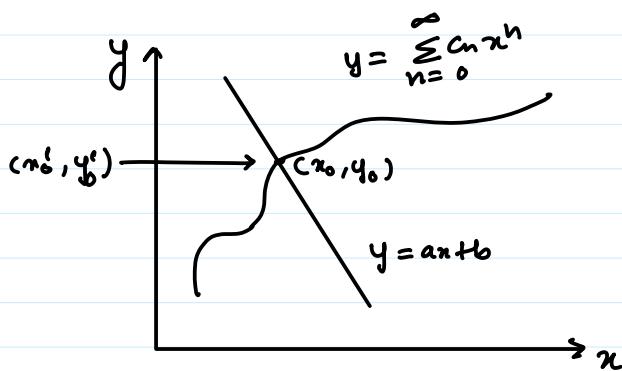


Numerical error:-

$$\epsilon_x = \frac{|x_0' - x_0|}{x_0}$$

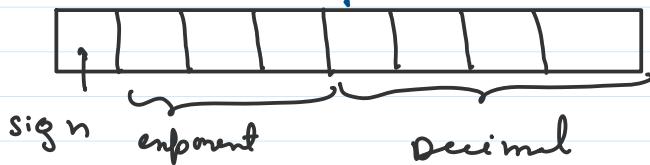
$$\epsilon_y = \frac{|y_0' - y_0|}{y_0}$$



Truncation or round off error :- for non-integer numbers.

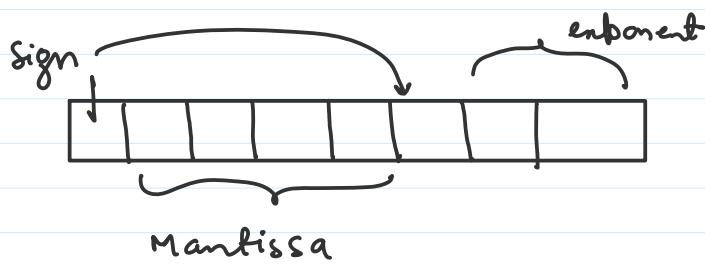
→ 8-bit system:-

Hypothetical computer in base of 10 :-



$$999.9999 \rightarrow 10^3$$

$$000.0001 \rightarrow 10^{-4}$$



$$\begin{aligned}
 80.9 &= 0.809 \times 10^2 && \text{Exponent} \\
 0.33 &= 0.33 \times 10^0 && \\
 -0.000019 &= -0.19 \times 10^{-4} &&
 \end{aligned}
 \rightarrow \begin{aligned}
 &+ 8090 + 02 \\
 &+ 3300 + 00 \\
 &- 1900 - 04
 \end{aligned}$$

Highest no. we can stored = 9999 E+99

Lowest " " " " " = 1000 E-99

→ Mathematical Rules:-

1) Addition :- exponents has to be made equal

$$\begin{array}{rcl}
 0.3780 \times 10^{16} & \rightarrow & \text{keep the larger one unchanged} \\
 0.1953 \times 10^{14} & \xrightarrow{\substack{\text{last digit} \\ \downarrow}} & 0.3780 \times 10^{16} \\
 & \xrightarrow{\substack{\text{last digit} \\ \downarrow}} & 0.0019 \times 10^{16}
 \end{array}$$

$$\begin{array}{r}
 0.1953 \times 10^{-14} \\
 \xrightarrow{\text{lost digit}} \\
 \text{which is truncated}
 \end{array}
 \quad
 \begin{array}{r}
 0.5780 \times 10^{-16} \\
 0.0019 \times 10^{-16} \\
 \hline
 0.3799 \times 10^{-16}
 \end{array}$$

$$\begin{array}{r}
 0.4546 \times 10^3 \\
 0.5433 \times 10^7
 \end{array}
 \rightarrow
 \begin{array}{r}
 0.0000 \times 10^3 \\
 0.5433 \times 10^7 \\
 \hline
 0.5433 \times 10^7
 \end{array}$$

$$0.4546 \times 10^3 = 454.6$$

$$\begin{array}{r}
 0.5433 \times 10^7 \\
 + 5433000.6 \\
 \hline
 5433454.6 \rightarrow 0.5433 \times 10^7
 \end{array}$$

$$\begin{array}{r}
 0.6389 \times 10^9 \\
 0.4117 \times 10^9 \\
 \hline
 1.0506 \times 10^9
 \end{array}
 \rightarrow$$

0.1050×10^{10} → out of range a.k.a overflow error

$$\begin{array}{r}
 0.1512 \times 10^{-12} \\
 - 0.3573 \times 10^{-13} \\
 \hline
 0.1155 \times 10^{-12}
 \end{array}
 \rightarrow$$

$$\begin{array}{r}
 0.5452 \times 10^{-31} \\
 0.5424 \times 10^{-31} \\
 \hline
 0.0028 \times 10^{-31}
 \end{array}
 \rightarrow$$

$$\begin{array}{r}
 0.5452 \times 10^{-99} \\
 0.5424 \times 10^{-99} \\
 \hline
 0.0028 \times 10^{-99}
 \end{array}
 \rightarrow$$

0.2800×10^{-101} → underflow error

$$0.0028 \times 10^{-99} \rightarrow 0.2800 \times 10^{-101} \rightarrow \boxed{\text{underflow error}}$$

2) Multiplication:- Multiply mantissa, add exponents

$$0.7500 \times 10^{-32}$$

$$0.9248 \times 10^{-33}$$

$$\overline{0.6936 \times 10^{-65}}$$

$$0.1111 \times 10$$

$$\overline{0.1231 \times 10^5}$$

$$\overline{0.1367 \times 10^{25}} \rightarrow 0.1367 \times 10^{24}$$

3) Division :- Divide the mantissa, subtract the exponent of Denominator

$$0.1111 \times 10$$

$$\div 0.1231 \times 10^3$$

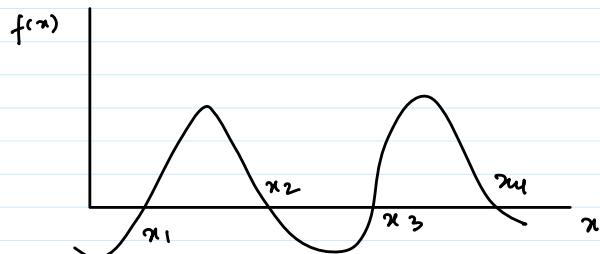
$$\overline{0.9025 \times 10^7}$$

→ Log table :-

$$\begin{aligned} 27.262 &\rightarrow 2.7262 \times 10^1 \rightarrow \frac{\text{Characteristics}}{1} \\ 5530 &\rightarrow 5.53 \times 10^2 \rightarrow 2 \\ 0.0058939 &\rightarrow 5.8939 \times 10^{-3} \rightarrow -3 \end{aligned}$$

	Characteristic	Mantissa	$\log_{10}(x)$
2726	1	$2726 \Rightarrow 4846 + 9 = 4855$	1.4355
5530	2	$5530 \Rightarrow 7427$	2.7427
5894	-3	$5894 \Rightarrow 7701 + 3 = 7704$	$-3 + 0.7704 = -2.2296$

→ Root finding :-



$$f(x_1) = f(x_2) = f(x_3) = f(x_4) = 0$$

→ User defined error limit

$$d^i < \epsilon, \text{ then stop} \quad d^i = \text{width of the } i^{\text{th}} \text{ iteration}$$

internal d^i after i^{th} iteration.

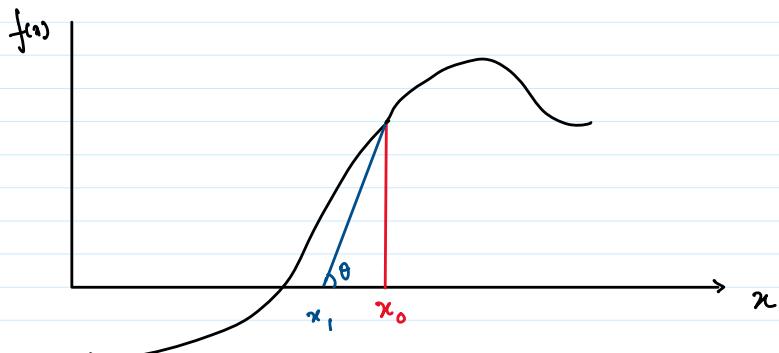
Q. Try the bisection method for different values of ϵ -

$$\epsilon \equiv \ell = 0.1 \quad 0.01 \quad 0.001 \quad 0.0001 \quad \dots$$

$$\text{no. of iteration, } N = N_1, \quad N_2, \quad N_3, \quad N_4, \quad \dots$$



→ Newton - Raphson method :-



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

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

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

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

→ Discussion on Newton-Raphson :-

1) Issue with convergence

2) Error propagation :-

Let us assume that the root is found in $(i+1)$ iteration

$$f(x_{i+1}) = 0 = f(x_i + h)$$

$$\therefore x_{i+1} = x_i + h \quad (h \rightarrow \text{increment in } x)$$

$$\text{Now, } f(x_i + h) = f(x_i) + hf'(x_i) + \frac{h^2}{2!} f''(x_i) + \frac{h^3}{3!} f'''(x_i) + \dots$$

$$0 \approx f(x_i) + hf'(x_i)$$

$$\Rightarrow 0 \approx f(x_i) + (x_{i+1} - x_i) f'(x_i)$$

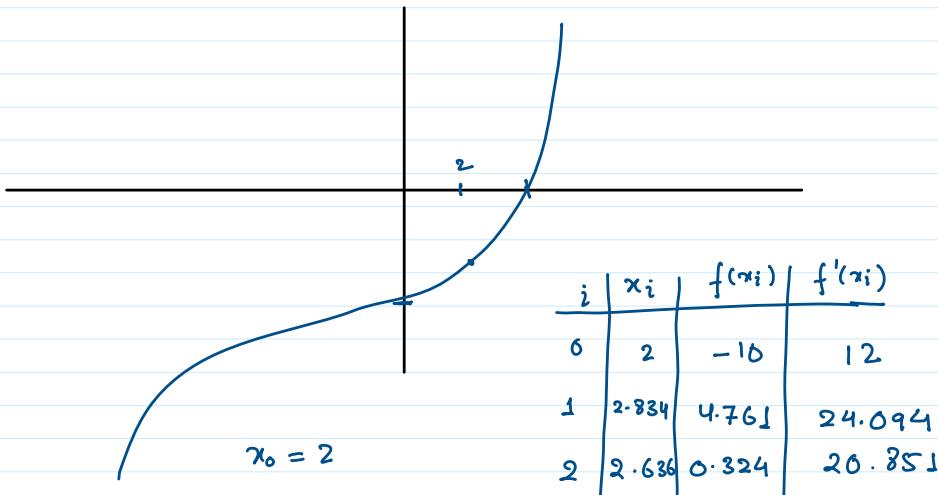
$$\Rightarrow x_{i+1} - x_i = -\frac{f(x_i)}{f'(x_i)}$$

$$\therefore x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)} \Rightarrow 1^{\text{st}} \text{ order convergence}$$

2) Presence of differential.

Ex:- $\sqrt[3]{18}$ using Newton-Raphson.

$$f(x) : x^3 - 18 = 0$$



$x_0 = 2$	1	2.834	4.761	24.094
$f'(x) = 3x^2$	2	2.636	0.324	20.851
$x_{i+1} = x_i - \frac{f(x_i)}{f'(x_i)}$	3	2.620	-0.005	20.600

$$x_1 = 2 + \frac{10}{12} = \frac{34}{12} = \frac{17}{6} = 2.834$$

$$\Rightarrow x_2 = 2.834 - \frac{4.761}{24.094} = 2.636$$

$$x_3 = 2.636 - \frac{0.324}{20.851} = 2.620$$

$$\sqrt[3]{18} = 2.620$$

now, for $x_0 = 10$;

i	x_i	$f(x_i)$	$f'(x_i)$
0	10	982	300
1	6.726	286.368	135.744
2	4.616	86.379	63.933
3	3.358	19.893	33.845

$$x_1 = 10 - \frac{982}{300} = 6.726$$

$$x_2 = 6.726 - \frac{286.368}{135.744} = 4.616$$

$$x_3 = 4.616 - \frac{86.379}{63.933} = 3.358$$

→ Gauss - Seidel method :-

system of linear equations

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = c_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = c_2$$

$$\vdots \quad \vdots \quad \ddots \quad \vdots \quad \vdots$$

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = c_n$$

→ Write in matrix form :-

$$a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn}x_n = b_n$$

→ Write in matrix form:-

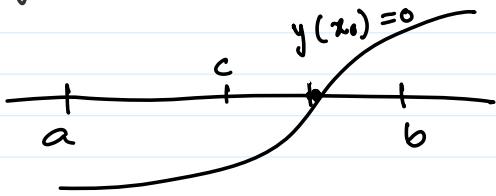
$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix} = \begin{pmatrix} c_1 \\ c_2 \\ \vdots \\ c_n \end{pmatrix}$$

$$AX = C$$

$$\Rightarrow \boxed{X = A^{-1}C}$$

$A \rightarrow$ has to be diagonally dominant

→ Convergence of bisection method:-



$$0^{\text{th}} \quad \epsilon_{\max}^0 \approx |b-a| = \Delta$$

$$1^{\text{st}} \quad \epsilon_{\max}^1 \approx |c-b| = \frac{\Delta}{2}$$

:

:

:

:

$$n^{\text{th}} \quad \epsilon_{\max}^n \approx \dots = \frac{\Delta}{2^n}$$

→ Bisection method (the series always converges).

→ General formula for error propagation:-

$$\epsilon^{k+1} = C(\epsilon^k)^m$$

↑ order of convergence
↓ constant

for method of bisection:-

$$\epsilon_b^{k+1} = \frac{\epsilon_b^k}{2} \quad m=1$$

of error propagation

so, the order of convergence is 1.

→ for a user defined error ϵ_{user} :-

the iteration terminate at

$$\frac{\Delta}{2^n} \leq \epsilon_{\text{user}}$$

Ex:

$$\Delta = 0.8, \quad \epsilon = 0.00001$$

the iteration terminate at

$$\frac{\Delta}{2^n} = \epsilon \Rightarrow n \log 2 = \log \Delta - \log \epsilon$$

$$n = \frac{\log(\frac{\Delta}{\epsilon})}{\log(2)} = 16.28$$

$$\Rightarrow (n)_{\text{actual}} = \text{Round}(n) + 1$$

→ Order of Convergence for N-R:-

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

$$\text{Actual root} = x_0$$

$$\text{Error in } (i+1)^{\text{th}} \text{ iteration} = \epsilon_{i+1}$$

$$\text{" " } i^{\text{th}} \text{ " } = \epsilon_i$$

$$\Rightarrow (x_0 + \epsilon_{i+1}) = x_0 + \epsilon_i - \frac{f(x_0 + \epsilon_i)}{f'(x_0 + \epsilon_i)}$$

Taylor expansion of the function \Rightarrow

$$\epsilon_{i+1} = \epsilon_i - \frac{f(x_0) + \epsilon_i f'(x_0) + \frac{\epsilon_i^2}{2} f''(x_0)}{f'(x_0) + \epsilon_i f''(x_0) + \cancel{\mathcal{O}(\epsilon_i^2)}}$$

$$= \epsilon_i - \epsilon_i \left[1 + \frac{\epsilon_i}{2} \frac{f''(x_0)}{f'(x_0)} \right]$$

$$\left[1 + \epsilon_i \frac{f''(x_0)}{f'(x_0)} \right]$$

$$= \epsilon_i - \epsilon_i \left[\left(1 + \frac{\epsilon_i}{2} \frac{f''(x_0)}{f'(x_0)} \right) \left(1 - \epsilon_i \frac{f''(x_0)}{f'(x_0)} \right) \right]$$

$$\Rightarrow \epsilon_{i+1} = \epsilon_i - \epsilon_i \left[1 + \frac{\epsilon_i}{2} \frac{f''(x_0)}{f'(x_0)} - \epsilon_i \frac{f''(x_0)}{f'(x_0)} - \frac{\epsilon_i^2}{2} \left\{ \frac{f''(x_0)}{f'(x_0)} \right\}^2 \right]$$

$$\Rightarrow \epsilon_{i+1} = \epsilon_i - \epsilon_i \left[1 + \frac{\epsilon_i}{2} \frac{f'(x_0)}{f''(x_0)} - \epsilon_i \frac{f''(x_0)}{f'(x_0)} - \frac{\epsilon_i^2}{2} \left\{ \frac{f''(x_0)}{f'(x_0)} \right\} \right]$$

$$= \epsilon_i - \epsilon_i - \frac{\epsilon_i^2}{2} \frac{f''(x_0)}{f'(x_0)} + \epsilon_i^2 \frac{f''(x_0)}{f'(x_0)} - \frac{\epsilon_i^3}{2} ()$$

$$\epsilon_{i+1} \approx \epsilon_i^2 \left[\frac{1}{2} \frac{f''(x_0)}{f'(x_0)} \right]$$

$\Rightarrow m=2$
 \Rightarrow 2nd order convergence.

→ Solution of simultaneous linear equations :-

$$A_{n \times n} \bar{X}_{n \times 1} = \bar{b}_{n \times 1}$$

let us take a system of two linear eqn:-

$$x_{10} \quad | \quad x_1 + x_2 = 2$$

$$\begin{array}{r} 3x_1 - 10x_2 = 3 \\ \hline 13x_1 = 23 \\ x_1 = 23/13 \end{array} \quad | \quad 10x_2 = 3 - 3 \cdot \frac{23}{13} \\ = \frac{39 - 69}{13} \\ = -\frac{30}{13}$$

$$\Rightarrow x_1 = \frac{23}{13} = 1.769$$

$$x_2 = \frac{3}{13} = 0.231$$

Now,

$$x_1 + x_2 = 2 \Rightarrow x_1 = 2 - x_2 \quad -\textcircled{A}$$

$$3x_1 - 10x_2 = 3 \Rightarrow x_2 = \frac{-3 + 3x_1}{10} \quad -\textcircled{B}$$

Iteration	x_1	x_2	$ \epsilon_{x_1} $	$ \epsilon_{x_2} $
0	0	$0 \xrightarrow{+\textcircled{A}}$	1.769	0.231
1	2	$3/10 \xrightarrow{+\textcircled{A}}$		

$$2 \quad \frac{17}{10} \quad \frac{21}{100}$$

$$3. \quad \frac{179}{100} \quad \frac{237}{1000}$$

$$4. \quad \frac{1763}{1000} \quad \frac{2289}{10000}$$

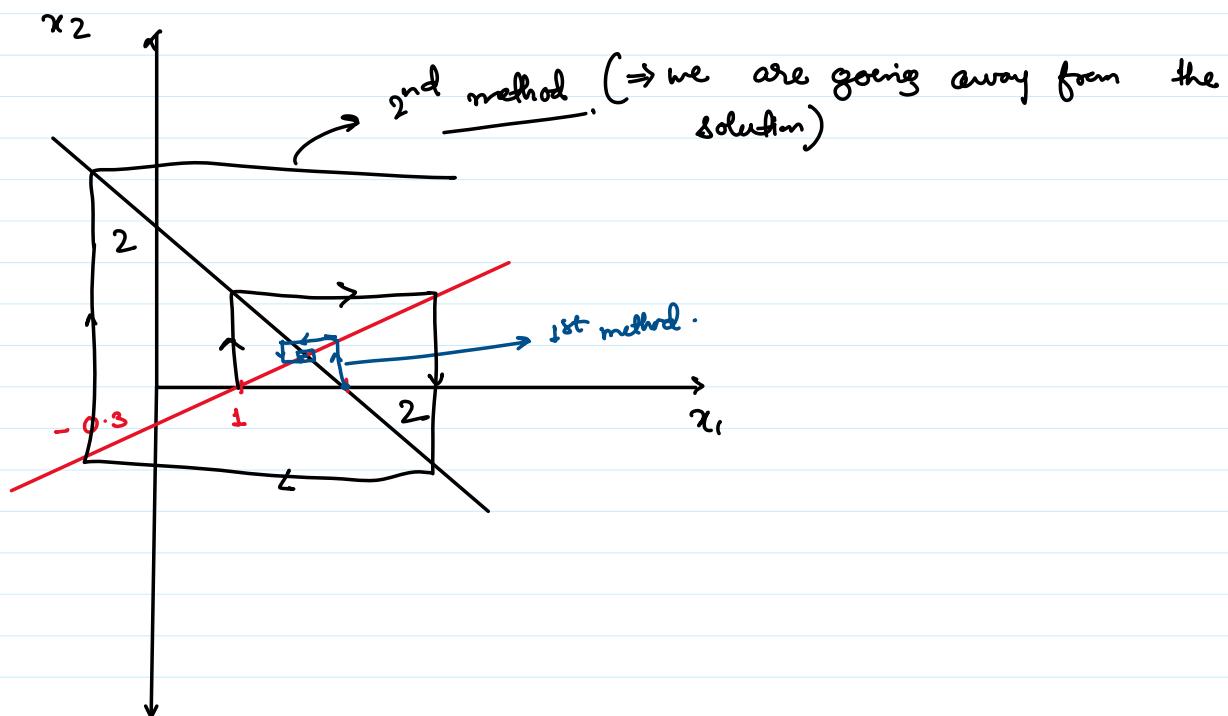
Now, (Switching the order)

$$3x_1 - 10x_2 = 3 \Rightarrow x_1 = \frac{3 + 10x_2}{3} \quad \text{--- (A)}$$

$$x_1 + x_2 = 2 \Rightarrow x_2 = 2 - x_1 \quad \text{--- (B)}$$

Method - ②

Iteration	x_1	x_2	
0	0	0	
1	1	1	
2	$\frac{13}{3}$	$-\frac{7}{3}$	→ Won't come to the solution
3	$-\frac{61}{9}$	$\frac{79}{9}$	
4.	$\frac{817}{27}$	$-\frac{763}{27}$	



$$\begin{pmatrix} 3 & -10 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 3 \\ 2 \end{pmatrix} \rightarrow \text{Method 2}$$

A_1 \bar{x} \bar{b}

$$\begin{pmatrix} 1 & 1 \\ 3 & -10 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 2 \\ 3 \end{pmatrix} \rightarrow \text{Method 1}$$

A_2 \bar{x} \bar{b}

for diagonal dominance:-

$$|A_{ii}| > \sum_{\substack{i=1 \\ i \neq j}}^n |A_{ij}| \quad (\text{if this criteria is satisfied then convergence is guaranteed})$$

for, $A = \begin{pmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ a_{21} & a_{22} & a_{23} & \dots & a_{2n} \\ \vdots & \vdots & \vdots & & \vdots \\ \vdots & \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & a_{nn} \end{pmatrix}_{n \times n}$

$$A_{n \times n} \bar{x}_{n \times 1} = \bar{b}_{n \times 1}$$

Ex:- $9x_1 + 2x_2 + 4x_3 = 20$

$$x_1 + 10x_2 + 4x_3 = 6$$

$$2x_1 - 4x_2 + 10x_3 = -15$$

$$\begin{pmatrix} 9 & 2 & 4 \\ 1 & 10 & 4 \\ 2 & -4 & 10 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} 20 \\ 6 \\ -15 \end{pmatrix}$$

↓

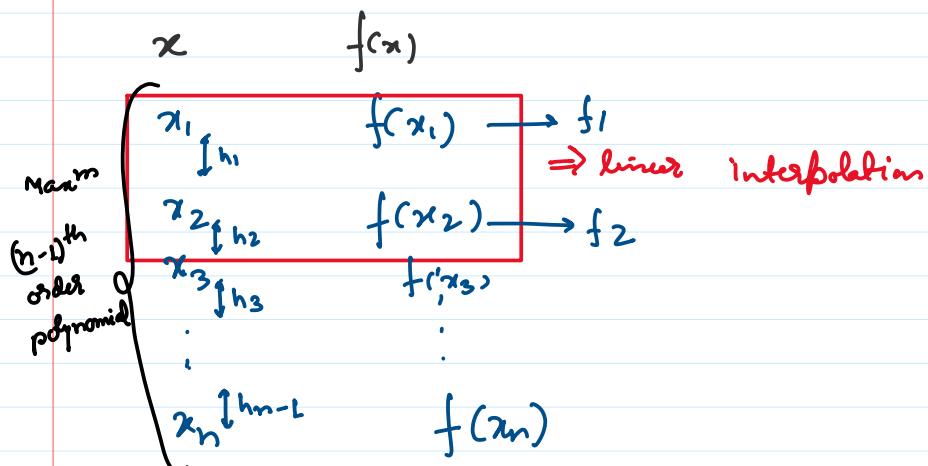
diagonally dominant \Rightarrow convergence.

$$x_1 = \frac{20 - 2x_2 - 4x_3}{9}$$

$$x_2 = \frac{6 - x_1 - 4x_3}{10}$$

$$x_3 = \frac{-15 - 2x_1 + 4x_2}{10}$$

Iteration	x_1	x_2	x_3
0	0	0	0
1	2.7367	0.9872	-1.6525



- $f(x) = a + bx \quad (\text{linear})$

$$f(x) = c_1(x - x_2) + c_2(x - x_1)$$

$$\Rightarrow c_1 = \frac{f_1}{(x_1 - x_2)}, \quad c_2 = \frac{f_2}{(x_2 - x_1)}$$

$$\Rightarrow f(x) = \frac{f_1(x - x_2)}{(x_1 - x_2)} + \frac{f_2(x - x_1)}{(x_2 - x_1)} \quad \rightarrow \text{Lagrange representation.}$$

- Quadratic: for three points x_1, x_2, x_3

$$f(x) = a + bx + cx^2$$

$$f(x) = c_1(x - x_2)(x - x_3) + c_2(x - x_1)(x - x_3) + c_3(x - x_1)(x - x_2)$$

$$c_1 = \frac{f_1}{(x_1 - x_2)(x_1 - x_3)} ; \quad c_2 = \frac{f_2}{(x_2 - x_1)(x_2 - x_3)} ;$$

$$c_3 = \frac{f_3}{(x_3 - x_1)(x_3 - x_2)}$$

$$c_3 = \frac{t_3}{(x_3 - x_1)(x_3 - x_2)}$$

$$\Rightarrow f(x) = \frac{f_1(x-x_2)(x-x_3)}{(x_1-x_2)(x_1-x_3)} + \frac{f_2(x-x_1)(x-x_3)}{(x_2-x_1)(x_2-x_3)} + \frac{f_3(x-x_1)(x-x_2)}{(x_3-x_1)(x_3-x_2)}$$

Generalization :-

$$f(x) = \sum_{i=1}^n \prod_{\substack{j=1 \\ i \neq j}}^N \frac{f_i(x-x_j)}{(x_i-x_j)}$$

$$= \sum_{i=1}^n f_i \prod_{j \neq i} \frac{(x-x_j)}{(x_i-x_j)}$$



→ Differentiation and Integration :-

$$\frac{df}{dx} = \lim_{h \rightarrow 0} \frac{f_{n+1} - f_n}{h_n}$$

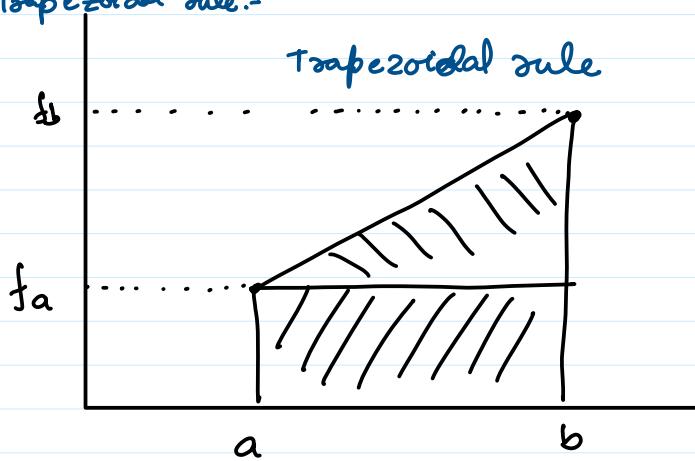
$$\left. \frac{df}{dx} \right|_{in n^{th} place} \approx \frac{f_{n+1} - f_n}{h_n}$$

$$\approx \frac{f_n - f_{n-1}}{h_{n-1}}$$

$$\approx \frac{h_{n-1}}{2} \cdot f_1 + f_2 + f_3 + \dots + f_{n-1}$$

→ Integration :-

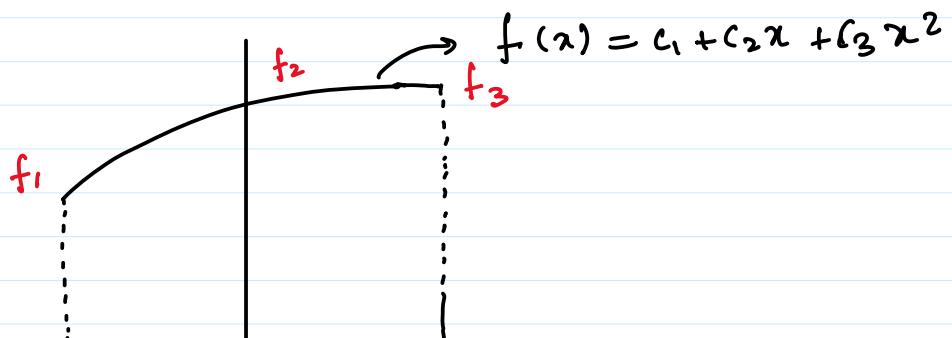
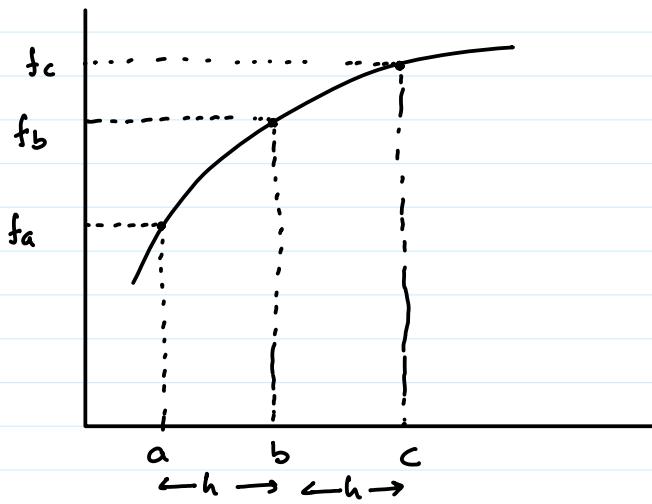
1. Trapezoidal rule :-

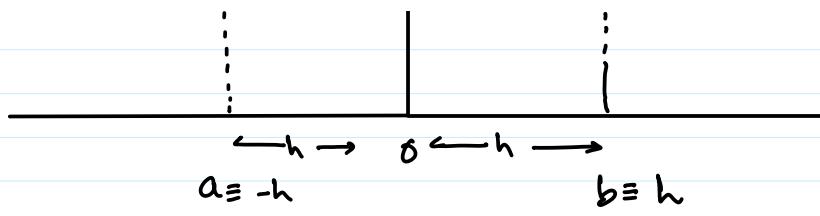


$$\text{Area } A = f_a(b-a) + \frac{1}{2}(b-a)(f_b - f_a)$$

$$= \frac{(b-a)}{2} (f_b + f_a) = \frac{h}{2} (f_b + f_a)$$

2. Simpson's $\frac{1}{3}$ rd rule





$$\begin{aligned} \int_{-h}^h f(x) dx &= \int_{-h}^h (c_1 + c_2 x + c_3 x^2) dx \\ &= c_1 x \Big|_{-h}^h + c_2 \frac{x^2}{2} \Big|_{-h}^h + c_3 \frac{x^3}{3} \Big|_{-h}^h \\ &= 2c_1 h + \frac{2c_3}{3} h^3 \end{aligned}$$

$$\text{Now, } f(-h) = c_1 - c_2 h + c_3 h^2 = f_1$$

$$f(0) = c_1 = f_2 \Rightarrow c_1 = f_2$$

$$f(h) = c_1 + c_2 h + c_3 h^2 = f_3$$

$$3c_1 + 2c_3h^2 = f_1 + f_2 + f_3$$

三

$$c_3 = \frac{f_1 - 2f_2 + f_3}{2h^2}$$

$$c_2 = \frac{f_3 - f_1}{2h}$$

$$\Rightarrow \int_{-h}^h f(x) dx = 2C_1 h + \frac{2C_3}{3} h^3$$

$$= 2f_2 h + \frac{f_1 + f_3 - 2f_2}{3} h^5$$

$$= \frac{h}{3} \left[6f_2 + f_1 + f_3 - 2f_2 \right]$$

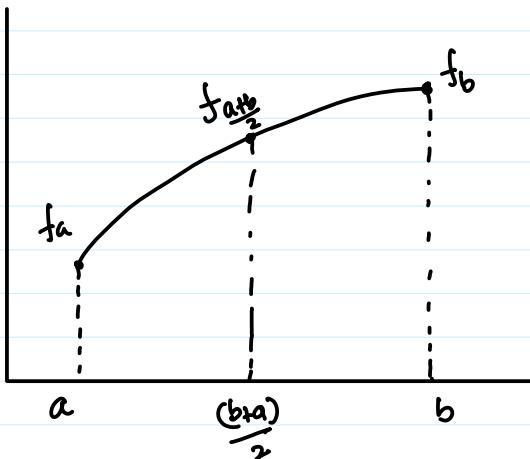
$$\int_{-h}^h f(x) dx = \frac{h}{3} [f_1 + 4f_2 + f_3]$$

→ Simpson's formula.

3. Midpoint rule

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

→ Numerical Integration summary :-



$$T \approx \left(\frac{b-a}{2}\right)(f_b + f_a)$$

$$M \approx (b-a) f_{\frac{a+b}{2}}$$

$$S \approx \frac{(b-a)}{6} \left(f_a + 4f_{\frac{a+b}{2}} + f_b \right)$$

$$\Rightarrow S = \frac{b-a}{6} \left[\frac{2T}{b-a} + 4 \frac{M}{b-a} \right]$$

$$= \left(\frac{T}{3} + \frac{2M}{3} \right)$$

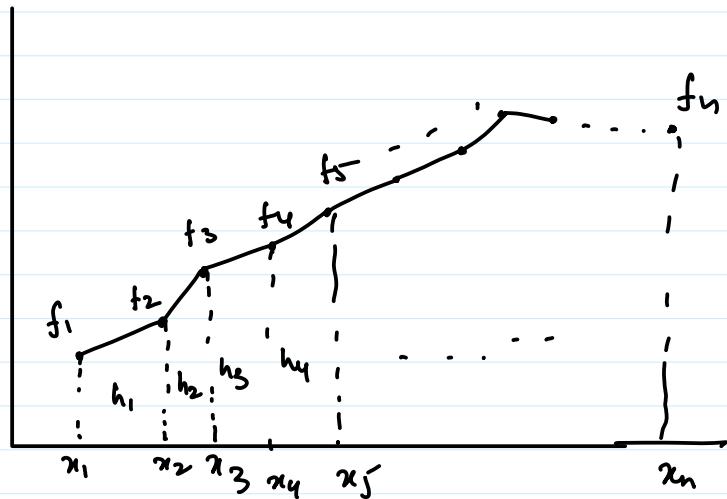
∴

$$S = \frac{2M+T}{3}$$

→ Numerical Integration :-

Composite formulae

x	$f(x)$
x_1	f_1
x_2	f_2
x_3	f_3
\vdots	\vdots
\vdots	\vdots



$$\rightarrow I \approx T = \frac{h_1}{2} (f_1 + f_2) + \frac{h_2}{2} (f_2 + f_3) + \frac{h_3}{2} (f_3 + f_4) + \dots$$

in special case when $h_1 = h_2 = h_3 = \dots = h$

$$\Rightarrow T = \frac{h}{2} [f_1 + f_2 + f_3 + f_4 + \dots + f_{n-1} + f_n]$$

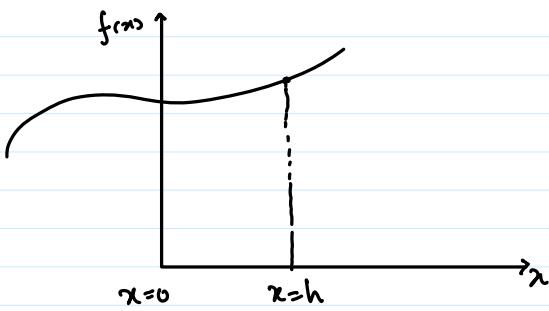
$$T = \frac{h}{2} [f_1 + 2f_2 + 2f_3 + \dots + 2f_{n-1} + f_n]$$

$$\rightarrow I \approx S = \frac{h}{3} [(f_1 + 4f_2 + f_3) + (f_3 + 4f_4 + f_5) + (f_5 + 4f_6 + f_7) + \dots]$$

$$S = \frac{h}{3} [f_1 + 4f_2 + 2f_3 + 4f_4 + 2f_5 + 4f_6 + 2f_7 + \dots + 4f_{n-1} + f_n]$$

$$\rightarrow I \approx M = 2h [f_2 + f_4 + f_6 + \dots + f_{n-1}]$$

→ Error in numerical integrations:-



$$\text{Actual : } E = \int_0^h f(x) dx$$

$$I \approx T = \frac{h}{2} (f(0) + f(h))$$

In order to calculate the error $e_T = |E - T|$ we need to expand $f(x)$ in Taylor expansion.

$$f(x) = f(0) + x f'(0) + \frac{x^2}{2!} f''(0) + \frac{x^3}{3!} f'''(0) + \dots$$

$$\Rightarrow E(x) = h f(0) + \frac{h^2}{2} f'(0) + \frac{h^3}{6} f''(0) + \frac{h^4}{24} f'''(0) + \frac{h^5}{120} f^{(v)}(0)$$

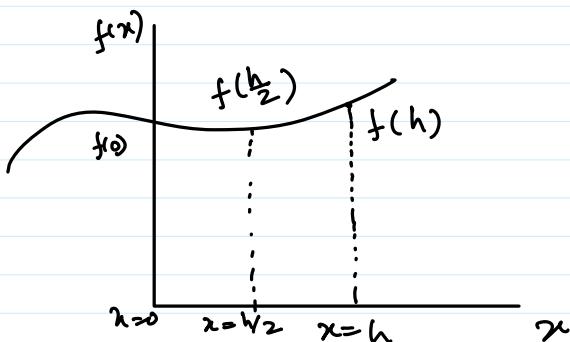
$$\begin{aligned} \text{and } T &= \frac{h}{2} [f(0) + f(0) + h f'(0) + \frac{h^2}{2!} f''(0) + \frac{h^3}{3!} f'''(0) + \dots] \\ &= h f(0) + \frac{h^2}{2} f'(0) + \frac{h^3}{4} f''(0) + \frac{h^4}{12} f'''(0) + \frac{h^5}{48} f^{(v)}(0) \end{aligned}$$

$$\begin{aligned} \Rightarrow e_T = |E - T| &= h^3 \left(\frac{1}{4} - \frac{1}{6} \right) f''(0) + h^4 \left(\frac{1}{12} - \frac{1}{24} \right) f'''(0) \\ &\quad + h^5 \left(\frac{1}{48} - \frac{1}{120} \right) f^{(v)}(0) \end{aligned}$$

∴

$$e_T = \frac{h^3}{12} f''(0) + \frac{h^4}{24} f'''(0) + \frac{h^5}{80} f^{(v)}(0)$$

- Error for mid-point rule:-



$$I \approx M = h f\left(\frac{h}{2}\right)$$

$$= h \left[f(0) + \frac{h}{2} f'(0) + \frac{h^2}{8} f''(0) + \frac{h^3}{48} f'''(0) + \frac{h^4}{384} f^{IV}(0) \right]$$

$$\Rightarrow \varepsilon_M = E - M$$

$$= h^3 \left(\frac{1}{6} - \frac{1}{8} \right) f''(0) + h^4 \left(\frac{1}{24} - \frac{1}{48} \right) f'''(0) \\ + h^5 \left(\frac{1}{120} - \frac{1}{384} \right) f^{IV}(0)$$

$$\boxed{\varepsilon_M = \frac{h^3}{24} f''(0) + \frac{h^4}{24} f'''(0) + \frac{11 \times h^5}{1920} f^{IV}(0)}$$

• For Simpson's rule :-

$$\varepsilon_S = \frac{2\varepsilon_M + \varepsilon_T}{3} \quad (\because S = \frac{2M+T}{3})$$

$$= \frac{1}{3} \left[h^3 f''(0) \left(\frac{1}{12} - \frac{1}{12} \right) + h^4 f'''(0) \cancel{\left(\frac{1}{24} - \frac{1}{24} \right)} + \right. \\ \left. + h^5 f^{IV}(0) \left(\frac{11}{960} - \frac{1}{80} \right) \dots \right]$$

$$= -\frac{1}{3} h^5 \frac{1}{960} f^{IV}(0)$$

$$\Rightarrow \varepsilon_S = \frac{-1}{2880} h^5 f^{IV}(0)$$

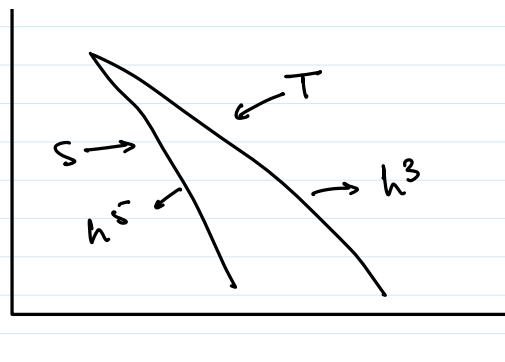
for $h = 2H$:-

$$\boxed{\varepsilon_S = -\frac{h^5}{90} f^{IV}(0)}$$

Err



$$f(x) = 6x^5 \quad I = 1$$



$$f(x) = 6x^5 \quad I = 1$$

$$\varepsilon_T = \left| \frac{T-1}{1} \right|$$

$$\varepsilon_s = \left| \frac{s-1}{1} \right|$$

→ Gauss quadrature :- (Gauss - Legendre formula)

Let say we have the integral :- $\int_{-1}^1 f(x) dx$

so, $\int_{-1}^1 f(x) dx = \sum_{i=1}^n w_i f(x_i)$

\downarrow
weight

↳ nodes (nodes of n^{th} order
Legendre polynomial)

$$w_i = \frac{2}{(1-x_i^2) [P_n'(x_i)]^2}$$

- This formula is exact upto $(2n-1)^{th}$ order.

$$\int_a^b f(x) dx = \int_{-1}^1 f(y) dy$$

$$y = px + q$$

$$= \frac{b-a}{2} \int_{-1}^1 f\left(\frac{b-a}{2}y + \frac{b+a}{2}\right) dy$$

no. of points	Points (x_i)	w_i
1	0	2
2	$\pm \sqrt{1/3}$	1
3	$0, \pm \sqrt{3/5}$	$8/9, 5/9$

4

$$\pm 0.339981$$

$$0.652145$$

$$\pm 0.861136$$

$$0.347853$$

Ans:- $I = \int_{-5}^9 (7x^3 - 9y) dx = 10,136$

$$b=9, a=-5$$

$$\frac{b-a}{2} = 2, \frac{b+a}{2} = 7$$

$$\Rightarrow I = 2 \int_{-1}^1 f(2y+7) dy$$

$$f(2y+7) = 7(2y+7)^3 - 9(2y+7)$$

$$= 2 \int_{-1}^1 (56y^3 + 588y^2 + 2040y + 2338) dy$$

$$= 7(84^3 + 343 + 3 \cdot 7 \cdot 4y^2 + 3 \cdot 72 \cdot 2y) - 18y - 63$$

$$I(\text{for one point}) \approx 2 [2f(0)] = 4 \times 2338 \\ = 9354$$

$$= 56y^3 + 588y^2 + 2338 + 2040y$$

$$I(\text{for two point Gauss quadrature}) = 2 \left[1 \cdot f\left(\frac{1}{\sqrt{3}}\right) + 1 \cdot f\left(-\frac{1}{\sqrt{3}}\right) \right] \\ = 2 \left[(588 \cdot \frac{1}{3} + 2338) \times 2 \right] \\ = 10,136$$

$$I(\text{for 3 point}) = 2 \left[\frac{8}{9} f(0) + \frac{5}{9} f\left(\sqrt{\frac{3}{5}}\right) + \frac{5}{4} f\left(-\sqrt{\frac{3}{5}}\right) \right]$$

$$= 2 \left[\frac{8}{9} \times 2338 + \frac{10}{4} (2338 + 588 \times \frac{3}{5}) \right]$$

Example

$$I = \int_2^3 \frac{dx}{1+x}, \quad 5 \text{ point Simpson} \\ h = \frac{b-a}{n-1} = 0.25$$

$$= \ln(1+x) \Big|_2^3$$

$$= \ln \frac{4}{3} = 0.287682 .$$

$$f(2) = \frac{1}{1+2} = 0.333333$$

$$f(2.25) = \frac{1}{1+2.25} = 0.307692$$

$$f(2.5) = \frac{1}{1+2.5} = 0.285714$$

$$f(2.75) = 0.266666$$

$$f(3) = \frac{1}{4} = 0.25$$

$$s = \frac{h}{3} \left[0.33333 + 4(0.307692 + 0.266666) + 2(0.285714) + 0.25 \right]$$

$$= 0.287682 .$$

→ with 2 point Gauss - quadrature.

$$I = \int_{-1}^1 \frac{dx}{1+x}$$

$$= \int_{-1}^1 f\left(\frac{5}{2} + \frac{1}{2}y\right) dy$$

$$f(y) = \frac{1}{2} \frac{\frac{2}{2+\frac{5}{2}+\frac{1}{2}y}}{7+y} = \frac{1}{7+y}$$

$$\Rightarrow G^1 = 2 \times \frac{1}{7} = 0.285714$$

$$G^2 = 1 \left(\frac{1}{7+\sqrt{43}} + \frac{1}{7-\sqrt{43}} \right)$$

$$= 0.287671$$

$$G^2 = \frac{8}{9} \times \frac{1}{7} + \frac{5}{9} \left(\frac{1}{7 + \sqrt{3}/5} + \frac{1}{7 - \sqrt{3}/5} \right)$$

$$= 0.126984 + 0.160697$$

$$= 0.287682$$

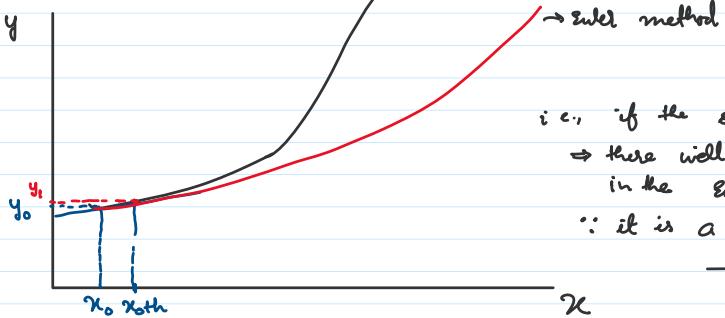
→ Solution of ODE :-

I.C. :-
 $y(x_0) = y_0 \quad \frac{dy}{dx} = f(x, y) \rightarrow y = y(x)$
 $y(x_0) = y_0 \quad \frac{d^2y}{dx^2} = g(x, y, y')$
 $y'(x_0) = y'_0$

Taylor expansion → Euler's first order approximation

$$y(x_0 + h) = y(x_0) + \frac{h}{1!} y'(x_0) + \frac{h^2}{2!} y''(x_0) + \mathcal{O}(h^3)$$

Geometrically →



i.e., if the slope change
 → there will be huge deviation
 in the Euler's method.
 ∵ it is a first order approximation.

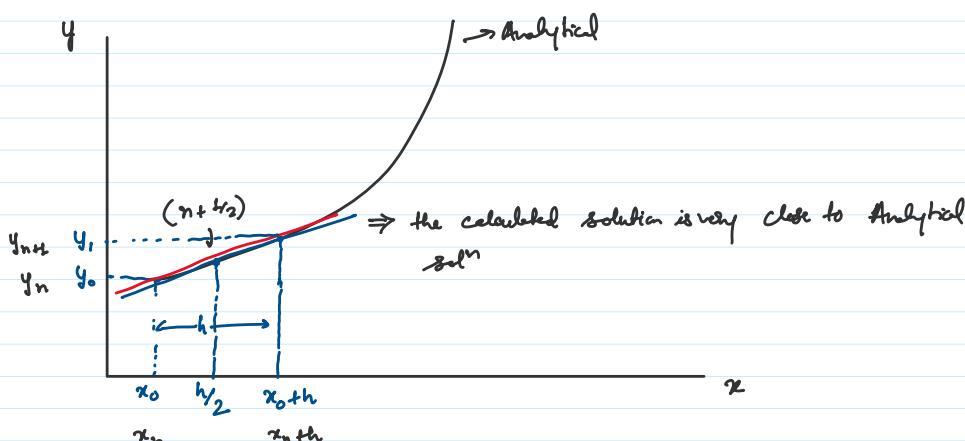
Euler Method :-

$$y_{n+1} = y_n + h f(x_n, y_n)$$

• Improvement in the method → second order approximation.

$$y(x_0 + h) = y(x_0) + h y'(x_0) + \frac{h^2}{2!} y''(x_0) + \mathcal{O}(h^3)$$

Runge - Kutta 2nd order



• Runge - Kutta 2nd order method / Midpoint method / modified Euler method
 Let us assume that we know y_{n+1} analytically

- Runge - Kutta - order method / Adams method / modified Euler method
 let us assume that we know $y_{n+\frac{h}{2}}$ analytically

$$y_n \equiv y_{n+\frac{h}{2}} \left(x - \frac{h}{2} \right) = y_{n+\frac{h}{2}} - \frac{h}{2} y'_{n+\frac{h}{2}} + \frac{h^2}{8} y''_{n+\frac{h}{2}} + O(h^3)$$

$$\Rightarrow y_{n+\frac{h}{2}} \equiv y_{n+\frac{h}{2}} \left(x + \frac{h}{2} \right) = y_{n+\frac{h}{2}} + \frac{h}{2} y'_{n+\frac{h}{2}} + \frac{h^2}{8} y''_{n+\frac{h}{2}} + O(h^3)$$

$$\Rightarrow y_{n+\frac{h}{2}} - y_n = h y'_{n+\frac{h}{2}}$$

$$\Rightarrow y_{n+\frac{h}{2}} = y_n + h y'_{n+\frac{h}{2}} + O(h^3)$$

where $y'_{n+\frac{h}{2}}$ can be calculated using Euler's Method :-

$$\text{Now, } y'_{n+\frac{h}{2}} = f(x_{n+\frac{h}{2}}, y_{n+\frac{h}{2}})$$

$$\Rightarrow y'_{n+\frac{h}{2}} = f\left(x_{n+\frac{h}{2}}, y_n + \underbrace{\frac{h}{2} f(x_n, y_n)}_{\substack{\text{Euler method} \\ (\because \text{it is called modified Euler's method})}}\right)$$

$$\text{Let } \begin{cases} s_1 = f(x_n, y_n) \rightarrow \text{Euler} \\ s_2 = f(x_{n+\frac{h}{2}}, y_n + \frac{h}{2} s_1) \end{cases}$$

$$\Rightarrow y'_{n+\frac{h}{2}} \equiv s_2$$

$$\Rightarrow y_{n+\frac{h}{2}} = y_n + h s_2 \rightarrow \text{final recursion relation}$$

where s_1 and s_2 are defined.

Now, from Taylor expansion, we have

$$y_{n+\frac{h}{2}} \approx y_n + h y'_n + \frac{h^2}{2} y''_n$$

$$\text{also, } y''_n = \frac{y'_{n+\frac{h}{2}} - y'_n}{h}$$

$$\Rightarrow y_{n+1} \approx y_n + \frac{h}{2} (y'_n + y'_{n+1})$$

$$= y_n + \frac{h}{2} (f(x_n, y_n), f(x_{n+1}, y_{n+1}))$$

∴ $y_{n+1} = y_n + \frac{h}{2} (f(x_n, y_n) + f(x_n + h, y_n + h f(x_n, y_n)))$

$\boxed{\begin{array}{l} S_1 \\ S_2' \quad (\text{Euler}) \end{array}}$

$$y_{n+1} = y_n + \frac{h}{2} (S_1 + S_2')$$

$$S_1 = f(x_n, y_n)$$

$$S_2' = f(x_n + h, y_n + h f(x_n, y_n))$$

• R-K 4th order approximation :-

$$y_{n+1} = y_n + \frac{h}{6} (k_1 + 2k_2 + 2k_3 + k_4)$$

$k_i = \text{slope}$

$$k_1 = f(x_n, y_n)$$

$$k_2 = f(x_n + \frac{h}{2}, y_n + \frac{h}{2} k_1)$$

$$k_3 = f(x_n + \frac{h}{2}, y_n + \frac{h}{2} k_2)$$

$$k_4 = f(x_n + h, y_n + h k_3)$$

$$\text{Error} \sim O(h^5)$$

Ex -

$$\frac{dy}{dx} = -xy, \quad y(0) = 1 \quad \left| \begin{array}{l} \ln y = -\frac{x^2}{2} \\ \Rightarrow y = e^{-\frac{x^2}{2}} \\ \Rightarrow y(0.1) = 0.995012 \end{array} \right.$$

$$f(x, y) = -xy, \quad h = 0.1$$

Mid-point method

x	x_{n+1}	y	$S_1 = f(x, y)$	$S_2 = f(x_{n+1}, y_n + h \frac{S_1}{2})$	$y_{n+1} = y_n + h S_2$	y_{n+1} Analytical value
0	0.05	1	0	$-(0.05)(1+0) = -0.05$	0.995	0.995012
0.1	0.15	0.995	-0.0995	$-(0.15)(0.995 - 0.05 \times 0.995) = -0.1485$	0.98015	0.980159
0.2	0.25	0.98015	-0.19603	$-(0.25)(0.98015 - 0.05 \times 0.19603) = -0.45599$	0.9559	0.95599

$$0.2 \quad 0.25 \quad 0.98015 \quad -0.19603 \quad - (0.25) (0.98015 - 0.05 \times 0.19603) \\ = -0.2425 \quad 0.9559 \quad 0.95599$$

→ R-K 2nd order:-

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

$$S_1 = f(x_n, y_n)$$

$$S_2 = f(x_n + dh, y_n + dh S_1)$$

weight factor

$$y_{n+1} = y_n + h \cdot (\omega_1 S_1 + \omega_2 S_2)$$

$$\omega_1 + \omega_2 = 1 \quad \text{and} \quad d = \frac{1}{2}$$

- for mid-point method :-

$$\omega_1 = 0, \quad \omega_2 = 1 \quad \text{and} \quad d = \frac{1}{2}$$

- for Heun's method :-

$$\omega_1 = \omega_2 = \frac{1}{2} \quad \text{and} \quad d = \frac{1}{2}$$

→ Coupled ODE:-

$$\left. \begin{array}{l} \frac{dy}{dx} = f(x, y, z) \\ \frac{dz}{dx} = g(x, y, z) \end{array} \right\} \begin{array}{l} \text{I.C. :-} \\ \text{at } x=x_0, y=y_0, z=z_0 \end{array}$$

$$S = f(x_n, y_n, z_n)$$

$$P = g(x_n, y_n, z_n)$$

— Using Euler's method the solution for 1st eqn using S is given by :-

$$y_{n+1} = y_n + h S$$

Similarly for the 2nd eqn using P :-

Similarly for the 2nd eq using P :-

$$z_{n+1} = z_n + hP$$

$$\text{and, } x_{n+1} = x_n + h$$

- Using RK2 (Heun's method) :-

$$S_1 = f(x_n, y_n, z_n)$$

$$P_1 = g(x_n, y_n, z_n)$$

$$S_2 = f(x_n + h, y_n + h S_1, z_n + h P_1)$$

$$P_2 = g(x_n + h, y_n + h S_1, z_n + h P_1)$$

$$y_{n+1} = y_n + \frac{h}{2} (S_1 + S_2)$$

$$z_{n+1} = z_n + \frac{h}{2} (P_1 + P_2)$$

$$x_{n+1} = x_n + h$$

- For 2nd order ODE :-

$$a \frac{d^2y}{dx^2} + b \frac{dy}{dx} + cy = d \rightarrow \text{we solve by breaking it into two coupled ODE.}$$

$$\frac{dy}{dx} = z = f(x, y, z) \quad \text{--- (1)}$$

$$a \frac{dz}{dx} + bz + cy = d$$

$$\frac{dz}{dx} = \frac{d - cy - bz}{a} = g(x, y, z) \quad \text{--- (2)}$$

Ex: $\frac{d^2y}{dx^2} + 5 \frac{dy}{dx} + 4y = 5 \quad , \quad y(0) = 0 = y'(0)$

$$\text{Analytical soln: } y_a = \frac{5}{4} - \frac{5}{3} e^{-x} + \frac{5}{12} e^{-4x}$$

$$\frac{dy}{dx} = z \rightarrow f(x, y, z) = S \quad \text{--- (1)}$$

$$\frac{dz}{dx} + 5z + 4y = 5 \quad \rightarrow g(x, y, z) = P$$

$$\therefore \frac{dz}{dx} = 5 - 4y - 5z \quad \text{--- (2)}$$

(1) Euler's method

$$x_0 = y_0 = z_0 = 0$$

$$S = z = f(x, y, z)$$

$$P = 5 - 4y - 5z = g(x, y, z)$$

$$h = 0.1$$

$$x_1 = 0 + 0.1 = 0.1$$

$$S_0 = 0$$

$$P_0 = 5$$

$$y_1 = y_0 + hS_0 = 0$$

$$z_1 = z_0 + hP_0 = 0.1 \times 5 = 0.5$$

$$x_2 = 0.1 + 0.1 = 0.2$$

$$S_1 = 0.5$$

$$P_1 = 5 - 4y_1 - 5z_1$$

$$= 5 - 5(0.5) = 2.5$$

$$y_2 = y_1 + hS_1 = 0 + 0.1 \cdot 0.5 = 0.05$$

$$z_2 = z_1 + hP_1 = 0.5 + 0.1 \times 2.5 = 0.75$$

$$x_3 = 0.2 + 0.1 = 0.3$$

$$s_2 = 0.25$$

$$P_2 = 5 - 4 \times 0.05 - 5 \times 0.25$$

$$= 5 - 0.2 - 3.75$$

$$= 5 - 3.95 = 1.05$$

$$y_3 = y_2 + h s_2 = \frac{0.05 + 0.1 \times 0.25}{0.125}$$

$$z_3 = z_2 + h P_2 = \frac{0.25 + 0.1 \times 1.05}{0.125}$$

i	0	1	2	3
x	0	0.1	0.2	0.3
y	0	0	0.05	0.125
z	0	0.5	0.25	0.855

② RK2 method :- $x_0 = y_0 = z_0 = 0$

$$s_1 = f(x, y, z) = z$$

$$P_1 = g(x, y, z) = 5 - 4y - 5z$$

$$s_2 = f(x_n + h, y_n + hs_1, z_n + hP_1) = z + hP_1$$

$$P_2 = g(x_n + h, y_n + hs_1, z_n + hP_1)$$

$$= 5 - 4(y_n + hs_1) - 5(z_n + hP_1)$$

$$= 5 - 4y_n - 5z_n - h(4s_1 + 5P_1)$$

$$y_{n+1} = y_n + \frac{h}{2}(s_1 + s_2)$$

$$z_{n+1} = z_n + \frac{h}{2} (P_1 + P_2)$$

$$z_{n+1} = z_n + h$$

$$x_1 = 0.1$$

$$S_1 = 6$$

$$P_1 = 5$$

$$S_2 = 0 + (0.1) \times 5 = 0.5$$

$$\begin{aligned} P_2 &= 5 - 0 - 5(0 + 0.1 \times 5) \\ &= 5 - 2.5 = 2.5 \end{aligned}$$

$$y_1 = 0 + \frac{0.1}{2} \times (0.5) = \frac{0.05}{2} = 0.025$$

$$z_2 = 0 + \frac{0.1}{2} \times (5 + 2.5) = \frac{0.75}{2} = 0.375$$

$$x_2 = 0.2$$

$$S_2 = 0.375$$

$$\begin{aligned} P_1 &= 5 - 4(0.025) - 5(0.375) \\ &= 5 - 0.1 - 1.875 \\ &= +3.025 \end{aligned}$$

$$\begin{aligned} S_2 &= 0.375 + 0.1 \times 3.025 \\ &= 0.6725 \end{aligned}$$

$$\begin{aligned} P_2 &= 5 - 4 \times (0.025) - 5 \times (0.375) - 0.1(4(0.375) + 5(3.025)) \\ &= 5 - 0.1 - 1.875 - 1.6625 \\ &= 1.3625 \end{aligned}$$

$$y_2 = 0.025 + \frac{0.1}{2} (0.375 + 0.625) \\ = 0.077625$$

$$z_2 = 0.375 + \frac{0.1}{2} (3.025 + 1.375) \\ = 0.594375$$

$$x_3 = 0.3$$

s_1

i	0	1	2	3
x	0	0.1	0.2	0.3
y	0	0.025	0.077625	0.145650625
z	0	0.375	0.594375	0.711309375
y_a	0	0.021	0.074	0.14

RK 2 - method is more accurate than Euler's method.

→ Finite difference method:-

① In Euler/RK method a differential eqn is converted into a recursion relation.

However, in finite difference method we convert the differential eqn to a set of linear/non-linear eqn.

② In Euler/RK - I.C. are given. $\rightarrow y'' + w^2 y = 0$ | $y(0) = a$; $y'(0) = b$

But FD method deals with I.C. and Boundary values (B.V.).



$$\underline{y'' + w^2 y = 0}$$

$$y'' + \omega^2 y = 0$$

$$y(0) = a, \quad y(\pi) = -a$$

\rightarrow Euler method cannot work with this boundary condition.

③ FD method can be applicable for PDEs.

\rightarrow Finite difference Method:-

$$\begin{cases} f(x) \rightarrow f(x+h) = f(x) + h f'(x) + \frac{h^2}{2!} f''(x) + \mathcal{O}(h^3) & -\textcircled{1} \\ f(x-h) = f(x) - h f'(x) + \frac{h^2}{2!} f''(x) - \mathcal{O}(h^3) & -\textcircled{2} \end{cases}$$

from -\textcircled{1}, truncating the series to order 2

Forward difference

$$\Rightarrow f'(x) \approx \frac{f(x+h) - f(x)}{h} + \mathcal{O}(h) \quad -\textcircled{3}$$

Similarly from -\textcircled{2}

Backward difference

$$f'(x) \approx \frac{-f(x-h) + f(x)}{h} + \mathcal{O}(h) \quad -\textcircled{4}$$

Now,

$$f(x+h) - f(x-h) = 2h f'(x) + \mathcal{O}(h^3)$$

Central difference

$$\Rightarrow f'(x) = \frac{f(x+h) - f(x-h)}{2h} + \mathcal{O}(h^2) \quad -\textcircled{5}$$

Also

$$f(x+h) + f(x-h) = 2f(x) + \frac{h^2}{2!} f''(x) + \mathcal{O}(h^4)$$

$$\Rightarrow f''(x) = \frac{f(x+h) - 2f(x) + f(x-h)}{h^2} + \mathcal{O}(h^2) \quad -\textcircled{6}$$

Central difference method

1.0 and 0.5 0.5

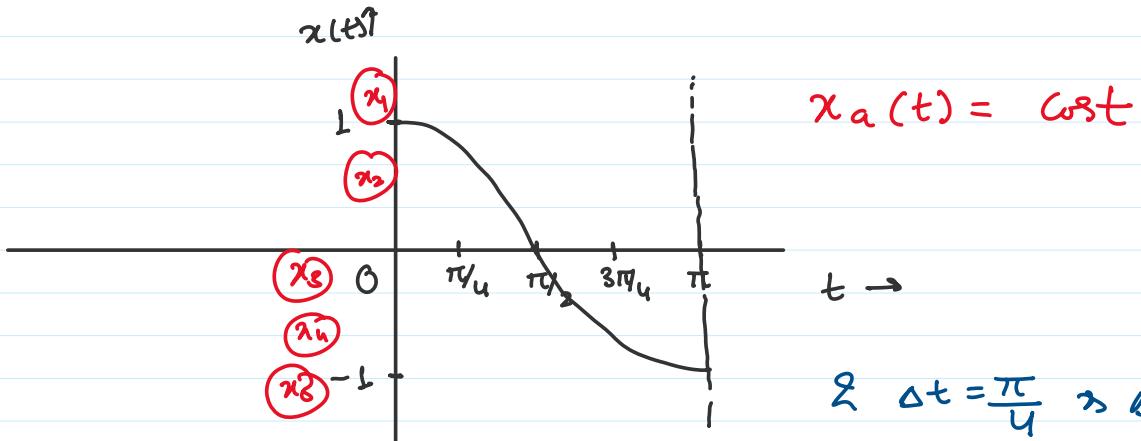
Central difference method
for 2nd derivative

— (6)

Ex:

$$\ddot{x} + \omega_0^2 x = 0, \quad \omega_0 = 1$$

$$\therefore \ddot{x} + x = 0, \quad x(0) = 1, \quad x(\pi) = 1, \quad \Delta t = \frac{\pi}{4}$$



$$2 \Delta t = \frac{\pi}{4} \Rightarrow \Delta t^2 = \frac{\pi^2}{16}$$

Using — (6) we can have the expression for \ddot{x}

Hence, our differential eqn becomes -

$$\frac{x_{n+1} - 2x_n + x_{n-1}}{\Delta t^2} + x_n = 0$$

$$x_{n+1} + x_n (\Delta t^2 - 2) + x_{n-1} = 0$$

$$\text{for } n=2 : - x_3 + x_2 \left(\frac{\pi^2}{16} - 2 \right) + x_1 = 0 \quad = -1.383$$

$$\therefore x_3 - 1.383 x_2 + 1 = 0 \quad - (1)$$

$$n=3 : - x_4 - 1.383 x_3 + x_2 = 0 \quad - (2)$$

$$n=4 : - x_5 - 1.383 x_4 + x_3 = 0$$

$$\text{or } -1 - 1.383 x_4 + x_3 = 0 \quad - (3)$$

$$x_2 = \frac{1 + x_3}{1.383}$$

$$1.383 \left(\cancel{x_4} - 1.383 x_3 + \frac{1}{1.383} + \frac{x_3}{1.383} = 0 \right)$$

$$-1 - 1.383 \cancel{x_4} + x_3 = 0$$

$$\Rightarrow x_3 = 0$$

$$\therefore x_2 = \frac{1}{1.383} = 0.7230658$$

$$x_4 = 1.383 x_3 - x_2 = -0.7230658.$$

Now for n eqns :-

$$x_1 = a, \quad x_N = -a$$

$$\begin{pmatrix} 1 & 0 & 0 & \cdots & 0 \\ x & x & x & \cdots & 0 \\ 0 & x & x & \cdots & 0 \\ 0 & 0 & x & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \cdots & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_N \end{pmatrix} = \begin{pmatrix} a \\ c_2 \\ \vdots \\ -a \end{pmatrix}$$

Spur Matrix.

→ Finite difference method:- (FD) method

$$\begin{aligned}
 y' &= \frac{y_{n+1} - y_n}{h} + \Theta(h) \\
 &= -\frac{y_{n-1} + y_n}{h} + \Theta(h) \\
 &= \frac{y_{n+1} - y_{n-1}}{2h} + \Theta(h^2) \\
 y'' &= \frac{y_{n+1} - 2y_n + y_{n-1}}{h^2} + \Theta(h^2)
 \end{aligned}$$

- SHO :-

$$\ddot{x} + \omega^2 x = 0, \quad \left\{ \begin{array}{l} x(\theta_0) = a_0 \\ x(\theta_n) = a_n \end{array} \right.$$

Dirichlet Boundary Condition

or we can have

$$\left\{ \begin{array}{l} x(\theta_0) = a_0 \\ \dot{x}(\theta_0) = b_0 \end{array} \right.$$

Neumann B.C.

Sparse matrix

$$\begin{pmatrix}
 1 & 0 & 0 & & \cdots & 0 \\
 x & x & x & & & \\
 0 & x & x & x & & \\
 \vdots & & x & x & x & \\
 0 & 0 & 0 & \cdots & \cdots & 1
 \end{pmatrix}
 \begin{pmatrix}
 x_0 \\
 x_1 \\
 x_2 \\
 \vdots \\
 x_n
 \end{pmatrix}
 =
 \begin{pmatrix}
 a_0 \\
 ; \\
 ; \\
 ; \\
 a_n
 \end{pmatrix}$$

- damped Harmonic Oscillator:-

$$m\ddot{x} + \gamma \dot{x} + kx = 0$$

$$\ddot{x} + 2\beta \dot{x} + \omega^2 x = 0$$

$$\frac{x_{n+1} - 2x_n + x_{n-1}}{(\Delta t)^2} + 2\beta \frac{x_{n+1} - x_{n-1}}{2\Delta t} + \omega_0^2 x_n = 0$$

$$\Rightarrow x_{n+1}(1 + \beta \Delta t) + x_n(\omega_0^2 \Delta t^2 - 2) + x_{n-1}(1 - \beta \Delta t) = 0$$

$$x_{n+1} = -a_1 x_n - a_2 x_{n-1} \quad \text{--- (1)}$$

$$\text{where } a_1 = \frac{\omega_0^2 \Delta t^2 - 2}{1 + \beta \Delta t}, \quad a_2 = \frac{1 - \beta \Delta t}{1 + \beta \Delta t}$$

from Neumann condition:-

$$x(0) = a_0 \Rightarrow x_0 = a_0$$

$$\dot{x}(0) = b_0 \Rightarrow \frac{x_1 - x_0}{h} = b_0$$

$$\Rightarrow x_1 = h b_0 + a_0$$

- determination of Δt :-

$$x_{n+1} = z x_n \quad \text{where } z \text{ is the propagator}$$

$$\text{and } x_n = z x_{n-1}$$

$$\Rightarrow x_{n+1} = z^2 x_{n-1}$$

} --- (2)

\Rightarrow from (1)

$$(z^2 + a_1 z + a_2) x_{n-1} = 0$$

$$\Rightarrow z = \frac{-a_1 \pm \sqrt{a_1^2 - 4a_2}}{2}$$

\Rightarrow from (2)

$$z = \frac{1}{2} [-a_1 \pm i D]$$

$$D = \sqrt{4a_2 - a_1^2}$$

for decaying solution:- $|z| < 1$

$$\Rightarrow \frac{1}{4}(\alpha_1^2 + D^2) < 1$$

$$\Rightarrow (\alpha_1^2 + 4\alpha_2 - \alpha_1^2) < 4$$

$$\Rightarrow \alpha_2 < 1$$

$$\Rightarrow \frac{1-\beta\Delta t}{1+\beta\Delta t} < 1, \quad \beta\Delta t > 0 \Rightarrow \beta > 0.$$

2nd condition:- $\text{Im}(z) \neq 0$

$$\Rightarrow D \neq 0$$

$$\Rightarrow 4\alpha_2 - \alpha_1^2 > 0$$

$$\Rightarrow 4\alpha_2 > \alpha_1^2$$

$$\Rightarrow \frac{4(1-\beta\Delta t)}{1+\beta\Delta t} > \frac{(w_0^2 \Delta t^2 - 2)^2}{(1+\beta\Delta t)^2}$$

$$\Rightarrow \Delta t < \frac{2\sqrt{w_0^2 - \beta^2}}{w_0^2} \rightarrow \text{Stability Criteria.}$$



$$y' = f(x, y)$$

$$\frac{y_{n+1} - y_n}{h} = f(x_n, y_n)$$

$$y_{n+1} = y_n + h f(x_n, y_n)$$

[Explicit / forward Euler]

Now, using backward method:-

$$\frac{y_n - y_{n-1}}{h} = f(x_n, y_n)$$

$$\Rightarrow \frac{y_{n+1} - y_n}{h} = f(x_{n+1}, y_{n+1})$$

$$\Rightarrow \frac{y_{n+1} - y_n}{h} = f(x_{n+1}, y_{n+1})$$

$$\Rightarrow y_{n+1} = y_n + h f(x_{n+1}, y_{n+1}) \quad \boxed{\text{Implicit / Backward Euler}}$$

Ex: $\frac{dy}{dx} = \lambda y \quad y(x=0) = 1$

$$\Rightarrow \ln y = \lambda x + C$$

$$\because C = 0 \Rightarrow \boxed{y = e^{\lambda x}}$$

Explicit

$$y_{n+1} = y_n + h \lambda y_n$$

$$\text{or } y_{n+1} = y_n \underbrace{(1 + h\lambda)}_z$$

$\lambda < 0 \rightarrow \text{decaying}$

$$\Rightarrow -1 < (1 + h\lambda) < 1$$

$$\Rightarrow h\lambda > -2$$

$$h > \frac{2}{|\lambda|}$$

Stability Criteria

for $\lambda > 0$:-

unconditionally
stable

Implicit

$$y_{n+1} = y_n + h \lambda y_{n+1}$$

$$\text{or } y_{n+1} = \underbrace{\left(\frac{1}{1 - h\lambda} \right)}_z y_n$$

$\lambda < 0$

$$z = \frac{1}{1 + h|\lambda|}$$

for $\lambda > 0$:-

$$h < \frac{1}{\lambda} \quad \boxed{\text{stability Criteria}}$$

Ex. $\frac{dy}{dx} = 2y - 3x \quad y(0) = 1$

explicit :-

$$y_{n+1} - y_n$$

$$\frac{y_{n+1} - y_n}{h} = 2y_n - 3x_{n+1}$$

Implicit :-

$$\frac{y_{n+1} - y_n}{h} = 2y_n - 3x_n$$

$$\frac{y_{n+1} - y_n}{h} = 2y_{n+1} - 3x_{n+1}$$

$$\Rightarrow y_{n+1} = y_n(2h+1) - 3hx_n$$

$$y_{n+1} = \frac{y_n - 3h(x_n + h)}{1 - 2h}$$

for $h = 0.1$

$$y_0 = 1$$

$$\begin{aligned} y_1 &= 1(0.2+1) - 3(0.1) \cdot 0 \\ &= 1.2 \end{aligned}$$

$$\begin{aligned} y_1 &= \frac{1 - 3(0.1)(0+0.1)}{1 - 2(0.1)} \\ &= \frac{1 - 0.03}{1 - 0.2} = 1.2125 \end{aligned}$$

$$\begin{aligned} y_2 &= 1.2(2 \times 0.1 + 1) - 0.3 \times 0.1 \\ &= 1.41 \end{aligned}$$

$$\begin{aligned} y_2 &= \frac{1.2125 - 0.3(0.2)}{0.8} \\ &= 1.4406 \end{aligned}$$

$$\text{---}^{\circ}\text{---}$$

$$\frac{dy}{dx} - 2y = -3x$$

$$\text{I.F.} = e^{\int -2 dx} = e^{-2x}$$

$$ye^{-2x} = \int -3x e^{-2x} dx + C$$

$$= -3 \left[x \frac{e^{-2x}}{-2} + \frac{1}{2} \frac{e^{-2x}}{(-2)^2} \right] + C$$

$$y = \left(\frac{3x}{2} + \frac{3}{4} \right) + C e^{2x}$$

$$1 = \frac{3}{4} + C \Rightarrow C = \frac{1}{4}$$

$$\therefore y_{\text{an}} = \frac{1}{4} e^{2x} + \frac{3x}{2} + \frac{3}{4}$$

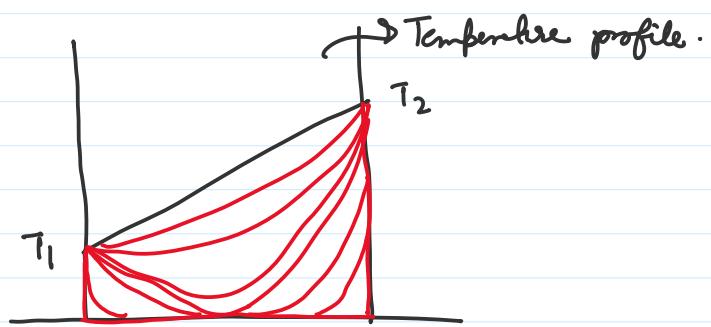
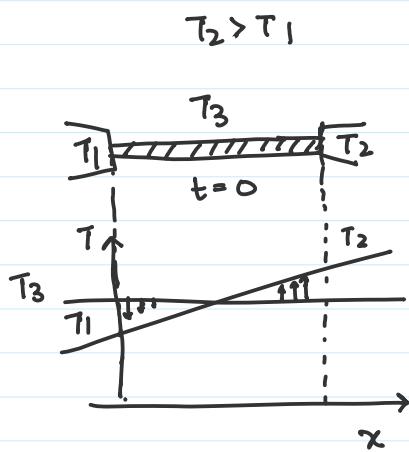
$$y(0.1) = 1.20535$$

$$y_2(0.2) = 1.42295$$

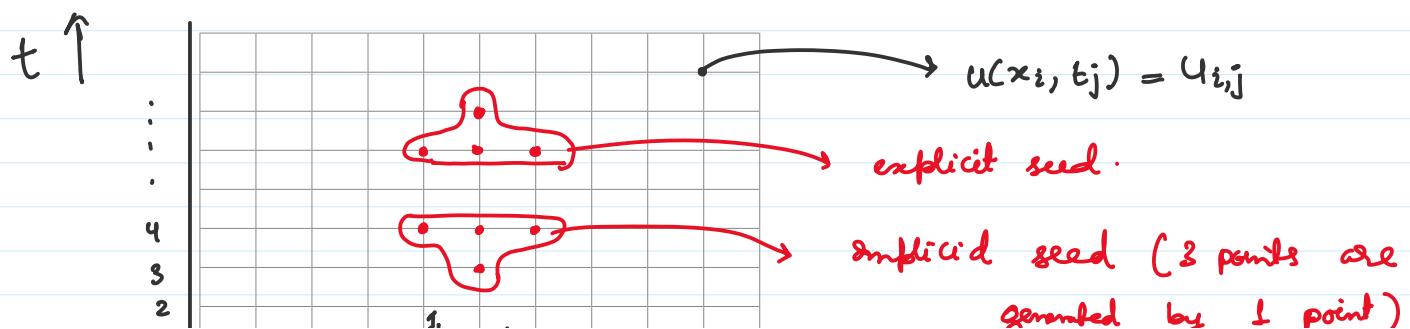
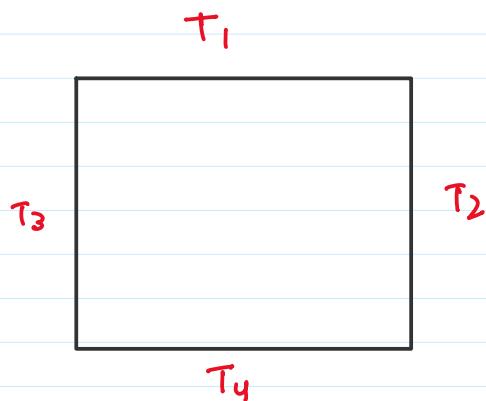
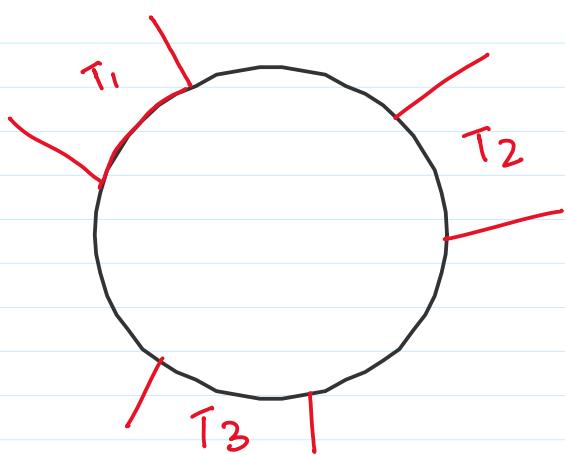
→ Partial differential equations (PDEs) :-

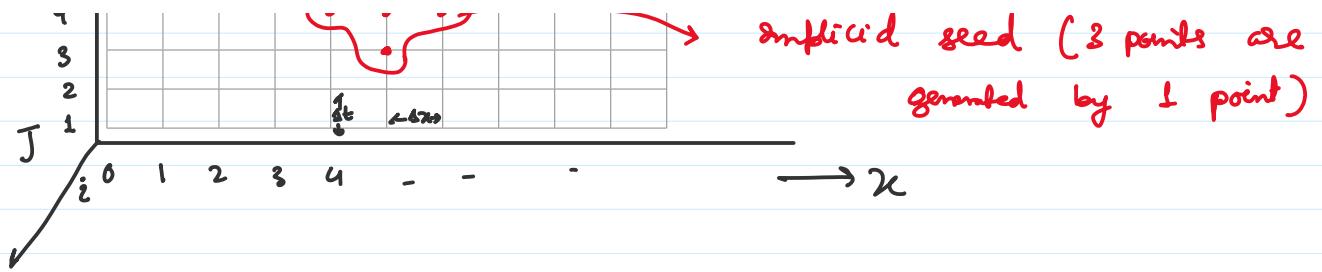
1. 1D - heat eqn :- $\frac{\partial u(x,t)}{\partial t} = \alpha \frac{\partial^2 u(x,t)}{\partial x^2}$

2. Laplace eqn :- $\frac{\partial^2 u(x,y)}{\partial x^2} + \frac{\partial^2 u(x,y)}{\partial y^2} = 0$



for 2D :-





Explicit scheme :-

Forward difference in time

Central in space

$$\frac{\partial u}{\partial t} = \frac{u(x, t + \Delta t) - u(x, t)}{\Delta t}$$

$$\frac{\partial^2 u}{\partial x^2} = \frac{u(x + \Delta x, t) - 2u(x, t) + u(x - \Delta x, t)}{(\Delta x)^2}$$

$$\text{let } \sigma = \frac{\alpha \Delta t}{(\Delta x)^2}$$

$$u(x, t + \Delta t) = \sigma u(x + \Delta x, t) - (2\sigma - 1) u(x, t) + \sigma u(x - \Delta x, t)$$

$$\therefore u(x, t + \Delta t) = \underbrace{A^E(\sigma)}_{\text{is propagator}} u(x, t)$$

$$A = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & - & - & - & \cdot \\ 0 & -2\sigma + 1 & 0 & 0 & 0 & 0 & - & \cdot & \cdot \\ 0 & 0 & -2\sigma + 1 & 0 & 0 & 0 & - & \cdot & \cdot \\ \vdots & \vdots & \ddots & & & & & & 1 \end{bmatrix}$$

$$u(x, 0) = \begin{pmatrix} u_0^0 \\ u_1^0 \\ u_2^0 \\ \vdots \\ \vdots \\ u_n^0 \end{pmatrix} \rightarrow u(x, \Delta t) = \begin{pmatrix} u_0' \\ u_1' \\ u_2' \\ \vdots \\ \vdots \\ u_n' \end{pmatrix}$$

for N iterations :-

$$u(x, N\Delta t) = \left(A^{E(\sigma)} \right)^N u(x, 0) \rightarrow \text{Conditionally Stable}$$

$$0 < \sigma < \gamma_2$$

→ FDM - Heat equation (1-D):-

$$\frac{\partial u(x,t)}{\partial t} = \alpha \frac{\partial^2 u(x,t)}{\partial x^2}$$

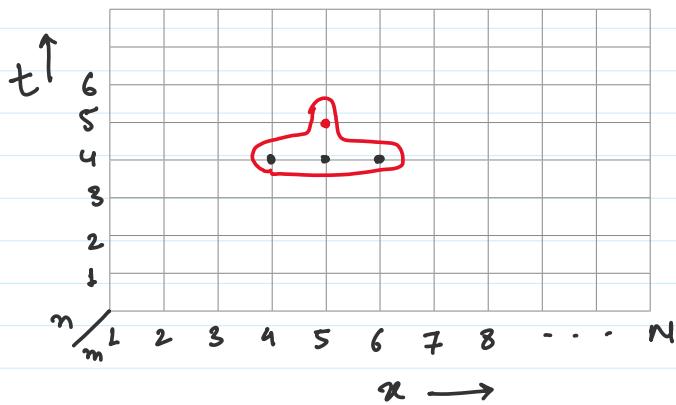
→ Explicit :-

forward difference in time at x
Central " " space at t

Thermal diffusivity
(mm^2/sec)

$$u(x, t + \Delta t) = \sigma u(x + \Delta x, t) + (1 - 2\sigma) u(x, t) + \sigma u(x - \Delta x, t)$$

$$u(m, n+1) = \sigma u(m+1, n) + (1 - 2\sigma) u(m, n) + \sigma u(m-1, n)$$



$$A^E = \begin{pmatrix} 1 & 0 & 0 & 0 & \dots & \dots \\ \sigma & 1-2\sigma & \sigma & 0 & \dots & \dots \\ 0 & \sigma & 1-2\sigma & \sigma & 0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \dots & 1 \end{pmatrix}$$

→ Implicit :-

Forward difference in time at x

Central difference in space at $t + \Delta t$

$$u(x, t) = -\sigma u(x + \Delta x, t + \Delta t) + (1 + 2\sigma) u(x, t + \Delta t) - \sigma u(x - \Delta x, t + \Delta t)$$

$$u(m, n) = -\sigma u(m+1, n+1) + (1 + 2\sigma) u(m, n+1) - \sigma u(m-1, n+1)$$

$\alpha \rightarrow$ Thermal diffusivity

$$\alpha = \frac{k}{\rho C_p} \rightarrow \text{thermal conductivity}$$

$$\sigma = \frac{\alpha \Delta t}{(\Delta x)^2} \quad (\text{dimensionless})$$

$$A^I = \begin{pmatrix} 1 & 0 & 0 & 0 & \dots & \dots \\ -\sigma & 1+2\sigma & -\sigma & 0 & \dots & \dots \\ 0 & -\sigma & 1+2\sigma & -\sigma & \dots & \dots \\ \vdots & & & & & \\ 0 & 0 & - & - & - & 1 \end{pmatrix}$$

$$\Rightarrow u(x, t) = A^I u(x, t + \Delta t)$$

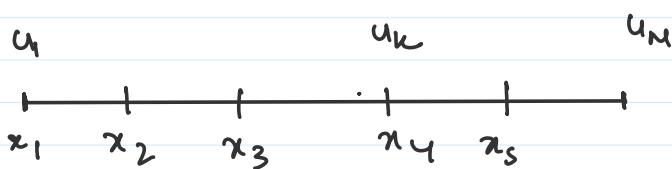
$$\Rightarrow u(x, t + \Delta t) = (A^I)^{-1} u(x, t)$$

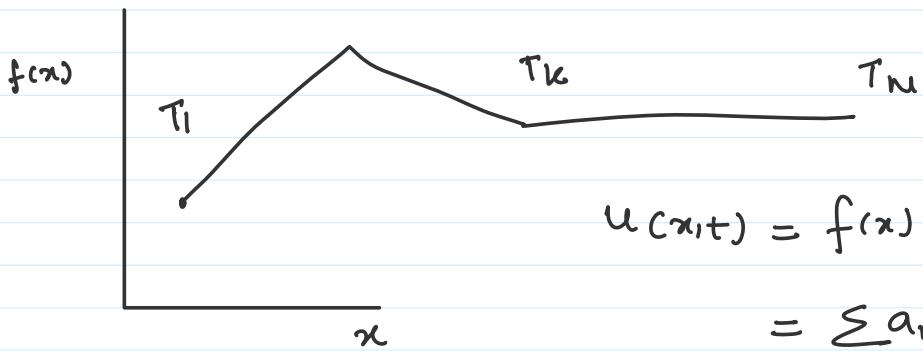
$$\text{or } u^k = (A^I)^{-1} u^{k-1}$$

$$= (A^I)^{-k} u^0$$

- Implicit scheme is unconditionally stable.

→ Instability of Explicit scheme:-





$u(x, t) = f(x) \rightarrow \text{random functions}$

$$= \sum_n a_n e^{ik_n x}$$

→ Von-Neumann stability analysis:-

PDE with periodic b.c. has general spatial solution

$$u_x = e^{ikx}, \quad k \rightarrow \text{constant}$$

$$u(x, t) = u_x u_t$$

Assume: time evolution to be exponential

$$u_t = e^{at}$$

$$\Rightarrow u(x, t) = e^{ikx} e^{at}$$

$$u(m, n) = e^{ikm\Delta x} e^{an\Delta t}$$

$x \rightarrow m\Delta x$
 $t \rightarrow n\Delta t$

$$u(m-1, n) = e^{ik(m-1)\Delta x} e^{an\Delta t}$$

$$u(m+1, n) = e^{ik(m+1)\Delta x} e^{an\Delta t}$$

$$u(m, n+1) = e^{ikm\Delta x} e^{a(n+1)\Delta t}$$

Substituting these values in explicit form gives:-

$$u(m, n+1) = \sigma u(m+1, n) + (1 - 2\sigma) u(m, n) + \sigma u(m-1, n)$$

$$\Rightarrow \underbrace{e^{ikm\Delta x} e^{a(n+1)\Delta t}}_{u(m, n+1)} = \sigma \left[e^{ik(m+1)\Delta x} e^{an\Delta t} \right] + \\ \left. \left(1 - \sigma \right) \right] e^{ikm\Delta x} e^{an\Delta t} +$$

$u(m, n+1)$

$$\begin{aligned}
 & (1-2\sigma) \left[e^{ikm\Delta x} e^{a\Delta t} \right] + \\
 & \sigma \left[e^{ik(m-1)\Delta x} e^{a\Delta t} \right] \\
 & = \frac{e^{a\Delta t} e^{ikm\Delta x}}{u(m, n)} \left[\sigma (e^{ik\Delta x} + e^{-ik\Delta x}) - (2\sigma - 1) \right]
 \end{aligned}$$

Amplitude factor, $G = \frac{u(m, n+1)}{u(m, n)}$

 \Rightarrow

$$G = 1 - 2\sigma + 2\sigma \cos(k\Delta x)$$

$$|G| = 1 - 4\sigma \quad \text{for} \quad k\Delta x = n\pi ; n = \text{odd}$$

So, one condition is :- $|G| < 1$ (as we need converging solⁿ)
 $\Rightarrow -1 < (1 - 4\sigma) < 1$ \hookrightarrow i.e., the amplitude factor < 1

$$\Rightarrow -2 < -4\sigma < 0$$

$$\Rightarrow \boxed{\frac{1}{2} > \sigma > 0}$$

$$\boxed{0 < \sigma < 0.5}$$

\Rightarrow the amplitude factor is not increasing and the solⁿ is converging.

similarly, for implicit scheme:-

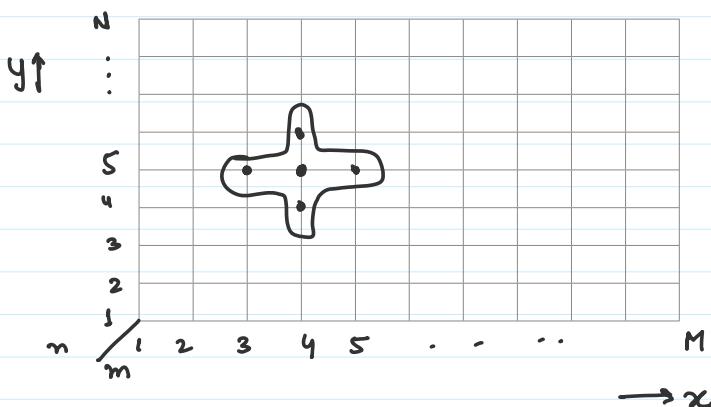
$$u(m, n) = -\sigma u(m+1, n+1) + (1+2\sigma) u(m, n+1) - \sigma u(m-1, n+1)$$

$$G_1 = \frac{u(m, n+1)}{u(n, m)} = \frac{1}{1+2\sigma - 2\sigma \cos k\Delta x}$$

$$= \frac{1}{1+4\sigma} < 1 \Rightarrow \text{Implicit scheme is always stable.}$$

→ Laplace Eqn :- (2D)

$$\frac{\partial^2 u(x, y)}{\partial x^2} + \frac{\partial^2 u(x, y)}{\partial y^2} = 0$$



$$\rightarrow \frac{u(m+1, n) - 2u(m, n) + u(m-1, n)}{(\Delta x)^2} + \frac{u(m, n+1) - 2u(m, n) + u(m, n-1)}{(\Delta y)^2} = 0$$

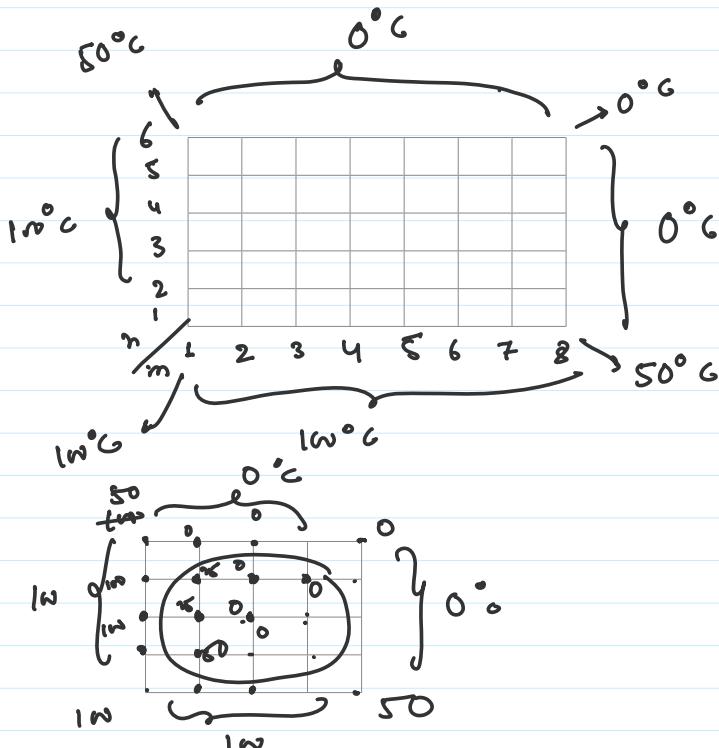
$$\rightarrow u(m, n) = \frac{u(m-1, n) + u(m+1, n) + \beta^2 (u(m, n+1) + u(m, n-1))}{2(1+\beta^2)}$$

$$\beta = \frac{\Delta x}{\Delta y}$$

$$P = \frac{1}{\Delta y}$$

for $\Delta x = \Delta y$ ($\beta = 1$)

$$u(m,n) = \frac{1}{9} \left[u(m+1,n) + u(m-1,n) + u(m,n+1) + u(m,n-1) \right]$$



Average temp. for corner points

$$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{pmatrix} \rightarrow \begin{pmatrix} 25 & 0 & 0 \\ 25 & 0 & 0 \\ 50 & 25 & 25 \end{pmatrix}$$



$$\begin{pmatrix} 31.25 & 6.25 & 0 \\ 43.75 & 12.5 & 6.25 \\ 62.5 & 43.75 & 31.25 \end{pmatrix}$$

difference, $D^{k+1}(m,n) = u^{k+1}(m,n) - u^k(m,n)$

user defined precision limit $\rightarrow \epsilon$

if $\max\{|D(m,n)|\} \leq \epsilon$

then stop.