

## Unit 4. Markov chain and queuing

i) Define probability vector

ii) Define stochastic matrix

iii) Define regular stochastic matrix

iv) unique fixed probability vector

i) Probability vector :- A vector  $v = [v_1, v_2, \dots, v_n]$  is said to be probability vector if

$$v_i \geq 0$$

$$\sum v_i = 1$$

Ex:- i)  $v = \left[ \frac{1}{2}, \frac{1}{2} \right]$

ii)  $v = [0, 1, 0]$

iii)  $v = \left[ \frac{1}{3}, \frac{1}{3}, \frac{1}{3} \right]$

iv)  $v = \left[ \frac{1}{4}, \frac{1}{4}, \frac{1}{2} \right]$

Notes:-

If  $v$  is not probability vector but each one of  $v_i$  are non-negative then  $\lambda v$  is probability vector where the  $\lambda = \frac{1}{\sum v_i}$

Ex:-  $v = [1, 2, 3]$

$$\lambda = \frac{1}{\sum v_i} = \frac{1}{1+2+3} = \frac{1}{6}$$

$$\lambda v = \frac{1}{6} [1, 2, 3]$$

ii) stock

$$P =$$

vector

Ex:-

iii) Regu

P is a  
enteries

Ex:-

$$P^2 =$$

$$\lambda v = \left[ \frac{1}{6}, \frac{2}{6}, \frac{3}{6} \right]$$

$$\lambda v = 1$$

$\lambda v$  is a probability vector.

ii] Stochastic Matrix :- A square matrix

$P = [P_{ij}]$  with each row being a probability vector is called stochastic matrix

Ex:- ]  $P = \begin{bmatrix} 1 & 0 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$

]  $A = \begin{bmatrix} 1 & 0 & 0 \\ \frac{1}{2} & \frac{1}{4} & \frac{1}{4} \\ \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \end{bmatrix}$

iii] Regular stochastic matrix :- A stochastic matrix  $P$  is said to be regular stochastic if all the entries of  $P^n$  are positive

Ex:- ]  $A = \begin{bmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$

$$A^2 = A \cdot A = \begin{bmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$A^2 = \begin{bmatrix} 0 + \frac{1}{2} & 0 + \frac{1}{2} \\ 0 + \frac{1}{4} & \frac{1}{2} + \frac{1}{4} \end{bmatrix}$$

$$A^2 = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{3}{4} \end{bmatrix}$$

Q] A stochastic matrix which is not a regular stochastic.

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

$$A^2 = A \cdot A = \begin{bmatrix} 1 & 0 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ \frac{3}{4} & \frac{1}{4} \end{bmatrix}$$

$$A^3 = A^2 \cdot A = \begin{bmatrix} 1 & 0 \\ \frac{3}{4} & \frac{1}{4} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$= \begin{bmatrix} 1+0 & 0+0 \\ \frac{3}{4} + \frac{1}{8} & \frac{3}{8} + \frac{1}{8} \end{bmatrix}$$

$$= \begin{bmatrix} 1 & 0 \\ \frac{5}{8} & \frac{1}{2} \end{bmatrix}$$

iv] Unique fixed probability vector:

A regular stochastic matrix  $p$  is said to be unique for fixed probability vector if  $p$  satisfies

i)  $x+y+z=1$

ii)  $xp = v$

# Stochastic

## Problem

Show that the matrix  $A = \begin{bmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix}$  is a regular stochastic matrix and hence find the unique fixed vector.

Given stochastic matrix

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

$$A^2 = A \cdot A = \begin{bmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} = \begin{bmatrix} 0 + \frac{1}{2} & 0 + \frac{1}{2} \\ 0 + \frac{1}{4} & \frac{1}{2} + \frac{1}{4} \end{bmatrix}$$

$$A^2 = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{3}{4} \end{bmatrix}$$

All the entries of  $A^2$  are positive.

$\therefore A$  is the regular stochastic matrix of order  $n=2$ .

unique fixed probability vector:

Let  $v = (x, y)$  be the unique fixed probability vector

such that i]  $x+y=1$  - ①

ii]  $VA = v$  - ②

$$\begin{bmatrix} x, y \end{bmatrix} \begin{bmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{bmatrix} = \begin{bmatrix} x, y \end{bmatrix}$$

$$\begin{bmatrix} 0x + \frac{1}{2}y, x + \frac{1}{2}y \end{bmatrix} = \begin{bmatrix} x, y \end{bmatrix}$$

$$0x + \frac{1}{2}y = x \quad | \quad x + \frac{1}{2}y = y$$

$$-x + \frac{1}{2}y = 0 \quad \text{---} \quad |$$

$$x + (-1 - 1)y = 0 \quad \text{---}$$

$$x = \frac{1}{2}y \quad \text{---} \textcircled{2}$$

put  $y = \frac{2}{3}$  in eqn  $\textcircled{2}$

$$x = \frac{1}{2} \left( \frac{2}{3} \right)$$

$$x = \frac{1}{3}$$

$$\boxed{x = \frac{1}{3}}$$

$$\frac{1}{2}y + y = 1$$

$$\frac{3}{2}y = 1$$

$$3y = 2$$

$$\boxed{y = \frac{2}{3}}$$

$$V = [x, y] = \left[ \frac{1}{3}, \frac{2}{3} \right] = 1$$

If  $A = \begin{bmatrix} a_1 & a_2 \\ b_1 & b_2 \end{bmatrix}$  is a stochastic matrix and  $v = (v_1, v_2)$  is a probability vector. Show that  $VA$  is also probability vector.

Soln By data  $A = \begin{bmatrix} a_1 & a_2 \\ b_1 & b_2 \end{bmatrix}$  be the stochastic matrix and  $v = (v_1, v_2)$  be the probability vector

$$VA = [v_1 \ v_2] \begin{bmatrix} a_1 & a_2 \\ b_1 & b_2 \end{bmatrix}$$

$$VA = [v_1 \ v_2] \begin{bmatrix} a_1 & a_2 \\ b_1 & b_2 \end{bmatrix}$$

We have prove that  $(v_1 a_1 + v_2 b_1) + (v_1 a_2 + v_2 b_2) = 1$

$$\text{LHS} = v_1 a_1 + v_2 b_1 + v_1 a_2 + v_2 b_2$$

$$= v_1(a_1 + a_2) + v_2(b_1 + b_2) \quad (\because \text{By the defn of S.M})$$

$$= v_1(1) + v_2(1) \quad (\text{By the defn of pr})$$

$$= 1 = \text{RHS}$$

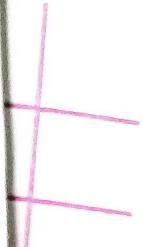
Q3 If  $A$  is a square matrix of order 'n' whose rows are each the same vector  $a = \{a_1, a_2, a_3, \dots, a_n\}$  and  $v = \{v_1, v_2, \dots, v_n\}$  is a probability vector prove that  $VA = a$

Soln By data

$$A = \begin{bmatrix} a_1 & a_2 & \dots & a_n \\ a_1 & a_2 & \dots & a_n \\ \vdots & & & \\ a_1 & a_2 & \dots & a_n \end{bmatrix}$$

g lecture? You could look, text underneath

is & Crosses



and  $v = [v_1, v_2, \dots, v_n]$  be the probability vector

$$vA = [v_1, v_2, \dots, v_n] \begin{bmatrix} a_1 & a_2 & \dots & a_n \\ a_1 & a_2 & \dots & a_n \\ \vdots & & & \\ a_1 & a_2 & \dots & a_n \end{bmatrix}$$

$$= [v_1 a_1 + v_2 a_1 + \dots + v_n a_1, v_1 a_2 + v_2 a_2 + \dots + v_n a_2, \dots, v_1 a_n + v_2 a_n + \dots + v_n a_n]$$

$$= [a_1(v_1 + v_2 + \dots + v_n), a_2(v_1 + v_2 + \dots + v_n), \dots, a_n(v_1 + v_2 + \dots + v_n)]$$

$$= [a_1(1), a_2(1), \dots, a_n(1)] = a$$

Q] Which of the following matrices are stochastic matrices? Soln

i)  $\begin{bmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix}$  If we add sum should be 1  
Eg: it should be square matrix.  $3 \times 3 = 1$

$\rightarrow$  It is not a square matrix. So it is not a stochastic matrix.

ii)  $\begin{bmatrix} \frac{15}{16} & \frac{1}{16} \\ \frac{1}{3} & \frac{2}{3} \end{bmatrix}$

$\rightarrow$  This is <sup>not</sup> a stochastic matrix because  $\frac{2}{3} + \frac{2}{3} \neq 1$

$$\text{ox} + \frac{1}{6} y$$

$$\boxed{\frac{1}{6} y}$$

i)  $\begin{bmatrix} 1 & 0 \\ \gamma_2 & \gamma_2 \end{bmatrix}$   $\rightarrow$  It is a stochastic matrix, because it is square,  $1, 0, \gamma_2, \gamma_2 \geq 0$  &  $1+0=1, \gamma_2+\gamma_2=1$

ii)  $\begin{bmatrix} \gamma_2 & -\gamma_2 \\ \gamma_4 & 3/4 \end{bmatrix}$

$\rightarrow$  It is not a stochastic matrix

$-\gamma_2$  is a negative value

$$\frac{1}{2} + (-\frac{1}{2}) \neq 1$$

Find the unique fixed probability vector of the

matrix  $A = \begin{bmatrix} 0 & 1 & 0 \\ \gamma_6 & \gamma_2 & \gamma_3 \\ 0 & \gamma_3 & \gamma_3 \end{bmatrix}$

above soln let  $v = [x, y, z]$  be the unique fixed probability vector.

$$[x, y, z] \begin{bmatrix} 0 & 1 & 0 \\ \gamma_6 & \gamma_2 & \gamma_3 \\ 0 & \gamma_3 & \gamma_3 \end{bmatrix} = [x, y, z]$$

$$[0x + \gamma_6 y + 0z, 1x + \gamma_2 y + \gamma_3 z, 0x + \gamma_3 y + \gamma_3 z] = [x, y, z]$$

$$0x + \frac{1}{6}y + 0z = x$$

$$\frac{1}{6}y = x \quad \text{---(1)}$$

$$y = 6x$$

$$1x + \frac{1}{2}y + \frac{2}{3}z = y$$

$$\frac{6x + 3y + 4z}{6} = y$$

$$6x + 3y + 4z = 6y$$

$$6x - 3y + 4z = 0$$

$$0x + \frac{1}{3}y + \frac{1}{3}z = z$$

$$\frac{y + z}{3} = z$$

$$y + z = 3z$$

$$y - 2z = 0$$

Substituting eqn ② and eqn ③ in eqn ①

$$x+y+z=1$$

$$\frac{y}{6} + y + \frac{y}{2} = 1$$

$$\frac{y+6y+3y}{6} = 1$$

$$10y = 6$$

$$\boxed{y = \frac{6}{10}} = \frac{3}{5} - \textcircled{3}$$

$$\textcircled{2} \Rightarrow x = y \cdot \frac{1}{6} = \frac{1}{6} \left( \frac{3}{5} \right) = \frac{1}{10}$$

$$z = \frac{y}{2} = \frac{1}{2} \left( \frac{3}{5} \right) = \frac{3}{10}$$

$$v = (x, y, z) = \left( \frac{3}{10}, \frac{1}{10}, \frac{3}{10} \right)$$

$$\text{i)} \quad \frac{3}{5} + \frac{1}{10} + \frac{3}{10} \geq 0$$

$$\text{ii)} \quad \frac{3}{5} + \frac{1}{10} + \frac{3}{10} = 1$$

$$\text{iii)} \quad P = \begin{bmatrix} 0 & 1/2 & v_2 \\ y_3 & 2/3 & 0 \\ 0 & 1 & 0 \end{bmatrix}$$

Soln Let  $v = [x, y, z]$  be the unique fixed probability vector

$$x+y+z=1 - \textcircled{1}$$

$$\text{and } vp = v$$

$$[x, y, z] \begin{bmatrix} 0 & v_2 & v_3 \\ v_3 & 0 & v_1 \\ 0 & v_1 & 0 \end{bmatrix} = [x, y, z]$$

$$[v_2x + v_3y + v_1z, \quad v_3x + v_1y + v_2z, \quad v_1x + v_2y + v_3z] \cdot [x, y, z]$$

$$v_2x + v_3y + v_1z = x$$

$$\boxed{\frac{1}{3}y = x} - ②.$$

$$v_3x + \frac{2}{3}y + v_1z = y$$

$$\frac{3x + 4y + 6z}{6} = y$$

$$v_1x + v_2y + v_3z = z$$

$$\boxed{\frac{1}{2}x = z}$$

$$\boxed{3x + 4y + 6z = 6y}$$

③

④

Substituting eqn ② & eqn ④ in eqn ⑥

$$x + y + z = 1$$

$$\frac{1}{3}y + y + \frac{1}{2}x = 1$$

$$\frac{1}{3}y + y + \frac{1}{2}\left(\frac{1}{3}y\right) = 1$$

$$\frac{1}{3}y + y + \frac{1}{6}y = 1$$

$$\frac{2y + 6y + 1y}{6} = 1$$

$$\frac{9y}{6} = 1$$

$$9y = 6$$

$$y = \frac{6}{9}$$

$$\boxed{y = \frac{2}{3}}$$

Substitute  $y = \frac{2}{3}$  in eqn ②

$$\frac{1}{3}y = x \Rightarrow \frac{1}{3}\left(\frac{2}{3}\right) = x \quad \boxed{x = \frac{2}{9}}$$

eqn ④  $\Rightarrow$

$$\frac{1}{2}x = z$$

$$\frac{1}{2}\left(\frac{2}{9}\right) = z$$

$$z = \frac{2}{18}$$

$$\boxed{z = \frac{1}{9}}$$

$$v = (x, y, z) = \left(\frac{2}{9}, \frac{2}{3}, \frac{1}{9}\right)$$

i)  $\frac{2}{9}, \frac{2}{3}, \frac{1}{9} \geq 0$

ii)  $\frac{2}{9} + \frac{2}{3} + \frac{1}{9} = 1$

iii)  $A = \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{4} & \frac{1}{4} \end{bmatrix}$

Soln Let  $v = [x, y, z]$  be the unique fixed probability vector

$$x + y + z = 1 \quad \text{--- ①}$$

$$\text{and } vp = v$$

$$[x, y, z] \begin{bmatrix} 0 & 1 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{4} & \frac{1}{4} \end{bmatrix} = [x, y, z]$$

$$\begin{bmatrix} p & q \\ p & q \end{bmatrix} x = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$[0x + \frac{1}{2}y + \frac{1}{2}z, 1x + 0y + \frac{1}{4}z, 0x + \frac{1}{2}y + \frac{1}{4}z] = [x, y, z]$$

$$0x + \frac{1}{2}y + \frac{1}{2}z = xc \quad 1x + 0y + \frac{1}{4}z = y \quad 0x + \frac{1}{2}y + \frac{1}{4}z = z$$

$$\boxed{y + z = xc} \quad \textcircled{2}$$

$$\frac{4x + 1z}{4} = y$$

$$\frac{2y + 1z}{4} = z$$

$$\boxed{\frac{4x + 1z}{4} = y} \quad \textcircled{3}$$

$$2y + 1z = 4z - 1z$$

$$\boxed{\frac{2y}{3} = z} \quad \textcircled{4}$$

$$2y = 3z$$

Substituting eqn  $\textcircled{2}$  and eqn  $\textcircled{4}$  in eqn  $\textcircled{3}$

$$x + y + z = 1$$

$$y + z + y + \frac{2}{3}y = 1$$

$$y + \frac{2}{3}y + y + \frac{2}{3}y = 1$$

$$2y + \frac{4}{3}y = 1 \quad \textcircled{5}$$

$$\frac{6y + 4y}{3} = 1$$

$$\boxed{z = \frac{2}{10}}$$

eqn  $\textcircled{2}$

$$y + z = xc$$

$$\frac{3}{10} + \frac{2}{10} = xc$$

$$\boxed{\frac{8}{10} = xc \quad xc = \frac{1}{2}}$$

$$\boxed{y = \frac{3}{10}}$$

$$\boxed{10y = 3}$$

$$\frac{10y}{3} = 1$$

$$10y = 3$$

$$v = (x; y, z) = \left( \frac{1}{2}, \frac{3}{10}, \frac{2}{10} \right)$$

i]  $\frac{1}{2}, \frac{3}{10}, \frac{2}{10} \geq 0$

ii]  $\frac{1}{2} + \frac{3}{10} + \frac{2}{10} = 1$

iii]  $A = \begin{bmatrix} 3/4 & y/4 \\ y/2 & y/2 \end{bmatrix}$

Soln. Let  $v = [x, y]$  be the unique fixed probability vector

$x+y=1$  and

$vP=v$

$$[x, y] \begin{bmatrix} 3/4 & y/4 \\ y/2 & y/2 \end{bmatrix} = [x, y]$$

$$\left[ \frac{3}{4}x + \frac{1}{2}y, \frac{1}{4}x + \frac{1}{2}y \right] = [x, y]$$

$$\frac{3}{4}x + \frac{1}{2}y = x \quad \frac{1}{4}x + \frac{1}{2}y = y$$

$$\frac{3x+2y}{4} = x$$

$$\frac{1x+2y}{4} = y$$

$$3x+2y = 4x$$

$$x+2y = 4y \quad \text{---(3)}$$

$$2y = 4x - 3x$$

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

eqn ② in ①

$$x + y = 1$$

$$2y + y = 1$$

$$3y = 1$$

$$\boxed{y = \frac{1}{3}}$$

$$\text{eqn ②} \Rightarrow 2y = 1$$

$$2\left(\frac{1}{3}\right) = x$$

$$\boxed{x = \frac{2}{3}}$$

$$v = (x, y) = \left(\frac{2}{3}, \frac{1}{3}\right)$$

i]  $\frac{2}{3}, \frac{1}{3} \geq 0$

ii]  $\frac{2}{3} + \frac{1}{3} = 1$

Q) Which of the following is a probability vector and give reason

i]  $v_1 = \left(\frac{1}{3}, 0, -\frac{1}{6}, \frac{1}{2}, \frac{1}{3}\right)$  → It is not a probability vector because  $-\frac{1}{6}$  is not greater than or equal to zero.

ii]  $v_2 = \left(\frac{1}{3}, 0, \frac{1}{6}, \frac{1}{2}, \frac{1}{3}\right)$   $v_i \neq 0$

$\Rightarrow v_i \geq 0$

(i.e.,  $-\frac{1}{6} \neq 0$ )

$$\sum v_i = \frac{1}{3} + 0 + \frac{1}{6} + \frac{1}{2} + \frac{1}{3} = 1.3 \neq 1$$

∴ It is not a probability vector

iii]  $v_3 = \left(\frac{1}{3}, 0, 0, -\frac{1}{6}, \frac{1}{2}\right)$

$$v_i \geq 0$$

$$\sum v_i = \frac{1}{3} + 0 + 0 + \frac{1}{6} + \frac{1}{2} = 1 \therefore v_3 \text{ is a probability vector}$$

10)

Show that the matrix

$A = \begin{bmatrix} 0 & 1 & 0 \\ \gamma_2 & 0 & \gamma_1 \\ \frac{1}{2} & \gamma_1 & 0 \end{bmatrix}$  is a regular stochastic matrix and hence find the unique fixed probability vector.

Soln

$$A = \begin{bmatrix} 0 & 1 & 0 \\ \gamma_2 & 0 & 1 \\ \gamma_1 & \gamma_2 & 0 \end{bmatrix}$$

$$A^2 = A \cdot A = \begin{bmatrix} 0 & 1 & 0 \\ \gamma_2 & 0 & 1 \\ \gamma_1 & \gamma_2 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ \gamma_2 & 0 & 1 \\ \gamma_1 & \gamma_2 & 0 \end{bmatrix}$$

$$= \begin{bmatrix} 0 + \gamma_2 + 0 & 0 + 0 + 0 & 0 + 1 + 0 \\ 0 + 0 + \gamma_2 & 0 + 0 + \gamma_2 & 0 + 0 + 0 \\ 0 + 0 + 0 & 0 + \frac{1}{2} + 0 & 0 + \frac{1}{2} + 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 1 \\ \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

$$A^3 = A^2 \cdot A = \begin{bmatrix} 0 & 0 & 1 \\ \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ \gamma_2 & 0 & 1 \\ \gamma_1 & \gamma_2 & 0 \end{bmatrix}$$

$$A^3 = \begin{bmatrix} 0+0+\frac{1}{2} & 0+0+\frac{1}{2} & 0+0+0 \\ 0+0+0 & 0+\frac{1}{2}+0 & 0+\frac{1}{2}+0 \\ 0+0+\frac{1}{4} & 0+0+\frac{1}{4} & 0+\frac{1}{2}+0 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \end{bmatrix}$$

$$A^4 = A^3 \cdot A = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \gamma_2 & \gamma_2 & 0 \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & 0 & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & 0 \end{bmatrix}$$

$$A^5 = \begin{bmatrix} 0+0+0 & \frac{1}{2}+0+0 & 0+\frac{1}{2}+0 \\ 0+0+\frac{1}{4} & 0+0+\frac{1}{4} & 0+\frac{1}{2}+0 \\ 0+0+\frac{1}{4} & \frac{1}{4}+0+\frac{1}{4} & 0+\frac{1}{4}+0 \end{bmatrix} = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & 0 & 0 \\ \frac{1}{4} & \frac{1}{4} & 0 \end{bmatrix}$$

$$A^5 = A^4 \cdot A = \begin{bmatrix} 0 & \frac{1}{2} & 0 \\ \frac{1}{4} & \frac{1}{2} & 0 \\ \frac{1}{4} & \frac{1}{4} & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \gamma_2 & \gamma_2 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 0 \\ \gamma_2 & \gamma_2 & 0 \end{bmatrix}$$

$$A^5 = \begin{bmatrix} 0+0+\frac{1}{4} & 0+\frac{1}{2}+0 & 0+\frac{1}{2}+0 \\ 0+0+\frac{1}{4} & 0+0+\frac{1}{4} & 0+\frac{1}{4}+0 \\ 0+0+\frac{1}{8} & 0+0+\frac{1}{8} & 0+\frac{1}{2}+0 \end{bmatrix} = \begin{bmatrix} \frac{1}{4} & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ \frac{1}{8} & \frac{1}{8} & 0 \end{bmatrix}$$

All the entries of  $A^5$  are +ve  
 All the entries of the matrix of the order  $n=5$   
 $A$  is a stochastic matrix

unique probability vector

Let  $v = [x, y, z]$  be the unique probability vector

$$x + y + z = 1 \quad - \textcircled{1}$$

and  $\nabla \rho = v$

$$[x, y, z] = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

$$\left[ \begin{matrix} 0x + 0y + \frac{1}{2}z \\ 0x + 1y + 0z \end{matrix} \right] = \left[ \begin{matrix} x, y, z \end{matrix} \right]$$

$$\begin{aligned} z &= z_0 + h_1 + xc_0 \\ h &= z^{\frac{1}{2}} + h_0 + xc_1 \\ x &= z^{\frac{1}{2}} + h_0 + xc_0 \end{aligned}$$

$$[1] \quad \frac{d}{dx} \left( \frac{1}{x^2} \right) = -\frac{2}{x^3}$$

$$\frac{2x+12}{2} = 24 \quad (3)$$

gutekula ② zu @ in ①

$$54 = 2$$

8/5  
-  
4

$$\frac{1}{2}(q) + q + q = 1$$

$$\frac{1}{2}y + 2y = 1$$

$$= \frac{1}{\beta_1 + \beta_2}$$

— 11 —

$$\text{eqn ②} \rightarrow y = z$$

$$\boxed{\frac{3}{5} = z}$$

$$\text{eqn ②} \rightarrow \frac{1}{2}z = x$$

$$\frac{1}{2}\left(\frac{3}{5}\right) = x$$

$$x = \frac{2}{10}$$

$$\boxed{x = \frac{1}{5}}$$

$$v = (x, y, z) = \left( \frac{1}{5}, \frac{2}{5}, \frac{3}{5} \right)$$

i)  $\frac{1}{5}, \frac{2}{5}, \frac{3}{5} \geq 0$

ii)  $\frac{1}{5} + \frac{2}{5} + \frac{3}{5} = 1$

What does this require  
Markov chain: It is a set of random variables  $x(t)$  defined on  $S$  with parameter  $t$  which is such that the generation of probability distribution depends only on the present state is called markov process.

Markov chain: A stochastic process which is such that the generation of probability distribution depends only on the present state is called a

If this state space is discrete is called markov chain.

Transition probability matrix: The  $p_{ij}$ , which are non-negative real numbers form a square matrix of order  $m$  is called the transition probability matrix and it satisfies

$$\text{i) } 0 \leq p_{ij} \leq 1$$

$$\text{ii) } \sum_{j=1}^m p_{ij} = 1$$

$$i = 1, 2, 3, \dots, m$$

$$P = [p_{ij}] = \begin{bmatrix} p_{11} & p_{12} & \dots & p_{1m} \\ p_{21} & p_{22} & \dots & p_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ p_{m1} & p_{m2} & \dots & p_{mm} \end{bmatrix}$$

With the help TPM we predict the moment of the system from one state to other state.  
 Ex:- Three boy A, B, C throwing the ball to each other. A always throws ball to B, B always throws ball to C and C just as likely to throw ball to B as to A.

Form the TPM

$$\text{Ans} P = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix}$$

Higher tra  
The probabil  
a, to the  
by  $P^n = [$

Let  $P^{(0)} = [$   
distribution.

Let  $P^{(0)} = [$   
probability

$P^{(0)} = P^{(0)}$

$P^{(0)} = P^{(0)}$

$P^{(0)} = P^{(0)}$

If the ar  
absorbing

state if pi  
just

matrix

Higher transition probability matrix

The probability that the system changes from state  $a$  to the state  $a_j$  in exactly  $m$  steps is defined

$$\text{by } P^n = [P_{ij}^{(n)}] = \begin{bmatrix} p_{11}^{(n)} & p_{12}^{(n)} & \dots & p_{1m}^{(n)} \\ p_{21}^{(n)} & p_{22}^{(n)} & \dots & p_{2m}^{(n)} \\ \vdots & \vdots & \ddots & \vdots \\ p_{m1}^{(n)} & p_{m2}^{(n)} & \dots & p_{mm}^{(n)} \end{bmatrix}$$

Let  $P^{(0)} = [p_1^{(0)}, p_2^{(0)}, \dots, p_n^{(0)}]$  denote initial probability distribution.

Let  $P^{(n)} = [p_1^{(n)}, p_2^{(n)}, \dots, p_n^{(n)}]$  denotes the  $n$ th step probability distribution

the

$$\begin{aligned} P^{(0)} &= P^{(0)} \\ P^{(1)} &= P^{(0)} P^1 \\ P^{(2)} &= P^{(0)} P^2 \\ P^{(3)} &= P^{(0)} P^3 \\ &\vdots \\ P^{(n)} &= P^{(0)} P^n \end{aligned}$$

Inreducible:— The markov chain is said to be irreducible if the associated  $P_m$  is regular.

A absorbing state:— A state  $i$  is called an absorbing state if the transition probability  $P_{ij}$  are such that  $P_{ij} = \begin{cases} 1 & \text{if } i=j \\ 0 & \text{otherwise} \end{cases}$

$$P = \begin{bmatrix} 0 & a_1 & a_2 \\ a_1 & 0 & 0 \\ a_2 & 0 & 0 \end{bmatrix}$$

Problem  
 ] write the  
 Also verify the

$a_2$  state is the  
absorbing state

$a_1$  and  $a_3$  are the  
absorbing states.

$$P = \begin{bmatrix} 1 & a_1 & a_2 \\ a_1 & 0 & 0 \\ a_2 & 0 & 0 \end{bmatrix}$$

Transient state: A state is said to be transient state if the system is in this state at some step and there is a chance it will not return to that state.

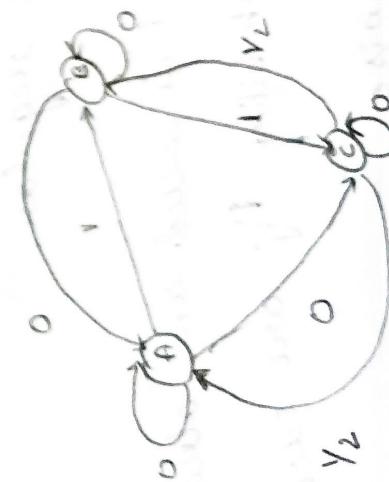
State transition diagram:

A markov chain is usually shown by a state transition diagram.

The pictorial representation of transition probabilities from state  $a_i$  to  $a_j$  is called state transition diagram.

$$\text{Ex: } P = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

There are three states A, B and C



$$P = \begin{bmatrix} 1 & a_1 & a_2 \\ a_1 & 0 & 0 \\ a_2 & 0 & 0 \end{bmatrix}$$

Also verify the

$$P^2 = P \cdot P =$$

$$P^2 = \begin{bmatrix} 1 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 1 & 0 \\ \frac{1}{2} & 0 & 1 \end{bmatrix}$$

### Problems

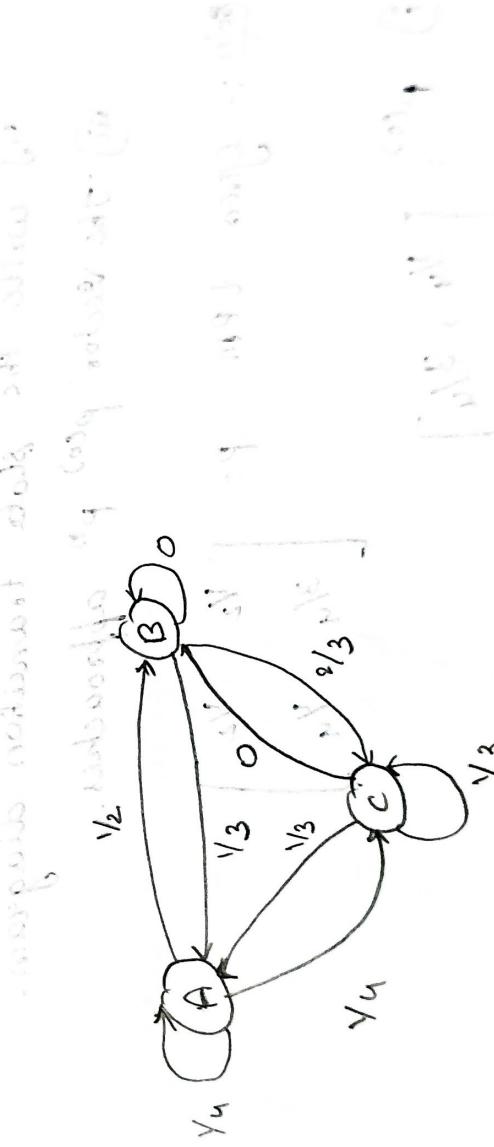
- i) write the state transition diagram of the following T.P.M P  
Also verify that P is irreducible

$$P = \begin{bmatrix} \gamma_4 & \gamma_2 & \gamma_4 \\ \gamma_3 & 0 & 2/3 \\ \gamma_2 & 0 & 1/2 \end{bmatrix}$$

means

Step State transition diagram  
we have three state space states

$$P = \begin{bmatrix} A & \gamma_4 & \gamma_2 \\ B & \gamma_3 & 0 & 2/3 \\ C & \gamma_2 & 0 & 1/2 \end{bmatrix}$$



like

now

$$P^2 = P \cdot P = \begin{bmatrix} \gamma_4 & \gamma_2 & \gamma_4 \\ \gamma_3 & 0 & 2/3 \\ \gamma_2 & 0 & 1/2 \end{bmatrix} \begin{bmatrix} \gamma_4 & \gamma_2 & \gamma_4 \\ \gamma_3 & 0 & 2/3 \\ \gamma_2 & 0 & 1/2 \end{bmatrix} = \begin{bmatrix} \frac{1}{16} + \frac{1}{6} + \frac{1}{8} & \frac{1}{8} + 0 + 0 & \frac{1}{16} + \frac{1}{3} + \frac{1}{8} \\ \frac{1}{12} + 0 + \frac{1}{3} & \frac{1}{6} + 0 + 0 & \frac{1}{12} + 0 + \frac{1}{3} \\ \frac{1}{2} + 0 + \frac{1}{4} & \frac{1}{4} + 0 + 0 & \frac{1}{2} + 0 + 0 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{16} + \frac{1}{6} + \frac{1}{8} & \frac{1}{8} + 0 + 0 & \frac{1}{16} + \frac{1}{3} + \frac{1}{8} \\ \frac{1}{12} + 0 + \frac{1}{3} & \frac{1}{6} + 0 + 0 & \frac{1}{12} + 0 + \frac{1}{3} \\ \frac{1}{2} + 0 + \frac{1}{4} & \frac{1}{4} + 0 + 0 & \frac{1}{2} + 0 + 0 \end{bmatrix}$$

### Problems

1) construct the state transition diagram of the following T.P.M P  
Also verify that P is irreducible

$$P = \begin{bmatrix} \gamma_4 & \gamma_2 & \gamma_4 \\ \gamma_3 & 0 & 2/3 \\ \gamma_2 & 0 & \gamma_2 \end{bmatrix}$$

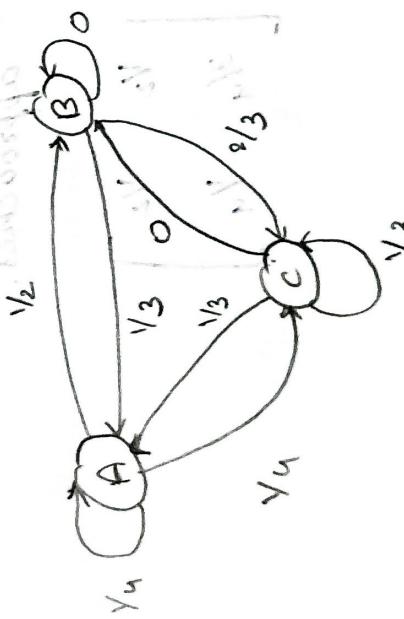
Ans:

Step 1  
Step 2

State transition diagram  
we have three state space states

$$P = A \begin{bmatrix} \gamma_4 & \gamma_2 & \gamma_4 \\ \gamma_3 & 0 & 2/3 \\ \gamma_2 & 0 & \gamma_2 \end{bmatrix}$$

States  
No.



$$P^2 = P \cdot P = \begin{bmatrix} \gamma_4 & \gamma_2 & \gamma_4 \\ \gamma_3 & 0 & 2/3 \\ \gamma_2 & 0 & \gamma_2 \end{bmatrix} \begin{bmatrix} \gamma_4 & \gamma_2 & \gamma_4 \\ \gamma_3 & 0 & 2/3 \\ \gamma_2 & 0 & \gamma_2 \end{bmatrix} = \begin{bmatrix} \frac{1}{16} + \frac{1}{6} + \frac{1}{8} & \frac{1}{8} & \frac{1}{16} + \frac{1}{3} + \frac{1}{8} \\ \frac{1}{12} + 0 + \frac{1}{3} & \frac{1}{6} + 0 + 0 & \frac{1}{12} + 0 + \frac{1}{3} \\ \frac{1}{8} + 0 + 0 & \frac{1}{4} + 0 + 0 & \frac{1}{8} + 0 + 0 \end{bmatrix}$$

$$\begin{bmatrix} 17/48 & 1/8 & 25/48 \\ \frac{1}{6} & \frac{5}{12} & 3/8 \\ \frac{5}{12} & \frac{1}{4} & 3/8 \end{bmatrix}$$

All the entries are +ve  
 $\therefore P$  is irreducible for  $n=2$

Q) The transition probability matrix of markov chain is given by  $P = \begin{bmatrix} 1/2 & \gamma_2 \\ 3/4 & \gamma_4 \end{bmatrix}$  with initial probability

Distribution  $\left( \frac{1}{4}, \frac{3}{4} \right)$

$$\text{Find } \begin{bmatrix} P_{11}^{(2)} \\ P_{12}^{(2)} \end{bmatrix}, P_{21}^{(2)}, P_{22}^{(2)}$$

- i) write the state transition diagram.  
 ii) The vector  $p^{(0)}$  approaches.

$$\text{Soln: Given } t \text{ pm } P = \begin{bmatrix} \gamma_2 & \gamma_2 \\ 3/4 & \gamma_4 \end{bmatrix}$$

$$;\quad P^{(0)} = \begin{bmatrix} \gamma_4 & 3/4 \end{bmatrix}$$

$$P^2 = P \cdot P = \begin{bmatrix} \gamma_2 & \gamma_2 \\ 3/4 & \gamma_4 \end{bmatrix} \begin{bmatrix} \gamma_2 & \gamma_2 \\ 3/4 & \gamma_4 \end{bmatrix} = \begin{bmatrix} \frac{1}{4} + \frac{3}{8} & \frac{1}{4} + \frac{1}{8} \\ \frac{3}{8} + \frac{3}{16} & \frac{3}{8} + \frac{1}{16} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{5}{8} & \frac{9}{16} \\ \frac{9}{16} & \frac{7}{16} \end{bmatrix} = \begin{bmatrix} P_{11}^{(2)} & P_{12}^{(2)} \\ P_{21}^{(2)} & P_{22}^{(2)} \end{bmatrix}$$

Comparing  $P_{21}^{(2)} = \frac{9}{16}$   $P_{12}^{(2)} = \frac{3}{8}$

WRT  $P^{(n)} = P^{(0)} P^n$

$$P^{(0)} = P^{(0)} P^2$$

$$P^2 = \left[ \frac{1}{4}, \frac{3}{4} \right] \begin{bmatrix} \frac{5}{8} & \frac{3/8}{1/16} \\ 9/16 & 7/16 \end{bmatrix}$$

Markov chain

$$\text{probability} = \left[ \frac{5}{32} + \frac{27}{64}, \frac{3}{32} + \frac{27}{64} \right]$$

$\rightarrow$   $P_1^{(2)} = \frac{3/4}{64}$

$$P_2^{(2)} = \left[ \frac{27}{64}, \frac{27}{64} \right] = P^{(2)}$$

$$= \left[ \frac{3/4}{64}, \frac{3/4}{64} \right]$$

Comparing:

$$P_1^{(2)} = \frac{3/4}{64} \quad P_2^{(2)} = \frac{27}{64}$$

[ii] State transition diagram. Ans



[iii] The vector  $P^{(0)} P^n$  approaches to the unique fixed probability vector.

$$\begin{bmatrix} \frac{1}{4} + \frac{1}{8} \\ \frac{3}{8} + \frac{1}{16} \end{bmatrix} \xrightarrow{n \rightarrow \infty} \begin{bmatrix} P_{11}^{(2)} & P_{12}^{(2)} \\ P_{21}^{(2)} & P_{22}^{(2)} \end{bmatrix}$$

Let  $\mathbf{v} = [x, y]$  be the unique fixed probability vector.  $\textcircled{3}$  The

such that  $x + y = 1$   $\textcircled{1}$

and  $\mathbf{v}P = \mathbf{v}$   $\textcircled{2}$

$$[x, y] \begin{bmatrix} v_1 & v_2 \\ v_3 & v_4 \end{bmatrix} = [x, y]$$

$$\left[ \frac{1}{2}x + \frac{3}{4}y, \quad \frac{1}{2}x + \frac{1}{4}y \right] = [x, y]$$

$$P^2 = P$$

$$\frac{1}{2}x + \frac{3}{4}y = x \quad \textcircled{3}$$

$$\frac{1}{2}x + \frac{1}{4}y = y \quad \textcircled{4}$$

$$\frac{3x + 3y}{4} = x$$

$$= \begin{bmatrix} \frac{1}{4} & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{8} \end{bmatrix}$$

$$2x + 3y = 4x$$

$$3y = 4x - 2x$$

$$3y = 2x$$

$$x = \frac{3}{2}y \quad \textcircled{5}$$

earn  $\textcircled{3}$  in  $\textcircled{2}$ .  
x value in  $\textcircled{4}$

$$x + y = 1$$
  
$$\frac{3}{2}y + y = 1$$

$$3y + 2y = 2$$

$$5y = 2$$

$$y = \frac{2}{5} \quad \textcircled{6}$$

$$x = \frac{3}{2} \left( \frac{2}{5} \right)$$

$$x = \frac{6}{10}$$

compa-

nate

(\*) The T PM of a markov chain  $P$  given by

$P = \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 1 & 0 & 0 \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \end{bmatrix}$  and the initial probability distribution  $p^{(0)} = \left(\frac{1}{2}, \frac{1}{2}, 0\right)$  find  $p_{13}^{(1)}$  and  $p_{11}^{(2)}$

$$p_{11}^{(1)} \text{ given } P = \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 1 & 0 & 0 \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \end{bmatrix}$$

$$P^2 = P \cdot P = \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 1 & 0 & 0 \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \end{bmatrix} \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} \\ 1 & 0 & 0 \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{4} + 0 + \frac{1}{8} & 0 + 0 + \frac{1}{4} & \frac{1}{4} + 0 + \frac{1}{8} \\ \frac{1}{2} + 0 + 0 & 0 + 0 + 0 & \frac{1}{2} + 0 + 0 \\ \frac{1}{8} + \frac{1}{2} + \frac{1}{16} & 0 + 0 + \frac{1}{8} & \frac{1}{8} + 0 + \frac{1}{16} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{3}{8} & \frac{1}{4} & \frac{3}{8} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{5}{16} & \frac{1}{8} & \frac{3}{16} \end{bmatrix} = P = \begin{bmatrix} p_{11}^{(0)} & p_{12}^{(0)} & p_{13}^{(0)} \\ p_{21}^{(0)} & p_{22}^{(0)} & p_{23}^{(0)} \\ p_{31}^{(0)} & p_{32}^{(0)} & p_{33}^{(0)} \end{bmatrix}$$

$$\text{composing } p_{13}^{(2)} = \frac{3}{8} p_{23}^{(1)} : \frac{1}{2}$$

$$\text{now } P^2 = p^{(0)} = p^{(0)} p^n$$

$$P^2 = \left(\frac{1}{2}, \frac{1}{2}, 0\right) (P^2)$$

$$P^2 = \begin{pmatrix} 1/2 & 1/2 & 0 \end{pmatrix} \begin{bmatrix} 3/8 & 1/4 & 3/8 \\ 1/2 & 0 & 1/2 \\ 1/16 & 1/8 & 3/16 \end{bmatrix}$$

since  
we have  
 $P^{10}$

$$\begin{aligned} &= \begin{bmatrix} \frac{3}{16} + \frac{1}{4} + 0, & \frac{1}{8} + 0 + 0, & \frac{3}{16} + \frac{1}{4} + 0 \end{bmatrix} \\ &= \begin{bmatrix} \frac{7}{16} & \frac{1}{8} & \frac{7}{16} \end{bmatrix} = \begin{bmatrix} P_1^{(0)}, & P_2^{(0)}, & P_3^{(0)} \end{bmatrix} \end{aligned}$$

$$\text{Comparing } P_1^{(0)} = \frac{7}{16}$$

4] Three boys A, B, C are throwing ball to each other. If A always throws the ball to B and B always throws the ball to C. C is just as likely to throw the ball to B as to A. If C was the first person to throw the ball. Find the probabilities that after three throws.

- i] A has the ball
- ii] B has the ball
- iii] C has the ball.

Given the state space {A, B, C} the associated TPM P is

as follows.

$$P = \begin{array}{c} A \\ B \\ C \end{array} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix}$$

Given person A throws the ball, the associated initial probability vector is given by

$$p^{(0)} = [0, 0, 1]$$

Since the probabilities are desired after three throws we have to find

$$p^{(3)} = p^{(0)} P^3 - \textcircled{1}$$

$$\begin{aligned} P \cdot P &= \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \gamma_2 & \gamma_2 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \gamma_2 & \gamma_2 & 0 \end{bmatrix} \\ &= \begin{bmatrix} 0 & 0 & 1 \\ \gamma_2 & \gamma_2 & 0 \\ 0 & \gamma_2 & 0 \end{bmatrix} \end{aligned}$$

$$\begin{aligned} P^3 &= P^2 \cdot P = \begin{bmatrix} 0 & 0 & 1 \\ \gamma_2 & \gamma_2 & 0 \\ 0 & \gamma_2 & 0 \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \gamma_2 & \gamma_2 & 0 \end{bmatrix} \\ &= \begin{bmatrix} \gamma_2 & \gamma_2 & 0 \\ \gamma_2 & \gamma_2 & \gamma_2 \\ \gamma_4 & \gamma_4 & \gamma_2 \end{bmatrix} \end{aligned}$$

$$\text{eqn } \textcircled{1} \rightarrow p^{(3)} = p^{(0)} P^3$$

$$[0, 0, 1] \begin{bmatrix} \gamma_2 & \gamma_2 & 0 \\ \gamma_2 & \gamma_2 & \gamma_2 \\ \gamma_4 & \gamma_4 & \gamma_2 \end{bmatrix}$$

$$[0, 0, 1] = [\gamma_a, \gamma_b, \gamma_c]$$

$$P^{(3)} = \begin{bmatrix} \gamma_a & \gamma_b & \gamma_c \end{bmatrix}$$

5] A student's studying habit are as follows. If he studies one night, he is 70% sure not to study the next night on the other hand if he does not study one night, he is 60% sure not to study the next night. In the long run how often does he study?

Soln The state space of the system is  $\{A, B\}$

where A: studying

B: not studying.

The associated  $P$  is

$$P = \begin{bmatrix} A & B \\ B & A \end{bmatrix} = \begin{bmatrix} 0.3 & 0.7 \\ 0.4 & 0.6 \end{bmatrix}$$

In order to find the happening in the long run we have to find the unique fixed probability vector  $v(P)$  of  $P$

Let  $v = (x, y)$  be the unique fixed probability vector such that  $x + y = 1$  and

$$\begin{bmatrix} x & y \end{bmatrix} \begin{bmatrix} 0.3 & 0.7 \\ 0.4 & 0.6 \end{bmatrix} = \begin{bmatrix} x & y \end{bmatrix}$$

$$[0.3x + 0.4y, 0.7x + 0.6y] = [x, y]$$

$$0.3x + 0.4y = x \quad \textcircled{1}$$

$$0.7x + 0.6y = y \quad \textcircled{2}$$

$$\frac{0.4}{0.7}y = x \quad \textcircled{3}$$

$$y = \frac{0.7}{0.4}x \quad \textcircled{4}$$

$$y = \frac{7}{4}x \quad \textcircled{5}$$

In long run  
He will

question  
A line  
waiting  
of deci

provide  
input  
arrival

sys  
hop

eqn @ in eqn ①

the  
y one  
at night.

$$x + y = 1$$
$$x + \frac{0.4x}{0.4} = 1$$

$$y = \frac{7}{11}x$$

$$y = \frac{7}{11} \times \frac{4}{11}$$

$$\frac{0.4x + 0.7x}{0.4} = 1$$

$$y = \frac{21}{44} = \frac{7}{11}$$

$$\frac{0.11x}{0.4} = 1$$
$$\frac{11}{4}x = 1$$

$$v = \left( \frac{4}{11}, \frac{7}{11} \right)$$

$$(P_A, P_B)$$

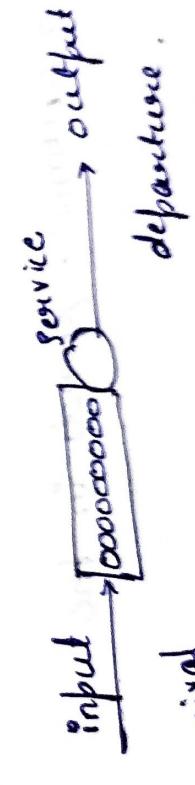
$$x = \frac{4}{11}$$

In long run  
He will stay  $\frac{4}{11} = 0.36 \text{ or } 36\%$ .

e have  
of p

queuing theory  
A line of people or vehicle awaiting for their turn.

Queuing Theory - It is the mathematical study of waiting line / queue. This technique provide basis of decision making about the resources needed to provide service.



customer system  
patient hospital  
service doctor or nurse.

## Queue discipline.

- i] first come first serve
- ii] last come first serve
- iii] service in random order
- iv] Emergency

## Queuing model

### Deterministic queuing model

$\lambda$  = mean no of arrivals per unit time

$\mu$  = mean no of services per unit time

~~in~~ A deli service store employees has 2 cashiers at its counter. If customers arrived on an average of every 5 minutes & one cashier can serve 10 consumers in 5 mins. Assuming poisson's distribution of arrival rate and exponential distribution for service rate find.

i] Average number of customers in the system

ii] Average queue length

iii] Average time a customer spends in the system

iv] Average time a customer waits before being served.

Ques  $\lambda$  = Arrival rate of a customer = 9 per 5 minutes.

$\mu$  = Service rate of customer = 10 per 5 minutes.

i] Average no. of customer in system

$$L_S = \frac{\lambda}{\mu - \lambda} = \frac{9}{10 - 9} = 9$$

ii] Average queue length.

$$L_Q = \frac{\lambda^2}{\mu(\mu - \lambda)} = \frac{9^2}{10(10 - 9)} = \frac{81}{10} = 8.1 \text{ customers per 5 minutes}$$

iii] Average time a customer spends in a system

$$W_S = \frac{1}{\mu - \lambda} = \frac{1}{10 - 9} = 1 \text{ unit.}$$

iv] Average time a customer waits before being served

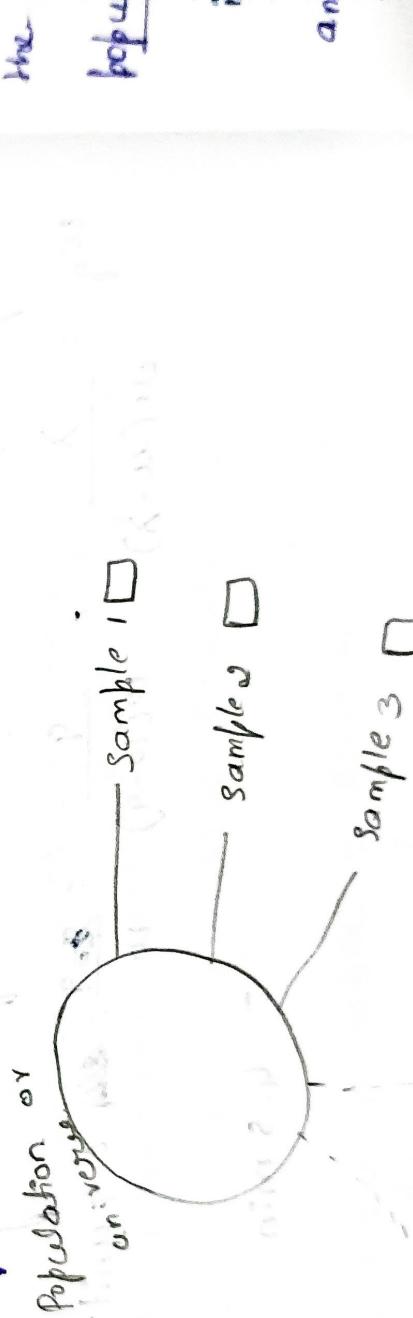
$$W_Q = \frac{1}{\mu(\mu - \lambda)} = \frac{1}{10(10 - 9)} = \frac{1}{10} = 0.1 \text{ min}$$

new desk  $\theta_1$  line Q

## Unit 5 - Sampling and Statistical Inference

more  
sample

- \* A large collection of individuals or attributes or numerical data can be understood as a population or universe.
  - \* A finite subset of the population is called a sample.
  - \* The process of selecting a sample from the population is called a sampling.
  - \* The number of individuals in a population is the size of the population. And is denoted by 'n'.
  - \* The numbers of individual in a sample is called its sample size and is denoted by 'n'.
  - \* The selection of an individual or item from the population in such a way that each has same chance of being selected is called as random sample.



## Sampling distribution

Sampling means members of population may be selected more than once is called as sampling with replacement. On other hand if a member of population is drawn more than once is called as sampling without replacement.

Population is

fixed.

size of

called

chance

Random Sampling with replacement.

\* Here the items are drawn one by one and are

put back to the population before the next draw.

\* If  $n$  is the size of the finite population and  $n$  is

the size of the sample then we have  $n^n$  samples.

\* The mean  $\mu_{\bar{x}}$  of the frequency distribution of the sample means will be equal to the mean of the population

$\sigma^2$  of the frequency distribution of the sample means will be equal to the variance of the population ( $\sigma^2$ )

$$\text{i.e. } \mu_{\bar{x}} = \mu$$

$$\text{and } \frac{\sigma^2}{n} = \sigma^2$$

where  $\sigma^2$

is called standard error of the mean.

a) specific a

Random Sampling with replacement

one by one and are not

\* Here the items are drawn before the next draw.

b) lrb to the population

Sampling where a member of population may be selected more than once is called as sampling with replacement.

On other hand if a member of population cannot be chosen more than once is called as sampling without replacement.

### Random Sampling with replacement.

\* Here the items are drawn one by one and are put back to the population before the next draw.

If  $n$  is the size of the finite population and  $n$  is the size of the sample then we have  $N^n$  samples.

The mean  $\mu_{\bar{x}}$  of the frequency distribution of the sample means will be equal to the mean of the population ( $\mu^2$ )

The variance  $\frac{\sigma^2}{n}$  of the frequency distribution of the sample means will be equal to the variance of the population ( $\sigma^2$ )

$$\text{i.e. } \mu_{\bar{x}} = \mu$$

and  $\frac{\sigma^2}{n} = \sigma^2$   
where  $\frac{\sigma^2}{n}$  is called standard error of the mean.

### Random Sampling without replacement

\* Here the items are drawn one by one and are not put back to the population before the next draw.

- \* If  $n$  is the size of finite population &  $m$  is the size of sample then we have  $N_m^n$  samples.
- \* In this case we have the following results.

$$i) \quad M_{\bar{x}} = M$$

$$ii) \quad \frac{\sigma^2}{\bar{x}} = \left[ \frac{n-n}{n-1} \right] \frac{\sigma^2}{n}$$