# 171 Midterm 1 Solutions, Spring 2017<sup>1</sup>

### 1. QUESTION 1

True/False

(a) Let **P** be a probability law on a sample space C. Let  $A_1, A_2, \ldots$  be sets in C which are increasing, so that  $A_1 \subseteq A_2 \subseteq \cdots$ . Then

$$\lim_{n \to \infty} \mathbf{P}(A_n) = \mathbf{P}(\bigcap_{n=1}^{\infty} A_n).$$

FALSE. If  $A_1 = \emptyset$ , and  $A_2 = A_3 = \cdots = \mathcal{C}$ , then the left side is 1, while the right side is  $\mathbf{P}(\emptyset) = 0.$ 

(b) The Markov Chain with transition matrix  $P = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$  has exactly two recurrent

states.

TALSE. All three states are recurrent. Since  $P^2 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ , for any  $x \in \{1, 2, 3\}$  we

have  $\mathbf{P}_x(X_2 = x) = 1$ . So,  $\mathbf{P}_x(T_x \le 2) = 1$ , and  $\mathbf{P}_x(T_x < \infty) = 1$ .

(c) Let X, Y be discrete random variables such that

$$\mathbf{P}(X \le x, Y = y) = \mathbf{P}(X \le x)\mathbf{P}(Y = y), \qquad \forall x, y \in \mathbb{R}.$$

Then

$$\mathbf{P}(X \le x, Y \le y) = \mathbf{P}(X \le x)\mathbf{P}(Y \le y), \qquad \forall x, y \in \mathbb{R}$$

TRUE. For any  $t \in \mathbb{R}$ , let  $A_t = \{Y = t\}$ . Then  $A_{t_1} \cap A_{t_2} = \emptyset$  if  $t_1 \neq t_2$ , and  $\bigcup_{t \leq y} A_t = \{Y = t\}$ .  $\{Y \leq y\}$ , so

$$\sum_{t \le y} \mathbf{P}(X \le x, Y = t) = \sum_{t \le y} \mathbf{P}(\{X \le x\} \cap A_t) = \mathbf{P}(\{X \le x\} \cap (\bigcup_{t \le y} A_t)) = \mathbf{P}(X \le x, Y \le y).$$

Similarly,  $\sum_{t \le y} \mathbf{P}(Y = t) = \mathbf{P}(Y \le y)$ . So, summing both sides of the equality  $\mathbf{P}(X \le y)$ . x, Y = t =  $\mathbf{P}(X \le x)\mathbf{P}(Y = t)$  over all  $t \le y$  proves the assertion.

## 2. Question 2

For any  $x \in \mathbb{R}$ , define

$$\phi(x) := \max\left(-x - 1, 0, x - 1\right).$$

Prove that  $\phi \colon \mathbb{R} \to \mathbb{R}$  is convex.

(In this problem, unlike the other problems, you are allowed to use results from the homework.)

Solution. From Homework 1, Exercise 1, it suffices to show: for any  $y \in \mathbb{R}$ , there exists  $a \in \mathbb{R}$  such that  $L(x) := a(x-y) + \phi(y)$  satisfies  $L(y) = \phi(y)$  and  $L(x) \le \phi(x)$  for all  $x \in \mathbb{R}$ . We break into three cases.

Case 1.  $y \in [-1,1]$ . In this case we choose a = 0. Then L(x) = 0 for all  $x \in \mathbb{R}$ ,  $L(y) = \phi(y) = 0$ , and  $L(x) = 0 \le \phi(x)$  for all  $x \in \mathbb{R}$  by definition of  $\phi$ .

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Case 2. y > 1. In this case we choose a = 1. Then L(x) = (x - y) + (y - 1) = x - 1 for all  $x \in \mathbb{R}$ ,  $L(y) = y - 1 = \phi(y)$ , and  $L(x) \le \phi(x)$  for all  $x \in \mathbb{R}$ , since  $L(x) = \phi(x)$  when  $x \ge 1$ , and if x < 1, then  $L(x) < 0 \le \phi(x)$  since  $\phi(x) \ge 0$  by definition of  $\phi$ .

Case 3. y < -1. In this case we choose a = -1. Then L(x) = -(x-y)+(-y-1) = -x-1 for all  $x \in \mathbb{R}$ ,  $L(y) = -y-1 = \phi(y)$ , and  $L(x) \le \phi(x)$  for all  $x \in \mathbb{R}$ , since  $L(x) = \phi(x)$  when  $x \le -1$ , and if x > -1, then  $L(x) < 0 \le \phi(x)$  since  $\phi \ge 0$  by definition of  $\phi$ .

#### 3. QUESTION 3

Suppose we have a Markov chain  $X_0, X_1, \ldots$  with finite state space  $\Omega$ . Let  $y \in \Omega$ . Define  $L_y := \max\{n \ge 0 \colon X_n = y\}$ . Is  $L_y$  a stopping time? Prove your assertion.

Solution. No,  $L_y$  is not a stopping time. We argue by contradiction. Let  $\Omega := \{1, 2\}$ . If  $L_1$  were a stopping time, then there exists  $B \subseteq \Omega^2$  such that  $\{L_1 = 1\} = \{(X_0, X_1) \in B\}$ . But  $\{L_1 = 1\} = \{X_1 = 1, 2 = X_2 = X_3 = X_4 = \cdots\}$ . That is, the *B* as defined before does not exist.

## 4. QUESTION 4

Suppose we have a Markov Chain  $(X_0, X_1, ...)$  with state space  $\Omega = \{1, 2, 3, 4, 5\}$  and with the following transition matrix

$$P = \begin{pmatrix} 1/2 & 1/2 & 0 & 0 & 0\\ 1/2 & 1/2 & 0 & 0 & 0\\ 0 & 1/3 & 1/3 & 1/3 & 0\\ 0 & 0 & 0 & 1/2 & 1/2\\ 0 & 0 & 0 & 1/2 & 1/2 \end{pmatrix}.$$

Classify state 3 as either transient or recurrent.

Is this Markov Chain irreducible? Prove your assertions.

Solution. State 3 is transient since P(3,2) > 0, while  $P^n(2,3) = 0$  for all  $n \ge 1$ . In fact, it follows by induction on n and the definition of P that  $P^n(i,j) = 0$  for all  $1 \le i \le 2$  and  $3 \le j \le 5$ . By the definition of matrix multiplication, by the inductive hypothesis, and by definition of P, we have

$$P^{n+1}(i,j) = \sum_{k=1}^{5} P^n(i,k) P(k,j) = \sum_{k=1}^{2} P^n(i,k) P(k,j) = \sum_{k=1}^{2} P^n(i,k) \cdot 0 = 0.$$

So, if  $X_1 = 2$ , we know that  $X_n \in \{1, 2\}$  for all  $n \ge 1$  with probability 1. Therefore,

$$\mathbf{P}_3(T_3 = \infty) \ge \mathbf{P}_3(X_1 = 2, X_n \in \{1, 2\} \,\forall \, n \ge 2) = \mathbf{P}_3(X_1 = 2) = P(3, 2) = 1/3 > 0.$$

The Markov chain is not irreducible, since as we mentioned above,  $P^n(1,3) = 0$  for all  $n \ge 1$ .

#### 5. Question 5

Give an example of a Markov chain on the state space  $\Omega = \{1, 2\}$  such that state 1 is recurrent and state 2 is transient. Prove your assertions.

Solution. Define

$$P := \begin{pmatrix} 1 & 0 \\ 1 & 0 \end{pmatrix}.$$

Then P is a stochastic matrix, so P defines a Markov chain on  $\{1,2\}$ . Since P(1,1) = 1,  $\mathbf{P}_1(T_1 < \infty) \geq \mathbf{P}_1(T_1 = 1) = P(1,1) = 1$ . That is,  $\mathbf{P}_1(T_1 < \infty) = 1$ , so the state 1 is recurrent. On the other hand,  $\mathbf{P}_2(T_2 = \infty) = \mathbf{P}_2(1 = X_1 = X_2 = X_3 = \cdots) = \lim_{n \to \infty} P(2,1)[P(1,1)]^n = 1$ . That is,  $\mathbf{P}_2(T_2 < \infty) = 0$ , so that state 2 is transient.