

Lecture 15 :

Infinite State Spaces

Def 1. The matrix $L := P - I$ is called Laplace Matrix.

For the exit distribution $h(x) = P_x(V_A < V_B)$, it satisfies the Laplace Equation :

$$\left\{ \begin{array}{l} Lh = 0, \quad \text{on } C = X \setminus (A \cup B); \\ h = 1, \quad \text{on } A; \\ h = 0, \quad \text{on } B. \end{array} \right\} \text{Boundary Conditions}$$

For the exit time $g(x) = E_x T$, it solves the Poisson Equation :

$$\left\{ \begin{array}{l} Lg = -1 \quad \text{on } C; \\ g = 0 \quad \text{on } X \setminus C. \end{array} \right. \text{--- Boundary Conditions}$$

Def 2. A state $x \in X$ is said to be positive recurrent

if $E_x T_x < \infty$; is said to be null recurrent

if it is recurrent but not positive recurrent, i.e.,

$P_x(T_x < \infty) = 1$ but $E_x T_x = \infty$.

Ex 1. (Reflecting Random Walk on \mathbb{Z}). Let $\mathcal{X} = \{0, 1, 2, \dots\}$.

The transition probability is

$$P(X_1 = y | X_0 = x) = \begin{cases} p, & y = x+1; \\ 1-p, & y = x-1 \ (x > 0); \\ 1-p, & y = x=0; \\ 0, & \text{else.} \end{cases}$$

Here $0 < p < 1$. Thus, this chain is irreducible.

Case I. $0 < p < \frac{1}{2}$.

Notice

$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 & 3 & \dots \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \\ 3 \\ \vdots \end{matrix} & \begin{bmatrix} 1-p & p & 0 & 0 & \dots \\ 1-p & 0 & p & 0 & \dots \\ 0 & 1-p & 0 & p & \dots \\ 0 & 0 & 1-p & 0 & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix} \end{matrix}$$

Solving

$$\begin{cases} \vec{\pi} P = \vec{\pi} & \Rightarrow (\vec{\pi}_{i+1} = \frac{p}{1-p} \vec{\pi}_i) \\ \vec{\pi}_i \geq 0, \quad \forall i \in \mathbb{N} \\ \sum_{i=0}^{\infty} \vec{\pi}_i = 1 \end{cases}$$

gives

$$\vec{\pi}_i = \frac{1-2p}{1-p} \cdot \left(\frac{p}{1-p}\right)^i > 0, \quad \forall i \in \mathbb{N}.$$

Since $P_{00} > 0$, the state 0 has period 1.

Thus, 0 is aperiodic. Because this chain is irreducible, it is aperiodic.

Lemma 11.1 implies that the chain is recurrent.

i.e., $P_{ii} = P_i(\tau_i < \infty) = 1, \forall i \in \mathcal{N}$.

Theorem 9.3 implies that

$$\mathbb{E}_i \tau_i = \frac{1}{\pi_i} < \infty.$$

Case II. $\frac{1}{2} < p < 1$.

Let $h(x, N) := P_x(\tau_0 < \tau_N)$, then it satisfies

$$(*) \begin{cases} h(x, N) = (1-p)h(x-1, N) + ph(x+1, N), \forall 1 < x < N-1; \\ h(1, N) = (1-p) + ph(2, N); \\ h(N-1, N) = (1-p)h(N-2, N). \end{cases}$$

Solving this linear system of equations, one has

$$h(x, N) = 1 - \frac{1 - \theta^x}{1 - \theta^N}, \quad \forall 1 \leq x \leq N-1.$$

Here $\theta = \frac{1-p}{p} < 1$.

Taking $N \rightarrow \infty$ at both sides yields

$$P_x(\tau_0 < \infty) = \theta^x < 1, \quad \forall x \geq 1.$$

In particular, $P_1(\tau_0 < \infty) < 1$. Since 0 communicates with 1, we know 0 is transient. (Thm 1 in Lecture 5)

why?

← And thus the chain is transient.

Case III. $p = \frac{1}{2}$.

Let $h(x, N) := P_x(\tau_0 < N)$, then it satisfies (*).

Solving this linear system of equations, one has

$$h(x, N) = \frac{N-x}{N}, \quad \forall 1 \leq x \leq N-1.$$

Taking $N \rightarrow \infty$ at both sides yields

$$P_x(\tau_0 < \infty) = 1, \quad \forall x \geq 1.$$

Thus, $P_{00} = P_0(\tau_0 < \infty) = \frac{1}{2} + \frac{1}{2} \cdot P_1(\tau_0 < \infty) = 1$. This

implies 0 is recurrent and so is the chain.

For any fixed $N > 3$, let $T_N := \min \{n \geq 1 : X_n \in \{0, N\}\}$.

For any $1 \leq x \leq N-1$, define $g(x, N) = \mathbb{E}_x T_N$. Then it solves

$$\begin{cases} g(x, N) = 1 + \frac{1}{2}g(x-1, N) + \frac{1}{2}g(x+1, N), & \forall 1 < x < N-1; \\ g(1, N) = 1 + \frac{1}{2}g(2, N); \\ g(N-1, N) = 1 + \frac{1}{2}g(N-2, N). \end{cases}$$

Solving this linear system of equations, one has

$$g(x, N) = x(N-x), \quad \forall 1 \leq x \leq N-1.$$

Notice that for any $1 \leq x \leq N-1$, $T_N \leq T_0$. Thus,

$$\forall 1 \leq x \leq N-1, \quad \mathbb{E}_x T_0 \geq \mathbb{E}_x T_N = g(x, N) = x(N-x).$$

In particular, $\mathbb{E}_1 T_0 \geq N-1$, for any $N > 3$.

Taking $N \rightarrow \infty$ at both sides yields $\mathbb{E}_1 T_0 = \infty$.

Thus, $\mathbb{E}_0 T_0 = 1 + \frac{1}{2} \mathbb{E}_1 T_0 = \infty$.

In sum, if $p \in (\frac{1}{2}, 1)$, D is transient;

if $p = \frac{1}{2}$, D is null recurrent.

if $p \in (0, \frac{1}{2})$, D is positive recurrent.

Theorem 15.1 For an irreducible chain, the following are equivalent:

- (i). Some state is positive recurrent.
- (ii). There is a stationary distribution.
- (iii). All states are positive recurrent.

Proof. We will show (i) \Rightarrow (ii) \Rightarrow (iii) \Rightarrow (i).

(i) \Rightarrow (ii). Recall that (in the proof of Proposition 11.1)

the stationary measure $\vec{\mu}^x$ has a total mass

$$\vec{\mu}_y^x = \sum_{n=0}^{\infty} \mathbb{P}_x(X_n=y, \tau_x > n)$$

$$\begin{aligned} \sum_{y \in X} \vec{\mu}_y^x &= \sum_{y \in X} \sum_{n=0}^{\infty} \mathbb{P}_x(X_n=y, \tau_x > n) \\ &= \sum_{n=0}^{\infty} \sum_{y \in X} \mathbb{P}_x(X_n=y, \tau_x > n) \\ &= \sum_{n=0}^{\infty} \mathbb{P}_x(\tau_x > n) \\ &= \mathbb{E}_x \tau_x. \end{aligned}$$

Thus, if there exists x with $\mathbb{E}_x \tau_x < \infty$, then

$\frac{\vec{\mu}^x}{\sum_{y \in X} \vec{\mu}_y^x}$ defines a stationary distribution.

(ii) \Rightarrow (iii). Since $\sum_y \bar{\pi}_y = 1$, there exists at least one state, say x , s.t. $\bar{\pi}_x > 0$. For any fixed $y \in \mathcal{X}$, since the chain is irreducible, one has x communicates with y . That is, $\exists m$, s.t. $[P^m]_{xy} > 0$. Thus,

$$\bar{\pi}_y = [\bar{\pi} P^m]_y = \sum_{z \in \mathcal{X}} \bar{\pi}_z [P^m]_{zy} \geq \bar{\pi}_x [P^m]_{xy} > 0.$$

Therefore, $\bar{\pi}_y > 0$ for any $y \in \mathcal{X}$.

Thm 9.3

If I & S hold, then

$$\bar{\pi}_y = \frac{1}{\mathbb{E}_y T_y}, \quad \forall y \in \mathcal{X}.$$

Theorem 9.3 implies

$$\mathbb{E}_y T_y = \frac{1}{\bar{\pi}_y}, \quad \forall y \in \mathcal{X}.$$

Thus, $\mathbb{E}_y T_y < \infty$, $\forall y \in \mathcal{X}$.

(iii) \Rightarrow (i). Trivial. \square

why?

Remark 13.1. Positive recurrent states are recurrent, i.e.,

for any $x \in \mathcal{X}$ s.t. $\mathbb{E}_x T_x < \infty$, one has $\mathbb{P}_x(T_x < \infty) = 1$.

This is the end of this lecture!