

# STAT 433: HOMEWORK 3

DANIEL PFEFFER

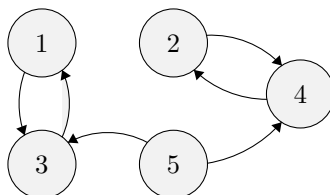
**Problem 1.** Consider the following transition matrices. Identify the transient and recurrent states, and the irreducible closed sets in the Markov chains.

(a)	<b>1</b>	<b>2</b>	<b>3</b>	<b>4</b>	<b>5</b>	(b)	<b>1</b>	<b>2</b>	<b>3</b>	<b>4</b>	<b>5</b>	<b>6</b>
<b>1</b>	.4	.3	.3	0	0	<b>1</b>	.1	0	0	.4	.5	0
<b>2</b>	0	.5	0	.5	0	<b>2</b>	.1	.2	.2	0	.5	0
<b>3</b>	.5	0	.5	0	0	<b>3</b>	0	.1	.3	0	0	.6
<b>4</b>	0	.5	0	.5	0	<b>4</b>	.1	0	0	.9	0	0
<b>5</b>	0	.3	0	.3	.4	<b>5</b>	0	0	0	.4	0	.6
						<b>6</b>	0	0	0	0	.5	.5

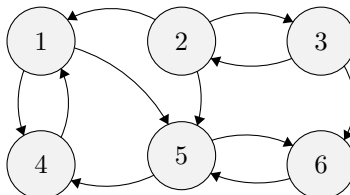
(c)	<b>1</b>	<b>2</b>	<b>3</b>	<b>4</b>	<b>5</b>	(d)	<b>1</b>	<b>2</b>	<b>3</b>	<b>4</b>	<b>5</b>	<b>6</b>
<b>1</b>	0	0	0	0	1	<b>1</b>	.8	0	0	.2	0	0
<b>2</b>	0	.2	0	.8	0	<b>2</b>	0	.5	0	0	.5	0
<b>3</b>	.1	.2	.3	.4	0	<b>3</b>	0	0	.3	.4	.3	0
<b>4</b>	0	.6	0	.4	0	<b>4</b>	.1	0	0	.9	0	0
<b>5</b>	.3	0	0	0	.7	<b>5</b>	0	.2	0	0	.8	0
						<b>6</b>	.7	0	0	.3	0	0

(a) *Solution.* The directed graph representation is



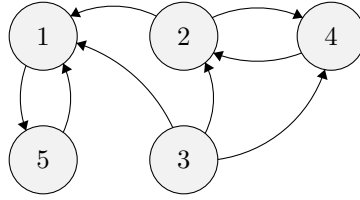
Since  $1 \rightarrow 2$  but  $2 \not\rightarrow 1$ , state 1 is transient. Since  $5 \rightarrow 4$  but  $4 \not\rightarrow 5$ , state 5 is transient. Since  $3 \rightarrow 2$  but  $2 \not\rightarrow 3$ , state 3 is transient. So, states 2 and 4 are recurrent. The closed irreducible set is  $\{2,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

(b) *Solution.* The directed graph representation is



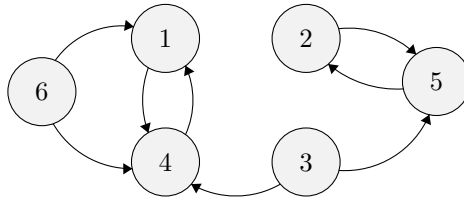
Since  $2 \rightarrow 1$  but  $1 \not\rightarrow 2$ , state 2 is transient. Since  $3 \rightarrow 6$  but  $6 \not\rightarrow 3$ , state 3 is transient. So states 1, 4, 5, and 6 are recurrent. The closed irreducible set is  $\{1, 4, 5, 6\}$ .  $\square$

(c) *Solution.* The directed graph representation is



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

(d) *Solution.* The directed graph representation is



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

**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

$$\begin{aligned} \rho_{xy} &\geq \frac{1}{\deg(x)} \cdot \frac{1}{\deg(x_1)} \cdots \frac{1}{\deg(x_{n-1})} > 0 \\ \rho_{yx} &\geq \frac{1}{\deg(y)} \frac{1}{\deg(x_{n-1})} \cdots \frac{1}{\deg(x_1)} > 0. \end{aligned}$$

Therefore  $x \leftrightarrow y$ .  $\square$

**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  $x \not\leftrightarrow y$  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  $V_1 = \{1, \dots, k\}$  and  $V_2 = \{k+1, \dots, k+\ell\}$  be the set of vertices in  $G_1$  and  $G_2$ . Show that  $P$  is of the following block diagonal form

$$P = \begin{pmatrix} P_1 & 0 \\ 0 & P_2 \end{pmatrix}. \quad (1)$$

Note a block diagonal matrix is a *reducible* matrix.

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

(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 stationary 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  $0 \leq \lambda \leq 1$ , 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  $G$ .  $\square$

**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

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

$\square$

(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

$$\begin{aligned}
 P(X_{n+1} = 1) &= \sum_i P(X_{n+1} = 1 \mid X_n = i)P(X_n = i) \\
 &= (1-a)P(X_n = 1) + b(1 - P(X_n = 1)) \\
 &= P(X_n = 1) - aP(X_n = 1) + b - bP(X_n = 1) \\
 &= (1-a-b)P(X_n = 1) + b \\
 &= (1-a-b)P(X_n = 1) + \frac{b(a+b)}{a+b} \\
 &= (1-a-b)P(X_n = 1) + \frac{ab}{a+b} + \frac{b^2}{a+b}.
 \end{aligned}$$

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

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

and the second base case  $n = 1$  holds since

$$\begin{aligned}
 P(X_1 = 1) &= \sum_i P(X_1 = 1 \mid X_0 = i)P(X_0 = i) \\
 &= P(X_1 = 1 \mid X_0 = 1)P(X_0 = 1) + P(X_1 = 1 \mid X_0 = 2)P(X_0 = 2) \\
 &= (1-a)P(X_0 = 1) + b(1 - P(X_0 = 1)) \\
 &= \frac{b}{a+b} + (1-a-b) \left( P(X_0 = 1) - \frac{b}{a+b} \right)
 \end{aligned}$$

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

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

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

$$\begin{aligned} \lim_{n \rightarrow \infty} P(X_n = 1) &= \lim_{n \rightarrow \infty} \left\{ \frac{b}{a+b} + (1-a-b)^n \left( P(X_0 = 1) - \frac{b}{a+b} \right) \right\} \\ &= \frac{b}{a+b} + \lim_{n \rightarrow \infty} (1-a-b)^n \left( P(X_0 = 1) - \frac{b}{a+b} \right), \end{aligned}$$

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