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Subject: Stochastic process

Assignment: Homework Eight

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1. EXERCISE 4.10

(a)

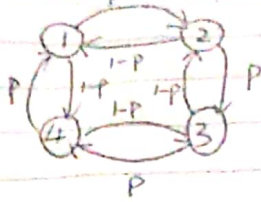


figure (a), we can get the transformation matrix

$$\begin{cases} \pi_1 = \pi_2(1-P) + \pi_4 P \\ \pi_2 = \pi_3(1-P) + \pi_1 P \\ \pi_3 = \pi_4(1-P) + \pi_2 P \\ \pi_4 = \pi_1(1-P) + \pi_3 P \end{cases} \Rightarrow \pi_1 = \pi_2 = \pi_3 = \pi_4 = \frac{1}{4}$$

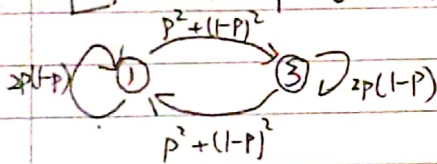
For figure (b), we know state 4 has two direction: ① to state 1 ② to state 5
 $P_{41} = P_{51}$, we conclude that $P_{41} = P_{51} = \frac{1}{2}$, for others, they are all 1

$$\begin{cases} \pi_4 = \pi_1 \times 1 + \pi_3 \times 1 \\ \pi_5 = \pi_6 \times \frac{1}{2} \\ \pi_1 = \pi_4 \times \frac{1}{2} \\ \pi_9 = \pi_8 = \pi_7 = \pi_6 = \pi_5 \\ \pi_3 = \pi_2 = \pi_1 \end{cases} \Rightarrow \sum \pi_i = 1$$

$$\Rightarrow \pi_4 = \frac{1}{5}, \pi_1 = \pi_2 = \pi_3 = \pi_5 = \pi_6 = \pi_7 = \pi_8 = \pi_9 = \frac{1}{10}$$

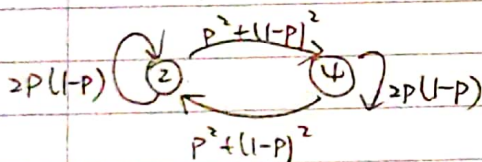
(b) For the first figure

$$\begin{bmatrix} 0 & P & 0 & 1-P \\ 1-P & 0 & P & 0 \\ 0 & 1-P & 0 & P \\ P & 0 & 1-P & 0 \end{bmatrix} \begin{bmatrix} 0 & P & 0 & 1-P \\ 1-P & 0 & P & 0 \\ 0 & 1-P & 0 & P \\ P & 0 & 1-P & 0 \end{bmatrix} = \begin{bmatrix} 2P(1-P) & 0 & P^2+(1-P)^2 & 0 \\ 0 & 2P(1-P) & 0 & P^2+(1-P)^2 \\ P^2+(1-P)^2 & 0 & 2P(1-P) & 0 \\ 0 & P^2+(1-P)^2 & 0 & 2P(1-P) \end{bmatrix}$$



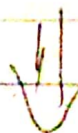
$$\pi_1 = \pi_3 = \frac{1}{2}$$

$$\Rightarrow \text{or } \pi_2 = \pi_4 = \frac{1}{2}$$

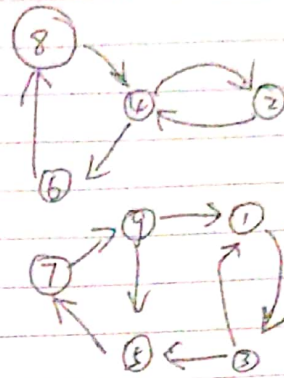


For the second figure

$$\begin{bmatrix} 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \end{bmatrix}$$



$$\begin{bmatrix} 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \\ 0 & \frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ \frac{1}{2} & 0 & 0 & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \end{bmatrix}$$



$$(1) \pi_2 = \pi_6 = \pi_8 = \frac{1}{5}, \pi_4 = \frac{2}{5}$$

or

$$\pi_1 = \pi_3 = \pi_5 = \pi_7 = \pi_9 = \frac{1}{5}$$

(c) According to part (b), for the first figure,

$$\lim_{n \rightarrow \infty} [P^{2n}] = \lim_{n \rightarrow \infty} \begin{bmatrix} 2P(1-P) & 0 & P^2+(1-P)^2 & 0 \\ 0 & 2P(1-P) & 0 & P^2+(1-P)^2 \\ P^2+(1-P)^2 & 0 & 2P(1-P) & 0 \\ 0 & P^2+(1-P)^2 & 0 & 2P(1-P) \end{bmatrix} = \begin{bmatrix} \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} \end{bmatrix}$$

$$\begin{cases} \pi_1 = 2P(1-P)\pi_1 + [P^2+(1-P)^2]\pi_3 \\ \pi_2 = 2P(1-P)\pi_2 + [P^2+(1-P)^2]\pi_4 \\ \pi_3 = [P^2+(1-P)^2]\pi_1 + 2P(1-P)\pi_3 \\ \pi_4 = [P^2+(1-P)^2]\pi_2 + 2P(1-P)\pi_4 \\ \sum_{i=1}^4 \pi_i = 1 \end{cases}$$

$$\Rightarrow 2P(1-P) = P^2+(1-P)^2$$

$$P = \frac{1}{2}$$

For the second figure

$$\lim_{n \rightarrow \infty} [P^{2n}] = \begin{bmatrix} \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} \\ 0 & \frac{1}{5} & 0 & \frac{2}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 \\ \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} \\ 0 & \frac{1}{5} & 0 & \frac{2}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 \\ \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} \\ 0 & \frac{1}{5} & 0 & \frac{2}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 \\ \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} \\ 0 & \frac{1}{5} & 0 & \frac{2}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 \\ \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} & 0 & \frac{1}{5} \end{bmatrix}$$

similar like (b)



2.

- (a) Through the above information from question, we can find this is aperiodic.
 $\Pr(X_{1000}=j, X_{1001}=k, X_{2000}=k \mid X_0=i)$ (using the definition of conditional)
 $= \Pr(X_{2000}=k, X_{1001}=k \mid X_{1000}=j, X_0=i) \cdot \Pr(X_{1000}=j \mid X_0=i)$
 $= \Pr(X_{2000}=k \mid X_{1001}=k, X_{1000}=j, X_0=i) \cdot \Pr(X_{1001}=k \mid X_{1000}=j, X_0=i) \cdot \Pr(X_{1000}=j \mid X_0=i)$
 Then we can simplify that like this
 $= \Pr(X_{2000}=k \mid X_{1001}=k) \cdot \Pr(X_{1001}=k \mid X_{1000}=j) \cdot \Pr(X_{1000}=j \mid X_0=i)$
 $= P_{kk}^{999} \cdot P_{jk}^{999} \cdot P_{ij}$

- (b) For part (b), we can find it is posterior probability,
 Using the bayes law, we get

$$\Pr(X_{1000}=i \mid X_{1001}=j) = \frac{\Pr(X_{1001}=j \mid X_{1000}=i) \cdot \Pr(X_{1000}=i)}{\Pr(X_{1001}=j)}$$

$$= \frac{\Pr(X_{1001}=j) \cdot \Pr(X_{1000}=i)}{\sum_{m=1}^n \Pr(X_{1001}=j \mid X_{1000}=m) \cdot \Pr(X_{1000}=m)}$$

Because it is recurrent class

and aperiodic, it must satisfy ergodic

Owing to the definition of ergodic chain, we can get $P_{kj} = P_{ij}$, for any k .

So, the result is
$$\frac{P_{ij} \cdot \pi_i}{n \cdot P_{ij}} = \frac{\pi_i}{n}$$

3. (a) For this process, we can know X_n can be two different directions, which can simplify like that

$$X_n = \begin{cases} X_{n-1} + 1, & \text{with probability of } p. \\ 0, & \text{with probability of } (1-p). \end{cases}$$

And then, we can get the transformation equation ($\pi = \pi[P]$)

$$\begin{cases} \pi_1 = \pi_0 p \\ \pi_2 = \pi_1 p = \pi_0 p^2 \\ \vdots \\ \pi_i = \pi_{i-1} p = \pi_0 p^i \end{cases} \Rightarrow \text{using } \sum_{i=0}^{\infty} \pi_i = 1$$

$$\pi_0 = \frac{1}{1-p} \quad \text{so, } \pi_i = \frac{p^i}{1-p}$$

- (b) Followed the statement in part (b)

T_k means the first time to get state k , so there are two possible conditions: ① it stays at state $(k-1)$, and next step is to state k with probability of p ② it stays at state $(k-1)$ but next step



is to redirect to state 0, and then a new process to get state k .

$$T_k = \begin{cases} T_{k-1} + 1 & \text{with probability of } p \\ T_{k-1} + 1 + T_k' & \text{with probability of } (1-p). \end{cases}$$

And then, we can get $E[T_k | T_{k-1}] = p \cdot (T_{k-1} + 1) + (1-p) [T_{k-1} + 1 + E[T_k' | T_{k-1}]]$
 $= T_{k-1} + 1 + (1-p) E[T_k' | T_{k-1}]$

T_k' is a renewal process like T_k , so $T_k' \stackrel{d}{=} T_k$

$$E[T_k' | T_{k-1}] = E[T_k | T_{k-1}]$$

Then we can get $p E[T_k | T_{k-1}] = T_{k-1} + 1$

use the total expectation $E[p E[T_k | T_{k-1}]] = E[T_{k-1} + 1]$

$$p \cdot E[T_k] = E[T_{k-1}] + 1$$

As desired. For T_1 $T_1 = \begin{cases} 1, & \text{with probability of } p \\ 1 + T_1', & \text{with probability of } (1-p) \end{cases}$

$$E[T_1] = 1 \cdot p + E[1 + T_1'] \cdot (1-p)$$

$$= p + 1 + E[T_1'] - p - p E[T_1']$$

$$p E[T_1] = 1 \quad (E[T_1'] = E[T_1])$$

$$E[T_1] = \frac{1}{p}$$

(c) Follow part (b), we can get $\begin{cases} p \cdot E[T_k] = E[T_{k-1}] + 1 \\ E[T_1] = \frac{1}{p} \end{cases}$

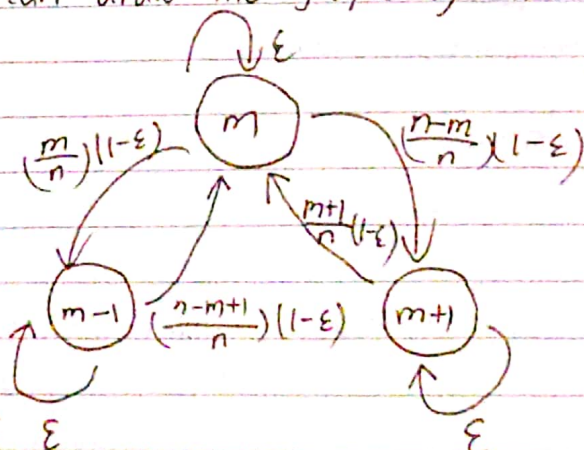
$$E[T_2] = \frac{1}{p^2} + \frac{1}{p}$$

$$E[T_3] = \frac{1}{p^3} + \frac{1}{p^2} + \frac{1}{p}$$

\vdots

$$E[T_k] = \frac{1}{p^k} + \frac{1}{p^{k-1}} + \dots + \frac{1}{p} = \sum_{i=1}^k \left(\frac{1}{p}\right)^i$$

4. We can draw the graph of this markov chain



we suppose state m means there are m white ball
we can get this balance equation

$$\pi_i \frac{n-i}{n} = \pi_{i+1} \cdot \frac{i+1}{n}$$

$$\pi_{i+1} = \frac{n-i}{i+1} \pi_i$$



Follow this balance equation, we can get below equation

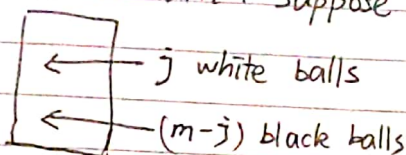
$$\begin{cases} \pi_0 = \frac{n}{1} \pi_0 \\ \pi_1 = \frac{n-1}{2} \pi_0 = \frac{(n-1)n}{2 \times 1} \pi_0 \\ \vdots \\ \pi_i = \frac{n!}{(n-i)! i!} \pi_0 = \binom{n}{i} \pi_0 \end{cases} \Rightarrow \sum_{i=0}^n \pi_i = 1$$

And then we can get $\sum_{i=0}^n \binom{n}{i} 1^{n-i} \cdot 1^i = (1+1)^n \pi_0 = 1$

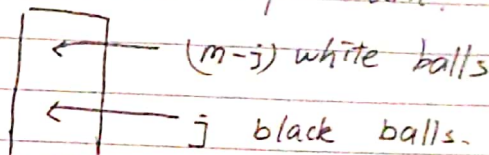
$$\text{so } \pi_0 = \left(\frac{1}{2}\right)^n$$

$$\pi_i = \binom{n}{i} \left(\frac{1}{2}\right)^n$$

5. With the above information in this question five, we should know the white balls in the first urn is image condition of the second urn. Suppose j white balls in the first urn.



First urn



Second urn

For state j , there are three conditions.

- ① it can add one white ball, and then becomes state $(j+1)$
- ② it can keep same, so state j is steady
- ③ it can minus one white ball, and then becomes state $(j-1)$

$$P_{j,j+1} = \left(\frac{m-j}{m}\right) \cdot \left(\frac{m-j}{m}\right) = \left(\frac{m-j}{m}\right)^2$$

$$P_{j,j} = \left(\frac{j}{m}\right) \left(\frac{m-j}{m}\right) + \left(\frac{m-j}{m}\right) \left(\frac{j}{m}\right) = \frac{2j(m-j)}{m^2}$$

$$P_{j,j-1} = \left(\frac{j}{m}\right) \cdot \left(\frac{j}{m}\right) = \frac{j^2}{m^2}$$

According to the transformation, we can get balance equation

$$\pi_j \left(\frac{m-j}{m}\right)^2 = \pi_{j+1} \left(\frac{j+1}{m}\right)^2 \quad j=0, 1, \dots, m-1$$

Follow this equation, we can get

$$\begin{cases} \pi_1 = \frac{m^2}{1^2} \pi_0 \\ \pi_2 = \left(\frac{m-1}{2}\right)^2 \pi_1 = \left(\frac{m-1}{2}\right)^2 \left(\frac{m}{1}\right)^2 \pi_0 \\ \vdots \end{cases} \Rightarrow \text{using } \sum_{j=0}^m \pi_j = 1$$

$$\pi_j = \left(\frac{m-j+1}{j}\right)^2 \pi_{j-1} = \left(\frac{m}{j}\right)^2 \pi_0$$

$$\sum_{j=0}^m \binom{m}{j}^2 = \binom{2m}{m}$$

$$\text{So } \pi_0 = \frac{1}{\binom{2m}{m}}$$

$$\pi_j = \left(\frac{m}{j}\right)^2 \cdot \frac{1}{\binom{2m}{m}}$$

