

Definition:

A random process in which the future value depends only on the present value and not on the past value is called MARKOV PROCESS.

Markov chain:

$$\begin{aligned} & \text{Markov chain if } P[x_n = a_n / x_{n-1} = a_{n-1}, x_{n-2} = a_{n-2}, \dots, x_0 = a_0] \\ & = P[x_n = a_n / x_{n-1} = a_{n-1}] \end{aligned}$$

Then the random variables $\{x_n\}$ is called Markov chain, $n=0, 1, 2, \dots$ etc and a_1, a_2, \dots, a_n are called states of Markov chain.

Transition probability:

$P[x_m = a_i] = P_i(m)$ represents the probability that at time $t = t_m$, the system occupies the state a_i .

$P[x_n = a_j / x_m = a_i]$, $P_{ij}(m,n)$, represent the prob that the system goes into state a_j at $t = t_n$ given that it was in state a_i at $t = t_m$. The no's

$P_{ij}(m,n)$ represent the transition property of the Markov chain from state a_i to a_j .

One step transition probability

The conditional probability

$P[x_{n+1} = a_j / x_n = a_i]$ is called one step transition property.

from a_i at time t_n to state a_j at time t_{n+1} in one step.

one step transition prob. denoted by $P_{ij}^{(n, n+1)}$.
A markov chain is said to be homogeneous in time if one step transition prob. is independent of the step.

n step transition probability.

The conditional prob $P[x_n = a_j | x_0 = a_i]$ is called n step transition prob. and it is denoted by $P_{ij}^{(n)}$.

Probability vector and stochastic matrix
A vectorial (row matrix) $[P_1, P_2, \dots, P_n]$ is called prob. vector.

i) $P_{ij} \geq 0, \forall i, j \in \{1, 2, \dots, n\}$
ii) $\sum_{i=1}^n P_{ij} = 1, \forall j \in \{1, 2, \dots, n\}$

A square matrix is called a stochastic matrix if each row of P is a prob. vector.

Chapman - Kolmogorov equation:

The n^{th} step transition prob. can be computed using EK equation which can be stated as $P_{ij}^{(M+n)} = \sum_{k=0}^{\infty} P_{ik}^{(n)} P_{kj}^{(M)}$,
 $\forall n, m, i, j \geq 0$

Note: If P is a transition prob matrix of a regular markov chain then $\pi P = \pi$ where $\pi = (\pi_1, \pi_2)$ and $\pi_1 + \pi_2 = 1$ given. The invariant prob or the stationary probability.

Remark:

when Markov chain shows steady state or stationary behaviour. Then system in the long run has an invariant prob.

Classification of Markov chain:

i) Irreducible Markov chain : A markov chain is said to irreducible if every state is reachable from every other state i.e $P_{ij}^{(n)} > 0$, for some n and i, j .

ii) Non-null persistent : A finite and irreducible markov chain is non-null persistent markov chain.

iii) A-periodic (m-periodic) :-
The period d_i of a return state i is the GCD of all integer m such that $P_{ii}^{(m)} > 0$, $d_i = \text{GCD}$ { $m | P_{ii}^{(m)} > 0$ }.

The state i is periodic with period d_i state i is A-periodic if $d_i = 1$.

iv) Ergodic :-
A non-null persistent, A-periodic markov chain is called Ergodic Markov chain.

STAT

Problems:

1) Find the invariant probability for Markov chain

$\{X_n\}$, $n \geq 1$ with the state space $\{0, 1, 2\}$ and

one step TPM is given by $P = \begin{bmatrix} 0 & 1 & 1 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{bmatrix}$

Solution:

The invariant prob is $\pi_p = \pi$ anywhere $\pi = (\pi_1, \pi_2)$

and $\pi_1 + \pi_2 = 1$

$$\text{and } (\pi_1, \pi_2) \begin{pmatrix} 0 & 1 & 1 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{pmatrix} = (\pi_1, \pi_2) \quad \pi_p = \frac{\pi}{3}$$

$$\text{With } \left(\frac{\pi_2}{2}, \pi_1 + \pi_2\right) = C(\pi_1, \pi_2) \quad P = \begin{pmatrix} \pi_1 & \pi_2 \\ \frac{\pi_2}{2} & \pi_1 + \pi_2 \end{pmatrix}$$

$$\begin{cases} \pi_1 = \frac{\pi_2}{2} \\ \pi_2 = \pi_1 + \frac{\pi_2}{2} \end{cases} \quad \Rightarrow \quad \begin{cases} \pi_2 = 2\pi_1 \\ \pi_2 = 3\pi_1 \end{cases} \quad \Rightarrow \quad \pi_1 = \frac{1}{3}, \pi_2 = \frac{2}{3}$$

$$\pi_1 + 2\pi_1 = 1.$$

$$3\pi_1 = 1$$

$$\pi_1 = 1/3$$

$$\pi_1 + \pi_2 = 1/3 + 2/3 = 1$$

$$\text{State 0 is a sing. state} \quad \pi_1 = 1/3 \quad \text{is a sing. state}$$

$$1 = \pi_2 = 2/3 \quad \text{is a sing. state}$$

$$\text{State 1 is a sing. state} \quad \pi_2 = 2/3 \quad \text{is a sing. state}$$

$$\text{State 2 is a sing. state} \quad \pi_1 = 1/3 \quad \text{is a sing. state}$$

Q) Let $P = \begin{pmatrix} 3/4 & 1/4 \\ 1/2 & 1/2 \end{pmatrix}$ be the pm of a state Markov chain. Find the stationary prob. of the chain.

Solution

The stationary prob. $\pi p = \pi$ where $\pi = (\pi_1, \pi_2)$ and $\pi_1 + \pi_2 = 1$.

$$(\pi_1, \pi_2) \begin{pmatrix} 3/4 & 1/4 \\ 1/2 & 1/2 \end{pmatrix} = (\pi_1, \pi_2)$$

$$\left(\frac{3\pi_1}{4} + \frac{\pi_2}{2}, \frac{\pi_1}{4} + \frac{\pi_2}{2} \right) = (\pi_1, \pi_2)$$

$$\pi_1 = \frac{3\pi_1}{4} + \frac{\pi_2}{2} \quad \pi_2 = \frac{\pi_1}{4} + \frac{\pi_2}{2}$$

$$\pi_2 = \frac{\pi_1}{4} + \frac{\pi_2}{2}$$

$$(\pi_1 - \pi_2) = \frac{3\pi_1}{4} - \frac{\pi_1}{4} \Rightarrow \frac{2\pi_1}{4} = \frac{2\pi_1}{2}$$

$$\pi_1 - \pi_2 = \frac{\pi_1}{2}$$

$$-\pi_2 = \frac{\pi_1}{2} - \pi_1$$

$$\pi_2 = \frac{\pi_1}{2} \Rightarrow \pi_1 = 2\pi_2$$



$$\begin{aligned} \pi_1 + \pi_2 &= 1 \\ 2\pi_2 + \pi_2 &= 1 \\ \pi_2 &= 1/3 \end{aligned}$$

$$\pi = (\pi_1, \pi_2)$$

$$= (2/3, 1/3)$$

3) A college student x has the following study habits. If he studies one night he is 70% sure not to next night. If he doesn't study one night he is only 60% sure not to study next night also.

i) find tpm.

ii) How often he studies in a long time?

$$\text{long time} = \frac{59}{8} + \frac{192}{8} = 27$$

Solution:

Since the study pattern depends on the present not on the past it is

a markov chain where the states

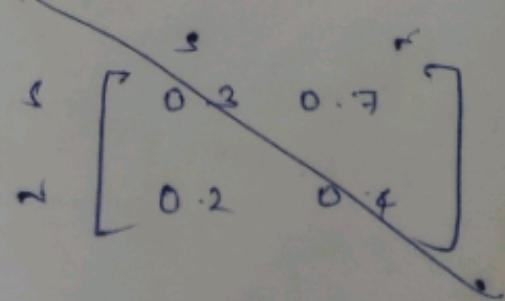
are studying (s), are not studying (n).

i) Find tpm: $s \rightarrow s$

$$P = \begin{bmatrix} s \\ n \end{bmatrix}$$

$(s \rightarrow s) = p$

$(s \rightarrow n)$:



i) If π :

$$\text{then } \begin{bmatrix} s & n \\ 0.3 & 0.7 \\ 0.4 & 0.6 \end{bmatrix}$$

Let $\pi = (\pi_1, \pi_2)$ be the long run then

Tip = π .

while $\pi_1 + \pi_2 = 1$.

for long run we have to find stationary.

$$(\pi_1, \pi_2) \begin{pmatrix} 0.3 & 0.7 \\ 0.4 & 0.6 \end{pmatrix} = (\pi_1, \pi_2)$$

$$0.3\pi_1 + 0.4\pi_2 = 0.7\pi_1 + 0.6\pi_2 = (\pi_1, \pi_2)$$

$$\pi_1 = 0.3\pi_1 + 0.4\pi_2$$

$$\pi_1 - 0.3\pi_1 = 0.4\pi_2 \Rightarrow$$

$$0.7\pi_1 = 0.4\pi_2$$

$$\frac{0.7\pi_1}{0.4} = \frac{\pi_2}{\pi_1} \Rightarrow \pi_1 = \frac{0.4\pi_2}{0.7}$$

0.4 value is to be .. what

$$0.7\pi_1 + 0.6\pi_2 = \pi_2$$

$$0.7\pi_1 = 0.4\pi_2$$

$$1.7\pi_1 = 0.571\pi_2$$

$$= 0.571 \times 0.636$$

$$\frac{0.4\pi_2}{0.7} = \frac{\pi_2}{\pi_1}$$

$$1.041\pi_2 = 0.363$$

$$0.4\pi_2 + \pi_2 = 1$$

$$\pi = (\pi_1, \pi_2) = (0.363, 0.636)$$

$$\frac{0.4\pi_2}{0.7} + \pi_2 = 1$$

$$(1) \quad 0.4\pi_2 + 0.7\pi_2 = 1 \Rightarrow \pi_2 = 0.6364$$

$$1.571 = 1 \Rightarrow \pi_2 = 0.6364$$

$$1.571 = 1 \Rightarrow \pi_2 = 0.6364$$

//

$$\begin{pmatrix} 1 & 0 \\ 0.7 & 1 \end{pmatrix} \begin{pmatrix} 0.4 & 0.3 \\ 0.3 & 0.6 \end{pmatrix} =$$

$$\begin{pmatrix} 1 & 0 \\ 0.7 & 1 \end{pmatrix} \cdot (0.4 + 0.3, 0.3 + 0.6) =$$

A man either drives a car or catches a train to go to office every day. He never goes in a row by train. But if he drives one day, then the next day is just as likely to drive again as he is to travel by train. Suppose that on the first day of the week the man tossed a fair die, and ~~then~~ to work if 6 appeared.

Find the prob. that it takes a train on the third day. Prob. that he drives to work is the long run.

Solution
The travel pattern of tomorrow depends on today. \therefore It is a Markov chain.

$$\text{TPM} \quad P = \begin{bmatrix} T & C \\ C & C \end{bmatrix} = \begin{bmatrix} 0 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}$$

Initial state probability distribution is obtained by throwing a die.

$$P = \left(\frac{5}{6} \quad \frac{1}{6} \right)$$

ii) P (takes a train on the 3rd day)

$$P^{(2)} = P^{(1)} \cdot P$$

$$= \left(\frac{5}{6} \quad \frac{1}{6} \right) \begin{pmatrix} 0 & 1 \\ 1/2 & 1/2 \end{pmatrix}$$

$$= \left(\frac{1}{12} \quad \frac{5}{6} + \frac{1}{6} \right) = \left(\frac{1}{12} \quad \frac{11}{12} \right)$$

$$P^{(2)} = P^{(1)} \cdot P$$

$$= \left(\frac{1}{12}, \frac{11}{12} \right) \left(\frac{1}{12}, \frac{1}{12} \right)$$

$$= \left(\frac{1}{24}, \frac{1}{12}, \frac{11}{24} \right) \text{ will be } P$$

$$= \left(\frac{11}{24}, \frac{13}{24} \right) \text{ will be } P$$

$$P(\text{train in the 3rd day}) = \frac{13}{24}$$

Let $\pi = (\pi_1, \pi_2)$ be the probability long run.

$$\therefore \pi P = \pi \text{ where } \pi_1 + \pi_2 = 1. \quad (1)$$

$$(P_{11}, P_{12}) \begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{12} & \frac{11}{12} \end{pmatrix} = (\pi_1, \pi_2)$$

$$\left(\frac{\pi_1}{2} + \frac{\pi_2}{12}, \frac{\pi_1 + \pi_2}{2} \right) = (\pi_1, \pi_2) \quad (2)$$

$$\frac{\pi_2}{2} = \pi_1 \text{ and } \frac{\pi_1 + \pi_2}{2} = \pi_2$$

$$\begin{pmatrix} \pi_1 & \pi_2 \\ \pi_2 & \pi_1 \end{pmatrix} = (2\pi_1, 1) \quad \pi_2 = 2\pi_1$$

$$\text{from (1). } \pi_1 + \pi_2 = 1.$$

$$\pi_1 + 2\pi_1 = 1 \quad \pi_1 = 1/3. \quad \therefore \pi_2 = 2/3.$$

$$\begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{12} & \frac{11}{12} \end{pmatrix} \begin{pmatrix} \frac{1}{3} & \frac{2}{3} \\ \frac{2}{3} & \frac{1}{3} \end{pmatrix} = P(\text{drives}) = \frac{2}{3} //$$

$$\begin{pmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{12} & \frac{11}{12} \end{pmatrix}$$

$\therefore \frac{1}{2} = \text{(prob prob in 10^8 & prob)}$

- 5) Suppose that the probability of a dry day following rainy day is $\frac{1}{3}$ and the prob of rainy day following a dry day is $\frac{1}{2}$. If May 1st is a dry day, find the prob. that
- i) May 3rd is a dry day again.
 - ii) May 5th is a dry day.

$$P = \begin{bmatrix} D & R \\ R & D \end{bmatrix} \quad P^0 = \begin{bmatrix} 1/2 & 1/2 \\ 1/3 & 2/3 \end{bmatrix}$$

Initial state probability distribution is obtained

by May $p^{(0)}$ is a dry day:

$$p^{(0)} = \begin{bmatrix} D \\ R \end{bmatrix} \quad p^2 = p^0 \cdot p$$

$$(i) \quad P(\text{May } 3^{\text{rd}} \text{ is a dry day}) = p^2 = p^0 \cdot p^2 = p^4$$

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

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

$$p^{(4)} = p^{(2)} \cdot p^2$$

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

$$= \begin{pmatrix} 5/12 & 7/12 \end{pmatrix}$$

$$P(\text{May } 3^{\text{rd}} \text{ is dry day}) = 5/12$$

dry day
the prob of
 $\frac{1}{2}$.

the prob that

ii) $P(\text{May } 5^{\text{th}} \text{ is a dry day})$

$$P^{(4)} = P^{(3)} \cdot P$$

$$= \begin{pmatrix} 2/12 & 7/12 \\ 1/12 & 11/12 \end{pmatrix} \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix}$$

$$= \begin{pmatrix} 29/72 & 43/72 \\ 43/72 & 29/72 \end{pmatrix}$$

$$P^{(5)} = P^{(4)} \cdot P$$

$$= \begin{pmatrix} 29/72 & 43/72 \\ 43/72 & 29/72 \end{pmatrix} \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix}$$

$$= \begin{pmatrix} 173/432 & 259/432 \\ 259/432 & 173/432 \end{pmatrix}$$

$P(\text{May } 5^{\text{th}} \text{ is dry})$

$$= \frac{173}{432}$$

6) The TPM of a Markov process of $\{x_n\}, n = 1, 2, 3, \dots$
having 3 states is

$$P = \begin{bmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{bmatrix}$$

and

initial distribution $p^{(0)} = [0.7 \ 0.2 \ 0.1]$

find.

$$\text{i) } P[x_2 = 3, x_1 = 3, x_0 = 2]$$

$$\text{ii) } P[x_3 = 2, x_2 = 3, x_1 = 3, x_0 = 2]$$

$$\text{iii) } P[x_2 = 3]$$

Solution

$$\text{i) } P[x_2 = 3, x_1 = 3, x_0 = 2]$$

$$\begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\begin{aligned}
 &= P[X_3 = 3 / X_1 = 3] \cdot P[X_1 = 3 / X_0 = 2] \cdot P[X_0 = 2] \\
 &= \frac{P_{33}^{(1)}}{0.3} \cdot \frac{P_{23}^{(1)}}{0.2} \cdot P[X_0 = 2] \\
 &= 0.3 \times 0.2 \times 0.2 \\
 &= 0.012
 \end{aligned}$$

$$(ii) P[X_3 = 2, X_2 = 3, X_1 = 3 / X_0 = 2]$$

$$\begin{aligned}
 &= P[X_3 = 2 / X_1 = 3] \cdot P[X_2 = 3 / X_1 = 3] \cdot P[X_1 = 3 / X_0 = 2] \cdot P[X_0 = 2] \\
 &= P_{32}^{(1)} \left(\frac{P_{33}^{(1)}}{0.3} \cdot \frac{P_{23}^{(1)}}{0.2} \right) \cdot P[X_0 = 2] \\
 &= 0.4 \cdot 0.3 \cdot 0.2 \times 0.2 = 0.0048
 \end{aligned}$$

$$(iii) P[X_2 = 3]$$

$$\begin{aligned}
 &= \sum_{i=1}^3 P[X_2 = 3 / X_0 = i] \cdot P[X_0 = i] \\
 &= P[X_2 = 3 / X_0 = 1] \cdot P[X_0 = 1] + P[X_2 = 3 / X_0 = 2] \\
 &\quad \left[\begin{array}{ccc} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{array} \right] \cdot P[X_0 = 2] \\
 &\quad + P[X_2 = 3 / X_0 = 3] \cdot P[X_0 = 3].
 \end{aligned}$$

$$= P_{13}^{(2)} = P[X_0 = 1] + P_{23}^{(2)} \cdot P[X_0 = 2]$$

$$\left[\begin{array}{ccc} 0 & 0 & 1 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{array} \right] + P_{33}^{(2)} \cdot P[X_0 = 3]$$

now $P^{(2)} = P \cdot P$

$$[E = eX] \quad (ii)$$

$$\begin{aligned}
 &= \left[\begin{array}{ccc} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{array} \right] \cdot \left[\begin{array}{ccc} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.2 & 0.4 & 0.3 \end{array} \right]
 \end{aligned}$$

$$P^{(2)} = \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} \begin{bmatrix} 0.43 & 0.31 & 0.26 \\ 0.24 & 0.42 & 0.34 \\ 0.36 & 0.35 & 0.29 \end{bmatrix}$$

$$\text{(i) } P[x_2=2 | x_0=2] = 0.26(0.7) + 0.34(0.2) + 0.29(0.1) \\ = 0.279.$$

7) The tpm of a Markov process of $\{x_n\}_{n=1,2,3\dots}$ having 3 states $i=0,1,2$ is $P = \begin{bmatrix} 0 & 3/4 & 1/4 \\ 1 & 1/4 & 1/2 & 1/4 \\ 2 & 0 & 3/4 & 1/4 \end{bmatrix}$
and the initial distribution
 $P^0 = (1/3, 1/3, 1/3)$

Find:

$$\text{i) } P[x_3=2 | x_2=1]$$

$$\text{ii) } P[x_3=1, x_2=2, x_1=1 | x_0=2]$$

$$\text{iii) } P[x_3=2, x_1=1, x_0=2]$$

$$\text{iv) } P[x_2=2]$$

Solution

$$\text{i) } P[x_3=2 | x_2=1]$$

$$P_{21} = 1/4$$

$$\text{ii) } P[x_3=1, x_2=2 | x_1=1, x_0=2]$$

$$= P[x_3=1 | x_2=2] \cdot P[x_2=2 | x_1=1] \cdot P[x_1=1 | x_0=2]$$

$$= P_{21} \cdot P_{12} \cdot P_{12} \cdot P[x_2=2]$$

$$= 3/4 \cdot 1/4 \cdot 3/4 \cdot 1/3 = 1/64$$

$$= \frac{9}{64} \approx \frac{3}{64} //$$

$$\begin{aligned}
 \text{iii)} \quad & P[X_2=2, X_1=1, X_0=2] \\
 & = P[X_2=2/X_1=1] \cdot P[X_1=1/X_0=2] \cdot P[X_0=2] \\
 & \stackrel{(i)}{=} P_{12} \cdot P_{12} \cdot P_{22} \\
 & = \frac{3}{4} \cdot \frac{3}{4} = \frac{1}{3}
 \end{aligned}$$

$\left[\begin{array}{ccc} 3 & 1 & 0 \\ 1 & 3 & 0 \\ 0 & 0 & 2 \end{array}\right] \cdot \left[\begin{array}{ccc} \frac{3}{16} & \frac{1}{16} & 0 \\ \frac{1}{16} & \frac{3}{16} & 0 \\ 0 & 0 & \frac{1}{2} \end{array}\right] = \left[\begin{array}{ccc} \frac{9}{16} & \frac{1}{16} & 0 \\ \frac{1}{16} & \frac{9}{16} & 0 \\ 0 & 0 & \frac{1}{4} \end{array}\right]$

$$\text{iv)} \quad P[X_2=2]$$

$$= \sum_{i=0}^2 P[X_2=2/X_0=i] \cdot P[X_0=i]$$

$$= P[X_2=2/X_0=0] \cdot P[X_0=0] +$$

$$P[X_2=2/X_0=1] \cdot P[X_0=1] +$$

$$P[X_2=2/X_0=2] \cdot P[X_0=2]$$

$$= P_{02}^2 \cdot P[X_0=0] + P_{12} \cdot P[X_0=1] + P_{22}^2 \cdot P[X_0=2]$$

$$\Rightarrow P^2 = P \times P \quad \text{[using (i)]}$$

$$= \begin{bmatrix} \frac{3}{4} & \frac{1}{4} & 0 \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \\ 0 & \frac{3}{4} & \frac{1}{4} \end{bmatrix} \cdot \begin{bmatrix} \frac{9}{16} & \frac{1}{16} & 0 \\ \frac{1}{16} & \frac{9}{16} & 0 \\ 0 & 0 & \frac{1}{4} \end{bmatrix}$$

$$= \begin{bmatrix} 0.625 & 0.125 & 0.0625 \\ 0.125 & 0.5 & 0.1875 \\ 0.1875 & 0.375 & 0.25 \end{bmatrix}$$

$$\begin{aligned}
 \text{(i)} \quad & = 0.625 \times \frac{1}{3} + 0.1875 \times \frac{1}{3} + 0.25 \times \frac{1}{3} \\
 & = 0.1666\overline{6}
 \end{aligned}$$

$$2] \sim P[X_0 = 2]$$

8) A gambler has Rs 2. He bets Rs 1 at a time and wins Rs 1 with prob $\frac{1}{2}$. He stops playing if he loses Rs 2 or wins Rs 4.

i) what is the TPM related to Markov chain

ii) what is the prob that he has lost his

money at the end of 5th play.

iii) what is the prob that the game last more

than 7 plays

Sol states are $\{0, 1, 2, 3, 4, 5\}$

Initial prob dist is $p(0) = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$

TPM

| | 0 | 1 | 2 | 3 | 4 | 5 | 6 |
|---|---------------|---------------|---------------|---------------|---------------|---------------|---|
| 0 | 1 | 0 | 0 | 0 | 0 | 0 | 0 |
| 1 | $\frac{1}{2}$ | 0 | $\frac{1}{2}$ | 0 | 0 | 0 | 0 |
| 2 | 0 | $\frac{1}{2}$ | 0 | $\frac{1}{2}$ | 0 | 0 | 0 |
| 3 | 0 | 0 | $\frac{1}{2}$ | 0 | $\frac{1}{2}$ | 0 | 0 |
| 4 | 0 | 0 | 0 | $\frac{1}{2}$ | 0 | $\frac{1}{2}$ | 0 |
| 5 | 0 | 0 | 0 | 0 | $\frac{1}{2}$ | 0 | 0 |
| 6 | 0 | 0 | 0 | 0 | 0 | 0 | 1 |

ii) $P[\text{that he lost money at the end of 5th play}]$

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

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

$$P^{(2)} = P^{(1)} \cdot P$$

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

$$\begin{aligned}
 p^{(3)} &= P^{(2)} \cdot P \\
 &= \left[\frac{1}{4} \quad \frac{1}{4} \quad 0 \quad \frac{3}{8} \quad 0 \quad \frac{1}{8} \quad 0 \right] \\
 p^{(4)} &= P^{(3)} \cdot P \\
 &= \left[\frac{3}{8} \quad \frac{1}{4} \quad 0 \quad \frac{5}{16} \quad 0 \quad \frac{1}{4} \quad 0 \quad \frac{1}{16} \right] \\
 p^{(5)} &= P^{(4)} \cdot P \\
 &= \left[\frac{3}{8} \quad \frac{5}{32} \quad 0 \quad \frac{9}{32} \quad 0 \quad \frac{1}{8} \quad 0 \quad \frac{1}{16} \right]
 \end{aligned}$$

Q) A
the
V
sol

(PCHe lost his money at the end of 5th play)

$= \frac{9}{16}$

iii) P[game last more than 7 plays]

$$\begin{aligned}
 p^{(6)} &= P^{(5)} \cdot P \\
 &= \left[\frac{29}{64} \quad 0 \quad \frac{7}{64} \quad 0 \quad \frac{13}{64} \quad 0 \quad \frac{1}{8} \right] \\
 &= \left[\frac{29}{64} \quad 0 \quad \frac{7}{32} \quad 0 \quad \frac{3}{64} \quad 0 \quad \frac{1}{8} \right] \\
 p^{(7)} \cdot p^{(6)} \cdot P & \\
 &= \left(\frac{29}{64} \quad \frac{7}{64} \quad 0 \quad \frac{27}{128} \quad 0 \quad \frac{13}{128} \quad \frac{1}{8} \right)
 \end{aligned}$$

P[game last more than
the 7 play]
 $i.e. > 7$

$$= \frac{7}{64} + 0 + \frac{27}{128}$$

$$+ 0 + \frac{13}{128}$$

$$= \frac{54}{128} \Rightarrow \frac{27}{64}$$

q) A die is tossed repeatedly. Let X_n denote the maximum of the numbers occurring in the tosses. i) find also P^2 such that $P[X_2 = 6]$.

Sol states are $\{1, 2, 3, 4, 5, 6\}$.
 P^0 states are $\{1, 2, 3, 4, 5, 6\}$

Individual prob $p^0 = [1/6, 1/6, 1/6, 1/6, 1/6, 1/6]$

| | 1 | 2 | 3 | 4 | 5 | 6 |
|---|-----|-----|-----|-----|-----|-----|
| 1 | 1/6 | 1/6 | 1/6 | 1/6 | 1/6 | 1/6 |
| 2 | 0 | 2/6 | 1/6 | 1/6 | 1/6 | 1/6 |
| 3 | 0 | 0 | 3/6 | 1/6 | 1/6 | 1/6 |
| 4 | 0 | 0 | 0 | 4/6 | 1/6 | 1/6 |
| 5 | 0 | 0 | 0 | 0 | 5/6 | 1/6 |
| 6 | 0 | 0 | 0 | 0 | 0 | 1 |

$$\begin{aligned} \Rightarrow & P(X_2 = 6) \\ &= \sum_{i=1}^6 P(X_2 = 6 | X_0 = i) \cdot P(X_0 = i) \\ \Rightarrow & P(X_2 = 6 | X_0 = 1) \cdot P(X_0 = 1) + \\ & P(X_2 = 6 | X_0 = 2) \cdot P(X_0 = 2) + \dots + \\ & P(X_2 = 6 | X_0 = 6) \cdot P(X_0 = 6) \\ \Rightarrow & P_{16}^2 \cdot P(X_0 = 1) + P_{26}^2 \cdot P(X_0 = 2) + \dots + \\ & P_{66}^2 \cdot P(X_0 = 6) \end{aligned}$$

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

$$= \frac{1}{36} \left[\begin{array}{cccccc} 1 & 3 & 5 & 7 & 9 & 11 \\ 0 & 4 & 5 & 7 & 9 & 11 \\ 0 & 0 & 9 & 7 & 9 & 11 \\ 0 & 0 & 0 & 16 & 9 & 11 \\ 0 & 0 & 0 & 0 & 25 & 11 \\ 0 & 0 & 0 & 0 & 0 & 36 \end{array} \right]$$

$$\begin{aligned} &= \frac{1}{36} \cdot \frac{1}{6} + \frac{1}{36} \cdot \frac{1}{6} + \frac{1}{36} \cdot \frac{1}{6} + \frac{1}{36} \cdot \frac{1}{6} \\ &\quad + \frac{1}{36} \cdot \frac{1}{6} + \frac{1}{36} \cdot \frac{1}{6} \end{aligned}$$

$$P[X_2 = 6] = \frac{9}{216}$$

Classification:

i) Irreducible

$p_{ij}^{(m)} > 0$, & i, j and for some m.

ii) Non-pnull persistent should be b
educible and finite

iii) Aperiodic (or) periodic $p_{ij}^{(m)} > 0$.

$$d_i = \inf \{ m / p_{ij}^{(m)} > 0 \}$$

$$= \inf \{ 2, 4 \}$$

$d_i > 1 \rightarrow$ periodic

$d_i = 1 \rightarrow$ Aperiodic

iv) Ergodic:

2nd type and (i) Aperiodic

10) find the nature of state of markov chain with TPM given

$$P = \begin{bmatrix} 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \\ 1 & 1 & 0 \end{bmatrix} \text{ and the}$$

states are $\{0, 1, 2\}$

$$P^2 = \begin{bmatrix} 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \\ 1/2 & 0 & -1/2 \end{bmatrix} \quad P^3 = \begin{bmatrix} 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\text{and } P^4 = \begin{bmatrix} 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \\ 1/2 & 0 & -1/2 \end{bmatrix} \quad P^5 = \begin{bmatrix} 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \end{bmatrix}$$

$$P^2 = P^{2n}, n=1, 2, 3.$$

$P = P^{2n+1}$ would be periodic
but P is not a periodic matrix

i)

State p_{ij} belongs to NTP
 $p_{00}^2 > 0, p_{01}^1 > 0, p_{11}^1 > 0,$
 $p_{10}^1 > 0, p_{11}^2 > 0, p_{12}^1 > 0,$
 $p_{20}^2 > 0, p_{21}^1 > 0, p_{22}^2 > 0,$

\therefore It is irreducible. Since states are $\{0, 1, 2\}$

ii) $p_{00}^2 > 0 \Rightarrow p_{00}^4 > 0 \Rightarrow p_{00}^6 > 0.$

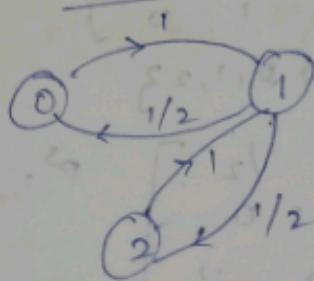
\therefore for all i , states of chain are periodic with 2. $\therefore d_i > 1$.

State 1: $\Rightarrow p_{11}^{(2)} > 0 = p_{11}^{(4)} > 0 = p_{11}^{(6)} > 0$.
 $d_1 = 2$.

State 2: $p_{22}^{(2n)} > 0, n=1, 2 \Rightarrow$ periodic

w) Non null persistent but it's not Aperiodic
as it's not ergodic

Transition diagram



ii) Three boys A, B, C are throwing a ball
to each other. A always throws to B,

B always throws to C. But C is just

as likely to throw ball to A or B.

Show that process is not ergodic. find

TPM and classify state.

| | A | B | C |
|---|-----|-----|---|
| A | 0 | 1 | 0 |
| B | 0 | 0 | 1 |
| C | 1/2 | 1/2 | 0 |

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

$$P^4 = \begin{bmatrix} 0 & 1/2 & 1/2 \\ 1/4 & 1/4 & 1/2 \\ 1/4 & 1/2 & 1/4 \end{bmatrix}, P^5 = \begin{bmatrix} 1/2 & 1/2 & 1/2 \\ 1/4 & 1/2 & 1/4 \\ 1/8 & 3/8 & 1/2 \end{bmatrix}$$

so - m

$$\begin{array}{ccc} 5 & 5 & 5 \\ \hline P_{AA} > 0 & P_{AB} > 0 & P_{AC} > 0 \\ P_{BA} > 0 & P_{BB} > 0 & P_{BC} > 0 \\ P_{CA} > 0 & P_{CB} > 0 & P_{CC} > 0 \end{array}$$

Irreducible.

i) Irreducible, states $\{A, B, C\}$
 \therefore non-null persistent.

iii) State A:

$$P_{AA}^a > 0, P_{AA}^b > 0, P_{AA}^c > 0, P_{AA}^T > 0.$$

$$d_i = \text{GCD } \{ M | P_{ij}^{(m)} > 0 \}$$

$\leq 1 \Leftrightarrow A$ periodic.

State B: $d_i = 1$ A-periodic

State C: $d_i = 1$ A-periodic

ω : Non-null ω A-periodic
gt is ergodic

