

E_t , True Error = True value - Approximate value

Relative true error, $E_t = \frac{\text{True error}}{\text{True value}} = \frac{E_t}{\text{True Value}}$

E_a = approximate error = Present appxn. - Previous appxn.

E_a = relative approximate error = $\frac{\text{approximate error } (E_a)}{\text{Present approximation}}$

Absolute relative approximate error = $|E_a|$

FDD $f'(x) \approx \frac{f(x+\Delta x) - f(x)}{\Delta x}$

BDD $f'(x) \approx \frac{f(x) - f(x-\Delta x)}{\Delta x}$

CDD $f'(x) \approx \frac{f(x+4\Delta x) - f(x-4\Delta x)}{2\Delta x}$

$f''(x_i) \approx \frac{f(x_{i+2}) - 2f(x_{i+1}) + f(x_i)}{(\Delta x)^2}$ [FDD] [not important]

$ E_a \%$	m
≈ 5	1
≈ 0.5	2
≈ 0.05	3
≈ 0.005	4

Bisection Method

$$x_L = \dots$$

$$x_U = \dots$$

~~f(x)~~

if $f(x_L) f(x_U) > 0$:

return "no root between x_L & x_U . break"

$$x_m = \frac{x_L + x_U}{2}$$

for i in range(∞):

while (!0):

$$x_{old} = x_m$$

if ($f(x_L) f(x_m) < 0$):

$$x_U = x_m$$

$$x_L = x_L - \frac{(x_L + x_U)}{2}$$

else if ($f(x_L) f(x_m) > 0$):

$$x_L = x_m$$

else if ($f(x_L) f(x_m) == 0$):

$$x_{new}^m = \frac{x_L + x_U}{2}$$

root found! break

$$|\epsilon_a| = \left| \frac{x_{new}^m - x_{old}^m}{x_{new}^m} \right|$$

100%

0.1%

0.01%

0.001%

0.0001%

Newton - raphson

$$x_0 = \dots$$

$$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$$

$$|\epsilon_a| = \left| \frac{x_{i+1} - x_i}{x_{i+1}} \right| \times 100\%$$

if $|\epsilon_a| \leq \epsilon_s$, stop

$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)}$$

$$x_2 = x_1 - \frac{f(x_1)}{f'(x_1)}$$

$$x_3 = x_2 - \frac{f(x_2)}{f'(x_2)}$$

$$|\epsilon_a|_1 = \left| \frac{x_1 - x_0}{x_1} \right| \times 100\%$$

Secant method

$$x_{-1} = \dots$$

$$x_0 = \dots$$

$$x_{i+1} = x_i - \frac{f(x_i)(x_0 - x_{-1})}{f(x_i) - f(x_{-1})}$$

$$|\epsilon_a| = \left| \frac{x_{i+1} - x_i}{x_{i+1}} \right| \times 100\%$$

$$x_1 = x_0 - \frac{f(x_0)(x_0 - x_{-1})}{f(x_0) - f(x_{-1})}$$

$$x_2 = x_1 - \frac{f(x_1)(x_1 - x_0)}{f(x_1) - f(x_0)}$$

$$x_3 = x_2 - \frac{f(x_2)(x_2 - x_1)}{f(x_2) - f(x_1)}$$

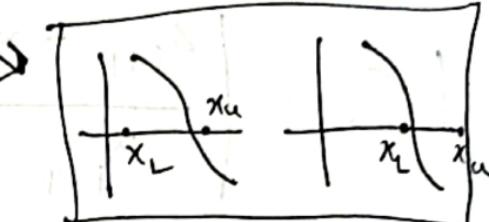
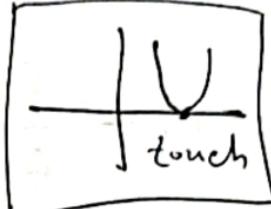
Advantage

guaranteed
to converge

bracket
halved at
each iteration

disadvantage

slow convergence

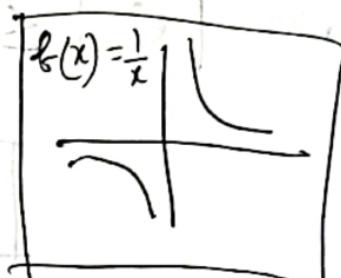


a guy with a plan?

Do I really look like

very poor choice of words

It's not about
the money
It's about sending
a message



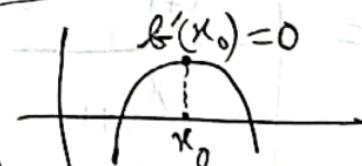
Fast
convergence

only one
initial
guess

Newton
Raphson

divergence @
inflection point

division by zero



Oscillation near local
maxima / minima

root jumping

Advantage

fast
convergence
(if b)

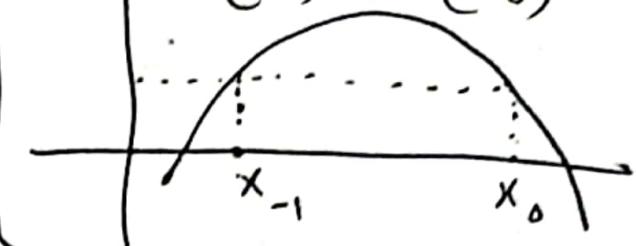
secant
method

two initial
guesses don't
have to bracket
root

disadvantage

division by zero

$$f(x_{-1}) = f(x_0)$$



root jumping

$$\frac{d}{dt} s = v$$

$$\frac{d}{dt} v = a$$

$$v = \int a dt$$

$$s = \int v dt$$

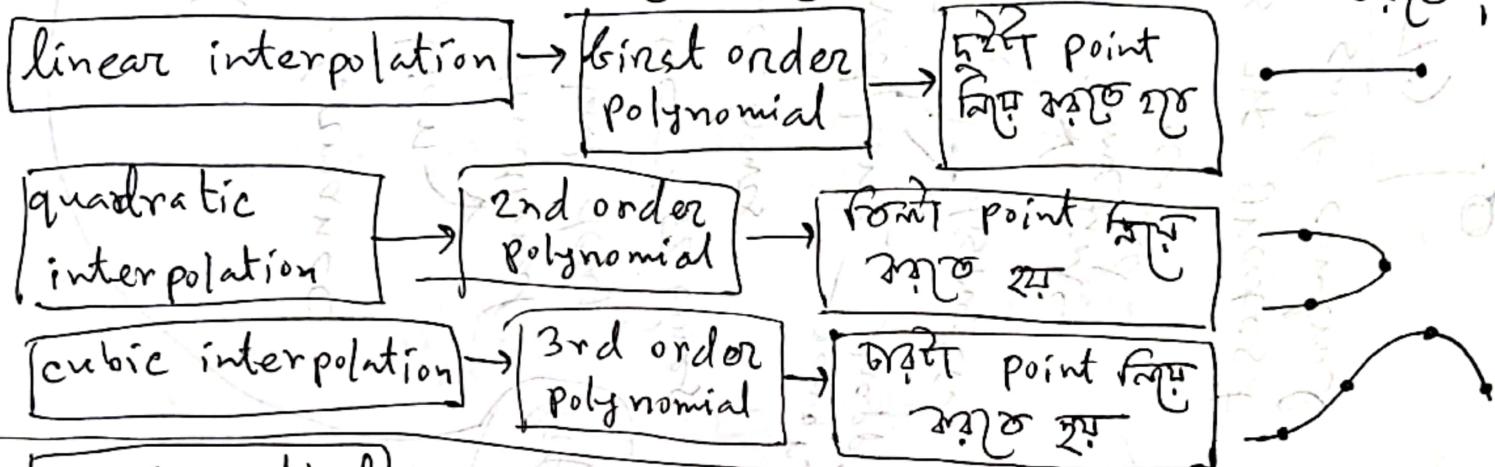
Interpolation

Direct method

Newton's
Divided difference
method

Lagrangian
interpolation

दो दिए गए फल (x, y) के लिए यह ताकि $x = \dots$ पर उन्हें interpolation करें।



Direct method: n -th order polynomial निम्न अवधि $n+1$ point interpolant a अवधि system solve करते हैं a_0, a_1, \dots, a_{n+1} के लिए

order	interpolant
1st order	$v(t) = a_0 + a_1 t$
2nd order	$v(t) = a_0 + a_1 t + a_2 t^2$
3rd order	$v(t) = a_0 + a_1 t + a_2 t^2 + a_3 t^3$

Newton's Divided difference → quadratic interpolation

$$b_1(x) = b_0 + b_1 (x - x_0)$$

$$b_0 = b(x_0) = y_0$$

$$b_1 = \frac{b(x_1) - b(x_0)}{x_1 - x_0} = \frac{y_1 - y_0}{x_1 - x_0}$$

linear
inter
polation

$$f_2(x) = b_0 + b_1(x - x_0) + b_2(x - x_0)(x - x_1)$$

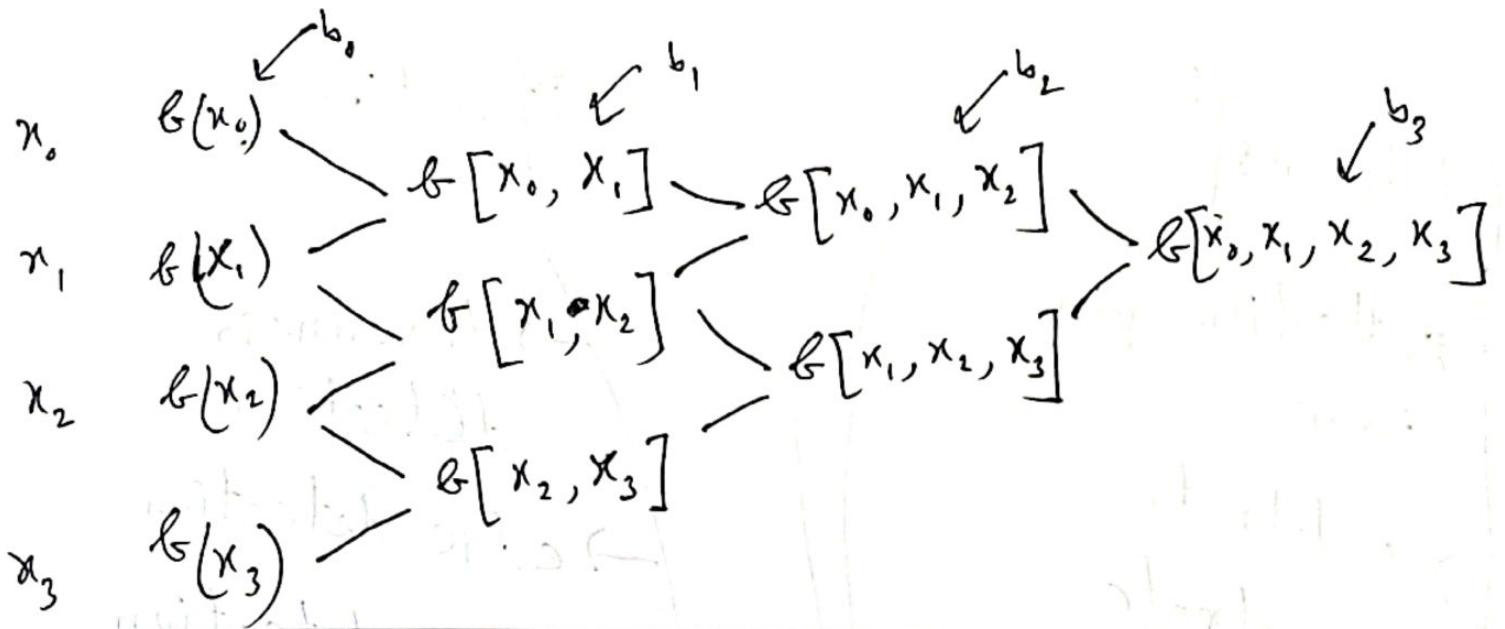
$$b_0 = f(x_0) = y_0$$

$$b_1 = \frac{f(x_1) - f(x_0)}{x_1 - x_0}$$

$$b_2 = \frac{\frac{f(x_2) - f(x_1)}{x_2 - x_1} - \frac{f(x_1) - f(x_0)}{x_1 - x_0}}{x_2 - x_0}$$

given points	
x_0	y_0
x_1	y_1
x_2	y_2

given points	
x_0	y_0
x_1	y_1



Lagrange's Interpolation

$$f_n(x) = \sum_{i=0}^n L_i(x) f(x_i)$$

$$L_i(x) = \prod_{\substack{j=0 \\ j \neq i}}^n \frac{x - x_j}{x_i - x_j}$$

linear interpolation n=1

$$v(t) = L_0(t)v(t_0) + L_1(t)v(t_1)$$

$$L_0(t) = \frac{t - t_1}{t_0 - t_1} \quad L_1(t) = \frac{t - t_0}{t_1 - t_0}$$

quadratic interpolation n=2

$$v(t) = L_0(t)v(t_0) + L_1(t)v(t_1) + L_2(t)v(t_2)$$

$$L_0(t) = \left(\frac{t - t_1}{t_0 - t_1} \right) \left(\frac{t - t_2}{t_0 - t_2} \right)$$

$$L_2(t) = \left(\frac{t - t_0}{t_2 - t_0} \right) \left(\frac{t - t_1}{t_2 - t_1} \right)$$

$$L_1(t) = \left(\frac{t - t_0}{t_1 - t_0} \right) \left(\frac{t - t_2}{t_1 - t_2} \right)$$

- 2nd vs 1st $\rightarrow |E_a| = \left| \frac{2nd - 1st}{2nd} \right| \times 100\%$

Naïve Gaussian Elimination

division by zero

(pitfalls)
larger and often errors

An $n \times n$ or U matrix
forward elimination
is $n-1$ step process

$$A_{11}x_1 + A_{12}x_2 + A_{13}x_3 = C_{11}$$

$$A_{21}x_1 + A_{22}x_2 + A_{23}x_3 = C_{22}$$

$$A_{31}x_1 + A_{32}x_2 + A_{33}x_3 = C_{33}$$

System of linear
equation

$$\Rightarrow \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} C_1 \\ C_2 \\ C_3 \end{bmatrix} \quad [\text{in } [A][x] = [c] \text{ form}]$$

$$\Rightarrow \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ 0 & A'_{22} & A'_{23} \\ 0 & A'_{32} & A'_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} C_1 \\ C'_2 \\ C'_3 \end{bmatrix} \quad \begin{aligned} r'_2 &= r_2 - \frac{A_{21}}{A_{11}} \times r_1 \\ r'_3 &= r_3 - \frac{A_{31}}{A_{11}} \times r_1 \end{aligned} \quad \begin{bmatrix} \text{forward} \\ \text{elimination} \\ \text{step 1} \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ 0 & A'_{22} & A'_{23} \\ 0 & 0 & A''_{33} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} C_1 \\ C'_2 \\ C''_3 \end{bmatrix} \quad \begin{aligned} r''_3 &= r'_3 - \frac{A'_{32}}{A'_{22}} r'_2 \end{aligned} \quad \begin{bmatrix} \text{forward} \\ \text{elimination} \\ \text{step 2} \end{bmatrix}$$

Back substitution: x_3 এর মান দ্বারা x_2 , x_1 এর মান প্রেরণ করা হয়।
Prevents division by zero

partial pivoting: forward elimination এর প্রতীক্ষিত step কোথাও যায়না
সেটা step s র কলুম্ন এর $n-(s-1)$ টি element এর মধ্যবাসন
এর মধ্যে maximum টা \rightarrow you দে ~~বে~~ swap করে দেওয়া লাগে,

$$A = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}_{n \times n} \xrightarrow{\text{naive gaussian elimination}} U = \begin{bmatrix} U_{11} & U_{12} & U_{13} \\ U_{21} & U_{22} & U_{23} \\ U_{31} & U_{32} & U_{33} \end{bmatrix} \quad \begin{aligned} \text{here,} \\ U_{21} &= 0 \\ U_{31} &= 0 \\ U_{32} &= 0 \end{aligned}$$

$$\det(A) = |A| = \prod_{i=1}^n U_{ii} \quad \text{in this case, } \det(A) = U_{11} \times U_{22} \times U_{33}$$

$$\det(A) = \det(U)$$

LU decomposition

$$[A][X] = [C] \quad (\text{for gaussian elimination } \Rightarrow [U])$$

[C] \Rightarrow lower triangular matrix

$$[U] = \begin{bmatrix} U_{11} & U_{12} & U_{13} \\ 0 & U_{22} & U_{23} \\ 0 & 0 & U_{33} \end{bmatrix} \quad [L] = \begin{bmatrix} 1 & 0 & 0 \\ l_{21} & 1 & 0 \\ l_{31} & l_{32} & 1 \end{bmatrix}$$

$$\text{where } l_{21} = \frac{a_{21}}{a_{11}} ; \quad l_{31} = \frac{a_{31}}{a_{11}} ; \quad l_{32} = \frac{a_{32}}{a_{22}}$$

How to solve SLE using LU decomposition

given, $[A][X] = [C]$

① decompose, $[A] = [L][U]$

② solve for $[Z]$ from $[L][Z] = [C]$

③ solve for $[X]$ from $[U][X] = [Z]$

How to find inverse of $A = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{bmatrix}$ using LU decomposition

① decompose, $[A] = [L][U]$

② solve for $[Z]$ $\begin{bmatrix} L \\ Z_1 \\ Z_2 \\ Z_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \rightarrow [U] \begin{bmatrix} b_{11} \\ b_{21} \\ b_{31} \end{bmatrix} = \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix}$ solve for z_1

③ bind column-wise values of b $\begin{bmatrix} L \\ Z_1 \\ Z_2 \\ Z_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix} \rightarrow [U] \begin{bmatrix} b_{12} \\ b_{22} \\ b_{32} \end{bmatrix} = \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix}$ solve for z_2

$U \times B \times U^{-1} = B$ $\begin{bmatrix} L \\ Z_1 \\ Z_2 \\ Z_3 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \rightarrow [U] \begin{bmatrix} b_{13} \\ b_{23} \\ b_{33} \end{bmatrix} = \begin{bmatrix} z_1 \\ z_2 \\ z_3 \end{bmatrix}$ solve for z_3

Gauss seidel method

step-1

given SLE,

$$ax_1 + bx_2 + cx_3 = d \dots (i)$$

$$ex_1 + fx_2 + gx_3 = h \dots (ii)$$

$$ix_1 + jx_2 + kx_3 = l \dots (iii)$$

$$(i) \text{ for } x_1 \text{ eqn, } x_1 = \frac{d - bx_2 - cx_3}{a} \dots (1)$$

$$(ii) \text{ " } x_2 \text{ " " , } x_2 = \frac{h - ex_1 - gx_3}{f} \dots (2)$$

$$(iii) \text{ " } x_3 \text{ " " , } x_3 = \frac{l - ix_1 - jx_2}{k} \dots (3)$$

initial guess $\begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

step-3

~~x_1, x_2, x_3~~ एवं मान $\begin{array}{l} (1) \text{ अपेक्षित} \\ (2) \text{ अपेक्षित} \\ (3) \text{ अपेक्षित} \end{array}$

x_1, x_2, x_3 एवं latest मान $\begin{array}{l} (2) \text{ अपेक्षित} \\ (3) \text{ अपेक्षित} \end{array}$

x_2, x_1 " " " $\begin{array}{l} (3) \text{ अपेक्षित} \end{array}$

$|E_a| \leq |E_s|$ iteration वर्तमान,

step-4 go to step 3

i-th unknown का absolute approximate error,

$$|E_a|_i = \left| \frac{x_i^{\text{new}} - x_i^{\text{old}}}{x_i^{\text{new}}} \right| \times 100$$

$$|a_{ij}| > \sum_{j=1}^n |a_{ij}| \text{ for all } i$$

(2) अन्तिम अंत नाम $(\text{if } i=2)$

$$|a_{22}| > |a_{21}| + |a_{23}|$$

(1) तथा (2) अन्त मान्य A diagonally dominant.

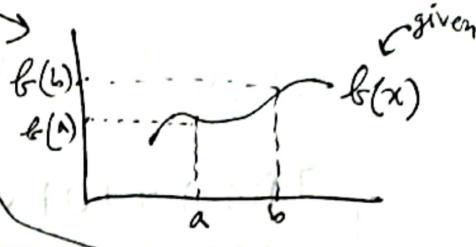
$$\begin{aligned} (1) |a_{11}| &\geq |a_{12}| + |a_{13}| \\ &\& \text{for } 2 \\ &\& \text{for } 3 \\ &\& \text{for } 2 \\ &\& \text{for } 3 \\ &\& \text{for } 2 \\ &\& \text{for } 3 \end{aligned}$$

simpson's one third rule

single segment

$$\rightarrow x = \int_a^b f(x) dx$$

$$\approx \left(\frac{b-a}{6} \right) \left[f(a) + 4f\left(\frac{a+b}{2}\right) + f(b) \right]$$

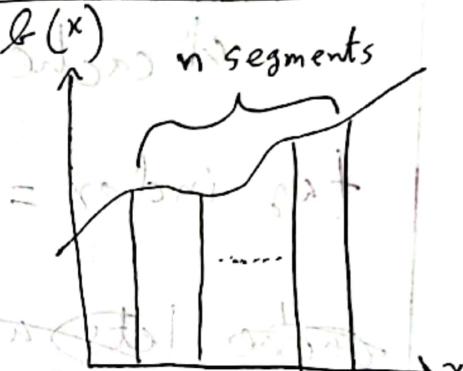


Multiple segment / n-segment

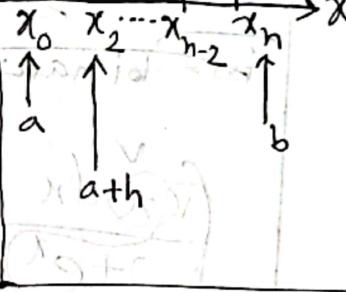
$$h = \frac{b-a}{n}$$

upper limit
lower limit
number of segments

$$x_0 = a ; x_1 = a+h ; x_2 = a+2h$$



$$\int_a^b f(x) dx = \frac{b-a}{3n} \left[f(x_0) + 4 \sum_{i=1, \text{ odd}}^{n-1} f(x_i) + 2 \sum_{i=2, \text{ even}}^{n-2} f(x_i) + f(x_n) \right]$$



Gauss quadrature rule

one-point

two-point

$$\int_a^b f(x) dx \approx c_1 f(x_1) = (b-a) f\left(\frac{b+a}{2}\right)$$

$$\int_a^b f(x) dx \approx c_1 f(x_1) + c_2 f(x_2)$$

$$\frac{b-a}{2} f\left(\frac{b-a}{2} \left(-\frac{1}{\sqrt{3}}\right) + \frac{b+a}{2}\right) + \frac{b-a}{2} f\left(\frac{b-a}{2} \left(\frac{1}{\sqrt{3}}\right) + \frac{b+a}{2}\right)$$

n-point Gauss quadrature rule

এখন n-point এর জন্য nটি c ওর্ডস c_1, c_2, \dots, c_n

এবং nটি x ওর্ডস x_1, x_2, \dots, x_n এর জন্য একটি ফাংশন,

given $\int_a^b f(x) dx$ or $\int_{-1}^1 g(t) dt$ convert করুন

বিলু করুন।

$$\int_a^b f(x) dx = \frac{b-a}{2} \int_{-1}^1 f\left(\frac{b-a}{2}t + \frac{b+a}{2}\right) dt$$

$$\approx \frac{b-a}{2} c_1 g(x_1) + \frac{b-a}{2} c_2 g(x_2) + \dots + \frac{b-a}{2} c_n g(x_n)$$

এখন c_1, c_2, \dots, c_n ও x_1, x_2, \dots, x_n এর জন্য বের করুন

ans. বের করুন।

Trapezoidal rule

multiple segment

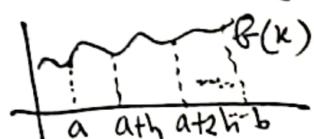
$$\int_a^b f(x) dx = (b-a) \left[\frac{f(a) + f(b)}{2} \right]$$

single segment

$$\int_a^b f(x) dx = \frac{b-a}{2n} \left[f(a) + 2 \left\{ \sum_{i=1}^{n-1} f(a+ih) \right\} + f(b) \right]$$

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

number of segments



Trapezoidal

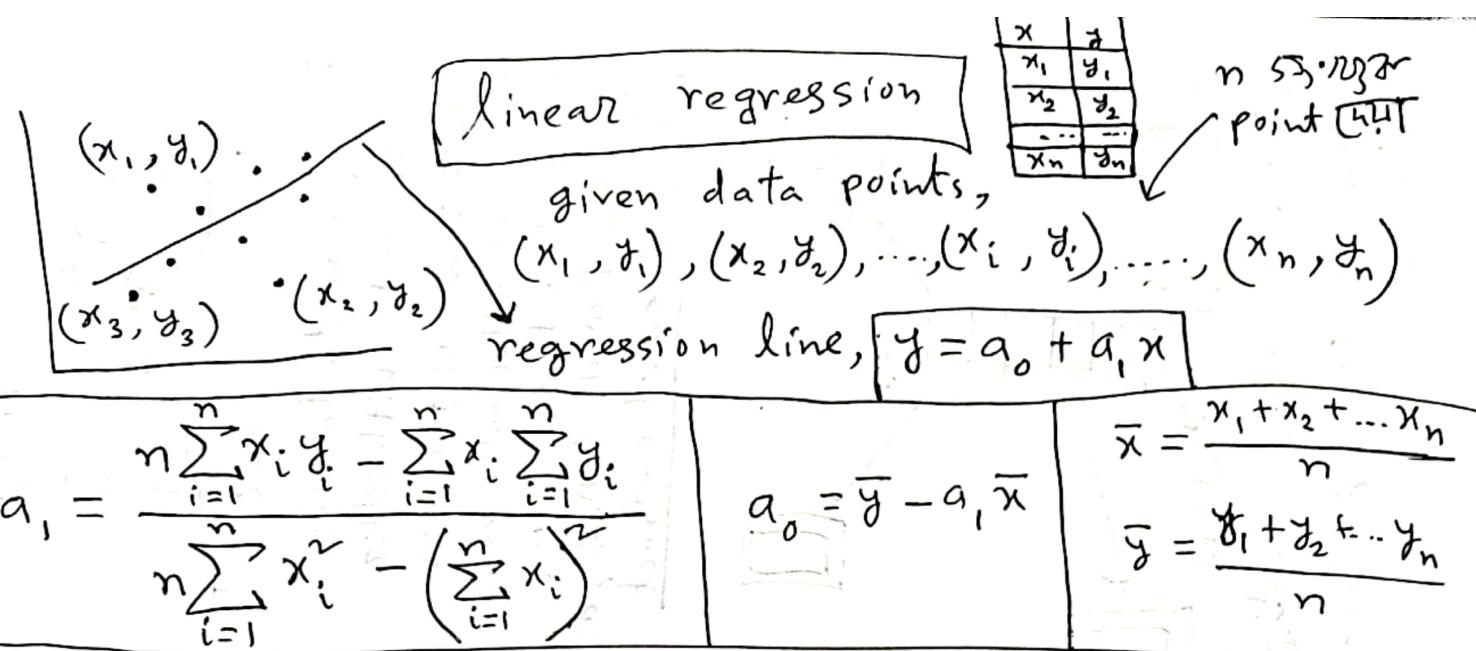
$$\int_a^b f(x) dx \approx (b-a) \left[\frac{f(a) + f(b)}{2} \right]$$

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

$$\int_a^b f(x) dx \approx \frac{b-a}{2n} \left[f(a) + 2 \sum_{i=1}^{n-1} f(a+ih) + f(b) \right]$$

$$\frac{b-a}{2n} \left[f(a) + 2 \sum_{i=1}^{n-1} f(a+ih) + f(b) \right]$$

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



$$a_1 = \frac{n \sum_{i=1}^n x_i y_i - \sum_{i=1}^n x_i \sum_{i=1}^n y_i}{n \sum_{i=1}^n x_i^2 - (\sum_{i=1}^n x_i)^2}$$

$$a_0 = \bar{y} - a_1 \bar{x}$$

$$\bar{x} = \frac{x_1 + x_2 + \dots + x_n}{n}$$

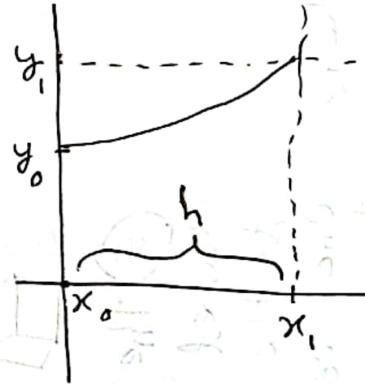
$$\bar{y} = \frac{y_1 + y_2 + \dots + y_n}{n}$$

Euler's Method

step size = h given
given,

$$\frac{dy}{dx} = f(x, y), \quad y(0) = y_0$$

differential equation initial condition



y_0 पहली बार, y_1 दूसरी बार y_2 तीसरी बार तक इसी रिटर्न,

$$y_{i+1} = y_i + f(x_i, y_i)h$$

$$y_1 = y_0 + f(x_0, y_0)h$$

$$y_2 = y_1 + f(x_1, y_1)h$$

$$y_3 = y_2 + f(x_2, y_2)h$$

