

Numerical Analysis

$$P(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0 = 0$$

↳ Algebraic eqn

$$\text{Ex: } x^3 - 5x - 7 = 0$$

$$\tan x - x^2 - 5 = 0$$

↳ Transcendental

functions

* An eqn of the form,

$$f(x) = a_n x^n + a_{n-1} x^{n-1} + \dots + a_1 x + a_0 = 0$$

is called algebraic equation.

Bisection Method

Let $f: \mathbb{R} \rightarrow \mathbb{R}$ be continuous function

Step 1: Find $(a, b) \in \mathbb{R}^2$ such that $f(a), f(b) < 0$

Step 2: Choose c as $\frac{a+b}{2}$

Step 3: If :

* $f(c) = 0$ then c is a root

* $f(a)f(b) < 0$ root lies in $[a, b]$
put $a = c$ repeat

* $f(c)f(b) < 0$ root lies in $[c, b]$
put $b = c$

Step 4: Repeat step 2 & 3 till

$$f(c) = 0$$

example

Consider,

$$f(x) = x^3 - 3x + 1 = 0$$

find root of eqn.

$$a=0 \quad f(0) = 1$$

$$b=1 \quad f(1) = 1-3+1 = -1$$

$$c_1 = 1/2$$

$$\begin{aligned} f(c_1) &= 1/8 - 3/2 + 1 \\ &= 9/8 - 12/8 = -3/8 \end{aligned}$$

$$f(c_1) \cdot f(a) < 0$$

$$\therefore b = c_1$$

$$b = 1/2$$

$$c_2 = 1/4$$

$$f(c_2) = \frac{1}{64} - \frac{3}{4} + 1 > 0$$

$$f(c_2) \cdot f(b) < 0$$

$$\text{root } \in [c_1, c_2] \quad a = c_2$$

$$c_3 = \frac{1/2 + 1/4}{2} = 3/8$$

~~$a = 1 \quad f(1) = -1$~~

~~$b = 2 \quad f(2) = 3$~~

~~$c = \frac{1+2}{2} = 3/2$~~

~~$f(3/2) = 5/8 < 0 \quad c_1 = 3/2$~~

~~$f(3/2) > 0$~~

~~$f(1) \cdot f(3/2) < 0$~~

root exists b/w $[1, 3/2]$

$$c_2 = \frac{1+3/2}{2}$$

$$= \frac{5/4}{2} = 5/8$$

$$f(5/4) = 5/8 > 0 \Rightarrow \text{root}$$

$$0 > -\frac{5}{8} - 5/4 = -15/8$$

$$x^3 - 3x + 1 = 0$$

$$f(1) = -1$$

$$f(2) = 3$$

$$x_1 = \frac{1+2}{2} = 3/2$$

$$f(x_1) = (3/2)^3 - 3 \cdot 3/2 + 1$$

$$= 27/8 - 9/2 + 1$$

$$= -4/8 > -27/8 = 7/2 > 0$$

$$f(x_1) \cdot f(1) < 0$$

$$\therefore \text{root } (1, x_1)$$

$$x_2 = \frac{1+x_1}{2} = \frac{1+3/2}{2} = 5/4$$

$$f(x_2) = (5/4)^3 - 5/4 \cdot 3 + 1$$

$$= 125/64 - 15/4 < 0$$

$$f(x_2) \cdot f(2) < 0 \quad (x_2, x_1) \rightarrow \text{root}$$

$$\begin{array}{r} 1087 \\ 8 \\ \hline 30 \\ 24 \\ \hline 6 \end{array}$$

$$x_3 = \frac{1+x_2}{2} = \frac{1+5/4}{2} = 9/8$$

$$f(x_3) = (9/8)^3 - 3 \cdot 9/8 + 1$$

$$= 729/512 - 27/8 + 1$$

$$x_3 = \frac{x_2 + x_1}{2}$$

$$\text{root at } 3^{\text{rd}} \text{ approx } \frac{3/2 + 5/4}{2} = \frac{11/8}{2} = 11/8$$

Prob. 28.1

27-6-5

1) i) $x^3 - 2x - 5 = 0$

$$\begin{array}{ll} \cancel{x=0} & a=0 \quad f(a) = -5 \\ b=3 & f(b)=16 \end{array}$$

$$f(a) \cdot f(b) < 0$$

$$x_1 = 3/2$$

$$f(x_1) = \frac{27}{8} - 3 - 5 < 0$$

$$f(x_1) \cdot f(3) < 0$$

root $\in (x_1, 3)$

$$\bar{x}_2 = \frac{x_1 + 3}{2}$$

$$\bar{x}_2 = \frac{3/2 + 3}{2} = \frac{9/4}{2}$$

$$f(\bar{x}_2) = (9/4)^3 - 2 \cdot 9/4 - 5$$

$$= \frac{729}{64} - \frac{18}{4} - 5$$

$$= \frac{729 - 608}{64} > 0$$

$$f(x_1) \cdot f(\bar{x}_2) < 0$$

root $\in (x_1, \bar{x}_2)$

$$\bar{x}_3 = \frac{x_1 + \bar{x}_2}{2} = \frac{9/4 + 3/2}{2} = \frac{15}{8} = 1.875$$

$$f(\bar{x}_3) = \left(\frac{15}{8}\right)^3 - 2 \cdot \frac{15}{8} - 5$$

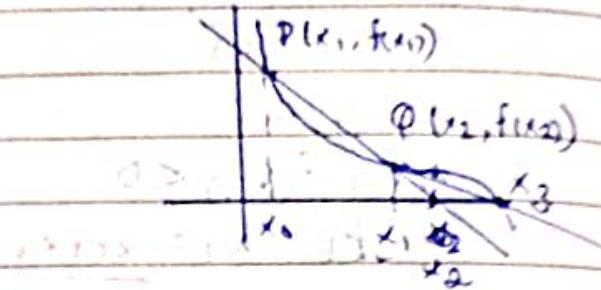
$$f(\bar{x}_3) = \frac{3375}{512} - \frac{15}{4} - 5 \approx -0.125 < 0$$

$(\bar{x}_2, \bar{x}_3) \rightarrow$ root

$$x_4 = \frac{\bar{x}_2 + \bar{x}_3}{2} = 2.0625$$

Secant Method

Let $f(x) = 0$



Algorithm

- Step 1 Consider $x_0 \& x_1$ as initial approximation of a root $f(x) = 0$
 Denote P for $(x_0, f(x_0))$ and φ for $(x_1, f(x_1))$

Step 2 Find eqn for for st. line - $P\varphi$

$$\frac{y - f(x_0)}{x - x_1} = \frac{f(x_1) - f(x_0)}{x_1 - x_0} \quad \text{--- (1)}$$

- Step 3 Put $y = 0$ in (1) & solve for x

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

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

$$\left[x_2 = x_1 - \frac{f(x_1)(x_1 - x_0)}{f(x_1) - f(x_0)} \right]$$

- Step 4 : Repeat procedure for $x_1 \& x_2$

eg →

Find a root of $x^3 - 2x - 5 = 0$ using Secant method. $x_0 = 2, x_1 = 3$

$$f(x_0) = 8 - 4 - 5 = -1$$

$$f(x_1) = 27 - 6 - 5 = 16$$

$$\frac{16}{3-2} = x$$

$$16 + 1 = -16$$

$$17x - 51 = -16$$

$$1^{st} \text{ Approx} \quad x_2 = \frac{2.0588}{2}$$

$f(x_2) \approx -0.3907$ and so on.

$$\frac{16.3907}{x - 0.9412} \approx 0.3907$$

$$17.414x - 25.853 = 0.3677$$

$$2^{nd} \text{ Approx } x_3 = 2.079$$

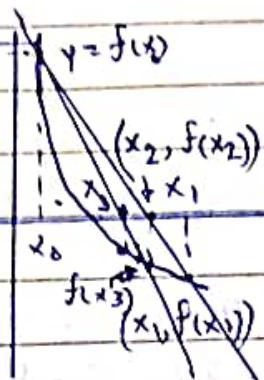
Criteria to stop $\rightarrow \epsilon = 0.001$

$$|x^* - x_1| < \epsilon$$

$$(2^{nd} \text{ stopping criteria}) |x_2 - x_1| < \epsilon$$

$$|x_3 - x_2| < \epsilon_2 = 0.0005$$

Regula - Falsi Method



Step 1 : Consider x_0 & x_1 as 2 approximations of initial root such that $f(x_1) \cdot f(x_0) < 0$

Repeat for step 2 of the Secant method

Step 4 : Check whether $f(x_0) f(x_1) < 0$ or $f(x_1) f(x_2) < 0$

Step 5 : Repeat

Eg : Find a root of $x^3 - 2x - 5$ using Regula - Falsi Method.

$$x_0 = 2$$

$$x_1 = 3$$

$$f(x_1) \cdot f(x_0) < 0$$

$$x_2 = 3 - \frac{16(0)}{16+1}$$

$$= 2.0588$$

$$\underset{\text{1st Approx}}{x_2 = 3 - \frac{16}{12} = 2.0588}$$

$$\therefore f(x_2) = -0.3907 < 0$$

$$f(x_1) \cdot f(x_2) < 0$$

$$\therefore x_3 = 2.0588 + \frac{(0.3907)}{(-0.3907)}$$

$$= 2.081$$

$$\underset{\text{2nd Approx}}{x_3 = 2.081}$$

Secant Method \rightarrow root can be found outside the domain of $f(x)$, unlike the other 2 methods.

Problem 28.1

2) i) $\cos x = x e^x$ $\Rightarrow \cos x - x e^x = 0$

~~one~~ $f(x) =$

$x e^x - \cos x = 0$

$x=0$ $f(0) = 1$ $\neq 0$ \therefore $x \neq 0$

ii) $2 \log x = 1.2$

$a=2$ $f(x) = 2 \log x - 1.2 < 0$

$b=3$ $f(b) = 2 \log 3 - 1.2 > 0$

$f(a) \cdot f(b) < 0$ \therefore 1 root

$x_1 = \frac{2 \cdot 3}{2} = 2.5$ \therefore 1 root

$f(x_1) < 0$

$\therefore f(x_1) \cdot f(b) < 0$

$x_2 = \frac{2.5 + 3}{2} = 2.75$

$f(x_2) > 0$

$\therefore f(x_1) \cdot f(x_2) < 0$

$x_3 = \frac{x_1 + x_2}{2} = 2.625$

$\therefore f(x_3) < 0$

~~so~~ $f(x_3) \cdot f(x_2) < 0$

$\therefore x_4 = \frac{x_2 + x_3}{2} = 2.6725$

Newton-Raphson Method

* function must be differentiable

$$y = f(x)$$

eqn of tangent

$$\text{slope} = \frac{dy}{dx} = f'(x) \quad |_{(x_0, y_0)}$$

$$\cancel{x=x_0} \quad \frac{y-y_0}{x-x_0} = f'(x)$$

$$y - y_0 = f'(x)(x - x_0)$$

$$y = y_0 + f'(x)(x - x_0)$$

Eqn of
tangent at
(x_0, y_0)

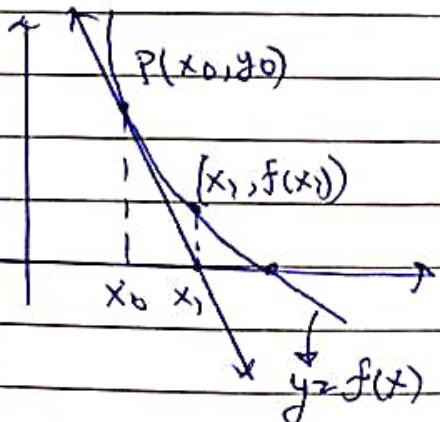
$$y = \underbrace{f(x_0)x}_{m} - \underbrace{f'(x_0)(x_0)}_{c} + y_0$$

$$\text{put } y=0$$

$$\cancel{y=0} \quad \frac{f'(x)x}{f'(x_0)x_0}$$

$$f'(x)x = f'(x)(x_0) + y_0$$

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



: General form is :

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

Q9 :

$$x^3 - 3x - 5 = 0$$

$$f(0) = -5$$

27

$$\frac{-9}{18} \\ -5$$

$$f(x) = x^3 - 3x - 5$$

$$f(1) = -7$$

13

$$f'(x) = 3x^2 - 3$$

$$f(2) = -3$$

13

$$x_{n+1} = x_n + \frac{f(x_n)}{f'(x_n)}$$

$$f(3) = 13$$

13

$$x_{n+1} = \frac{x_n^3 - 3x_n - 5}{3x_n^2 - 3}$$

$$x_0 = 2$$

$$x_1 = 2 + \frac{-3}{9}$$

$$2 + \frac{1}{3} = \frac{7}{3} = 2.33$$

~~$$f(7/3) = 0.6593$$~~

~~$$x_2 = \frac{7}{3} + \frac{0.6593}{4.33}$$~~

~~$$f'(7/3) = 4.33 \approx 13.2867$$~~

~~$$f(x_2) = 3.0075$$~~

~~$$f'(x_2)$$~~

~~$$x_3 = 2.33 + \frac{0.6593}{13.2867} = 2.3796$$~~

~~$$x_{n+1} = x_n + f'(x_n)$$~~

~~$$(x_{n+1} - x_n) \approx 0.0001$$~~

~~$$x_{n+1} - x_n = 0.0001$$~~

Rate of convergence

An iterative method is said to be of order p if
rate of convergence p if p is the largest +ve real
number for which there exist a finite constant c
is non-zero such that

$$|E_{k+1}| < c |E_k|^p$$

where $E_k = x_k - \varphi$

error in k^{th} iteration and c is called
as the asymptotic error constant.

a) c usually depends upon ~~domain~~ $f'(x)$ at $x =$

Problems 28.1

6) i) $x^3 - 3x + 1 = 0$

~~for~~ $f(x) = x^3 - 3x + 1$

$f'(x) = 3x^2 - 3$

$f(0) = 1$

$f(1) = -1$

$x_0 = 0.5$

$f(0.5) \approx 0.125$

~~f'~~

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

$x_1 = 0.33$

$$x_2 = x_1 + \frac{f(x_1)}{f'(x_1)} = 0.34718$$

$$x_3 = x_2 - \frac{f(x_2)}{f'(x_2)} = 0.34729$$

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

$$= 2.7935$$

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

$$= 2.7983$$

(1)

(2)

Rate of convergence

$$\rightarrow |x_n - \gamma| < \epsilon$$

$$\rightarrow |x_{n+1} - x_n| < \epsilon$$

$$\rightarrow |f(x_n)| < \epsilon$$

let γ be a root of $f(x) = 0$

Now,

$$\epsilon_k = x_k - \gamma$$

error

$x_k \rightarrow k^{\text{th}}$ iteration

let ϕ be the ~~the~~ largest real number such that

$$|\epsilon_{k+1}| \leq c |\epsilon_k|^\phi$$

where ϕ is called ratio of convergence

i) Secant

Method

$$\left\{ \begin{array}{l} x_{k+1} = x_k - \frac{f(x_k)(x_k - x_{k-1})}{f(x_k) - f(x_{k-1})} \end{array} \right\} - (1)$$

Assume γ be root of $f(x) = 0$

let

$$\epsilon_{k+1} = x_{k+1} - \gamma$$

$$\epsilon_k = x_k - \gamma$$

$$\epsilon_{k-1} = x_{k-1} - \gamma$$

Substitute value of ① x_{k+1}, x_k, x_{k-1}
 $f(\epsilon_k + \gamma)$.

$$\Leftrightarrow \epsilon_{k+1} + \gamma = \epsilon_k + \gamma - \frac{((\epsilon_k + \gamma) - (\epsilon_{k-1} + \gamma))}{f(\epsilon_k + \gamma) - f(\epsilon_{k-1} + \gamma)} \quad (2)$$

Using Taylor series

$$f(\epsilon_k + \gamma) = f(\gamma) + \epsilon_k f'(\gamma) + \frac{\epsilon_k^2}{2!} f''(\gamma) + \dots$$

$$f(\epsilon_{k-1} + \gamma) = f(\gamma) + \epsilon_{k-1} f'(\gamma) + \frac{\epsilon_{k-1}^2}{2!} f''(\gamma) \dots$$

$$\begin{aligned} & f(\epsilon_k + \gamma) - f(\epsilon_{k-1} + \gamma) \\ &= (\epsilon_k - \epsilon_{k-1}) f'(\gamma) + \frac{(\epsilon_k^2 - \epsilon_{k-1}^2) f''(\gamma)}{2!} \end{aligned} \quad (3)$$

Use ③ in ②

$$\begin{aligned} \epsilon_{k-1} + \gamma &= \epsilon_k + \gamma - \frac{(\epsilon_k - \epsilon_{k-1})(f(\gamma) + \epsilon_k f'(\gamma) + \frac{\epsilon_k^2 f''(\gamma)}{2!})}{(f'(\gamma) + \frac{(\epsilon_k + \epsilon_{k-1}) f''(\gamma)}{2!})} \\ &= \epsilon_k - \frac{(\epsilon_k - \epsilon_{k-1})(f(\gamma) + \epsilon_k f'(\gamma) + \frac{\epsilon_k^2 f''(\gamma)}{2!})}{f'(\gamma) + \frac{(\epsilon_k + \epsilon_{k-1}) f''(\gamma)}{2!}} \end{aligned}$$

$$\epsilon_{k+1} = \epsilon_k - \frac{(\epsilon_k - \epsilon_{k-1})(f(\gamma) + \frac{\epsilon_k^2 f''(\gamma)}{2!})}{f'(\gamma) + \frac{(\epsilon_k + \epsilon_{k-1}) f''(\gamma)}{2!}}$$

$$= \epsilon_k - \frac{f'(\gamma)(\epsilon_k + \frac{\epsilon_k^2 f''(\gamma)}{2!})}{f'(\gamma) \left(1 + \frac{(\epsilon_k + \epsilon_{k-1}) f''(\gamma)}{2 f'(\gamma)} \right)}$$

use binomial expansion of $(1+\alpha)^{-1}$

$$\varepsilon_k = \left[1 - \frac{1}{\alpha} (\varepsilon_{k+1} + \varepsilon_{k-1}) \frac{f''(\tau)}{f'(\tau)} + \dots \right] \left[\varepsilon_{k+1} + \frac{1}{\alpha} \frac{f''(\tau)}{f'(\tau)} \varepsilon_k^2 \right]$$

$$= \varepsilon_k = \varepsilon_{k+1} + \frac{1}{\alpha} \varepsilon_{k+1} \varepsilon_{k-1} \frac{f''(\tau)}{f'(\tau)}$$

$$|\varepsilon_{k+1}| = \frac{1}{\alpha} |\varepsilon_k| |\varepsilon_{k-1}| \frac{f''(\tau)}{f'(\tau)}$$

$$\frac{|\varepsilon_{k+1}|}{|\varepsilon_{k-1}|} \leq c |\varepsilon_k|^{\phi}$$

$$|\varepsilon_{k-1}| \simeq c |\varepsilon_k|^{\phi}$$

$$|\varepsilon_{k-1}| = c^{-1/\phi} |\varepsilon_k|^{1/\phi}$$

$$c |\varepsilon_k|^{\phi} = |\varepsilon_{k+1}|^{-1/\phi}$$

$$= \frac{1}{2} \left| \frac{f''(\tau)}{f'(\tau)} \right| c |\varepsilon_k| |\varepsilon_{k+1}|^{1/\phi}$$

$$= \frac{1}{2} \left| \frac{f''(\tau)}{f'(\tau)} \right|^{-1/\phi} c^{1+1/\phi}$$

$$\phi_B = 1 + \frac{1}{\phi}$$

$$\phi \simeq 1.42$$

$$\underline{f(\tau) = f(\tau) + \phi \varepsilon}$$

HW

calculate the rate of convergence for Regula-Falsi Method

- i) Obtain the value of $\sqrt[3]{25}$ correct for two decimal places

using Newton-Raphson

$$x = \sqrt[3]{25}$$

$$x^3 - 25$$

$$x^3 - 25 = 0$$

$$f(x) = x^3 - 25$$

$$f'(x) = 3x^2$$

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

$$\approx 2.925$$

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

$$\approx 2.924$$

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

$$\approx 2.924$$

$$\therefore \boxed{x = 2.92}$$

- Q) How many steps would you expect to carry out to solve:

i) $e^x - 3x = 0$ $a=0, b=1$

ii) $x^3 - 3x + 1 = 0$ $a=1, b=2$

correct to two decimal places

Hint $I \in (a_0, b_0)$ $I_1 \in (a_1, b_1) \rightarrow I_2 \in (a_2, b_2)$

[for bisection method]

$$x_1 = \frac{a_0 + b_0}{2}$$

$$|x_1 - z| \leq \frac{1}{2}(b_0 - a_0)$$

$$\therefore |x_2 - z| \leq \frac{1}{2}(b_1 - a_1) = \frac{1}{2^2}(b_0 - a_0)$$

$$\therefore |x_n - z| \leq \frac{1}{2^n}(b_0 - a_0) < 0.01$$

$\in \downarrow$

$$2^n > \frac{b_0 - a_0}{\epsilon} \quad \text{2 decimal places}$$

$$\left[n > \frac{\log(b_0 - a_0) - \log \epsilon}{\log 2} \right]$$

Ex 28.1

$$(12) \quad r = \sqrt{N}$$

$$r^2 = N$$

$$r^2 - N = 0$$

$$\therefore f(x) = x^2 - N$$

$$\text{P i) } f(x) = x^2 - 13$$

$$\{x_0 = 4\} \quad f(4) = 16 - 13 = 3$$

$$f'(x) = 2x$$

$$f'(4) = 8$$

$$x_1 = x_0 - \frac{f(x_0)}{f'(x_0)} = 4 - \frac{3}{8}$$

$$\approx 3.625$$

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

$$= 3.6056$$

$$x_3 = x_2 - \frac{f(x_2)}{f'(x_2)} = 3.6055 \approx 3.605$$

$$(15) \quad r = (28)^{-1/4}$$

$$\therefore \frac{1}{r^4} = (28)^{1/4}$$

$$\frac{1}{r^4} = 28$$

$$28r^4 - 1 = 0$$

$$f(x) = 28x^4 - 1$$

$$f'(x) = 112x^3$$

$$x_0 = 1/3$$

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

$$= 0.4959$$

$$x_2 = 0.4451$$

$$x_3 = 0.4350$$

$$x_4 =$$

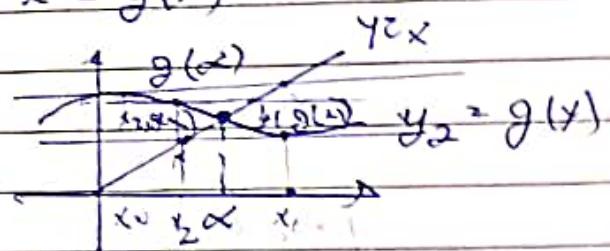


Fixed point iteration

Let g be a continuous function. A point $\alpha \in \mathbb{R}$ is said to be fixed point of g if $[g(\alpha) = \alpha]$

$$[f(x) = 0] \rightarrow x = g(x)$$

$$g: \mathbb{R} \rightarrow \mathbb{R}$$



Write $f(x) = 0 \Leftrightarrow x = g(x)$ for some $g(x)$ does
to initial approximation x_0 .

$$\boxed{x_{n+1} = g(x_n)}$$

e.g.: $f(x) = x^3 - 5x + 1 = 0$

$$f(2) = -1 < 0$$

$$f(3) = 13 > 0$$

$$x_0 = 2$$

$$f(x) = x^3 - 5x + 1 = 0$$

$$\therefore x = g(x)$$

$$x^3 = 5x - 1 \quad \text{or} \quad x^3 + 1 = 5x \quad \text{or} \quad x^3 - 4x + 1 = 0$$

$$x = (5x - 1)^{1/3} \quad \xrightarrow{\bullet} \quad x = \frac{x^3 + 1}{5} \quad \approx x$$

$$g_1(x)$$

$$g_2(x)$$

$$g_3(x)$$

$$x_1 = (5x_0 - 1)^{1/3}$$

$$x_1 = \frac{x_0^3 + 1}{5}$$

$$x_1 = x_0^3 - 4x_0^{2/3}$$

$$x_1 = 2.22$$

$$\approx 1.8$$

$$= 1$$

$$x_2 = (5x_1 - 1)^{1/3}$$

$$x_2 = \frac{x_1^3 + 1}{5}$$

$$x_2 = -2$$

$$x_2 = (5x_1 - 1)^{1/3}$$

$$x_2 = 1.364$$

$$\text{Not } 4/3$$

$$x_3 = (5x_2 - 1)^{1/3}$$

$$x_3 = 2.121$$

$$= 2$$

$$x_3 = (5x_2 - 1)^{1/3}$$

$$x_3 = 2.121$$

$$\text{discard}$$

$$x_4 = (5x_3 - 1)^{1/3}$$

$$x_4 = 2.1257$$

$$= 2$$

$$x_4 = (5x_3 - 1)^{1/3}$$

$$x_4 = 2.1257$$

$$= 2$$

$$x_4 = (5x_3 - 1)^{1/3}$$

$$x_4 = 2.1257$$

$$= 2$$

\Rightarrow How to pick g

Assume g is differentiable

$$\varepsilon_{k+1} = x_{0+1}^k - \gamma$$

$$\Rightarrow g(x_k) - g(\gamma)$$

$$\left\{ \frac{f(c) - f(1)}{c-1} = f'(c) \right\}$$

M-II

Finding

$$= (x_k - \gamma) g'(\alpha_k) \quad \{ \gamma < \alpha_k < x_k \}$$

$$= \varepsilon_k g'(\alpha_k)$$

or $g(\gamma)$

$$\Rightarrow [g(x_{k+1}) - g(\gamma)] g'(\alpha_k)$$

$$= (x_{k+2} - \gamma) g'(\alpha_k) g'(\alpha_{k+2})$$

$$\dots x_0 (x_0 - \gamma) g'(\alpha_0) g'(\alpha_1) \dots g'(\alpha_k)$$

$$|\varepsilon_{k+1}| \leq \varepsilon_0 c^{k+1}$$

$$|g'(x)| \leq c$$

if $c < 1$ then

then

$$|\varepsilon_{k+1}| \leq c^{k+1} \cdot \varepsilon_0$$

choose function $g(x)$ where $f(x) = 0$ can be written as $x = g(x)$ such that $\max |g'(x)| \leq c < 1$

Using same in previous example

$$g_1(x) = (5x-1)^{1/3}$$

$$g'_1(x) = \frac{5}{3(5x-1)^{2/3}} \text{ at } x=2 < 1$$

M-2

Finding g

$$f(x) = 0 \quad x = g(x)$$

$$g(x) = x + \alpha f(x)$$

$$(1) \quad x = x + \alpha f(x)$$

$$(2) \quad |g'(x)| < 1$$

$$|1 + \alpha f'(x)| < 1$$

Choose $\alpha \in \mathbb{R}$ in such a way that

e.g/

$$f(x) = x^2 - 5$$

$$x = x + \alpha f(x)$$

$$2x + \alpha(x^2 - 5) = (x+5)^2$$

$$|1 + \alpha f'(x)| < 1$$

$$|1 + \alpha(2x)| < 1$$

$$x_0 \in (2, 3)$$

$$x_0 \approx 2.5$$

$$|1 + 2\alpha x_0| < 1$$

$$|1 + 5\alpha| < 1$$

$$-1 < 1 + 5\alpha < 1$$

$$\left\{ \begin{array}{l} -2 < \alpha < 0 \\ \frac{-1}{5} < \alpha < \frac{1}{5} \end{array} \right\}$$

$$\text{let } \alpha = -0.1$$

$$g(x) = x + -0.1(x^2 - 5) \approx 0.9$$

$$-0.5x^2 + 0.9$$

Newton Raphson rate of convergence

$$E_{n+1} = x_{n+1} - \gamma$$

$$|E_{n+1}| \leq c |E_n|^p$$

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

$$E_{n+1} + \gamma = E_n + \gamma - \frac{f(E_n + \gamma)}{f'(E_n + \gamma)}$$

let f be twice differentiable

$$f(E_n + \gamma) = f(\gamma) + E_n f'(\gamma) + \frac{E_n^2 f''(\gamma)}{2!} \dots$$

$$f'(E_n + \gamma) = f'(\gamma) + E_n f''(\gamma) + \frac{E_n^2 f'''(\gamma)}{2!} \dots$$

$$E_{n+1} = E_n - \frac{f(\gamma) + E_n f'(\gamma) + \frac{E_n^2}{2!} f''(\gamma) + \dots}{f'(\gamma) + E_n f''(\gamma) + \frac{E_n^2}{2!} f'''(\gamma) + \dots}$$

$$= E_n - \frac{f(\gamma)}{f'(\gamma)} \left[E_n + \frac{E_n^2}{2!} \frac{f''(\gamma)}{f'(\gamma)} + \dots \right]$$

$$= E_n - \frac{f(\gamma)}{f'(\gamma)} \left[1 + \frac{E_n}{2!} \frac{f''(\gamma)}{f'(\gamma)} + \frac{E_n^2}{2!} \frac{f'''(\gamma)}{f'(\gamma)} + \dots \right]$$

$$= E_n - \frac{\cancel{E_n^3}}{2!} \frac{f(\gamma)}{f'(\gamma)} \left[E_n - \frac{E_n}{2!} \frac{f''(\gamma)}{f'(\gamma)} + \frac{E_n^2}{2!} \frac{f'''(\gamma)}{f'(\gamma)} - \frac{E_n^3}{2!} \frac{f''''(\gamma)}{f'(\gamma)^2} \right]$$

$$e_{n+1} \leq \left| \frac{e_n^2}{2} \frac{|f''(\gamma)|}{|2f'(\gamma)|} \right|$$

$$c = \frac{|f''(m)|}{|2f'(\gamma)|}$$

[P=2]

second

Newton-Raphson method has 2 order of convergence

$$x_{n+1} = \phi(n)$$

Using Newton's formula

$$x_{n+1} = g(x_n)$$

$$e_{n+1} = x_{n+1} - \gamma \quad \text{where } \begin{cases} f(\gamma) = 0 \\ g(\gamma) = \gamma \end{cases}$$

$$e_{n+1} + \gamma = g(e_n + \gamma)$$

$$e_{n+1} + \gamma = g(\gamma) + e_n g'(\gamma) + \frac{e_n^2}{2!} g''(\gamma)$$

$$e_{n+1} = e_n g'(\gamma) + \frac{e_n^2}{2!} g''(\gamma)$$

We know that if maximum of $|g'(x)| \leq c < 1$
then the sequence $\{x_n\}$ converges to γ .

$$|g(\gamma)| \leq 1$$

Now, case 1 : $g'(\gamma) \neq 0$ { Neglects higher order terms }

case 2 : $g'(\gamma) = 0$ but $g''(\gamma) \neq 0$

case 3 : $g'(\gamma) = 0, g''(\gamma) = 0 \text{ & } g'''(\gamma) \neq 0$

case 4: $g'(r) = 0, g''(r) = \dots, g^{(p-1)}(r) = 0$ but $g^{(p)}(r) \neq 0$

case 1: $|E_{n+1}| \leq |g(x)| |E_n|$

case 2: $g'(\tau) = 0$ but $g''(\tau) \neq 0$
 $|E_{n+1}| \leq |E_n|^2 \left| \frac{g''(\tau)}{2!} \right|$

case 3: $|E_{n+1}| \leq |E_n|^3 \left| \frac{g'''(\tau)}{3!} \right|$

$|E_{n+1}| \leq \left| \frac{g^{(p)}}{p!} \right| |E_n|^p$

System of Linear Equations

$$2x + 3y = 7$$

$$4x + 5y = 8$$

$$\begin{bmatrix} 2 & 3 \\ 4 & 5 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} 7 \\ 8 \end{bmatrix}$$

Gauss Crayen Method rule

$$x = \frac{\Delta_1}{\Delta_0} \quad \left| \Delta_1 = \begin{bmatrix} 7 & 3 \\ 8 & 5 \end{bmatrix} \right.$$

$$y = \frac{\Delta_2}{\Delta_0} \quad \left| \Delta_2 = \begin{bmatrix} 2 & 7 \\ 4 & 8 \end{bmatrix} \right.$$

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

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

$$a_{11}(x_1) + a_{12}(x_2) + \dots + a_{1n}(x_n) = b_1$$

$$a_{21}(x_1) + a_{22}(x_2) + \dots + a_{2n}(x_n) = b_2$$

$$\left[\begin{array}{cccc|c} a_{11} & a_{12} & \dots & a_{1n} & x_1 \\ a_{21} & a_{22} & \dots & a_{2n} & x_2 \\ a_{31} & a_{32} & \dots & a_{3n} & \vdots \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{m1} & a_{m2} & \dots & a_{mn} & x_n \end{array} \right] = \left[\begin{array}{c} b_1 \\ b_2 \\ \vdots \\ b_m \end{array} \right]$$

$$AX = b$$

m × n

r_{1,2} no of LI column vectors

$$r_1 + r_2 + \dots$$

r_{2,2} no of LI row vectorsr_{1,2} no of LI row vectorsdim(R²) = 2 → any elementbasis(R²) = {(0)(1)} → can be written as

linear combination of these two elements

$$\begin{bmatrix} 1 & 2 & 3 & 4 & 5 \\ 0 & 3 & 4 & 7 & 8 \end{bmatrix}$$

2 × 5

$$\begin{cases} r_2 = 1 \\ r_1 = 2 \end{cases}$$

No of LI column vector ~ No of LI

$$8 = 5 + 3x_1 + x_2$$

$$5x_1 + 3x_2 + x_3 =$$

$$5x_1 + 3x_2 + x_3$$

$$2x_1 + 4x_2 + 3x_3 =$$

$$10x_1 + 7x_2 + 5x_3 =$$

$$10x_1 + 7x_2 + 5x_3 =$$

$$x_1 + 2x_2 + x_3$$

Solutions of Linear System

i) A linear system of m equations in n -unknowns is of the form -

$$a_{11}x_1 + a_{12}x_2 + a_{13}x_3 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + a_{23}x_3 + \dots + a_{2n}x_n = b_2$$

$$a_{m1}x_1 + a_{m2}x_2 + a_{m3}x_3 + \dots + a_{mn}x_n = b_m$$

$a_{jk} \in \mathbb{R}$ $1 \leq j \leq m$, $1 \leq k \leq n \rightarrow$ coefficients, $b_j \in \mathbb{R}$
 \rightarrow constants A , x_1, x_2, \dots, x_n are unknown

An element $\underline{x} \in (x_1, x_n)$ is said to be a solution of ① if $A\underline{x} = b$

where

$$A = \begin{bmatrix} a_{11} & a_{12} & a_{13} & \dots & a_{1n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{m1} & a_{m2} & a_{m3} & \dots & a_{mn} \end{bmatrix}$$

$$b = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_m \end{bmatrix}$$

case I (Homogeneous System) The system ① is called homogeneous system of $b_1 = b_2 = b_3 = \dots = b_m = 0$

In this case, solution has at least 1 solution (zero), $x_1 = x_2 = \dots = x_n = 0$

Case 2: (non-homogeneous system).

The system is said non-homogeneous system if at-least one of b_i is non-zero.

In this case

Options =

→ may not have solution $\begin{cases} 2x+4y=5 \\ 4x+6y=8 \end{cases}$

→ may have unique solution $\begin{cases} y-x=0 \\ y-2x=0 \end{cases}$

→ may have infinite many solution

$$3x+y=5$$

$$10x+\frac{2}{3}y=25$$

[Ans]

Special Case:

Assume

$$a_{11} \neq 0$$

$$a_{22} \neq 0 \therefore a_{mn} \neq 0$$

$$a_{11}x_1 + a_{12}x_2 \cdots a_{1n}x_n = b_1$$

$$a_{22}x_2 + a_{23}x_3 \cdots a_{2n}x_n = b_2$$

$$a_{mn-1}x_{m-1} + a_{mn}x_n = b_{m-1}$$

$$a_{mn}x_n = b_m$$

By back substitution,

$$x_n = \frac{b_m}{a_{mn}}$$

$$a_{mn}$$

$$x_{m-1} = \frac{1}{a_{m-1}} \left(b_{m-1} - \frac{b_m}{a_{mn}} a_{m-1} \right)$$

$$\textcircled{1} \rightarrow x_1 = a_{11} - a_{12}x_2 - \dots - a_{1m-1}x_{m-1}$$

eg Consider the linear system,

$$x_1 - x_2 + x_3 = 0$$

$$-x_1 + x_2 - x_3 = 0 \quad \textcircled{1}$$

$$10x_2 + 25x_3 = 90 \quad \textcircled{2}$$

$$20x_1 + 10x_2 = 80 \quad \textcircled{3}$$

$$-20x_1 + 20x_2 - 20x_3 = 0$$

Step ①:
Eliminate x_1 from ② & ③

$$30x_2 - 20x_3 = 80 \quad \textcircled{4} \text{ eqn}$$

$$\begin{pmatrix} 3x_2 - 2x_3 & 30 \\ 10x_2 + 25x_3 & 90 \end{pmatrix} \xrightarrow{\text{Row 2} \times 2}$$

$$\frac{75x_2}{2} + 10 \left(\frac{2x_3}{2} \right) = 100 \quad | \cdot 2$$

$$\frac{95x_2}{2} = 190$$

$$x_2 = 4$$

$$x_3 = \frac{50}{25} = 2$$

$$\begin{bmatrix} x_1 = ? \\ x_2 = 4 \\ x_3 = 2 \end{bmatrix}$$

Gauss Elimination

$$x_1 - x_2 + x_3 = 0 \quad | \cdot 1 \quad (1)$$

$$10x_2 + 25x_3 = 90 \quad | \cdot 2 \quad (2)$$

$$20x_1 + 10x_2 = 80 \quad | \cdot 3 \quad (3)$$

Eliminate x_1 from (3)

$$x_1 - x_2 + x_3 = 0 \quad | \cdot 1 \quad (4)$$

$$10x_2 + 25x_3 = 90 \quad | \cdot 2 \quad (5)$$

$$30x_2 - 20x_3 = 80 \quad | \cdot 3 \quad (6)$$

Eliminate x_2 from (4) & (6)

$$x_1 - x_2 + x_3 = 0$$

$$10x_2 + 25x_3 = 90$$

$$95x_3 = 190$$

Using back-substitution

$$x_3 = 2$$

$$x_2 = \frac{190 - 90}{50 \times 25} = 4$$

$$x_1 = 2$$

Consider the augmented matrix

$$[A|b] = \left[\begin{array}{ccc|c} 1 & -1 & 1 & 0 \\ -1 & 10 & -1 & 0 \\ 0 & 10 & 25 & 90 \\ 0 & 20 & 10 & 80 \end{array} \right]$$

$$R_2 \rightarrow R_2 + R_1$$

$$R_4 \rightarrow R_4 - 20 \times R_1$$

$$\left[\begin{array}{ccc|c} 1 & -1 & 1 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 10 & 25 & 90 \\ 0 & 30 & -20 & 80 \end{array} \right]$$

$$R_2 \leftrightarrow R_4$$

$$\left[\begin{array}{ccc|c} 1 & -1 & 1 & 0 \\ 0 & 30 & -20 & 80 \\ 0 & 10 & 25 & 90 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$$R_3 \rightarrow R_3 - R_2/3$$

$$x_1 - 4 + 2 = 0$$

$$x_1 = 2$$

$$\frac{95}{3} x_3 = 190/2$$

$$x_3 = 2$$

$$30 x_2 - 20 \cdot 2 = 80$$

$$x_2 = 4$$

$$\left[\begin{array}{ccc|c} 1 & -1 & 1 & 0 \\ 0 & 30 & -20 & 80 \\ 0 & 0 & 95/3 & 190/3 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

$$\left[\begin{array}{c} x_1 \\ x_2 \\ x_3 \\ x_4 \end{array} \right] = \left[\begin{array}{c} 2 \\ 4 \\ 2 \\ 0 \end{array} \right]$$

Such a matrix is called,
Row-Echelon form

$$A_1 x = b_1$$

$$\left[\begin{array}{ccc|c} 1 & -1 & 1 & x_1 \\ 0 & 10 & 25 & x_2 \\ 0 & 0 & -95 & x_3 \\ 0 & 0 & 0 & 0 \end{array} \right]$$

A matrix A is said to be in row-Echelon form if the following conditions are satisfied.

- i) If i^{th} row contains all zeros it is true for all subsequent rows.
- ii) If i^{th} and $(i+1)^{\text{th}}$ row are both non-zero rows initial non-zero entry of $(i+1)^{\text{th}}$ row appears in a later column than that of the i^{th} row.
- iii) Rows containing all zeros occur only after all non-zero rows.

Def: If a row is non-zero in a matrix Then the first non-zero element in that row is called pivot.

Elementary Row Operations

- (i) Interchange of any two rows ($R_i \leftrightarrow R_j$)

(a) Addition of a scalar multiple λ to a row or another row: $R_1 \leftrightarrow R_1 + \lambda R_2$

(b) Multiplication / Division of any row by non-zero scalar $\lambda \neq 0$ ($\lambda \neq 0$)

Ex/ Find rank of matrix $A = \begin{bmatrix} 1 & 3 & 5 \\ 2 & -1 & 4 \\ -2 & 3 & 2 \end{bmatrix}$
by reducing it to its row-echelon form.

$$\begin{bmatrix} 1 & 3 & 5 \\ 2 & -1 & 4 \\ -2 & 3 & 2 \end{bmatrix} \xrightarrow[R_2 \rightarrow R_2 - 2R_1]{R_3 \rightarrow R_3 + 2R_1} \begin{bmatrix} 1 & 3 & 5 \\ 0 & -7 & -6 \\ 0 & 9 & 12 \end{bmatrix}$$

$$\xrightarrow[R_2 \rightarrow R_2 + 2R_3]{R_3 \rightarrow R_3 - 3R_1} \begin{bmatrix} 1 & 3 & 5 \\ 0 & -7 & -6 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\text{Rank} = 2$$

Ex/ $\mathbb{R}^3 = \left\{ \begin{pmatrix} 1 \\ 3 \\ 5 \end{pmatrix} \right\} \oplus \text{...}$ along with R_1

$$x_1 = \begin{pmatrix} 1 \\ 3 \\ 5 \end{pmatrix}^T$$

$$x_2 = \begin{pmatrix} 2 \\ -1 \\ 4 \end{pmatrix}^T$$

$$x_{3,4} = \begin{pmatrix} -2 \\ 3 \\ 2 \end{pmatrix}^T$$

Check whether they are linearly independent?

Linearly dependent (not spanned)?

Steps :

* reduce to row-Echelon

* look at row-Echelon form
and look at no. of non-zero rows

$$A = \begin{pmatrix} 1 & 3 & 5 \\ 2 & -1 & 4 \\ -2 & 8 & 2 \end{pmatrix}$$

$$R_2 \rightarrow R_2 - 2R_1$$

$$\begin{pmatrix} 1 & 3 & 5 \\ 0 & -7 & -6 \\ 0 & 14 & 12 \end{pmatrix} \quad R_3 \rightarrow R_3 + 2R_2$$

$$E = \begin{pmatrix} 1 & 3 & 5 \\ 0 & -7 & -6 \\ 0 & 0 & 0 \end{pmatrix}$$

$$R_3 \rightarrow R_3 + 2R_2$$

If no. of non zero rows \neq no. of rows of matrix then eqns are linearly dependent.

else, they are independent.

Gauss Elimination Method

This is an algorithm to transform a system of linear equations into its row-echelon form and then back-substitution method is applied to solve.

- Steps :
- ① Step up augmented matrix
 - ② Transform to its row-echelon form
 - ③ Use back-substitution to solve

eg: Use Gauss Elimination method to solve:

$$6x_2 + 13x_3 = 61$$

$$6x_1 - 8x_3 = -38$$

$$13x_1 - 8x_2 = 79$$

$$\left[\begin{array}{ccc|c} 0 & 6 & 13 & x_1 \\ 6 & 0 & -8 & x_2 \\ 13 & -8 & 0 & x_3 \end{array} \right] = \left[\begin{array}{c} 61 \\ -38 \\ 79 \end{array} \right]$$

$$\left[\begin{array}{ccc|c} 0 & 6 & 13 & 61 \\ 6 & 0 & -8 & -38 \\ 13 & -8 & 0 & 79 \end{array} \right]$$

$$R_3 \leftrightarrow R_1$$

$$\left[\begin{array}{ccc|c} 13 & -8 & 0 & 79 \\ 6 & 0 & -8 & -38 \\ 0 & 6 & 13 & 61 \end{array} \right]$$

$$6x_2 + 13x_3 = 61$$

$$R_2 \leftrightarrow R_2 - \frac{6}{13} R_1$$

$$\begin{aligned} & \frac{494}{474} \\ & \frac{968}{268} \\ & \frac{38}{114} \end{aligned}$$

$$\left[\begin{array}{ccc|c} 13 & -8 & 0 & 79 \\ 0 & 0 & 48/13 & 208/13 \\ 0 & 6 & 13 & 61 \end{array} \right]$$

$$\begin{aligned} & \frac{2}{79} \\ & \frac{3}{237} \\ & \frac{4}{24} \\ & \frac{237+61}{481} \end{aligned}$$

$$R_2 \leftrightarrow R_3$$

$$\left[\begin{array}{ccc|c} 13 & -8 & 0 & 79 \\ 0 & 6 & 13 & 61 \\ 0 & 0 & 48/13 & 228/13 \end{array} \right]$$

$$R_2 \rightarrow R_2 + \frac{3}{4} R_1$$

$$\left[\begin{array}{ccc|c} 13 & -8 & 0 & 79 \\ 0 & 6 & 13 & 481/13 \\ 0 & 0 & 48/13 & 288/13 \end{array} \right]$$

$$\left[\begin{array}{ccc|c} 0 & 6 & 13 & 61 \\ 6 & 0 & -8 & -38 \\ 13 & -8 & 0 & 79 \end{array} \right]$$

$$\left[\begin{array}{ccc|c} 13 & -8 & 0 & 79 \\ 6 & 0 & -8 & -38 \\ 0 & 6 & 13 & 61 \end{array} \right]$$

$$R_2 \rightarrow R_2 - \frac{6}{13} R_1$$

$$\left[\begin{array}{ccc|c} 13 & -8 & 0 & 79 \\ 0 & \frac{-48}{13} & -8 & -\frac{968}{13} \\ 0 & 6 & 13 & 61 \end{array} \right]$$

$$0 + \frac{6 \times 8}{13} = \frac{38}{13}$$

$$-38 - \frac{79 \times 6}{13} = \frac{38}{13}$$

$$-38 - \frac{474}{13} = \frac{494}{13}$$

$$R_3 \rightarrow R_3 - \left(\frac{13 \times 6}{48} \right) R_2$$

$$\left[\begin{array}{ccc|c} 13 & -8 & 0 & 79 \\ 0 & \frac{48}{13} & -8 & -\frac{968}{13} \\ 0 & 0 & 26 & 182 \end{array} \right]$$

$$13 + \left(\frac{13 \times 6}{48} \right) 8 = 61 + \left(\frac{13 \times 6}{48} \right) 8$$

$$26 \times 3 = 182$$

$$x_3 = 7$$

$$x_2 = \left(\frac{-968 + 8 \times 7}{13} \right) \frac{13}{48} = \frac{728}{240} = \frac{48}{15}$$

$$= -\frac{240}{48} = -5$$

$$13x_1 + 40 = 79 \quad x_1 = 39/13 = 3$$

$$AX = b$$

$$A\bar{x}_1 = \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \Rightarrow A^{-1} \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} \quad \left. \begin{array}{l} \text{Gauss} \\ \text{Jordan} \end{array} \right\}$$

$$A\bar{x}_2 = \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} = A^{-1} \begin{pmatrix} 0 \\ 1 \\ 0 \end{pmatrix} \quad \text{Method}$$

$$A\bar{x}_3 = \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} = A^{-1} \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$$

$$\bar{x}_1 = \begin{pmatrix} x_{11} \\ x_{12} \\ x_{13} \end{pmatrix}$$

$$[A\bar{x}_1 \ A\bar{x}_2 \ A\bar{x}_3]^{-1} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{pmatrix} \Rightarrow I$$

$$\begin{vmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{vmatrix} \xrightarrow{\text{Row operations}} \begin{vmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{vmatrix} \xrightarrow{\text{Row operations}} \begin{vmatrix} 1 & 0 & 0 & 1 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{vmatrix}$$

A^{-1}

~~eg/~~

$$A = \begin{bmatrix} -2 & 4 & -1 \\ -2 & 0 & 3 \\ 7 & -12 & 2 \end{bmatrix}$$

$$\det(A) \neq 0$$

convert this to

identity matrix

$$(A | I) = \left[\begin{array}{ccc|ccc} -2 & 4 & -1 & 1 & 0 & 0 \\ -2 & 0 & 3 & 0 & 1 & 0 \\ 7 & -12 & 2 & 0 & 0 & 1 \end{array} \right]$$

$\xrightarrow{\text{ } (I | X)}$

$\xrightarrow{\text{this becomes inverse}}$

$$\left[\begin{array}{ccc|ccc} -2 & 4 & -1 & 1 & 0 & 0 \\ 0 & -1 & 1 & -1 & 1 & 0 \\ 0 & 2 & -3/2 & 7/2 & 0 & 1 \end{array} \right]$$

$$R_3 \rightarrow R_3 + R_2$$

$$\left[\begin{array}{ccc|ccc} -2 & 4 & -1 & 1 & 0 & 0 \\ 0 & -1 & 1 & -1 & 1 & 0 \\ 0 & 0 & 1/2 & 3/2 & 2 & 1 \end{array} \right]$$

$$R_1 \rightarrow -1/2 R_1$$

$$R_2 \rightarrow -R_2$$

$$R_3 \rightarrow 2R_3$$

$$R_2 \rightarrow R_2 - 2R_1$$

$$\left[\begin{array}{ccc|ccc} 1 & -2 & 1/2 & -1/2 & 0 & 0 \\ 0 & 1 & -1 & 1 & -1 & 0 \\ 0 & 0 & 0 & 3 & 4 & 1 \end{array} \right]$$

Prob

6
6)

2
8

(A) b)

$$R_1 \rightarrow R_1 + 2R_2$$

$$-1/2 + 2 \cdot 1 \\ 0 + (-2)$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & -3/2 & 3/2 & -2 & 0 \\ 0 & 1 & -1 & 1 & -1 & 0 \\ 0 & 0 & 1 & 3 & 4 & 2 \end{array} \right]$$

$$R_1 \rightarrow R_1 + 3/2 R_3$$

$$R_2 \rightarrow R_2 + R_3$$

$$\left[\begin{array}{ccc|ccc} 1 & 0 & 0 & 6 & 4 & 3 \\ 0 & 1 & 0 & 4 & 3 & 2 \\ 0 & 0 & 1 & 3 & 4 & 2 \end{array} \right]$$

$$3/2 + 3/2 \cdot 3$$

$$3/2 + 9/2$$

$$-2 + 3/2 \cdot 4^2$$

$$3/2 \cdot 0 + 3/2 \cdot 2$$

$$1 + 1$$

$$-1 + 1$$

$$A^{-1} \left[\begin{array}{ccc} 6 & 4 & 3 \\ 4 & 3 & 2 \\ 3 & 4 & 2 \end{array} \right]$$

LU - Decomposition Method

→ factorization method

→ triangulation method

In this method, we decompose the coefficient matrix A is $A = LU$ & L is lower triangular matrix U is upper triangular matrix of the form.

$$L = \begin{Bmatrix} l_{11} \cdot 0 \dots 0 \\ 0_{21} l_{22} 0 \dots 0 \\ 0_{31} l_{32} l_{33} 0 \dots 0 \\ \vdots l_{m1} l_{m2} \dots l_{mn} \end{Bmatrix}$$

$$U = \begin{pmatrix} u_{11} & u_{12} & \dots & (0) u_{1n} \\ 0 & u_{22} & \dots & 0 & u_{2n} \\ 0 & 0 & u_{33} & \dots & u_{3n} \\ \vdots & 0 & 0 & \dots & 0 & u_{nn} \end{pmatrix}$$

Doolittle's Method

In this method we choose $l_{11} = l_{22} = \dots = l_{nn} = 1$
 $l_{ij} = 1 \quad \forall 1 \leq i \leq n$

Gaussian Method

In this method we choose $u_1 = u_{22} = \dots = u_{nn}$
 $u_{ii} = 1 \quad \forall 1 \leq i \leq n$

Choleski's Method

In this method we choose $U = L^T$

(Q) Solve system of linear equations by using
Doolittle Method:

$$\begin{aligned} 1.8x_1 + 2.6x_2 &= 13.2 \\ 0.36x_1 + 3.782x_2 &= 12.24 \end{aligned} \quad \left. \begin{array}{l} \\ \end{array} \right\} \textcircled{1}$$

Steps

① Write eqn as $Ax = b$

$$A = \begin{pmatrix} 1.8 & 2.6 \\ 0.36 & 3.782 \end{pmatrix}$$

$$b = \begin{pmatrix} 13.2 \\ 12.24 \end{pmatrix}$$

$$A = LU$$

$$L = \begin{pmatrix} 1 & 0 \\ l_{21} & 1 \end{pmatrix}$$

$$U = \begin{pmatrix} u_{11} & u_{12} \\ 0 & u_{22} \end{pmatrix}$$

$$LU = \begin{pmatrix} u_{11} + 0 & u_{12} + 0 \\ l_{21}u_{11} & l_{21}u_{12} + u_{22} \end{pmatrix}$$

$$= \begin{pmatrix} 1.8 & 2.6 \\ 0.36 & 3.782 \end{pmatrix}$$

$$u_{11} = 1.8$$

$$u_{12} = 2.6$$

$$l_{21} = \frac{0.36}{1.8} = \frac{36 \times 10^{-2}}{18 \times 10^{-2}} = 0.2$$

$$u_{22} = 3.782$$

Step 2 :

$$Ux = z$$

$$Ax = b$$

$$LUx = b$$

$$Lz = b$$

$$\left(\begin{matrix} z \\ z \end{matrix} \right) \left[\begin{array}{cc|c} 1 & 0 & 13.2 \\ 0.2 & 1 & 12.24 \end{array} \right] \quad \left(\begin{matrix} 1 & 0 \\ 0.2 & 1 \end{matrix} \right) \left(\begin{matrix} z_1 \\ z_2 \end{matrix} \right) = \left(\begin{matrix} 13.2 \\ 12.24 \end{matrix} \right)$$

$$R_2 \leftrightarrow R_1$$

$$\left(\begin{matrix} z_1 \\ 0.2z_1 + z_2 \end{matrix} \right) = \left(\begin{matrix} 13.2 \\ 12.24 \end{matrix} \right)$$

$$\left\{ \begin{array}{l} z_1 = 13.2 \\ z_2 = 9.6 \end{array} \right.$$

$$\left[\begin{array}{cc|c} 0.2 & 1 & 12.24 \\ 1 & 0 & 13.2 \end{array} \right]$$

$$R_2 \rightarrow R_2 - 0.2R_1$$

$$Ux = z$$

$$\left[\begin{array}{cc|c} 0.2 & 1 & 12.24 \\ 0.2 & 0 & 2.64 \end{array} \right]$$

$$R_2 \rightarrow R_2 - R_1$$

$$\left(\begin{array}{cc|c} 1.8 & 2.6 & x_1 \\ 0 & 3.202 & x_2 \end{array} \right) = \left(\begin{matrix} 13.2 \\ 9.6 \end{matrix} \right)$$

$$\left[\begin{array}{cc|c} 0.2 & 1 & 12.24 \\ 0 & 0 & 2.64 \end{array} \right]$$

~~last step~~

$$\left\{ \begin{array}{l} 1.8x_1 + 2.6x_2 = 13.2 \\ 3.202x_2 = 9.6 \end{array} \right. \quad \left\{ \begin{array}{l} x_2 = 2.998 \\ x_1 = 3.003 \end{array} \right.$$

SOL

SOL

$$\therefore x_1 = 3.003 \quad x_2 = 2.998$$

Summary for Doolittle / Crout Method

- {
 (i) Write $A = LU$
 (ii) Solve $L \otimes Z = b$ for Z
 (iii) $Ux = Z$ for x
}

Solve system of linear eqns using Crout's M.

$$3x_1 + 2x_2 = 18$$

$$18x_1 + 17x_2 = 123$$

$$Ax = b$$

$$A = \begin{pmatrix} 3 & 2 \\ 18 & 17 \end{pmatrix}$$

$$b = \begin{pmatrix} 18 \\ 123 \end{pmatrix}$$

$$A = LU$$

$$L = \begin{pmatrix} l_{11} & 0 \\ l_{21} & l_{22} \end{pmatrix}$$

$$U = \begin{pmatrix} 1 & u_{12} \\ 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} 3 & 2 \\ 18 & 17 \end{pmatrix} = LU = \begin{pmatrix} l_{11} + 0 & l_{11} \cdot u_{12} \\ l_{21} & l_{21} \cdot u_{12} + l_{22} \end{pmatrix}$$

$$l_{11} = 3$$

$$l_{21} = 18$$

$$u_{12} = 2/3$$

$$18 \cdot 2/3 + l_{22} = 17 \Rightarrow l_{22} = 5$$

$$L = \begin{pmatrix} 3 & 0 & 0 \\ 0 & 18 & 5 \end{pmatrix} \quad U = \begin{pmatrix} 1 & 2/3 \\ 0 & 1 \end{pmatrix}$$

$$Ax = b$$

$$\underbrace{LUx}_{=Z} = b$$

$$Lz = b$$

$$\begin{pmatrix} 3 & 0 \\ 18 & 5 \end{pmatrix} \begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \begin{pmatrix} 18 \\ 123 \end{pmatrix}$$

$$3z_1 = 18 \Rightarrow$$

$$18z_1 + 5z_2 = 123$$

$$\begin{cases} z_1 = 6 \\ z_2 = 3 \end{cases}$$

$$Ux = Z$$

$$\begin{pmatrix} 2/3 & 1 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 6 \\ 3 \end{pmatrix}$$

$$x_1 + \frac{2}{3}x_2 = 6$$

$$x_1 + 2 = 6$$

$$x_2 = 3$$

$$x_1 = 4$$

$$x_1 = \frac{52}{3} = 17.33$$

$$x_1 = 17.33$$

$$x_1 = -17$$

(Q)

Solve the system of equation by using Choleski method.

$$4x_1 + 6x_2 + 8x_3 = 0$$

$$6x_1 + 34x_2 + 52x_3 = -160$$

$$8x_1 + 52x_2 + 129x_3 = -452$$

$$A = \begin{pmatrix} l_{11} & 0 & 0 \\ l_{21} & l_{22} & 0 \\ l_{31} & l_{32} & l_{33} \end{pmatrix} \begin{pmatrix} 0 & l_{12} & l_{21} & l_{31} \\ 0 & 0 & l_{22} & l_{32} \\ 0 & 0 & 0 & l_{33} \end{pmatrix}$$

$$\therefore \begin{pmatrix} l_{11}^2 & l_{11}l_{21} & l_{11}l_{31} \\ l_{21}l_{11} & l_{21}^2 + l_{22}^2 & l_{21}l_{31} + l_{22}l_{32} \\ l_{31}l_{11} & l_{31}l_{21} + l_{32}l_{22} & l_{31}^2 + l_{32}^2 + l_{33}^2 \end{pmatrix}$$

$$= \begin{pmatrix} 4 & 6 & 8 \\ 6 & 34 & 52 \\ 8 & 52 & 129 \end{pmatrix}$$

$$L = \begin{pmatrix} 2 & 0 & 0 \\ 3 & 5 & 0 \\ 4 & 8 & 7 \end{pmatrix}$$

$$L = \begin{pmatrix} 2 & 0 & 0 \\ 3 & 5 & 0 \\ 4 & 8 & 7 \end{pmatrix}$$

$$\left\{ \begin{array}{l} Lz = b \\ LTx = z \end{array} \right\}$$

Note : Choleski's Method is not possible if

- (1) a_{nn} is non-negative
- (2) Matrix is not symmetric
- (3) Principle matrix Δ is negative

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$$\begin{aligned} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 \dots + a_{1n}x_n &= b_1 \quad (i) \\ a_{21}x_1 + a_{22}x_2 \dots + a_{2n}x_n &= b_2 \quad (ii) \\ a_{n1}x_1 + a_{n2}x_2 \dots + a_{nn}x_n &= b_n \quad (iii) \end{aligned}$$

$$x_1 = \frac{1}{a_{11}} (b_1 - a_{12}x_2 - \dots - a_{1n}x_n)$$

$$x_2 = \frac{1}{a_{22}} (b_2 - a_{21}x_1 - a_{23}x_3 - \dots - a_{2n}x_n)$$

$$x_n = \frac{1}{a_{nn}} (b_n - a_{n1}x_1 - a_{n2}x_2 - \dots - a_{n-1}x_{n-1})$$

iterative formula

$$x_1^{(k+1)} = \frac{1}{a_{11}} (b_1 - a_{12}x_2^k - a_{13}x_3^k - \dots - a_{1n}x_n^k)$$

$$x_n^{(k+1)} = \frac{1}{a_{nn}} (b_n - a_{n1}x_1^k - a_{n2}x_2^k - \dots - a_{n-1}x_{n-1}^k)$$

$$\bar{x}_0 = (x_1^0, x_2^0, x_3^0, \dots, x_n^0)$$

$$\begin{aligned} \bar{x}_1 &= (x_1^1, x_2^1, x_3^1, \dots, x_n^1) \\ \bar{x}_2 &= (x_1^2, x_2^2, x_3^2, \dots, x_n^2) \end{aligned}$$

This Method is called Gauss-Jacobi

Iterative Method

$$x_{k+1} = \phi(x_k)$$

Consider the linear system,

$Ax = b$ —① where A is $n \times n$ matrix whose diagonal elements are non-zero. Aim is to write ① in the form

$$x^{k+1} = Hx^k + c \quad k = 0, 1, 2, \dots —②$$

for some $n \times n$ matrix H & some vector c

→ Gauss Jordan

→ Gauss Seidel

Here H is the called the iterative matrix.

Remark

The iterative formula ② converges to solution of ① if & only if $\rho(H) < 1$

~~$\rho(H) \rightarrow$~~ $\rho(H) \rightarrow$ spectral radius of H

$$\rho(H) = \max \{ |\lambda| : \lambda \text{ is eigenvalue of } H \}$$

Note A suitable initial approximation for ② can be taken as $x_i = \frac{b_i}{a_{ii}}$ $1 \leq i \leq n$

Stopping criterion: A stopping criterion can be chosen as

$$|x_i^{k+1} - x_i^k| < \epsilon \quad \forall i \in n$$

Gauss Jacobi

(For both methods)

Diagonal Dominance

Consider the linear system

(in Matrix form)

$$a_{11}x_1 + a_{12}x_2 + \dots + a_{1n}x_n = b_1$$

$$a_{21}x_1 + a_{22}x_2 + \dots + a_{2n}x_n = b_2$$

:

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

with the assumption that $a_{ii} \neq 0 \quad \forall i \in \{1, 2, \dots, n\}$

Now consider iterative formula,

$$x_1^{k+1} = \frac{1}{a_{11}} (b_1 - (a_{12}x_2^k + a_{13}x_3^k + \dots))$$

$$x_2^{k+1} = \frac{1}{a_{22}} (b_2 - (a_{21}x_1^k + a_{23}x_3^k - a_{2n}x_n^k))$$

$$\vdots$$

$$x_n^{k+1} = \frac{1}{a_{nn}} (b_n - (a_{n1}x_1^k + \dots + a_{nn}x_n^k))$$

The iterative formula can be written as

$$x^{k+1} = Hx^k + c$$

$$d = x(u + v + \dots)$$

$$\text{where } H = I - (a_{11}^{-1}a_{12}, a_{11}^{-1}a_{13}, \dots)$$

$$- a_{21}^{-1}a_{22}, \dots$$

$$d' = [x(u + v + \dots)]$$

$$(D - H) \bar{x} + [R] = [x(u + v + \dots)]$$

$$c_2 \begin{pmatrix} b_1/c_{11} \\ b_2/c_{22} \\ b_3/c_{33} \\ \vdots \end{pmatrix}$$

Let ① can be written as

$$Ax = b$$

where $b = b_1 + b_2 + \dots + b_n$

$$A = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix}$$

$$(x_1 + x_2 + \dots + x_n) - b = 0$$

$$\begin{matrix} L+D+U \\ (x_1 + x_2 + \dots + x_n) \end{matrix} = \begin{pmatrix} 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \end{pmatrix} \begin{pmatrix} a_{11} & 0 & 0 & \dots & 0 \\ a_{21} & a_{22} & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ a_{n1} & a_{n2} & a_{n3} & \dots & 0 \end{pmatrix} \begin{pmatrix} 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \end{pmatrix}$$

$$(x_1 + x_2 + \dots + x_n) + \begin{pmatrix} 0 & a_{12} & \dots & a_{1n} \\ 0 & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & 0 \end{pmatrix}$$

To invert a_{ii} we have to eliminate a_{ij} for $j \neq i$

$$(L+D+U)x = b$$

D^{-1} $\rightarrow (L+U)x + b = H$ where
 (D) is invertible since $a_{ii} \neq 0$

$$\lambda = D^{-1} \underbrace{[(L+U)x]}_{H} + D^{-1}b$$

$$H = [x_{i+1}] = [x_i] + D^{-1}(b - H)$$

Thus the iterative formula, (Gauss-Siedel)

$$x^{k+1} = -D^{-1}(L+U)x^k + D^{-1}b \quad k=0, 1, 2, \dots$$

program to solve

$$a_1x_1 + b_1x_2 + c_1x_3 = d_1$$

$$(1) \quad a_2x_1 + b_2x_2 + c_2x_3 = d_2 \text{ left } \rightarrow$$

$$a_3x_1 + b_3x_2 + c_3x_3 = d_3 \text{ right}$$

to solve for x_1, x_2, x_3

$x_1^{(0)}, x_2^{(0)}, x_3^{(0)} \rightarrow$ initial guess

$$\text{id. no. of } x_1^{(0)} = \text{left} \left(d_1 - b_2x_2^{(0)} - c_1x_3^{(0)} \right) \text{ right}$$



$$x_2^{(1)} = \frac{1}{a_2} \left(d_2 - b_2x_1^{(0)} - c_2x_3^{(0)} \right) \text{ put this value}$$

$$x_3^{(1)} = \frac{1}{a_3} \left(d_3 - a_3x_1^{(0)} - b_3x_2^{(1)} \right)$$

$$\vdots = \vdots$$

$$(L+D)x = b$$

$$(L+D)x^k = Ux^k + b^k, [x^{k+1}] = Q[L^{-1}](b^k - Ux^k)$$

but a_{ii} don't work for iterative elimination method

Let's $x^{k+1} = Hx^k + c$ minimum to

$$\text{where } H = - (L+D)^{-1}U$$

x \rightarrow $L+D$ is invertible

iff $\exists c \in \mathbb{C} \Rightarrow ((L+D)^{-1})c = 0$ but not if

Definition — A $n \times n$ matrix is said to be diagonally dominant if

$$\sum_{j \neq i} |a_{ij}| < |a_{ii}| \quad 1 \leq i \leq n$$

$$\text{eg} \quad \begin{pmatrix} 13 & 2 & 0 \\ 4 & 15 & 3 \\ 7 & 9 & 20 \end{pmatrix} \quad \begin{array}{l} 13 > 2+0 \\ 15 > 4+3 \\ 20 > 9+7 \end{array}$$

Note $\rho(H) < 1$ is automatically satisfied if the matrix A is diagonally dominant.

Rate of convergence

If the iterative formula converges ($\rho(A) < 1$) then

$\alpha = -\log_{10}(\rho(H))$ is called rate/order of convergence.

e.g., Write the iterative formula for Gauss Jacobi method for the system

$$\begin{aligned} (i) \quad & 7x_1 - 15x_2 - 21x_3 = -2 \\ (ii) \quad & 7x_1 - x_2 - 5x_3 = -3 \end{aligned}$$

$$J = \rho(I + (A - J))$$

$$(I - [J])^{-1} = (1 + 7x_1 + 5x_2 + 2x_3)^{-1}$$

Without performing iterations show that sequence does not converge to exact soln of system.

Youtube

Jacobi iteration $(I - J)^{-1} \epsilon$ should be diagonally dominant

so that $|a_{ii}| \geq |a_{ij}|$ for all i, j and $i \neq j$

if $|a_{ii}| > |a_{ij}|$ for all i, j and $i \neq j$

$$\sum_{j=1}^{n-1} |a_{ij}| \leq |a_{ii}|$$

$$|a_{ii}| \leq \frac{1}{2} \sum_{j=1}^n |a_{ij}|$$

③ System of non-linear equations

Let us consider $f(x, y) = 0$ (and) $g(x, y) = 0$ be two non-linear equations with unknown x and y .
 let (x_0, y_0) be an initial approximation of ①
 write expansion Approximate f and g using Taylor's formula.

$$f(x) = f(a) + (x-a) f'(a) + \frac{(x-a)^2}{2!} f''(a) + \dots \quad ①$$

$$\begin{aligned} f(x, y) &= f(x_0, y_0) + \left[(x-x_0) f'_x(x_0) + (y-y_0) f'_y(y_0) \right] \\ &\quad + \left(\frac{(x-x_0)^2}{2!} f''_{xx}(x_0) + \frac{(y-y_0)^2}{2!} f''_{yy}(y_0) + \right. \\ &\quad \left. + (x-x_0)(y-y_0) f''_{xy}(x_0, y_0) \right) + \dots \end{aligned}$$

linear approximation

$$\begin{aligned} f(x, y) &\approx f(x_0, y_0) + \left[(x-x_0) f'_x(x_0) + (y-y_0) f'_y(y_0) \right] \\ g(x, y) &\approx g(x_0, y_0) + \left[(x-x_0) g'_x(x_0) + (y-y_0) g'_y(y_0) \right] \end{aligned}$$

Equate RHS quantities to zero.

$$\left. \begin{aligned} f(x_0, y_0) + (x-x_0) f'_x(x_0) + (y-y_0) f'_y(y_0) &= 0 \\ g(x_0, y_0) + (x-x_0) g'_x(x_0) + (y-y_0) g'_y(y_0) &= 0 \end{aligned} \right\} \quad ③$$

Solve ③ for x & y and denote them as x_1 & y_1 , respectively.

$$\text{From } \frac{f(x_0, y_0)}{f'_x(x_0)} \cdot g'_x(x_0) + (x - x_0) g''_x(x_0) + \frac{(y - y_0) f'_y(y_0)}{f'_x(x_0)} g'_x(x_0) = 0$$

$$(x - x_0) g'(x_0) + (y - y_0) g'_y(y_0) = 0$$

$$\frac{f(x_0, y_0)}{f'_x(x_0)} g'_x(x_0) + \frac{(y - y_0) f'_y(y_0) g'_x(x_0)}{f'_x(x_0)} - g(x_0, y_0) = (y - y_0) g'_y(y_0) = 0$$

$$y - y_0 = \left[\frac{f(x_0, y_0) g'_x(x_0)}{f'_x(x_0)} - g(x_0, y_0) \right] \frac{f'_x(x_0)}{f'_y(y_0) g'_x(x_0) + g'_y(y_0) f'_x(x_0)}$$

$$y = y_0 + \frac{f(x_0, y_0) g'_x(x_0) - g(x_0, y_0) f'_x(x_0)}{g'_y(y_0) f'_x(x_0) - f'_y(y_0) g'_x(x_0)}$$

$$x = x_0 + \frac{f(x_0, y_0) g'_y(x, y_0) - g(x, y_0) f'_y(x, y_0)}{f'_x(x_0, y_0) g'_y(x, y_0) - g'_x(x_0, y_0) f'_y(x, y_0)}$$

\downarrow
 Δx

$$\left. \begin{array}{l} x_1 = x_0 + \Delta x \\ y_1 = y_0 + \Delta y \end{array} \right\} \textcircled{5}$$

$$\begin{bmatrix} f_x(x_0, y_0) & f_y(x_0, y_0) \\ g_x(x_0, y_0) & g_y(x_0, y_0) \end{bmatrix} \begin{bmatrix} \Delta x_0 \\ \Delta y_0 \end{bmatrix} = - \begin{bmatrix} f(x_0, y_0) \\ g(x_0, y_0) \end{bmatrix}$$

$\downarrow J_0$

$$J_0 \frac{\Delta x^{(0)}}{\Delta y^{(0)}} = -F(x_0, y_0)$$

Assume $J_0 \rightarrow \text{invertible}$

$$J_0^{-1} \frac{\Delta x^{(0)}}{\Delta y^{(0)}} = -F(x_0, y_0), \text{ } \textcircled{6}$$

From $\textcircled{5}$ & $\textcircled{6}$ we obtain

$$\begin{bmatrix} x_1 \\ y_1 \end{bmatrix} = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix} + \begin{bmatrix} \Delta x_0 \\ \Delta y_0 \end{bmatrix} = \begin{bmatrix} x_0 \\ y_0 \end{bmatrix} - J_0^{-1} F(x_0, y_0)$$

General iterative formula

$$\begin{bmatrix} x_{k+1} \\ y_{k+1} \end{bmatrix} = \begin{bmatrix} x_k \\ y_k \end{bmatrix} - J_k^{-1} F(x_k, y_k)$$

$x \Delta$

$$\begin{array}{|c|c|c|} \hline 1 & 3 & 1 \\ \hline 2 & 5 & 0 \\ \hline 1 & 3 & 1 \\ \hline 0 & 1 & 1 \\ \hline 1 & 3 & 1 \\ \hline 0 & 1 & 1 \\ \hline \end{array}$$

$$J_k = \begin{bmatrix} f_x(x_k, y_k) & f_y(x_k, y_k) \\ g_x(x_k, y_k) & g_y(x_k, y_k) \end{bmatrix}$$

$$F(x_k, y_k) = \begin{bmatrix} f(x_k, y_k) \\ g(x_k, y_k) \end{bmatrix}$$

Example — The system of equation

$$f(x, y) = y \cos(xy) + 1 = 0$$

$$g(x, y) = \sin(xy) + x - y = 0$$

has one solution close to $(1, 2)$. Calculate

approximate solution using Newton-Raphson method
(2 iterations). $\rightarrow x_0, x_1, x_2$

$$x_0 = 1$$

$$y_0 = 2$$

$$\begin{bmatrix} x_1 \\ y_1 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} - J_0^{-1} F(x_0, y_0) \quad \begin{bmatrix} F(x_0, y_0) \\ g(x_0, y_0) \end{bmatrix} = \begin{bmatrix} 2\cos 2 + 1 \\ \sin 2 - 1 \end{bmatrix}$$

$$J_0 = \begin{bmatrix} f_x(x_0, y_0) & f_y(x_0, y_0) \\ g_x(x_0, y_0) & g_y(x_0, y_0) \end{bmatrix} \quad f_x = -y^2 \sin(xy) \quad f_y = \cos(xy)$$

$$\phi = y^2 \sin(xy)$$

$$f_x = -y^2 \sin(xy) + 1 \quad f_y = \cos(xy) - 1$$

$$R_1 \rightarrow R_2 - 2R_1 \\ R_1 \rightarrow R_1 - 3R_2$$

$$\left[\begin{array}{ccc|c} 1 & 3 & 1 & 0 \\ 2 & 1 & 0 & 1 \end{array} \right]$$

$$\left[\begin{array}{ccc|c} 1 & 3 & 1 & 0 \\ 0 & 6 & -2 & 1 \end{array} \right] \quad \left[\begin{array}{ccc|c} 1 & 3 & 1 & 0 \\ 0 & 1 & -\frac{1}{3} & \frac{1}{6} \end{array} \right]$$

$$\left[\begin{array}{ccc|c} 1 & 0 & 1 & -2 \\ 0 & 1 & -2 & 1 \end{array} \right]$$

$$\begin{bmatrix} -4\sin(2) & \cos(2) - 2\sin(2) \\ 2\cos(2) + 1 & \cos(2) - 1 \end{bmatrix}$$

$$J_0^{-1} = \frac{1}{10} \begin{bmatrix} \cos(2) - 1 & 2\sin(2) - 8\cos(2) \\ 2\cos(2) - 1 & -4\sin(2) \end{bmatrix}$$

1	575	—	—	—
2	5239	—	—	—

General form: $|z - z_0| = r$

11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30

Consider, \exists \forall 's outside and \exists 's inside.

$$\text{Implies } \alpha_1 + \alpha_2 + \dots + \alpha_n = 0.$$

$$\sum_{i=1}^n x_1 x_2 x_3 \cdots (x_i) = 0$$

→

$$\text{Definition: } \exists n \in \mathbb{N} \quad f_n(x_1, x_2, \dots, x_n) = 0 \quad \checkmark$$

For solving n -system of non-linear equation

$$\begin{bmatrix} x_1^{(k)} \\ x_2^{(k)} \\ \vdots \\ x_n^{(k)} \end{bmatrix} = J_K^{-1} F(x_1^{(k)}, x_2^{(k)}, \dots, x_n^{(k)})$$

(k=0, 1, 2, ..., iterations)

$$\begin{aligned} & \text{Left side: } (s+2)(s-2) - (s^2 + 4) = s^2 + 2s - 2s - 4 - s^2 - 4 = -8 \\ & \text{Right side: } 1 - (s^2 + 4) = 1 - s^2 - 4 = -s^2 - 3 \end{aligned}$$

$$J = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_{21}}{\partial x_2} & \cdots & \frac{\partial f_{n1}}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \cdots & \frac{\partial f_{2n}}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \frac{\partial f_n}{\partial x_2} & \cdots & \frac{\partial f_n}{\partial x_n} \end{bmatrix} \quad (1)$$

$$F(x_1^{(k)}, x_2^{(k)}, \dots, x_n^{(k)}) = \begin{bmatrix} f_1(x_1^{(k)}, x_2^{(k)}, \dots, x_n^{(k)}) \\ f_2(x_1^{(k)}, x_2^{(k)}, \dots, x_n^{(k)}) \\ \vdots \\ f_n(x_1^{(k)}, x_2^{(k)}, \dots, x_n^{(k)}) \end{bmatrix}$$

Note : A necessary and sufficient condition for convergence of (8) is

$$\|J\|_2 = \sqrt{\lambda_1^2 + \lambda_2^2 + \dots + \lambda_n^2} \leq R$$

$$\rho(J^{-1}) < 1$$

↳ spectral radius

Stopping criterion (for given tolerance)

$$\|x_i^{k+1} - x_i^k\| \leq \epsilon$$

x

x

(2)

$$0.1E = 5 - 4.5E - 0.5E$$

$$-0.5E = 5E + 4E + 0.5E$$

$$0.5E = 5.5E + 4.5E + 0.5E$$

$$A/E = 5.5E + 4.5E + 0.5E$$

Eigen Value Problems

Defⁿ - Let $A \in \mathbb{R}^{n \times n}$ be a general $n \times n$ matrix

A number λ is said to be an eigen value of A if there exists a non zero vector $v \in \mathbb{R}^n$ such that $(Av - \lambda v) = 0$

$$\det(A - \lambda I) = 0 \quad \text{or} \quad Av = \lambda v$$

\downarrow
polynomial in λ of order 'n'

Dominant Eigen Value - An eigen value λ of $n \times n$ matrix is said to be dominant eigen value if $|\lambda| = \max\{|z| : z \in \sigma(A)\}$ is an eigen value of A .

Power Method - This method is used to obtain a specific eigen value called dominant eigen value and the corresponding eigen vector for a given $n \times n$ matrix.

Steps -

1) Let x_0 be the initial vector which is generally chosen as vector with all components as 1, that is $x_0 = \begin{bmatrix} 1 \\ 1 \\ \vdots \\ 1 \end{bmatrix}$ or $\begin{bmatrix} 1 \\ 0 \\ \vdots \\ 0 \end{bmatrix}$ or $\begin{bmatrix} 0 \\ 1 \\ \vdots \\ 0 \end{bmatrix}$ or $\begin{bmatrix} 0 \\ 0 \\ \vdots \\ 1 \end{bmatrix}$

2) Calculate Ax_0 and write it in the form $Ax_0 = \lambda_1 x_1$, where x_1 is normalized by taking out the largest component of Ax_0 .

3) Calculate AX_1 and express AX_1 as $\lambda_2 X_2 + \lambda_3 Y_2$

where X_2 is normalized as Y_2 in step 2.

by taking out largest (in magnitude) component of AX_1 .

Continue the process till the diagonal result is

obtained.

4) Then we have a sequence of eqns:

$$AX_0 = \lambda_1 X_1 + \lambda_2 Y_1, \quad AX_1 = \lambda_2 X_2 + \lambda_3 Y_2, \quad \dots \quad \text{approximate}$$

The λ_n is an approximation of dominant eigen value λ_{dom} of eqn.

Q) Find the largest (in magnitude) eigen value if

$$A = \begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix} \text{ using power method and compare}$$

your result with explicit solution.

Sol: $X_0 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ as initial vector.

$$\text{for } AX_0 = \begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = \begin{bmatrix} 5 \\ 4 \end{bmatrix} = 5 \begin{bmatrix} 1 \\ 4/5 \end{bmatrix}$$

$$AX_1 = \lambda_2 X_2$$

$$\text{now } \begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 4/5 \\ 1/5 \end{bmatrix} = \begin{bmatrix} 4+4/5 \\ 1+1/5 \end{bmatrix} = \begin{bmatrix} 4.8 \\ 1.2 \end{bmatrix} = 4.8 \begin{bmatrix} 1 \\ 1/4 \end{bmatrix}$$

$$\text{so } AX_2 = \begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 1/4 \end{bmatrix} = \begin{bmatrix} 4+0.25 \\ 1+0.75 \end{bmatrix} = \begin{bmatrix} 4.25 \\ 1.75 \end{bmatrix} = 4.25 \begin{bmatrix} 1 \\ 1/1.75 \end{bmatrix}$$

$$\begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 0.6637 \end{bmatrix} = \begin{bmatrix} 4.6637 \\ 2.9242 \end{bmatrix}$$

$$4.6637 \begin{bmatrix} 1 \\ 0.63 \end{bmatrix}$$

12th iteration

$$\begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} 1 \\ 0.63 \end{bmatrix} = \begin{bmatrix} 4.6637 \\ 2.9242 \end{bmatrix}$$

$$(A - \lambda I) \begin{bmatrix} 4-\lambda & 1 \\ 1 & 3-\lambda \end{bmatrix} \begin{bmatrix} 1 \\ 1 \end{bmatrix} = 0$$

$$6.12 + \lambda^2 - 7\lambda - 1 = 0$$

$$\lambda^2 - 7\lambda + 11 = 0$$

$$\lambda = 7 \pm \frac{\sqrt{49-44}}{2}$$

$$\begin{bmatrix} 2 & 1 \\ 1 & 1 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \end{bmatrix} = \frac{1}{2} \begin{bmatrix} 5 \\ 3 \end{bmatrix} = 4.618$$

$$\lambda_1 = 2.38 \quad \lambda_2 = 4.618$$

* Note:

For a given tolerance ϵ one can stop when
 $|x_{i+1} - x_i| < \epsilon$ or
 $|d_{i+1} - d_i| < \epsilon$

$$\begin{bmatrix} 2805 & 1200 \\ 1200 & 1200 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 11 \\ 12 \end{bmatrix} \rightarrow \text{final value using dig. values}$$

$$Ax = b \quad \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 11 \\ 12 \end{bmatrix}$$

Find the largest eigen value of $A = \begin{bmatrix} 3 & 0 & 0 \\ -4 & 6 & 2 \\ 16 & -15 & -5 \end{bmatrix}$
using power method.

$$\lambda_1 = 1 \quad \lambda_2 = 0 \quad \lambda_3 = 3$$

$$X_0 = \begin{bmatrix} 1 \\ 0.5 \\ 0.25 \end{bmatrix}$$

$$AX_0 = \begin{bmatrix} 3 \\ -4+3+0.5 \\ 40-15-5 \end{bmatrix} = \begin{bmatrix} 3 \\ -4.5 \\ 16-0.5 \times 5 - 0.25 \times 5 \end{bmatrix}$$

$$= \begin{bmatrix} 3 \\ -0.5 \\ 7.25 \end{bmatrix} = \frac{1}{7.25} \begin{bmatrix} 0.413 \\ -0.0689 \\ 1 \end{bmatrix}$$

largest value $\rightarrow A^T X_1$

$$AX_1 = \lambda_2 X_2$$

$$\begin{bmatrix} 3 & 0 & 0 \\ -4 & 6 & 2 \\ 16 & -15 & -5 \end{bmatrix} \begin{bmatrix} 0.413 \\ -0.0689 \\ 1 \end{bmatrix} = \begin{bmatrix} 1.246 \\ -0.0689 \\ 2.655 \end{bmatrix}$$

$$\text{largest value} \rightarrow A^T X_2$$

$$\begin{bmatrix} 3 & 0 & 0 \\ -4 & 6 & 2 \\ 16 & -15 & -5 \end{bmatrix} \begin{bmatrix} 0.4637 \\ 0.0359 \\ 1 \end{bmatrix} = \begin{bmatrix} 0.4637 \\ 0.0359 \\ 1 \end{bmatrix}$$

$$AX_2 = \lambda_3 X_3$$

$$\begin{bmatrix} 3 & 0 & 0 \\ -4 & 6 & 2 \\ 16 & -15 & -5 \end{bmatrix} \begin{bmatrix} 0.4637 \\ 0.0359 \\ 1 \end{bmatrix} = \begin{bmatrix} 0.4637 \\ -0.00905 \\ 1 \end{bmatrix}$$

$$AX = \begin{bmatrix} 1.99952 \\ -0.08333 \\ 2.99930 \end{bmatrix} \quad \lambda = 2.9990 \quad \begin{bmatrix} 0.69395 \\ 0.53666 \\ 1.00000 \end{bmatrix}$$

 (d_{12}) \downarrow

0.01100

close to

actual A_2

= 3

close to λ_2

7.16811

\checkmark eigen vector

 $v = \begin{bmatrix} 1 \\ 0 \\ 2 \end{bmatrix} = 2 \begin{bmatrix} 0.5 \\ 0 \\ 1 \end{bmatrix}$

$P^{-1} = \begin{bmatrix} 2 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{bmatrix}$

A:

square

If A is a real matrix

↳ eigen values are real? A

No

$$\text{eg } A = \begin{bmatrix} 3 & 2 & 1 \\ 2 & 4 & 0 \\ 1 & 0 & 2 \end{bmatrix} \quad \lambda = \begin{cases} 1+2i \\ 1-2i \\ 3 \end{cases}$$

If A is a real symmetric matrix

↳ Are all eigen values real?

Yes

↳ orthogonal matrix is

$$S = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad S^T A S = D$$

↓
diagonal matrix

$$S = [v_1, v_2, v_3, v_4]$$

$\{v_1, v_2, v_3, v_4 \rightarrow \text{eigen vectors}\}$

Jacobi Method for real symmetric matrix

Let A be a real symmetric matrix. Then the eigen values of A are real and there exists a real orthogonal matrix S such that

$$S^{-1}AS = D, \text{ where } D \text{ is diagonal matrix.}$$

These processes can be done by applying a series of orthogonal matrices S_1, S_2, \dots

Simple case -

Let $A = \begin{bmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{bmatrix}$ be a real symmetric matrix.

$$i) \quad S = \begin{bmatrix} \cos\theta & \sin\theta \\ \sin\theta & \cos\theta \end{bmatrix} \quad (ii) \quad S^{-1}AS = D$$

To find S_1 i.e. $\begin{bmatrix} \cos\theta & -\sin\theta \\ \sin\theta & \cos\theta \end{bmatrix}$ where θ is angle to be determined.

Calculate $S^{-1}AS$

$$\begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} \\ a_{12} & a_{22} \end{bmatrix} \begin{bmatrix} \cos\theta & -\sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix}$$

$$= \begin{bmatrix} a_{11}\cos^2\theta + a_{12}\sin 2\theta + a_{22}\sin^2\theta, & (a_{22}-a_{11})\sin\theta\cos\theta \\ (a_{22}-a_{11})\sin\theta\cos\theta + a_{12}\cos 2\theta, & a_{11}\sin^2\theta + a_{22}\cos^2\theta \end{bmatrix}$$

$$- a_{12}\sin 2\theta$$

To make off-diagonal entries zero, we choose θ in such a way that

$$\frac{1}{2}(a_{22} - a_{11})2\sin\theta\cos\theta + a_{12}\cos^2\theta = 0.$$

$$\frac{1}{2}(a_{22} - a_{11})\sin 2\theta + a_{12}\cos 2\theta = 0.$$

$$\tan 2\theta = \frac{2a_{12}}{a_{11} - a_{22}}$$

$$\theta = \frac{1}{2}\tan^{-1}\left(\frac{2a_{12}}{a_{11} - a_{22}}\right)$$

Let A be $m \times n$ real symmetric matrix.

Let $A = (a_{ij})_{1 \leq i, j \leq n}$ where $a_{ij} \in \mathbb{R}$ and $a_{ij} = a_{ji}$

Step 1: Find out the largest (in magnitude) of the off-diagonal element in A , a_{ij}

$$A = \begin{pmatrix} 2 & 3 & 4 \\ 3 & 8 & 7 \\ 4 & 7 & -9 \end{pmatrix}$$

Step 2: Calculate $\theta = \frac{1}{2}\tan^{-1}\left(\frac{2a_{ij}}{a_{ii} - a_{jj}}\right)$

Step 3: Set $S_1 = \begin{pmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & \cos\theta & -\sin\theta & 0 & 0 \\ 0 & \sin\theta & \cos\theta & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{pmatrix}$

$$\left. \begin{array}{l} S_{ii} = S_{jj} = \cos\theta, \text{ other diagonal entries } = 1 \\ S_{ij} = -\sin\theta, S_{ji} = \sin\theta, \text{ all other elements } = 0 \end{array} \right\}$$

Observe that S_1 is invertible

Step 4 :

Compute $D_1 = S_1^{-1} A S_1$

Step 5 : Let ϵ be the given tolerance - if sum of squared of off-diagonal matrix entries is less than ϵ :

$$\sum (a_{ij})^2$$

Or else repeat steps 1-4 for matrix D_1

$$A = \begin{pmatrix} 1 & 2 & -1 \\ 2 & 1 & 2 \\ -1 & 2 & 3 \end{pmatrix}$$

Step 5 $a_{ij} = 2$

$$\theta = \tan^{-1} \left(\frac{a_{212}}{a_{11} - a_{22}} \right)$$

$$\theta = \pi/4$$

$$S_1 = \begin{pmatrix} \cos\pi/4 & -\sin\pi/4 & 0 \\ \sin\pi/4 & \cos\pi/4 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$S^{-1}AS^{-1} = \begin{pmatrix} 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ -1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 & -1 \\ 2 & 1 & 2 \\ -1 & 2 & 1 \end{pmatrix} \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{2} & 0 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$D = \begin{pmatrix} 3 & 0 & 1/\sqrt{2} \\ 0 & -1 & 3/\sqrt{2} \\ 1/\sqrt{2} & 3/\sqrt{2} & 1 \end{pmatrix}$$

(At least 2 more iterations)

(Q) Using Jacobi Method to find all eigen values of $A = \begin{bmatrix} 1 & (2)^{-1} & 1 \\ 2 & 1 & 2 \\ -1 & 2 & 1 \end{bmatrix}$

$$\theta = \tan^{-1} \left| \frac{2(2)}{2+1-1} \right| = \frac{1}{2} \left(\frac{\pi}{2} \right) = \frac{\pi}{4}$$

$$S_1 = \begin{pmatrix} \cos\theta & 0 & 0 \\ 0 & \sin\theta & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \sqrt{2}/2 & \sqrt{2}/2 \\ 0 & \sqrt{2}/2 & -\sqrt{2}/2 \end{pmatrix}$$

$$S^{-1} = \begin{pmatrix} 1/\sqrt{2} & -1/\sqrt{2} & 0 \\ 1/\sqrt{2} & 1/\sqrt{2} & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

$$S^{-1}AS^{-1} = \begin{pmatrix} 3 & 0 & 1/\sqrt{2} \\ 0 & -1 & 3/\sqrt{2} \\ 1/\sqrt{2} & 3/\sqrt{2} & 1 \end{pmatrix}$$

$$D = S^{-1}AS \Rightarrow \begin{bmatrix} 3 & 0 & 0 \\ 0 & -1 & 2.1213 \\ 0 & 2.1213 & 1 \end{bmatrix}$$

$$\rightarrow \begin{bmatrix} 3 & 0 & 1/\sqrt{2} \\ 0 & -1 & 3/\sqrt{2} \\ 1/\sqrt{2} & 3/\sqrt{2} & 1 \end{bmatrix} \xrightarrow{d_{23}}$$

$\theta = 1/2 + \tan^{-1}\left(\frac{2 \cdot 3/\sqrt{2}}{-1 - 1}\right)$

$$\theta = \left(\frac{\pi}{6}\right) - 0.565$$

$$\begin{aligned} \cos \theta &= 0.84 \\ \sin \theta &= -0.565 \end{aligned}$$

$$S = \begin{bmatrix} 0.84 & 0.565 \\ -0.565 & 0.84 \end{bmatrix}$$

$$S^2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0.84 & 0.565 \\ 0 & -0.565 & 0.84 \end{bmatrix}$$

$$D_2 = S^{-1}AS \Rightarrow \begin{bmatrix} 3 & -0.378 & 0.5939 \\ -0.381 & -2.345 & -0.009 \\ 0.5938 & -0.009 & 2.345 \end{bmatrix}$$

Eigen values

$$D_6^2 = \begin{bmatrix} 3 & 0.00048 & 0 \\ 0.0048 & -2.37 & 0 \\ 0 & 0 & 1.99 \end{bmatrix}$$

Extract eigenvalues

$$\text{Given } \left\{ \begin{array}{l} \lambda_1 = 3.37 \\ \lambda_2 = -2.37 \\ \lambda_3 = 2 \end{array} \right\} \rightarrow 2A^T + 9$$

(Q)

$$A = \begin{bmatrix} 3 & 2 & 2 \\ 2 & 3 & 2 \\ 2 & 2 & 3 \end{bmatrix}$$

$$\begin{pmatrix} -2.37 & 0 & 1.37 \\ 0 & 3.37 & 0 \\ 0 & 0 & 2.37 \end{pmatrix}$$

$$\theta = \tan^{-1} \left(\frac{2e^{i\phi}}{\omega_{11} - \omega_{33}} \right) = \pi/4$$

$$S = \begin{bmatrix} 1/\sqrt{2} & 0.8 \cdot 0 - 1/\sqrt{2} \\ 0 & 1/\sqrt{2} \\ 0 & 0.8 \cdot 0 + 1/\sqrt{2} \\ 0 & 4.8 \cdot 0 - 2\sqrt{2}, 0 \end{bmatrix}$$

$$A_2 = D_2^{-1} \begin{bmatrix} 5 & 2.8284 & 0 \\ 2.8284 & 5 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$\theta = 30.5^\circ + \tan^{-1} \left(\frac{-1}{4} \right) = \pi/4$$

$$P_{00,0} = 853.0 - 82 \rightarrow 281.8 = 1$$

$$P_{00,0} = 853.0 - 82 \rightarrow 281.8 = 1$$

$$P_{00,0} = \begin{bmatrix} 1/\sqrt{2} & -1/\sqrt{2} & 0 \\ 0 & 1/\sqrt{2} & 0 \\ 0 & 0 & 1 \end{bmatrix} D_2 \rightarrow \begin{bmatrix} 7.8284 & 0 \\ 0 & 2.1716 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 84000.0 & 82 \\ 0 & 18.5 - 64000.0 & 281.8 \\ 0 & 0 & 0 \end{bmatrix}$$

Eigen values of A are diagonal elements

$$\lambda_1 = 7.8284$$

$$\lambda_2 = 2.1716$$

$$\lambda_3 = 1.00$$

- (Q) Use Power Method to find smallest eigen value for some matrix A

$$A = \begin{bmatrix} 3 & 2 & 2 \\ 2 & 3 & 2 \\ 2 & 2 & 3 \end{bmatrix}$$

$$AX_0 = \begin{pmatrix} 7 \\ 7 \\ 7 \end{pmatrix} = 7 \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$$

$$A = \begin{bmatrix} 2.05 & -1 & 0 \\ -1 & 2 & -1 \\ 0 & -1 & 2 \end{bmatrix}$$

Concept $\rightarrow A^{\alpha} = ((\lambda_1)^{\alpha}, (\lambda_2)^{\alpha}, (\lambda_3)^{\alpha})$ $A^{-1} = (1/\lambda_1, 1/\lambda_2, 1/\lambda_3)$

$$|1/\lambda_3| < |1/\lambda_1|$$

$$A^{-1} = \begin{bmatrix} 0.75 & 0.5 & 0.25 \\ 0.5 & 1 & 0.5 \\ 0.25 & 0.5 & 0.75 \end{bmatrix}$$

elements belongs to A for matrix $A = \begin{bmatrix} 1.5 & 2 & 1.5 \\ 2 & 1 & 0.75 \\ 1.5 & 0.75 & 1 \end{bmatrix}$

$$A^{-1} x_2 =$$

steps //

①

Check $\det(A) \neq 0$

②

Find largest value of A^{-1} using power method

③

x_1 will be smallest

$$A^{-1} x_2 = \begin{bmatrix} 1.5 \\ 2 \\ 1.5 \end{bmatrix} = 2 \begin{bmatrix} 0.75 \\ 1 \\ 0.75 \end{bmatrix} + x_2$$

$$A^{-1} x_2 = \begin{bmatrix} 1.5 \\ 2 \\ 1.5 \end{bmatrix} = \begin{bmatrix} 1.25 \\ 1.75 \\ 1.25 \end{bmatrix} = \begin{bmatrix} 0.213 \\ 1.25 \\ 0.213 \end{bmatrix} + x_2$$

$$A^{-1} x_3 = \begin{bmatrix} 1.213 \\ 1.713 \\ 1.213 \end{bmatrix} = \begin{bmatrix} 0.708 \\ 1 \\ 0.708 \end{bmatrix} + x_3$$

$$A^{-1} x_4 = \begin{bmatrix} 1.208 \\ 1.708 \\ 1.208 \end{bmatrix} = \begin{bmatrix} 0.707 \\ 1 \\ 0.707 \end{bmatrix} + x_4$$

$$(A - \lambda_1 I)^{-1} = A^{-1} (I - \lambda_1^{-1} A) = A^{-1} (I - k_1^{-1} A)$$

$$|(k_1 A)| > |(A)|$$

$$\begin{bmatrix} 25.0 & 2.0 & 2.0 \\ 2.0 & 1 & 0 \\ 2.0 & 0 & 25.0 \end{bmatrix} \quad \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

Numerical Solution of ordinary differential equation (ODE)

Let us consider the following ODE

$$y' = f(x, y) \quad \text{--- (1)}$$

\downarrow \downarrow
 dependent independent

where $y(x)$ is unknown, $f: \mathbb{R} \times \mathbb{R} \rightarrow \mathbb{R}$ is a given function $y(x) = \frac{dy}{dx}$.

The aim is to find a solution of (1) with the initial condition

$$y(x_0) = y_0 \quad \text{--- (2)}$$

Finalizing the solution of (1) with initial condition

(2) is called initial value problem (IVP).

Lemma: A continuous function y defined in an interval I containing the point x_0 is a solution of IVP (1) if and only if y satisfies the integral equation.

$$y(x) = y_0 + \int_{x_0}^x f(s, y(s)) ds \quad \forall x \in I$$

We partition the interval $I = [x_0, b]$ into the points $x_0 < x_1 < x_2 \dots < x_N = b$ with $x_i = x_0 + ih$, $i = 1, 2, 3 \dots N$.

Single Step Method

$$\rightarrow \text{explicit } y(x_{n+1}) = \phi(x_0, y_0, f(x_0, y_0, h))$$

$$\rightarrow \text{implicit } y(x_{n+1}) = \phi(x_n, y_n, h, f(x_{n+1}, y_{n+1}))$$

Taylor Series MethodTaylor Series expansion of $y(x)$ around x_0

$$y(x) = y(x_0) + (x-x_0) y'(x_0) + \frac{(x-x_0)^2}{2!} y''(x_0) + \dots$$

$$\frac{(x-x_0)^2}{2!} y''(x_0)$$

or Taylor's method expansion of $n!$

$$\Phi) \quad y' = -2xy^2 \quad y(1) = 1$$

$$\text{here } x_0 = 1 \text{ and } y = 1$$

$$h = 0.1 \quad x_1 = 1.1 \quad x_2 = 1.2 \quad x_3 = 1.3$$

$$y''(x_0) = \frac{4xy^3}{(x-2y)^2}$$

a

$$y(x_1) = y(x_0) + h y'(x_0) + \frac{h^2}{2!} y''(x_0)$$

$$= 1 + (0.1)(-2 \times 1 \times 1) + \frac{(0.1)^2}{2!} \times 6$$

$$= 1 - 0.2 + 0.03$$

$$= 0.8 + 0.03$$

$$= 0.83$$

$$y(x_2) = y(x_1) + hy'(x_1) + \frac{h^2}{2!} y''(x_1)$$

$$= 0.83 + (0.1) \left(-2 \times (0.83)^2 \times 1.1 \right) + \frac{(0.1)^2}{2!} \left[-4(0.1)(0.83)(1.1) - 2(0.83)^2 \right]$$

$$y(x_2) = 0.5967$$

Put $x_n = x_{n+1}$

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

$$y(x_{n+1}) = y(x_n) + hy'(x_n) + \frac{h^2}{2!} y''(x_n)$$

If we retain terms upto the path power of h in the above equation, we get an approximate

$$y_{n+1} = y_n + hy_n + \frac{h^2}{2!} y''_n + \dots + \frac{h^p}{p!} y^{(p)}_n$$

and error as

$$E = \frac{h^{p+1}}{(p+1)!} y^{(p+1)}(x_n) + \frac{h^{p+2}}{(p+2)!} y^{(p+2)}(x_n)$$

$$E = \frac{h^{p+1}}{(p+1)!} y^{(p+1)}(x_\xi) \quad \text{where } x_n < \xi < x_{n+1}$$

The path order Taylor series method for solving an ODE of the form ① is $y(x) = F(x, y)$
 $y(x_0) = y_0$

is given as

Consider the ODE, $y(x) = f(x, y)$ Runge Kutta Method

$$y(x_0) = y_0$$

Assume that $y(x_n)$ is known. Use Taylor Series expansion of $y(x_{n+1})$ around x_n upto second order

$$y(x_{n+1}) = y(x_n) + y'(x_n) + \frac{h}{2!} y''(x_n) + O \quad (1)$$

$$y'(x) = \frac{\partial F}{\partial x}(x, y) + \frac{\partial F}{\partial y}(x, y) \cdot \frac{\partial y}{\partial x}$$

$$y''(x) = \frac{\partial^2 F}{\partial x^2}(x, y) + \frac{\partial^2 F}{\partial y^2}(x, y) + f(x, y) \quad (2)$$

Substitute (2) in (1)

$$y(x) = y(x_{n+1})$$

$$y(x_{n+1}) = y(x_n) + h f(x_n, y_n) + \frac{h^2}{2!} \left[\frac{\partial^2 F}{\partial x^2}(x_n, y_n) + \frac{\partial^2 F}{\partial y^2}(x_n, y_n) f(x_n, y_n) \right] + (O) h^3$$

$$= y(x_n) + \frac{h}{2} (x_n y_n) + \frac{h}{2} \left[f(x_n, y_n) + \frac{h}{2} \frac{\partial^2 F}{\partial x^2}(x_n, y_n) + \frac{h}{2} \left[h f(x_n, y_n) \frac{\partial^2 F}{\partial y^2}(x_n, y_n) + O(h^3) \right] \right]$$

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

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

$$= y(x_n) + \frac{h}{2} f(x_n, y_n) + \frac{h}{2} f(x_n+h, y_n+h) f(x_n, y_n) + O(h^2)$$

lets denote $k_1 = h f(x_n, y_n)$

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

$$k_2 = h f(x_{n+1}, y_n + k_1)$$

Thus we can approximate $y(x_{n+1})$ by

$$y_{n+1} = y_n + \frac{1}{2}(k_1 + k_2) \quad \text{--- (2)}$$

where k_1, k_2 are in (4)

This equation is called as the
Runge - kutta Method of order two

$$\text{Error} = \frac{h^3}{3!} y'''(\xi) \quad x_n \leq \xi < x_{n+1}$$

$\left\{ \begin{array}{l} \text{Rate of convergence} = 3 \\ \text{order} = 3 \end{array} \right\}$

The truncation error of Runge - kutta method
is of order $O(h^3)$.

Example: Solve the ODE

$$\begin{aligned} y' &= y \\ y(0) &= 1 \end{aligned}$$

using R-K method of order 2 with

step size $R = 0.01$

$$\left\{ \begin{array}{l} f(x, y) = 0 \\ y_0 = 1 \end{array} \right.$$

n	x	y	k_1	k_2	
0	0	1	0.01	0.0101	$y_1 = y_0 + \frac{1}{2}(k_1 + k_2)$
1	0.01	1.01005	0.0101005	0.010201505	$k_1 = h f(x_1, y_1) = 0.01$
2	0.02	1.020201005	0.010201005	0.0105000	$k_2 = h f(x_2, y_1 + k_1) = 0.01$
3	0.03	1.030454	0.010202	0.010304	$f(0.01, 1.010201505) = 0.01$
4	0.04	1.040510	0.010305	0.010408	$^2(0.01)(1.01)$

$$y_1 = y_0 + \frac{1}{2}(k_1 + k_2)$$

$$= 1 + \frac{1}{2}(0.01 + 0.0101)$$

$$k_1 = h f(x_1, y_1) = h y_1$$

$$k_2 = h f(x_2, y_1 + k_1) = 0.01 \times \begin{cases} 1.01005 \\ + 0.0101005 \end{cases}$$

$$= 0.010201505$$

$$y_2 = y_1 + \frac{1}{2}(k_1 + k_2)$$

Runge Kutta Method of order four

We state deviation - the formula for the Runge - Kutta method of order four.

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

How to solve it bottom part

$$10 - 9 = 1$$

where

$$k_1 = h f(x_0, y_0)$$

$$k_2 = h f(x_0 + k_1, y_0 + k_1)$$

$$k_3 = h f(x_0 + k_2/2, y_0 + k_2/2)$$

$$k_4 = h f(x_0 + h, y_0 + k_3)$$

HW

e.g/ Solve the ODE,

$$y' = y \quad y(0) = 1$$

using Runge - Kutta method of order four
with step-size $h = 0.01$. Compare with exact
solution and find the error.

x ————— x —————

Ex 32.1

5) $y' = 1 - 2xy$ $x_0 = 0$ $x = x_0 + nh$
 $y'(0) = 0$ $h = 0.2$ $= 0.2$

$$y_{1g} = y(x) + \frac{h y'(x)}{2!} + \frac{h^2 y''(x)}{3!} + \frac{h^3 y'''(x)}{4!}$$

$$y'(0) = 1$$

$$y''(x) = 1 - 2(y + xy')$$

$$y''(0) = 1 - 2(y(0) + 0) = 1$$

$$y'''(x) = 1 - 2(y' + y' + xy'')$$

$$y'''(0) = 1 - 2(1 + 1 + 0) \\ = 1 - 4 = -3$$

$$y(0.2) = 0 + (0.2)(1) + \frac{(0.2)^2}{2}(1) + \frac{(0.2)^3}{3}(-3) \\ = 0.216 \approx$$

Runge-Kutta Method of order four

The local truncation error of the R-K method of order 4 is $O(h^5)$.

$$\begin{aligned} y' &= y \\ y_{(0)} &= 1 \\ h &= 0.01 \\ y(0.04) &=? \end{aligned} \quad \left. \begin{array}{l} \text{Solve} \\ \text{with} \\ \text{assignment} \end{array} \right\}$$

Euler's Method

Recall the Taylor series expansion

$y(x_{n+1})$ is around x_n as

$$y(x_{n+1}) = y(x_n) + hy'(x_n) + \frac{h^2}{2!} y''(x_n) \dots + \frac{h^p}{p!} y^{(p)} + O(h^p)$$

Take $p < 1$ and approximate $y(x_{n+1})$ as

$$y_{n+1} = y_n + hy'_n$$

$$= y_n + hf(x_n, y_n) \quad \left\{ x = 0, 1, 2, \dots \right\}$$

Truncation error

$$T.E = \frac{h^2}{2!} y''(\xi_p) \quad x_n \leq \xi_p < x_{n+1}$$

Eg/ Consider the IVP,

$$y'(x) = y$$

$$y(0) = 1$$

and solve using Euler's method to find the approximate value of $y(0.04)$ with step size $h=0.01$. Compare with exact solution and find the error.

Sol/ Consider $f(x, y) = y$ then $x_0 = 0, y_0 = 1$

The Euler formula is given by

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

$$y_n + h y_n \quad \text{y}'(x) = y : y = e^x$$

n	x_n	y_n	Exact values	Error
0	0	1	1	0
1	0.01	1.01	1.01005	0.00005
2	0.02	1.0201	1.02020	0.00020
3	0.03	1.0303	1.03045	0.00154
4	0.04	1.0406	1.04081	0.00207

Modified Euler Method

Consider the IVP,

$$y'(x) = f(x, y) \quad \text{①}$$

$$y(x_0) = y_0$$

Integral representation of ① over $[x_n, x_{n+1}]$

$$y(x_{n+1}) = y(x_n) + \int_{x_n}^{x_{n+1}} f(x, y) dx$$

$$y(x_{n+1}) = y(x_n) + \int_{x_n}^{x_{n+1}} f(x, y) dx \quad \textcircled{1}$$

$$y(x_n) + (x_{n+1} - x_n) f(x_n, y) \quad \text{approximate} \quad \textcircled{2}$$

If we approximate $\textcircled{1}$ by $\textcircled{2}$ then

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

$$y(x_{n+1}) = y(x_n) + h f(x_n, y)$$

Using the integral formula over the interval $[x_n, x_{n+1}]$ and using midpoint formula

$$\int_{x_n}^{x_{n+1}} f(x, y) dx \approx f(x_n, y_n) \cdot (x_{n+1} - x_n)$$

$$= 2h f(x_n, y_n)$$

We get Euler's midpoint

$$y_{n+1} = y_n + 2h f(x_n, y_n) \quad n = 0, 1, 2, 3, \dots$$

$$y(x_{n+1}) = y(x_n) + \int_{x_n}^{x_{n+1}} f(x, y) dx$$

We approximate

$$\int_{x_n}^{x_{n+1}} f(x, y) dx = (x_{n+1} - x_n) f(x_n, y_n)$$

$$= 2h f(x_0, y_0)$$

is given by

$$y_{n+1} = y_n + \Delta h (f(x_n, y_n)) \quad \textcircled{1}$$

put $n=1$ in $\textcircled{1}$

$$y_2 = y_0 + 2h(x, y_1)$$

$$= y_0 + 2hy_1 = 1 + 2 \times 0.01 \times 0.61$$

put $n=2$ in $\textcircled{1}$

$$\left\{ \begin{array}{l} y_3 = y_1 + 2h y_2 \\ \hline \end{array} \right.$$

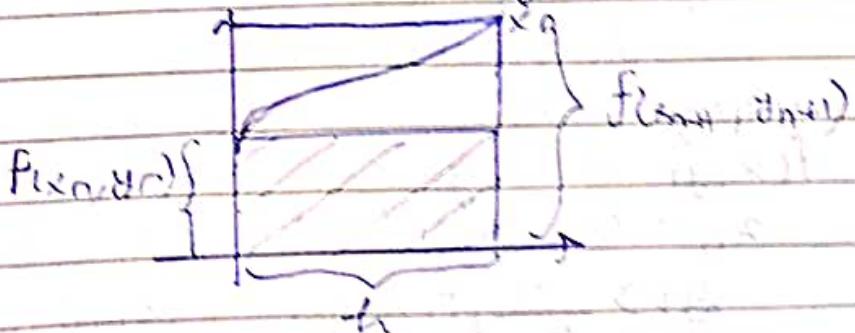
x_n	y_n	$f(x_n, y_n)$	y_{n+1}
0	0	1	
1	0.01	1.01	
2	0.02	1.0201	1.0202
3	0.03	1.0201	1.0304

Consider the IVP

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

$$y(x_0) = y_0$$

The integral of $\textcircled{1}$ over the interval $[x_0, x_{n+1}]$ is $y(x_{n+1}) = y(x_0) + \int_{x_0}^{x_{n+1}} f(x_n, y_n) dx$



If we approximate the integral of

$$\int_{x_n}^{x_{n+1}} f(x, y_n) \approx h(x_n, y_n)$$

x_n

We approximate y_{n+1} as

$$y_{n+1} = y_n + h(x_n, y_n) \cdot h \quad \text{[Euler's method]}$$

Now, if we approximate as $\int_{x_n}^{x_{n+1}} f(x, y_n) dx = h f(x_n, y_n)$

We can approximate y_n as

$$y_{n+1} = y_n + h f(x_n, y_n) \quad [Backward Euler's Method]$$

If we take $\int_{x_n}^{x_{n+1}} f(x, y) \approx \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1})]$

We approximate value of y_n as

$$y_{n+1} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1})] \quad [n=0, 1, 2]$$

[Euler's trapezoidal Method]

eg,

Find the approximate value of $y(0.4)$ of the IVP.

$$y' = xy$$

$$y(0) = 1$$

using Euler's trapezoidal method with $h=0.2$

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

$$x_0 = 0, y_0 = 1, x_1 = 0.2, x_2 = 0.4$$

$$y_1 = \frac{h}{2} [f(x_0, y_0) + f(x_1, y_1)] + y_0$$

$$y_1 = \frac{0.2}{2} [0 + (0.2)y_1] + 1$$

$$y_1 = 0.1 (0.2)y_1 + 1$$

$$y_1 = 0.02y_1 + 1$$

$$0.98y_1 = 1$$

$$y_1 = \frac{100}{98} = 1.0204$$

$$y_2 = y_1 + \frac{h}{2} (f(x_n, y_n) + f(x_{n+1}, y_{n+1}))$$

$$= 1.0204 + 0.05 (0.1)(0.2)(1.0204) + 0.4y_1$$

$$y_2 = 1.0204 + 0.0204 + 0.04y_2$$

$$0.96y_2 = 1.0408$$

$$\boxed{y_2 = 1.0841}$$

$\left\{ \begin{array}{l} y \\ y(0.4) = 1.0841 \end{array} \right\}$ Hence approximate value
of $y(0.4) = 1.0842$

$$y_{n+1}^{(0)} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1})]$$

$$y_1^{(1)} = y_0 + h [f(x_0, y_0) + f(x_1, y_1)]$$

more \Downarrow
accuracy $y_1^{(2)} = y_0 + h [f(x_0, y_0) + f(x_1, y_1)]$

$\left\{ \begin{array}{l} \downarrow \\ \text{correction} \\ \text{step} \end{array} \right\}$

Predictor - Corrector Method for
Euler's trapezoidal

$$P: y_{n+1}^{(0)} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1})]$$

$$C: y_{n+1}^{(1)} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1}^{(0)})]$$

Q9

Obtain an approximate value of

$$y(0.2) \text{ for IVP}$$

$$y' = x^2 + y^2$$

$$y(0) = 1$$

using Euler trapezoidal Predictor corrector method
with $h = 0.1$

(Q9)

$$y_1 = y_0 + \frac{h}{2} [f(x_0, y_0) + f(x_1, y_1)]$$

$$y_1 = 1 + \frac{0.1}{2} \left[1 + (0.1)^2 + y_1^2 \right]$$

(Q9)

$$\text{solve } 1 + 0.05 + 0.0005 + 0.05y^2 = 1 + 0.1^2 + y_1^2$$

$$\Rightarrow 0.05 + 0.05y^2 - y + 0.05005 = 0$$

$$y_1 \approx 1.11$$

$$y_1^{(1)} = y_0 + \frac{h}{2} [f(x_0, y_0) + f(x_1, y_1^{(0)})]$$

$$= 1 + \frac{0.1}{2} \left[1 + (0.1)^2 + (1.1)^2 \right]$$

$$y_1^{(1)} \approx 1.111$$

$$y_1^{(2)} = 1 + \frac{0.1}{2} \left[1 + 0 + (0.1)^2 + (1.11)^2 \right]$$

$$y_1^{(2)} \approx 1.112$$

$$y_2 = y_1 + \frac{h}{2} [f(x_1, y_1) + f(x_2, y_2)]$$

$$y_2 = 1.112 + 0.1 \left(\frac{0.1}{1.112} \right)$$

$$y_2 \approx 1.236$$

$$y_2^{(1)} = y_0 + \frac{h}{2} (f(x_1, y_1) + f(x_2, y_2))$$

$$= 1.112 + \frac{0.1}{2} ((0.1)^2 + (1.1)^2 + (0.2)^2 + (1.234)^2)$$

$$y_2^{(1)} = 1.2527$$

$$y_2^{(2)} = y_1 + \frac{h}{2} (f(x_1, y_1) + f(x_2, y_2))$$

$$= 1.254$$

| Predictor Correction method for Backward Euler

~~$$P: y_{n+1} = y_n + \frac{h}{2} [f(x_n, y_n) + f(x_{n+1}, y_{n+1})]$$~~

~~$$L: y_n$$~~

$$\underline{P}: y_{n+1} = y_n + h[f(x_n, y_n)]$$

$$\underline{C}: y_{n+1} = y_n + h[f(x_{n+1}, y_{n+1})]$$

Consider the IVP,

$$y' = f(x, y) \quad (1)$$

$$y(x_0) = y_0$$

Solve (1) on $[x_0, b]$ into n parts

Divide $[x_0, b]$ into n -sub intervals

$$x_0 < x_1 < x_2 < \dots < x_n = b$$

Milne's P-C method

$$y_{n+1}^{(P)} = y_{n-3} + \frac{4h}{3} \left[2f(x_n, y_n) - f(x_{n-1}, y_{n-1}) + 2f(x_{n-2}, y_{n-2}) \right] \\ n = 3, 4, 5, \dots$$

$$y_{n+1}^{(C)} = y_{n-1} + \frac{h}{3} \left[f(x_n, y_n) + 4f(x_{n-1}, y_{n-1}) + 2f(x_{n-2}, y_{n-2}) \right]$$

We have,

$$y_4^{(P)} = y_0 + \frac{4h}{3} \left[2f(x_3, y_3) - f(x_2, y_2) + 2f(x_1, y_1) \right]$$

~~eg~~ Find an approximate value of $y(0.8)$ using Milne's P-C method for IVP.

$$y = (x^2 + ty^2) \\ y(0) = 1$$

with $h = 0.2$.

$$\rightarrow h = 0.2$$

$$x_0 = 0 \quad x_1 = 0.2 \quad x_2 = 0.4 \quad x_3 = 0.6 \quad x_4 = 0.8 \\ y_0 = 1$$

Using Euler's method,

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

to calculate y_1, y_2, y_3

$$f = y_1 = y_0 + hf(x_0, y_0) \\ = 1 + (0.2)(0^2 + 1^2) = 1.2$$

$$y_2 = y_1 + hf(x_1, y_1)$$

$$= 1.2 + (0.2)(0.2^2 + 1.2^2) \approx 1.496$$

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

$$\begin{aligned} &= 1.496 + (0.2) \left(0.4^2 + 1.496^2 \right) \\ &= 1.9756 \end{aligned}$$

put $n=3$

~~Step~~

$$y_4^{(0)} = y_0 + \frac{4h}{3} \left[2f(x_3, y_3) - f(x_2, y_2) + 2f(x_1, y_1) \right]$$

$$= 1 + \frac{4 \times 0.2}{3} \left[2 \times 1.9756 - 1.496 + 2 \times 1.2 \right]$$

$$= 3.424$$

~~k+1~~

$$y_4^{(1)} = y_2 + \frac{h}{3} \left[f\left(\frac{x_2+x_3}{2}, y_2\right) + 4f(x_3, y_3) + f(x_4, y_4^{(0)}) \right]$$

$$= 3.6169$$

$\{k=0, 1, 2, \dots\}$

$$y_4^{(2)} = y_0 + \frac{h}{3} \left[f(x_2, y_2) + 4f(x_3, y_3) + f(x_4, y_4^{(1)}) \right]$$

$$y_4^{(2)} = 3.7074$$

$$y_4^{(3)} = 3.7078$$

$$y_4^{(8)} = 3.7958$$

$y_1 \Rightarrow \text{approximate}$

$y(x_1)$

$y_2 \Rightarrow y(x_2)$

$y_n \Rightarrow y(x_n)$

To compute for y_5 first

$y_1, y_2, y_3 \rightarrow \text{Euler}$

$y_4 \rightarrow P-C \text{ millinear method}$

then use y_5 from \uparrow method

Adam-Basch & Adam Mid-Point Methodmethod

$$P: y_{n+1} = y_n + \frac{h}{24} \left[55f(x_n, y_n) - 51f(x_{n-1}, y_{n-1}) + 37f(x_{n-2}, y_{n-2}) - 9f(x_{n-3}, y_{n-3}) \right] \quad (1)$$

$n = 3, 4, 5, \dots$

$$C: y_{n+1} = y_n + \frac{h}{24} \left[f(x_n, y_n) - 5f(x_{n-1}, y_{n-1}) + 19f(x_{n-2}, y_{n-2}) + 9f(x_{n-3}, y_{n-3}) \right] \quad (2)$$

$n = 3, 4, 5, \dots \quad k = 0, 1, 2$

Use Euler's method to find ~~y_0, y_1, y_2, y_3~~ Use (1) to find $y_4^{(0)}$ $y_4^{(k+1)}$ can be computed by (2) $y_{4+k+1}^{(0)}$ can be computed by (1) using $y_4^{(0)}$ $y_{4+k+1}^{(1)}$ can be computed by (2) using $y_4^{(0)}$ (Q) Find approximate value of $y(1.4)$ using Adam's P-C
method for IVP

$$y' = x^2 + y^2$$

$$y(1) = 1$$

with $h = 0.1$

[Assignment 4]

Tut

To find eigen value of A closest to 5 (using power method)

so $B = A - 5I$

smallest eigen value of B is closest eigen value of A

$\therefore B^{-1}$ (power method)

Q if $\lambda_1, \lambda_2, \lambda_3 \dots \lambda_n \Rightarrow$ eigen values of A
 $\lambda_1 - 5, \lambda_2 - 5, \lambda_3 - 5 \dots \lambda_n - 5 \Rightarrow$ eigen values of B

x ————— x ————— B x ————— x

Boundary Value Problem (BVP)

Consider the differential equation

$$y'' = f(x, y, y') \quad x \in (a, b)$$

with boundary

Type 1: $y(a) = A$

$$y(b) = B$$

Type 2: $y'(a) = A$

$$y'(b) = B$$

Type 3:

$$\begin{cases} c_1 y(a) + c_2 y'(b) = A \\ b_1 y(b) + b_2 y'(b) = B \end{cases} \quad \text{Mixed}$$

$c_1, c_2, b_1, b_2, A, B \rightarrow$ constants

$$\boxed{y''(x_n) = \frac{y_{n+1} - 2y_n + y_{n-1}}{h^2}}$$

$$y'(x_n) \approx \frac{y_{n+1} - y_{n-1}}{2h}$$

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Finite Difference Method (FDM)

Consider the BVP,

$$c(x)y'' + b(x)y' + a(x) = d(x) \quad x \in [a, b]$$

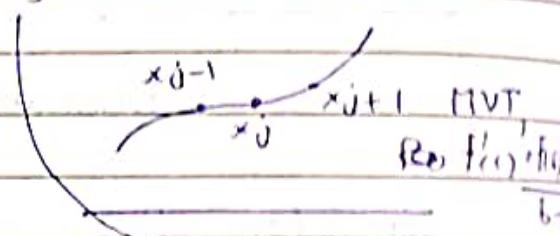
$$y(a) = A$$

$$y(b) = B$$

(1)

Let $a = x_0 < x_1 < \dots < x_N < x_{N+1} = b$ be a uniform partition of $[a, b]$ such that

$$x_{j+1} = x_j + h$$



$$\therefore y(x_{j+1}) = y(x_j) + hy'(x_j) + \frac{h^2}{2!} y''(x_j) \quad y'(x_j) \approx y(x_{j+1}) - y(x_{j-1})$$

Euler

$$\text{Substitute: } y'(x_j) = \frac{y(x_{j+1}) - y(x_{j-1})}{2h}$$

$$y'(x_j) = \frac{y(x_{j+1}) - y(x_{j-1})}{2h}$$

$$\text{Add: } y(x_{j+1}) + y(x_{j-1}) = 2y(x_j)$$

$$2y(x_j) + h^2 y''(x_j)$$

$$+ h^2 y'(x_j)$$

$$y'(x_j) = \frac{y(x_{j+1}) - 2y(x_j) + y(x_{j-1})}{h^2}$$

$$\text{Approx: } y_{j-1} - (2 + h^2) y_j + y_{j+1} = h^2 x_j$$

After rearranging, we get

Let $y''(x)$

We can approximate $y''(x_j)$ by $\frac{y_{j+1} - 2y_j + y_{j-1}}{h^2}$

consider $y(x_j)$ by

$$\rightarrow \frac{y_{j+1} - y_{j-1}}{2h} \quad (\text{Central fr})$$

$$\rightarrow \frac{y_{j+1} - y_j}{h} \quad (\text{Euler})$$

$$\rightarrow \frac{y_j - y_{j-1}}{h} \quad (\text{Backward Euler})$$

Thus we can approximate ① by

$$a(x_j) \left[\frac{y_{j+1} - 2y_j + y_{j-1}}{2h^2} \right] + b(x_j) \left[\frac{y_{j+1} - y_{j-1}}{2h} \right]$$

$$+ c(x_j)y_j = d(x_j) \quad j=1, 2, 3, \dots, N$$

$$\left[a(x_j) + \frac{h}{2} b(x_j) \right] y_{j+1} + \left[-\frac{h}{2} 2a(x_j) + h^2 c(x_j) \right] y_j$$

$$+ \left[a(x_j) - \frac{h}{2} b(x_j) \right] y_{j-1} = d(x_j) \quad \forall j=1, 2, \dots$$

There are N eqns and N unknown variables

which can be solved by using my known technique.

example — Solve the BVP,

$$y'' = y + x_j \quad x \in (0, 1)$$

$$y(0) = 0$$

$$y(1) = 0$$

with $h = 1/4$ using second order FDM.

$$\rightarrow \text{Let } x_0 = 0, x_1 = 0.25, x_2 = 0.5, x_3 = 0.75, x_4 = 1 \\ y_0 = 0 \quad \underbrace{y_1, y_2, y_3}_{\text{Unknown}} \quad \dots \quad y_4 = 0$$

Here y_1, y_2, y_3 are unknowns that approximate y at x_1, x_2, x_3 respectively

The approximate system will be,

$$\frac{y_{j+1} - 2y_j + y_{j-1}}{h^2} = y_j + x_j \quad j = 1, 2, 3$$

$$\frac{y_2 - 2y_1 + y_0}{h^2} = y_1 + x_1$$

$$y_2 - 2y_1 + \frac{1}{16} = (y_1 + 1/4)$$

$$y_2 - 2y_1 - \frac{y_1}{16} = \frac{1}{64}$$

$$y_2 - \frac{33y_1}{16} = \frac{1}{64} \quad \text{--- (1)}$$

$$\frac{y_3 - 2y_2 + y_1}{h^2} = y_2 + x_2$$

$$y_3 - 2y_2 + y_1 \approx \frac{1}{16} \left(y_2 + \frac{1}{2} \right)$$

$$y_3 - 2y_2 + y_1 - \frac{1}{16} y_2 = \frac{1}{32}$$

$$\frac{y_3 - 33y_2}{16} + y_1 = \frac{1}{32} \quad \text{--- (2)}$$

$$\frac{y_4 - 2y_3 + y_2}{h^2} = y_3 + x_3$$

$$0 - 2y_3 + y_2 \approx \frac{1}{16} (y_3 + 3/4)$$

$$\frac{y_2 - 2y_3}{16} = \frac{1}{16} y_3 = \frac{3}{64}$$

$$\frac{y_2 - 33y_3}{16} = 3/64 \quad \text{--- (3)}$$

$$\begin{cases} y_1 = -0.035 \\ y_2 = -0.035 \\ y_3 = -0.050 \end{cases}$$

Example: Solve BVP

$$y'' = xy + 1 \quad x \in [0, 1]$$

$$y'(0) = y(0) \approx 1 \quad (\text{or } y'(0) = 7)$$

$$y(1) \approx 1 \quad (\text{or } y(1) + 3y'(1) \approx 5)$$

use FDM, to find approximate value of
 $y(0), y(0.25), y(0.5), y(0.75)$

$$\text{Soln: } h = 0.25$$

$$x_0 = 0 \quad x_1 = 0.25 \quad x_2 = 0.5 \quad x_3 = 0.75$$

$$y_0 \approx y(0)$$

$$y_1 \approx y(0.25)$$

$$y_2 \approx y(0.5)$$

$$y_3 \approx y(0.75)$$

$$y_{j+1} = y_j + h$$

$$\frac{y_{j+1} - 2y_j + y_{j-1}}{2h} = x_j y_j + r$$

$$\frac{y_2 - 2y_1 + y_0}{2h} = x_1 y_1 + r$$

$$\frac{y_2 - 2y_1 + y_0}{2h} = \frac{y_1 + 1}{2}$$

$$y_2 - 2y_1 + y_0 = \frac{y_1 + 1}{2}$$

$$y_2 - 17y_1 + y_0 = \frac{1}{2} - ①$$

$$\frac{y_3 - 2y_2 + y_1}{2h} = -x_2 y_2 + 1$$

$$\frac{y_3 - 2y_2 + y_1}{2h} = \frac{1}{2} y_2 + 1$$

$$y_3 - 2y_2 - \frac{1}{2} y_2 + y_1 = \frac{1}{2}$$

$$y_3 - \frac{9}{4} y_2 + y_1 = \frac{1}{2} - ②$$

$$\underline{y_4 - 2y_3 + y_2 = x_3 y_3 + 1}$$

$$\alpha h = \frac{1}{2}$$

$$y_4 - 2y_3 + y_2 = \frac{3}{8} y_3 + \frac{1}{2}$$

$$y_4 - \frac{19}{8} y_3 + y_2 = \frac{1}{2} - \textcircled{3}$$

Use $y'(0) \stackrel{\text{MVT}}{=} y'(0) + y(0) = 1$

$$\underline{y_1 - y_{-1} = + y_0 = 1} \quad \text{(crossed out)}$$

$$\cancel{2} \cdot \underline{y_1 - 2y_0 + y_{-1} = + 2y_0 = 2} \quad \text{--- } \textcircled{3}$$

$$\cancel{2} \cdot \underline{y_1 - 2y_0 + y_{-1} = x_0 y_0 + 1}$$

$$y_1 - 2y_0 - y_{-1} = \frac{1}{2} \cdot (0) + 1$$

$$(x_0 - x_1) \cdot y_1 - 2y_0 - y_{-1} = 1 \quad \text{--- } \textcircled{4}$$

$\therefore 8 \cdot (2)$

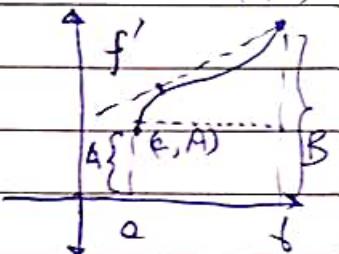
Shooting Method

Consider a second order BVP

$$y'' + p(x)y' + q(x)y = r(x) \quad x \in (a, b)$$

$$y(0) = A, \quad y(b) = B$$

\therefore eq: $y'' = 6y^2 - x$
 $y(0) = 1, \quad y(1) = 5 \quad h = 1/3$



Sol: Choose $y'(0) = \alpha_0 = 1.2$

$$\text{let } y' = z$$

$$z' = 6y^2 - x$$

$$z(0) = 1.2$$

We can use R.K. method, Euler method, Taylor series
to solve I.V.P.S.

How we can use R.K. method

$$\text{Here } z_0 = 0 \quad y(x_1) = 1/3, \quad x_2 = 2/3 \dots$$

$$z' = 6y^2 - x$$

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

$$z_{n+1} = z_n + h(6y^2 - x)$$

$$z_{n+1} = z_n + h \left[1 + \frac{1}{3} (1.2) \right]$$

$$y_1 = 1 + \frac{1}{3} (1.2) \\ = 1.4$$

$$z_1 = z_0 + h(6y_1^2 - x_1) \\ = 1.2 + \frac{1}{3} (6) = 3.2$$

$$y_2 = y_1 + h z_1$$

$$= 1.4 + \frac{1}{3} (3.2)$$

$$= 3.466$$

$$z_2 = z_1 + h(6y_2^2 - x_2) \\ = 3.2 + \frac{1}{3} (6 \times 1.4) =$$

$$= 7.01$$

$$y_3 = y_2 + h z_2$$

$$= 3.466 + \frac{1}{3} \times 7.01$$

$$= 4.7956$$

$$y(x_3) = 4.7956 \quad \{ \text{approx.} \}$$

$$y(x_3) = 5$$

$$\phi(\alpha_0) = y(1) - y(1; \alpha_0)$$

$$= 5 - 4.7966$$

$$\text{Error} = 0.2034$$

Now consider the initial guess

$$y'(0) = 1.5$$

Again solve ① with $\alpha_1 = 1.5$ by replacing α_0 by α_1

$$y' = z \quad y(0) = 1$$

$$y_1 = y_0 + \frac{h}{3}(2z_0 + z_1)$$

$$= 1 + \frac{1}{3}x^{1.5}$$

$$= 1.5$$

$$y_2 = y_1 + h z_1 = 1.166$$

$$= 1.5 + \frac{1}{3}x^{3.5}$$

$$= 1.5 + 1.167$$

$$= 1.667$$

$$y_3 = y_2 + h z_2$$

$$= 1.67 + \frac{1}{3}x^{6.5}$$

$$= 5.29$$

$$z_{n+1} = z_n + h(6y_n - x_n)$$

$$z_1 = z_0 + h(6y_0 - x_0)$$

$$= 1.5 + \frac{1}{3}x^{6.2}$$

$$= 3.5$$

$$z_2 = z_1 + h(6y_1 - x_1)$$

$$= 3.5 + \frac{1}{3}(8 \times 1.5)$$

$$= 6.5$$

$$\phi(\alpha_0) = 5 - 5.29$$

$$= -0.29$$

Note - If it is asked to use N.R method

$$\alpha_1 = \alpha_0 - \frac{\phi(\alpha_0)}{\phi'(\alpha_0)}$$

using secant method,

$$\alpha_2 = \alpha_1 - \frac{(\alpha_1 - \alpha_0) f(\alpha_1)}{f(\alpha_1) - f(\alpha_0)}$$

$$\Rightarrow \alpha_2 = 1.323$$

$$z(0) = \alpha_2 = 1.32$$

$$y' = \exists \quad y(0) = 1$$

$$z' = 6y^2 - x$$

$$y_1 = y_0 + h z_0 \\ = 1 + \frac{1}{3} \times 1.32 \\ = 1.44$$

$$z_1 = z_0 + h(6y_0 - x_0) \\ = 1.32 + \frac{1}{3}(6)$$

$$y_2 = y_1 + h z_1 \\ = 1.44 + \frac{1}{3} \times 3.32 \\ = 2.546$$

$$z_2 = z_1 + h(6y_1 - x_1) \\ = 3.32 + \frac{1}{3}(6)$$

$$y_3 = y_2 + h z_2 \\ = 2.546 + \frac{1}{3} \times 7.316 \\ = 4.98$$

$$z_3 = z_2 + h(6y_2 - x_2) \\ = 7.316 + \frac{1}{3}(6)$$

Q) $\frac{d^2y}{dx^2} - y = 3x \quad x \in (0, 6)$
 $y(0) = 0$
 $y(6) = 0$

using shooting method, $h = 2$

use Euler's method to solve IVP.

Use linear-interpolation to obtain subsequent initial condition $y'(0) = 3$

$$y \quad y''(0) = 3$$

Let $z = y'$

$$\text{I.V.P. 2: } y' = z$$

$$y(0) = 0$$

I.V.P. 2:

$$\begin{aligned} z' - y &= 3x \\ z(0) &= 3 \end{aligned}$$

$$\begin{aligned} z_0 &= 0 & z_1 &= 2 & z_2 &= 4 & z_3 &= 6 \\ h &= 2 \end{aligned}$$

I.V.P. 1

$$y' = z$$

$$y(0) = 0$$

I.V.P. 2

$$z' - y = 3x$$

$$z(0) = 3$$

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

$$y_1 = y_0 + h z_0$$

$$= 0 + 2 \times 3$$

$$= 6$$

$$y_2 = y_1 + h z_1$$

$$y_2 = 6 + 2 \times 3$$

$$= 12$$

$$y_3 = y_2 + h z_2$$

$$= 12 + 2 \times 27$$

$$\begin{array}{r} 54 \\ 12 \\ \hline 66 \end{array}$$

$$\begin{aligned} z_{n+1} &= z_n + (y_n + 3x_n)h \\ z_1 &= z_0 + (y_0 + 3x_0)(2) \\ &= 3 + (0 + 0)(2) \\ &= 3 \end{aligned}$$

$$z_2 = z_1 + h(y_1 + 3x_1)$$

$$= 3 + 2(6 + 0)$$

$$= 27$$

For $y'(0) = -3$

$$y' = z$$

$$y_1 = z$$

$$y(0) = 0$$

$$y_1 = y_0 + hz_0 = -6,$$

$$y_2 = y_1 + hz_1$$

$$= -6 + 2 \times (-3)$$

$$= -18$$

$$y_3 = y_2 + hz_2$$

$$= -18 + 2 \times (-3)$$

$$= -24$$

$$\approx -18$$

$$z' = y' + 3x$$

$$z(0) = -3$$

$$z_1 = z_0 + h(y_0 + 3x_0)$$

$$= -3 + 2(0 + 0)$$

$$= -3$$

$$z_2 = z_1 + h(y_1 + 3x_1)$$

$$= -3 + 2(-6 + 3(2))$$

$$= -3$$

$$\alpha_0 = 3, \alpha_1 = -3$$

$$\xi_0 = 66 - 0, \xi_1 = 11 - 0$$

$$= 66$$

$$= 18$$

Linear Interpolation Formula

$$\alpha_x = \alpha_0 + \frac{(\alpha_1 - \alpha_0)}{(\xi_1 - \xi_0)} (y(6) - y_0)$$

$$\approx \cancel{-18} - \cancel{56}$$

$$= -1.7914$$

$$y' = z'$$

$$y'(0) = 0$$

$$y_1 = y_0 + hz_0$$

$$= 0 + 2 \times (-1.7914)$$

$$= -3.428$$

$$y_2 = y_1 + hz_1$$

$$= -3.428 + 2(-1.714)$$

$$= -6.856$$

$$z' = y' + 3x$$

$$z(0) = -1.714$$

$$z_1 = z_0 + h(y_0 + 3x_0)$$

$$= -1.714 + 2(0 + 3(0))$$

$$= -1.714$$

$$z_2 = z_1 + h\alpha_1 (y_0 + 3x_0)$$

$$= -1.714 + 2(\cancel{-1.714})$$

$$= -3.429$$

$$y_3 = y_2 + h y_2 \\ \Rightarrow 0.04$$

Shooting method using Newton-Raphson method

let $y'(a) = \alpha$ be the additional initial form.

let $y(b, \alpha) \rightarrow$ approximate value of $y(b)$ using shooting method with $y'(a) = \alpha$

$$\text{let } \phi(\alpha) = y(b, \alpha) - \gamma$$

$$\left\{ \begin{array}{l} b \Rightarrow \text{approximate value} \\ \alpha \Rightarrow \text{initial value guess} \end{array} \right\} \quad \left\{ \begin{array}{l} y(b) \\ \text{desired value} \end{array} \right\}$$

Compute $\phi'(\alpha)$

$$\text{Denote } \frac{d\phi}{d\alpha} = \frac{dy}{dx}(b, \alpha) - \frac{dy}{dx}^0 \\ \Rightarrow \frac{dy}{dx}(b, \alpha)^2 \eta_\alpha(b)$$

$$\eta_\alpha'' = \frac{df}{dy} \eta_\alpha + \frac{df}{dy'} \eta_{\alpha'}$$

$$\eta_\alpha(0) = 0 \\ \eta_{\alpha'}(0) = 1$$

Example :- $y'' = y$

$$y(0) = 1, y(1) = 0 \\ f(x, y, y') = y$$

$$\left\{ \begin{array}{l} \eta_\alpha'' = \eta_\alpha \\ \eta_\alpha(0) = 0, \eta_\alpha(1) = 0 \end{array} \right\}$$

Let $\alpha \sim 1/3$,

$$\mu = n'_\alpha$$

$$n'_\alpha = \mu$$

$$n_\alpha(\alpha) = 0$$

$$H(\alpha) = -n'_\alpha T(\alpha) \leq$$

$$n_{\alpha_1} = n_{\alpha_0} + h\mu_0$$

$$= 0 + \frac{1}{3}(1)$$

$$= 1/3$$

$$n_{\alpha_2} = n_{\alpha_1} + hH_1$$

$$= 1/3 + 1/3$$

$$= 2/3$$

$$n_{\alpha_3} = n_{\alpha_2} + hH_2$$

$$= 2/3 + 1 \times \frac{10}{9}$$

$$= \frac{18}{27} + \frac{10}{27}$$

$$= \frac{28}{27} \rightarrow \phi'(n'_\alpha)$$

$$\alpha_{n+1} = \alpha_n - \frac{\phi(\alpha_n)}{\phi'(\alpha_n)}$$

Probability

Random experiments

A random experiment is an experiment such that

- (1) All outcomes of experiment is known in advance.
- (2) Any performance of results is not known in advance.
- (3) The experiment can be repeated under identical condition.

Sample Space

A set S that consists of all possible outcome of a random experiment is called sample space and each outcome is called sample point.

Ex we toss a dice,

$$\text{Sample space} = \{1, 2, 3, 4, 5, 6\}$$

$$S_2 = \{\text{odd, even}\}$$

Event -

An event is a subset of the sample space, it is a subset of all possible outcomes.

example: $A = \{1, 6\} \rightarrow$ is an event.

Class: Set of all subsets of a sample space

$$C \xrightarrow{\phi} [0, 1]$$

$$\downarrow P(A_i)$$

$C \rightarrow \{\frac{1}{2}, 1, 0\}$
 \downarrow
 $S = \{(H,H), (H,T), (T,H), (T,T)\}$

$$0 \leq P(A) \leq 1$$

Axioms of Probability

To each event A in the class C of the ~~outcomes~~ we associate a real no. $P(A)$ called probability of A provided the following axioms are satisfied.

A1: For ~~suit~~ every event A in C , $P(A) \geq 0$

A2: For sure/certain event C , $P(C) = 1$

A3: For no. of mutually exclusive events A_1, A_2, \dots , $P(A_1 \cup A_2 \cup \dots \cup A_n) = P(A_1) + P(A_2) + \dots + P(A_n)$

Theorems : i) If $A_1, A_2 \in C$ then $P(A_1 \cap A_2) = P(A_1)P(A_2)$
 $P(A_1/A_2) = P(A_1) - P(A_1 \cap A_2)$

- 2) For every event A , $0 \leq P(A) \leq 1$
- 3) For impossible event \emptyset , $P(\emptyset) = 0$
- 4) If A^c is complement of A , then $P(A^c) = 1 - P(A)$

5) If

$$P(A \cup B) = \begin{cases} P(A) + P(B) & \text{A \& B mutually exclusive} \\ P(A) + P(B) - P(A \cap B) & \text{A \& B are not mutually exclusive} \end{cases}$$

$A \& B$ are not
mutually exclusive

$$\begin{aligned} P(A_1 \cup A_2 \cup A_3) &= P(A_1) + P(A_2) + P(A_3) \\ &\quad - P(A_1 \cap A_2) - P(A_2 \cap A_3) \\ &\quad - P(A_3 \cap A_1) + P(A_1 \cap A_2 \cap A_3) \end{aligned}$$

For any two events A & B

$$P(A) = P(A \cap B) + P(A \cap B^c)$$

once

Q) A simple die is tossed. Find the probability of 2 & 4 turn up.

$$P(1) = 1/6$$

$$P(4) = 1/6$$

$$P(1) + P(4) = 1/3$$

Conditional Probability

Let A and B be two events such that $P(A) > 0$
 and $P(B|A) = P(B \cap A)/P(A)$ is

The probability of B given that A has occurred
 Since A is known to have occurred it becomes
 the new sample space replacing the original

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

example : Find the probability that a single toss
 of a dice will result in a number less
 than 4 if

- (a) No prior information is given
- (b) an odd number

$$P(a) = \cancel{3/6} = 1/2$$

$$P(b) = 2/3$$

Baye's Theorem

Suppose A_1, \dots, A_n are mutually exclusive events
 whose union is sample space S . Then if n
 is any event we have:

$$P(A_k | A) = \frac{P(A_k) P(A | A_k)}{\sum_{j=1}^n P(A_j) P(A | A_j)}$$

eg

A coin is tossed twice. Possible outcomes are -

$$1 \rightarrow H \text{ or } T$$

$$2 \rightarrow T \text{ or } H$$

— ①

$$\begin{matrix} S = \{HH, HT, TH, TT\} \\ \text{Sample Space} \end{matrix}$$

each element is called event

$$P(1) = 1/4$$

$$P(2) = 1/4$$

$$P(3) = 1/4$$

$$P(4) = 1/4$$

Random Variable

Let, be a random experiment and S be the corresponding sample space. A random variable function is : $X : S \rightarrow \mathbb{R}$

Suppose a coin is tossed twice solⁿ the possible sample spaces are :

$$S = \{HH, HT, TH, TT\}$$

Let X represent the number of heads that can come up with

$$X(HH) = 2$$

$$X(HT) = 1$$

$$X(TH) = 1$$

$$X(TT) = 0$$

Discrete Random Variable

A random variable that takes on a finite or countably infinite number of values is called discrete random variable.

Discrete Probability Distribution

Let X be a discrete random variable and let assume that if it takes the value x_1, x_2, \dots, x_n . Define.

$$P(X = x_i) \stackrel{=}{\sim} f(x_i)$$

Therefore is called as discrete function probability.

$$\begin{aligned} P(HH) &= \left\{ \begin{array}{l} f(HH) = 1/4 \\ f(HT) = 1/4 \end{array} \right. \\ &\quad \left. \begin{array}{l} f(TH) = 1/4 \\ f(TT) = 1/4 \end{array} \right\} \text{ refer to } ① \end{aligned}$$

$$\sum f(x_i) = 1$$

Distribution of functions

$$x_i \geq 0$$

$$\sum f(x_i) = 1$$

$$P(X \geq 1) = 3/4 \quad \text{refer to } ①$$

$$P(Y \leq 1) = 3/4$$

tails

Distribution

Discrete functions

Let X be a random variable. Then a distribution function F on X is defined as

$$F(x) = P(X \leq x) \quad \text{when } x \in \mathbb{R}$$

$$F(x_2) = 1/2 //$$

Properties:

- 1) $F(x)$ is non decreasing
- 2) $\lim_{x \rightarrow -\infty} F(x) = 0$ $\lim_{x \rightarrow \infty} F(x) = 1$

- 3) $F(x)$ is continuous from right

$$\lim_{h \rightarrow 0^+} F(x+h) = F(x)$$

Let X be a discrete variable. and f denotes the probability function. Then the distribution function is determined by —

$$F(x) = P(X \leq x) = \sum_{u \leq x} f(u)$$

If f takes only finitely many number of value x_1, x_2, x_3, \dots (in some order) then distribution function $F(x)$ is defined as —

$$F(x) = \begin{cases} 0 & x < x_1 \\ f(x_1) & x_1 \leq x < x_2 \\ f(x_1) + f(x_2) & x_2 \leq x < x_3 \\ \vdots & \vdots \\ f(x_1) + f(x_2) + \dots + f(x_n) & x_n \leq x < \infty \end{cases}$$

According to ①

$$F(x) = \begin{cases} 0 & \text{if } x < 0 \\ 1/4 & 0 \leq x < 1 \\ 3/4 & 1 \leq x < 2 \\ 1 & 2 \leq x < \infty \end{cases}$$

Continuous Random Variable

A random variable X is said to be continuous random variable if it takes all possible values in a given interval $\in \mathbb{R}$.

Example: Age, height in a large population

Probability function

$$f(x) = P(X=x)$$

if (i) $f(x) \geq 0$

$$(ii) \int_{-\infty}^{\infty} f(x) dx = 1$$

Distribution function

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(t) dt$$

$$P(0 \leq x \leq b) = \int_0^b f(x) dx$$

Q. Find the constant c such that the function

$$f(x) = \begin{cases} cx^2 & 0 < x < 3 \\ 0 & \text{otherwise} \end{cases}$$

is a probability function. Also compute $P(1 < x < 2)$

$$\int_0^3 cx^2 dx = 1$$

$$c \frac{x^3}{3} \Big|_0^3 = 1$$

$$c = \frac{1}{\frac{3 \cdot 3 \cdot 3}{3}} = \frac{1}{9}$$

$$\begin{aligned} \int_1^2 cx^2 dx &= \frac{c}{3} (x^3) \Big|_1^2 \\ &= \left(\frac{8-1}{3}\right) \cdot \frac{1}{9} \\ &= \frac{7}{27} \end{aligned}$$

$$F(x) = \int_{-\infty}^x cx^2 dx = \frac{1}{9} \frac{x^3}{3} \Big|_{-\infty}^x$$

$$= \frac{x^3}{27}$$

Suppose that a pair of fair dice are thrown and X denotes the random variable of the sum of the outcome. Find the probability function and the distribution function.

Soln:

$$P(X) \left\{ \begin{array}{ll} 0 & x=1 \\ 1/36 & x=2 \\ 2/36 & x=3 \\ 3/36 & x=4 \\ 4/36 & x=5 \\ 5/36 & x=6 \\ 6/36 & x=7 \\ 5/36 & x=8 \\ 4/36 & x=9 \\ 3/36 & x=10 \\ 2/36 & x=11 \\ 1/36 & x=12 \end{array} \right.$$

$$P(X) \left\{ \begin{array}{ll} 0 & x \leq 1 \\ 1/36 & x=2, 12 \\ 2/36 & x=3, 11 \\ 3/36 & x=4, 10 \\ 4/36 & x=5, 9 \\ 5/36 & x=6, 8 \\ 6/36 & x=7 \end{array} \right.$$

$$F(x) \left\{ \begin{array}{ll} 0 & x < 2 \\ 1/36 & 2 \leq x < 3 \\ 3/36 & 3 \leq x < 4 \\ 6/36 & 4 \leq x < 5 \\ 10/36 & 5 \leq x < 6 \\ 15/36 & 6 \leq x < 7 \\ 21/36 & 7 \leq x < 8 \\ 26/36 & 8 \leq x < 9 \\ 30/36 & 9 \leq x < 10 \\ 33/36 & 10 \leq x < 11 \\ 35/36 & 11 \leq x < 12 \end{array} \right.$$

Eg: A random variable x has the following probability function

x :	0	1	2	3	4	5	6	7
$f(x)$:	0	$2k$	$2k$	$2k$	$3k^2$	$8k^2$	$2k^2$	$37k^2 + k$

(i) Find k

$$(ii) \text{ Evaluate } P(X \geq 6), P(X < 5)$$

$$P(X \leq c) \geq \frac{1}{2}, P(0 < X < 5)$$

$$\sum f(x) = 1$$

$$0 + 3k^2 + 7k^2 + k = 1$$

$$10k^2 + 9k - 1 = 0$$

$$10k^2 + 5k + 4k - 1 = 0$$

$$10k^2 + 10k - k - 1 = 0$$

$$10k(k+1) - 1(k+1) = 0$$

$$k = -1 \text{ or } k = 0.1$$

$$P(X < 6) = P(0) + P(1) + P(2) + P(3) + P(4) + P(5)$$

$$= 0 + 0.1 + 0.2 + 0.2 + 0.3 + 0.01$$

$$= 0.81$$

$$P(X \geq 6) = 1 - P(X < 6)$$

$$= 0.19$$

$$P(X \leq c) \geq \frac{1}{2}$$

$$P(0) + P(1) \dots P(c) \geq \frac{1}{2}$$

$$0 + 0.1 + 0.2 + 0.2$$

$$F(x) \begin{cases} 0.1 & 0 < x \leq 1 \\ 0.3 & 1 < x \leq 2 \\ 0.5 & 2 < x \leq 3 \\ 0.5 & 3 < x \leq 4 \\ 0.7 & 4 < x \leq 5 \\ 1 & x > 5 \end{cases}$$

$\times \oplus$

$$C = 4$$

Joint DistributionDiscrete Case

discrete

Let X and Y be 2 random variables, we define the joint probability on X and Y by

$$f(x, y) = P(X=x, Y=y), \text{ where } f(x, y) \text{ satisfies}$$

the condition

$$\text{i)} f(x, y) \geq 0$$

$$\text{ii)} \sum_{x \in A} \sum_{y \in B} f(x, y) = 1$$

Let X can assume x_1, x_2, \dots, x_n and Y can assume

y_1, y_2, \dots, y_n

Then

$$P(X=x_j, Y=y_k) = f(x_j, y_k)$$

$$P(X=x_j) = P(X=x_j, Y=y_1) + P(X=x_j, Y=y_2) + \dots + P(X=x_j, Y=y_n) = \sum_{k=1}^n f(x_j, y_k)$$

$$\begin{aligned} P(Y=y_k) &= P(X=x_1, Y=y_k) + P(X=x_2, Y=y_k) + \dots + P(X=x_n, Y=y_k) \\ &= \sum_{j=1}^n P(X=x_j, Y=y_k) \end{aligned}$$

~~Joint - distribution~~

Let X and Y be two discrete random variables. Then

$X \setminus Y$	y_1	y_2	y_3	\dots	y_n	Total
x_1	$f(x_1, y_1)$	$f(x_1, y_2)$	\dots	\dots	$f(x_1, y_n)$	$\sum f(x_1, y_i)$
x_2	$f(x_2, y_1)$	$f(x_2, y_2)$	$f(x_2, y_3)$	\dots	$f(x_2, y_n)$	$\sum f(x_2, y_i)$
x_3	$f(x_3, y_1)$	$f(x_3, y_2)$	$f(x_3, y_3)$	\dots	$f(x_3, y_n)$	$\sum f(x_3, y_i)$
\vdots						
x_n	$f(x_n, y_1)$	$f(x_n, y_2)$	$f(x_n, y_3)$	\dots	$f(x_n, y_n)$	$\sum f(x_n, y_i)$
						$\sum f(x_k, y_1) + \sum f(x_k, y_2) + \sum f(x_k, y_3) + \dots + \sum f(x_k, y_n)$

Eg/

let the joint probability function f of two discrete random variables X and Y be:

$$f(x,y) = \begin{cases} C(2x+y) & \text{if } 0 \leq x \leq 2 \\ 0 & \text{otherwise} \end{cases}$$

when $x, y \in \mathbb{Z}$

① Find C

② Find $P(X=2, Y=1)$, $P(X \geq 1, Y \leq 2)$

$$\begin{aligned} & \int_0^3 \int_0^1 c(2x+y) dx dy \\ & c \left[\int_0^1 [x^2 + xy] \right]_0^1 dy \end{aligned}$$

$$\sum_{y=0}^3 \sum_{x=0}^1 c(2x+y)$$

	0	1	2	3
0	0	c		
1	$2c$	$3c$	$4c$	$5c$
2	$4c$	$5c$	$6c$	$7c$
	$6c$	$9c$	$12c$	$15c$

$$40c = 1$$

$$\{c = 1/40\}$$

$$\sum_{k=1}^n f(x_i, y_k)$$

$$P(X=2, Y=1) = f(2, 1) \stackrel{?}{=} 5c \quad , \quad 5/42$$

$$P(X \geq 1, Y \leq 2)$$

$$\begin{aligned} &= f(1, 0) + f(1, 1) + f(1, 2) + f(2, 0) \\ &= 4c + 5c + 6c + 7c + f(2, 1) + f(2, 2) \\ &= 25c \end{aligned}$$

$$= 2c + 3c + 4c + 4c + 5c + 6c$$

$$= 5c + 8c + 11c$$

$$= 24c$$

$$= 24/42$$

$$= 4/7$$

~~For~~ The joint distribution function ~~or~~ / ~~and~~ cumulative joint distribution function is

$$F(x, y) = P(X \leq x, Y \leq y) = \sum_{u \leq x} \sum_{v \leq y} f(u, v)$$

Continuous Case

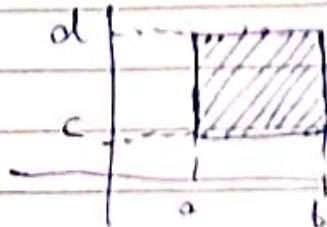
Let X and Y be two continuous functions random variables. Then the probability that X lies in between a and b , Y lies between c and d . Then the probab.

$$\text{is given by, } P(a < x < b, c < y < d) = \iint_{a \leq x \leq b} f(x,y) dx dy$$

Here $f(x,y)$ satisfies :

$$(i) f(x,y) \geq 0$$

$$(ii) \iint_{-\infty}^{\infty} f(x,y) dx dy = 1$$



Note : If A represents any event there will be a region R_A in $x-y$ plane such that

$$P(A) = \iint_R f(x,y) dx dy$$

Example 2:

Let the joint distribution function of random variables X and Y be:

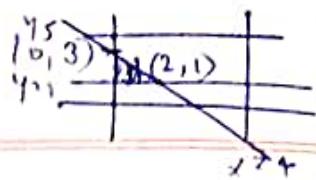
$$f(x,y) = \begin{cases} \frac{1}{96} xy & 0 < x < 4 \\ & 1 < y < 5 \\ 0 & \text{otherwise} \end{cases}$$

$$\text{Find } P(X+Y < 3)$$

~~$$4 \int_0^{3-x} \int_1^{3-x} \frac{1}{96} xy dy dx$$~~

$$= \left[\frac{1}{96} \left(\frac{xy^2}{2} \right) \right]_1^{3-x} dx$$

$$\begin{aligned} y &= 2-x+3 \\ 1 &= -x+3 \\ x &= 2 \end{aligned}$$



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$$\frac{1}{96 \times 2} \int_0^4 x [(3-x)^2 - 1] dx$$

$$\frac{1}{96 \times 2} \int_0^4 x (9+x^2-6x) - x dx$$

$$\frac{1}{96 \times 2} \int_0^4 8x + x^3 - 6x^2 dx$$

$$\frac{1}{96 \times 2} \left(\frac{8x^2}{2} + \frac{x^4}{4} - \frac{6x^3}{3} \right) \Big|_0^4$$

$$\frac{1}{96 \times 2} [4(4)^2 + 4^4 - 2(4)^3]$$

$$64 + 64 \neq -128$$

$$\int_0^2 \int_{3-x}^1 \frac{1}{96} xy dy dx$$

$$\frac{1}{96} \int_0^2 x \left\{ \left(\frac{y^2}{2} \right) \right\}_{3-x}^1 dx$$

$$\frac{1}{96 \times 2} \int_0^2 x ((3-x)^2 - 1) dx$$

$$\frac{1}{96 \times 2} \int_0^2 8x + x^3 - 6x^2 dx$$

$$\frac{1}{2} \times \frac{1}{96} \left(4x^2 + \frac{x^4}{4} - 62x^3 \right)_0^2$$

$$\frac{1}{2} \times \frac{1}{96} \left[4 \times 4 + \cancel{8} \frac{4 \times 4}{4} - 2 \times 4 \times 2 \right]$$

$$\frac{1}{2} \times \frac{1}{96} [16 + 4 - 16]$$

$$\frac{1}{2} \times \frac{1}{24} \quad , \quad \frac{1}{48}$$

Note : $\frac{\partial^2 F}{\partial x \partial y} = f(x, y)$

$$P(X \leq x) = \int_{-\infty}^x \int_{-\infty}^y f(u, v) du dv$$

$$P(Y \leq y) = \int_{-\infty}^y \int_{-\infty}^x f(u, v) du dv$$

When $F_1(x)$ and $F_2(y)$ are called marginal distribution function or distribution function of x and y .

Q) Find distribution function and marginal distribution function.

$$x < 0 \quad y < 1$$

$$x < 0 \quad 1 \leq y \leq 5$$

$$x < 0 \quad y > 5$$

$$0 < x < 4 \quad y < 1$$

$$0 < x < 4 \quad 1 \leq y \leq 5$$

$$0 < x < 4$$

$$y > 5$$

$$F(x,y) = \int_{-\infty}^x \int_{-\infty}^y f(x,y) dy dx$$

$$F(x,y) = \int_{-\infty}^x \int_y^\infty f(x,y) dy dx$$

$$F(x,y) = 0$$

$$F(x,y) = 0$$

$$F(x,y) = \int_0^x \int_0^y xy dy dx$$

$$= \int_0^x \int_0^y \frac{xy}{96} dy dx$$

$$\frac{1}{2 \times 96} \int_0^x u (y^2 - 1) du$$

$$\frac{1}{4 \times 96} x^2 (y^2 - 1)$$

$$\frac{1}{384} x^2 (y^2 - 1)$$

~~4~~

$$y > 5$$

$$\int_0^4 \int_{\frac{5}{\sqrt{x}}}^{\infty} \frac{xy}{96} dy dx$$

$$= \frac{x^2}{16}$$

$$x > 4, y < 1$$

$$F(x,y) = \iiint x^2 y / 96 > 0,$$

$$x > 4, 1 < y < 5$$

$$\int \int \frac{x^2 y}{96} dy dx$$

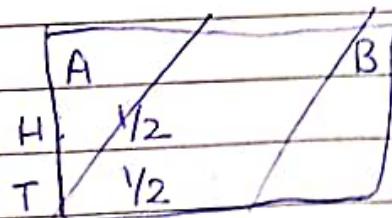
$$\int \frac{x^2}{2 \times 96} dx$$

$$\frac{x^3}{4 \times 96}, \frac{16 y^2}{4 \times 96}$$

$$y^2 / 24$$

$$x > 4, y > 5$$

$$F(x,y) = 1,$$



Mathematical Expectation

Let x be a discrete random variable and let $f(x)$ be its corresponding probability mass function.

$$f(x) = P(X=x)$$

$$\text{or } f(x) = P(X=x_k) \text{ where } k=1, 2, 3, \dots$$

and if $\sum x_i f(x_i) < \infty$ we say that the expected value/mean/mathematical expression of X exists and is defined as

$$E(X) = \sum_{i=1}^{\infty} x_i f(x_i)$$

Note that the series $\sum x_i f(x_i)$ may not converge but this series $\sum x_i^k f(x_i)$ converges. In this case we can say $E(X)$ does not exist.

e.g. Two players A and B, play a coin tossing game. A gives B one dollar if a head turns up otherwise B gives A one dollar. If probability of coin showing head is P then find expected gain of A.

$$\begin{array}{lll} x_i & p(x_i) & p_i x_i \\ H: +2 & P & +P \\ T: -1 & (1-P) & -(1-P) \\ \hline \sum p_i x_i = & 1-2P & \end{array}$$

let X denote the random variable of amount paid to A

$$X \begin{cases} 1 & P \\ -1 & (1-P) \end{cases}$$

$$1-2P > 0$$

when

$$P > 1/2$$

$$1-2P < 0$$

when $P < 1/2$

at $P = 1/2$

$$1 - \alpha p^2 = 0$$

[so unbiased coin is used]

Example 2 :

The probability density function for a random variable is given by

$$f(x) = \begin{cases} \frac{1}{2}x & 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$

Then the expected value of x is $\int_{-\infty}^{\infty} x f(x) dx$

$$\int_0^2 x \cdot \frac{1}{2}x dx = \frac{x^2}{4} \Big|_0^2$$

$$\frac{1}{2} \cdot \frac{2^3}{3} \Big|_0^2 = \frac{2^2}{3} \cdot \frac{2}{2} = \frac{4}{3}$$

$$\frac{8}{6} = \frac{4}{3}$$

Theorem : Let x and y be two random variables. Then :

$$\textcircled{1} \quad E(cx) = cE(x) \quad \text{for any constant } c$$

$$\textcircled{2} \quad E(x+y) = E(x) + E(y)$$

$$\textcircled{3} \quad E(xy) = E(x) \cdot E(y) \quad \text{provided } x \text{ and } y \text{ are independent}$$

Exercise

Suppose that a game is to be played with a single fair die. In this game a player wins ₹20 if 2 turns up, ₹40 if 4 turns up, losses ₹30 if 6 turns up, and neither wins/loses if any other face turns up.

Let X be the random variable denoted which player wins money

$$X = \begin{cases} 20 & \text{if } Y=2 \\ -30 & \text{if } Y=6 \\ 40 & \text{if } Y=4 \\ 0 & \text{if } Y=1, 3, 5 \end{cases}$$

$P(1)$

$$P(X) = \frac{1}{6} \Rightarrow P(2) = P(4) = P(6) = P(1) = \frac{1}{6}$$

$$\mathbb{E}(X) = \frac{1}{6} \times 20 + \frac{1}{6} \times 40 - \frac{1}{6} \times 30$$

$$\frac{10}{6} - \frac{5}{6} = \frac{5}{6}$$

Definition: Let μ denote the mean/expected expectation of a random variable X .

Then variance of X is defined and denoted by

$$V(X) = [E(X - \mu)^2]$$

$$\text{Var}(X) = E(X^2) - E(X)^2$$

and its positive root is called standard deviation & denoted by

$$\sigma_x = \sqrt{\text{Variance}} = \sqrt{E[(X-\mu)^2]}$$

- Q) Find the variance and standard deviation of the pdf:

$$f(x) = \begin{cases} 2/2 & 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned} E(X) &= \int_{-\infty}^{\infty} x f(x) dx = \int_0^2 x \cdot \frac{2}{2} dx \\ &\quad + \frac{1}{2} \int_0^2 x^2 dx \\ &\quad + \frac{x^3}{6} \Big|_0^2 = 8/3 = 4/3 \\ \mu &= 4/3 \end{aligned}$$

$$E(X^2) =$$

$$\text{Var}(X) = \sqrt{E((X-\mu)^2)}$$

$$= \int_0^2 (x - 4/3)^2 f(x) dx$$

$$= \int_0^2 (x - 4/3)^2 \cdot 2/2 dx$$

$$= \int_0^2 \frac{x^2 - 2x + 1}{2} dx$$

$$= \frac{x^3}{6} - \frac{2x^2}{4} \Big|_0^2 = \frac{8}{6} - \frac{16}{6} = -\frac{8}{6}$$

$$\int \left(x^2 + \frac{16}{9} - \frac{8x}{3} \right) \frac{x}{2} dx$$

$$\int_0^2 \frac{x^3}{2} + \frac{8x}{9} - 4x^2 dx$$

$$\left. \frac{x^4}{8} + \frac{4x^2}{9} - \frac{4x^3}{9} \right|_0^2$$

$$\frac{16}{8} + \frac{4 \cdot 4}{9} - \frac{4 \cdot 4 \cdot 2}{9}$$

$$2 + \frac{16}{9} - \frac{32}{9}$$

$$2 + \frac{16}{9} - \frac{16}{9}$$

$$\text{Var}(x) = 2/9$$

$$\sigma = \sqrt{2/9} = \sqrt{2}/3 //$$

Theorem: Let x be a random variable with μ as mathematical expectation mean. Then -

$$\textcircled{1} \quad \sigma_x^2 = [E(x-\mu)^2] = \frac{E(x^2) - (\mu)^2}{E(x^2) - [E(x)]^2}$$

$$\textcircled{2} \quad \text{Var}(cx) = c^2 \text{Var}(x) \quad \text{for any constant } c$$

$$\textcircled{3} \quad \min [E(x - \bar{x})^2] = [E(x - \mu)^2]$$

$$(4) \text{Var}(x+y) = \text{Var}(x) + \text{Var}(y)$$

$\sigma_{x+y}^2 \rightarrow \sigma_x^2 + \sigma_y^2$

$$(5) \text{Var}(x-y) = \text{Var}(x) + \text{Var}(y)$$

$\sigma_x^2 + \sigma_y^2$

provided x & y are independent

~~P(x →)~~

Independent random Variable

Suppose that X and Y are two discrete random variables. If the events $X=x$ and $Y=y$ are independent for all x and y we say that X and Y are independent random variables. In this case,

$$P(X=x, Y=y) = P(X=x) P(Y=y)$$

$$f(x,y) = f_1(x) \cdot f_2(y)$$

Conversely, if for all x & y , $f(x,y)$ can be written as $f(x,y) = f_1(x) \cdot f_2(y)$ for some f_1 and f_2 , X and Y are independent, otherwise dependent.

In case of X and $Y \rightarrow$ independent

$$P(X \leq x, Y \leq y) = P(X \leq x) \cdot P(Y \leq y)$$

$$F(x,y) = F(x) \cdot F(y)$$

Theorem: For 2 independent random variables

X and Y

$$E(XY) = E(X) \cdot E(Y)$$

Proof: Let $f(x,y) = P(X \leq x, Y \leq y)$ be the corresponding pdf and pmf. Since X and Y are independent

$$f(x,y) \equiv f(x) \cdot f(y) \quad \text{when}$$

$f(x) \rightarrow \text{pdf for } x$
 $f(y) \rightarrow \text{pdf for } y$

$$E(XY) = \sum \sum xy f(x,y)$$

$$= \sum \sum xy f(x) \cdot f(y)$$

$$= \sum x f(x) \sum y f(y)$$

Hence Proved

$$E(XY) = E(X) \cdot E(Y)$$

Theorem: If $\mu = E(X)$ then prove that
 $E(X-\mu) = 0$

(*)

$$E(X-\mu) = E(X) - E(\mu)$$

$$= \mu - \sum f(x_i) \cdot \mu$$

$$= \mu - \mu$$

$$= 0$$

Hence Proved

$$(E(X) + E(X) - \mu)$$

? ~~for~~

$$\text{(To prove: } E(X^2) - (E(X))^2 = [E(X-\mu)^2])$$

$$[E(X-\mu)^2] = E(X^2 + \mu^2 - 2X\mu)$$

$$= E(X^2) + E(\mu^2) - 2E(X\mu)$$

$$= E(X^2) + 1 - E(2\mu X)$$

$$= E(X^2) + 1 - E(X) - 1$$

$$= E(X^2) - E(X)$$

2) $\text{Var}(cx) = c^2 \text{Var}(x)$

Proof:

$$\text{Var}(cx) = \sqrt{E((x-\mu)^2)}$$

$$= c^2 \text{Var}(x)$$

3) $\min_{\alpha \in \mathbb{R}} E[(x-\alpha)^2] = E[(x-\mu)^2]$

Proof:

$$E[(x-\alpha)^2] = E(x^2 + \alpha^2 - 2\alpha x)$$

$$= E(x^2) + E(\alpha^2) - E(2\alpha x)$$

$$= E(x^2) + E(\alpha^2) - [E(x) - E(x) - E(\alpha)]$$

$$[E(x - \mu + \mu - \alpha)^2]$$

$$= E[(x - \mu) + (\mu - \alpha)^2]$$

$$= E(x - \mu)^2 + E(\mu - \alpha)^2 + E(2(\mu - \alpha)(x - \mu))$$

$$= E(x - \mu)^2 + 2(\mu - \alpha)E(x - \mu) + E(\mu - \alpha)^2$$

$$= E(x - \mu)^2 + (\mu - \alpha)^2$$

$$\min_{\alpha \in \mathbb{R}} E[(x - \mu)^2]$$

Hence Proved

$\text{Var}(x+y) = \text{Var}(x) + \text{Var}(y)$ provided x & y are independent

$$\underline{\mathbf{E}(x+y) \text{ Proof: } E[(x+y) - (\mu_x + \mu_y)]^2 \rightarrow \text{Var}(x+y)}$$

$$\text{Var}(x+y) = E[(x+y) - (\mu_x + \mu_y)]^2$$

$$= E[(x+y)^2] + E[(\mu_x + \mu_y)^2] - 2(\mu_x + \mu_y)E(x+y)$$

$$= E[(x+y)^2] + (\mu_x + \mu_y)^2 - 2(\mu_x + \mu_y)E(x+y)$$

$$E[(x-\mu_x) + (y-\mu_y)]^2$$

$$= E[(x-\mu_x)^2] + E[(y-\mu_y)^2]$$

$$= E[(x-\mu_x)^2 + (y-\mu_y)^2 + 2(x-\mu_x)(y-\mu_y)]$$

$$= E[(x-\mu_x)^2] + E[(y-\mu_y)^2] + 2[E(x-\mu_x)E(y-\mu_y)]$$

$$= E[(x-\mu_x)]^2 + E[(y-\mu_y)]^2$$

$$= \text{Var}(x) + \text{Var}(y)$$

Suppose $X \rightarrow$ random variable

$\mu \rightarrow E(x) \rightarrow$ mean, $\sigma \rightarrow$ standard deviation

[Standardized $\leftarrow X^* = \frac{x-\mu}{\sigma}$
random variable]

$$E(X^*) = 0$$

$$\sigma_{X^*} = \sqrt{E((X^* - \mu)^2)} = 1$$

$$g \rightarrow X: S \rightarrow \mathbb{R} \subset \mathbb{R}$$

$g = X \rightarrow$ is also random variable

Probability (contd.)

Function of random variable

Let X be a discrete random variable with pmf f and $g: R \rightarrow R$ be a given function. Then $y = g \circ x = g(x)$ is also a random variable and its pmf is given by

$$h(y) = P(Y=y) = \sum_{\{x: g(x)=y\}} P(X=x) = \sum_{\{x: g(x)=y\}} f(x) \geq 0$$

$$\sum_y h(y) = 1$$

Proof: $\sum_y h(y) = \sum_y P(Y=y) = \sum_y \sum_{\{x: g(x)=y\}} f(x) = 1$

$$E(Y) = \sum_{i=1}^{\infty} y_i h(y_i) = \sum_{i=1}^{\infty} g(x_i) h(y_i)$$

$$h(y_i) = \sum_{\{x: g(x)=y_i\}} f(x)$$

Repeat the same for continuous random variable

(Q) Compute $E(3x^2 - 2x)$ for $f(x) \begin{cases} 1/2 & 0 < x < 2 \\ 0 & \text{otherwise} \end{cases}$

$$E(3x^2 - 2x^3) = \int_{-\infty}^{x} (3x^2 - 2x^3) f(x) dx$$

$$= \int_{-\infty}^{x} (3x^2 - 2x^3) \frac{1}{2} x dx$$

$$\frac{1}{2} \int_{-\infty}^{x} 3x^3 - 2x^2 dx$$

$$\frac{1}{2} \left(\frac{3x^4}{4} - \frac{2x^3}{3} \right) \Big|_{-\infty}^{x}$$

$$E(3x^2 - 2x^3) = \frac{3x^4}{8} - \frac{x^3}{3}$$

$$E(3x^2 - 2x)$$

$$\begin{cases} 0 & x < 1 \\ 1/24 & 1 \leq x < 2 \\ 10/3 & 2 \leq x \end{cases}$$

Let x be a continuous random variable with pdf f and $g: \mathbb{R} \rightarrow \mathbb{R}$

$$E(Y) = E(g(x)) = \int_{-\infty}^{\infty} g(x) \cdot f(x) dx$$

In case of joint distribution

$$E(g(x,y)) = \iint_{-\infty}^{\infty} g(x,y) f(x,y) dx dy$$

Moments

Let X be a random variable and μ is the mean
then r^{th} central moment is defined as
 $\mu_r = E[(X-\mu)^{2r}]$ $r=0, 1, 2, \dots$

$$= \begin{cases} \sum (x-\mu)^r f(x) & \text{(discrete case)} \\ \int_{-\infty}^{\infty} (x-\mu)^r f(x) dx & \text{(continuous case)} \end{cases}$$

The r -th moment of X about origin also called r^{th} raw moment, is defined by -

$$\mu_r = E(X^r), \quad r=0, 1, 2, 3, \dots$$

$$\mu_0 = 1$$

$$\mu_1 = 0$$

μ_2 = Variance

Prove that :

$$\mu_r = \mu'_r - \binom{r}{1} \mu'_{r-1} \mu + \dots + (-1)^j \binom{r}{j} \mu'_{r-j} \mu^j$$

$$\dots + (-1)^r \mu'_0 \mu^r$$

Central Moment	Raw Moment	moment round:
$\mu_r = E((x-\mu)^r)$	$\mu'_r = E(x^r)$	$\hat{\mu}_r = E(x^r)$
$\mu_0 = \sigma^2$	$\mu'_0 = \sigma^2$	$\hat{\mu}_0 = 1$
$\mu_1 = \mu$	$\mu'_1 = \mu$	$\hat{\mu}_1 = \mu - E\mu$
$\mu_2 = E((x-\mu)^2) = \mu'_2 - 2C_1\mu'_1 + C_2\mu'_2$		

Q2 Find the first four moment

① about the origin

② about mean

for a random variable X with density

$$f(x) = \begin{cases} \frac{4}{81} x(9-x)^2 & 0 \leq x \leq 3 \\ 0 & \text{otherwise} \end{cases}$$

① $\mu'_0 \approx \mu'_0 = 1.3$

$$\mu'_1 = \int (x-\mu) \frac{4}{81} x(9-x)^2 dx$$

$$\int_0^3 \frac{4}{81} x^2 (9-x)^2 dx$$

$$= \int_0^3 \frac{4}{81} x^2 (81x^3 - 54x^2 + 9x) dx$$

$$= \frac{4}{81} \left[\frac{3x^3 - x^5}{5} \right]_0^3 = \frac{4}{81} \left(\frac{81 - 243}{5} \right) = -\frac{162}{81} = -2$$

$$\mu'_2 = \frac{4}{81} \times \frac{(3-1)^2}{5} = \frac{4}{81} \times \frac{4}{5} = \frac{16}{405}$$

$$V_{03} = \frac{4}{3} \int_0^2 \left(\frac{4}{5} - \frac{4}{5}x^2 \right) dx$$

$$= \frac{4}{3} \left[\frac{4}{5}x - \frac{4}{15}x^3 \right]_0^2$$

$$= \frac{4}{3} \left[\left(\frac{8}{5} - \frac{16}{15} \right) \right]$$

$$= \frac{4}{3} \left(\frac{24}{15} - \frac{16}{15} \right)$$

$$= \frac{4}{3} \left(\frac{8}{15} \right)$$

$$= \frac{32}{45}$$

$$\therefore R \left(\frac{32}{45} \right) = \frac{12}{5} = 2\frac{2}{5}$$

$$123 \quad V'_{03} = \frac{4}{3} \int_0^2 (9-x^2) dx$$

$$= \frac{4}{3} \int_0^2 (9x^2 - x^4) dx$$

$$= \frac{4}{3} \left(\frac{9x^3}{3} - \frac{x^5}{5} \right)_0^2$$

$$= \frac{4}{3} \left(\frac{72}{3} - \frac{32}{5} \right)$$

$$= \frac{4}{3} \left(\frac{240}{15} - \frac{96}{15} \right) = \frac{216}{35}$$

$$\mu_0 = 27/2$$

$$\mu_0 = 1$$

$$\mu_1 > 0$$

$$\mu_2 = 11/25 \Rightarrow \mu_2' = 27/10$$

$$\mu_2' = \mu_2^1 - 2\mu_1^1 \cdot \mu_1 \quad \{ \text{Formula} \}$$

Moment Generating Function

The moment generating function of a random variable x is defined by

$$M_x(t) = E(e^{xt}) = \int_{-\infty}^{\infty} e^{tx} f(x) dx$$

$$\text{Theorem: } M_x(t) = 1 + \mu t + \frac{\mu' t^2}{2!} + \dots + \frac{\mu' t^r}{r!}$$

$$\mu'_r = \left. \frac{d^r}{dt^r} M_x(t) \right|_{t=0}$$

$$M_x(t) = E(e^{xt}) = E\left(1 + tx + \frac{t^2 x^2}{2!} + \frac{t^3 x^3}{3!} + \dots\right)$$

$$= 1 + tE(x) + \frac{t^2}{2!} E(x^2) + \frac{t^3}{3!} E(x^3) + \dots$$

$$= 1 + \mu'_1 t + \frac{t^2}{2!} \mu'_2$$

(q) Prove that if X and Y are independent random variables.

$$M_{x+y}(t) = M_x(t) \cdot M_y(t)$$

$$\begin{aligned} M_{x+y}(t) &\rightarrow E(e^{tx+y}) \\ &= E(e^{tx}) \cdot E(e^{ty}) \end{aligned}$$

(q) The random variable X can assume the value 1 and -1 with probability $1/2$ each.

① Find MGf

② Find first four moments about origin

$$X \{ -1, 1 \}$$

~~P(X = -1)~~ $P(X = -1) = 1/2$

$P(X = 1) = 1/2$

$$M_X(t) = e^{tx_1} f(x_1) + e^{tx_2} f(x_2)$$

$$= \frac{1}{2} e^{-t} + e^{t}$$

$$= \frac{1}{2} (e^{-t} + e^t)$$

$$\mu'_0 = E(X-0) = \int_{-\infty}^{\infty} x \cdot \frac{1}{2} (e^{-t} + e^t) dx$$

$$= 1$$

~~$\frac{1}{2}$~~ $= \frac{1}{2} (e^{-t} + e^t)$
 $= \frac{1}{2} \left(1 - t + \frac{t^2}{2!} - \frac{t^3}{3!} + \dots + 1 + t + \frac{t^2}{2!} + \frac{t^3}{3!} + \dots \right)$

$$M_X(t) = \frac{1}{2} \left(2 + \cancel{2} \cdot \frac{t^2}{2!} + \cancel{2} \cdot \frac{t^4}{4!} + \cancel{2} \cdot \frac{t^6}{6!} + \dots \right)$$

$$H_1' = 0 = H_3' + H_5' \dots$$

$$H_2' = 1 = H_4' + H_6' \dots$$

B

Same question with

$$f(x) \begin{cases} 2e^{-2x} & x > 0 \\ 0 & x \leq 0 \end{cases}$$

Find MGF & first four moments about origin

~~$$M(t) = \int_{-\infty}^{\infty} e^{tx} \cdot 2e^{-2x} dt$$~~

Theorem: If a and b are constants, the MGF of $\frac{x+a}{b}$ is

$$M_{\frac{x+a}{b}}(t) = e^{at/b} M_a(t/b)$$

⑥ Suppose X and Y are random variable having moment generating function $M_X(t)$ and $M_Y(t)$. Then X and Y have the same probability distribution $M_X(t)$, $M_Y(t)$

$$M_x(t) = \int_0^{\infty} 2e^{-2x} \cdot e^{tx} dt dx \quad (t < 2)$$

$$\left[2e^{-2x} \middle/ -\frac{e^{tx}}{2} \right]_0^{\infty}$$

$$2 \int_0^{\infty} e^{x(t-2)} dx$$

$$2 \left. \frac{e^{x(t-2)}}{(t-2)} \right|_0^{\infty} \quad \begin{matrix} 1 \\ (t-2) \end{matrix} \quad 2(0 - i) \quad \cancel{2} \cancel{(t-2)}$$

$$2 \left(\cancel{0} \cancel{(t-2)} \right) \quad \cancel{2} \cancel{(t-2)} = \frac{2}{t-2}$$

$$= \frac{1}{1 - \frac{t}{2}}$$

$$= \left(1 - \frac{t}{2} \right)^{-1}$$

$$M_x(t) = 1 + t/2 + t^2/4! + t^3/8! \dots$$

$$\left\{ \begin{array}{l} \mu_0 = 1 \\ \mu_1 = 1/2 \\ \mu_2 = 1/2 \\ \mu_3 = 3/2 \end{array} \right\}$$

Recall :

$$P(X, Y) = f(x, y)$$

$$P(X_1 = x_k) = \sum_{i=1}^n f(x_k, y_i) \geq f(x_k, y_i)$$

$$P(Y = y_k) = \sum_{i=1}^n f(x_i, y_k) \leq f(x_i, y_k)$$

$$P(a < x < b, c < y < d) = \int_c^d \int_a^b f(x, y) dx dy$$

$$A \subseteq \mathbb{R}^2 \quad P(A) = \iint_A f(x, y) dx dy$$

$$F(x, y) = \int_{-\infty}^x \int_{-\infty}^y f(x, y) dy dx$$

$$\frac{\partial^2 F}{\partial x \partial y} = f(x, y)$$

Marginal distribution

$$\text{Marginal density function } f_1(x) = \int_{-\infty}^{\infty} f(x, v) dv$$

$$f_2(y) = \int_{-\infty}^{\infty} f(u, y) du$$

Variance and expectation of joint distribution

Let X and Y be two random variables with pmf/pdf $f(x,y)$ mean w.r.t x (or y) is given by

$$\mu_x = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} x f(x,y) dx dy$$

$$\mu_y = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} y f(x,y) dx dy$$

Variance

$$\sigma_x^2 = \text{Var}(x) = E((x - \mu_x)^2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x - \mu_x)^2 f(x,y) dx dy$$

$$\sigma_y^2 = \text{Var}(y) = E((y - \mu_y)^2) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (y - \mu_y)^2 f(x,y) dx dy$$

Covariance

$$\sigma_{xy} = \text{covar}(X, Y)$$

$$= E((x - \mu_x)(y - \mu_y)) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} (x - \mu_x)(y - \mu_y) f(x,y) dx dy$$

For discrete case:

$$\mu_x = \sum \sum x f(x,y) \quad \mu_y = \sum \sum y f(x,y)$$

$$\sigma_{xy} = \sum \sum (x - \mu_x)(y - \mu_y) f(x,y)$$

Theorem : $\sigma_{xy} = E(xy) - E(x)E(y)$

$$\begin{aligned} & E((x-\mu_x)(y-\mu_y)) \\ & E(x-\mu_x) \cdot E(y-\mu_y) \\ & [E(x) - E(\mu_x)] [E(y) - E(\mu_y)] \end{aligned}$$

$$\begin{aligned} \sigma_{xy}^2 &= \sum_x \sum_y (x-\mu_x)(y-\mu_y) f(x,y) \\ &= \sum_x \sum_y (xy - x\mu_y - y\mu_x + \mu_x\mu_y) f(x,y) \\ &\quad \cancel{\times \cancel{\times}} \end{aligned}$$

$$\begin{aligned} & E(xy - x\mu_y - y\mu_x + \mu_x\mu_y) \\ &= E(xy) - \mu_x E(y) - \mu_y E(x) + \mu_x \mu_y \\ &= E(xy) - E(x) \cdot E(y) - E(y) \cdot E(x) \\ &\quad + E(x) E(y) \\ &= E(xy) - E(x) \cdot E(y) \end{aligned}$$

Theorem : $\text{Var}(x \pm y) = \text{Var}(x) + \text{Var}(y) \pm 2 \cancel{\times} \text{Cov}(x, y)$

$$\begin{aligned} \text{Var}(x \pm y) &= E((x \pm y)^2) - (\mu_x \pm \mu_y)^2 \\ &= E((x - \mu_x) \pm (y - \mu_y))^2 \\ &= E((x - \mu_x)^2) + E(y - \mu_y)^2 \pm 2(x - \mu_x)(y - \mu_y) \\ &= E(x - \mu_x)^2 + E(y - \mu_y)^2 \pm 2 E((x - \mu_x)(y - \mu_y)) \end{aligned}$$

$$\text{Var}(x) = \text{Var}(x) + \text{Var}(y) \pm 2\text{Cov}(x, y)$$

Q) What happens to σ_{xy} when $x=y$?

$$\sigma_{xy} = \begin{cases} 0 & \text{if } x \neq y \text{ are independent} \\ \sigma_x \cdot \sigma_y & x = y \end{cases}$$

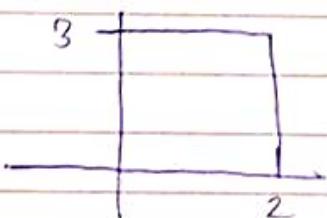
Correlation coefficients

$$\rho = \frac{\sigma_{xy}}{\sigma_x \cdot \sigma_y} = \begin{cases} 0 & \text{if } x \neq y \text{ are independent} \\ 1 & \text{if } x = y \end{cases}$$

Example: The joint p.m.f of x and y

$$f(x, y) = \begin{cases} \frac{1}{42} (2x + y) & 0 \leq x \leq 2 \\ & 0 \leq y \leq 3 \\ 0 & x, y \in \mathbb{Z} \end{cases}$$

Find: $E(x), E(y), E(xy)$
 $E(x^2), E(y^2), \sigma_x \cdot \sigma_y$
 $\text{Var}(x), \text{Var}(y), \rho$



$$\begin{aligned} \rightarrow E(x) &= \sum_{y=0}^3 \sum_{x=0}^2 x \cdot f(x, y) \\ &= \sum_{y=0}^3 \sum_{x=0}^2 \frac{2x + y}{42} \\ &= \frac{1}{42} \sum_{y=0}^3 2 \cdot (0^2 + 1^2 + 2^2) + (0 + y + 2y) \\ &= \frac{1}{42} \sum_{y=0}^3 10 + 3y \Rightarrow \frac{1}{42} [10(4) + 3(0+1+2+3)] \\ &= \frac{40 + 18}{42} = \frac{58}{42} \end{aligned}$$

$$E(Y) = \sum \sum y \cdot f(y)$$

$$= \frac{1}{42} \sum_{y=0}^3 2y + y^2$$

$$\Rightarrow \frac{1}{42} \sum_{y=0}^3 2y(1+2) + 3y^2$$

$$\Rightarrow \frac{1}{42} \sum_{y=0}^3 6y + 3y^2$$

$$\Rightarrow \frac{1}{42} [6(0+1+2+3) + 3(0+1^2+2^2+3^2)]$$

$$\Rightarrow \frac{1}{42} [36 + 42]$$

$$= 78/42 // \Rightarrow 13/7 //$$

$$E(XY) = \sum \sum xy \cdot f(x,y)$$

$$= \frac{1}{42} \sum \sum 2x^2y + xy^2$$

$$\Rightarrow \frac{1}{42} \sum_{y=0}^3 2x^2y(0+1+2+3) + xy^2(0+1+2+3)$$

$$\Rightarrow \frac{1}{42} \sum_{y=0}^3 10y + 30y^2$$

$$\Rightarrow \frac{1}{42} [10(0+1+2+3) + 30(0+1+2+3)]$$

$$\Rightarrow \frac{1}{42} \left(\frac{60}{60} + \frac{3 \times 14}{84} \right) // \Rightarrow 102/42 //$$

$$E(X^2) = 17/7$$

$$\text{Var}(X) = 17/7 - 58/42$$

~~$$E(Y^2) \rightarrow E(Y^2) = 33/7$$~~

$$C_{XY} = E(XY) - E(X) \cdot E(Y)$$

Q: If X and Y are continuous random variable with pdf:

$$f(x, y) = \begin{cases} \frac{1}{210} (2x+y), & 2 < x < 5 \\ 0, & 0 < y < 5 \\ 0, & \text{otherwise} \end{cases}$$

$$E(X), E(Y), E(XY)$$

$$E(X^2), E(Y^2), C_{XY}$$

$$\text{Var}(X), \text{Var}(Y), f$$

Conditional Distribution

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

$$f_{XY|X}(y|x) = \frac{P(Y=y | X=x)}{\frac{P(Y=y, X=x)}{P(X=x)}}$$

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

$$f_1(x) = \int_{-\infty}^{\infty} f(x, y) dy$$

$$P(B|A) = \frac{P(A \cap B)}{P(A)}$$

If X and Y are two random variables and $\{A: x\}$, and $\{B: Y=y\}$ be two events is -

$$P(Y=y | X=x) = \frac{P(X=x, Y=y)}{P(X=x)} = \frac{f(x,y)}{f_1(x)}$$

let $f(x,y)$ denote pmf we can write
the above inequality as

$f_1(y|x) = \frac{f(x,y)}{f_1(x)}$, where $f_1(x)$ is marginal density/mass function of X .
we call $f_1(y|x)$ as conditional mass function of Y for a given X .

Similarly we can define

$$f_2(x|y) = \frac{f(x,y)}{f_2(y)}$$

$f_2(y) \rightarrow$ marginal probability mass function

$$P((c < y < d) | (x_1 < x < x_2)) = \int_c^d f_1(y|x) dy$$

Conditional expectation :-

$$f(x,y) \rightarrow \begin{cases} 1/42 (2x+y) & 0 \leq x \leq 2 \\ 0 & 0 \leq y \leq 3 \\ & \text{otherwise} \end{cases} \quad x, y \in \mathbb{R}$$

$$f_1(x) = \sum_{y=0}^3 f(x_1, y)$$

$$= \frac{1}{42} [(2x+0) + (2x+1) + (2x+2) + (2x+3)]$$

$$= \frac{8x+6}{42}$$

Find $f(y|x)$, $f(x|y)$.

$$f(y|x) = \frac{f(x,y)}{f_1(x)}$$

$$= \frac{\sum_{x=0}^2 \sum_{y=0}^3 f(x,y)}{(2x+6)} \quad \cancel{42(2x+6)}$$

$$= \frac{(8x+6)}{42}$$

$$= \frac{2x+y}{8x+6}$$

$$f_2(y) = \sum_{x=0}^2 f(x_k, y)$$

$$= \frac{1}{42} [(2(0)+y) + (2(1)+y) + (2(2)+y)]$$

$$f(x|y) = \frac{2x+y}{3y+6} \quad \cancel{4x+6} \quad \frac{1}{42} [y+2y+4y+y] \quad \frac{3y+6}{42}$$

Conditional Expectation

$$E(Y|X=x) = \int_{-\infty}^{\infty} y f(y|x) dy = \int_{-\infty}^{\infty} y \frac{f(x,y)}{f(x)} dy$$

Calculate expectation of Y given X=2

$$E(Y|X=2) = \sum_{y=-\infty}^{\infty} y f(y|2)$$

$$= \sum_{y=0}^{3} y \left(\frac{(2)}{8+6} \right)$$

$$= \sum_{y=0}^{3} y \left(\frac{4+y}{16+6} \right)$$

$$\frac{1}{22} \sum_{y=0}^{3} 4y + y^2$$

$$\frac{1}{22} [4(1+2+3) + (0+1+4+9)]$$

$$\frac{1}{22} (24 + 14)$$

$$\frac{38}{22} = \frac{19}{11}$$

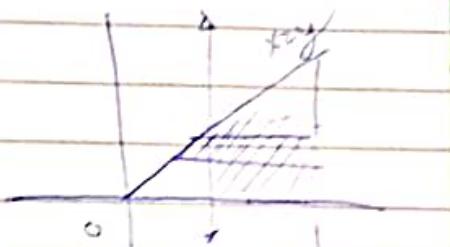
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Example: The joint density function of random variables X and Y is given by

$$f(x,y) = \begin{cases} 8xy & 0 \leq x \leq 1 \\ 0 & 0 < y \leq x \\ 0 & \text{otherwise} \end{cases}$$

- (i) Find marginal density function of x and y
 Find conditional density function of x & y
 Find conditional expectation of y and y to sum y .

$$f_1(x) = \int_0^x 8xy \, dy \\ = 8x \cdot \frac{x^2}{2} = 4x^3 //$$



$$f_2(y) = \int_0^y 8xy \, dx$$

$$= 8y \cdot \frac{y^2}{2} = 4y^3 // \otimes$$

$$f_2(y) = \int_0^1 8xy \, dx \\ = 8y \left(\cancel{4x^2} \right) \cdot \frac{(1-y^2)}{2}$$

$$= 4y - 4y^3 //$$

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

$$\begin{cases} \frac{8xy}{4x^3} = \frac{2y}{x^2} & 0 < y \leq x \\ 0 & \text{otherwise} \end{cases}$$

$$f(x|y) = \frac{f(x,y)}{f_2(y)} = \begin{cases} \frac{2x^2}{1-y^2} & 0 < y < 1 \\ 0 & \text{otherwise} \end{cases}$$

$$\begin{aligned}
 E(y|x) &= \int_{-\infty}^{\infty} y f(y|x) dy \\
 &= \int_0^x y \frac{2y}{x^2} dy \\
 &= \left. \frac{2}{x^2} \frac{y^3}{3} \right|_0^x \\
 &= \frac{2}{3} \frac{x^2}{x} \\
 &= \frac{2}{3} x
 \end{aligned}$$

$$\begin{aligned}
 E(x|y) &= \int_{-\infty}^{\infty} x f(x|y) dx \\
 &= \int_0^1 x \cdot \frac{2x}{1-y^2} dx \\
 &= \left. \frac{1}{1-y^2} \frac{2x^3}{3} \right|_0^1 \\
 &= \frac{2}{3(1-y^2)} (1-y^3)
 \end{aligned}$$

eg

The number of items produced in a factory during a week is a minimum variable with mean 50 and standard deviation 5. What can be said about the probability that the weeks production will be between 45 and 60?

Chernoff's Inequality

Suppose X is a random variable with mean μ ,
and variance σ^2 . Then if $\epsilon > 0$ be any number
then

$$P(|X - \mu| \geq \epsilon) \leq \frac{\sigma^2}{\epsilon^2} \quad \text{or if we take } \epsilon = k\sigma$$

$$P(|X - \mu| \geq k\sigma) \leq 1/k^2$$

$$\mu = 50$$

$$\sigma = 5$$

$$P(40 \leq X \leq 60) = P(-10 \leq X - 50 \leq 10)$$

$$= P(|X - 50| \leq 10)$$

$$= P(|X - \mu| \leq 10)$$

$$= 1 - P(|X - \mu| \geq 10)$$

$$\epsilon = 10$$

$$\therefore 1 - P(|X - \mu| \geq 10) \leq \sigma^2/\epsilon^2$$

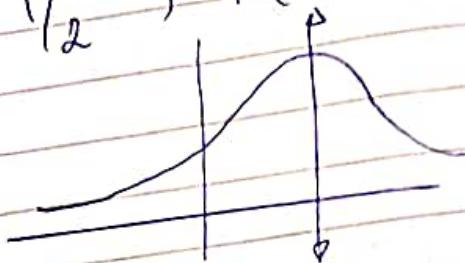
$$\leq 25/100 \\ \leq 0.25$$

$$\therefore 1 - P(|X - \mu| \geq 10)$$

$$= 1 - 0.25$$

$$P(40 \leq X \leq 60) = 0.75$$

α Median $\rightarrow x$ such that
 $P(X \leq x) = 1/2, P(X \geq x) = 1/2$



$$\alpha = \beta/2$$

Median

The median is the values of x such that
 $P(X \leq x) = 1/2$

$$P(X \geq x) = 1/2$$

In particular median divides the density curve into two parts having area = $1/2$.

e.g.: Find median of

$$p.d.f = \frac{1}{(x+1)(x+2)}$$

$$x = 0, 1, 2$$

$$\int_{-\infty}^x \frac{1}{(x+1)(x+2)} dx = \frac{1}{2}$$

$$\int_0^x \left(\frac{1}{x+1} - \frac{1}{x+2} \right) dx = \frac{1}{2}$$

$$\ln \left(\frac{x+1}{x+2} \right) = \frac{1}{2}$$

$$\ln \left(\frac{x+1}{x+2} \right) = \ln \left(\frac{1}{2} \right)$$

$$P(x) = \sum_{x=0}^{\infty} \frac{1}{(x+1)(x+2)}$$

$x \geq 0$

$$P(x) \rightarrow P(0) = 1/2$$

Skewness

Often distribution function is symmetric about any number. Instead it may have tails longer than other side. If longer tail occurs,

The density function is called right skew,

If ~~tail~~ longer tail occurs on left
the density function is called left skew.

$$\text{sk } \alpha_3 = \frac{E[(x-\mu)^3]}{\sigma^3} = \begin{cases} > 0 & \text{right skew} \\ < 0 & \text{left skew} \\ = 0 & \text{symmetric} \end{cases}$$

$$\% A = \frac{A_1}{\int_{-\infty}^{\infty} f(x) dx} \times 100 = 100 A_1 \rightarrow \text{percentile}$$

Let $\alpha \in [0, 1]$ then $\alpha \in \mathbb{R}$
for which

$$\alpha = \int_{-\infty}^{\infty} f(x) dx$$

Then the area left to $x = x_\alpha$ is $100 \times \alpha \%$. is called 100^th percentile

Median = 50th percentile.

Example: We toss a coin 6-times. What is the probability of getting exactly ~~top~~² heads.

$${}^6C_2 \cdot \left(\frac{1}{2}\right)^2 \cdot \left(\frac{1}{2}\right)^4$$

$\frac{6 \times 5}{2} \times \frac{1}{2^6} = \frac{15}{64}$

$$\left[(p+q)^n = \sum_{r=0}^n {}^n C_r p^r q^{n-r} \right]$$

$$E(ae^{tx}) = 1 + \mu t + \frac{\mu' t^2}{2!} + \frac{\mu'_3 t^3}{3!} + \dots$$

a) $\mu_2 \rightarrow$ central
 $\mu'_2 \sigma \rightarrow$ raw moment

Replace t by iw

Characteristic eqn

Replace ' t ' in Moment generating function by iw .
 the function we obtain is Characteristic function

$$\Phi_x(t)(w) = M_x e^{(iwx)}$$

$$\sum_{x=-\infty}^{\infty} e^{inx} f(x)$$

$$\int_{-\infty}^{\infty} e^{inx} f(x) dx$$

Continuous

$$\Phi_X = \begin{cases} \sum_{\omega} |f(x)| dx = \sum f(x) = 1 \\ \int f(x) dx = 1 \end{cases}$$

$$\Phi_X = 1 + i\omega f(w) - H_2 \frac{\omega^2}{2!} - \dots - i^r H_r \frac{\omega^r}{r!}$$

$$H_n' = (-1)^n i^n \left. \frac{d^n}{dw^n} \Phi_X(w) \right|_{w=0}$$

Exer: Find characteristic functions

$$f(x) = \begin{cases} 1/2a & |x| < a \\ 0 & \text{otherwise} \end{cases}$$

$$\Phi_X(w) = \int_{-\infty}^{\infty} e^{iwx} f(x) dx$$

$$= \frac{1}{2a} \int_{-a}^a e^{iwx} dx$$

$$= \frac{1}{2a} \frac{e^{iwa} - e^{-iwa}}{iw}$$

$$= \frac{\sin(wa)}{aw}$$

$$\mathbb{E}(X - \mu) = \sum (\lambda - \mu) f(x) = \sum \lambda f(x) - \mu f(x) = \mu - \mu = 0 //$$

Mean deviation: $\sum |x - \mu| f(x) \neq 0$

Q) Find the probability that in five tosses of a fair coin a 4 will appear : (i) exactly twice (ii) at least 2. $\sum f(x) = (p+q)^n = 1$

$$P(X=2) \hat{=} {}^5C_2 \left(\frac{1}{6}\right)^2 \left(\frac{5}{6}\right)^3$$

$$\begin{aligned} P(P(X \geq 2)) &= 1 - P(X < 2) \\ &= 1 - P(X=0) - P(X=1) \\ &= 1 - {}^5C_0 \left(\frac{5}{6}\right)^5 - {}^5C_1 \left(\frac{1}{6}\right) \left(\frac{5}{6}\right)^4 \end{aligned}$$

Note

Prop

Binomial Distribution

Let ϕ be the probability that an event will happen then $q = 1-p$ is the probability that the event won't happen.

Suppose, we are having n -trials. The probability that it will happen exactly x -times in n trials is given by,

$$f(x) = P(X=x) = {}^nC_x p^x q^{n-x}$$

$$f(x) = \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

The discrete probability function $f(x)$ in ① is called Binomial distribution since for $x=1, 2, 3 \dots n$ it corresponds to successive terms in the Binomial expansion.

$$\sum f(x) = (p+q)^n = q^n + {}^n C_1 q^{n-1} p + {}^n C_2 q^{n-2} p^2 \dots + {}^n C_n q^{n-p} p^p$$

Note for $n=1 \rightarrow$ distribution is called ^{Bernoulli's} _{Binomial distribution}

Properties :

1) Mean : $\mu = \sum_{x=0}^n x f(x) = np$

when $x=1$

$$\underline{\mu = 1}$$

$$= \sum \frac{n n!}{(n-x)! x!} p^x q^{n-x}$$

$$= \sum \frac{n!}{(x-1)! (n-x)!} p^x q^{n-x}$$

$$= \sum \frac{n (n-1) \dots (x)!}{(x-1)! (n-1-(x-1))!} p^x q^{n-x}$$

$$= np \sum \frac{(n-1)!}{(x-1)! (n-1-(x-1))!} p^{x-1} q^{(n-1)-(x-1)}$$

$$= np \sum {}^{n-1} C_{x-1} p^{x-1} q^{(n-1)-(x-1)} \quad \boxed{[(p+q)^{n-1}] = 1}$$

$$\underline{= np}$$

2) Variance

$$\sigma^2 = E(X-\mu)^2 = \sum (x-np)^2 \frac{n!}{n!(n-x)!} p^x q^{n-x}$$

$$= \sum \frac{x^2 n!}{x!(n-x)!} p^x q^{n-x}$$

$$\sum (x^2 + n^2 p^2 - 2np) \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

$$\sum x^2 \frac{n!}{x!(n-x)!} p^x q^{n-x} + \cancel{\sum n^2 p^2} - \cancel{2np^2}$$

$$\sum_{x=0}^n \frac{x^2 n!}{(x-1)!(n-x-1)!} p^x q^{n-x} / \cancel{2} - \cancel{n^2 p^2}$$

$$\sum x^2 \frac{n!}{(x-1)!(n-x)!} p^x q^{n-x} - n^2 p^2$$

$$\sum \frac{((x-1)+1)n!}{(x-1)!(n-x)!} p^x q^{n-x} - n^2 p^2$$

$$\sum \frac{(x-1)n! p^x q^{n-x}}{(x-1)!(n-x)!} + \frac{n! p^x q^{n-x}}{(x-1)!(n-x)!} - n^2 p^2$$

$$\sum \frac{n! p^x q^{n-x}}{(x-2)!(n-x)!} + np - n^2 p^2$$

$$n \cancel{p^2} \rightarrow p^2 n(n-1) \sum \frac{(n-2)! p^x q^{n-x}}{(x-2)!(n-2)!(x-2)!} - np$$

$$p^2(n^2 - n) + np - n^2 p^2$$

~~$$p^2 n^2 - np^2 + np - n^2 p^2$$~~

$$= np - np^2$$

$$\sigma^2 = np(1-p)$$

Standard Deviation

$$\therefore \sigma = \sqrt{npq}$$

4) * coefficient of skewness $\frac{\mu_3}{\sigma^3}$

$$\frac{\mu_3}{\sigma^3} = \frac{1}{\sigma^3} \sum_{x=0}^n (x-np)^3 \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

$$= \frac{q-p}{\sqrt{npq}}$$

5) Coefficient of kurtosis

$$\alpha_4 = \frac{E((X-\mu)^4)}{\sigma^4} \quad \text{Definition}$$

measures peakness of a density / mean distribution function

$$\alpha_4 = \frac{1}{\sigma^4} \sum (n-np)^4 \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

$$= 3 + \frac{1 - \frac{6pq}{npq}}{npq}$$

Moment Generating Function

$$M_x(t) = E(e^{tx}) = \sum_{x=0}^n e^{tx} {}^n C_x p^x q^{n-x}$$

$$= e^p \sum_{x=0}^n {}^n C_x (pe^t)^x q^{n-x}$$

$$= (pe^t + q)^n$$

Characteristic function

$$\phi(w) = E(e^{iwx}) = \sum_{x=0}^n e^{iwx} {}^n C_x p^x q^{n-x}$$

$$= (q + pe^{iw})^n$$

e.g., If 20% of the bolts produced by a machine are defective. Determine the probability that out of 4 chosen at random less than 2 bolts are defective.

$$p = 0.8$$

$$q = 0.2$$

$$P(X \leq 2) = {}^4 C_2 (0.2)^2 (0.8)^2$$

$$\frac{2 \cdot 4 \times 3}{2} / 0.04 (0.064)$$

$$(0.24)(0.64) = 0.1536$$

$$P = b \quad 4C_0 p^0 q^4 + 4C_1 p^1 q^3 \\ = 0.8192$$

eg : If the probability that an individual will suffer a bad reaction from an injection of a given serum is 0.001. Determine the probability that out of 2000 individuals

- (a) exactly 3
 - (b) more than 2
- individual will suffer a bad reaction.

$$P(X=3) = {}^{2000}C_3 (0.001)^3 (0.999)^{1997} \approx 0.1805$$

$$P(X > 2) = 1 - P(X \leq 2)$$

$$= 1 - {}^{2000}C_0 (0.001)^0 (0.999)^{2000} \\ - {}^{2000}C_1 (0.001)^1 (0.999)^{1999} \\ - {}^{2000}C_2 (0.001)^2 (0.999)^{1998}$$

Poisson's Distribution

Let X be a discrete random variable such that values $0, 1, 2, 3, \dots$ such that the p.m.f

$$f(x) = P(X=x) = \frac{\lambda^x e^{-\lambda}}{x!} \quad x=0, 1, 2, 3, \dots$$

where λ is a given constant. Then the distribution is called Poisson's Distribution.

Property

① Mean: $\mu = E(X) = \sum_{x=0}^{\infty} x \lambda^x e^{-\lambda}$

$$= \sum_{x=0}^{\infty} \frac{\lambda^x e^{-\lambda}}{(x-1)!}$$

$$= e^{-\lambda} \sum_{x=0}^{\infty} \frac{\lambda^x}{(x-1)!}$$

$$= e^{-\lambda} \sum_{y=-1}^{\infty} \frac{\lambda^{y+1} \lambda^{y+1}}{y!}$$

$$= e^{-\lambda} \lambda \sum_{y=0}^{\infty} \frac{\lambda^y}{y!} \rightarrow e^{\lambda}$$

$$= \lambda$$

In the previous problem

Mean : $np = (0.001)(2000) = 2$

$$\therefore \lambda = 2$$

$$P(X=3) = \frac{\lambda^3 e^{-\lambda}}{3!} \rightarrow \frac{8 \times e^{-2}}{3!} = 0.18045$$

② Variance $\sigma^2 = E((x-\lambda)^2) \Rightarrow \sum (x-\lambda)^2 \frac{\lambda^x e^{-\lambda}}{x!}$

$$\sum_{x=0}^{\infty} (x^2 + \lambda^2 - 2x\lambda) \frac{\lambda^x e^{-\lambda}}{x!}$$

$$\sum_{x=0}^{\infty} x^2 \frac{\lambda^x e^{-\lambda}}{x!} + \lambda^2 \sum_{x=0}^{\infty} \frac{\lambda^{x-1} e^{-\lambda}}{x!} - 2\lambda \sum_{x=0}^{\infty} x \frac{\lambda^x e^{-\lambda}}{x!}$$

$$\sum \frac{x \lambda^x e^{-\lambda}}{(x-1)!} + \lambda^2 - 2\lambda(\lambda)$$

$$x-1 = y$$

$$\sum \frac{(y+1)\lambda^{y+1} e^{-\lambda}}{y!} - \cancel{\lambda^3} = \lambda^2$$

$$\sum \frac{y\lambda^{y+1} e^{-\lambda}}{y!} + \sum \frac{\lambda^{y+1} e^{-\lambda}}{y!} - \lambda^2$$

$$\lambda \sum \frac{y\lambda^y e^{-\lambda}}{y!} + y\lambda \sum \frac{\lambda^y e^{-\lambda}}{y!} - \lambda^2$$

$$\cancel{\lambda^2} + \cancel{y\lambda} - \lambda^2 = \lambda^2$$

③ Standard deviation : $\sigma = \sqrt{\lambda}$

④ skewness : $\alpha_3 = \frac{E((x-\lambda)^3)}{\sigma^3} + \frac{1}{\sqrt{\lambda}}$

⑤ Kurtosis : $\alpha_4 = \frac{E((x-\lambda)^4)}{\sigma^4} = 3 + \frac{1}{\lambda}$

Q) MGF

$$\begin{aligned}
 M(t) &= E(e^{tx}) = \sum \frac{e^{tx} \lambda^x e^{-\lambda}}{x!} \\
 &= e^{-\lambda} \sum \frac{e^{tx} \lambda^x}{x!} = e^{-\lambda} e^{t\lambda} \\
 &= e^{t\lambda - \lambda} = e^{\lambda(t-1)}
 \end{aligned}$$

7) Characteristics: $\phi(\omega) = E(e^{j\omega x}) = e^{\lambda(e^{j\omega}-1)}$

Relation b/w poission and Binomial Distribution

Binomial Distribution	poisson
--------------------------	--------------------

$$P(X=x) = \frac{n!}{x!(n-x)!} p^x q^{n-x}$$

Let $\lambda = np \rightarrow$ mean

$$P(X=x) = \frac{n!}{x!(n-x)!} \left(\frac{\lambda}{n}\right)^x \left(1-\frac{\lambda}{n}\right)^{n-x}$$

$$\frac{1^x}{x!} \frac{n!}{(n-x)!} \frac{1}{n^x} \left(1-\frac{\lambda}{n}\right)^{n-x} \left(1-\frac{1}{n}\right)^x$$

$$\left[\frac{\lambda^x}{x!} \left(1-\frac{\lambda}{n}\right)^n \right] \text{to } e^{(\lambda-\lambda/n)n}$$

$$\lim_{n \rightarrow \infty} \left[\frac{n(n-1) \dots (n-x+1)}{n^x} \cdot \left(1-\frac{1}{n}\right)^x \right]$$

$$\left[\frac{\lambda^x}{x!} \left(1 - \frac{\lambda}{n}\right)^n \right] \left[1 \cdot \left(1 - \frac{1}{n}\right) \cdots \left(1 - \frac{x+1}{n}\right) \left(\frac{n-\lambda}{n}\right)^x \right]$$

$\lim_{n \rightarrow \infty}$

$$P(X=x) = \frac{\lambda^x e^{-\lambda}}{x!} \quad (1)$$

$$\therefore P(X=2) = \frac{\lambda^2 e^{-\lambda}}{2!}$$

Example : let us toss a coin until head appears
what is the probability that first head will appear
in the 5th attempt.

$$\text{Soln} : P(X=5) = \frac{1}{2} \left(\frac{1}{2}\right) \left(\frac{1}{2}\right) \left(\frac{1}{2}\right) \left(\frac{1}{2}\right)$$

1 1
32 /

Example : let us toss a die until a 4 appears what is the
probability that a 4 will appear in 5 attempts

$$P_f(x) = \left(\frac{5}{6}\right)^5 \left(\frac{1}{6}\right)^1$$

if prob. probability that 4 will appear = p

$$P(x) = (1-p)^5 p^1$$

1 1
5

Sum of these 2 numbers is no. of attempts

Geometric Distribution

Suppose that a pmf for a discrete random variable X has the following

$$P(X=k) = (1-p)^{k-1} \cdot p \quad k=1, 2, 3, \dots$$

Then X is called to have Geometric distribution with parameter p (possibility of success in each trial)

Properties

Mean $\mu = E(X) = \sum_{k=0}^{\infty} k (1-p)^{k-1} \cdot p$
 $= p [0 + (1-p)^0 + 2(1-p) + 3(1-p)^2 + \dots]$

$$S = 1 - (1-p)^0 + 2(1-p) + 3(1-p)^2 + 4(1-p)^3 + \dots$$
$$S(1-p) = (1-p)^1 + 2(1-p)^2 + 3(1-p)^3 + \dots$$

$$S - S(1-p) = 1 + (1-p) + (1-p)^2 + (1-p)^3 + \dots$$

$$Sp = \frac{1}{1 - (1-p)}$$

$$Sp = \frac{1}{p}$$

$$S = \frac{1}{p^2}$$

$$\therefore \mu = p \cdot S = \frac{1}{p}$$

2) Variance

$$E(X - 1/p)^2 = \sum_{k=1}^{\infty} \left(k - \frac{1}{p}\right)^2 (1-p)^{k-1} \cdot p$$

$$= \frac{1-p}{p^2}$$

3) S.D.:

$$\sigma = \sqrt{\frac{1-p}{p}}$$

4) Skewness:

$$\alpha_3 = \frac{2-p}{\sqrt{1-p}}$$

5) Kurtosis:

$$\alpha_4 = 6 + \frac{p^2}{1-p}$$

6) ~~MGF~~ MGF

$$M(t) = \sum_{k=0}^{\infty} e^{kt} (1-p)^{k-1} p$$

$$= pe^{pt} \left(1 + (1-p)e^t + (1-p)^2 e^{2t} \dots \right)$$

$$= \frac{pe^t}{1 - (1-p)e^t}$$

~~\approx~~ ~~$2e^t$~~

7) Characteristic functions

$$\phi(\omega) = \frac{pe^{i\omega}}{1 - (1-p)e^{i\omega}}$$

Normal Distribution

The most important continuous probability distribution & the density function is given by

$$f(x) = \frac{1}{s\sqrt{2\pi}} e^{-\frac{(x-m)^2}{2s^2}}, \quad -\infty < x < \infty$$

where $m = \mu$ is the mean
 $s = \sigma$ is standard deviation

(Q) Expectation: $E(f(x)) = \frac{1}{s\sqrt{2\pi}} \int_{-\infty}^{\infty} x e^{-\frac{(x-m)^2}{2s^2}} dx$

$$\begin{aligned} & y = \frac{x-m}{s} \\ & \int_{-\infty}^{\infty} (sy + m) e^{-y^2/2} dy \\ & \frac{1}{s\sqrt{2\pi}} \left[\int_{-\infty}^{\infty} sy e^{-y^2/2} dy + m \int_{-\infty}^{\infty} e^{-y^2/2} dy \right] = 1 \end{aligned}$$

$\int e^{-y^2/2} dy = \sqrt{\pi}/2$

$$\begin{aligned} & y^2/2 = P \\ & y dy = e^{-P} \end{aligned}$$

$$= s \int e^{-P} dP + m$$

$$\begin{aligned} & \frac{se^{-P}}{P} \\ & = \frac{2se^{-y^2/2}}{y^2} \Big|_{-\infty}^{\infty} + m \end{aligned}$$

$$\boxed{\mu = E(x) = m}$$

Variance :

$$\sigma^2 = E(x-\mu)^2 = \int_{-\infty}^{\infty} \frac{1}{s\sqrt{2\pi}} (x-\mu)^2 e^{-\frac{(x-\mu)^2}{2s^2}} dx$$

$$\frac{1}{s\sqrt{2\pi}} \int (x^2 + \mu^2 - 2\mu x) e^{-\frac{(x-\mu)^2}{2s^2}} dx$$

$$\left[\frac{y^2 - \mu^2}{s^2} \right] \int \frac{x^2}{s\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2s^2}} dx + \int \frac{\mu^2}{s\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2s^2}} dx = -2\mu \int \frac{xe^{-\frac{(x-\mu)^2}{2s^2}}}{s\sqrt{2\pi}} dx$$

$$\frac{1}{s\sqrt{2\pi}} \int (sy + \mu) e^{-\frac{(y-\mu)^2}{2s^2}} dy + \mu^2 = 2\mu^2$$

$$\frac{1}{s\sqrt{2\pi}} \int (sy + \mu^2) e^{-\frac{(y-\mu)^2}{2s^2}} dy = \mu^2$$

$$\frac{1}{s\sqrt{2\pi}} \int (s^2y^2 + \mu^2 + 2\mu sy) e^{-\frac{(y-\mu)^2}{2s^2}} dy = \mu^2$$

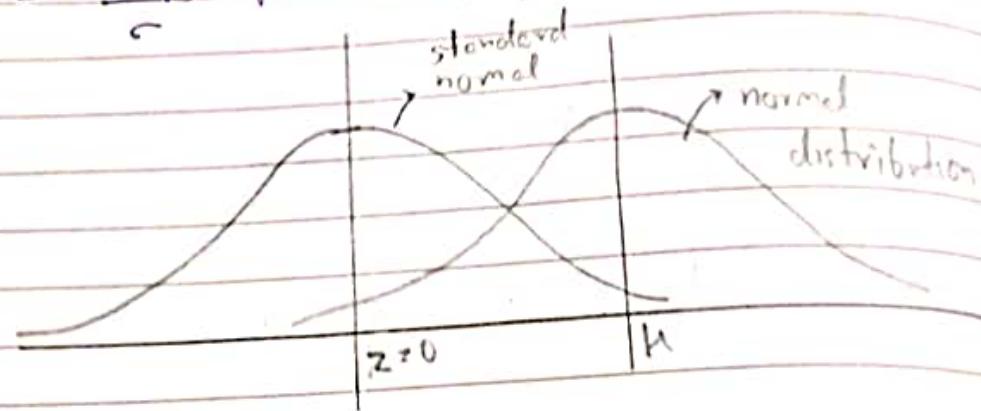
$$s^2\mu^2 + 2\mu^2 - \mu^2$$

$$s^2 \Rightarrow -2\mu^2 + 2\mu^2$$

$$s^2$$

Let

$$Z = \frac{X-\mu}{\sigma} \text{ for } Z, \text{ p } E(Z) = 0, \sigma_Z = 1,$$



$$f(z) = \frac{1}{\sigma\sqrt{2\pi}} e^{-z^2/2}$$

$$\textcircled{1} \quad P(-z \leq X \leq z) = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{(t-\mu)^2}{2\sigma^2}} dt$$

$$P(a \leq Z \leq b) = \frac{1}{\sigma\sqrt{2\pi}} \int_a^b e^{-\frac{z^2}{2}} dz$$

$$P(-1 \leq Z \leq 1) = 0.6827$$

$$P(-2 \leq Z \leq 2) = 0.9541$$

$$P(-3 \leq Z \leq 3) = 0.9973$$

MGF :

$$M_X(t) = \int_{-\infty}^{\infty} e^{xt} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$y = \frac{x-\mu}{\sigma}$$

$$\int \frac{1}{\sigma\sqrt{2\pi}} e^{(ys+\mu)t} e^{-y^2/2} dy$$

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{ys + \mu - y^2/2} dy$$

$$\frac{1}{\sqrt{2\pi}} e^{\mu} \int_{-\infty}^{\infty} e^{ys - y^2/2} dy$$

$$\int e^{ys - y^2/2} = \left[y \cdot e^{ys - y^2/2} \right]_{-\infty}^{\infty} - \int y(ys - y) e^{ys - y^2/2} dy$$

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{(ys + \mu)t} \cdot e^{-y^2/2} dy$$

$$\frac{e^{\mu t}}{\sqrt{2\pi}} \int e^{ys t - y^2/2} dy$$

$$\frac{e^{\mu t}}{\sqrt{2\pi}} \int e^{y^2 t - 2ys t + s^2 t^2 - s^2 t^2} dy$$

$$\frac{e^{\mu t}}{\sqrt{2\pi}} \int e^{-\frac{1}{2}(y^2 - 2ys t + s^2 t^2)} dy$$

$$\frac{e^{\mu t}}{\sqrt{2\pi}} \int e^{-\frac{1}{2}(y - st)^2 + \frac{s^2 t^2}{2}} dy$$

$$\frac{e^{\mu t + s^2 t^2/2}}{\sqrt{2\pi}} \int e^{-\frac{1}{2}(y - st)^2} dy$$

$$M_x(t) = e^{\mu t + s^2 t^2/2}$$

Coeff of skewness : $\alpha_3 = 0$

Coeff of kurtosis : $\alpha_4 = 3$

Characteristic function : $\varphi(t) = e^{i\mu t - \frac{1}{2}\sigma^2 t^2}$

Relation b/w Binomial & Normal Characteristics

when $n \rightarrow$ large and $p \rightarrow$ small

Binomial distribution

$$\mu = np$$

$$\sigma = \sqrt{np(1-p)} \quad z = \frac{x-np}{\sqrt{np(1-p)}} \quad \left[\frac{x-\mu}{\sigma} \right]$$

Suppose we take np & nq greater than 5

$$\lim_{n \rightarrow \infty} z = \frac{x-np}{\sqrt{np(1-p)}}$$

$$\therefore \lim_{n \rightarrow \infty} \varphi_p \left(a \leq \frac{x-np}{\sqrt{np(1-p)}} \leq b \right)$$

$$= \frac{1}{\sigma \sqrt{2\pi}} \int_a^b e^{-z^2/2} dz$$

Assignment 8

$$(Q12) \quad f(x|y=y) = \frac{|y| e^{-\lambda^2 y^2}}{\sqrt{\pi}}$$

$$f(y) = \frac{\lambda e^{-\lambda^2 y^2}}{\sqrt{\pi}}$$

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

$$f(x,y) = f(x|y=y) \cdot f(y)$$

$$= \frac{|y| e^{-x^2 y^2}}{\sqrt{\pi}} \cdot \frac{\lambda e^{-\lambda^2 y^2}}{\sqrt{\pi}}$$

$$\begin{aligned} f_x(x) &= \int_{-\infty}^{\infty} f(x,y) dy \\ &= \frac{\lambda}{\sqrt{\pi}} \int_{-\infty}^{\infty} |y| e^{-x^2 y^2} \cdot e^{-\lambda^2 y^2} dy \\ &= \frac{2\lambda}{\pi} \int_0^{\infty} y e^{-x^2 y^2} \cdot e^{-\lambda^2 y^2} dy \end{aligned}$$

$$\text{let } (x^2 + \lambda^2) y^2 = t$$

$$2(x^2 + \lambda^2) y dy = dt$$

$$\begin{aligned} (\lambda^2 + x^2)^{-\frac{1}{2}} \int_0^{\infty} e^{-t} dt &= \frac{1}{\pi(x^2 + \lambda^2)} \left[e^{-t} \right]_0^{\infty} \\ &= \frac{1}{\pi(x^2 + \lambda^2)} \end{aligned}$$

Hence Proved

Stochastic Process

Let T be a index/ordered set of \mathbb{R}
a stochastic process is a collection of random variables $\{X(t) : t \in T\}$ or $\{X_t : t \in T\}$

If T is countable \rightarrow we say that $\{X(t) : t \in T\}$ is a discrete time stochastic process.

If T is uncountable \rightarrow we say that $\{X(t) : t \in T\}$ is a continuous time stochastic process.

Note: The space S = Possible outcome is called state space.

1) Discrete time & discrete space

A motor insurance company review claims yearly. Three levels of discounts are possible: $\{0\%, 20\%, 40\%\}$ depending on the number of accidents by the insuree.

$$\text{Here } T = \{0, 1, 2, 3, \dots\}$$

$$S = \{0, 20, 40\}$$

2) Discrete Time and continuous time space

Let X_n denote the temperature of the sun on n^{th} day of calendar year at noon.

$$n = 1, 2, 3, \dots, 365$$

$$S = (10, 60)$$

3) Continuous time & discrete space

Modify example 2 by removing the noise

4) Continuous time & discrete space

Number of persons entering in a room at a certain time.

Def (Mean)

For stochastic process,
 $\{x(t) : t \in T\}$ the mean is calculated
 $\mu(t) = E(x(t)) \quad \forall t \in T$

Def (Auto-correlation)

Auto-correlation $\{x(t) : t \in T\}$ denoted by
 $R(t_1, t_2)$ for any $t_1, t_2 \in T$ as
expectation of product of $x(t_1) \& x(t_2)$
is $R(t_1, t_2) = E(x(t_1), x(t_2))$

Def (Auto-covariance)

Auto covariance of $\{x(t) : t \in T\}$ denoted by $C(t_1, t_2)$

$$C(t_1, t_2) = E[(x(t_1) - \mu(t_1))(x(t_2) - \mu(t_2))]$$
$$= E[x(t_1)x(t_2)] - \mu(t_1)\mu(t_2) = R(t_1, t_2) - \mu(t_1)\mu(t_2)$$

$$\Rightarrow R(t_1, t_2) = C(t_1, t_2) + \mu(t_1)\mu(t_2)$$

Correlation

$$f(t_1, t_2) = \frac{C(t_1, t_2)}{\sqrt{\text{Var}(X_{t_1})} \cdot \sqrt{\text{Var}(X_{t_2})}}$$

where $\text{Var}(X_{t_i}) = E[(X_{t_i}) - \mu(t_i)]^2$

Def (counting process)

Counting process is a stochastic process with state space $S = \{0, 1, 2, \dots\}$ such that $X(t)$ is non-decreasing in t .

Example:

- (i) $X(t) \rightarrow$ number of students present at time t
 - (ii) $X(t) \rightarrow$ number of students entering class at time t
- ↓
not non-decreasing

Note : A counting process $X(t)$ must satisfy

- (i) $X(t) \geq 0$
- (ii) $X(t)$ is integeral value
- (iii) if $s \leq t$, $X(s) \leq X(t)$

Increment in stochastic process in (t_1, t_2) is computed as $X(t_2) - X(t_1)$

Increment

- independent increments
- b. strictly stationary increment
- weakly

Independent Increments

if for m : $t_1 \leq t_2 \leq t_3 \leq t_4$.

$X(t_2) - X(t_1)$ and $X(t_4) - X(t_3)$ are independent to each other.

$X(t) \rightarrow$ number of babies born at time t ,
 $T \begin{cases} \nearrow \text{small} \\ \searrow \text{very large} \end{cases}$

Strictly stationary process if joint distribution of $X_{t_1}, X_{t_2}, \dots, X_t$ and X_{t+h}, Y_{t+h} , $+ X_{h+t}, \dots$ are identical for any t if $t+h \in T$ and for any number n .

Weakly stationary process

A stochastic process of $\{X(t) : T\}$ with $E(X_t^2) < \infty \quad \forall t \in T$ is called w.s.p.

If
 ① $E(X_t)$ is constant $\forall t \in T$
 ② $\text{Cov}(X_t, X_{t+k})$ depends only on k .

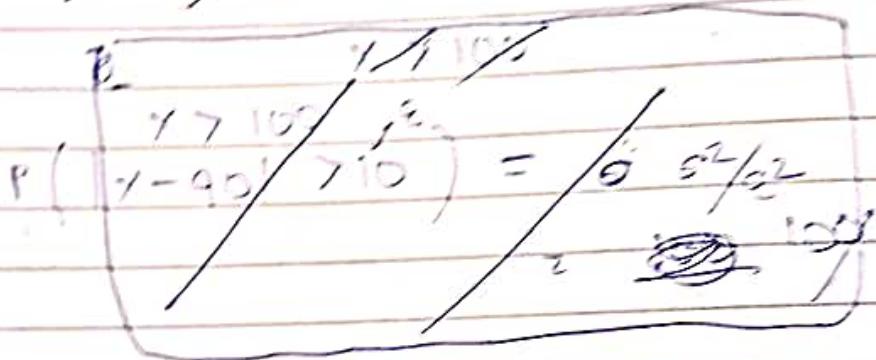
example: Let X_n $n \geq 1$ be independent random variable with $E(X_n) = 0 \quad \forall n \in \mathbb{N}$
 $\sigma_{X_n}^2 = 1 \quad \forall n \in \mathbb{N}$

Then $\text{Cov}(X_n, X_{n+m}) = E(X_n X_{n+m}) - E(X_n) E(X_{n+m})$
 $= E(X_n) \cdot E(X_{n+m}) - E(X_n) \cdot E(X_{n+m})$
 $= 0 \quad \text{if } m \neq 0$

It is a W.S.S.P. $\left\{ \begin{array}{ll} 1 & \text{if } m = 0 \\ ? & \text{if } m \neq 0 \end{array} \right.$

The speed of car at a point is recorded drawn with a mean speed 90 km/h and standard deviation 10 km/h . Probability that car is moving faster than 100 km/h .

$$P(X > 100) = ?$$



$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-(x-\mu)^2/2\sigma^2}$$

$$P(X > 100) = 1 - P(X \leq 100)$$

$$\int_{-\infty}^{100} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$

$$\int_{-\infty}^{100} \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{100-\mu/\sigma} e^{-t^2/2} dt$$

$$\frac{1}{2\sqrt{\pi}} \int_0^\infty e^{-m^2/2} dm$$

$$= \frac{1}{2\sqrt{\pi}}$$

$$dp = \frac{m}{\sqrt{2\pi}} dm$$

$$\frac{1}{2\sqrt{\pi}}$$

Note

$$\frac{1}{c\sqrt{2\pi}} \int e^{-(x-\mu)^2/2\sigma^2} dx$$

$$P(x > 100)$$

$$P\left(\frac{x-90}{10} > 1\right)$$

$$P(z > 1)$$

convert in
terms of z

$$\therefore \frac{1}{\sqrt{2\pi}} \int_1^\infty e^{-z^2/2} dz \quad \textcircled{1}$$

Similarly,

$$P(60 < x < 70)$$

$$P\left(\frac{-30}{10} < \frac{x-90}{10} < \frac{-20}{10}\right)$$

$$P(-3 < z < -2)$$

$$\frac{1}{\sqrt{2\pi}} \int_{-3}^{-2} e^{-z^2/2} dz$$

Using $\textcircled{1}$

$$\frac{1}{\sqrt{2\pi}} \int_{-\infty}^\infty e^{-z^2/2} dz + \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{-\infty} e^{-z^2/2} dz$$

$$1 - \frac{1}{\sqrt{2\pi}}$$

$$\frac{1}{\sqrt{2\pi}} \int_0^\infty e^{-z^2/2} dz - \frac{1}{\sqrt{2\pi}} \int_0^{-\infty} e^{-z^2/2} dz = 0.5 - \frac{1}{\sqrt{2\pi}}$$

$$\frac{0.6827}{2} = P(0.5 < z < 1)$$

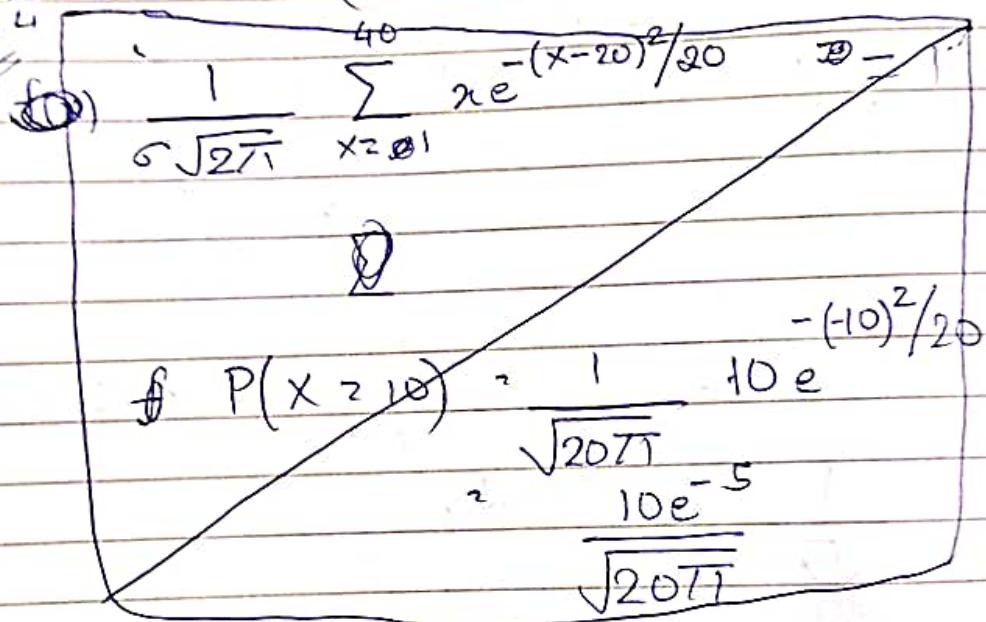
(Q) Find number of heads. Let X be number of times that a fair coin flipped 40 times lands heads. Find $P(X \geq 10)$ using normal distribution & compare with exact soln.

$$\text{Exact soln: } P = 40 C_{10} \left(\frac{1}{2}\right)^{10} \cdot \left(\frac{1}{2}\right)^{30}$$

$$\mu = np = 20$$

$$\sigma = \sqrt{npq} \\ = \sqrt{40 \cdot \frac{1}{2}} \\ = \sqrt{10}$$

$$\left\{ P = 40 C_{10} \frac{1}{2^{40}} \right\}$$



$$P(9.5 \leq X \leq 10.5) = \int_{9.5}^{10.5} \frac{1}{\sqrt{2\pi}} e^{-\frac{(x-20)^2}{20}} dx$$

$$P\left(\frac{-10.5}{\sqrt{10}} \leq Z \leq \frac{-9.5}{\sqrt{10}}\right) = \int_{-9.5/\sqrt{10}}^{-10.5/\sqrt{10}} \frac{1}{\sqrt{2\pi}} e^{-z^2/2} dz$$

$$= 2.334 \times 10^{-3}$$