3. Discrete Time Markov Chain

Discrete Time Markov Chain

Let $\{X_n, n \geq 0\}$ be a stochastic process taking values in a state space S that has N states. such a stochastic process is a Markov processes if it satisfies a following property:

$$P(X_{n+1}=k_{n+1}|X_n=k_n,X_{n-1}=k_{n-1},...X_1=k_1)=P(X_{n+1}=k_{n+1}|X_n=k_n)$$

For a markov process, the *future state* only depends on the *present state* and not on the *past states*.

If the state space of a Markov process is discrete, it's called a Markov Chain.

To understand the behaviour of this process, we will need to calculate probabilities like,

$$P[X_0 = i_0, X_1 = i_1, ..., X_n = i_n]$$

..(1)

 $:P(A,B) = P(A) \cdot P(B|A)$, this can be computed by multiplying conditional probabilities as follows.

$$=P(X_0=i_0)\cdot P(X_1=i_1|X_0=i_0)\cdot P(X_2=i_2|X_1=i_1,X_0=i_0)...$$

$$P(X_n = i_n | X_{n-1} = i_{n-1}, X_{n-2} = i_{n-2}, ..., X_0 = i_0)$$

..(2)

From the markovian property,

$$=P(X_0=i_0)\cdot P(X_1=i_1|X_0=i_0)\cdot P(X_2=i_2|X_1=i_1)...P(X_n=i_n|X_{n-1}=i_{n-1})...(3)$$
 ...(3)

State Transition Probabilities

For a discrete time Markov Chain $\{X_n:n=1,2,...\}$ with discrete state space $S=\{0,1,2,...\}$ where this set may be finite or infinite, if $X_n=i$ then the Markov Chain is said to be in state i at time n(or the nth step)

One Step Transition Probability

A discrete time Markov Chain $\{X_n: n=1,2,\ldots\}$ is characterized by

$$P[X_{n+1} = i_{n+1} | X_n = i_n, ..., X_0 = i_0] = P[X_{n+1} = i_{n+1} | X_n = i_n]$$

Where $P[X_{n+1} = j | X_n = i]$ is called one step transition probability

If $P[X_{n+1} = j | X_n = i]$ is independent of n then the Markov Chain is said to possess stationary transition probabilities and the process is reffered to as a homogeneous Markov Chain. Otherwise the process is called a non-homogeneous Markov Chain.

Transition Probability Matrix

The matrix called the state transition matrix $(\mathbf{t.p.m})$ or transition probability matrix is usually denoted by P.

Let $\{X_n:n=1,2,\ldots\}$ be a homogenous Markov Chain with a discrete finite state space $S=\{0,1,2,\ldots,m\}$ then

$$p_{ij} = P[X_{n+1} = j | X_n = i] \quad i \ge 0, j \ge 0$$

regardless of the value of n.

A t.p.m of $\{X_n\}$ is defined by

$$P = \begin{bmatrix} p_{ij} \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} & \dots & p_{1m} \\ p_{21} & p_{22} & \dots & p_{2m} \\ p_{31} & p_{32} & \dots & p_{3m} \\ \vdots & \ddots & & & \\ p_{m1} & p_{m2} & \dots & p_{mm} \end{bmatrix}$$

Where

$$p_{ij} \ge 0$$

and

$$\sum_{i=1}^{m} p_{ij} = 1, \quad i = 1, 2, ..., m$$

State Transition Diagram

A Markov Chain is usually shown by a state transition diagram. Consider a Markov Chain with three possible states $S=\{1,2,3\}$ and the following transition probabilities

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

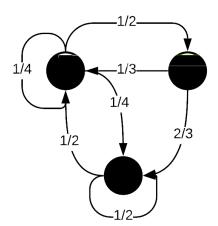
Which satisfies the two criterias, i.e.

$$p_{ij} \geq 0$$

and

$$\sum_{j=1}^{3} p_{ij} = 1, \quad i = 1, 2, 3$$

The figure below shows the state transition diagram for this Markov Chain



n-step Transition Probability

Consider a Markov Chain $\{X_n:n=0,1,2,...\}$ if $X_0=i$ then $X_1=j$ with probability p_{ij} is the probability of going from state i to state j in one step.

Now suppose we're interested in finding the probability of going from state i to state j in two steps, i.e.

$$p_{ij}^{(2)} = P(X_2 = j | X_0 = i)$$

$$p_{ij}^{(2)} = P(X_{n+2} = j | X_n = i)$$

We can find the probability by applying the law of total probability X_1 can take one of the possible values of S

$$\begin{split} p_{ij}^{(2)} &= P(X_2 = j | X_0 = i) \\ &= \sum_{k \in S} P(X_2 = j | X_1 = k, X_0 = i) \cdot P(X_1 = k | X_0 = i) \end{split}$$

[by law of total probability]

$$= \sum_{k \in S} P(X_2 = j | X_1 = k) \cdot P(X_1 = k | X_0 = i)$$

[by markovian property]

$$= \sum_{k \in S} p_{kj} \cdot p_{ik}$$
$$= \sum_{k \in S} p_{ik} \cdot p_{kj}$$

$$: p_{ij}^{(2)} = P(X_2 = j | X_0 = i) = \sum_{k \in S} p_{ik} \cdot p_{kj}$$

Which means that in order to get to state j from i, we need to pass through some intermediate state k.

$$i \longrightarrow k \longrightarrow j$$

n-step Transition Probability Matrix

We can define the ${f two-step}$ transition matrix as

$$P^{(2)} = \left[\begin{array}{cccc} p_{11}^{(2)} & p_{12}^{(2)} & \dots & p_{1m}^{(2)} \\ p_{21}^{(2)} & p_{22}^{(2)} & \dots & p_{2m}^{(2)} \\ p_{31}^{(2)} & p_{32}^{(2)} & \dots & p_{3m}^{(2)} \\ \vdots & \ddots & & & & \\ p_{m1}^{(2)} & p_{m2}^{(2)} & \dots & p_{mm}^{(2)} \end{array} \right]$$

We conclude that two-step transition matrix can be obtained by squaring the state transition matrix.

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

Similarly,

$$P^{(3)}=P\cdot P^2=P\cdot P^{(2)}$$

Generally we can define the *n*-step transition probability $p_{ij}^{(n)}$ as

$$p_{ij}^{(n)} = P(X_n = j | X_0 = i), \qquad i = 0, 1, 2, \dots$$

In order to get from state i to state j, we need to pass through n-1 intermediate states $k_1,k_2,...,k_{n-1}$

$$i \longrightarrow k_1 \longrightarrow k_2 \longrightarrow \dots \longrightarrow k_{n-1} \longrightarrow j$$

The n-step transition matrix is defined as follows

$$P^{(n)} = \left[\begin{array}{cccc} p_{11}^{(n)} & p_{12}^{(n)} & \dots & p_{1r}^{(n)} \\ p_{21}^{(n)} & p_{22}^{(n)} & \dots & p_{2r}^{(n)} \\ p_{31}^{(n)} & p_{32}^{(n)} & \dots & p_{3r}^{(n)} \\ \vdots & \ddots & & & & \\ p_{r1}^{(n)} & p_{r2}^{(n)} & \dots & p_{rr}^{(n)} \end{array} \right]$$

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

Let m and n be two positive integers and assume $X_0 = i$. In order to get to state j in (m+n) steps, the chain will be at some intermediate state k after m steps.

To obtain

$$\begin{split} p_{ij}^{(m+n)} &= P\left[X_{n+m} = j | X_0 = i\right] \\ &= \sum_{k \in S} p_{ik}^{(n)} \cdot p_{kj}^{(m)} \end{split}$$

This equation is called the Chapman–Kolmogorov Equation