

Chapter II Stationary Distributions

1 SD and its computations

- Recall that for a MC $\{X_n\}_{n=0}^\infty$,

$$\vec{\pi}_{n+1} = \vec{\pi}_n P, \quad \vec{\pi}_n = \vec{\pi}_0 P^n, \quad n = 0, 1, 2, \dots \quad (1)$$

where $\vec{\pi}_n$, $n \geq 0$, denote the p.d.f. of X_n .

- Consider a MC with P and S (for instance, $S = \{0, 1, 2, \dots, N\}$ with N finite or infinite). $\vec{\pi} := [\pi(0), \pi(1), \dots, \pi(N)]$, or denoted by $\pi(x)$, $x \in S$, is called a **stationary distribution** for P if

(i) $\vec{\pi}$ is a distribution, i.e., $\pi(x) \geq 0$, $\forall x \in S$, and $\sum_{x \in S} \pi(x) = 1$.

(ii) $\vec{\pi}$ is stationary: $\vec{\pi}P = \vec{\pi}$, i.e.,

$$\sum_{x \in S} \pi(x) P(x, y) = \pi(y), \quad \forall y \in S. \quad (2)$$

Here, (ii) means that if the chain starts from the distribution $\vec{\pi}$, then all X_n , $n \geq 1$ have the same distributions as $\vec{\pi}$.

- We have to notice:

(a) Given an initial distribution $\vec{\pi}_0$, if the limit distribution exists, i.e., $\lim_{n \rightarrow \infty} \vec{\pi}_0 P^n$ exists, denoted by $\vec{\pi}$, then $\vec{\pi}$ satisfies

$$\vec{\pi} = \left(\lim_{n \rightarrow \infty} \vec{\pi}_0 P^{n-1} \right) \cdot P = \vec{\pi}P, \quad (3)$$

i.e., the limit distribution $\vec{\pi}$ is stationary and hence $\vec{\pi}$ is a SD. Moreover, if $\vec{\pi} = \vec{\pi}P$ has a unique distribution solution then the limit distribution is independent of the initial distribution.

(b) If

$$\lim_{n \rightarrow \infty} P^n = \begin{pmatrix} \vec{\pi} \\ \vec{\pi} \\ \vdots \\ \vec{\pi} \end{pmatrix} \quad (4)$$

for some distribution $\vec{\pi}$, then the limit distribution exists and is independent of the initial distribution. We will discuss the long-term behavior of P^n in the last subsection.

- In case S is finite, we have some general conditions to assure the existence and uniqueness. In fact, let P be a Markov matrix with **finite** state space S . Assume

- (i) the left 1-eigenvector (which must exist; *why?*) can be chosen to have all nonnegative entries.

- (ii) 1 is a simple eigenvalue.

- (iii) all other eigenvalues: $|\lambda_i| < 1$.

Then P has a unique SD $\vec{\pi}$, and (4) holds true. In particular, **if for some n , P^n has all entries strictly positive**, then three conditions above can be satisfied and the conclusion is true for the chain.

In the future lecture, we will show that *an irreducible MC with finite state space must have a unique SD* (but (4) may NOT hold true!).

- In the general situation that S is finite or infinite, we will discuss the existence and uniqueness of SD later on.

- Computation issues on SD, as well as the limit of P^n **if it exists**:

- In case S is finite and P is irreducible, apply Row Operators to $P^T - I$ to get the upper diagonal form.

- In case S is finite and P is reducible, apply the State Decomposition, for instance, $S = C_1 \cup C_2 \cup S_T$, re-write P as the canonical form, and then try to find the limit of P^n as $n \rightarrow \infty$, **if it exists**. See the tutorial and exercises for examples.

- In case S is infinite, see the lectures for two additional examples:

- (a) Find SD of an irreducible birth and death chain.

- (b) Find SD of a telephone exchange model with new calls satisfying the Poisson distribution (or a general queuing chain model with the service given by the rule that each person at the beginning of a unit time has the probability q to be served and leave the waiting line by the end of the unit time).

2 Average number of visits

- Given a MC with S and P , let $N_n(y)$ be the NO of visits to y in n -steps (i.e., during times $m = 1, 2, \dots, n$). We are interested in determining

$$\frac{N_n(y)}{n}, \quad \frac{E_x(N_n(y))}{n}, \quad \text{as } n \rightarrow \infty. \quad (5)$$

Note:

- (i) $\frac{N_n(y)}{n}$ is a r.v., denoting the **proportion of the first n units of time that the chain visits y** , and the limit of $\frac{N_n(y)}{n}$ as $n \rightarrow \infty$ (if exists) means the **average NO of visits to y (per unit time)** or the **frequency** that the chain visits y . We can compute $N_n(y)$ as

$$N_n(y) = \sum_{m=1}^n 1_y(X_m). \quad (6)$$

- (ii) $\frac{E_x(N_n(y))}{n}$ is the **expected value of $\frac{N_n(y)}{n}$ for a chain starting from x** , and hence its limit value (if exists) means the **expected average NO of visits to y (per unit time)** or the **expected frequency** that the chain visits y . We can compute $E_x(N_n(y))$ as

$$E_x(N_n(y)) = \sum_{m=1}^n P^m(x, y). \quad (7)$$

Thus, to determine $\lim_{n \rightarrow \infty} \frac{E_x(N_n(y))}{n}$ is equivalent to determine

$$\lim_{n \rightarrow \infty} \frac{\sum_{m=1}^n P^m(x, y)}{n}. \quad (8)$$

Note that it could occur that the above limit exists but $\lim_{n \rightarrow \infty} P^n(x, y)$ may not exist!

- In case y is transient, it is direct to see

$$\lim_{n \rightarrow \infty} N_n(y) = N(y) < \infty \text{ with prob 1, } \lim_{n \rightarrow \infty} E_x(N_n(y)) = \frac{\rho_{xy}}{1 - \rho_{yy}} < \infty, \quad (9)$$

and hence

$$\lim_{n \rightarrow \infty} \frac{N_n(y)}{n} = 0 \text{ with prob 1, } \lim_{n \rightarrow \infty} \frac{E_x(N_n(y))}{n} = 0. \quad (10)$$

This means that in the long run, the average NO of visits to a transient state is zero, and its expected value is also zero.

- In case y is recurrent, we can show the following result. For simplicity we consider an irreducible recurrent MC only. Then, for any $y \in S$,

$$\lim_{n \rightarrow \infty} \frac{N_n(y)}{n} = \frac{1}{m_y} \text{ with prob 1, } \lim_{n \rightarrow \infty} \frac{E_x(N_n(y))}{n} = \frac{1}{m_y}, \quad x \in S, \quad (11)$$

where $m_y := E_y(T_y)$ denotes the **mean return time to y for a chain starting from y** . m_y can be understood to be the **mean waiting time**. Thus, two limits mean that *the visit frequency and the waiting time are reciprocal to each other!!!* It is heuristically obvious; see the lectures for the rigorous proof.

3 Waiting time and existence of stationary distribution

- $0 < m_x := E_x(T_x) \leq \infty$ for a recurrent state x . Note: If x is recurrent, then $P_x(T_x = \infty) = 0$ and $P_x(T_x < \infty) = 1$, so there is $k_0 \geq 1$ such that $P_x(T_x = k_0) > 0$, hence $m_x = E_x(T_x) = \sum_{k=1}^{\infty} k P_x(T_x = k) \geq k_0 P_x(T_x = k_0) > 0$.

- A recurrent state x is called **positive recurrent** if $(0 <) m_x < \infty$, and **null recurrent** if $m_x = \infty$. Thus, a positive recurrent state comes back *in finite waiting time*, and a null recurrent state comes back *very rarely*.

- We can also discuss communications between positive recurrent states. In fact, one can prove that *if a positive recurrent state x leads to y then y is also a positive recurrent state*.

Recall that an irreducible MC with finite state space is recurrent. One can further show that *an irreducible MC with finite state space does not admit any null recurrent state*, and hence it is positive recurrent.

Recall that given S and P , we have the state decomposition

$$S = S_R \cup S_T = (\cup_{i=1}^k C_i) \cup S_T, \quad (12)$$

where k can be finite or infinite. Then, for each i , C_i is either positive recurrent or null recurrent. Moreover, if C_i is finite, then C_i must be positive recurrent.

- The waiting time m_x of a recurrent state x , or the frequency $1/m_x$ of the chain visiting x , would be connected with the stationary solution of the chain. In fact, one can show that *an irreducible positive recurrent MC has a unique stationary distribution $\vec{\pi}$, given by*

$$\pi(x) = \frac{1}{m_x} \in (0, 1), \quad x \in S. \quad (13)$$

Notice that the theorem gives us a way to find the value of waiting time m_x of any state x . Here are a few immediate consequences:

- An irreducible MC with finite state space has a unique SD $\vec{\pi}$ with $\pi(x) = 1/m_x$, $x \in S$.
- We may further show that *if a general MC has no positive recurrent state (i.e., any state is either null recurrent or transient), then the chain has no SD*. Therefore, *for an irreducible MC, it has a SD if and only if it is positive recurrent*. Exercise: Apply it to determine if a general irreducible birth and death chain is either positive recurrent, or null recurrent, or transient; see the criterion discussed in the lecture.
- Let C be an irreducible closed set of positive recurrent states. Then, the chain has

a unique SD $\vec{\pi}$ **concentrated on** C :

$$\pi(x) = \begin{cases} \frac{1}{m_x} & \text{if } x \in C, \\ 0 & \text{otherwise.} \end{cases} \quad (14)$$

Moreover, if $\{C_i\}_{i=1}^N$ is a collection of irreducible closed set consisting of positive recurrent states, then $\pi := \sum_{i=1}^N \lambda_i \pi_i$ is a SD of the chain for any $0 \leq \lambda_i \leq 1$ with $\sum_{i=1}^N \lambda_i = 1$, where each π_i is the unique SD concentrated on C_i .

4 Periodicity

• Recall that it could occur that the chain admits a SD but $\lim P^n$ does not exist (hence the long-term behavior of the chain seems unclear!). For instance,

$$P = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}. \quad (15)$$

The SD exists, given by $\vec{\pi} = [1/2, 1/2]$. For such P you can compute

$$P^{2m} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \quad P^{2m+1} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}. \quad (16)$$

Thus, $\lim P^n$ does not exist, but you can still determine the long-term behavior of the chain in the following way

$$\lim_{m \rightarrow \infty} P^{2m} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}, \quad \lim_{m \rightarrow \infty} P^{2m+1} = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}. \quad (17)$$

We can discuss such property by using the periodicity of the chain.

• The **period** d_x of a state $x \in S$ is defined by

$$d_x = g.c.d. \{n \geq 1 : P^n(x, x) > 0\}. \quad (18)$$

Note that d_x is a positive integer with $1 \leq d_x \leq \min\{n \geq 1 : P^n(x, x) > 0\}$. If $P(x, x) > 0$ then $d_x = 1$.

For the chain with P given by (15),

$$\{n \geq 1 : P^n(0, 0) > 0\} = \{2, 4, 6, \dots\} = \{n \geq 1 : P^n(1, 1) > 0\}. \quad (19)$$

Thus,

$$d_0 = d_1 = 2. \quad (20)$$

• For an irreducible MC, $A_x := \{n \geq 1 : P^n(x, x) > 0\}$ must be non-empty, thus $d_x = g.c.d. A_x$ is finite, and it further holds that all states have the same period $d \geq 1$ (see

the lecture for the proof), called the period of the irreducible MC. If $d = 1$, the chain is said to be **aperiodic**.

- We can make connection between the **long-term behavior of $P^n(x, y)$** and SD $\vec{\pi}$ in the following way (the proof was omitted in the lecture; please refer to the textbook). Consider an irreducible positive recurrent MC. We know such chain must have a SD, denoted by $\vec{\pi}$. Then, we have

(a) if the chain is aperiodic, then

$$\lim_{n \rightarrow \infty} P^n(x, y) = \pi(y), \quad (21)$$

for any $x, y \in S$.

(b) if the chain is periodic with period $d \geq 2$, then for any $x, y \in S$, there exists an integer

$$r \in \{0, 1, 2, \dots, d-1\},$$

generally depending on x, y , such that

$$P^n(x, y) = 0 \quad (22)$$

for all n except that $n = md + r$ ($m \geq 0$ is an integer) for which

$$\lim_{m \rightarrow \infty} P^{md+r}(x, y) = d\pi(y). \quad (23)$$

This result tells that in case $d \geq 2$, we are able to determine the limits of subsequences

$$P^{md}, \quad P^{md+1}, \dots, P^{md+(d-1)} \quad (24)$$

as $m \rightarrow \infty$. Precisely, for any given x, y ,

$$P^{md}(x, y), \quad P^{md+1}(x, y), \dots, P^{md+(d-1)}(x, y) \quad (25)$$

are zeros except that exactly one of them tends to $d\pi(y)$ as $m \rightarrow \infty$.

—End, Updated on March 17—