpha_{-j+1}f_{i-j+1} + \dots + \alpha_0f_i + \dots + \alpha_{k-1}f_{i+k-1} + \alpha_kf_{i+k}$$

$$= \beta_0f_i + \beta_1f'(x_i) + \dots + \beta_mf^{(m)}(x_i) + \dots + \beta_{j+k}f^{(j+k)}(x_i) + O(\alpha h^{j+k+1})$$

- choose $\{\alpha_i\}_{i=-j}^k$ such that

$$\begin{cases} \beta_m = 1 \\ \beta_i = 0 \quad \text{for } i = 0, 1, \dots, j+k; \quad i \neq m \end{cases}$$

- dimensional consistency: $\alpha_i = O(1/h^m)$

$$f^{(m)}(x_i) \approx \alpha_{-j}f_{i-j} + \alpha_{-j+1}f_{i-j+1} + \dots + \alpha_0f_i + \dots + \alpha_{k-1}f_{i+k-1} + \alpha_kf_{i+k}$$

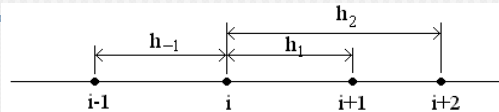
$$= f^{(m)}(x_i) + O(\alpha h^{j+k+1}) = f^{(m)}(x_i) + O(h^{j+k+1-m})$$

order = $j+k+1-m$

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example: $m = 2, j = 1$, and $k = 2$

$$f''(x_i) = \alpha_{-1}f_{i-1} + \alpha_0f_i + \alpha_1f_{i+1} + \alpha_2f_{i+2}$$



define $h_{-1} = |x_{i-1} - x_i|$, $h_1 = |x_{i+1} - x_i|$, $h_2 = |x_{i+2} - x_i|$

$$f_{i-1} = f(x_{i-1}) = f(x_i - h_{-1}) = f_i - h_{-1}f'_i + \frac{h_{-1}^2}{2}f''_i - \frac{h_{-1}^3}{6}f'''_i + O(h^4)$$

$$f_{i+1} = f(x_{i+1}) = f(x_i + h_1) = f_i + h_1f'_i + \frac{h_1^2}{2}f''_i + \frac{h_1^3}{6}f'''_i + O(h^4)$$

$$f_{i+2} = f(x_{i+2}) = f(x_i + h_2) = f_i + h_2f'_i + \frac{h_2^2}{2}f''_i + \frac{h_2^3}{6}f'''_i + O(h^4)$$

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$$f''(x_i) = \alpha_{-1}f_{i-1} + \alpha_0f_i + \alpha_1f_{i+1} + \alpha_2f_{i+2}$$

$$= \alpha_{-1} \left\{ f_i - h_{-1}f'_i + \frac{h_{-1}^2}{2}f''_i - \frac{h_{-1}^3}{6}f'''_i + O(h^4) \right\}$$

$$+ \alpha_0f_i$$

$$+ \alpha_1 \left\{ f_i + h_1f'_i + \frac{h_1^2}{2}f''_i + \frac{h_1^3}{6}f'''_i + O(h^4) \right\}$$

$$+ \alpha_2 \left\{ f_i + h_2f'_i + \frac{h_2^2}{2}f''_i + \frac{h_2^3}{6}f'''_i + O(h^4) \right\}$$

$$\begin{aligned} & \beta_0 = 0 \qquad \beta_1 = 0 \\ & (\alpha_{-1} + \alpha_0 + \alpha_1 + \alpha_2) f_i + (-\alpha_{-1}h_{-1} + \alpha_1h_1 + \alpha_2h_2) f'_i \\ & \beta_2 = 1 \qquad \beta_3 = 0 \\ & \left(\frac{\alpha_{-1}h_{-1}^2 + \alpha_1h_1^2 + \alpha_2h_2^2}{2} \right) f''_i + \left(\frac{-\alpha_{-1}h_{-1}^3 + \alpha_1h_1^3 + \alpha_2h_2^3}{2} \right) f'''_i + O(\alpha h^4) \end{aligned}$$

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$$f''(x_i) = \alpha_{-1}f_{i-1} + \alpha_0f_i + \alpha_1f_{i+1} + \alpha_2f_{i+2} + O(\alpha h^4) \quad \text{if}$$

$$\begin{cases} \beta_0 = \alpha_{-1} + \alpha_0 + \alpha_1 + \alpha_2 = 0 \\ \beta_1 = -\alpha_{-1}h_{-1} + \alpha_1h_1 + \alpha_2h_2 = 0 \\ \beta_2 = \frac{\alpha_{-1}h_{-1}^2 + \alpha_1h_1^2 + \alpha_2h_2^2}{2} = 1 \\ \beta_3 = -\alpha_{-1}h_{-1}^3 + \alpha_1h_1^3 + \alpha_2h_2^3 = 0 \end{cases} \quad \begin{matrix} O(h^2) \\ \\ \rightarrow \alpha \sim O(h^{-2}) \end{matrix}$$

• special case: uniformly spaced nodes: $h_{-1} = h_1 = h$ and $h_2 = 2h$

$$\begin{cases} \alpha_{-1} + \alpha_0 + \alpha_1 + \alpha_2 = 0 \\ -\alpha_{-1} + \alpha_1 + 2\alpha_2 = 0 \\ \alpha_{-1} + \alpha_1 + 4\alpha_2 = 2/h^2 \\ -\alpha_{-1} + \alpha_1 + 8\alpha_2 = 0 \end{cases} \Rightarrow \begin{cases} \alpha_{-1} = 1/h^2 \\ \alpha_0 = -2/h^2 \\ \alpha_1 = 1/h^2 \\ \alpha_2 = 0 \end{cases} \quad \begin{matrix} \text{(central difference)} \\ f''(x_i) = \frac{f_{i-1} - 2f_i + f_{i+1}}{h^2} + O(h^2) \end{matrix}$$

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§ Finite Difference Approximation

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

• unavoidable rounding error: $f_{i+1} = fl(f_{i+1}) + e_{i+1}$

$$f_{i-1} = fl(f_{i-1}) + e_{i-1}$$

$$\text{output}(f'_i) = \frac{fl(f_{i+1}) - fl(f_{i-1})}{2h} = \frac{f_{i+1} - f_{i-1}}{2h} - \frac{e_{i+1} - e_{i-1}}{2h}$$

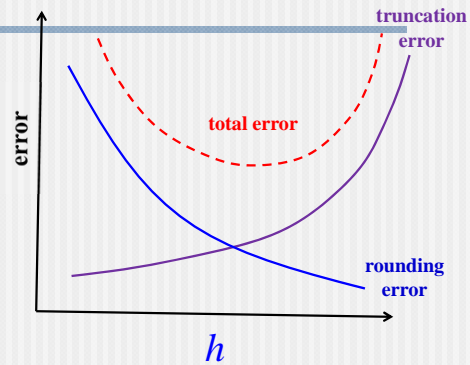
$$= f'_i + O(h^2) - \frac{e_{i+1} - e_{i-1}}{2h}$$

= exact value + truncation error + rounding error

↓ as h ↓ ↑ as h ↓

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§ Finite Difference Approximation



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§ Numerical Integration

Given: $\{(x_i, f_i)\}_{i=1}^N$ Want: $I = \int_a^b f(x) dx$

- polynomial approximation (global)

$$f(x) \approx P_{N-1}(x) = \sum_{i=1}^N f_i L_{N-1,i}(x)$$

$$I = \int_a^b f(x) dx \approx \sum_{i=1}^N f_i \int_a^b L_{N-1,i}(x) dx = \sum_{i=1}^N c_i f_i$$

$$c_i \equiv \int_a^b L_{N-1,i}(x) dx$$

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§ Numerical Integration --- given $f(x)$ for $a \leq x \leq b$

$$I = \int_a^b f(x) dx \approx \sum_{i=1}^N c_i f(x_i)$$

Question: What a choice of the nodes $\{x_i\}_{i=1}^N$ and a choice of the coefficients $\{c_i\}_{i=1}^N$ will give a "best" approximation?

- A "best" choice is recognized as the one that produces the exact result for the largest class of polynomials, $\Pi(R)$.

For any $f(x) \in \Pi(R)$, then $I = \int_a^b f(x) dx = \sum_{i=1}^N c_i f(x_i)$ instead of " \approx "

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§ Numerical Integration --- given $f(x)$ for $a \leq x \leq b$

$$I = \int_a^b f(x) dx \approx \sum_{i=1}^N c_i f(x_i)$$

- degrees of freedom = $2N$
- candidate of the largest class of polynomials: $\Pi_{2N-1}(R)$
i.e. polynomials of degree $\leq 2N-1$

$$P_{2N-1}(x) = a_{2N-1}x^{2N-1} + a_{2N-2}x^{2N-2} + \dots + a_1x + a_0 = \sum_{m=0}^{2N-1} a_m x^m$$

$$\int_a^b P_{2N-1}(x) dx = \sum_{i=1}^N c_i P_{2N-1}(x_i)$$

$$\sum_{m=0}^{2N-1} \int_a^b a_m x^m dx = \sum_{i=1}^N c_i \left(\sum_{m=0}^{2N-1} a_m x_i^m \right)$$

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$$\sum_{m=0}^{2N-1} a_m \left(\int_a^b x^m dx \right) = \sum_{m=0}^{2N-1} a_m \left(\sum_{i=1}^N c_i x_i^m \right)$$

"=" must hold for arbitrary $a_m, m=0,1,\dots,2N-1$

$$\sum_{i=1}^N c_i x_i^m = \int_a^b x^m dx = \frac{b^{m+1} - a^{m+1}}{m+1}$$

for $m = 0, 1, 2, \dots, 2N-1$

~ $2N$ nonlinear equations for $2N$ unknowns ~

- With the solutions of the above system of equations, we intend to approximate

$$I = \int_a^b f(x) dx \approx \sum_{i=1}^N c_i f(x_i)$$

for any given function $f(x)$.

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§ Alternative way to solve $\{x_i\}_{i=1}^N$ and $\{c_i\}_{i=1}^N$

Let $f(x) \in \Pi_{2N-1}(R)$ and $f(x_i) = f_i$. Then $f(x)$ can be written as

$$f(x) = P_{N-1}(x) + \phi_N(x) \sum_{m=0}^{N-1} \alpha_m x^m$$

$$\phi_N(x) = \prod_{i=1}^N (x - x_i)$$

$$P_{N-1}(x) = \sum_{i=1}^N f(x_i) \cdot L_{N-1,i}(x) = \sum_{i=1}^N f(x_i) \cdot \frac{\phi_N(x)}{(x - x_i) \phi'_N(x_i)}$$

= the unique polynomial of degree $N-1$ passing through the N points

$$\phi'_N(x) = \sum_{k=1}^N \prod_{\substack{j=1 \\ j \neq k}}^N (x - x_j) \Rightarrow \phi'_N(x_i) = \sum_{k=1}^N \prod_{\substack{j=1 \\ j \neq k}}^N (x_i - x_j) = \prod_{\substack{j=1 \\ j \neq i}}^N (x_i - x_j)$$

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"best choice": $\sum_{i=1}^N c_i f(x_i) = \int_a^b f(x) dx$

$$= \int_a^b P_{N-1}(x) dx + \int_a^b \left(\sum_{m=0}^{N-1} \alpha_m x^m \right) \phi_N(x) dx$$

Without loss of generality, we assume $[a,b]=[-1,1]$.

$$\sum_{i=1}^N c_i f(x_i) = \int_{-1}^1 \sum_{i=1}^N f(x_i) \cdot \frac{\phi_N(x)}{(x - x_i) \phi'_N(x_i)} dx + \int_{-1}^1 \left(\sum_{m=0}^{N-1} \alpha_m x^m \right) \phi_N(x) dx$$

$$= \sum_{i=1}^N f(x_i) \cdot \int_{-1}^1 \frac{\phi_N(x)}{(x - x_i) \phi'_N(x_i)} dx + \int_{-1}^1 \left(\sum_{m=0}^{N-1} \alpha_m x^m \right) \phi_N(x) dx$$

$$= \sum_{i=1}^N c_i f(x_i) + \int_{-1}^1 \left(\sum_{m=0}^{N-1} \alpha_m x^m \right) \phi_N(x) dx$$

= 0 for any polynomial of degree $\leq N-1$ 23

$\Rightarrow \phi_N(x)$ is the N^{th} polynomial that is orthogonal to all polynomials of degree $\leq N-1$

$\Rightarrow \phi_N(x)$ is the Legendre polynomial of degree N

§ Gaussian-Legendre quadrature method ($-1 \leq x \leq 1$)

For any arbitrary given function $f(x)$:

$$I = \int_{-1}^1 f(x) dx \approx \sum_{i=1}^N c_i f(x_i)$$

$$\phi_N(x) = \prod_{i=1}^N (x - x_i) = L_N(x) \text{ (Legendre polynomial)}$$

$$c_i = \int_{-1}^1 \frac{L_N(x)}{(x - x_i) L'_N(x_i)} dx$$

- "=" instead of " \approx " when $f(x)$ is a polynomial of degree $\leq 2N-1$

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§ Legendre polynomials $L_N(x)$

- a polynomial of degree N

$L_0(x) = 1$

$L_1(x) = x$

$L_2(x) = \frac{3}{2}x^2 - \frac{1}{2}$

$nL_n(x) = (2n-1)xL_{n-1}(x) - (n-1)L_{n-2}(x)$

- orthogonal to any polynomial of degree $\leq N-1$

$$\int_{-1}^1 P_{N-1}(x)L_N(x)dx = 0$$

- $\int_{-1}^1 L_n(x)L_m(x)dx = \frac{2}{2n+1}\delta_{nm}$

$$nL_n(x) = (2n-1)xL_{n-1}(x) - (n-1)L_{n-2}(x)$$

$$L_n(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0$$

$$L_{n-1}(x) = b_{n-1} x^{n-1} + b_{n-2} x^{n-2} + \dots + b_1 x + b_0$$

$$L_{n-2}(x) = c_{n-2} x^{n-2} + c_{n-3} x^{n-3} + \dots + c_1 x + c_0$$

$$n(a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0)$$

$$= \begin{cases} (2n-1)(b_{n-1} x^n + b_{n-2} x^{n-1} + \dots + b_1 x^2 + b_0 x) \\ -(n-1)(c_{n-2} x^{n-2} + c_{n-3} x^{n-3} + \dots + c_1 x + c_0) \end{cases}$$

$$n(a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0) = \begin{cases} (2n-1)(b_{n-1} x^n + b_{n-2} x^{n-1} + \dots + b_1 x^2 + b_0 x) \\ -(n-1)(c_{n-2} x^{n-2} + c_{n-3} x^{n-3} + \dots + c_1 x + c_0) \end{cases}$$

$x^n : na_n = (2n-1)b_{n-1}$

$x^{n-1} : na_{n-1} = (2n-1)b_{n-2}$

$x^{n-2} : na_{n-2} = (2n-1)b_{n-3} - (n-1)c_{n-2}$

$x^k : na_k = (2n-1)b_{k-1} - (n-1)c_k \text{ for } 1 \leq k \leq n-2$

$x^0 : na_0 = -(n-1)c_0$

REAL*8 :: A(O:N),B(O:N),C(O:N-1)

C(0)=0

$L_1(x)$

DO m=3,N

C(1)=1

tmpB=2d0*DBLE(m)-1d0

B(0)=-0.5d0

tmpC=DBLE(m)-1d0

B(1)=0

$L_2(x)$

A(m)=tmpB*B(m-1) $ma_m = (2m-1)b_{m-1}$

B(2)=1.5d0

A(m-1)=tmpB*B(m-2) $ma_{m-1} = (2m-1)b_{m-2}$

DO k=m-2,1,-1

A(k)=tmpB*B(k-1)-tmpC*C(k)

ENDDO $ma_k = (2m-1)b_{k-1} - (m-1)c_k$

A(0)=-tmpC*C(0) $ma_0 = -(m-1)c_0$

C(O:m-1)=B(O:m-1)

B(O:m)=A(O:m)/DBLE(m)

ENDDO

A(O:N)=B(O:N)

$$c_i = \int_{-1}^1 \frac{L_N(x)}{(x-x_i)L'_N(x_i)} dx \quad \bullet \quad \int_{-1}^1 L_n(x)L_m(x)dx = \frac{2}{2n+1}\delta_{nm}$$

$$= \frac{1}{L'_N(x_i)} \int_{-1}^1 P_{N-1}(x) dx \quad = \frac{1}{L'_N(x_i)} \sum_{k=0}^{N-1} a_k \cdot 2\delta_{k0}$$

$$= \frac{1}{L'_N(x_i)} \int_{-1}^1 \sum_{k=0}^{N-1} a_k L_k(x) dx \quad = \frac{2a_0}{L'_N(x_i)}$$

$$= \frac{1}{L'_N(x_i)} \sum_{k=0}^{N-1} a_k \int_{-1}^1 L_k(x) dx$$

$$= \frac{1}{L'_N(x_i)} \sum_{k=0}^{N-1} a_k \int_{-1}^1 L_k(x)L_0(x) dx$$

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§ Weighted Gaussian quadrature methods

Given: $f(x)$ and $\omega(x)$ for $a \leq x \leq b$

Want: $\{x_i\}_{i=1}^N$ and $\{c_i\}_{i=1}^N$ such that $\int_a^b \omega(x)f(x)dx \approx \sum_{i=1}^N c_i f(x_i)$

Theorem: Let $\omega(x)$ be a positive weight function and $\phi_N(x)$ be a nonzero polynomial of degree N that is ω -orthogonal to $\Pi_{N-1}(R)$, i.e.

$$\int_a^b \omega(x)\phi_N(x)P_{N-1}(x)dx = 0$$

Then, $\{x_i\}_{i=1}^N$ are the zeros of $\phi_N(x)$ and $c_i = \int_a^b \frac{\omega(x)\phi_N(x)}{(x-x_i)\phi'_N(x_i)} dx$

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§ Weighted Gaussian quadrature methods

- $\omega(x)=1, (a,b)=(-1,1) : \phi_N(x)=L_N(x)$ (Legendre)
- $\omega(x)=1/\sqrt{1-x^2}, (a,b)=(-1,1) : \phi_N(x)=T_N(x)$ (Chebychev)
- $\omega(x)=\exp(-x^2), (a,b)=(-\infty,\infty) : \phi_N(x)=H_N(x)$ (Hermite)

Theorem:

- The quadrature formula holds exactly for all $f(x) \in \Pi_{2N-1}(R)$.
- For $f \in C^{2N}[a,b]$, the error term is

$$\frac{f^{(2N)}(\xi)}{(2N)!} \int_a^b \phi_N^2(x)\omega(x)dx \quad \text{for some } a < \xi < b$$

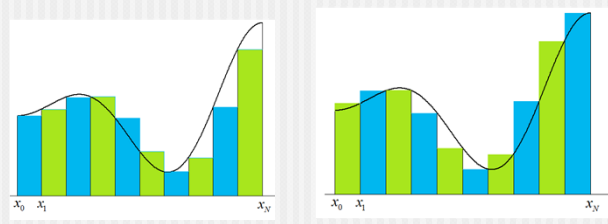
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§ Piecewise approaches

Given: $f_i = f(x_i)$ for $a = x_0 < x_1 < x_2 < \dots < x_{N-1} < x_N = b$

Want: $I = \int_a^b f(x)dx$

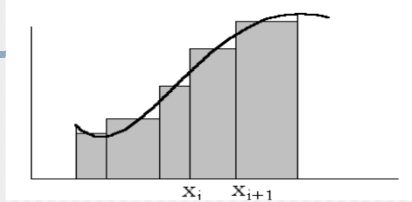
Simpliest way: $I \approx \sum_{i=0}^{N-1} f_i(x_{i+1} - x_i)$ or $\sum_{i=1}^N f_i(x_i - x_{i-1})$



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§ Piecewise approaches

- rectangular rule



$$I = \int_a^b f(x) dx = \sum_{i=0}^{N-1} \left(\int_{x_i}^{x_{i+1}} f(x) dx \right) \approx \sum_{i=0}^{N-1} \left\{ f\left(\frac{x_i + x_{i+1}}{2}\right) (x_{i+1} - x_i) \right\}$$

- error analysis: define $\bar{x}_i = (x_i + x_{i+1})/2$

$$\text{for } x \in [x_i, x_{i+1}]: f(x) = f(\bar{x}_i) + (x - \bar{x}_i) f'(\bar{x}_i) + \frac{(x - \bar{x}_i)^2}{2} f''(\bar{x}_i) + O(\Delta x^3)$$

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$$\int_{x_i}^{x_{i+1}} f(x) dx = \int_{x_i}^{x_{i+1}} \left\{ f(\bar{x}_i) + (x - \bar{x}_i) f'(\bar{x}_i) + \frac{(x - \bar{x}_i)^2}{2} f''(\bar{x}_i) + O(\Delta x^3) \right\} dx$$

$$= f(\bar{x}_i)(x_{i+1} - x_i) + \frac{f''(\bar{x}_i)}{24} \cdot (x_{i+1} - x_i)^3 + O(\Delta x^5)$$

$$I = \int_a^b f(x) dx = \sum_{i=0}^{N-1} \left(\int_{x_i}^{x_{i+1}} f(x) dx \right)$$

$$= \sum_{i=0}^{N-1} \left\{ f(\bar{x}_i)(x_{i+1} - x_i) + \frac{f''(\bar{x}_i)}{24} \cdot (x_{i+1} - x_i)^3 + O(\Delta x^5) \right\}$$

$$= \sum_{i=0}^{N-1} \left\{ f(\bar{x}_i)(x_{i+1} - x_i) + O(\Delta x^3) \right\}$$

$$= \sum_{i=0}^{N-1} \left\{ f(\bar{x}_i)(x_{i+1} - x_i) \right\} + N \cdot O(\Delta x^3)$$

$$\text{total truncation error} \sim \frac{(b-a)}{\Delta x} \cdot O(\Delta x^3) \sim (b-a) O(\Delta x^2)$$

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Problem: We have $f(x_i)$ but not $f(\bar{x}_i)$.

Solution: make sure $x_i = \frac{x_{i+1} + x_{i-1}}{2}$, that is $x_{i+1} - x_i = x_i - x_{i-1}$.

$$I = \int_a^b f(x) dx = \int_{x_0}^{x_2} f(x) dx + \int_{x_2}^{x_4} f(x) dx + \int_{x_4}^{x_6} f(x) dx + \dots + \int_{x_{N-2}}^{x_N} f(x) dx$$

$$\approx f(x_1)(x_2 - x_0) + f(x_3)(x_4 - x_2) + \dots + f(x_{N-1})(x_N - x_{N-2})$$

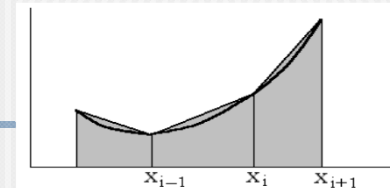
$$\approx \sum_{i=0}^{N/2-1} f(x_{2i+1})(x_{2i+2} - x_{2i})$$

$$\approx 2h \cdot \sum_{i=0}^{N/2-1} f(x_{2i+1}) \text{ if } (x_{2i+2} - x_{2i}) = \text{constant} = 2h$$

$$\text{total truncation error} \sim O((b-a)\Delta x^2)$$

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- Trapezoidal rule



$$I = \int_a^b f(x) dx = \sum_{i=0}^{N-1} \left(\int_{x_i}^{x_{i+1}} f(x) dx \right) \approx \sum_{i=0}^{N-1} \left\{ (x_{i+1} - x_i) \left(\frac{f(x_{i+1}) + f(x_i)}{2} \right) \right\}$$

total truncation error $\sim O((b-a)\Delta x^2)$

for $x \in [x_i, x_{i+1}]: f(x) = f(\bar{x}_i) + (x - \bar{x}_i) f'(\bar{x}_i) + \frac{(x - \bar{x}_i)^2}{2} f''(\bar{x}_i) + O(\Delta x^3)$

$$f(x_i) = f(\bar{x}_i) - \frac{(x_{i+1} - x_i)}{2} f'(\bar{x}_i) + \frac{(x_{i+1} - x_i)^2}{8} f''(\bar{x}_i) + O(\Delta x^3)$$

$$f(x_{i+1}) = f(\bar{x}_i) + \frac{(x_{i+1} - x_i)}{2} f'(\bar{x}_i) + \frac{(x_{i+1} - x_i)^2}{8} f''(\bar{x}_i) + O(\Delta x^3)$$

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- **Simpson rule** Provided $x_i = \frac{x_{i+1} + x_{i-1}}{2}$ or $x_{i+1} - x_i = x_i - x_{i-1} \equiv h_i$.

$$\int_{x_{i-1}}^{x_{i+1}} f(x) dx = \frac{h_i}{3} (f_{i-1} + 4f_i + f_{i+1}) + O(h_i^5)$$

total truncation error $\sim O((b-a)\Delta x^4)$

< show >

$$\int_{x_{i-1}}^{x_{i+1}} f(x) dx = \int_{x_{i-1}}^{x_{i+1}} \left\{ \begin{aligned} &f(x_i) + (x-x_i)f'(x_i) + \frac{(x-x_i)^2}{2}f''(x_i) \\ &+ \frac{(x-x_i)^3}{6}f'''(x_i) + \frac{(x-x_i)^4}{24}f^{(4)}(x_i) + O(\Delta x^5) \end{aligned} \right\} dx$$

$$= 2h_i f_i + \frac{1}{3} h_i^3 f_i'' + \frac{1}{60} h_i^5 f_i^{(4)} + O(h_i^7)$$

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central difference method: $f_i'' = \frac{f_{i+1} - 2f_i + f_{i-1}}{h_i^2} + O(h_i^2)$

$$\begin{aligned} \int_{x_{i-1}}^{x_{i+1}} f(x) dx &= 2h_i f_i + \frac{1}{3} h_i^3 f_i'' + \frac{1}{60} h_i^5 f_i^{(4)} + O(h_i^7) \\ &= 2h_i f_i + \frac{1}{3} h_i^3 \left\{ \frac{f_{i+1} - 2f_i + f_{i-1}}{h_i^2} + O(h_i^2) \right\} + \frac{1}{60} h_i^5 f_i^{(4)} + O(h_i^7) \\ &= 2h_i f_i + \frac{1}{3} h_i (f_{i+1} - 2f_i + f_{i-1}) + O(h_i^5) \\ &= \frac{h_i}{3} (f_{i-1} + 4f_i + f_{i+1}) + O(h_i^5) \end{aligned}$$

if $h_i = \text{constant} = h$:

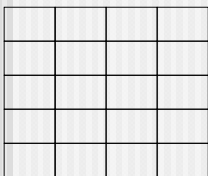
$$\int_a^b f(x) dx \approx \sum_{i=0}^{N/2-1} \frac{h}{3} (f_{2i+2} + 4f_{2i+1} + f_{2i}) + O((b-a)\Delta x^4)$$

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§ Multiple Integrals

- regular domains

$$\int_a^b dx \int_c^d f(x, y) dy \approx \int_a^b dx \sum_{j=0}^N \beta_j f(x, y_j)$$



$$= \sum_{j=0}^N \beta_j \int_a^b f(x, y_j) dx$$

$$\approx \sum_{j=0}^N \beta_j \left(\sum_{i=0}^M \alpha_i f(x_i, y_j) \right)$$

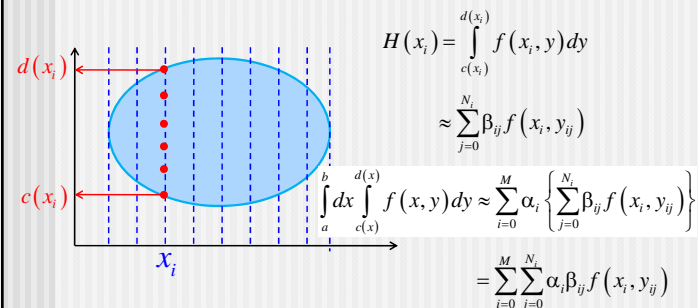
$$= \sum_{j=0}^N \sum_{i=0}^M \alpha_i \beta_j f(x_i, y_j)$$

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§ Multiple Integrals

- irregular domains

$$\int_a^b dx \int_{c(x)}^{d(x)} f(x, y) dy \equiv \int_a^b H(x) dx \approx \sum_{i=0}^M \alpha_i H(x_i)$$



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§ Richardson's extrapolation

- construct a better answer based on some unsatisfactory answers.

Given a 1st-order formula $N_1(h)$ which approximates a quantity M , i.e.

$$M = N_1(h) + C_1h + C_2h^2 + C_3h^3 + \dots$$

Want: an answer with a better accuracy

$$(1) \quad M = N_1(h) + C_1h + C_2h^2 + C_3h^3 + \dots$$

$$(2) \quad M = N_1\left(\frac{h}{2}\right) + C_1\frac{h}{2} + C_2\frac{h^2}{4} + C_3\frac{h^3}{8} + \dots$$

$$2 * (2) - (1) = (3): \quad M = 2N_1\left(\frac{h}{2}\right) - N_1(h) - C_2\frac{h^2}{2} - C_3\frac{3h^3}{4} + \dots$$

$$N_2(h) \equiv 2N_1\left(\frac{h}{2}\right) - N_1(h) \sim \text{2nd order accuracy}$$

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§ Richardson's extrapolation

$$(3) \quad M = N_2(h) - C_2\frac{h^2}{2} - C_3\frac{3h^3}{4} + \dots \quad N_2(h) = 2N_1\left(\frac{h}{2}\right) - N_1(h)$$

$$(4) \quad M = N_2\left(\frac{h}{2}\right) - C_2\frac{h^2}{8} - C_3\frac{3h^3}{32} + \dots \quad N_2\left(\frac{h}{2}\right) = 2N_1\left(\frac{h}{4}\right) - N_1\left(\frac{h}{2}\right)$$

$$\frac{4 * (4) - (3)}{3} : \quad M = \frac{4N_2\left(\frac{h}{2}\right) - N_2(h)}{3} + C_3\frac{3h^3}{8} + \dots$$

$$N_3(h) = \frac{4}{3}N_2\left(\frac{h}{2}\right) - \frac{1}{3}N_2(h)$$

Conclusion:

$$M = N_j(h) + O(h^j)$$

$$N_j(h) = N_{j-1}\left(\frac{h}{2}\right) + \frac{1}{2^{j-1}-1} \left(N_{j-1}\left(\frac{h}{2}\right) - N_{j-1}(h) \right) \text{ for } j \geq 2$$

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$$N_1(h)$$

$$N_1\left(\frac{h}{2}\right) \quad N_2(h)$$

$$N_2(h) = 2N_1\left(\frac{h}{2}\right) - N_1(h)$$

$$N_1\left(\frac{h}{4}\right) \quad N_2\left(\frac{h}{2}\right) \quad N_3(h)$$

$$N_2\left(\frac{h}{2}\right) = 2N_1\left(\frac{h}{4}\right) - N_1\left(\frac{h}{2}\right)$$

$$N_1\left(\frac{h}{8}\right) \quad N_2\left(\frac{h}{4}\right) \quad N_3\left(\frac{h}{2}\right) \quad N_4(h)$$

$$N_3(h) = \frac{4}{3}N_2\left(\frac{h}{2}\right) - \frac{1}{3}N_2(h)$$



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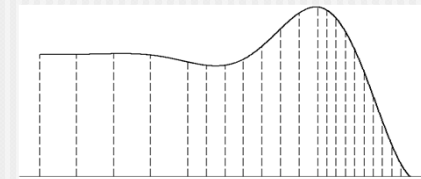
§ Adaptive Quadrature methods

Given: $f(x)$ and $\epsilon > 0$

Want: $I = \int_a^b f(x) dx$ within the specified tolerance ϵ

Idea: very dense grid points! \Rightarrow how dense?

Efficiency desired: more points in regions where $f(x)$ has large variations.

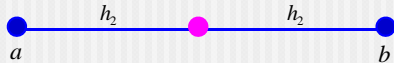


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Example: Simpson rule

$$\int_{x_{i-1}}^{x_{i+1}} f(x) dx = \frac{h_i}{3} (f_{i-1} + 4f_i + f_{i+1}) - \frac{h_i^5}{90} f^{(4)}(\mu) \quad \text{for some } \mu \in (x_i, x_{i+1})$$

Step 1: 3 equally spaced nodes, $h_2 \equiv (b-a)/2$



$$\begin{aligned} I &= \int_a^b f(x) dx = \frac{h_2}{3} (f(a) + 4f(a+h_2) + f(b)) - \frac{h_2^5}{90} f^{(4)}(\mu) \\ &\equiv S(a,b) - \frac{h_2^5}{90} f^{(4)}(\mu) \\ &\quad \text{for some } \mu \in (a,b) \end{aligned}$$

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Step 2: 5 equally spaced nodes



$$\begin{aligned} I &= \int_a^{a+h_2} f(x) dx + \int_{a+h_2}^b f(x) dx \\ &= \frac{h_2}{3} \left(f(a) + 4f\left(a + \frac{h_2}{2}\right) + f(a+h_2) \right) \\ &\quad + \frac{h_2}{3} \left(f(a+h_2) + 4f\left(a + \frac{3h_2}{2}\right) + f(b) \right) - 2 \times \left(\frac{h_2}{2} \right)^5 \frac{1}{90} f^{(4)}(\bar{\mu}) \\ &= S(a, a+h_2) + S(a+h_2, b) - \left(\frac{1}{16} \right) \frac{h_2^5}{90} f^{(4)}(\bar{\mu}) \end{aligned}$$

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$$\begin{cases} I = S(a,b) - \frac{h_2^5}{90} f^{(4)}(\mu) \\ I = S(a, a+h_2) + S(a+h_2, b) - \left(\frac{1}{16} \right) \frac{h_2^5}{90} f^{(4)}(\bar{\mu}) \end{cases}$$

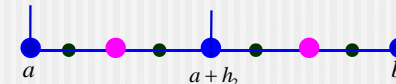
Assume $f^{(4)}(\mu) \approx f^{(4)}(\bar{\mu})$

$$\text{error} = \left| \left(\frac{1}{16} \right) \frac{h_2^5}{90} f^{(4)}(\bar{\mu}) \right| \approx \frac{1}{15} |S(a,b) - S(a, a+h_2) - S(a+h_2, b)|$$

$< \varepsilon$? yes \Rightarrow take the 5-point answer!

\sim The error is estimated by using the 3-point and 5-point answer for any given range $[a, b]$.

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$$I = \int_a^{a+h_2} f(x) dx + \int_{a+h_2}^b f(x) dx = I_{1/2}^{(1)} + I_{1/2}^{(2)}$$

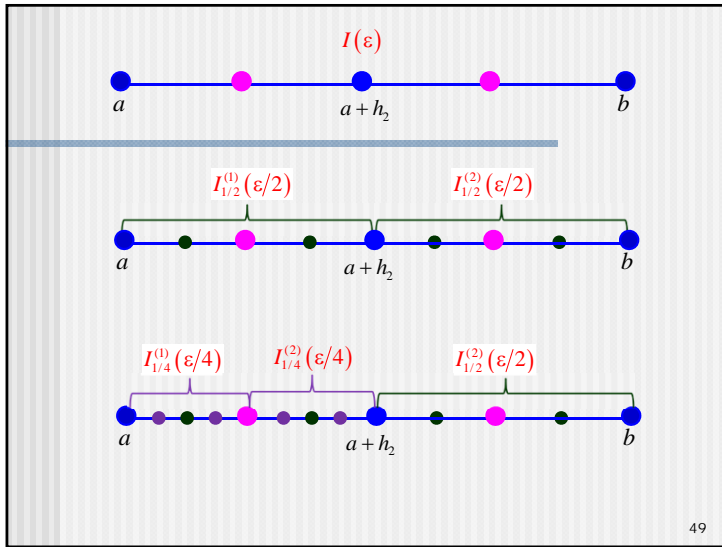
I within $\varepsilon \Rightarrow I_{1/2}^{(1)}$ within $\varepsilon/2$ and $I_{1/2}^{(2)}$ within $\varepsilon/2$

$$I_{1/2}^{(1)} = S\left(a, a + \frac{h_2}{2}\right) + S\left(a + \frac{h_2}{2}, a+h_2\right) + \text{error}_1$$

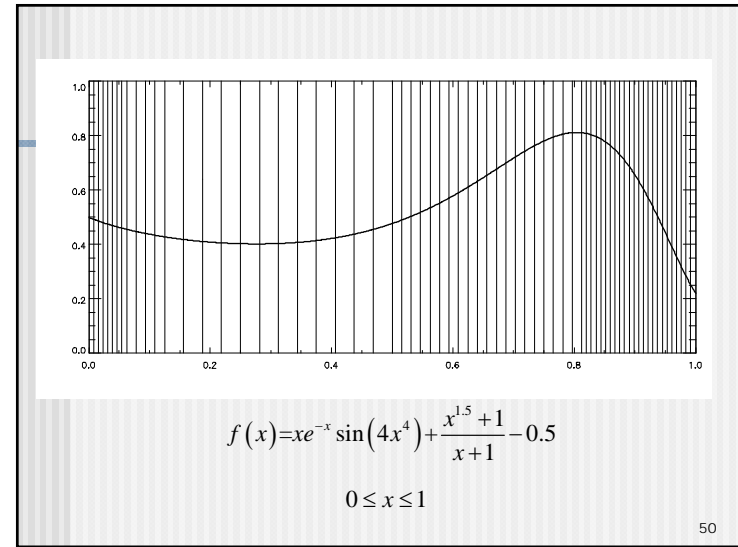
$$\text{error}_1 \approx \frac{1}{15} \left| S(a, a+h_2) - S\left(a, a + \frac{h_2}{2}\right) - S\left(a + \frac{h_2}{2}, a+h_2\right) \right| < \frac{\varepsilon}{2} ?$$

$$I_{1/2}^{(2)} = S\left(a+h_2, a + \frac{3h_2}{2}\right) + S\left(a + \frac{3h_2}{2}, b\right) + \text{error}_2$$

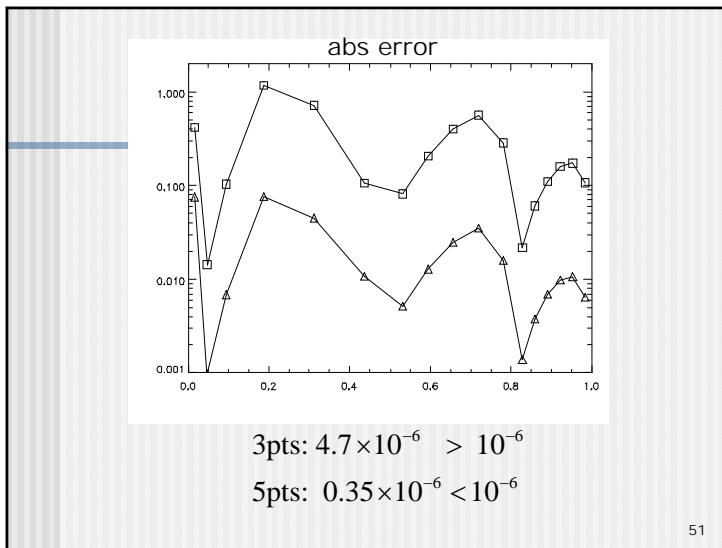
$$\text{error}_2 \approx \frac{1}{15} \left| S(a+h_2, b) - S\left(a+h_2, a + \frac{3h_2}{2}\right) - S\left(a + \frac{3h_2}{2}, b\right) \right| < \frac{\varepsilon}{2} \quad 48$$



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