

Stochastic process & Markov chains

stochastic (or) random process:

* A stochastic or random process is defined as a family of random variables. $\{x(t_n); n=1, 2, 3, \dots\}$.

* The Random variable $x(t)$ stands for the observation at time 't'. The No. of states 'n' may be finite or infinite depending on time range.

* for example, Let us consider the poisson distribution, $P_n(t) = \frac{e^{-\lambda t} (\lambda t)^n}{n!}$ where $n=1, 2, 3, \dots$

This distribution represent a stochastic process with infinite No. of state. In this example, the Rv 'n' denotes the No of occurrences between the interval '0' and 't'. Thus, the states of the system at any time is given by $n=0, 1, 2, 3, \dots$

* Markov process :-

Markov process is a stochastic process which has the property that the probability of transition from a given state to from any future states depends only on the present state and not on the manner in which it was reached.

for example, suppose a car rental company is running agencies in different cities. If car sent to one city may return to any city where the companies agency is available.

* If this situation is considered as the Markov process the different rental cities would be in the

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the probability that a car rented to city ' i ' would be returned to city ' j ', where j may be equal ' i '. The mathematical structure of this problem is to determine the expected long fraction of cars at each city and mean no of steps a car would make starting from city ' i ' before returning the location.

Definition :-

The stochastic or random process is called Markov process if the occurrence of the future state depends on the immediately preceding state and only on it. If $t_1 < t_2 < \dots < t_n$ represents the points in time scale then the family of RV $\{x(t_n)\}$ is said to be Markov process provides it holds the Markovian property

$$P \{ x(t_n) = x_n \mid x(t_{n-1}) = x_{n-1}, \dots, x(t_0) = x_0 \} \\ = P \{ x(t_n) = x_n \mid x(t_{n-1}) = x_{n-1} \}$$

for all $x(t_0), x(t_1), x(t_2), \dots$. A Markov process is a sequence of ' n ' experiments in which each experiment ' n ' possible outcomes x_1, x_2, \dots, x_n for individual outcome is called a state and the probability depends only on the probability of the outcome of the preceding experiment.

Characteristic of Markov process?

Markov analysis is based on the following characteristics:

- The states are both collectively exhaustive and

old be of them absorbing. If a customer would never switch to a particular brand.

The transition probabilities are stationary.

4. The probability of moving from one state to another depends only for the immediately preceding state.

5. The transition probabilities of moving to alternative state in the next time period. Given a state in the current time period must sum to unity.

6. The process has set of initial probabilities that may be either given or determined.

Uses :-

Markov process is widely used;

1. In examining and predicting the behaviour of consumers, in terms of their brand loyalty and their switching patterns to another brands.

2. In the study of equipment maintenance and failure problems.

3. In analyzing accounts receivable that will ultimately become bad debts.

4. To study the stock market price movements

and the Transition probability :- (2M) Many

* The probability of moving from one state to another or remaining in the same state during a single time period is called the transition probability.

* Mathematically the probability

$$P_{x_n, x_{n-1}} = P(x(t_n) = x_n | x(t_{n-1}) = x_{n-1})$$

The conditional probability of the system which is now in the state ' x_n ' at time ' t_n ' provided that it was previously in state ' x_{n-1} ' at time ' t_{n-1} ' is.

* Sometimes this probability is known as one-step transition probability, because it describes the system during the time interval (t_{n-1}, t_n) .

Transition probability matrix (def) $2^M \times 2^M$

The transition probabilities can be arranged in a matrix from such a matrix is called one-step transition probability matrix denoted by ' P '.

$$\text{is defined as } P = \begin{bmatrix} p_{11} & p_{12} & p_{13} & \dots & p_{1m} \\ p_{21} & p_{22} & p_{23} & \dots & p_{2m} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ p_{m1} & p_{m2} & p_{m3} & \dots & p_{mm} \end{bmatrix}$$

The matrix ' P ' is a square matrix whose each element p_{ij} is non-negative and sum of elements of each row is unity i.e.; $\sum_{j=1}^m p_{ij} = 1$.

$$\text{for } i = 1, 2, \dots, m$$

which construction of transition matrix = the following
provided procedure is Adapted for constructing a state
time transition matrix -

step-1 :- find first compute retention probability
which can be obtained by dividing the no. of customers
retained by the no. of customers originally served

Number retained = $\frac{\text{Original number served} - \text{Number lost}}{\text{Original number served}}$

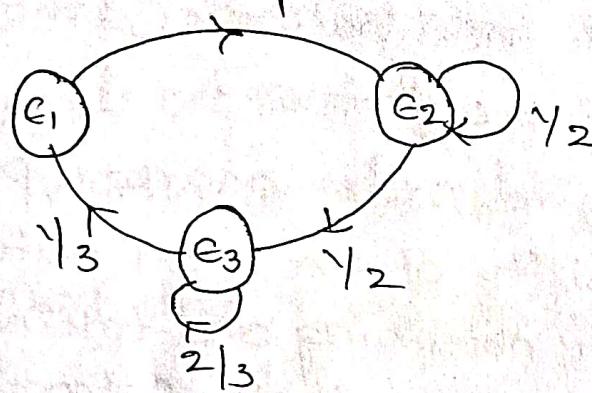
step 2 :- calculate the probabilities associated with
customers gains or losses, for customer gain with probabilities, divide the number of
customers each entry has gained by the number of
customers served by the source of the gain.

for customer loss probabilities divide the number of
customers in each entity has lost by the original no.
of customers of it served

Step 3 :- results of Step-1 and Step-2 are used
to construct the state transition matrix. The retention
probabilities are listed along the main diagonal
The remaining entries are obtained as follows -
if i. loss probabilities are used, insert them in the
appropriate row cells of the state transition matrix.
ii. Gain probabilities are used, insert them in the
appropriate column cells of the state transition matrix.
In both the cases, state transition matrices thus
obtained will be the same.

Diagrammatic representation of transition probabilities
 The transition probabilities can be represented by two types of diagrams i.e., transition diagram (Venn diagram), transition diagram shows the transition probabilities or shifts that can occur in any particular situation.

Eg:-



The arrows each state indicates the possible states to which a process can move from given state.

The matrix of transition probabilities which corresponds to the above diagram is:

$$E_1 \quad E_2 \quad E_3$$

$$P = \begin{matrix} E_1 & \begin{bmatrix} 0 & 1 & 0 \end{bmatrix} \\ E_2 & \begin{bmatrix} 0 & \gamma_2 & 1/2 \end{bmatrix} \\ E_3 & \begin{bmatrix} 2/3 & 0 & 1/3 \end{bmatrix} \end{matrix}$$

ii. Probability tree diagram :- As the name implies, this diagram emphasizes the probabilities and their movement from 1 step to another, along with all possible branches (or) paths that may connect the outcomes over a period of time.

first and higher order Markov process :- [2m]
 - 1st order Markov process is based on the following

probabilities. The set of possible outcomes is finite.
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 e transition probability parti-

- ii. The probability of the next outcome (state) depends only on the immediately preceding outcome.
- iii. The transition probabilities are constant over time.

The second order Markov process assumes that the probability of the next outcome (state) may depend on the two previous outcomes.

Similarly, a third order markov process assumes that the probability of the next outcome (state) can be calculated by obtaining and taking account of the outcomes of the past three outcomes.

n-step transition probabilities :-

Suppose the system which occupies state E_i at time $t=0$, then we may be interested in finding out the probability that the system moves to state E_j at time $t=n$.

These time periods are sometimes referred to as number of steps

If the n -steps transition probability is denoted by $P_{ij}^{(n)}$ then these transition probabilities can be represented in matrix form as given below.

$$P^{(n)} = \begin{matrix} & E_1 & E_2 & \dots & E_n \\ \begin{matrix} E_1 \\ E_2 \\ \vdots \\ E_m \end{matrix} & \left[\begin{matrix} P_{11}^{(n)} & P_{12}^{(n)} & \dots & P_{1m}^{(n)} \\ P_{21}^{(n)} & P_{22}^{(n)} & \dots & P_{2m}^{(n)} \\ \vdots & \vdots & \ddots & \vdots \\ P_{n1}^{(n)} & P_{n2}^{(n)} & \dots & P_{nm}^{(n)} \end{matrix} \right] \end{matrix}$$

which occupies state e_2 will move to state e_1 after n -steps.

Let $P_{ij}^{(n)}$ be the probability that the system occupies state e_i will move to state e_j in one transition. It should be noted that the transition probability p_{ij} is independent of time whereas the absolute probability $p_i(n)$ depends on time.

If the number of possible states be m , then

$$\sum_{i=1}^m p_i(n) = 1 \text{ and } \sum_{j=1}^m p_{ij}^{(n)} = 1, \text{ for all } i.$$

If all the state probabilities are known at time $t=n$, then the state probabilities at time $t=n+1$ can be determined by the equation,

$$p_j(n+1) = \sum_{i=1}^m p_i(n) P_{ij}^{(n)}, n = 0, 1, 2, \dots$$

Rewriting the equations for each states be ∞ , time $t=n+1$:

$$p_1(n+1) = p_1(n)p_{11} + p_2(n)p_{21} + \dots + p_m(n)p_{m1}$$

$$p_2(n+1) = p_1(n)p_{12} + p_2(n)p_{22} + \dots + p_m(n)p_{m2}$$

$$\vdots$$

$$p_n(n+1) = p_1(n)p_{1n} + p_2(n)p_{2n} + \dots + p_m(n)p_{mn}$$

This system of equations can be written in matrix form as $R(n+1) = R(n)P$

$$\begin{aligned} \text{① } (0, 1)^T R(n+1) &= \pi(n)P \\ a_{n+1} &= a_n P \end{aligned}$$

where $R(n+1)$ is the row vector of state probabilities at time $t=n+1$, $R(n)$ is the row vector of state probabilities at time $t=n$ and P is the transition probability at time $t=n$.

Note :- If the state probabilities at time $t=0$ are known, then we can find the state probabilities at any time by solving the matrix equation ①.

Markov chain :

Let $p_{ij}^{(0)}$ ($j=0, 1, 2, \dots$) be the absolute probability such that the system be in state E_i at time t_0 , where E_j ($j=0, 1, 2, \dots$) denote the exhaustive and mutually exclusive outcomes of a system at any time.

Also it is assumed that the system is Markovian

we now define $p_{ij} = P\{x(t_n) = j | x(t_{n-1}) = i\}$ as the one-step transition probability of going from state i to at time t_{n-1} to state j at time t_n .

It is also assumed here that these probabilities from state E_i to state E_j ($i=0, 1, 2, \dots; j=0, 1, 2, \dots$) are expressed in the matrix form as below.

$$P = \begin{bmatrix} p_{00} & p_{01} & p_{02} & \dots \\ p_{10} & p_{11} & p_{12} & \dots \\ p_{20} & p_{21} & p_{22} & \dots \\ \dots & \dots & \dots & \dots \end{bmatrix}$$

This matrix P is known as stochastic matrix or homogeneous matrix.

The probabilities p_{ij} must satisfy the boundary conditions $\sum p_{ij} = 1$, for all i , $p_{ij} \geq 0$, for all i and j

Definition : Markov chain :

The transition matrix P as defined above together with the initial probabilities $\{p_j^{(0)}\}$ associated with the states E_j ($j=0, 1, 2, \dots$) completely define a Markov chain.

- i. ergodic Markov chain ii. Regular Markov chain
- An Ergodic Markov chain is defined as a chain having the property that it is possible to pass from one state to another in a finite number of steps, regardless of present state.
- A regular Markov chain is defined as a chain having a transition matrix P such that for some power of P it has only non-zero positive probability values.
- Note
- i. A special type of ergodic markov chain is the regular markov chain.
 - ii. All the regular chains must be ergodic chains.
 - iii. To check if ergodic chain is regular, continue at squaring the transition matrix P until all zeros are removed.

Prblms. Markom

1. Two manufacturers A and B are competing with each other in a restricted market. Over the years, A's' customers have exhibited a high degree of loyalty as measured by the fact that customers using A's' product from 80% of time. Also former customers purchasing the product from B have switched back to A's' 60% of time.
- a. Construct and interpret the state transition matrix in terms of i. retention and loss ii. retention and gain
- b. calculate the probability of customer purchasing A's product at the end of the second period?
- Sol :-
- a. The transition probabilities can be arranged in a

chain present $P = \text{purchase}(n=0)$

A	0.80	0.20
B	0.60	0.40

↑ retention &
gain
→ retention losses

Clearly the probability of customers purchase at the next step ($n=1$) depends upon the product which a customer is having at present ($n=0$).

∴ Each probability in the above matrix must be conditional probability for passing from one state to another.

The conditional probabilities in the above matrix can be stated as:

$$\text{i. } P(B_0|A_1) = P_{21} = 0.60$$

This indicates that the customer now using B's product at $n=0$ (present purchase) will purchase A's product at $n=1$ (next purchase) is 0.60

This means loss to B's product.

$$\text{i. } P(A_0|A_1) = P_{11} = 0.80$$

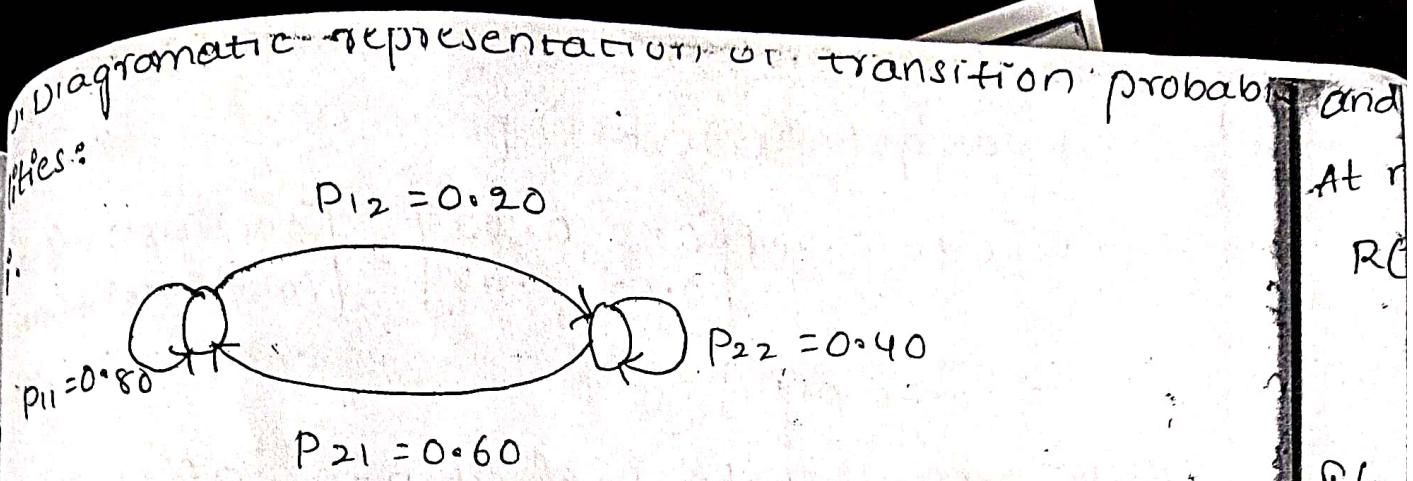
This indicates that the probability that the customer now using A's product at $n=0$ (present purchase) will again purchase A's product at $n=1$ (next purchase) is 0.80. This means retention to A's product.

$$\text{ii. } P(A_0|B_1) = P_{12} = 0.20$$

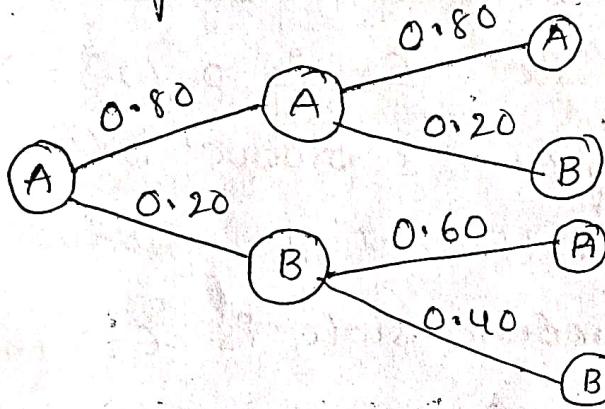
This indicates that the probability that the customer now using A's product at $n=0$ will purchase B's product at $n=1$ is 0.20. This means loss to A's product.

$$\text{iii. } P(B_0|B_1) = P_{22} = 0.40$$

This indicates that the probability that the customer now using B's product at $n=0$ will purchase B's product at $n=1$ is 0.40. This means retention to B's product.

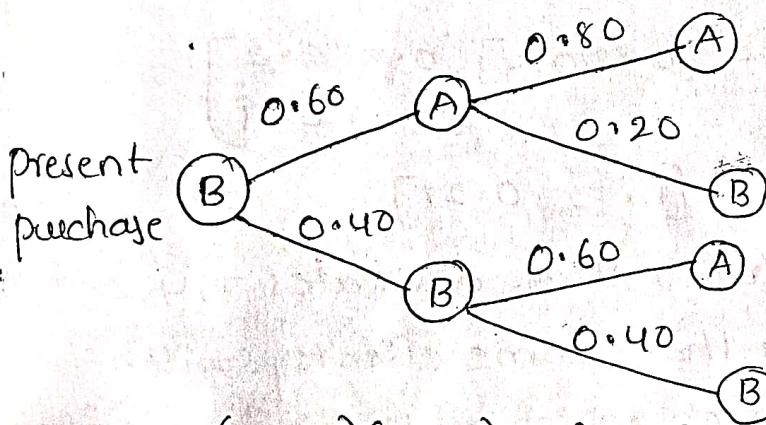


ii. Tree diagram



$$P_{11} = (0.80)(0.80) + (0.20)(0.60) = 0.76$$

$$P_{12} = (0.80)(0.20) + (0.20)(0.40) = 0.24$$



$$P_{22} = (0.60)(0.20) + (0.40)(0.40) = 0.28$$

$$P_{21} = (0.60)(0.80) + (0.40)(0.60) = 0.42$$

Calculations of probability:

We have that $R(n+1) = R(n) P$

At $n=0$, we get $R(1) = R(0) P$

$$\Rightarrow R(1) = [1 \ 0] \begin{bmatrix} 0.80 & 0.20 \\ 0.60 & 0.40 \end{bmatrix} = [0.80 \ 0.20]$$

$$\begin{cases} P_1(0) = P_2(0) = 1 \\ \Rightarrow R(0) = [1 \ 0] \end{cases}$$

probability and the probability at the next state E_2 is $P_2(1) = 0.24$.
 At $n=1$, we get $R(2) = R(1) \cdot P$

$$R(2) = [0.80 \ 0.20] \begin{bmatrix} 0.80 & 0.20 \\ 0.60 & 0.40 \end{bmatrix} = \begin{bmatrix} (0.80)(0.80) + (0.20)(0.60) \\ (0.80)(0.20) + (0.20)(0.40) \end{bmatrix} \\ = [0.76 \ 0.24]$$

If the present state is E_1 at $n=0$, after 2-steps, the probability of being in the state E_1 is $P_1(2) = 0.76$ and in the state E_2 is $P_2(2) = 0.24$.

i.e., The probability of A's product after the end of 2-steps is 76% and that of B's product is 24%.

Similarly if the present state is E_2 then.

$$\text{where } R(1) = [P_1(1) \ P_2(1)]$$

$$R(1) = [0.1] \begin{bmatrix} 0.8 & 0.2 \\ 0.6 & 0.4 \end{bmatrix} = [0.6 \ 0.4]$$

$$\text{and } R(2) = R(1) P = [0.6 \ 0.4] \begin{bmatrix} 0.8 & 0.2 \\ 0.6 & 0.4 \end{bmatrix} \\ = [0.72 \ 0.28]$$

2. In a small town with three advocates 1, 4 and 2 each advocate known that some clients switch back and forth, depending on which advocate is available at the time the client needs one.

There are no new clients in the current legal market, however none of the old clients are leaving the area.

$P_2(0) = 0$
 $= [1 \ 0]$ During a slack period, the three advocates collected data which identical the number of clients each advocate had seen during the preceding year.

Tables given below summarize the results of this study & the numbers in which clients were moved as new connections.

Data summary - Client Summary

Advocate	clients as of Jan 1, 2005	change during the year		clients as of Jan 1, 2006
		Gain	Loss	
X	400	75	50	425
Y	500	50	150	400
Z	500	100	25	573

Gain - loss summary

Advocate	clients as of Jan 1, 2005	Gains			Losses			clients as of Jan 1, 2006
		from X	Y	Z	To X	Y	Z	
X	400	0	50	25	0	50	0	425
Y	500	50	0	0	50	0	100	400
Z	500	0	100	0	25	0	0	573

Sol :- Calculation of probabilities :-

i) Retention probability :-

Retention probabilities are calculated by dividing the number of customers that were retained by the number of clients that were originally served.

No. of clients retained = The original number of clients minus the number of customers that were lost.

Retention probabilities

Advocate	clients as of Jan 1, 2005	Number lost	Number retained	probability of retention
X	400	50	$\frac{400-50}{400} = 350$	$350/400 = 0.875$
Y	500	150	$\frac{500-150}{500} = 350$	$350/500 = 0.70$

iii. Gain & loss probabilities :

as of
, 2006

The probabilities associated with client gain (loss) are calculated by dividing the number of clients gained (losses) by the number of clients served at the source of gain(loss)

Gain probability.

original number served	Advocates	probability of gain		
		from X	from Y	from Z
400	X	$P_{xx} = 0/400 = 0$	$P_{xy} = 50/500 = 0.1$	$P_{xz} = 25/500 = 0.05$
500	Y	$P_{xy} = 50/400 = 0.125$	$P_{yy} = 0/500 = 0$	$P_{yz} = 0/500 = 0$
500	Z	$P_{xz} = 0/400 = 0$	$P_{yz} = 100/500 = 0.2$	$P_{zz} = 0/500 = 0$
400				
575				

loss probability

original number served	Advocates	probability of loss		
		TO X	TO Y	TO Z
400	X	$0/400 = 0$	$50/400 = 0.125$	$0/400 = 0$
500	Y	$50/500 = 0.1$	$0/500 = 0$	$100/500 = 0.2$
500	Z	$25/500 = 0.05$	$0/500 = 0$	$0/500 = 0$

iii. Transition matrix :-

list the retention probabilities along the main diagonal.

If loss probabilities are used, insert them in appropriate row cells.

If gain probabilities are used, insert them in appropriate

Transition matrix

$$P = \begin{matrix} & X & Y & Z \\ X & \left[\begin{matrix} 0.875 & 0.125 & 0 \\ 0.1 & 0.7 & 0.2 \\ 0.05 & 0 & 0.95 \end{matrix} \right] \\ Y & \\ Z & \end{matrix}$$

Classification of states:

The states of Markov chain can be classified based on the transition probability p_{ij} .

(1) A state i is said to be absorbing if it returns to itself with certainty in one transition i.e., $p_{ii} = 1$.

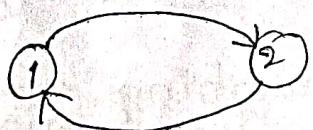


(2) A state i is said to be transient if it can reach another state but cannot itself be reached back from another state.

Mathematically, it will happen if $\lim_{n \rightarrow \infty} p_{ij}^{(n)} = 0$, for all i .



(3) A state i is said to be a recurrent state if the probability of being revisited from other state is one. This can happen if and only if the state is not transient.

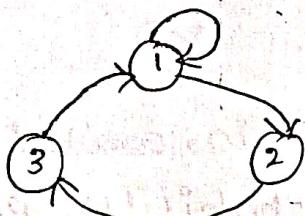


(4) A state i is said to be periodic with period $t > 1$, if a return is possible only in $t, 2t, 3t$ steps. This returns that $p_{ii}^{(n)} = 0$ whenever n is not divisible



If the chain is not periodic, then it is said to be "aperiodic".

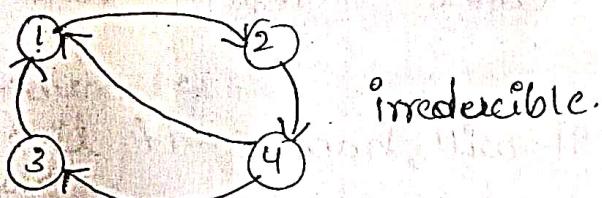
Ex :-



⑤ A Markov chain is irreducible if all states belong to one class (All the states communicate with each other).

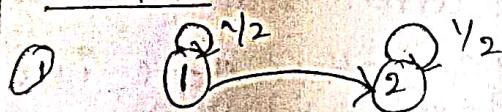
(or) If a chain does not contain any other proper closed subset other than the state space, then the chain is called irreducible, otherwise it is called reducible.

Ex :-



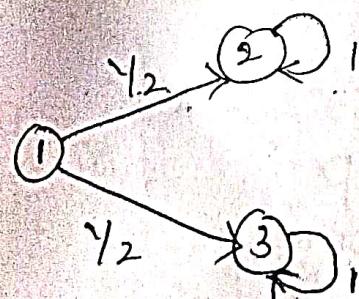
In irreducible markov chain every state can be reached from every other state.

Examples :-



It is absorbing, aperiodic and transient.

②

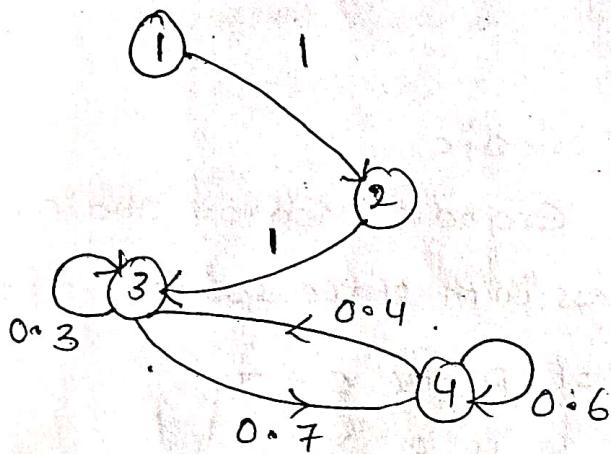


It is absorbing, not irreducible
(not transitive).

1. Classify the states of the following Markov chain.

$$P = \begin{bmatrix} 1 & 2 & 3 & 4 \\ 1 & 0 & 1 & 0 & 0 \\ 2 & 0 & 0 & 1 & 0 \\ 3 & 0 & 0 & 0.3 & 0.7 \\ 4 & 0 & 0 & 0.4 & 0.6 \end{bmatrix}$$

Sol :- The Markov chain is given below



States ① & ② are transient

State 3 & 4 are absorbing and constitute a closed set i.e., irreducible

The chain is aperiodic

Defn :- A closed Markov chain is said to ergodic

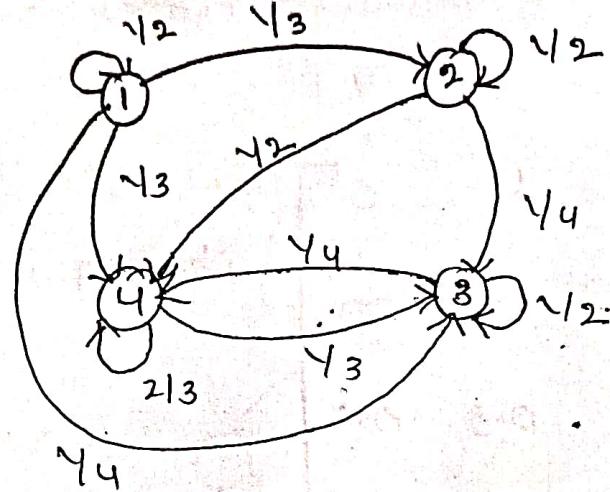
if all of its states are recurrent and aperiodic.

2. Determine if the following transition matrix is ergodic markov chain

	future state			
present	1	2	3	4
1	$\frac{1}{2}$	$\frac{1}{2}$	0	0
2	0	$\frac{1}{2}$	$\frac{1}{2}$	0
3	0	0	$\frac{1}{2}$	$\frac{1}{2}$
4	0	0	0	1

Sol: - The Markov chain is given below

chain.



Here the states 1, 2, 3, 4 are absorbing.

It is possible to go from one state to another state directly or indirectly.

i. It is irreducible.

Also the chain is ~~not~~ a periodic

Hence the given chain is Ergodic Markov chain.

3. Consider a Markov process with state space $S = \{0, 1, 2\}$ and transition matrix $P = \begin{bmatrix} p & q & r \\ r & p & q \\ q & r & p \end{bmatrix}$

a closed

a. what can you say about the values of p and q ?

b. calculate the transition probabilities $P_{ij}^{(3)}$.

c. Draw the transition graph for the process represented by P .

ergodic

Sol: - Since $\sum P_{ij} = 1$, we have

$$p + q + r = 1$$

∴ ergodic

$$P = 1 - \frac{1}{5} - \frac{7}{10} + \frac{1}{2} = \frac{10 - 2 + 5 - 7}{10} = \frac{6}{10} = \frac{3}{5}$$

$$q = 1 - \frac{3}{5} = \frac{5-3}{5} = \frac{2}{5}$$

$$P = \begin{bmatrix} 3/5 & 2/5 & 0 \\ 4/2 & 0 & 4/2 \\ 1/10 & 7/10 & 1/5 \end{bmatrix}$$

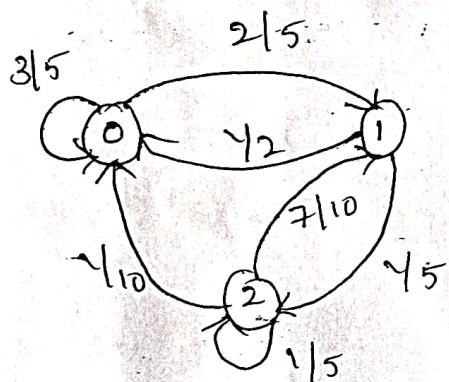
$$\left[\frac{3}{5} - \frac{1}{2} = \frac{6-5}{10} = \frac{1}{10} \right]$$

b. Transition probabilities of $P_{ij}^{(3)}$ are the elements of $(P^3)_{ij}$

$$P_3 = 0 \begin{bmatrix} 3/5 & 2/5 & 0 \\ 4/2 & 0 & 4/2 \\ 1/10 & 7/10 & 1/5 \end{bmatrix}^3 = \frac{1}{(10)^3} \begin{bmatrix} 6 & 4 & 0 \\ 5 & 0 & 5 \\ 1 & 7 & 2 \end{bmatrix}^3$$

$$= \begin{bmatrix} 0.467 & 0.364 & 0.16 \\ 0.495 & 0.21 & 0.295 \\ 0.387 & 0.445 & 0.168 \end{bmatrix} \quad [P^3 = P \times P \times P]$$

c. Transition graph :-



4. Consider a homogeneous Markov chain with state space

$S = \{1, 2, 3\}$ and transition matrix

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

calculate the 3-step transition matrix.

Sol :- Given $P = \begin{bmatrix} 4/4 & 1/2 & 1/4 \\ 1/2 & 0 & 4/2 \\ 3/4 & 1/4 & 0 \end{bmatrix}$

we have that the n-step transition probability matrix

e.g. $P_{ij}^{(3)} = \begin{bmatrix} 1/4 & 1/2 & 1/4 \\ 1/2 & 0 & 1/2 \\ 3/4 & 1/4 & 0 \end{bmatrix}^3$

$$P^3 = \left(\frac{1}{4}\right)^3 \begin{bmatrix} 1 & 2 & 1 \\ 2 & 0 & 2 \\ 3 & 1 & 0 \end{bmatrix}^3 = \frac{1}{64} \begin{bmatrix} 29 & 21 & 14 \\ 26 & 18 & 20 \\ 32 & 15 & 17 \end{bmatrix}$$

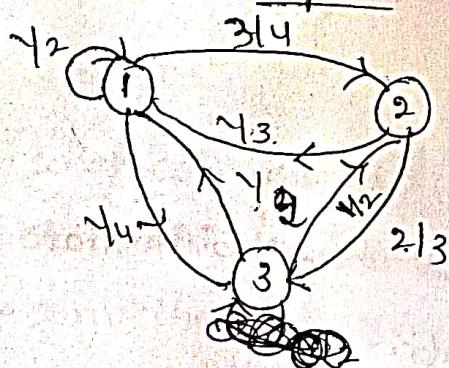
5. consider the Markov chain with 3 states $S = \{1, 2, 3\}$ that has the following transition matrix $P = \begin{bmatrix} 1/2 & 3/4 & 1/4 \\ 1/3 & 0 & 2/3 \\ 1/2 & 1/2 & 0 \end{bmatrix}$

a. Draw the state transition diagram for this chain.

b. If we know $P(x_1=1) = P(x_1=2) = 1/4$, find $P(x_1=3, x_2=x_3=1)$

So:- Given $P = \begin{bmatrix} 1/2 & 3/4 & 1/4 \\ 1/3 & 0 & 2/3 \\ 1/2 & 1/2 & 0 \end{bmatrix}$

Transition diagram.



⑥ Given $P(x_1=1) = P(x_1=2) = 1/4$

$$\therefore P(x_1=3) = 1 - P(x_1=1) - P(x_1=2)$$

$$= 1 - 1/4 - 1/4 = 1 - \frac{1}{2} = \frac{1}{2}$$

We can write $P(x_1=3, x_2=2, x_3=1) = P(x_1=3) P_{32} P_{21}$

$$= 1/2 \cdot 1/2 \cdot 1/3 = \frac{1}{12}$$

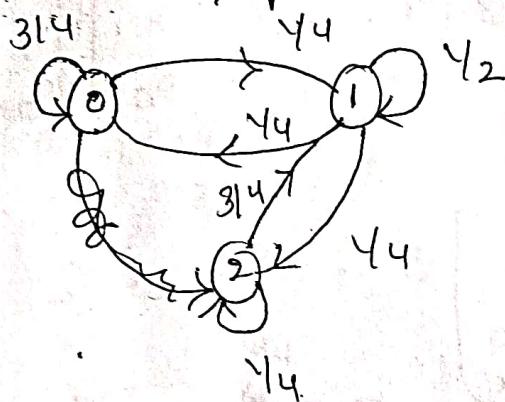
b. Inc transition probability from a markov chain with states 0, 1, 2 if $P = \begin{bmatrix} 3/4 & 1/4 & 0 \\ 1/4 & 1/2 & 1/4 \\ 0 & 3/4 & 1/4 \end{bmatrix}$ and the initial state distribution of the chain is $P(x_0 = i) = 1/3$, $i = 0, 1, 2$ obtain i. $P(x_2 = 2)$ ii. $P(x_2 = 2, x_1 = 1, x_0 = 2)$

Manu 10 M

Sol :- Given $P = \begin{bmatrix} 0 & 1 & 2 \\ 3/4 & 1/4 & 0 \\ 1/4 & 1/2 & 1/4 \\ 0 & 3/4 & 1/4 \end{bmatrix}$

Sol :

Transition diagram



Given $P(x_0 = i) = 1/3$, $i = 0, 1, 2$

$$\therefore P(x_0 = 0) = P(x_0 = 1) = P(x_0 = 2) = 1/3$$

$$P(x_1 = 0) = P(x_1 = 1) = P(x_1 = 2) = 1/3$$

$$P(x_2 = 0) = P(x_2 = 1) = P(x_2 = 2) = 1/3$$

$$\text{i. } P(x_2 = 2) = 1 - P(x_2 = 0) - P(x_1 = 1)$$

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

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

$$\text{ii. } P(x_2 = 2, x_1 = 1, x_0 = 2) = P(x_2 = 2) P_{21} P_{12}$$

from power
odd
Since
the
But
state

chain with initial state
 $x_0 = \vec{y} \theta = 1/3$, $x_0 = 2$)

Test the following transition matrix to see if the Markov chain is regular and ergodic where x is some positive p_{ij} value.

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

Sol :- we compute $P^2 =$

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

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

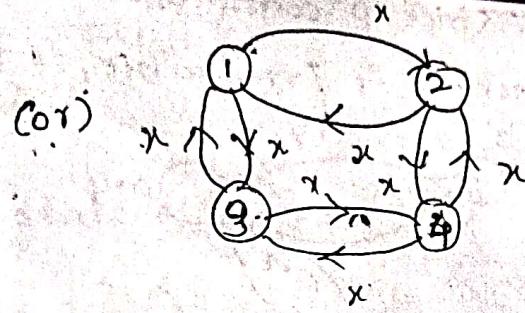
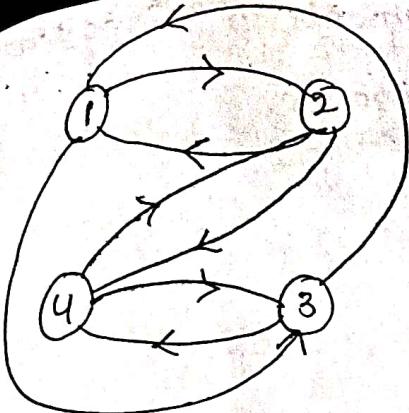
Similarly $P^4 =$

$$\begin{bmatrix} x & 0 & 0 & x \\ 0 & x & x & 0 \\ 0 & x & x & 0 \\ x & 0 & 0 & x \end{bmatrix} \quad P^8 = \begin{bmatrix} x & 0 & 0 & x \\ 0 & x & x & 0 \\ 0 & x & x & 0 \\ x & 0 & 0 & x \end{bmatrix}$$

from this we observe P raised to an even number power gives the result as above, while P raised to an odd numbers power will give the original matrix.

Since all the elements are not non-zero positive elements, the given matrix is not regular.

But, it is ergodic - since it is possible to go from one state to another.



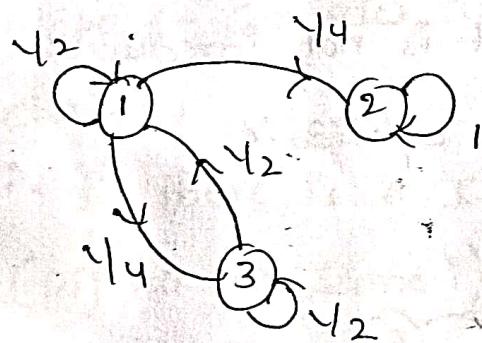
8. Which of the following stochastic matrix is regular and ergodic Markov chain.

$$i. P = \begin{bmatrix} 1/2 & 1/4 & 1/4 \\ 0 & 1 & 0 \\ 1/2 & 0 & 1/2 \end{bmatrix}$$

$$ii. P = \begin{bmatrix} 0 & 2/3 & 1/3 \\ 1/2 & 0 & 1/2 \\ 1/2 & 1/2 & 0 \end{bmatrix}$$

$$iii. \begin{bmatrix} 0.3 & 0.7 & 0 \\ 0.1 & 0.4 & 0.5 \\ 0 & 0.2 & 0.8 \end{bmatrix}$$

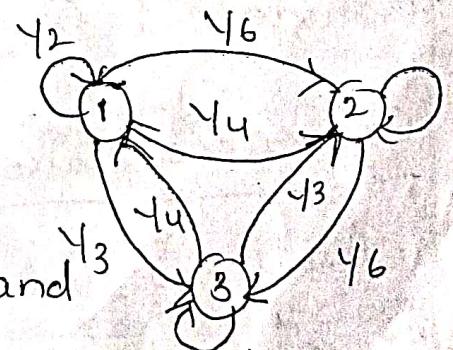
Sol :- Since $P_{22} = 1$, it is not regular



It is not irreducible state 3
1, 2 are not recurrent
No transition from state 2
to state 1, and from state
2 to state 3

It is not ergodic.

containing all +ve elements.

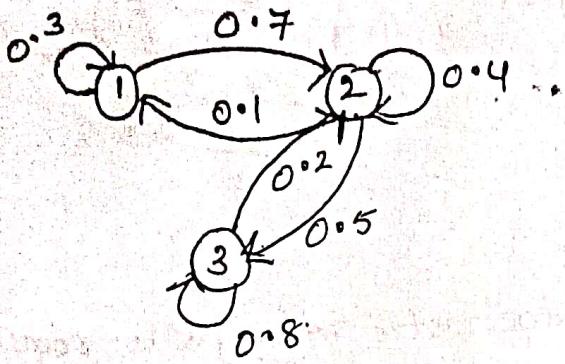


P is a regular Markov chain

Since all states are recurrent and

P is not periodic, it is an ergodic chain.

iii. Since P contains all the elements, P is regular
since P is irreducible, it is ergodic



regular and

Examples :-

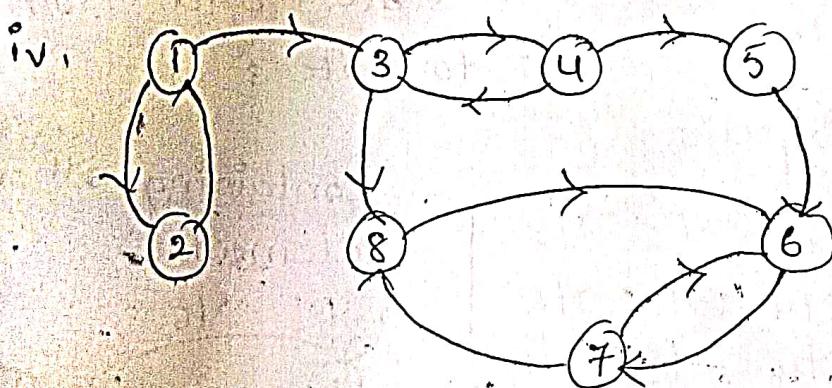
i. $P = \begin{bmatrix} 0 & 1 & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 0 & 1 & 0 \end{bmatrix}$ It is irreducible but not aperiodic.

ii. $P = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 0 \end{bmatrix}$ It is not irreducible and is aperiodic.

iii. $P = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & \frac{1}{2} & \frac{1}{2} \end{bmatrix}$ It is not irreducible and is aperiodic.

State 2
from State

all +ve



Equivalence classes :-

Class 1 = {State 1, State 2}

\therefore State 1 & State 2 communicate each other but not

7/13
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Class 2 = {state 3, state 4}

Class 3 = {state 5}

Class 4 = {state 6, state 7, state 8}

Class 1 = {state 1, state 2} is aperiodic since it has a self transition, $p_{22} > 0$

Class 2 = {state 3, state 4} is periodic with period 2.

Class 4 is aperiodic

Because, we can go from state 6 to state 6 in two steps (6-7-6) and in three steps (6-7-8-6)

Since $\gcd(2, 3) = 1$, we conclude that state 6 and Class 4 is aperiodic.

Note:- Consider a peri finite irreducible Markov chain.

i. If there is a self transition in the chain ($p_{ii} > 0$ for same i) then the chain is aperiodic.

ii. Suppose, you can go from state i to state j in l -steps i.e., $p_{ij}^{(l)} > 0$, also in m steps ($p_{ij}^{(m)} > 0$)

If $\gcd(l, m) = 1$, then the state i is aperiodic.

Steady state (Equilibrium) condition : (2M Manu)

The determination of "steady state" conditions is a regular ergodic Markov chain can be accomplished most readily by computing p^n for larger values of n .

large term probability that the system will be in state (or) any number of transitions applied to the system has no impact on the state vector i.e. the current behaviour of the system will continue into the future.

Mathematically,

$$\lim_{n \rightarrow \infty} R(n+1) = \lim_{n \rightarrow \infty} R(n) \quad (\text{or}) \quad R = RP \rightarrow ①$$

Therefore, as $n \rightarrow \infty$, $R(n)$ becomes constant (i.e., independent of time) and then the system is said to have reached to a steady state equilibrium.
Note:- As $n \rightarrow \infty$, the state probabilities at $(n+1)^{\text{th}}$ period are very close to those that n^{th} period.

problems

1. Consider a Markov chain with only two states s_1, s_2 and transition matrix $P = \begin{bmatrix} 1/2 & 1/2 \\ 1/3 & 2/3 \end{bmatrix}$ determine the stationary distribution of P .

Sol:- Let $R = (R_0, R_1)$ be the stationary distribution.

Given $P = \begin{bmatrix} 1/2 & 1/2 \\ 1/3 & 2/3 \end{bmatrix}$

We have $R = RP$ (or) $RP = R$, where $R_0 + R_1 = 1$

i.e.; $\begin{bmatrix} R_0 & R_1 \end{bmatrix} \begin{bmatrix} 1/2 & 1/2 \\ 1/3 & 2/3 \end{bmatrix} = \begin{bmatrix} R_0 & R_1 \end{bmatrix}$

$$\frac{1}{2}R_0 + \frac{1}{3}R_1 = R_0 \Rightarrow -\frac{R_0}{2} + \frac{R_1}{3} = 0 \rightarrow ①$$

$$\frac{1}{2}R_0 + \frac{2}{3}R_1 = R_1 \Rightarrow \frac{R_0}{2} - \frac{R_1}{3} = 0 \rightarrow ②$$

We have $R_0 + R_1 = 1 \Rightarrow R_1 = 1 - R_0$

$$\textcircled{1} \Rightarrow -\frac{R_0}{2} + \frac{1-R_0}{3} = 0 \Rightarrow \frac{R_0}{2} - \frac{R_0}{3} = -\frac{1}{3}$$

$$\Rightarrow \left(-\frac{1}{2} - \frac{1}{3}\right) R_0 = -\frac{1}{3} \Rightarrow \left[-\frac{2+2}{6}\right] R_0 = -\frac{1}{3}$$

$$\Rightarrow -\frac{5R_0}{6} = -\frac{1}{3} \Rightarrow R_0 = \frac{2}{5}$$

$$R_1 = 1 - \frac{2}{5} = \frac{3}{5} \Rightarrow R_1 = \frac{3}{5}$$

2. Consider the matrix of transition probabilities of a product available in the matrix in two brands A & B, $P = A \begin{bmatrix} A & B \\ 0.9 & 0.1 \\ B & 0.3 & 0.7 \end{bmatrix}$ determine the Market share of each brand in equilibrium position

Sol:- Given $P = A \begin{bmatrix} A & B \\ 0.9 & 0.1 \\ B & 0.3 & 0.7 \end{bmatrix}$ and $R = [A, B]$

At equilibrium condition, we have $RP = R$

$$\text{where } A+B=1 \rightarrow \textcircled{1}$$

$$[AB] \begin{bmatrix} 0.9 & 0.1 \\ 0.3 & 0.7 \end{bmatrix} = [A \ B]$$

$$(0.9)A + (0.3)B = A \rightarrow \textcircled{1}$$

$$(0.1)A + (0.7)B = B \rightarrow \textcircled{2}$$

from \textcircled{1}

$$B = 1 - A$$

$$\therefore \textcircled{1} \Rightarrow (0.9)A + (0.3)(1-A) = A$$

$$\Rightarrow (0.9)A + 0.3 - (0.3)A - A = 0$$

$$\Rightarrow -(0.4)A = -0.3$$

$$A = \frac{0.3}{0.4} = \frac{3}{4}$$

$$\text{Hence } B = 1 - \frac{3}{4} = \frac{1}{4}$$

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The market share of brand A is 75% and the brand B is 25%.

3. A manufacturing company has a certain piece of equipment that is inspected at the end of each day and classified as 'just overhauled', 'good', 'fair', or 'inoperative'. If the item is inoperative, it is overhauled a procedure that takes one day. Let us denote the four classifications as states 1, 2, 3 and 4 respectively. Assume that the working condition of the equipment follows a Markov process with the following transition matrix:

		Tomorrow			
		1	2	3	4
Today	1	0	3/4	1/4	0
	2	0	4/2	4/2	0
	3	0	0	1/2	1/2
	4	1	0	0	0

It costs Rs 125 to overhaul a machine (including lost time) on the average, and Rs 75 in production lost if a machine is found inoperative. Using steady state probabilities, compute the expected per day cost of maintenance.

Sol:- Since the given matrix 'P' is an ergodic regular

Let the steady state probabilities p_1, p_2, p_3 and p_4 represent the proportion of times that the machine will be in states 1, 2, 3 and 4 respectively. from the steady state equations, we have

$$R = RP$$

i.e; $(p_1 \ p_2 \ p_3 \ p_4) = (p_1 \ p_2 \ p_3 \ p_4) \begin{bmatrix} 0 & 3/4 & 1/4 & 0 \\ 0 & 1/2 & 1/2 & 0 \\ 0 & 0 & 1/2 & 1/2 \\ 1 & 0 & 0 & 0 \end{bmatrix}$

$$\Rightarrow p_1 = p_4 \rightarrow (1)$$

$$p_2 = \frac{3}{4} p_1 + \frac{1}{2} p_2 \rightarrow (2)$$

$$p_3 = \frac{1}{4} p_1 + \frac{1}{2} p_2 + \frac{1}{2} p_3 \rightarrow (3)$$

$$p_4 = \frac{1}{2} p_3 \rightarrow (4)$$

and $p_1 + p_2 + p_3 + p_4 = 1 \rightarrow (5)$

Solving these simultaneous equations we get

$$p_1 = \frac{2}{11}, p_2 = \frac{3}{11}, p_3 = \frac{4}{11} \text{ & } p_4 = \frac{2}{11}$$

On an average, 2 out of every 11 days the machine will be overhauled, 3 out of every 11 days it will be in good condition, 4 out of every 11 days it will be in fair condition and 2 out of every 11 days it will

be found inoperative at the end of the days.

Hence the expected (average) cost per day of maintenance will be given by

$$(2/11)(125) + (3/11)(75) = Rs. 36$$

Markov Analysis :-

and p_4
machine

- i. The rate of gains and losses for the current market.
- ii. The projected market share of each item in the system in the future.
- iii. The equilibrium state fact could exist.
The best example of Markov analysis is 'Brand switching' Model's, which enquires on the time behaviour of customers who make repeated purchase of a product class, but from time to time may switch over from one brand to another. Here the state is generally the customer's preference for a particular brand.

Brand switching Example :-

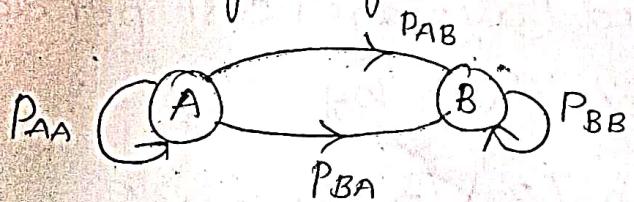
Let us consider a consumer sample distributed over two brands - A and B, the samples being the representative of the entire group the standpoint of their brand loyalty and their switching patterns.

The behaviour of the large groups can be better described in probabilistic terms. This probabilistic description can be represented by transition matrix as explained by the following diagram.

Transition matrix :

$$\text{from } \begin{matrix} & A & B \\ A & P_{AA} & P_{AB} \\ B & P_{BA} & P_{BB} \end{matrix}$$

Brand switching diagram :



In general, the transition matrix for n brands A_1, A_2, \dots, A_n is given by

$$P = \begin{bmatrix} A_1 & A_2 & \dots & A_j & \dots & A_n \\ A_1 & P_{11} & P_{12} & \dots & P_{1j} & \dots & P_{1n} \\ A_2 & P_{21} & P_{22} & \dots & P_{2j} & \dots & P_{2n} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ A_i & P_{i1} & P_{i2} & \dots & P_{ij} & \dots & P_{in} \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots & \vdots \\ A_n & P_{n1} & P_{n2} & \dots & P_{nj} & \dots & P_{nn} \end{bmatrix}$$

Where $\sum_{j=1}^n P_{ij} = 1, i = 1, 2, \dots, n$

Markovian Algorithm

The following algorithm gives a procedure for solving the application problems.

Step-1 :- from the given data, first determine the probabilities associated with retentions, gain and losses.

Step-2 :- Develop the appropriate state-transition matrix.

Step-3 :- With the help of the transition matrix obtained in step-2, determine the expected future market shares for the next period.

The procedure is as follows.

(Market shares beginning of period 1) \times (state-transition matrix) = Expected market shares beginning of period 2

Step-4 :- If more than one expected market share calculation is needed, then apply the follow

Step-2, find the equilibrium conditions for the current problem.

Problems

1. Suppose there are two market products of brand A and B respectively. Let each of these two brands have exactly 50% of the total market in same period and let the market be of a fixed size, the transition matrix is given as follows

$$\text{from } \begin{matrix} & \begin{matrix} A & B \end{matrix} \\ \begin{matrix} A \\ B \end{matrix} & \begin{bmatrix} 0.9 & 0.1 \\ 0.5 & 0.5 \end{bmatrix} \end{matrix}$$

If the initial market share breakdown is 50% for each brand then determine their market shares in the steady state

Let $P = \begin{bmatrix} & \begin{matrix} A & B \end{matrix} \\ \begin{matrix} A \\ B \end{matrix} & \begin{bmatrix} 0.9 & 0.1 \\ 0.5 & 0.5 \end{bmatrix} \end{bmatrix}$ be the given transition matrix

It is given that the initial market shares at the beginning of period 1 are 50% each. i.e., $R = (0.5 \ 0.5)$

Market shares of brands A & B in second period (or) after one steady period is given by

(Market shares in the beginning of period 1) \times (state transition matrix) = (expected market shares in the beginning of period 2)

$$\text{i.e., } (0.5 \ 0.5) \times \begin{bmatrix} 0.9 & 0.1 \\ 0.5 & 0.5 \end{bmatrix} = (0.5 \times 0.9 + 0.5 \times 0.5 + 0.5 \times 0.1 + 0.5 \times 0.5)$$

$$= (0.7 \ 0.3) = (70\% \ 30\%)$$

After two steady periods

$$(0.7 \ 0.3) \times \begin{bmatrix} 0.9 & 0.1 \\ 0.5 & 0.5 \end{bmatrix} = (0.7 \times 0.9 + 0.3 \times 0.5 + 0.7 \times 0.1 + 0.3 \times 0.5)$$

$$= (0.79 \ 0.21)$$

$$\text{Similarly } (0.78 \ 0.22) \begin{bmatrix} 0.4 & 0.1 \\ 0.5 & 0.5 \end{bmatrix} = (81.2 \ 18.8)$$

$$(81.2 \ 18.8) \begin{bmatrix} 0.9 & 0.1 \\ 0.5 & 0.5 \end{bmatrix} = (82.48 \ 17.52)$$

$$(82.48 \ 17.52) \begin{bmatrix} 0.9 & 0.1 \\ 0.5 & 0.5 \end{bmatrix} = (82.992 \ 17.008)$$

$$(82.992 \ 17.008) \begin{bmatrix} 0.9 & 0.1 \\ 0.5 & 0.5 \end{bmatrix} = (83.17)$$

After 6 time periods the resulting market shares are 83% and 17% respectively.

The equilibrium position of market shares of A and B will be $5/6$ and $1/6$ of the total market respectively.

Q. A Market research team has conducted a survey of customer buying habits with respect to three brands of talcum powder in an area. It estimates at present 20% of the customers buy brand A, 50% of the customers buy brand B and 30% of the customers buy brand C. In addition, the team has analysed its survey and has determined the following brand switching matrix.

Brand Next bought		A	B	C
Brand just bought	A	0.6	0.3	0.1
	B	0.4	0.5	0.1
C	0.2	0.1	0.7	

Determine the expected distribution of consumers two time periods later.

Sol :- Given transition matrix is

$$\begin{array}{c} A \quad B \quad C \\ \hline A & \begin{bmatrix} 0.6 & 0.3 & 0.1 \\ 0.4 & 0.5 & 0.1 \\ 0.2 & 0.1 & 0.7 \end{bmatrix} \\ B & \\ C & \end{array}$$

Given that the initial buying of the three brands, A, B & C are (20%, 50%, 30%)

i.e. (0.2, 0.5, 0.3)

After one steady period;

$$\begin{bmatrix} \text{expected distribution} \\ \text{in present period} \end{bmatrix} \times \begin{bmatrix} \text{Transition matrix} \end{bmatrix} = \begin{bmatrix} \text{expected distribution} \\ \text{in next period} \end{bmatrix}$$

$$(0.2 \quad 0.5 \quad 0.3) \begin{bmatrix} 0.6 & 0.3 & 0.1 \\ 0.4 & 0.5 & 0.1 \\ 0.2 & 0.1 & 0.7 \end{bmatrix} = (0.88 \quad 0.34 \quad 0.28)$$

After the steady periods;

$$(0.88 \quad 0.34 \quad 0.28) \begin{bmatrix} 0.6 & 0.3 & 0.1 \\ 0.4 & 0.5 & 0.1 \\ 0.2 & 0.1 & 0.7 \end{bmatrix} = (0.42 \quad 0.312 \quad 0.268)$$

Hence after two periods.

expected distribution of customers for three brands

A, B & C are 0.42, 0.312, 0.268 respectively

(or) 42%, 31.2% & 26.8% respectively

3. Suppose there are three dairies in a town, say A, B and C. They supply all the milk consumed in the town. It is known by all the dairies that consumers switched from dairy to dairy overtime because of advertising dissatisfaction with service and other reasons. All these dairies maintain records of the number of their customers and the dairy from which may

Dairy	June (1) Customer	Gain from			Losses to			July 1 Customer
		A	B	C	A	B	C	
A	200	0	35	25	0	20	20	220
B	500	20	0	20	35	0	15	490
C	300	20	15	0	25	20	0	290

We assume that the matrix of transition probabilities remain fairly stable and that the July market shares are

$$A = 22\% \quad B = 49\% \quad C = 29\%$$

Managers of these dairies are willing to know:

- i. market share of their dairies on 1st August & 1st Sep
- ii. their market shares in steady state.

Sol :- We have that the transition probability matrix

$$\text{P} = \begin{bmatrix} P_{AA} & P_{AB} & P_{AC} \\ P_{BA} & P_{BB} & P_{BC} \\ P_{CA} & P_{CB} & P_{CC} \end{bmatrix}$$

i.e., The probability transition matrix is

$$\begin{array}{ccc} & A & B & C \\ \begin{matrix} A \\ B \\ C \end{matrix} & \begin{bmatrix} \frac{160}{200} = 0.8 & \frac{20}{200} = 0.1 & \frac{20}{200} = 0.1 \\ \frac{35}{500} = 0.07 & \frac{450}{500} = 0.9 & \frac{15}{500} = 0.03 \\ \frac{95}{300} = 0.083 & \frac{20}{300} = 0.067 & \frac{25}{300} = 0.083 \end{bmatrix} \end{array}$$

Retention probabilities:

for brand A, $P_{AA} = \frac{\text{Number of retained}}{\text{Number of customer originally served}}$

$$\text{Number retained} = (\text{Original no} - \text{number lost})$$

July 1
customers
200

$$P_{AA} = \frac{160}{200} = 0.8$$

July $P_{BB} = \frac{450}{500} = 0.9$, $P_{CC} = \frac{255}{300} = 0.85$

Gain probabilities:

from B to A, $P_{BA} = \frac{\text{No. of customers gained from B}}{\text{Total no. of customers served for the source of gain (B)}}$

$$P_{BA} = \frac{35}{500} = 0.07$$

from C to A;

$$P_{CA} = \frac{\text{No. of customer gained from C}}{\text{Total no. of customer of C}} = \frac{25}{300} = 0.083$$

from A to B; $P_{AB} = \frac{20}{200} = 0.1$

from C to B; $P_{CB} = \frac{20}{300} = 0.067$

from A to C; $P_{AC} = \frac{20}{200} = 0.1$

from B to C; $P_{BC} = \frac{15}{500} = 0.03$

The transition matrix is,

$$\begin{bmatrix} 0.80 & 0.1 & 0.1 \\ 0.07 & 0.9 & 0.03 \\ 0.083 & 0.067 & 0.85 \end{bmatrix}$$

The market shares (on July 1st) are of the three dairy A, B & C are 22%, 49% and 29% respectively.

After one steady period (Market shares on 1st August.)

$$\begin{pmatrix} 0.22 & 0.49 & 0.29 \end{pmatrix} \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.07 & 0.9 & 0.03 \\ 0.083 & 0.067 & 0.85 \end{bmatrix} = \begin{pmatrix} 0.234 & 0.483 & 0.283 \end{pmatrix}$$

i. Co
two
ii. D
iii - C
beh
man
sol:

$$\begin{pmatrix} 0.234 & 0.483 & 0.283 \end{pmatrix} \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.07 & 0.9 & 0.03 \\ 0.083 & 0.067 & 0.85 \end{bmatrix} = \begin{pmatrix} 0.245 & 0.477 & 0.278 \end{pmatrix}$$

ii. The steady state market shares are given by

$$\begin{bmatrix} x & y & z \end{bmatrix} \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.07 & 0.9 & 0.03 \\ 0.083 & 0.067 & 0.85 \end{bmatrix} = \begin{bmatrix} x & y & z \end{bmatrix}$$

$$\Rightarrow (0.8)x + (0.07)y + (0.083)z = x \rightarrow (1)$$

$$(0.1)x + (0.9)y + (0.067)z = y \rightarrow (2)$$

$$(0.1)x + (0.03)y + (0.85)z = z \rightarrow (3)$$

$$\text{and } x+y+z=1$$

Solving (1) (2) (3) & (4) we get

$$x = 0.273 \quad y = 0.454 \quad z = 0.273$$

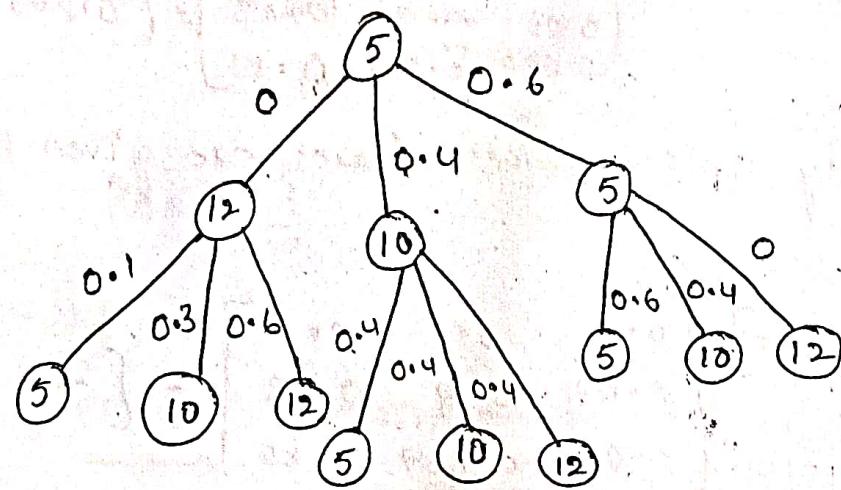
4. The number of units of an item that are withdrawn from inventory on a day-to-day basis is a Markov chain process in which requirements for tomorrow depend on today's requirements. A one day transition matrix is given below:

Number of units withdrawn from inventory

	5	10	12
5	0.6	0.4	0.0
Today	0.3	0.3	0.4

- 3 0.883) + sep) 177 0.278)
- Construct a tree diagram showing inventory requirements two consecutive days
 - Develop a two-day transition matrix
 - Comment on "how a two day transition matrix might be helpful to a manager who is responsible for inventory management".

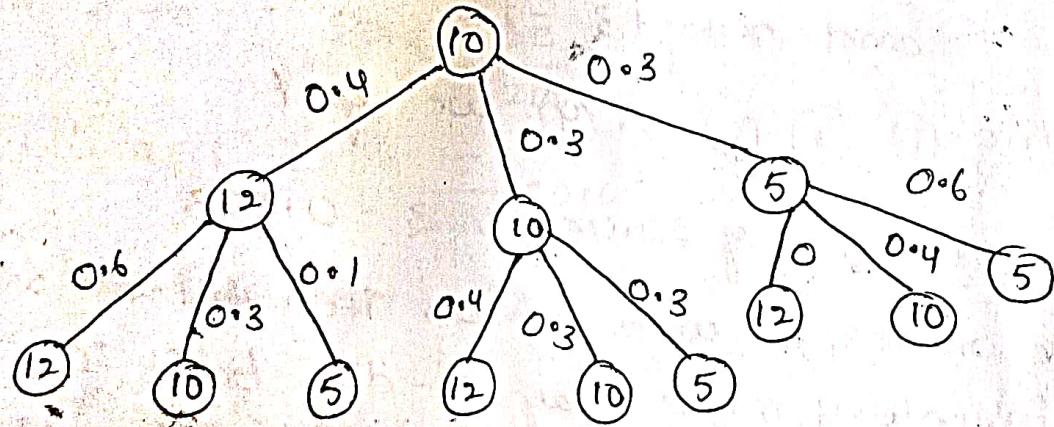
Sol:-



$$P_{11}^{(2)} = (0.6)(0.6) + (0.4)(0.3) + (0)(0.1) = 0.48$$

$$P_{12}^{(2)} = (0.6)(0.4) + (0.4)(0.3) + (0)(0.3) = 0.36$$

$$P_{13}^{(2)} = (0.6)(0) + (0.4)(0.4) + (0)(0.6) = 0.16$$

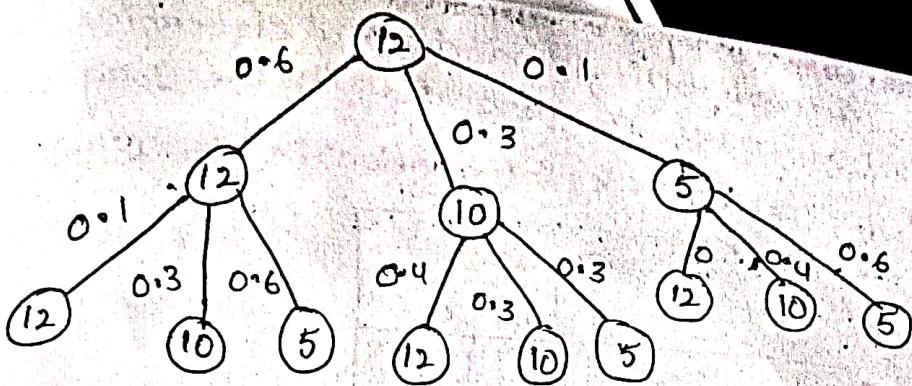


$$P_{22}^{(2)} = (0.3)(0.4) + (0.3)(0.3) + (0.4)(0.3) = 0.33$$

$$P_{21}^{(2)} = (0.3)(0.6) + (0.3)(0.3) + (0.4)(0.1) = 0.31$$

$$P_{23}^{(2)} = (0.3)(0) + (0.3)(0.4) + (0.4)(0.6) = 0.36$$

it draws
Markov
so depend
matrix P



$$P_{31}^{(2)} = (0.1)(0.6) + (0.3)(0.3) + (0.6)(0.1) = 0.21$$

$$P_{32}^{(2)} = (0.1)(0.4) + (0.3)(0.3) + (0.6)(0.3) = 0.31$$

$$P_{33}^{(2)} = (0.1)(0) + (0.3)(0.4) + (0.6)(0.6) = 0.48$$

ii. If the transition matrix is P , then the two-day transition matrix is given by

$$P^2 = \begin{bmatrix} 0.6 & 0.4 & 0.0 \\ 0.3 & 0.3 & 0.4 \\ 0.1 & 0.3 & 0.6 \end{bmatrix} \begin{bmatrix} 0.6 & 0.4 & 0.0 \\ 0.3 & 0.3 & 0.4 \\ 0.1 & 0.3 & 0.6 \end{bmatrix}$$

$$= \begin{matrix} & 5 & 10 & 12 \\ 5 & 0.48 & 0.36 & 0.16 \\ 10 & 0.31 & 0.33 & 0.36 \\ 12 & 0.21 & 0.31 & 0.48 \end{matrix}$$

iii. The two day transition matrix can be used for guiding ordering decisions.

for ex : if today the manager experiences a demand for five units, then two days later the probability of requiring 5 units is 0.48, 10 units is 0.36 and 12 units is 0.16.

5. A housewife buys three kind of cereals: A, B & C. She never buys the same cereal on successive weeks if she buys cereal A then the next week she buys cereal B. However, it

times as likely to buy now as the other brand. Obtain the transition probability matrix and determine how often she would buy each of the cereals in the long run.

Sol :- We present the information in the following table.

Cereals	Cereals		
	A	B	C
A	0	1	0
B	0.75	0	0.25
C	0.75	0.25	0

The transition matrix P_t given by

$$P_t = \begin{bmatrix} A & B & C \\ 0 & 1 & 0 \\ 0.75 & 0 & 0.25 \\ 0.75 & 0.25 & 0 \end{bmatrix}$$

If x, y and z represent the steady state probabilities of states A, B and C respectively, then we have

$$\begin{bmatrix} x & y & z \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0.75 & 0 & 0.25 \\ 0.75 & 0.25 & 0 \end{bmatrix} = \begin{bmatrix} x & y & z \end{bmatrix}$$

where $x+y+z=1$ and $x \geq 0, y \geq 0, z \geq 0$

The equations are

$$0 + (0.75)y + (0.75)z = x$$

$$x + 0 + (0.25)z = y$$

$$0 + (0.25)y + 0 = z$$

(or)

$$x - 0.75y - 0.75z = 0 \rightarrow (1)$$

$$x - y + 0.25z = 0 \rightarrow (2)$$

Solving these equations, we get:

$$x = 0.428, y = 0.457, z = 0.114$$

Hence, in the long run, the house wife would buy the cereal A, B and C with a frequency of 42.8%, 45.7% and 11.4% respectively.

6. The price of an equity share of a company may increase, decrease or remain constant on any given day. It is assumed that the change in price any day affects the change on the following day as described by the following transition matrix:

		Change Tomorrow		
		Increase	Decrease	Unchanged
Change Today	Increase	0.5	0.2	0.3
	Decrease	0.7	0.1	0.2
	Unchanged	0.4	0.5	0.1

- If the price of share increased today, what are chances that it will increase, decrease or unchanged tomorrow?
- If the price of the share decreased today, what are the chances that it will increase tomorrow?
- If the price of the share remained unchanged today, what are the chances that it will increase, decrease or remain unchanged day after tomorrow?

Sol:-

- Given that the probability of increase in price of the share today is 1.
∴ The probability of decreasing or unchanged is zero.

After one steady period,

$$\begin{bmatrix} \text{Today's probabilities} \end{bmatrix} \times \begin{bmatrix} \text{Transition matrix} \end{bmatrix} = \begin{bmatrix} \text{expected probabilities for tomorrow} \end{bmatrix}$$

$$\begin{bmatrix} 1.00 \end{bmatrix} \begin{bmatrix} 0.5 & 0.2 & 0.3 \\ 0.7 & 0.1 & 0.2 \\ 0.4 & 0.5 & 0.1 \end{bmatrix} = \begin{bmatrix} 0.5 & 0.2 & 0.3 \end{bmatrix}$$

This gives that the price will increase, decrease or unchanged tomorrow by 50%, 20% and 30% respectively.

ii. If the probability of decrease in price of a share is 1, then the probability of the other two events is zero.

∴ probabilities for today are (0 1 0).

After one steady period,

$$\begin{bmatrix} 0.10 \end{bmatrix} \begin{bmatrix} 0.5 & 0.2 & 0.3 \\ 0.7 & 0.1 & 0.2 \\ 0.4 & 0.5 & 0.1 \end{bmatrix} = \begin{bmatrix} 0.7 & 0.1 & 0.2 \end{bmatrix}$$

This indicates that the elements of price increase tomorrow is 70%.

iii. Let the probability of prices remain unchanged today be 1 and the probabilities of the other two events be zero.

then the probabilities for today are (0 0 1)

After one steady period

$$\begin{bmatrix} 0.01 \end{bmatrix} \begin{bmatrix} 0.5 & 0.2 & 0.3 \\ 0.7 & 0.1 & 0.2 \\ 0.4 & 0.5 & 0.1 \end{bmatrix} = \begin{bmatrix} 0.4 & 0.5 & 0.1 \end{bmatrix}$$

After two steady periods;

Q. On January 1 (this year), Bakery A had 40% of its local market share while the other two bakeries B and C had 40% and 20% respectively of the market share. Based upon a study by a marketing research firm, the following facts were compiled. Bakery A retains 90% of its customers while gaining 5% competitor B's customers and 10% of C's customers. Bakery B retains 85% of its customers while gaining 5% of A's customers and 7% of C's customers. Bakery C retains 83% of its customers and gains 5% of A's customers and 10% of B's customers. What will each firm's share be on January 1 (next year) and what will each firm's market share be at equilibrium?

Sol :- From the given data, we can form the state-transition matrix as given below.

$$\begin{array}{c}
 & & & \text{Bakery} \\
 & A & B & C \\
 \text{Bakery} & \begin{bmatrix} 0.90 & 0.05 & 0.05 \\ 0.05 & 0.85 & 0.10 \\ 0.10 & 0.07 & 0.83 \end{bmatrix} & \xrightarrow{\text{Retention and Gain}} & \text{Retention and Loss} \\
 & & & \xrightarrow{\quad\quad\quad} &
 \end{array}$$

Given that the initial (i.e., on January 1, present year) market shares of three Bakeries A, B & C are 40%, 40% and 20% respectively.

The expected market shares for A, B & C on January 1, next year are determined as follows.

$$\left[\begin{array}{l} \text{Market shares} \\ \text{on Jan 1} \\ \text{Present year} \end{array} \right] \times \left[\begin{array}{l} \text{state-transition} \\ \text{matrix} \end{array} \right] = \left[\begin{array}{l} \text{Expected market} \\ \text{shares on Jan 1,} \\ \text{next year} \end{array} \right]$$

$$\left[\begin{array}{l} 0.40 \ 0.40 \ 0.20 \end{array} \right] \times \left[\begin{array}{l} 0.90 \ 0.05 \ 0.05 \\ 0.05 \ 0.85 \ 0.10 \\ 0.10 \ 0.07 \ 0.83 \end{array} \right] = \left[\begin{array}{l} 0.400 \ 0.374 \ 0.226 \end{array} \right]$$

Hence the market shares of Bakeries A, B, and C on January 1 next year will be 40%, 37.4%, and 22.6% respectively.

If x, y, z be the market shares at equilibrium state, then we have,

$$[x \ y \ z] \begin{bmatrix} 0.90 & 0.05 & 0.05 \\ 0.05 & 0.85 & 0.10 \\ 0.10 & 0.07 & 0.83 \end{bmatrix} = [x \ y \ z]$$

and $x + y + z = 1$ where $x, y, z \geq 0$

The equations are

$$(0.90)x + (0.05)y + (0.10)z = x$$

$$(0.05)x + (0.85)y + (0.10)z = y$$

$$(0.05)x + (0.10)y + (0.83)z = z$$

The eqn's can be written as

$$-0.10x + 0.05y + (0.10)z = 0 \rightarrow (1)$$

$$0.05x - 0.15y + 0.07z = 0 \rightarrow (2)$$

$$0.05x + 0.10y - 0.17z = 0 \rightarrow (3) \quad \& \quad x + y + z = 1 \rightarrow (4)$$

Solving, we get $x = 0.43$, $y = 0.28$, $z = 0.29$

The market shares at equilibrium are

for Bakery A : 43% of the total market

for Bakery B = 28% of the total market

for Bakery C = 29% of the total market

Some important problems: [Markov chain]

Notation : $P(x_n)$; it means the probability of the state 'i' after 'n' time periods.

Ex :- $P(x_3) = 1$ means probability of the state 1 after 3 time periods (or) 3 states.

Note :- formula to calculate the probability of states:

$$\pi_n = \pi_0 p^n \quad (\text{or}) \quad \pi_{n+1} = \pi_n p$$

In other notation; $a_n = a_0 p^n$ (or) $a_{n+1} = \pi_n p$

Here a_0 (or) π_0 is the initial probability and p is a state transition matrix [or transition probability matrix (TPM)]

1. A man either uses his car or takes a bus or a train to work each day. The transition probability matrix of the markov chain with these three states 1 (car), 2 (bus) and 3 (train) is

$$\begin{array}{c} \text{C} & \text{B} & \text{T} \\ \text{present} & \text{P} = & \begin{matrix} \text{C} \\ \text{B} \\ \text{T} \end{matrix} \\ (\text{today}) & & \begin{bmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{bmatrix} \end{array}$$

and the initial probability is $(0.7, 0.2, 0.1)$ calculate $P(x_2 = 3)$

Sol :- Given TPM 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 probability is $a_0 = \begin{bmatrix} 0.7 & 0.2 & 0.1 \end{bmatrix}$

$$\begin{bmatrix} 0.7 & 0.2 & 0.1 \end{bmatrix}$$

probability after 2 time periods is given by $a_0 P^2 = a_2$

$$\text{i.e., } \begin{bmatrix} 0.7 & 0.2 & 0.1 \end{bmatrix} \begin{bmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{bmatrix} = \begin{bmatrix} \text{probability after} \\ 2 \text{ periods} \end{bmatrix}$$

$$\text{Now } P^2 = \begin{bmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{bmatrix} \begin{bmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{bmatrix} = \begin{bmatrix} 0.43 & 0.37 & 0.26 \\ 0.24 & 0.42 & 0.34 \\ 0.36 & 0.35 & 0.29 \end{bmatrix}$$

$$\therefore [0.7 \ 0.2 \ 0.1] \begin{bmatrix} 0.43 & 0.31 & 0.26 \\ 0.24 & 0.42 & 0.34 \\ 0.36 & 0.35 & 0.29 \end{bmatrix} = [0.385 \ 0.336 \ 0.279]$$

Hence the probability of state '3' after 2 time periods is $P(x_2 = 3) = 0.279$ (or) 27.9 %

- Q. Three boys A, B & C are throwing a ball to each other. A always throws the ball to B and B always throws the ball to C, but C is as likely to throw the ball to B as to A. If the 'initial' probability distribution of three states A, B and C is $(0.3, 0.4, 0.3)$ respectively. Find
 i. Transition matrix ii. $p(x_2 = B)$, iii. $p(x_3 = B, x_2 = C, x_1 = B, x_0 = A)$

Sol i. using the given information, we can write the transition probability matrix as

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

The initial probabilities are $\alpha_0 = [0.3 \ 0.4 \ 0.3]$

ii. To find $p(x_2 = B)$:-

i.e.; The probability of B after 2 time periods.

we have $\alpha_n = \alpha_0 P^n$

$$\therefore \alpha_2 = \alpha_0 P^2$$

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

∴ probability after '2' states (or) 2 time periods is

$$\begin{bmatrix} 0.3 & 0.4 & 0.3 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 0.5 & 0.5 & 0 \\ 0 & 0.5 & 0.5 \end{bmatrix} = \begin{bmatrix} 0.20 & 0.35 & 0.45 \end{bmatrix}$$

$$\therefore p(x_2 = B) = 0.35$$

ii. $p(x_3 = B, x_2 = C, x_1 = B, x_0 = A)$:

$$q_{V_0} \xrightarrow{(1)} A \xrightarrow{(1)} B \xrightarrow{(1)} C \xrightarrow{(1)} B$$

$$\begin{aligned} p(x_3 = B, x_2 = C, x_1 = B, x_0 = A) &= q_{V_0}(A) \cdot P_{AB} \cdot P_{BC} \cdot P_{CB} \\ &= 0.3 \times 1 \times 1 \times \frac{1}{2} \\ &= 0.15 \end{aligned}$$

[where $q_{V_0}(A)$ is the initial probability of A and we can find the values P_{AB}, P_{BC}, P_{CB} from the TPM]

iii. Steady-state probability (or) equilibrium (or) limiting probabilities of boys:

The steady state probability can be given by

$$\begin{bmatrix} x & y & z \end{bmatrix} \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} x & y & z \end{bmatrix}$$

$$x + 1z = x \rightarrow ①$$

$$x + 3z = y \rightarrow ②$$

$$y = z \rightarrow ③ \quad \text{and} \quad x + y + z = 1 \rightarrow ④$$

Solving, we get $x = 0.20, y = 0.40, z = 0.40$

3. The TPM of the Markov chain with three states 1, 2, 3

is $P = \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix}$ and the initial probability is $(0.5, 0.3, 0.2)$ calculate $p(x_3=3, x_1=1)$

Sol :- $p(x_3=3, x_1=1)$

$$q_0 \xrightarrow{(2)} 1 \xrightarrow{(2)} 3$$

$$\Rightarrow q_0(1) \cdot p_{13}^{(2)} = 0.5 \times p_{13}^{(2)}$$

To find $p_{13}^{(2)}$: Probability of moving state 1 to state 3 in 2 state periods is given by P^2 .

$$\therefore P^2 = \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix} \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix} = \begin{bmatrix} 0.27 & 0.39 & 0.34 \\ 0.20 & 0.48 & 0.32 \\ 0.23 & 0.39 & 0.38 \end{bmatrix}$$

$\therefore p_{13}^{(2)} = 0.34$

$$\text{Hence } p(x_3=3, x_1=1) = (0.5) \times (0.34) = 0.17$$

4. The TPM of the Markov chain with three states 0, 1, 2 is

$$P = \begin{matrix} & 0 & 1 & 2 \\ 0 & [0.2 & 0.3 & \dots] \\ 1 & [\dots & 0.6 & 0.3] \\ 2 & [0.4 & \dots & 0.3] \end{matrix} \text{ and the initial probability is } (0.5, 0.3, 0)$$

to fill the blanks in TPM

$$\text{i}. p(x_3=2, x_2=1, x_1=0, x_0=2) \quad \text{iii}. p(x_3=2, x_1=0, x_0=2)$$

$$\text{iv}. p(x_2=2)$$

Sol :- Since $\sum p_{ij} = 1$, we write TPM 'p' as

$$P = \begin{matrix} & 0 & 1 & 2 \\ 0 & [0.2 & 0.3 & 0.5] \\ 1 & [0.1 & 0.6 & 0.3] \\ 2 & [0.4 & 0.3 & 0.3] \end{matrix} \text{ and } q_0 = [0.5 \ 0.3 \ 0.2]$$

$$\text{i}. p(x_3=2, x_2=1, x_1=0, x_0=2) :$$

$$q_0 \xrightarrow{(1)} 2 \xrightarrow{(1)} 0 \xrightarrow{(1)} 1 \xrightarrow{(1)} 2$$

3. The TPM of the Markov chain with three states 1, 2, 3 is $P = \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix}$ and the initial probability is $(0.5, 0.3, 0.2)$. Calculate $P(x_3=3, x_1=1)$

Sol :- $P(x_3=3, x_1=1)$

$$q_0 \rightarrow 1 \xrightarrow{(2)} 3$$

$$\Rightarrow q_0(1) \cdot p_{13}^{(2)} = 0.5 \times p_{13}^{(2)}$$

To find $p_{13}^{(2)}$: Probability of moving state 1 to state 3 in 2 state periods is given by p^2

$$\therefore P^2 = \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix} \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix} = \begin{bmatrix} 0.27 & 0.39 & 0.34 \\ 0.20 & 0.48 & 0.32 \\ 0.23 & 0.39 & 0.38 \end{bmatrix}$$

$$\therefore p_{13}^{(2)} = 0.34$$

Hence $P(x_3=3, x_1=1) = (0.5) \times (0.34) = 0.17$

4. The TPM of the Markov chain with three states 0, 1, 2 is

$$P = \begin{matrix} 0 & 1 & 2 \\ \begin{bmatrix} 0.2 & 0.3 & - \\ - & 0.6 & 0.3 \\ 0.4 & - & 0.3 \end{bmatrix} \end{matrix} \text{ and the initial probability is } (0.5, 0.3, 0.2)$$

i. fill the blanks in TPM

ii. $P(x_3=2, x_2=1, x_1=0, x_0=2)$ iii. $P(x_3=2, x_1=0, x_0=2)$

iv. $P(x_2=2)$

Sol :- Since $\sum p_{ij} = 1$, we write TPM 'p' as

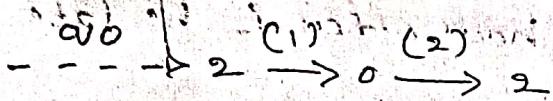
$$P = \begin{matrix} 0 & 1 & 2 \\ \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix} \end{matrix} \text{ and } q_0 = [0.5 \ 0.3 \ 0.2]$$

ii. $P(x_3=2, x_2=1, x_1=0, x_0=2)$:

$$q_0 \rightarrow 2 \xrightarrow{(1)} 0 \xrightarrow{(1)} 1 \xrightarrow{(0)} 2$$

$$= 0.2 \times 0.4 \times 0.3 \times 0.3 = 0.00172$$

iii. $P(x_3=2, x_1=0, x_0=2)$



$$= \alpha_{02}(2) \cdot P_{20} P_{02}^{(2)}$$

$$= 0.2 \times 0.4 \times P_{02}^{(2)}$$

To find $P_{02}^{(2)}$: (probability from state 0 to 2 after 2 time period)

$$P^2 = \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix} \begin{bmatrix} 0.2 & 0.3 & 0.5 \\ 0.1 & 0.6 & 0.3 \\ 0.4 & 0.3 & 0.3 \end{bmatrix} = \begin{bmatrix} 0.27 & 0.39 & 0.34 \\ 0.20 & 0.48 & 0.32 \\ 0.23 & 0.39 & 0.38 \end{bmatrix}$$

$$\therefore P_{02}^{(2)} = 0.34$$

$$\therefore P(x_3=2, x_1=0, x_0=2) = 0.2 \times 0.4 \times 0.34 \\ = 0.0272$$

iv. $P(x_2=2)$

i.e; The probability of '2' after '2' time periods

$$\Rightarrow \alpha_{22} = \alpha_{02} P^2$$

required probability is

$$\begin{bmatrix} 0.5 & 0.3 & 0.2 \\ 0.2 & 0.45 & 0.32 \\ 0.23 & 0.39 & 0.38 \end{bmatrix} \begin{bmatrix} 0.27 & 0.39 & 0.34 \\ 0.20 & 0.48 & 0.32 \\ 0.23 & 0.39 & 0.38 \end{bmatrix} = \begin{bmatrix} 0.241 & 0.408 \\ 0.342 & \end{bmatrix}$$

5. A professor tried not to be late for class too often, if he is late one day, he is 90% sure to be on time next day. If he is on time then the next time there is a 80% chance of his being late. In the long run, how often is he late for class?

Sol: - The required TPM is $P = \begin{bmatrix} L & O \\ O & L \end{bmatrix}$ | $L = \text{late}$ | $O = \text{on time}$

let $\alpha v = [x \ y]$ with $x+y=1$
be the stationary distribution such that $\alpha v P = \alpha v$

$$\text{i.e. } [x \ y] \begin{bmatrix} 0.1 & 0.9 \\ 0.3 & 0.7 \end{bmatrix} = [x \ y]$$

$$\Rightarrow (0.1)x + (0.3)y = x \Rightarrow -0.9x + 0.3y = 0 \rightarrow ①$$

$$(0.9)x + (0.7)y = y \Rightarrow 0.9x - 0.3y = 0 \rightarrow ②$$

$$\text{and } x+y=1 \rightarrow ③$$

solving ① & ③ we get

$$x = 0.25 \quad y = 0.75$$

Hence in the long run, there are 25% chances that he comes late for classes.

6. A man either drives his car or takes a train to work each day. suppose he never takes the train two days in a row, but he drives to work, then the next day he is just likely to drive again as he is to take the train. find transition matrix and also find the probability that he changes from going by train to driving exactly in four days?

Sol: - The required TPM is $P = \begin{bmatrix} T & C \\ C & T \end{bmatrix}$

because he will not travel by train for two days in a row.

The probability that he changes from going by train to driving exactly in four days is

$$P(x_4 = c | x_0 = T) = P_{TC}(4)$$

We have to compute P^4 and find the required probability P_{TC} .

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

$$P^4 = \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ -\frac{1}{4} & \frac{3}{4} \end{bmatrix} \begin{bmatrix} \frac{1}{2} & \frac{1}{2} \\ \frac{1}{4} & \frac{3}{4} \end{bmatrix} = \begin{bmatrix} \frac{3}{8} & \frac{5}{8} \\ \frac{5}{16} & \frac{11}{16} \end{bmatrix}$$

$$P^4 = \begin{matrix} T & C \\ \hline C & \begin{bmatrix} \frac{3}{8} & \frac{5}{8} \\ \frac{5}{16} & \frac{11}{16} \end{bmatrix} \end{matrix}$$

\therefore The required probability $P_{TC}(4)$ is $\frac{5}{8}$

4. In a certain market there are three brands of lipsticks A, B and C. Given that a lady last purchased lipstick of brand A, there is 70% chances that she would continue with brand A, 20% and 10% chances that she would shift to brands B and C respectively. Given that a lady last purchased lipstick of brand B, there is 50% chance that she would shift to brand A and 10% chances to brand C. Given that a lady last purchased lipstick brand C, there is 60%, 20% chance that she would shift to brand A and B respectively. The present market share of the three brands A, B & C is 60%, 30% & 10% respectively. Using this information find i. find transition probability matrix ii. probability that a customer who is currently a brand A purchases will purchase brand B after two time periods. iii. $P(x_2 = c | x_0 = A)$ iv. $P(x_2 = c | x_0 = c)$. v. probability that brand B purchase will purchase brand A three periods from now. vi. probability that a customer will purchase brand A, B, C three periods from now. vii. The market share of brands A, B, & C in the steady state (or) long run (or) equilibrium.

Sol :- i. The transition probability matrix P

$$P = \begin{bmatrix} 0.7 & 0.2 & 0.1 \\ 0.5 & 0.4 & 0.1 \\ 0.6 & 0.2 & 0.2 \end{bmatrix}$$

Given initial probabilities are $\pi_0 = [0.6 \ 0.1 \ 0.1]$

ii. required probability is $p(x_2 = B | x_0 = A) = P_{AB}^{(2)}$

$$\text{Now } P^2 = \begin{bmatrix} 0.7 & 0.2 & 0.1 \\ 0.5 & 0.4 & 0.1 \\ 0.6 & 0.2 & 0.2 \end{bmatrix} \begin{bmatrix} 0.7 & 0.2 & 0.1 \\ 0.5 & 0.4 & 0.1 \\ 0.6 & 0.2 & 0.2 \end{bmatrix}$$

$$= \begin{array}{c|ccc} & A & B & C \\ \hline A & 0.65 & 0.24 & 0.11 \\ B & 0.61 & 0.28 & 0.11 \\ C & 0.64 & 0.24 & 0.12 \end{array} \therefore P_{AB}^{(2)} = 0.24$$

iii. probability that a customer currently purchasing from A will purchase brand C after two time periods

$$\text{i.e., } p(x_2 = C | x_0 = A) = P_{AC}^{(2)} = 0.11$$

iv. probability that brand C will retain its customers after two time periods.

$$\text{i.e., } p(x_2 = C | x_0 = C) = P_{CC}^{(2)} = 0.12$$

v. probability that brand B purchases will purchase brand A three periods from now.

$$p(x_3 = A | x_0 = B) = P_{BA}^{(3)}$$

$$\text{Now } P^3 = P^2 \cdot P = \begin{bmatrix} 0.65 & 0.24 & 0.11 \\ 0.61 & 0.28 & 0.11 \\ 0.64 & 0.24 & 0.12 \end{bmatrix} \begin{bmatrix} 0.7 & 0.2 & 0.1 \\ 0.5 & 0.4 & 0.1 \\ 0.6 & 0.2 & 0.2 \end{bmatrix}$$

$$P^3 = \begin{matrix} & A & B & C \\ A & \left[\begin{matrix} 0.64 & 0.25 & 0.11 \end{matrix} \right] \\ B & \left[\begin{matrix} 0.63 & 0.26 & 0.11 \end{matrix} \right] \\ C & \left[\begin{matrix} 0.64 & 0.25 & 0.11 \end{matrix} \right] \end{matrix}$$

$$\therefore P_{BA}^{(3)} = 0.63$$

vii. probability that a customer will purchase brands A, B, C three periods from now:

required probability = $P(x_3 = A), P(x_3 = B), P(x_3 = C) \Rightarrow$

$$\left[\begin{matrix} 0.64 & 0.25 & 0.11 \end{matrix} \right] \cdot \left[\begin{matrix} 0.64 & 0.25 & 0.11 \\ 0.63 & 0.26 & 0.11 \\ 0.64 & 0.25 & 0.11 \end{matrix} \right] = \left[\begin{matrix} 0.637 & 0.253 & 0.11 \end{matrix} \right]$$

The market shares after 3 periods are:

$$\text{brand A} = 63.7\%$$

$$\text{brand B} = 25.3\%$$

$$\text{brand C} = 11\%$$

viii. Let $\pi = [x \ y \ z]$ be the stationary distribution such that

$$\pi P = \pi \text{ where } x + y + z = 1$$

$$\text{i.e. } \left[\begin{matrix} x & y & z \end{matrix} \right] \left[\begin{matrix} 0.7 & 0.2 & 0.1 \\ 0.5 & 0.4 & 0.1 \\ 0.6 & 0.2 & 0.2 \end{matrix} \right] = \left[\begin{matrix} x & y & z \end{matrix} \right]$$

$$\Rightarrow (0.7)x + (0.5)y + (0.6)z = x$$

$$(0.2)x + (0.4)y + (0.2)z = y$$

$$(0.1)x + (0.1)y + (0.2)z = z$$

$$\Rightarrow -0.3x + 0.5y + 0.6z = 0 \rightarrow ①$$

$$0.2x - 0.6y + 0.2z = 0 \rightarrow ②$$

$$0.1x + 0.1y - 0.8z = 0 \rightarrow ③$$

$$x + y + z = 1 \rightarrow ④$$

Solving ①, ②, ③ & ④ we get

$$x = 0.6389, y = 0.2500, z = 0.1111$$

Hence in the steady state, the market share of the brands A, B and C will be 63.89%, 25% & 11.11% respectively

H.W

Q. 80% of students who do maths work during one study period will do the same work at the next study period and 30% of students who do English work during one study period will do the English work in the next study period. Initially there were 60 students do maths and 40 students do English work. Calculate i. The transition probability matrix ii. The number of students who do maths work & English work for the next consequent 2 study periods.

Sol : i. State-Transition matrix is $\begin{bmatrix} 0.8 & 0.2 \\ 0.7 & 0.3 \end{bmatrix}$

Initial probabilities are (60, 40)

After one steady period

$$\begin{bmatrix} \text{expected distribution} \\ \text{in present study period} \end{bmatrix} \times \begin{bmatrix} \text{transition} \\ \text{matrix} \end{bmatrix} = \begin{bmatrix} \text{expected distribution} \\ \text{in next study period} \end{bmatrix}$$

$$\begin{bmatrix} 60 & 40 \end{bmatrix} \begin{bmatrix} 0.8 & 0.2 \\ 0.7 & 0.3 \end{bmatrix} = (76, 24)$$

After the steady period

$$(76, 24) \begin{bmatrix} 0.8 & 0.2 \\ 0.7 & 0.3 \end{bmatrix} = (77.6, 22.4)$$

Hence after two steady period.

The number of students who do maths & English works are 77.6% and 22.4%.

i. consider a bike share problem with only 3 stations A, B & C. Suppose that all bikes must be returned to the station at the end of the day, so that all the bikes are at some station each day, the distribution of bikes at each station changes as the bike get returned to different stations from where they are borrowed.

of the bikes borrowed from station A, 30% are return to station A, 50% end up at station B and 20% end up at station C.

of the bikes borrowed from station B, 10% end up at station A, 60% end up have been returned.

of the bikes borrowed from station C, 10% end up at station A, 10% end up at station B and 80% are returned to station C.

i. express this information as a transition probability matrix to determine the probabilities of bike being at a particular station after 2 days.

ii. suppose when we start observing the bike share program, 45% of the bikes at station A, 30% of the bikes are at station B and 25% are at station C, determine the distribution of bikes at the end of the next day & after 2 days.

Sol:- Transition matrix of A

$$P = \begin{bmatrix} 0.3 & 0.5 & 0.2 \\ 0.1 & 0.6 & 0.3 \\ 0.1 & 0.1 & 0.8 \end{bmatrix}$$

Transition probability matrix after 2 days

$$P^2 = \begin{bmatrix} 0.16 & 0.47 & 0.94 \\ 0.12 & 0.44 & 0.44 \\ 0.12 & 0.19 & 0.69 \end{bmatrix}$$

iii. Initial probabilities are $(0.3 : 0.45 : 0.25)$

$$(0.3 : 0.45 : 0.25) \begin{bmatrix} 0.3 & 0.5 & 0.2 \\ 0.1 & 0.6 & 0.3 \\ 0.1 & 0.1 & 0.8 \end{bmatrix} = (0.160, 0.445, 0.395)$$

∴ after 1 day 16% of bikes at station A, 44.5% at B, 39.5% at C.

After 2 days

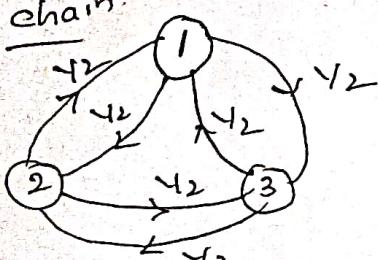
$$(0.160, 0.445, 0.395) \begin{bmatrix} 0.3 & 0.5 & 0.2 \\ 0.1 & 0.6 & 0.3 \\ 0.1 & 0.1 & 0.8 \end{bmatrix}^2 = (0.1320, 0.3865, 0.4815)$$

after 2 days, 13.20% of bikes at station A, 38.65% at B, 48.15% at C.

2. Three children denoted by 1, 2, 3 arranged in a circle to play a game of throwing a ball to one another. At each stage the child having the ball is equally likely to throw it to any one of the other 2 children. Suppose that x_0 denote the child who had the ball initially. and x_n denote the child who had the ball after n throws. $(n \geq 1)$ determine the transition probability matrix P . Calculate $\Pr\{x_2 = 1 | x_0 = 1\}$, $\Pr\{x_2 = 2 | x_0 = 3\}$, $\Pr\{x_2 = 3 | x_0 = 2\}$ and the probability that the child who had originally the ball will have it after 2 throws.

Sol:- Transition probability matrix $P =$

$$P = \begin{bmatrix} 1 & 2 & 3 \\ 0 & 0.5 & 0.5 \\ 0.5 & 0 & 0.5 \\ 0.5 & 0.5 & 0 \end{bmatrix}$$



$$\text{i. } \Pr \{ x_2 = 1 \mid x_0 = 1 \} = \frac{\Pr(x_2 = 1, x_0 = 1)}{\Pr(x_0 = 1)} =$$

Since the throws are equally likely at all stages, we have

$$\Pr(x_0 = 1) = \Pr(x_0 = 2) = \Pr(x_0 = 3) = \frac{1}{3} = 0.33$$

$$\Pr(x_1 = 1) = \Pr(x_1 = 2) = \Pr(x_1 = 3) = \frac{1}{3}$$

$$\Pr(x_2 = 1) = \Pr(x_2 = 2) = \Pr(x_2 = 3) = \frac{1}{3}$$

$$\therefore \Pr(x_0 = i), i = 1, 2, 3$$

$$\text{i. } \Pr \{ x_2 = 1 \mid x_0 = 1 \} = \frac{\Pr(x_2 = 1, x_0 = 1)}{\Pr(x_0 = 1)} = \frac{0}{\frac{1}{3}} = 0$$

$$\text{ii. } \Pr \{ x_2 = 2 \mid x_0 = 3 \} = \frac{\Pr(x_2 = 2, x_0 = 3)}{\Pr(x_0 = 3)} = \frac{0.5}{\frac{1}{3}} = 1.5$$

$$\text{iii. } \Pr \{ x_2 = 3 \mid x_0 = 2 \} = \frac{\Pr(x_2 = 3, x_0 = 2)}{\Pr(x_0 = 2)} = \frac{0.5}{\frac{1}{3}} = 1.5$$

probability that the child who had the ball ^{will} after 2 throws is given by the matrix P^2

$$\begin{bmatrix} 0 & 0.5 & 0.5 \\ 0.5 & 0 & 0.5 \\ 0.5 & 0.5 & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 & 0.5 & 0.5 \\ 0.5 & 0 & 0.5 \\ 0.5 & 0.5 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 0.5 & 0.25 & 0.25 \\ 0.25 & 0.5 & 0.25 \\ 0.25 & 0.25 & 0.5 \end{bmatrix}$$

3. The TPM of a markov process $\{x_n\}$ has 3 states 1, 2 & 3

is $P = \begin{bmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{bmatrix}$ with initial distribution $0.7, 0.2, 0.1$

i. respectively find i. $\Pr \{ x_3 = 2, x_2 = 3, x_1 = 3, x_0 = 2 \}$

ii. $\Pr \{ x_3 = 1 \mid x_2 = 2, x_0 = 2 \} [\Pr(x_3 = 1) | \Pr(x_2 = 2)]$

iii. $\Pr \{ x_2 = 3 \}$ probability of state 3 after 2 time period.

Sol:- Given Transition probability matrix P :

$$P = \begin{matrix} & 1 & 2 & 3 \\ 1 & \begin{bmatrix} 0.1 & 0.5 & 0.4 \end{bmatrix} \\ 2 & \begin{bmatrix} 0.6 & 0.2 & 0.2 \end{bmatrix} \\ 3 & \begin{bmatrix} 0.3 & 0.4 & 0.3 \end{bmatrix} \end{matrix}$$

and $v_0 = \begin{bmatrix} 0.7 & 0.2 & 0.1 \end{bmatrix}$

i. $P\{x_3=2, x_2=3, x_1=3, x_0=2\}$:

$$v_0 \rightarrow 2 \xrightarrow{(1)} 3 \xrightarrow{(2)} 3 \xrightarrow{(1)} 2$$

$$= v_0(2) P_{23} P_{33} P_{32}$$

$$= 0.2 \times 0.2 \times 0.3 \times 0.4 = 4.8 \times 10^{-3}$$

ii. $P\{x_3=1 | x_2=3, x_0=2\} = P(x_3=1 | x_2=3)$

$$\frac{P(x_3=1, x_2=3)}{P(x_2=3)} = \frac{0.04}{0.92} = 0.043 = v_0(3) P_{31}$$

$$= 0.1 \times 0.3 = 0.03$$

iii. $P(x_2=3)$

$$\begin{bmatrix} 0.7 & 0.2 & 0.1 \end{bmatrix} \begin{bmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{bmatrix} = \begin{bmatrix} 0.22 & 0.43 & 0.35 \end{bmatrix}$$

After 2 time period

$$\begin{bmatrix} 0.22 & 0.43 & 0.35 \end{bmatrix} \begin{bmatrix} 0.1 & 0.5 & 0.4 \\ 0.6 & 0.2 & 0.2 \\ 0.3 & 0.4 & 0.3 \end{bmatrix} = \begin{bmatrix} 0.385 & 0.336 & 0.22 \end{bmatrix}$$