

Chapter 4. Markov chain: Limits and stationarity

4.1. Stationary distribution

Suppose X_0 has a distribution π ,i.e.,

$$\mathbb{P}(X_0 = i) = \pi(i), \qquad i \in S.$$

Then,

$$\mathbb{P}(X_1 = i) = \sum_{j} \mathbb{P}(X_1 = i | X_0 = j) \mathbb{P}(X_0 = j)
= \sum_{j} \pi(j) p(j, i) = (\pi P)(i).$$



Definition

 π is a stationary distribution if $\pi P = \pi$.

Example: (Social mobility)

$$P = \left[\begin{array}{rrr} .7 & .2 & .1 \\ .3 & .5 & .2 \\ .2 & .4 & .4 \end{array} \right]$$

Find its stationary distribution.

Solution: Do it on board!



Theorem

Suppose that the $k \times k$ transition matrix is irreducible. Then, there is a unique solution to $\pi P = \pi$ with $\sum_{i=1}^k \pi_i = 1$. Further, $\pi_i > 0$, $\forall i$.

Proof: Omit



Proposition

All states y s.t. $\pi(y) > 0$ are recurrent.



4.2. Special examples

a) Double stochastic chains

Definition

Transition matrix P is *double stochastic* if its each column sums to 1,i.e.

$$\sum p(x,y) = 1, \qquad \forall \ y.$$

Theorem

If P is double stochastic with N states, then $\pi(x) = \frac{1}{N}$, $\forall x$ is a stationary distribution.



Example

Symmetric reflecting random walk on $\{0, 1, \dots, L\}$.

Take L=4.

$$P = \begin{bmatrix} .5 & .5 & 0 & 0 & 0 \\ .5 & 0 & .5 & 0 & 0 \\ 0 & .5 & 0 & .5 & 0 \\ 0 & 0 & .5 & 0 & .5 \\ 0 & 0 & 0 & .5 & .5 \end{bmatrix}$$

It is clearly double stochastic.

b) Detailed balance condition.

Definition

A distribution π satisfies the detailed balance condition if

$$\pi(x)p(x,y) = \pi(y)p(y,x), \quad \forall x,y \in S.$$



Proposition

If π satisfies the DBC, then π is a stationary distribution.



Example

Consider a random walk on $0, 1, \dots, 9$ given by

$$p(i, i+1) = \frac{2}{3}, \ p(i, i-1) = \frac{1}{3}, \qquad 1 \le i \le 8.$$

$$p(0,1) = 1, \quad p(9,8) = 1.$$

Does the DBC hold?

Solution: Do it on board!



Example

$$P = \begin{bmatrix} .5 & .5 & 0 \\ .3 & .1 & .6 \\ .2 & .4 & .4 \end{bmatrix}$$

Does the DBC hold?

Solution: Do it on board!



c) Reversibility.

Let p(i,j) be a transition probability with stationary distribution $\pi(i)$, $i \in S$.

Let X_n be the MC starting from the stationary distribution,

$$\mathbb{P}(X_0 = i) = \pi(i).$$

Let n be fixed and

$$Y_m = X_{n-m}, \qquad 0 \le m \le n$$

(look at the chain backward). Namely, we consider the process

$$X_n, X_{n-1}, \cdots, X_1, X_0.$$



Theorem

 (Y_m) is a MC with transition probability

$$\hat{p}(i,j) = \mathbb{P}(Y_{m+1} = j | Y_m = i) = \frac{\pi(j)p(j,i)}{\pi(i)}.$$



Definition

The MC is reversible if $\hat{p}(i,j) = p(i,j), \forall i,j \in S$.

Proposition

 (X_n) is reversible iff DBC holds.

Proof:

$$\hat{p}(i,j) = p(i,j) \Leftrightarrow \frac{\pi(j)p(j,i)}{\pi(i)} = p(i,j)$$

$$\Leftrightarrow \pi(j)p(j,i) = \pi(i)p(i,j).$$

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4.3. Limit behavior

Condition:

(I): P is irreducible.

(R): All states are recurrent.

(S): There is a stationary distribution π .



Recall: If y recurrent, then $N(y) = \infty$ a.s. Let

$$N_n(y) = \sum_{i=1}^n 1_{X_i = y}$$

be the # of visits to y before time n. How fast it goes to ∞ ?

Theorem

$$\lim_{n \to \infty} \frac{N_n(y)}{n} = \frac{1}{\mathbb{E}_y T_y}, \qquad a.s.$$



As a consequence, we have

Theorem

Under (I) and (S), we have

$$\pi(y) = \frac{1}{\mathbb{E}_y T_y},$$

and hence, the stationary distribution is unique.



Finally, we generalize Theorem ??.Recall

$$\frac{1}{n}\sum_{i=1}^{n}1_{X_i=y}\to\pi(y).$$

Rewrite

$$\frac{1}{n}\sum_{i=1}^{n}f(X_i)\to\sum_{x}f(x)\pi(x)$$

with
$$f(x) = 1_{x=y}$$
.

Theorem

Under (I), (S) and $\sum_{x} |f(x)| \pi(x) < \infty$, we have

$$\frac{1}{n}\sum_{m=1}^{n}f(X_m)\to\sum_{x}f(x)\pi(x),\qquad a.s.$$

Namely, long term average equals spatial average, i.e., ergodicity.



Example (Inventory chain)

Electronic store that sells a game system, with a potential for sales of 0, 1, 2, 3 of these units each day with prob. .3, .4, .2, .1.Each night, the store will bring inventory up to 3 items if it is x or less. Suppose the store makes \$12 for each unit sold and cost \$2 a day if it is stored over night. Which x is the best?

Solution: Do it on board!

4.4. Infinite state spaces

Example (Reflecting random walk)

A particle moves on $\{0, 1, 2, \dots\}$ with prob. p to the right and 1 - p to the left(at 0, it stays at 0 with prob. 1 - p instead of jumping to the left).

Solution:

Transition:

$$\begin{cases} p(i, i+1) = p, & i \ge 0, \\ p(i, i-1) = 1 - p, & i \ge 1, \\ p(0, 0) = 1 - p. \end{cases}$$

Stationary distribution π : Do it on board! (p < .5)



Next, we consider the case of p > .5.

Proposition

Let

$$V_x = \inf\{n \ge 0: \ X_n = x\}.$$

If $p \neq .5$ and $\theta = \frac{p}{1-p} \neq 1$, then

$$\mathbb{P}_x(V_N < V_0) = \frac{\theta^{-x} - 1}{\theta^{-N} - 1}.$$



We come back to our example for p > .5. Do it on board! Finally, we consider p = .5. Do it on board!



Proposition

Suppose p = .5. Then,

$$\mathbb{E}_x(T_0 \wedge T_N) = x(N - x), \qquad x \neq 0, N.$$



Note that

$$\mathbb{E}_x(T_0 \wedge T_N) = x(N - x), \qquad x \neq 0, N.$$

So

$$\mathbb{E}_1 T_0 = \lim_{N \to \infty} (N - x) = \infty.$$

Thus,

$$\mathbb{E}_0 T_0 = \frac{1}{2} \times 1 + \frac{1}{2} \mathbb{E}_1 (1 + T_0) = \infty.$$

Take a long time to come back to 0.



Definition

A recurrent state x is positive recurrent if $\mathbb{E}_x T_x < \infty$; it is null recurrent if $\mathbb{E}_x T_x = \infty$.

$$0 \text{ is } \begin{cases} \text{ positive recurrent } & p < .5 \\ \text{null recurrent } & p = .5 \\ \text{ transient } & p > .5 \end{cases}$$

Let $\mu_x(y)$ be the expected # of visits to y before returning to the initial position x.Namely

$$\mu_x(y) = \mathbb{E}_x \sum_{0 \le n \le T_x - 1} 1_{X_n = y}$$
$$= \sum_{n=0}^{\infty} \mathbb{P}_x(X_n = y, T_x > n).$$

Moving one-step in time (i.e. $1 \le n \le T_x$), the # of visits remains the same.

Theorem

Suppose (I) and (R) hold. Then, μ_x is a finite stationary measure.

Proof: Omit.

Recall

$$\pi(x) = \frac{1}{\mathbb{E}_x T_x} > 0$$
 if x is positive recurrent.

Theorem

For any irreducible chain the following are equivalent:

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



Example (Branching processes)

Recall $X_n = \#$ of individuals in nth generation. $Y_i = \#$ of children of ith individual.

$$\mathbb{P}(Y_i = k) = p_k, \quad k = 0, 1, 2, \cdots.$$

$$\mu = \mathbb{E}Y_i = \sum_{k=0}^{\infty} k p_k.$$

Find extinction probability.

Solution: Do it on board!



Lemma

 ρ is the smallest root of $x = \phi(x), \ 0 \le x \le 1$.



Example (Binary branching)

$$p_0 = 1 - a,$$
 $p_2 = a.$

Do it on board!



Proposition

If $\mu > 1$, then $\rho < 1$ (there is a chance of avoiding extinction).

Proof: Do it on board!

HW: Ch4, 18, 20, 28, 36, 42, 46, 56, 64, 74.