

Lecture 3-I

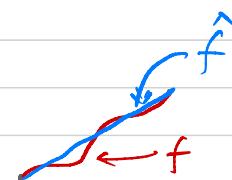
Curve fitting & interpolation

- 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 $\frac{1}{3}$ rd 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

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



$$x_i \quad x_{i+1}$$

\hat{f} linear interpolation using

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

we can show

$$E_t^i = -\frac{(x_{i+1}-x_i)^3}{12} f^{(2)}(z_i) \quad f^{(k)}(x) = \frac{d^k f}{dx^k}$$

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

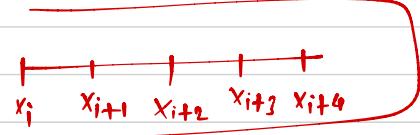
$$= -\frac{(b-a)^3}{12 n^3} \sum_{i=1}^n f^{(2)}(z_i)$$

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

- assuming

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

$$\overline{f}^{(k)} = \frac{1}{n} \sum_{i=1}^n f^{(k)}(z_i)$$



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

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

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



even segment

- odd points

↓
due to symmetry
higher accuracy

assuming

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

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

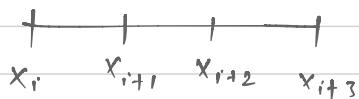
$$I = 2 \rightarrow (x_3, x_5)$$

.

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

} sum E_t^I to get total error

- Simpson's $\frac{3}{8}$ th rule



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

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

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

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

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 !!

$$x_{i+1} - x_i = x_{i+2} - x_{i+1}$$

$$= \dots = \frac{b-a}{n}$$

$$= h$$

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

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

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

$$\rightarrow [c_1 + c_2 = \int_a^b 1 dx = b-a]$$

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

$$\rightarrow [c_1 a + c_2 b = \int_a^b x dx = \frac{b^2 - a^2}{2}]$$

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

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

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

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

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

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

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

Solve for c_1 and c_2

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

Gauss Quadrature method

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

$$\text{find } c_1, c_2, x_1, x_2$$

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

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

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

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

Eq 1

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

Eq 2

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

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

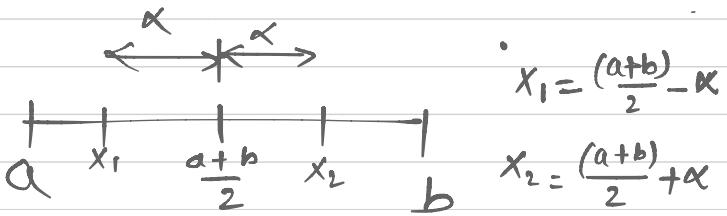
Eq 3

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

Eq 4

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

$$\cdot c_1 = c_2 = c$$



$$\cdot x_1 = \frac{(a+b)}{2} - x$$

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

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

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

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

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

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Proof that 2nd
order Gauss quadrature

is accurate upto cubic
polynomials

$$\begin{aligned}f = \alpha_1 + \alpha_2 x + \alpha_3 x^2 \\ + \alpha_4 x^3\end{aligned}$$

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

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

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

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.

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

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

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

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

we found that

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

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

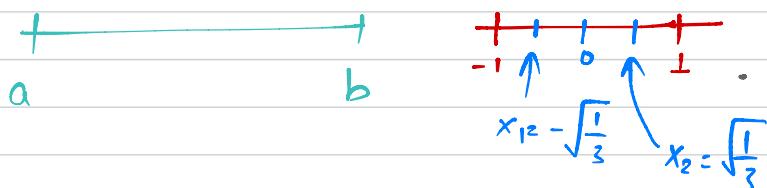
$$2\alpha^2 = \frac{1}{c} \frac{(b^3 - a^3)}{3} - \frac{(a+b)^2}{2}$$

Take $a = -1, b = 1$ then

$$c = c_2 = 1 = c \quad \leftarrow \text{weights}$$

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

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



$$x_1 = \frac{b+a}{2} - \alpha = -\sqrt{\frac{1}{3}}$$

$$x_2 = \frac{b+a}{2} + \alpha = \sqrt{\frac{1}{3}}$$

\leftarrow 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.

$$\Rightarrow x = \frac{z-\alpha}{\beta}, \quad dx = \frac{dz}{\beta} \quad \cdot \quad z = -1 \text{ at } x = a$$

• $z = 1$ at $x = b$

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

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

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

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

use quadrature formula that

we obtained for

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

$$I[g] \quad g(-\frac{1}{\sqrt{2}}) + g(\frac{1}{\sqrt{2}}) \Rightarrow$$

Higher order formula

we can consider

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

$$= \frac{1}{\beta} f\left(\frac{-\frac{1}{\sqrt{2}} - \alpha}{\beta}\right) + \frac{1}{\beta} f\left(\frac{\frac{1}{\sqrt{2}} - \alpha}{\beta}\right)$$

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

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

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

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

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

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

check

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

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

$$= -1$$

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

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

which has $2n$ unknowns so we can exactly

integrate upto $(2n-1)^{\text{th}}$ order polynomial

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

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



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

$\} \rightarrow 2n$ equations

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

Richardson's Extrapolation

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

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

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

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

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

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

$\hat{I}[h_2]$ is _____ using h_2

I is exact integral of f

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

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

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

$$E[h_1] = -\frac{(b-a)^2}{12} h_1^2 \bar{f}^{(2)}[h_1], \quad \bar{f}^{(2)}[h_1] = \frac{1}{n_1} \sum_{i=1}^{n_1} f^{(2)}(z_i)$$

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

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

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

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

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

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

\hat{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 $\frac{1}{3}$ rule approximations to get
higher order approximation !!!

Adaptive Quadrature Method

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

