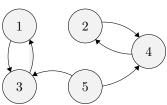
## STAT 433: HOMEWORK 3

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**Problem 1.** Consider the following transition matrices. Identify the transient and recurrent states, and the irreducible closed sets in the Markov chains.

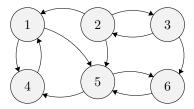
(a)	1	2	3	4	5	(b)	1	<b>2</b>	3	4	<b>5</b>	6
(a)	1			4		1	.1	0	0	.4	.5	0
1	.4	.3	.3	0	0	<b>2</b>	.1	.2	.2	0	.5	0
2	0	.5	0	.5	0	3	0	.1	.3	0	0	.6
3	.5	0	.5	0	0	4	.1	0	0	.9	0	0
<b>4</b>	0	.5	0	.5	0	5	0	0	0	.4	0	.6
5	0	.3	0	.3	.4	6	0	0	0	0	.5	.5
						U	U	U	U	U	.5	.5
						(d)	1	9	2	4	5	G
(c)	1	2	3	4	5	(d)	1	2	3	4	5	6
$\begin{pmatrix} c \end{pmatrix}$ 1	<b>1</b> 0	<b>2</b> 0	<b>3</b> 0	<b>4</b> 0	<b>5</b>	1	.8	0	0	.2	0	0
1	0	0	0	0	1	1 2	.8 0	0 .5	0	.2 0	0 .5	0
1 2	$0 \\ 0$	0.2	0	0 .8	1 0	1	.8	0	0	.2	0	0
1 2 3	0 0 .1	0 .2 .2	0 0 .3	0 .8 .4	1 0 0	1 2	.8 0	0 .5	0	.2 0	0 .5	0
1 2	$0 \\ 0$	0.2	0	0 .8	1 0	1 2 3	.8 0 0	$\begin{array}{c} 0 \\ .5 \\ 0 \end{array}$	0 0 .3	.2 0 .4	0 .5 .3	0 0 0

(a) Solution. The directed graph representation is



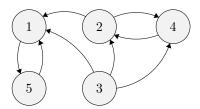
Since  $1 \to 2$  but  $2 \not\to 1$ , state 1 is transient. Since  $5 \to 4$  but  $4 \not\to 5$ , state 5 is transient. Since  $3 \to 2$  but  $2 \not\to 3$ , state 3 is transient. So, states 2 and 4 are recurrent. The closed irreducible set is  $\{2,4\}$ .

(b) Solution. The directed graph representation is



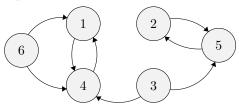
Since  $2 \to 1$  but  $1 \not\to 2$ , state 2 is transient. Since  $3 \to 6$  but  $6 \not\to 3$ , state 3 is transient. So states 1, 4, 5, and 6 are recurrent. The closed irreducible set is  $\{1, 4, 5, 6\}$ .

(c) Solution. The directed graph representation is



Since  $3 \to 1$  but  $1 \not\to 3$ , state 3 is transient. So state 1, 2, 4, and 5 are recurrent. The closed irreducible sets are  $\{1,5\}$  and  $\{2,4\}$ .

(d) Solution. The directed graph representation is



Since  $3 \to 4$  but  $4 \not\to 3$ , state 3 is transient, and since  $6 \to 1$  but  $1 \not\to 6$ , state 6 is transient. So state 1, 2, 4, and 5 are recurrent. The closed irreducible sets are  $\{1,4\}$  and  $\{2,5\}$ .

**Problem 2.** Let G be a connected graph. Let  $X_n$  be a simple random walk on G. Show that the Markov chain  $\{X_n\}$  is irreducible. (Hint: for any two vertices x and y in G, consider a path of consecutive nodes from x to y.)

Solution. Let x and y be any two vertices in G. Since G is connected, x and y are connected by with a path of edges formed by

$$x = x_0, x_1, x_2, \dots, x_n = y,$$

which implies that

$$\rho_{xy} \ge \frac{1}{\deg(x)} \cdot \frac{1}{\deg(x_1)} \cdots \frac{1}{\deg(x_{n-1})} > 0$$

$$\rho_{yx} \ge \frac{1}{\deg(y)} \frac{1}{\deg(x_{n-1})} \cdots \frac{1}{\deg(x_1)} > 0.$$

Therefore  $x \leftrightarrow y$ .

**Problem 3.** Let G be a graph with two disjoint components  $G_1$  and  $G_2$ . Let  $X_n$  be a simple random walk on G.

- (i) Prove  $\{X_n\}$  is not an irreducible Markov chain.
  - Solution. Since  $G_1 \cap G_2 = \emptyset$ , there does not exist an edge between some node  $x \in G_1$  and some other node  $y \in G_2$ , which shows  $G_1$  and  $G_2$  are disconnected. So  $x \not\to y$  and  $X_n$  on G is not irreducible.  $\square$
- (ii) Let P be the transition matrix of  $X_n$ . Let  $P_1$  and  $P_2$  be the transition matrices for the SRW on  $G_1$  and  $G_2$ , respectively. Let  $V_1 = \{1, \ldots, k\}$  and  $V_2 = \{k+1, \ldots, k+\ell\}$  be the set of vertices in  $G_1$  and  $G_2$ . Show that P is of the following block diagonal form

$$P = \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix}. \tag{1}$$

Note a block diagonal matrix is a reducible matrix.

Solution. We have p(x,y)=0 if  $1 \le x \le k$  and  $k+1 \le y \le k+\ell$ , or if  $k+1 \le x \le k+\ell$  and  $1 \le y \le k$ , which shoes that P is diagonal. We also have  $p(x,y)=p_1(x,y)$  if  $1 \le x,y,\le k$  and  $p(x,y)=p_2(x,y)$  if  $k+1 \le x,y,\le k+\ell$ , which establishes the fact that  $P=\operatorname{diag}(P_1,P_2)$ .

(iii) Show the the SRW  $\{X_n\}$  on G has infinitely many stationary distributions.

Solution. Since the disjoint graphs  $G_1$  and  $G_2$  are finite,  $P_1$  has a stationary distribution, call it  $\pi_1$ , and  $P_2$  has stationarity distribution, call it  $\pi_2$ . Note that  $\pi_1$  has dimension  $1 \times k$ ,  $\pi_2$  has dimension  $1 \times \ell$ . Let

$$\pi = \begin{cases} (\pi_1, 0) & \text{if } X_0 \in G_1\\ (0, \pi_2) & \text{if } X_0 \in G_2 \end{cases}$$

be the stationary distribution for the operator P on G, and note that  $\pi$  has dimension  $1 \times k + \ell$ . Note also that P has dimension  $(k + \ell) \times (k + \ell)$ . Now, for the chain started on any vertex in  $G_1$ , let  $\pi = (\pi_1, 0)$  where 0 is a  $1 \times \ell$  dimensional vector. Then

$$(\pi_1, 0)$$
 $\begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix} = (\pi_1 P_1, 0) = (\pi_1, 0),$ 

where the last equality follows since  $\pi_1$  was defined to be the stationary distribution for  $P_1$ . For the chain started on any vertex in  $G_2$ ,  $\pi = (0, \pi_2)$  where here 0 is a  $1 \times k$  vector. So

$$(0, \pi_2) \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix} = (0, \pi_2 P_2) = (0, \pi_2).$$

For  $0 \le \lambda \le 1$ , linear combinations of the form

$$(\lambda \pi_1 + (1 - \lambda)\pi_2) = (\lambda \pi_1 P + (1 - \lambda)\pi_2 P)$$

are also stationary distribution for P. Hence, there exists infinitely stationary for  $X_n$  on G.

**Problem 4.** Consider a Markov chain with state space  $S = \{1, 2\}$  and transition matrix

$$P = \begin{pmatrix} 1 - a & a \\ b & 1 - b \end{pmatrix},$$

where 0 < a < 1 and 0 < b < 1.

(i) Find its stationary distribution  $\pi$ .

Solution. Let  $\pi = (x, y)$ . Then

$$\pi P = \pi \implies \begin{cases} (1-a)x + by = x \\ ax + (1-b)y = y \end{cases} \implies \begin{cases} ax = by \\ x + y = 1 \end{cases} \implies \begin{cases} x = b/(a+b) \\ y = a/(a+b). \end{cases}$$

(ii) Use the Markov property to show that

$$P(X_{n+1} = 1) - \frac{b}{a+b} = (1-a-b)\left(P(X_n = 1) - \frac{b}{a+b}\right).$$

Solution. Start from the left hand side and condition on  $X_n$  to obtain

$$P(X_{n+1} = 1) = \sum_{i} P(X_{n+1} = 1 \mid X_n = i) P(X_n = i)$$

$$= (1 - a)P(X_n = 1) + b(1 - P(X_n = 1))$$

$$= P(X_n = 1) - aP(X_n = 1) + b - bP(X_n = 1)$$

$$= (1 - a - b)P(X_n = 1) + b$$

$$= (1 - a - b)P(X_n = 1) + \frac{b(a + b)}{a + b}$$

$$= (1 - a - b)P(X_n = 1) + \frac{ab}{a + b} + \frac{b^2}{a + b}.$$

Next, subtract b/(a+b) from both sides

$$P(X_{n+1} = 1) - \frac{b}{a+b} = (1-a-b)P(X_n = 1) + \frac{ab}{a+b} + \frac{b^2}{a+b} - \frac{b}{a+b}.$$

Simplifying the above expression gives

$$P(X_{n+1} = 1) - \frac{b}{a+b} = (1-a-b)\left(P(X_n = 1) - \frac{b}{a+b}\right).$$

(iii) Use mathematical induction to show that

$$P(X_n = 1) = \frac{b}{a+b} + (1-a-b)^n \left( P(X_0 = 1) - \frac{b}{a+b} \right)$$

Proof (by induction on n). The first base case n = 0 holds since

$$P(X_0 = 1) = \frac{b}{a+b} + (1-a-b)^0 \left( P(X_0 = 1) - \frac{b}{a+b} \right)$$
$$= \frac{b}{a+b} + P(X_0 = 1) - \frac{b}{a+b} = P(X_0 = 1),$$

and the second base case n = 1 holds since

$$P(X_1 = 1) = \sum_{i} P(X_1 = 1 \mid X_0 = i) P(X_0 = i)$$

$$= P(X_1 = 1 \mid X_0 = 1) P(X_0 = 1) + P(X_1 = 1 \mid X_0 = 2) P(X_0 = 2)$$

$$= (1 - a) P(X_0 = 1) + b(1 - P(X_0 = 1))$$

$$= \frac{b}{a + b} + (1 - a - b) \left( P(X_0 = 1) - \frac{b}{a + b} \right)$$

where the last equality uses the same logic as in (ii).

Assume the claim is true for any n. We want to show that the result is also true for n + 1. From part (i), we know that

$$P(X_{n+1} = 1) = \frac{b}{a+b} + (1-a-b)\left(P(X_n = 1) - \frac{b}{a+b}\right).$$

Substituting in for  $P(X_n = 1)$  gives

$$P(X_{n+1} = 1) = \frac{b}{a+b} + (1-a-b) \left\{ \left( (1-a-b)^n \left( P(X_0 = 1) - \frac{b}{a+b} \right) \right) - \frac{b}{a+b} \right\}$$
$$= \frac{b}{a+b} + (1-a-b)^{n+1} \left( P(X_0 = 1) - \frac{b}{a+b} \right).$$

Shifting the indices back one period gives

$$P(X_n = 1) = \frac{b}{a+b} + (1-a-b)^n \left( P(X_0 = 1) - \frac{b}{a+b} \right).$$

(iv) Show that  $P(X_n = 1)$  converges exponentially fast to  $\pi(1)$  for the  $\pi$  you found in (i). Solution. Since 0 < a < 1 and 0 < b < 1, the term |1 - a - b| < 1. Then, noting that  $(1 - a - b)^n$  is the only term depending on n, we have

$$\lim_{n \to \infty} P(X_n = 1) = \lim_{n \to \infty} \left\{ \frac{b}{a+b} + (1-a-b)^n \left( P(X_0 = 1) - \frac{b}{a+b} \right) \right\}$$
$$= \frac{b}{a+b} + \lim_{n \to \infty} (1-a-b)^n \left( P(X_0 = 1) - \frac{b}{a+b} \right),$$

which shows that convergence to the limiting distribution,  $P(X_n=1) \to \pi(1)$  as  $n \to \infty$ , is exponential at rate  $(1-a-b)^n$ .