

## 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

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

and hence

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

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  $y \in S$ ,

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

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:

(a) An irreducible MC with finite state space has a unique SD  $\vec{\pi}$  with  $\pi(x) = 1/m_x$ ,  $x \in S$ .

(b) We may further show that *if an irreducible chain 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 an irreducible birth and death chain is either positive recurrent, or null recurrent, or transient.

(c) Let  $C$  be an irreducible closed set of positive recurrent states. Then, the chain has

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

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

## 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, all states have the same period  $d \geq 1$  (*see the lecture for the proof*), and the chain is called **periodic** with period  $d \geq 1$ . 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 of Chapter 2, Updated on March 18—