



Markov Process:

It is the one in which the future value is independent of the past value given the present value.

Markovian:

A random process $x(t)$ is said to be Markovian if

$$P[x(t_{n+1}) \leq x_{n+1} / x(t_n) = x_n, x(t_{n-1}) = x_{n-1}, \dots, x(t_0) = x_0] \\ = P[x(t_{n+1}) \leq x_{n+1} / x(t_n) = x_n]$$

where $t_0 \leq t_1 \leq t_2 \leq \dots \leq t_n \leq t_{n+1}$ and

$x_0, x_1, \dots, x_n, x_{n+1}$ are called the states of the process.

Eg:

The probability of raining today depends on the previous weather conditions existed for the last two days and not on the past weather condition.

Markov chain:

If $P[x_n = a_n / x_{n-1} = a_{n-1}, x_{n-2} = a_{n-2}, \dots, x_0 = a_0] \\ = P[x_n = a_n / x_{n-1} = a_{n-1}]$, then the process $x_n, n=0, 1, 2, \dots$ is called a Markov chain.

One-step Transition Probability:

The conditional probability $P[x_n = a_j / x_{n-1} = a_i]$ is called the one-step transition probability from state a_i to state a_j at the n^{th} step and it is denoted by $p_{ij}(n-1, n)$



Homogeneous Markov chain:

If the one step transition probability does not depend on the step i.e., $P_{ij}(n-1, n) = P_{ij}(m-1, m)$, then the Markov chain is called the homogeneous Markov chain.

Transition Probability matrix [TPM]

When the Markov chain is homogeneous, the one step transition probability is denoted by P_{ij} . The matrix P_{ij} satisfies the following conditions

i). $P_{ij} \geq 0$

ii). $\sum P_{ij} = 1, \forall i$

Result:

i]. $P(x_i = a / x_j = b) = P_{ba}^{i-j}$

ii]. $P(x_n = j) = \sum_{i=0}^j P(x_n = j / x_0 = i) \cdot P(x_0 = i)$

Eg: $P(x_2 = 3 / x_0 = 1) = P_{13}^{2-0} = P_{13}^{(2)}$



7. The TPM of a Markov chain with three states 0, 1, 2 is

$$P = \begin{pmatrix} \frac{3}{4} & \frac{1}{4} & 0 \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \\ 0 & \frac{3}{4} & \frac{1}{4} \end{pmatrix} \text{ and the initial state}$$

distribution of the chain is $P(X_0 = i) = \frac{1}{3}, i = 0, 1, 2.$

- Find
- $P(X_2 = 2)$
 - $P(X_3 = 1, X_2 = 2, X_1 = 1, X_0 = 2)$
 - $P(X_2 = 1 / X_0 = 0)$

Soln.

$$\text{Let } P(X_0 = 0) = \frac{1}{3}; \quad P(X_0 = 1) = \frac{1}{3}; \quad P(X_0 = 2) = \frac{1}{3}$$

i) $P(X_2 = 2)$

$$\text{Now, } P(X_n = j) = \sum_{i=0}^2 P(X_n = j / X_0 = i) \cdot P(X_0 = i)$$

$$P(X_2 = 2) = \sum_{i=0}^2 P(X_2 = 2 / X_0 = i) P(X_0 = i)$$

$$= P(X_2 = 2 / X_0 = 0) P(X_0 = 0) + P(X_2 = 2 / X_0 = 1) P(X_0 = 1)$$

$$+ P(X_2 = 2 / X_0 = 2) P(X_0 = 2)$$

$$= P_{02}^{(2-0)} P_0^{(0)} + P_{12}^{(2-0)} P_1^{(0)} + P_{22}^{(2-0)} P_2^{(0)}$$

$$= P_{02}^{(2)} P_0^{(0)} + P_{12}^{(2)} P_1^{(0)} + P_{22}^{(2)} P_2^{(0)} \rightarrow (1)$$

Given

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



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

$$= \begin{pmatrix} 0 & 0.625 & 0.3125 & 0.0625 \\ 0.3125 & 0.5 & 0.1875 & \\ 0.1875 & 0.5625 & 0.25 & \end{pmatrix}$$

$$(i) \Rightarrow P(X_2 = 2) = 0.0625 \left(\frac{1}{3}\right) + 0.1875 \left(\frac{1}{3}\right) + 0.25 \left(\frac{1}{3}\right)$$

$$= 0.1639$$

$$ii). P(X_3 = 1, X_2 = 2, X_1 = 1, X_0 = 2)$$

$$= P(X_3 = 1 / X_2 = 2, X_1 = 1, X_0 = 2) P(X_2 = 2, X_1 = 1, X_0 = 2)$$

$$= P(X_3 = 1 / X_2 = 2) P(X_2 = 2 / X_1 = 1, X_0 = 2) P(X_1 = 1, X_0 = 2)$$

$$= P(X_3 = 1 / X_2 = 2) P(X_2 = 2 / X_1 = 1) P(X_1 = 1 / X_0 = 2)$$

$$P(X_0 = 2)$$

$$= P_{21}^{(1)} P_{12}^{(1)} P_{21}^{(1)} P_2^{(0)}$$

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

$$= 0.046$$

$$iii). P(X_2 = 1 / X_0 = 0) = P_{01}^{(2)} = 0.31$$



Classify the states :

Irreducible:

If for every i, j and for some n such that $P_{ij}^n > 0$, then every state can be reached from every other state, then the Markov chain is irreducible.

Periodic State:

Let $P_{ii}^{(m)} > 0$ for all m . Let i be a return state.

Then $d_i = \text{GCD} \{ m : P_{ii}^{(m)} > 0 \}$

where GCD stands for the greatest common divisor.

\Rightarrow If $d_i > 1$, then the state ' i ' is called periodic

\Rightarrow If $d_i = 1$, then the state ' i ' is called aperiodic.

Non-Null Persistent:

If a Markov chain is finite and irreducible then all the states are non-null persistent.

Ergodic:

A non-null persistent & aperiodic state is said to be ergodic.

Non Ergodic:

A non-null persistent & periodic state is said to be non-ergodic.

11. Three boys A, B, C are throwing a ball to each other. A always throw a ball to B, B always throw a ball to C, but C is just as likely to throw the ball to B as to A. Find TPM & classify the states, ^{draw} diagram.



Soln.

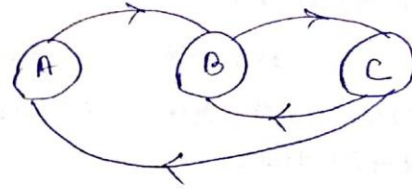
$$P = \begin{matrix} & \begin{matrix} A & B & C \end{matrix} \\ \begin{matrix} A \\ B \\ C \end{matrix} & \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ \frac{1}{2} & \frac{1}{2} & 0 \end{bmatrix} \end{matrix}$$

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

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

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

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



$$\therefore d_i = \text{G.C.D.} \{ m, P_{ii}^{(m)} > 0 \}$$

$$d_1 = \text{G.C.D.} \{ 3, 5 \} = 1$$

$$d_2 = \text{G.C.D.} \{ 2, 3, 4, 5 \} = 1$$

$$d_3 = \text{G.C.D.} \{ 2, 3, 4, 5 \} = 1$$

$$\therefore d_i = 1 \Rightarrow \text{aperiodic}$$

$$\therefore P_{11}^{(3)} > 0$$

$$P_{11}^{(5)} > 0$$

Now,

$$P_{11}^{(3)} > 0 \quad P_{12}^{(1)} > 0 \quad P_{13}^{(2)} > 0$$

$$P_{21}^{(2)} > 0 \quad P_{22}^{(2)} > 0 \quad P_{23}^{(1)} > 0$$

$$P_{31}^{(1)} > 0 \quad P_{32}^{(1)} > 0 \quad P_{33}^{(2)} > 0$$

\therefore The chain is irreducible.



Since we have 3 states, the chain is finite and irreducible.

\therefore All the states are non-null persistent.
Since all the states are aperiodic & non-null persistent,

\therefore It is ergodic.

eg. Let $P = \begin{bmatrix} 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \end{bmatrix}$, classify the states of the Markov chain.

Soln.

Given $P = \begin{bmatrix} 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \end{bmatrix}$

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

$\therefore P_{11}^{(2)} > 0, P_{22}^{(2)} > 0$ and $P_{33}^{(2)} > 0$

It is irreducible.

Since we've 3 states, the chain is finite and irreducible.
All the states are non-null persistent.

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

$$P^6 = \begin{bmatrix} 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \\ 0 & 1 & 0 \end{bmatrix}$$

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

$d_1 = \text{GCD} \{ 2, 4 \} = 2$

$d_2 = \text{GCD} \{ 2, 4 \} = 2$

$d_3 = \text{GCD} \{ 2, 4 \} = 2$

It is $\therefore d_i = 2 \Rightarrow$ Periodic



\therefore It is periodic & non-null persistent.

\therefore It is non-ergodic

Steady state distribution:

If P is TPM of Markov chain $\pi = \pi_1, \pi_2, \dots, \pi_n$
Steady state distribution is i). $\pi P = \pi$

$$\text{ii). } \sum_{i=1}^n \pi_i = 1$$

1. A house wife buys three kinds of cereals A, B, C. She never buys the same cereal in successive weeks. If she buys cereal A, the next week she buys cereal B. However if she buys either B or C, the next week she is three times as likely to buy A as the other cereal. How often she buys each of the 3 cereals?

Soln.

TPM:

$$P = \begin{matrix} & \begin{matrix} A & B & C \end{matrix} \\ \begin{matrix} A \\ B \\ C \end{matrix} & \begin{bmatrix} 0 & 1 & 0 \\ 3/4 & 0 & 1/4 \\ 3/4 & 1/4 & 0 \end{bmatrix} \end{matrix}$$

Let $\pi = (\pi_1, \pi_2, \pi_3)$

i). $\pi P = \pi$

$$(\pi_1, \pi_2, \pi_3) \begin{pmatrix} 0 & 1 & 0 \\ 3/4 & 0 & 1/4 \\ 3/4 & 1/4 & 0 \end{pmatrix} = (\pi_1, \pi_2, \pi_3)$$

$$\left(0 + \frac{3\pi_2}{4} + \frac{3\pi_3}{4} \quad \pi_1 + 0 + \frac{\pi_3}{4} \quad 0 + \frac{\pi_2}{4} + 0 \right) = (\pi_1, \pi_2, \pi_3)$$

$$\Rightarrow \frac{3\pi_2 + 3\pi_3}{4} = \pi_1 ; \quad \pi_1 + \frac{\pi_3}{4} = \pi_2 ; \quad \frac{\pi_2}{4} = \pi_3$$

$$4\pi_1 = 3\pi_2 + 3\pi_3 ; \quad 4\pi_1 + \pi_3 = 4\pi_2 ; \quad \pi_2 = 4\pi_3$$

$\hookrightarrow (1) \qquad \qquad \qquad \hookrightarrow (2) \qquad \qquad \qquad \hookrightarrow (3)$



Subst. (3) in (2),

$$4\pi_1 + \pi_3 - 4(4\pi_3) = 0$$

$$4\pi_1 + \pi_3 - 16\pi_3 = 0$$

$$4\pi_1 - 15\pi_3 = 0$$

$$\pi_1 = \frac{15}{4}\pi_3$$

ii). $\pi_1 + \pi_2 + \pi_3 = 1$

$$\frac{15}{4}\pi_3 + 4\pi_3 + \pi_3 = 1$$

$$\frac{15\pi_3 + 16\pi_3 + 4\pi_3}{4} = 1$$

$$\frac{35\pi_3}{4} = 1$$

$$\pi_3 = \frac{4}{35}$$

$$(3) \Rightarrow \pi_2 = 4\pi_3 = 4\left(\frac{4}{35}\right) = \frac{16}{35}; \quad \pi_1 = \frac{15}{4}\left(\frac{4}{35}\right) = \frac{15}{35}$$

$$\therefore \pi = \left(\frac{15}{35} \quad \frac{16}{35} \quad \frac{4}{35} \right)$$

2]. A man either drives a car or catches a train to his office each day. He never goes two days in row by train, but if he drives one day, then next day he is just as likely to drive again as he is to travel by train. Now suppose that on the first day drive to work if and only if a '6' appears. Find

- i). the probability that he takes a train on the 3rd day
- ii). the probability that he drives to work in the long run?

Scanned with CamScanner



Soln.

Let T be train and c be car.

Let (T, c) be a travel pattern.

TPM:

$$P = \begin{matrix} & T & c \\ \begin{matrix} T \\ c \end{matrix} & \begin{pmatrix} 0 & 1 \\ 1/2 & 1/2 \end{pmatrix} \end{matrix}$$

prob. of travelling by car = P[getting c in the 1st day] = $\frac{1}{6}$

prob. of travelling by train = $1 - \frac{1}{6} = \frac{5}{6}$

$$\therefore P^{(1)} = \begin{pmatrix} \frac{5}{6} & \frac{1}{6} \end{pmatrix}$$

$$P^{(2)} = P^{(1)} \cdot P = \begin{pmatrix} \frac{5}{6} & \frac{1}{6} \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1/2 & 1/2 \end{pmatrix} = \begin{pmatrix} \frac{1}{12} & \frac{11}{12} \end{pmatrix}$$

$$P^{(3)} = P^{(2)} \cdot P = \begin{pmatrix} \frac{1}{12} & \frac{11}{12} \end{pmatrix} \begin{pmatrix} 0 & 1 \\ 1/2 & 1/2 \end{pmatrix} = \begin{pmatrix} \frac{11}{24} & \frac{13}{24} \end{pmatrix}$$

i). P[the man travels by train on the 3rd day]
= $\frac{11}{24}$

ii). Steady state distribution: $\pi = (\pi_1, \pi_2)$

i). $\pi P = \pi$

$$(\pi_1, \pi_2) \begin{pmatrix} 0 & 1 \\ 1/2 & 1/2 \end{pmatrix} = (\pi_1, \pi_2)$$

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

$$\Rightarrow \frac{\pi_2}{2} = \pi_1 \quad \left\{ \begin{array}{l} \pi_1 + \frac{\pi_2}{2} = \pi_2 \\ \Rightarrow 2\pi_1 + \pi_2 = 2\pi_2 \\ \Rightarrow 2\pi_1 = \pi_2 \end{array} \right.$$

ii). $\pi_1 + \pi_2 = 1 \Rightarrow \pi_1 + 2\pi_1 = 1 \Rightarrow 3\pi_1 = 1 \Rightarrow \pi_1 = \frac{1}{3}$

$$\therefore \pi = \left(\frac{1}{3}, \frac{2}{3} \right)$$

$$\pi_2 = \frac{2}{3}$$



$$P[\text{the man travels by car in the long run}] = \frac{2}{3}$$