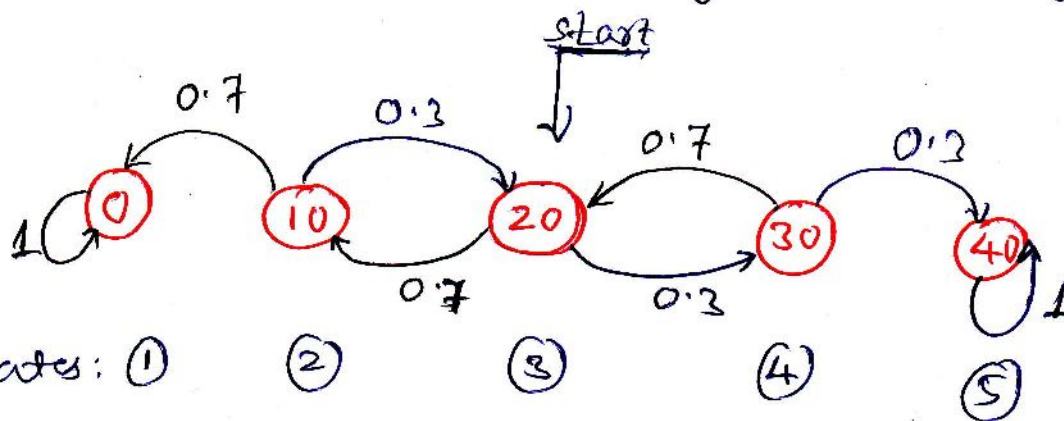


Ex.① Suppose Mr. X has ₹ 20 initially and he bets ₹ 10 each time on a gamble. Probability that Mr. X wins is 0.3 & lose with probability 0.7. If Mr. X has ₹ 0 or ₹ 40 the game is over.

Let X_t is the money Mr. X has after each gamble (each instant of time) & $X_0 = ₹ 20$



Sequence of RVs = $\{X_0, X_1, X_2, X_3, \dots\}$

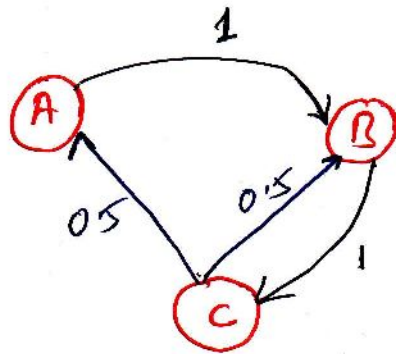
$$P = \begin{matrix} & \begin{matrix} \textcircled{1} & \textcircled{2} & \textcircled{3} & \textcircled{4} & \textcircled{5} \end{matrix} \\ \begin{matrix} \textcircled{1} \\ \textcircled{2} \\ \textcircled{3} \\ \textcircled{4} \\ \textcircled{5} \end{matrix} & \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0.7 & 0 & 0.3 & 0 & 0 \\ 0 & 0.7 & 0 & 0.3 & 0 \\ 0 & 0 & 0.7 & 0 & 0.3 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \end{matrix}$$

→ stochastic process

→ Stochastic matrix or Transition probability matrix (TPM)

Ex ② Three boys A, B & C are throwing ball to each other. A always throws the ball to B and B always throws the ball to C, but C is just likely to throw the ball to B as to A. C begins the game.

Let $X_t: (a, b, c)$, where $a, b \& c$ takes 1 or 0 depending on a, b or c has the ball or not



States: A, B, C

Sequence of RVs = $\{X_0, X_1, X_2, \dots\}$

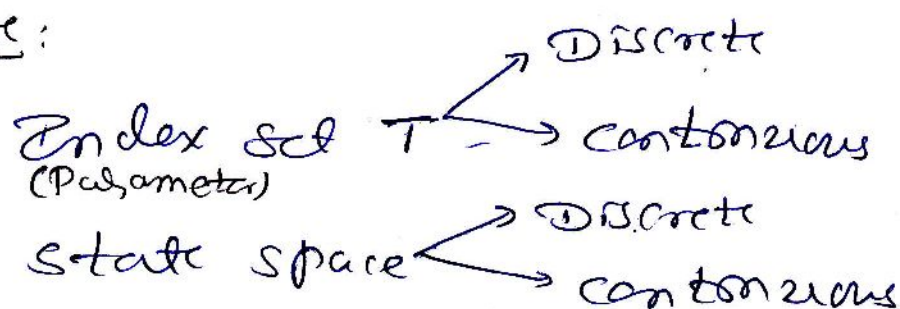
→ stochastic process

$$P = \begin{matrix} & \begin{matrix} A & B & C \end{matrix} \\ \begin{matrix} A \\ B \\ C \end{matrix} & \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 0.5 & 0.5 & 0 \end{bmatrix} \end{matrix} \rightarrow \text{stochastic matrix (TPM)}$$

Stochastic Process: A set or family of random variables $\{X_t; t \in T \subset \mathbb{R}\}$ defined on a sample space S with parameter t is called stochastic process & set T is called index set.

The values taken by random variable X_t are called states & set of all possible values is called state space.

Note:



A discrete space discrete parameter stochastic process $\langle X_t \rangle = \langle X_0, X_1, X_2, \dots \rangle$ is said to be a Markov chain if the probability of the state at time (parameter) $t+1$ depends only on the state at time (parameter) t and does not depend on the states before time t .

$$\text{i.e. } P[X_{t+1} = i_{t+1} | X_t = i_t, X_{t-1} = i_{t-1}, \dots, X_1 = i_1, X_0 = i_0] \\ = P[X_{t+1} = i_{t+1} | X_t = i_t]$$

Stochastic matrix: A square matrix

$P = (P_{ij})$ in which rows & columns represent states of the process is said to be stochastic matrix OR Transition Probability Matrix (TPM)

of

(i) $P_{ij} \geq 0 \quad \forall i, j$ (ii) Row sum: $\sum_j P_{ij} = 1 \quad \forall i$

Here P_{ij} represents probability of process moving from i th state to j th state in single instant of time.

Each row of a stochastic matrix is called probability vector OR any vector $v = \langle v_1, v_2, \dots, v_n \rangle$ is said to be probability vector if

(i) $v_i \geq 0 \quad \forall i$

(ii) $\sum_{i=1}^n v_i = 1$

Ex: $(0, 1)$, $(\frac{1}{2}, \frac{1}{2})$, $(\frac{1}{3}, \frac{1}{3}, \frac{1}{3})$, $(\frac{1}{4}, \frac{1}{2}, \frac{1}{4})$.

A stochastic matrix P is said to be regular if all entries of some power of P are strictly positive.

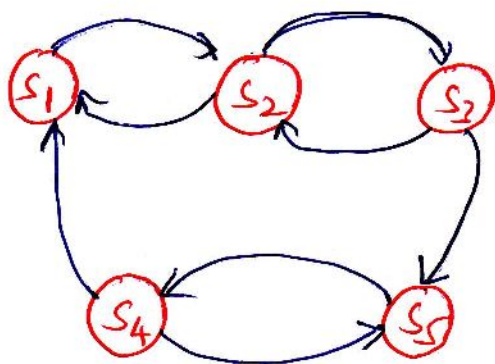
Ex (1) $P = \begin{pmatrix} 0 & 1 \\ \frac{1}{2} & \frac{1}{2} \end{pmatrix}$ is regular because $P^2 = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{3}{4} \end{pmatrix}$.

(2) $P = \begin{pmatrix} \frac{1}{2} & \frac{2}{3} \\ \frac{1}{4} & \frac{3}{4} \end{pmatrix}$ is regular as all entries are positive.

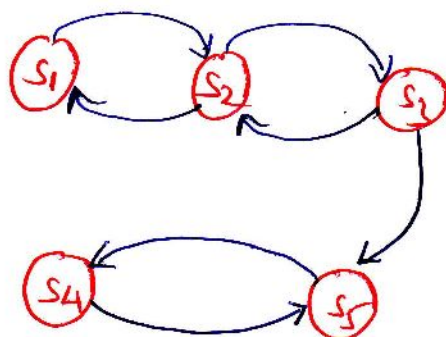
A Markov chain is said to be ~~regular~~ ^{irreducible} if corresponding stochastic matrix is regular. (5)

Meaning: If Markov chain is irreducible then regardless the present state one can reach any other state in finite time.

Ex

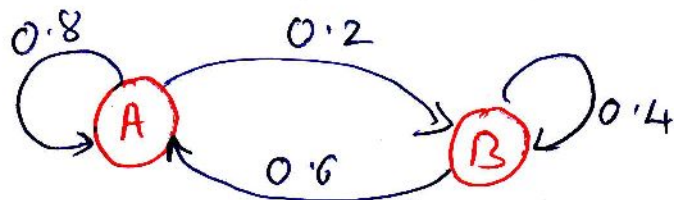


Irreducible



Not irreducible

Example: Throwing ball to each other



TPM: $P = \begin{pmatrix} 0.8 & 0.2 \\ 0.4 & 0.6 \end{pmatrix}$

A begins the game:

Initial Prob. vector: $P^{(0)} = (1 \ 0)$

After 1st throw: $P^{(1)} = P^{(0)} P = (1 \ 0) \begin{pmatrix} 0.8 & 0.2 \\ 0.4 & 0.6 \end{pmatrix} = (0.8 \ 0.2)$

After 2nd throw: $P^{(2)} = P^{(1)} P = (0.8 \ 0.2) \begin{pmatrix} 0.8 & 0.2 \\ 0.4 & 0.6 \end{pmatrix} = (0.72 \ 0.28)$

After 3rd throw: $P^{(3)} = P^{(2)} P = (0.688 \ 0.312)$

After 7th throw: $P^{(7)} = P^{(6)} P = (0.667 \ 0.333)$

After 8th throw: $P^{(8)} = P^{(7)} P = (0.667 \ 0.333)$

After 9th throw: $P^{(9)} = P^{(8)} P = (0.667 \ 0.333)$

?

(6)

Here $(0.667 \ 0.333) = (\frac{2}{3} \ \frac{1}{3})$ is called fixed probability vector or steady state probability vector.

Meaning: If game continues for a long or on a long run 66% of the time ball is with A & 33% of the time ball is with B.

Property of fixed probability vector

$$\rightarrow (\frac{2}{3} \ \frac{1}{3}) P = (\frac{2}{3} \ \frac{1}{3}) \begin{pmatrix} 0.8 & 0.2 \\ 0.4 & 0.6 \end{pmatrix} = (\frac{2}{3} \ \frac{1}{3})$$

In general, if v is a fixed probability vector of a process then $vP = v$.

Remember:

- (i) For 2 state process: $v = (x \ y), \ x + y = 1$
- (ii) For 3 state process: $v = (x \ y \ z), \ x + y + z = 1$

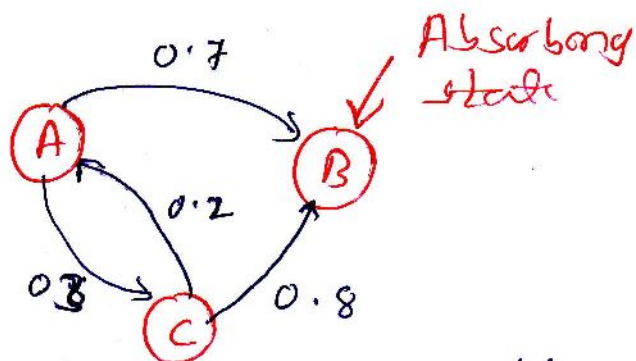
Different states

(7)

① Absorbing state

In a Markov chain,

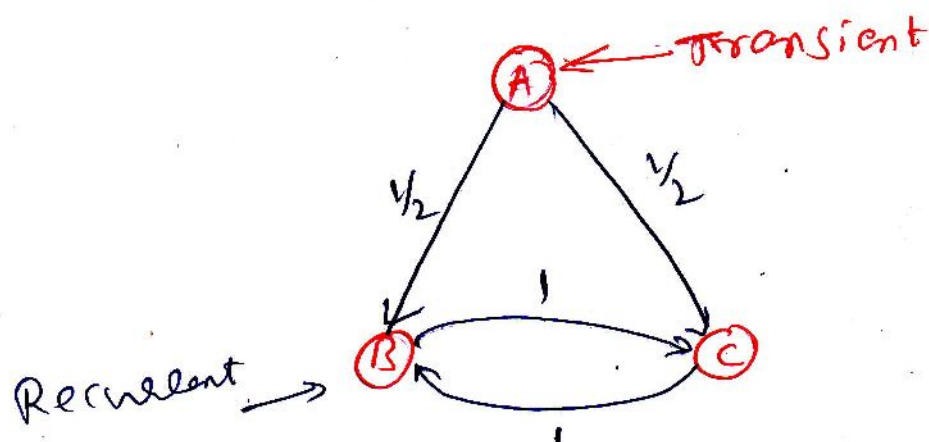
If the process reaches to a certain state after which it continues to remain in the same state is called absorbing state



② Recurrent state: A state i is said

to be recurrent if starting from state i the chain (process) eventually return to the state i with probability 1

③ Transient state: A state i is said to be transient (non-recurrent) if there is positive probability that the process will not return to that state.



Examples

Ex (1) Find fixed probability vector of a regular stochastic matrix

$$P = \begin{pmatrix} 1/4 & 3/4 \\ 1/2 & 1/2 \end{pmatrix}$$

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

$$\Rightarrow x + y = 1 \rightarrow \textcircled{1}$$

we have

$$vP = v$$

$$\Rightarrow (x \ y) \begin{pmatrix} 1/4 & 3/4 \\ 1/2 & 1/2 \end{pmatrix} = (x \ y)$$

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

$$\Rightarrow y/2 = 3x/4 \quad \& \quad \cancel{y/2} \frac{3x}{4} = y/2$$

$$\Rightarrow y = \frac{3x}{2} \quad \& \quad y = \frac{3x}{2} \quad \text{From } \textcircled{1} \quad x + y = 1$$

$$\Rightarrow x + \frac{3x}{2} = 1 \quad \therefore x = 2/5 \quad \& \quad y = 3/5$$

$$\therefore v = (x \ y) = (2/5 \ 3/5)$$

Ex (2) Find fixed probability vector of a regular stochastic matrix

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

Solⁿ Let $v = (x \ y \ z)$ be fixed probability vector of P & $x + y + z = 1 \rightarrow \textcircled{1}$

(9)

we have

$$\mathcal{V}P = \mathcal{V} \Rightarrow (x \ y \ z) \begin{pmatrix} 0 & 1 & 0 \\ 1/6 & 1/2 & 1/3 \\ 0 & 2/3 & 1/3 \end{pmatrix} = (x \ y \ z)$$

$$\Rightarrow \frac{y}{6} = x, \quad x + \frac{y}{2} + \frac{2z}{3} = y, \quad \frac{y}{3} + \frac{z}{3} = z$$

$$\Rightarrow \boxed{y = 6x} \quad x + \frac{2z}{3} = \frac{y}{2} \quad \frac{y}{3} = \frac{2z}{3}$$

$$\Rightarrow \boxed{y = 6x} \quad \boxed{y = 2z}, \quad \boxed{x = z/3}$$

From ① $x + y + z = 1 \Rightarrow x + 6x + 3x = 1$

$$\therefore x = 1/10, \quad y = 6/10, \quad z = 3/10$$

$$\therefore \mathcal{V} = (1/10 \ 6/10 \ 3/10)$$

ex ③ Prove that a Markov chain whose transition probability matrix is

$$P = \begin{pmatrix} 0 & 2/3 & 1/3 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} \text{ is irreducible and also}$$

determine the corresponding stationary probability vector.

Solⁿ $P = \begin{pmatrix} 0 & 2/3 & 1/3 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} = \frac{1}{6} \begin{pmatrix} 0 & 4 & 2 \\ 3 & 0 & 3 \\ 3 & 3 & 0 \end{pmatrix}$

$$P^2 = P \cdot P = \frac{1}{36} \begin{pmatrix} 0 & 4 & 2 \\ 3 & 0 & 3 \\ 3 & 3 & 0 \end{pmatrix} \begin{pmatrix} 0 & 4 & 2 \\ 3 & 0 & 3 \\ 3 & 3 & 0 \end{pmatrix}$$

$$= \frac{1}{36} \begin{pmatrix} 18 & 6 & 12 \\ 9 & 21 & 6 \\ 9 & 12 & 15 \end{pmatrix}$$

All entries of P^2 are strictly positive, hence P is regular & Markov chain is irreducible

(9)

Let $v = (x \ y \ z)$ be fixed probability vector
 & $x + y + z = 1$ — (1)

we have

$$vP = v \Rightarrow (x \ y \ z) \begin{pmatrix} 0 & 2/3 & 1/3 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} = (x \ y \ z)$$

$$\Rightarrow \frac{y}{2} + \frac{z}{2} = x; \quad \frac{2x}{3} + \frac{z}{2} = y; \quad \frac{x}{3} + \frac{y}{2} = z$$

$$\Rightarrow y + z = 2x \text{ — (2)} \quad 4x + 3z = 6y \text{ — (3)} \quad 2x + 3y = 6z \text{ — (4)}$$

From (1) & (2) $x + 2x = 1 \therefore x = 1/3$

From (3) & (4)

$$\begin{aligned} \frac{4}{3} + 3z &= 6y \Rightarrow 4 + 9z = 18y \\ \frac{2}{3} + 3y &= 6z \Rightarrow 2 + 9y = 18z \end{aligned}$$

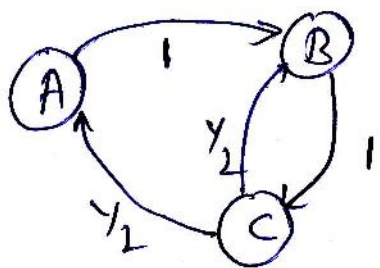
$$\begin{aligned} \Rightarrow 4 + 18y + 9z &= 0 \Rightarrow 8 - 27z = 0 \therefore z = 8/27 \\ 4 + 18y - 36z &= 0 \Rightarrow y = 10/27 \end{aligned}$$

$$\therefore v = \left(\frac{1}{3} \quad \frac{10}{27} \quad \frac{8}{27} \right)$$

Ex (4) Three boys A, B & C are throwing ball to each other. A always throws the ball to B & B always throws the ball to C, but C is just likely to throw the ball to B as to A. If C was the person to throw the ball, find the probabilities that (i) A has the ball (ii) B has the ball (iii) C has the ball, after 3 throws.

Solⁿ

(10)



$$P = \begin{matrix} & \begin{matrix} A & B & C \end{matrix} \\ \begin{matrix} A \\ B \\ C \end{matrix} & \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1/2 & 1/2 & 0 \end{pmatrix} \end{matrix}$$

Initial c has the ball $\therefore p^{(0)} = (0 \ 0 \ 1)$

After 1st throw: $p^{(1)} = p^{(0)} p$

$$\Rightarrow p^{(1)} = (0 \ 0 \ 1) \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1/2 & 1/2 & 0 \end{pmatrix} = (1/2 \ 1/2 \ 0)$$

After 2nd throw: $p^{(2)} = p^{(1)} p$

$$\Rightarrow p^{(2)} = (1/2 \ 1/2 \ 0) \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1/2 & 1/2 & 0 \end{pmatrix} = (0 \ 1/2 \ 1/2)$$

After 3rd throw: $p^{(3)} = p^{(2)} p$

$$\Rightarrow p^{(3)} = (0 \ 1/2 \ 1/2) \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1/2 & 1/2 & 0 \end{pmatrix} = (1/4 \ 1/4 \ 1/2)$$

After 3 throws, probability that

(i) A has ball $= 1/4 = 0.25$

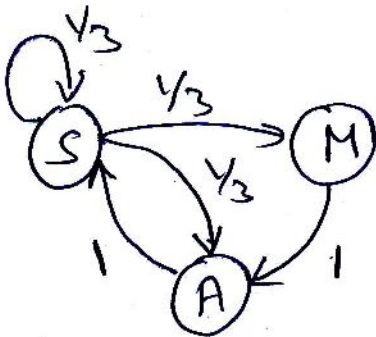
(ii) B has ball $= 1/4 = 0.25$

(iii) C has ball $= 1/2 = 0.5$

Ex(5) Every year a man trades his car for a new car. If he has Maruti, he trades it for an Ambassador, if he has an ambassador he trades it ~~for~~ for Santro, however if he has Santro, he is just likely to trade it for a new Santro as to trade for a Maruti or an Ambassador. In 2016, he bought his

first car which was santro, find the ⁽ⁱⁱ⁾
 probability that he has (i) 2018 santro
 (ii) 2018 Maruti.

Solⁿ



$$\therefore P = \begin{matrix} & \begin{matrix} M & A & S \end{matrix} \\ \begin{matrix} M \\ A \\ S \end{matrix} & \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1/3 & 1/3 & 1/3 \end{pmatrix} \end{matrix}$$

Initially (ie in 2016) he has Santro

$$\therefore P^{(0)} = (0 \ 0 \ 1)$$

After 1 year (in 2017): $P^{(1)} = P^{(0)} P$

$$\Rightarrow P^{(1)} = (0 \ 0 \ 1) \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1/3 & 1/3 & 1/3 \end{pmatrix} = (1/3 \ 1/3 \ 1/3)$$

After 2 years (in 2018): $P^{(2)} = P^{(1)} P$

$$\Rightarrow P^{(2)} = (1/3 \ 1/3 \ 1/3) \begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1/3 & 1/3 & 1/3 \end{pmatrix} = (1/9 \ 4/9 \ 4/9)$$

\therefore In 2018, probability of having

(i) Maruti = $1/9$

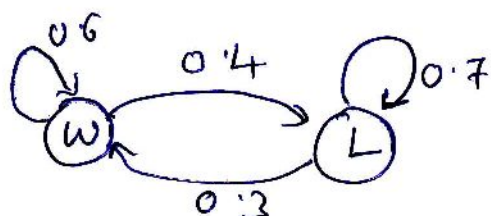
(ii) Ambassador = $4/9$

(iii) Santro = $4/9$

ex 6) A man's gambling luck follows a pattern. If he wins a game the probability of winning the next game is 0.6. However, if he loses ~~the~~ ^a game, the probability of losing the next game is 0.7. There is an even chance that he wins the first game.

- (i) Find prob of winning 3rd game
(ii) In long run how often he wins

Solⁿ



$$P = \begin{matrix} & \begin{matrix} W & L \end{matrix} \\ \begin{matrix} W \\ L \end{matrix} & \begin{pmatrix} 0.6 & 0.4 \\ 0.3 & 0.7 \end{pmatrix} \end{matrix}$$

(i) Initially (prob. he wins 1st game): $P^{(0)} = (\frac{1}{2} \ \frac{1}{2})$

After 1st game (prob. he wins 2nd game):

$$P^{(1)} = P^{(0)} P = (\frac{1}{2} \ \frac{1}{2}) \begin{pmatrix} 0.6 & 0.4 \\ 0.3 & 0.7 \end{pmatrix} = (0.45, 0.55)$$

After 2nd game (prob. of winning 3rd game):

$$P^{(2)} = P^{(1)} P = (0.45 \ 0.55) \begin{pmatrix} 0.6 & 0.4 \\ 0.3 & 0.7 \end{pmatrix} = (0.435, 0.565)$$

\therefore Probability of winning 3rd game = 0.435

(ii) Let $v = (x \ y)$ be fixed probability vector & $x + y = 1$.

$$\text{we have } vP = v \Rightarrow (x \ y) \begin{pmatrix} 0.6 & 0.4 \\ 0.3 & 0.7 \end{pmatrix} = (x \ y)$$

$$\Rightarrow 0.6x + 0.3y = x \text{ \& } 0.4x + 0.7y = y$$

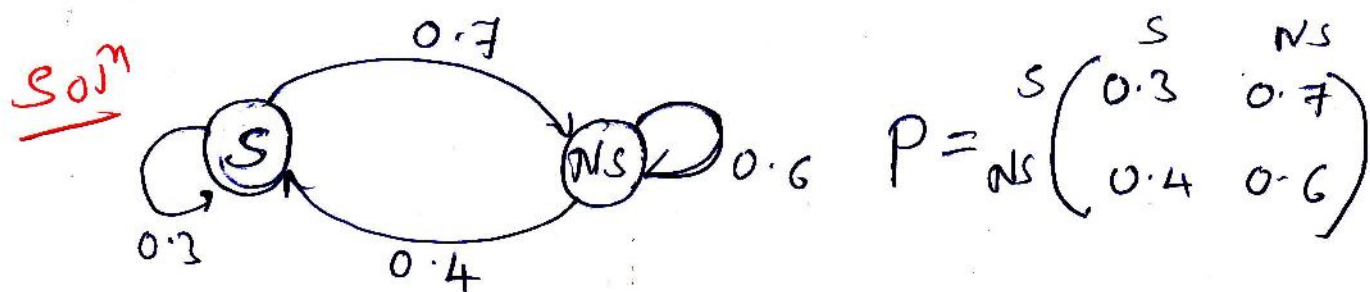
$$\Rightarrow 0.3y = 0.4x \text{ \& } 0.4x = 0.3y$$

$$\Rightarrow y = \frac{4}{3}x \text{ \& } x = \frac{3}{4}y$$

$$\text{From (i) } x + y = 1 \Rightarrow x + \frac{4}{3}x = 1 \therefore x = \frac{3}{7} \text{ \& } y = \frac{4}{7}$$

$\therefore v = (\frac{3}{7}, \frac{4}{7})$ \therefore In long run, he wins $\frac{3}{7}$ of times

Ex 7) A student's study habits are as follows, if he studies one night he is 70% sure not to study next night. on the other hand if he does not study one night he is 60% sure not to study the next night as well. In long run how often does he study?



Let $v = (x \ y)$ be fixed probability vector
 & $x + y = 1$

We have $vP = v \Rightarrow (x \ y) \begin{pmatrix} 0.3 & 0.7 \\ 0.4 & 0.6 \end{pmatrix} = (x \ y)$

$$\Rightarrow 0.3x + 0.4y = x \quad \& \quad 0.7x + 0.6y = y$$

$$\Rightarrow 0.4y = 0.7x \quad \& \quad 0.7x = 0.4y$$

$$\Rightarrow y = \frac{7}{4}x \quad \& \quad x = \frac{4}{7}y$$

From $x + y = 1 \Rightarrow x + \frac{7}{4}x = 1 \therefore x = \frac{4}{11} \quad \& \quad y = \frac{7}{11}$

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

In long run, he studies $\frac{4}{11}$ i.e. 36.36% of the time.