

Recall:- \rightarrow Probability - III - Topics in RWS \leftarrow Review

- Examples of Markov chains $\dots \rightarrow$ Martingale Theory

- Student talks during the course.

[Fixed goal
- Short two
page proof
a result]

- Set up time in the next two weeks, so

that we can agree on a topic via mutual consent.

V - Countable set [finite / c'tly infinite]

$E = \{ \{i, j\} \mid i, j \in V \}$ - Edge set of the graph.

$P = (V, E)$

[weights] $\mu: E \rightarrow (0, \infty)$

$i, j \in V$: $\mu_{ij} = \begin{cases} \mu(\{i, j\}) & \{i, j\} \in E \\ 0 & \text{otherwise} \end{cases}$

[Symmetry]
 $\mu_{ij} = \mu_{ji}$

$\mu(A) = \sum_{i \in A} \mu_i$
- c'tly additive
if $\mu(V) = 1$ - Probability

Assumptions: $i \in V$ $\mu_i = \sum_{j \in V} \mu_{ij} = \sum_{\{i, j\} \in E} \mu(\{i, j\})$

$0 < \mu_i < \infty$

Transition Matrix $P_{v \times v}$

$$P_{ij} = \frac{\mu_{ij}}{\mu_i}$$

$i, j \in V$

• $p_{ij} \geq 0$, $\sum_{j \in V} p_{ij} = \frac{\sum_{j \in V} \mu_{ij}}{\mu_i} = \frac{\mu_i}{\mu_i} = 1$

Markov chain : $\Omega = \mathcal{V}^{\mathbb{Z}_+} = \{\omega : \mathbb{Z}_+ \rightarrow \mathcal{V}\}$, \exists

$X_n : \Omega \rightarrow \mathcal{V}$ $X_n(\omega) = \omega(n)$

$i \in \mathcal{V}$
 $n \geq 1$

$$\mathbb{P}^i (X_0 = i_0, \dots, X_n = i_n) = \underbrace{1}_{\text{focus } X_0 = i} \prod_{k=1}^n p_{i_{k-1} i_k}$$

[Existence, Unique, ... ← Reading material]

λ is a probability on \mathcal{V}
 $X_0 \stackrel{\text{d}}{=} \lambda$, $\mathbb{P}^\lambda (X_0 = i_0, \dots, X_n = i_n) = \lambda(\{i_0\}) \prod_{k=1}^n p_{i_{k-1} i_k}$ (+)

Ex:-

$$\mathbb{P} (X_n = j \mid \underbrace{X_{n-1} = i}_{\text{Present}}, \underbrace{X_{n-2} = i_{n-2}, \dots, X_0 = i_0}_{\text{Past}}) = \mathbb{P}(X_n = j \mid X_{n-1} = i) = p_{ij}$$

future independent of given Present

M.C on (\mathcal{P}, μ) : [Reversible]

$$\mu_i p_{ij} = \cancel{\mu_i} \frac{\mu_{ij}}{\cancel{\mu_i}} = \frac{\mu_{ji}}{\mu_j} \mu_j = \mu_j p_{ji}$$

Ex:- $\mu_{i_0} \mathbb{P}^{i_0} (X_0 = i_0, \dots, X_n = i_n) = \mu_{i_n} \mathbb{P}^{i_n} (X_0 = i_n, \dots, X_n = i_0)$

$$\mu(A) = \sum_{i \in A} \mu_i$$

$$\mu(\mathcal{V}) = 1$$

(*)

Assume \otimes [Stationary]

$$P^\mu(X_1=j) = \sum_{i \in V} P^\mu(X_0=i, X_1=j) \stackrel{\oplus}{=} \sum_{i \in V} \mu(i) p_{ij}$$

$$= \sum_{i \in V} \cancel{\mu_i} \frac{\mu_j}{\cancel{\mu_i}} = \sum_{i \in V} \mu_{ji} = \mu_j = P(X_0=j)$$

$$\therefore X_0 \stackrel{d}{=} \mu \Rightarrow X_1 \stackrel{d}{=} \mu$$

Ex: Induction: $X_n \stackrel{d}{=} \mu \quad \forall n \geq 1$
or otherwise

we have shown: $\left. \begin{aligned} \sum_{i \in V} \mu_i p_{ij} &= \mu_j \\ \mu P &= \mu \end{aligned} \right\} \otimes$

we assumed \otimes $\mu(v) = \sum_{i \in V} \mu_i = 1$:

If not: notice always \otimes

$\pi(A) := \frac{\mu(A)}{\mu(v)}$ π - will be a probability

$$X_0 \stackrel{d}{=} \pi \Rightarrow X_n \stackrel{d}{=} \pi \quad \forall n \geq 1$$

Example 1 :- $m \geq 2$.

$$E = \{ \{i, j\} \mid |i-j|=1, i, j \in S \}$$

(a) $S = \{0, 1, 2, \dots, m-1\}$

$\{X_n\}_{n \geq 1}$ M.C. on (S, E)

$$\mu: E \rightarrow (0, \infty)$$

$$\mu(\cdot) \equiv 1.$$

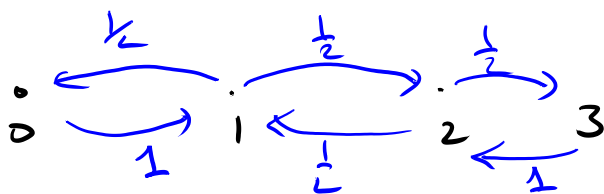
$m=4$:

$$\mu(S) = 2m$$

$$\mu_i = \begin{cases} 2 & i \neq 0, m-1 \\ 1 & i = 0, m-1 \end{cases}$$



$$p_{ij} = \frac{\mu_{ij}}{\mu_i}$$



X_n reflected random walk

π - stationary distributions ... comes from μ ... $\pi(\{i\}) = \begin{cases} \frac{1}{2m} & i=0, m-1 \\ \frac{2}{2m} & i \neq 0, m-1 \end{cases}$

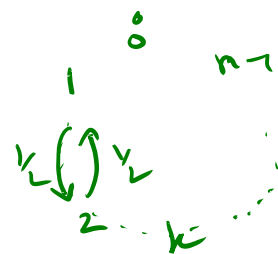
Ex:- irreducible, periodicity $\equiv 2$; Recurrent states $\equiv S$.

(b)

$$S = \mathbb{Z}_m$$

$$E' = \{0, m-1\} \cup E$$

$$p_{ij} = \begin{cases} \frac{1}{2} & i = j+1 \pmod{m} \\ \frac{1}{2} & i = j-1 \pmod{m} \\ 0 & \text{otherwise} \end{cases}$$



$\{X_n\}_{n \geq 1}$ M.C. with transition matrix P
Random walk on m -cycle

π -uniform(S)

Theorem - LIMIT :- $(\{X_n\}_{n \geq 0}, P)$ is a Markov chain S - finite / countable / infinite

• irreducible, stationary distribution π , aperiodic

$$P^n = [p_{ij}^n] = [P(X_n = j | X_0 = i)]$$

$$\forall j \in S. \quad \lim_{n \rightarrow \infty} p_{ij}^n = \pi(j)$$

Applications :- Convergence rate to stationarity $|p_{ij}^n - \pi(j)| = o(1)$

Distance between two Probabilities

(S, \mathcal{F}, P) and (S, \mathcal{F}, Q) two probability space

Total variation distance $\equiv \|P - Q\|_{TV} := \max_{A \in \mathcal{F}} |P(A) - Q(A)|$

Convergence rate

$|P(X_n \in A) - \pi(A)| \rightarrow 0$ at what rate?

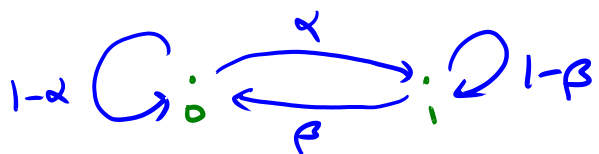
$|S| < \infty$ understand $\|P(X_n \in \cdot) - \pi(\cdot)\|_{TV} \rightarrow 0?$

Example 0 :-

$S = \{0, 1\}$

$0 < \alpha, \beta < 1$

$$P = \begin{bmatrix} 1-\alpha & \alpha \\ \beta & 1-\beta \end{bmatrix}$$



$\{X_n\}_{n \geq 1}$ m.c with transition matrix P on S

Stationary π . $\pi(\{0\}) = \beta/\alpha+\beta$
 $\pi P = \pi$

$\pi(\{1\}) = \frac{\alpha}{\alpha+\beta}$ [Ex.] - (XX)

Theorem-LIMIT:

X_n - aperiodic, irreducible, stationary
 $X_n \xrightarrow{d} \pi$ as $n \rightarrow \infty$

$P_{ij}^n \rightarrow \pi(\{j\})$ as $n \rightarrow \infty$

$\|P(X_n \in \cdot) - \pi(\cdot)\|_{TV} \rightarrow 0$ as $n \rightarrow \infty$

($\mathcal{S} = \{0,1\}$, $\mathcal{I} = \{\emptyset, \mathcal{S}, \{0\}, \{1\}\}$)

Ex:-
 $X_0 \stackrel{d}{=} \mu$

(J)

$P(X_n=0) = \beta/\alpha+\beta + (1-\alpha-\beta)^n (\mu(\{0\}) - \beta/\alpha+\beta)$

$P(X_n=1) = \frac{\alpha}{\alpha+\beta} + (1-\alpha-\beta)^n (\mu(\{1\}) - \frac{\alpha}{\alpha+\beta})$

$\|P(X_n \in \cdot) - \pi(\cdot)\|_{TV} = \max_{A \in \mathcal{I}} |P(X_n \in A) - \pi(A)|$

structure $\mathcal{I} = \max\{|P(X_n=0) - \pi(\{0\})|, |P(X_n=1) - \pi(\{1\})|\}$

(J), (XX) $\max\{|(1-\alpha-\beta)^n (\mu(\{0\}) - (\beta/\alpha+\beta))|, |(1-\alpha-\beta)^n (\mu(\{1\}) - \frac{\alpha}{\alpha+\beta})|\}$

$\mu(\{0\}) + \mu(\{1\}) = 1 \implies \|(1-\alpha-\beta)^n \cdot |\mu(\{0\}) - \beta/\alpha+\beta|\}$

Convergence to stationary is exponential as $0 < \alpha+\beta < 2$

Theorem - Convergence: $|S| < \infty$; $\{X_n\}_{n \geq 1}$, P irreducible, aperiodic, stationary

$\exists C > 0$ &
 $0 < \eta < 1$:

$$\sup \| P \cdot X_n^{-1} - \pi \|_{TV} \leq C \eta^n$$

μ - Probability on S

$$X_0 \stackrel{d}{=} \mu$$