

(M) Markov chains - $\begin{cases} \text{Discrete time } n = 0, 1, 2, \dots \\ \text{State space } S \equiv \text{finite or countable.} \end{cases}$

History: • Andrey Markov - 1906 - Motivation

- Weak law of large numbers for i.i.d. sequences
(Does this hold without independence?)

Example: Eugene Omezin - Pushkin
alternation of consonants & vowels.

- Theory :- was challenging to develop
[Post] - A.N. Kolmogorov formulation of modern probability

• Wide application: $\begin{cases} \text{Mathematics} \\ \text{physics} \\ \text{Biology} \end{cases}$

Notation :-

S - denote the set of all possible state of the system.
($|S| < \infty$ or S -countable)

$N_0 = \mathbb{N} \cup \{0\}$ - denote time

$X_n \in S \equiv$ state of the system at time n .

Refer to as Markov chain
 \equiv (M.C.) or chain

Motivational Examples / Questions :

- $\{X_n\}_{n \geq 0}$:
- # of searches on a google server in an hour ?
 - # of tigers at Nagarhole national park ?



- At each time step n , the chain $\{X_n\}_{n \geq 1}$ moves from state $i \in S$ to another state $j \in S$ with probabilities P_{ij} [$i=j$ is also allowed]

Need :- $P = [P_{ij}]_{i \in S, j \in S}$ [one step transition matrix]

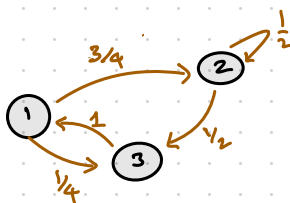


$$\sum_{j \in S} P_{ij} = 1, \quad P_{ij} \geq 0.$$

Example M1 :-

$$S = \{1, 2, 3\}$$

Intuitive way of defining M.C.



Directed, weighted graph

1 to 2	w.p.	$3/4$
1 to 3	w.p.	$1/4$
2 to 2	w.p.	$1/2$
2 to 3	w.p.	$1/2$
3 to 1	w.p.	1

Prescribe transitions

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

Transition matrix

let us define a Markov chain precisely.

Definition M-1 : let μ be a probability on S .

let $P = [p_{ij}]_{i,j \in S}$ be the matrix :

$$\forall i, j \in S \quad 0 \leq p_{ij} \leq 1 \quad \text{and} \quad \sum_{j \in S} p_{ij} = 1 \quad \forall i \in S.$$

let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space

$\mathbb{N} \ni n \geq 0$. let $X_n: \Omega \rightarrow S$ be a sequence of random variables whose

joint distribution is given by

$$\mathbb{P}(X_0 = i_0, X_1 = i_1, \dots, X_k = i_k)$$

$$= \mu(i_0) p_{i_0 i_1} p_{i_1 i_2} \dots p_{i_{k-1} i_k}$$

— (X)

$\forall k \geq 0$.

$\{X_n\}_{n \geq 0}$ is said to be a Markov chain on a state space S with initial distribution μ and transition matrix P .

⑥ (Markov Property) "Memoryless" \equiv $\begin{cases} \text{Past} \\ \text{future} \end{cases}$ are independent given present

Fix: $n \in \mathbb{N}$ $i_0, i_1, \dots, i_{n-2}, i, j \in S$. & assume

$$\mathbb{P}(X_0 = i_0, X_1 = i_1, \dots, X_{n-2} = i_{n-2}, X_{n-1} = i) > 0$$

$$\mathbb{P}(\underbrace{X_n = j}_{\text{future}} \mid \underbrace{X_0 = i_0, X_1 = i_1, \dots, X_{n-2} = i_{n-2}, X_{n-1} = i}_{\text{past}})$$

$$= \frac{\mathbb{P}(X_0 = i_0, X_1 = i_1, \dots, X_{n-2} = i_{n-2}, X_{n-1} = i, X_n = j)}{\mathbb{P}(X_0 = i_0, X_1 = i_1, \dots, X_{n-2} = i_{n-2}, X_{n-1} = i)}$$

$$= \frac{\mu(i_0) \prod_{k=1}^{n-2} p_{i_{k-1}, i_k} p_{i_{n-2}, i} p_{i, j}}{\mu(i_0) \prod_{k=1}^{n-2} p_{i_{k-1}, i_k} p_{i_{n-2}, i}}$$

$$= p_{i, j}$$

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

$$= \frac{\mathbb{P}\left(\bigcup_{\substack{i_k \in S \\ k=0, \dots, n-2}} \{X_0=i_0, X_1=i_1, \dots, X_{n-2}=i_{n-2}, X_{n-1}=i, X_n=j\}\right)}{\mathbb{P}\left(\bigcup_{\substack{i_k \in S \\ k=0, \dots, n-2}} \{X_0=i_0, X_1=i_1, \dots, X_{n-2}=i_{n-2}, X_{n-1}=i\}\right)}$$

$$= \frac{\sum_{\substack{i_k \in S \\ k=0, \dots, n-2}} \mu(\omega_t) \prod_{k=1}^{n-2} p_{i_{k-1}, i_k} p_{i_{n-2}, i} p_{i, j}}{\sum_{\substack{i_k \in S \\ k=0, \dots, n-2}} \mu(\omega_t) \prod_{k=1}^{n-2} p_{i_{k-1}, i_k} p_{i_{n-2}, i}}$$

$$= \frac{\sum_{\substack{i_k \in S \\ k=0, \dots, n-2}} \cancel{\mu(\omega_t)} \prod_{k=1}^{n-2} p_{i_{k-1}, i_k} p_{i_{n-2}, i}}{\sum_{\substack{i_k \in S \\ k=0, \dots, n-2}} \cancel{\mu(\omega_t)} \prod_{k=1}^{n-2} p_{i_{k-1}, i_k} p_{i_{n-2}, i}}$$

$$= p_{i, j}$$

We have shown:

$$\mathbb{P}(X_n=j \mid X_0=i_0, X_1=i_1, \dots, X_{n-2}=i_{n-2}, X_{n-1}=i)$$

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

(c) n th step Transition matrix

From (b), $n \geq 1$ $P(X_{n-1} = i) > 0$

$$\Rightarrow P(X_n = j \mid X_{n-1} = i) = P_{ij} \quad i, j \in S.$$

$P = [P_{ij}]$ - one step transition matrix

$n \geq 1$ $P(X_0 = i) > 0$

$$P(X_n = j \mid X_0 = i) = \frac{P(X_n = j)}{P(X_0 = i)}$$

$$= \frac{\sum_{\substack{i_k \in S \\ k=1, \dots, n-1}} P(X_0 = i, \dots, X_{n-1} = i_{n-1}, X_n = j)}{\mu(\{i\})}$$

$$= \frac{\sum_{\substack{i_k \in S \\ k=1, \dots, n-1}} \mu(\{i\}) p_{i i_1} \prod_{k=2}^{n-1} p_{i_{k-1}, i_k} p_{i_{n-1} j}}{\mu(\{i\})}$$

$$= \sum_{\substack{i_k \in S \\ k=1, \dots, n-1}} p_{i i_1} \prod_{k=2}^{n-1} p_{i_{k-1}, i_k} p_{i_{n-1} j}$$

observe $\equiv (i,j)^{\text{th}}$ entry of P^n

$$P(X_n=j \mid X_0=i) = P^n_{ij} \quad \text{--- } n^{\text{th}} \text{ step transition probability matrix.}$$

(d) (Chapman-Kolmogorov equation)

$$r, s, n \geq 1 \quad \& \quad r+s=n.$$

$$P^n = P^r P^s$$

Fact:
• $AB \leftarrow$ matrix multiplication
• holds for S countable

$$\Rightarrow P^n_{ij} = \sum_{k \in S} P^r_{ik} P^s_{kj}$$

\Rightarrow

$$P(X_n=j \mid X_0=i) = \sum_{k \in S} P(X_r=k \mid X_0=i) P(X_s=j \mid X_0=k)$$

i to j in steps \equiv i to $k \in S$ in r steps

AND
 $k \in S$ to j in s steps