

lecture 3-1

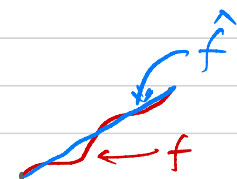
Curve fitting & interpolation

- Curve fitting
 - linear regression
 - general linear regression
 - interpolation
 - Direct polynomial method
 - Newton's interpolation method
 - Lagrange's interpolation method
 - piecewise interpolation
- Integration
- Trapezoidal rule (linear interpolation)
 - Simpson's $\frac{1}{2}$ nd rule (quadratic interpolation)
 - Simpson's $\frac{3}{8}$ th rule (cubic interpolation)
 - Integration using general order polynomial interpolation
 - Error in numerical integration
 - Integration of functions

• Error in numerical integration

Trapezoidal rule

$$E_t^i = \int_{x_i}^{x_{i+1}} f(x) dx - \int_{x_i}^{x_{i+1}} \hat{f}(x) dx$$



\hat{f} linear interpolation using

$$(x_i, f(x_i)), (x_{i+1}, f(x_{i+1}))$$

we can show

$$E_t^i = -\frac{(x_{i+1}-x_i)^3}{12} f^{(2)}(\xi_i)$$

$$f^{(k)}(x) = \frac{d^k f}{dx^k}$$

$$\therefore E_t = \sum_{i=1}^n E_t^i$$

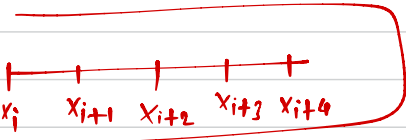
$$= -\frac{(b-a)^3}{12n^3} \sum_{i=1}^n f^{(2)}(\xi_i)$$

$$= -\frac{(b-a)^3}{12n^2} \bar{f}^{(2)}$$

assuming

$$x_{i+1} - x_i = h = \frac{(b-a)}{n}$$

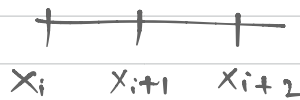
$$\bar{f}^{(k)} = \frac{1}{n} \sum_{i=1}^n f^{(k)}(\xi_i)$$



• Simpson's $\frac{1}{3}$ rd rule (quadratic interpolation)

$$E_t^I = -\frac{1}{90} \left(\frac{x_{i+2} - x_i}{2} \right)^5 f^{(4)}(\xi_i)$$

$$= -\frac{h^5}{90} f^{(4)}(\xi_i)$$



assuming

$$x_{i+1} - x_i = \frac{b-a}{n} = h$$

even segment
- odd points
↓
due to symmetry
higher accuracy

$$I = 1 \rightarrow (x_1, x_3)$$

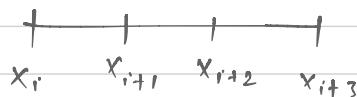
$$I = 2 \rightarrow (x_3, x_5)$$

⋮

$$I = m \rightarrow (x_{n-1}, x_{n+1})$$

sum E_t^I to get total error

• Simpson's $\frac{3}{8}$ th rule



$$E_t^I = -\frac{3}{80} \left(\frac{x_{i+3} - x_i}{3} \right)^5 f^{(4)}(\xi)$$

$$= -\frac{3}{80} h^5 f^{(4)}(\xi)$$

$$\begin{aligned} x_{i+1} - x_i &= x_{i+2} - x_{i+1} \\ &= \dots = \frac{b-a}{n} \\ &= h \end{aligned}$$

Compare this with error in Simpson's $\frac{1}{3}$ rd rule: Simpson's $\frac{1}{3}$ rd rule accuracy is comparable while using one order lower interpolation!!

$$I=1 \rightarrow (x_1, x_4)$$

$$I=2 \rightarrow (x_4, x_7)$$

Integration of function

$$f: [a, b] \rightarrow (-\infty, \infty)$$

we can compute f at discrete points (x_1, x_2, \dots, x_n)

$$(y_1 = f(x_1), y_2 = f(x_2), \dots, y_n = f(x_n))$$

Idea is to choose discrete points (x_1, x_2, \dots, x_n) more wisely to further reduce errors.

- Gauss Quadrature method
- Richardson's method
- Adaptive Quadrature

Gauss quadrature method

Example of 2nd order quadrature method

Usual interpolation method

$$\hat{I}[f] = c_1 f(a) + c_2 f(b)$$

$$\bullet \quad I[1] = \hat{I}[1]$$

$$\Rightarrow c_1 + c_2 = \int_a^b 1 \, dx = b-a$$

$$\bullet \quad I[x] = \hat{I}[x]$$

$$\Rightarrow c_1 a + c_2 b = \int_a^b x \, dx = \frac{b^2 - a^2}{2}$$

$$f = \alpha + \beta x$$

$$I[f] = \hat{I}[f]$$

$$\bullet \quad I[f] = \int_a^b (\alpha + \beta x) \, dx$$

$$= \alpha \int_a^b dx + \beta \int_a^b x \, dx$$

$$= \alpha I[1] + \beta I[x]$$

$$= \alpha \hat{I}[1] + \beta \hat{I}[x]$$

$$= \hat{I}[\alpha + \beta x]$$

Solve for c_1 and c_2

$$c_1 = c_2 = \frac{b-a}{2}$$

Gauss Quadrature method

$$\hat{I}[f] = c_1 f(x_1) + c_2 f(x_2)$$

find c_1, c_2, x_1, x_2

$$\bullet \quad I[1] = \hat{I}[1]$$

$$\bullet \quad I[x] = \hat{I}[x]$$

$$\bullet \quad I[x^2] = \hat{I}[x^2]$$

$$\bullet \quad I[x^3] = \hat{I}[x^3]$$

Eq 1

$$c_1 + c_2 = b-a \Rightarrow c_1 = \frac{b-a}{2}$$

Eq 2

$$c_1 x_1 + c_2 x_2 = \frac{b^2 - a^2}{2} \Rightarrow c(x_1 + x_2) = \frac{b^2 - a^2}{2}$$

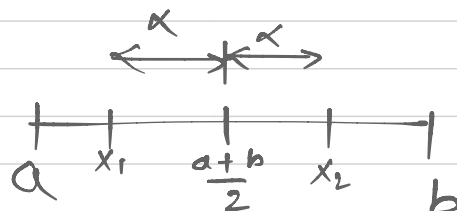
$$\checkmark c(a+b) = \frac{b^2 - a^2}{2}$$

Eq 3

$$c_1 x_1^2 + c_2 x_2^2 = \frac{b^3 - a^3}{3} \Rightarrow$$

Eq 4

$$c_1 x_1^3 + c_2 x_2^3 = \frac{b^4 - a^4}{4}$$



$$\bullet \quad c_1 = c_2 = c$$

$$\bullet \quad x_1 = \frac{(a+b)}{2} - \alpha$$

$$x_2 = \frac{(a+b)}{2} + \alpha$$

$$C(x_1^2 + x_2^2) = \frac{b^3 - a^3}{3}$$

$$\rightarrow C\left(\left(\frac{a+b}{2} - \alpha\right)^2 + \left(\frac{a+b}{2} + \alpha\right)^2\right) = \frac{b^3 - a^3}{3}$$

$$\rightarrow \left(\frac{a+b}{2}\right)^2 + \alpha^2 - \cancel{(a+b)\alpha} + \left(\frac{a+b}{2}\right)^2 + \alpha^2 + \cancel{(a+b)\alpha} = \frac{1}{c} \frac{(b^3 - a^3)}{3}$$

$$\rightarrow 2\alpha^2 = \frac{1}{c} \frac{(b^3 - a^3)}{3} - 2 \frac{(a+b)^2}{2}$$

\rightarrow

Proof that 2nd
order Gauss quadrature
is accurate upto cubic
polynomials

$$f = \alpha_1 + \alpha_2 x + \alpha_3 x^2 + \alpha_4 x^3$$

$$\begin{aligned} \mathcal{I}[f] &= \alpha_1 \mathcal{I}[1] + \alpha_2 \mathcal{I}[x] \\ &\quad + \alpha_3 \mathcal{I}[x^2] \\ &\quad + \alpha_4 \mathcal{I}[x^3] \end{aligned}$$

$$\begin{aligned} &= \alpha_1 \hat{\mathcal{I}}[1] + \alpha_2 \hat{\mathcal{I}}[x] \\ &\quad + \alpha_3 \hat{\mathcal{I}}[x^2] \\ &\quad + \alpha_4 \hat{\mathcal{I}}[x^3] \end{aligned}$$

$$= \hat{\mathcal{I}}[\alpha_1 + \alpha_2 x + \alpha_3 x^2 + \alpha_4 x^3]$$

Lecture 32

- Gauss Quadrature Method (Gauss-Legendre formulas)

for $\hat{I}[f] = c_1 f(x_1) + c_2 f(x_2)$ $f: [a, b] \rightarrow (-\infty, \infty)$

find c_1, c_2, x_1, x_2 s.t.

- $\hat{I}[1] = \hat{I}[1]$
- $\hat{I}[x] = \hat{I}[x]$
- $\hat{I}[x^2] = \hat{I}[x^2]$
- $\hat{I}[x^4] = \hat{I}[x^4]$

we found that

- $c_1 = c_2 = \frac{b-a}{2} = c$

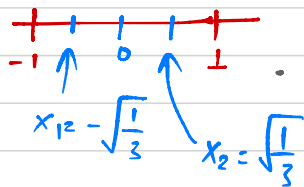
- $x_1 = \frac{b+a}{2} - \alpha, \quad x_2 = \frac{b+a}{2} + \alpha$ where

$$2\alpha^2 = \frac{1}{c} \frac{(b^3 - a^3)}{3} - \frac{(a+b)^2}{2}$$

Take $a = -1, b = 1$ then • $c_1 = c_2 = 1 = c$ ← weights

- $2\alpha^2 = \frac{1}{1} \frac{(1+1)}{3} - \frac{0^2}{2} = \frac{2}{3}$

$$\Rightarrow \alpha = \sqrt{\frac{1}{3}}$$



$$\begin{aligned} x_1 &= \frac{b+a}{2} - \alpha = -\sqrt{\frac{1}{3}} \\ x_2 &= \frac{b+a}{2} + \alpha = \sqrt{\frac{1}{3}} \end{aligned}$$
 ← quadrature points

$$\hat{I}[f] = f\left(-\frac{1}{\sqrt{3}}\right) + f\left(\frac{1}{\sqrt{3}}\right) \quad \text{for } f: [-1, 1] \rightarrow (-\infty, \infty)$$

Change of variable to approximate integration over any interval $[a, b]$

let $f: [a, b] \rightarrow (-\infty, \infty)$

$$I[f] = \int_a^b f(x) dx,$$

Consider

$$z = \alpha + \beta x$$

choose α, β s.t.

$$\Rightarrow x = \frac{z - \alpha}{\beta}, \quad dx = \frac{dz}{\beta}$$

- $z = -1$ at $x = a$

then

$$I[f] = \int_{-1}^1 f\left(\frac{z-\alpha}{\beta}\right) \frac{dz}{\beta}$$

$$= \frac{1}{\beta} \int_{-1}^1 f\left(\frac{z-\alpha}{\beta}\right) dz$$

define

$$g(z) = f\left(\frac{z-\alpha}{\beta}\right)$$

then

$$g: [-1, 1] \rightarrow (-\infty, \infty)$$

$$\therefore I[f] = \frac{1}{\beta} I[g]$$

$$= \frac{1}{\beta} \left(\int_{-1}^1 g(x) dx \right)$$

we quadrature formula that we obtained for

$$f: [-1, 1] \rightarrow (-\infty, \infty).$$

$$\hat{I}[g] = g\left(-\frac{1}{\sqrt{2}}\right) + g\left(\frac{1}{\sqrt{2}}\right) \Rightarrow$$

Higher order formula

$$I[f] \approx \frac{1}{\beta} \hat{I}[g] = \frac{1}{\beta} \left(g\left(-\frac{1}{\sqrt{2}}\right) + g\left(\frac{1}{\sqrt{2}}\right) \right) = \frac{1}{\beta} f\left(\frac{-\frac{1}{\sqrt{2}} - \alpha}{\beta}\right) + \frac{1}{\beta} f\left(\frac{\frac{1}{\sqrt{2}} - \alpha}{\beta}\right)$$

we can consider $\hat{I}[f] = G f(x_1) + G_2 f(x_2) + \dots + G_n f(x_n)$

which has 2n unknowns so we can exactly

integrate upto $(2n-1)$ th order polynomial

$$\bullet I[1] = \hat{I}[1]$$

$$\bullet I[x] = \hat{I}[x]$$

$$\bullet z = 1 \text{ at } x = b$$

$$\frac{1}{\beta} \left. \begin{aligned} \alpha + \beta a &= -1 \\ \alpha + \beta b &= 1 \end{aligned} \right\}$$

$$\alpha + \beta b = 1$$

$$\beta(a-b) = -1-1$$

$$\Rightarrow \beta = \frac{2}{b-a}$$

$$\therefore \alpha = 1 - \frac{2}{b-a} b$$

$$\therefore z = \frac{b-a-2b}{b-a} + \frac{2}{b-a} x$$

$$\Rightarrow z = -\frac{(a+b)}{b-a} + \frac{2}{b-a} x$$

check

$$\text{when } x=a, \quad z = -\frac{(a+b)}{b-a} + \frac{2a}{b-a}$$

$$= \frac{1}{b-a} (2a - a - b)$$

$$= -1$$

$$\text{when } x=b, \quad z = 1.$$

$$\left. \begin{array}{l} \vdots \\ \cdot \end{array} \right\} \rightarrow \text{an equation}$$

$$I[x^{2n-1}] = \hat{I}[x^{2n-1}]$$

Table 20.1 provides values of unknowns upto $n=6$

Richardson's Extrapolation

$$f: [a, b] \rightarrow (-\infty, \infty)$$

$$h_1, \quad x_1 = a, \quad x_2 = x_1 + h_1, \quad x_3 = x_2 + h_1, \dots, \quad x_{n_1} = b$$

$$h_2, \quad x_1 = a, \quad x_2 = x_1 + h_2, \quad x_3 = x_2 + h_2, \dots, \quad x_{n_2} = b$$

$$I[f] = \int_a^b f(x) dx \approx \hat{I}[f] = \frac{h_1}{2} [f(x_1) + f(x_{n_1}) + 2(f(x_2) + \dots + f(x_{n_1-1}))]$$

$$I[f] \approx \hat{I}[f] = \frac{h_2}{2} [f(x_1) + \dots]$$

lets $\hat{I}[h_1]$ is approximation of $I[f]$ using h_1

$\hat{I}[h_2]$ is \dots approximation of $I[f]$ using h_2

I is exact integral of f

$$I = \hat{I}[h_1] + E[h_1], \quad I = \hat{I}[h_2] + E[h_2]$$

Here $E[h_1], E[h_2]$ are errors due to approximation of integral $I[f]$

$$\hat{I}[h_1] + E[h_1] = \hat{I}[h_2] + E[h_2]$$

$$E[h_1] = -\frac{(b-a)^2}{12} h_1^2 \bar{f}^{(2)}[h_1], \quad \bar{f}^{(2)}[h_i] = \frac{1}{n_i} \sum_{i=1}^{n_i} f^{(2)}(z_{hi})$$

$$E[h_2] = -\frac{(b-a)^2}{12} h_2^2 \bar{f}^{(2)}[h_2]$$

Assume that $\bar{f}^{(2)}[h]$ is almost constant for different h

$$\frac{E[h_1]}{E[h_2]} \approx \frac{h_1^2}{h_2^2} \rightarrow E[h_1] \approx E[h_2] \frac{h_1^2}{h_2^2}$$

from $\hat{I}[h_1] + E[h_1] = \hat{I}[h_2] + E[h_2]$

$$\hat{I}[h_1] + \frac{h_1^2}{h_2^2} E[h_2] = \hat{I}[h_2] + E[h_2]$$

$$\Rightarrow E[h_2] = \frac{\hat{I}[h_1] - \hat{I}[h_2]}{1 - \frac{h_1^2}{h_2^2}}$$

use it with $I = \hat{I}[h_2] + E[h_2]$

$$I \approx \hat{I}[h_2] + \frac{\hat{I}[h_1] - \hat{I}[h_2]}{1 - \frac{h_1^2}{h_2^2}}$$

I assume $h_2 < h_1$

$h_1 = h, \quad h_2 = \frac{h}{2}$

$$I \approx \hat{I}[h/2] + \frac{\hat{I}[h] - \hat{I}[h/2]}{1 - 4}$$

$$= \hat{I}[h/2] - \frac{1}{3} (\hat{I}[h] - \hat{I}[h/2])$$

We have effectively built more accurate approximation of $I[f]$ using two Trapezoidal approximations

$$I \approx \frac{4}{3} \hat{I}[h/2] - \frac{1}{3} \hat{I}[h]$$

Idea can be used to similarly combine two Simpson's $\frac{1}{3}$ rd rule approximation to get higher order approximation !!!

Adaptive Quadrature Method



Choose more points in parts where function varies a lot and use fewer points elsewhere !!

