

Lecture 2 Markov chain (cont)

Review of Previous Lecture

- Stochastic process $\left\{ \begin{array}{l} \text{index set} \\ \text{state space} \\ \text{dependence relation} \end{array} \right.$
- Discrete time and continuous time
- chain
- Markov property

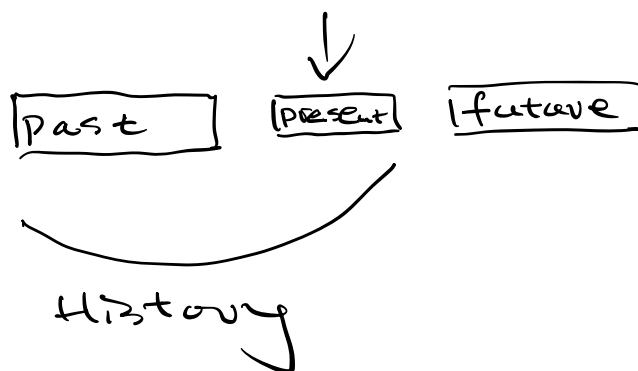
Assignment Project Exam Help

- Discrete time Markov chain

<https://powcoder.com>

$$P(X_{n+1}=j \mid X_0=x_0, \dots, X_n=i)$$

$$= P(X_{n+1}=j \mid X_n=i)$$



Example 1. Let $\{X_n : n=0,1,2,\dots\}$ be a process such that X_0, X_1, X_2, \dots are independent. Then $\{X_n : n=0,1,2,\dots\} \approx$ Markov.

Proof: For any $n \in \mathbb{T}$, $i_0, i_1, \dots, i_{n-1}, i, j \in S$,

$$P(X_{n+1}=j \mid X_0=i_0, \dots, X_{n-1}=i_{n-1}, X_n=i)$$

$$= \frac{P(X_0=i_0, \dots, X_n=i, X_{n+1}=j)}{P(X_0=i_0, \dots, X_n=i)}$$

Assignment Project Exam Help

$$= \frac{P(X_0=i_0, \dots, X_n=i) P(X_{n+1}=j)}{P(X_0=i_0, \dots, X_n=i)}$$

Add WeChat powcoder

$$= \frac{P(X_{n+1}=j)}{P(X_n=i)}$$

$$= \frac{P(X_n=i, X_{n+1}=j)}{P(X_n=i)}$$

$$= P(X_{n+1}=j \mid X_n=i)$$

Independent is a special Markov case.

Definition 1 If $P(X_{n+1}=j | X_n=i)$ is the same for all n , then the Markov chain has stationary transition probability.

$P(X_{n+1}=j | X_n=i)$, denoted by P_{ij} , is called the one-step transition probability from state i to state j .

Definition 2 the matrix

$P = (P_{ij})_{i,j \in S}$ is called the one-step transition matrix.

Definition 3 For each $i \in S$, define

$\pi_0(i) = P(X_0=i)$ and set

$\pi_0 = \{\pi_0(i) : i \in S\}$. clearly

$\sum_{i \in S} \pi_0(i) = 1$ and we call

π_0 the initial distribution of the Markov chain.

Proposition 1 Let $P = (P_{ij})_{i,j \in S}$ be two
 one-step stationary transition probability
 matrix. Then

① $P_{ij} \geq 0, i, j \in S$

② $\sum_{j \in S} P_{ij} = 1$ for all $i \in S$

Example 2: Determine which of the
 following are transition matrices.

(a) $\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$; (b) $\begin{pmatrix} 2 & 2 \\ 3 & 1 \end{pmatrix}$;

(c) $\begin{pmatrix} \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{8} & -\frac{1}{8} & 1 \\ \frac{1}{3} & 0 & \frac{2}{3} \end{pmatrix}$; (d) $\begin{pmatrix} \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \\ \frac{5}{8} & \frac{3}{8} & \frac{1}{16} \\ \frac{1}{7} & \frac{2}{7} & \frac{8}{14} \end{pmatrix}$

Given a discrete time Markov chain with stationary transition probability matrix P and initial distribution π_0 , One would like to know

<1> $P(X_n = j)$ for all n, j

the distribution at each time n

Assignment Project Exam Help

<2> $P(X_{n_1} = j_1, \dots, X_{n_k} = j_k)$ for
<https://powcoder.com>
all $n_1 < n_2 < \dots < n_k, j_1, \dots, j_k \in S$

Add WeChat powcoder

the joint distribution at times
 n_1, \dots, n_k

<3> $\lim_{n \rightarrow \infty} P(X_n = j)$ for all j

Asymptotic behaviour.

Proposition 2. For any $n \geq 1$, let P^n be the product of n P s, i.e.,

$$P^n = P \times \dots \times P$$

Denote element of P^n as P_{ij}^n . Then for any $j \in S$

$$P(X_n = j) = \sum_{i \in S} \pi_0(i) P_{ij}^n$$

Assignment Project Exam Help

where $P_{ij}^n = \sum_{i_1, \dots, i_{n-1} \in S} P_{i i_1} P_{i_1 i_2} \dots P_{i_{n-1} j}$

<https://powcoder.com>

Proof: we only consider cases $n=2$ or 3 .
the proof is similar for general n .

case 1: $n=2$

$$P(X_2 = j) = \sum_{i, i_1 \in S} P(X_2 = j, X_1 = i_1, X_0 = i)$$

Note:

$$P(A \cap B)$$

$$= P(A)P(B|A)$$

$$= \sum_{i \in S} \sum_{i_1 \in S} P(X_0 = i, X_1 = i_1, X_2 = j)$$

$$= \sum_{i \in S} \sum_{i_1 \in S} \left[P(X_2 = j | X_0 = i, X_1 = i_1) \cdot P(X_0 = i, X_1 = i_1) \right]$$

$$= \sum_{i \in S} \sum_{i_1 \in S} P(X_2=j | X_0=i, X_1=i_1) \cdot P(X_1=i_1 | X_0=i) P(X_0=i)$$

$$= \sum_{i \in S} \sum_{i_1 \in S} P(X_2=j | X_1=i_1) P(X_1=i_1 | X_0=i) \cdot P(X_0=i)$$

$$= \sum_{i \in S} \sum_{i_1 \in S} \pi_0(i) P_{i i_1} P_{i_1 j}$$

$$= \sum_{i \in S} \pi_0(i) \sum_{i_1 \in S} P_{i i_1} P_{i_1 j}$$

$$= \sum_{i \in S} \pi_0(i) P_{ij}^2$$

Add WeChat powcoder

Case 3 $n=3$

$$P(X_3=j) = \sum_{i \in S} \sum_{i_1 \in S} \sum_{i_2 \in S} \underline{P(X_0=i, X_1=i_1, X_2=i_2, X_3=j)}$$

$$\underline{P(X_0=i, X_1=i_1, X_2=i_2, X_3=j)}$$

$$= P(X_0=i) P(X_1=i_1 | X_0=i) P(X_2=i_2 | X_0=i, X_1=i_1)$$

$$\cdot P(X_3=j | X_0=i, X_1=i_1, X_2=i_2)$$

$$= P(X_0=i) P(X_1=i_1 | X_0=i) P(X_2=i_2 | X_1=i_1) \\ P(X_3=j | X_2=i_2)$$

$$= \pi_0(i) P_{i,i_1} P_{i_1,i_2} P_{i_2,j}$$

$$\Rightarrow P(X_3=j) = \sum_{i \in S} \pi_0(i) P_{i,j}^3$$

$$P_{i,j}^3 = \sum_{i_1 \in S} \sum_{i_2 \in S} P_{i,i_1} P_{i_1,i_2} P_{i_2,j}$$

<https://powcoder.com>

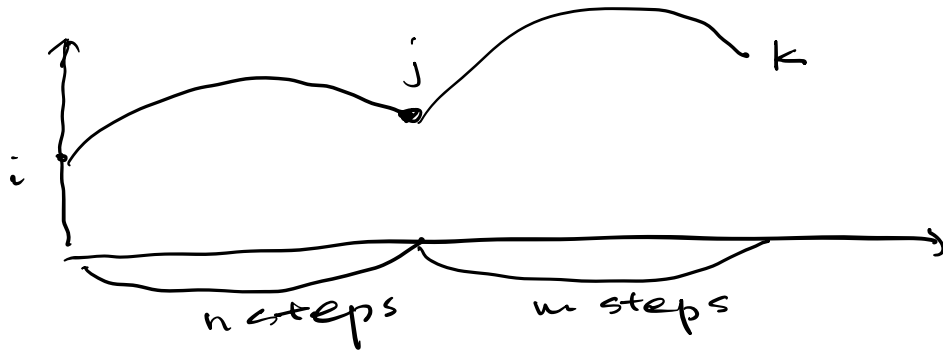
Proposition 3 For any k, n, m

$$P(X_{m+n}=k, X_n=j)$$

$$= \sum_{i \in S} \sum_{i_1, \dots, i_{n-1} \in S} \sum_{i_{n+1}, \dots, i_{m+n-1} \in S}$$

$$\pi_0(i) P_{i,i_1} \dots P_{i_{n-1},j} P_{j,i_{n+1}} \dots P_{i_{m+n-1},k}$$

$$= \sum_{i \in S} \pi_0(i) P_{i,j}^n P_{j,k}^m$$



Theorem. The Markov chain $\{X_n: n=0,1,\dots\}$ with initial distribution π_0 and one-step stationary transition probability matrix P is uniquely determined by (π_0, P) .

Add WeChat powcoder

Markov chain $\Leftrightarrow (\pi_0, P)$

Notation:

$$Y \Leftrightarrow S, \quad x, y \Leftrightarrow i, j$$

$$P(x, y) \Leftrightarrow P_{ij} \Leftrightarrow \underline{P(i, j)}$$

$$P_{ij}^n \Leftrightarrow P^n(i, j)$$