

# MATH 4740 HW4 Solution.

## Textbook Exercises:

1.14. A finite-state Markov Chain would converge if:

(i) it's irreducible

(ii) it's aperiodic.

(a). No, because each state has period 2.

(Also, notice it's an alternating transition matrix.)

(b). Yes, it's irreducible & aperiodic.

(c). No. Each state has period 3.

□

1.36 The transition matrix:

$$P = \begin{bmatrix} 0 & 0 & 1 \\ 0.05 & 0.95 & 0 \\ 0 & 0.02 & 0.98 \end{bmatrix}$$

$$(a) \quad \pi = (0 \ 0 \ 1) \begin{pmatrix} -1 & 0 & 1 \\ 0.05 & -0.05 & 1 \\ 0 & 0.02 & 1 \end{pmatrix}^{-1} = \left( \frac{1}{71} \quad \frac{20}{71} \quad \frac{50}{71} \right).$$

Therefore the longrun fraction of time spent

with 1 bulb is  $\frac{20}{71}$ .

$$(b). \quad E_0(T_0) = \frac{1}{\pi(0)} = 71.$$

□

1.48 (a). The transition matrix is given by:

$$P = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 & 8 & 9 & 10 & 11 & 12 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ 10 \\ 11 \\ 12 \end{matrix} & \left[ \begin{array}{cccccccccccc} & & & & & & & & & & & \frac{1}{2} \\ \frac{1}{2} & & & & & & & & & & & & \\ & \frac{1}{2} & & & & & & & & & & & \\ & & \frac{1}{2} & & & & & & & & & & \\ & & & \frac{1}{2} & & & & & & & & & \\ & & & & \frac{1}{2} & & & & & & & & \\ & & & & & \frac{1}{2} & & & & & & & \\ & & & & & & \frac{1}{2} & & & & & & \\ & & & & & & & \frac{1}{2} & & & & & \\ & & & & & & & & \frac{1}{2} & & & & \\ & & & & & & & & & \frac{1}{2} & & & \\ & & & & & & & & & & \frac{1}{2} & & \\ & & & & & & & & & & & \frac{1}{2} & \\ \frac{1}{2} & & & & & & & & & & & & \frac{1}{2} \end{array} \right] \end{matrix}$$

First, notice that this is a ~~so~~ doubly stochastic chain, (the rows and columns of  $P$  both sum to 1).

By Thm 1.24,  $\pi(x) = \frac{1}{N} \forall x \in \mathcal{X}$ . ( $N = \# \text{ states}$ ).

$$\text{So } E_{\pi}(T_x) = \frac{1}{\pi(x)} = \frac{1}{\frac{1}{N}} = N = 12.$$

□.

### Additional Problems:

1. (a).  $\pi = \left(\frac{1}{3} \quad \frac{2}{3}\right)$

(b). 
$$\begin{aligned} \mathbb{P}_b(T_a = k) &= \mathbb{P}(X_k = a, X_{k-1} = b, X_{k-2} = b, \dots, X_1 = b \mid X_0 = b) \\ &= \mathbb{P}(X_k = a \mid X_{k-1} = b) \mathbb{P}(X_{k-1} = b \mid X_{k-2} = b) \cdots \\ &\quad \cdot \mathbb{P}(X_1 = b \mid X_0 = b) \\ &= 0.3 \cdot 0.7^{k-1} \end{aligned}$$

$$\begin{aligned} \mathbb{P}_b(T_a \geq k) &= \mathbb{P}(\cancel{X_k = a}, X_{k-1} = b, X_{k-2} = b, \dots, X_1 = b \mid X_0 = b) \\ &= 0.7^{k-1} \end{aligned}$$

(c). 
$$\begin{aligned} \mathbb{E}_b(T_a) &= \sum_{k=1}^{\infty} \mathbb{P}_b(T_a \geq k) \\ &= \sum_{k=1}^{\infty} 0.7^{k-1} \\ &= \frac{1}{1-0.7} \\ &= \frac{10}{3} \end{aligned}$$

(d). Starting from state a ( $X_0 = a$ ), it will stay at a w/ prob. 0.4 ( $X_1 = a$ ) or it will go to b w/ prob. 0.6. So  $\mathbb{E}_a(T_a) = 0.4 \cdot 1 + 0.6 \cdot (\mathbb{E}_b(T_a) + 1)$   
 $= 3.$

$$\pi(a) = \frac{1}{3} = \frac{1}{\mathbb{E}_a(T_a)}. \quad \checkmark$$

□

2. (a). Starting from state  $y$ , it has two possibilities:

it will never visit state  $x$  ( $N(x) = 0$ ) or

it will visit  $x$  at some time point  $k$ .

In the latter case, once it's at state  $x$ , the number of visits to  $x$  after time  $k$  has the same distribution as  $N(x)$  for the chain started at  $x$ , by strong Markov property.

$$\text{So } \mathbb{E}_y[N(x)] = 0 \cdot (1 - p_{yx}) + p_{yx} \cdot (1 + \mathbb{E}_x[N(x)]).$$

(b) Suppose  $x$  is transient, then  $\mathbb{E}_x[N(x)] < \infty$ .

$$\text{So } \mathbb{E}_y[N(x)] = p_{yx} (1 + \mathbb{E}_x[N(x)]) < \infty$$

$$\text{Also recall that } \mathbb{E}_y[N(x)] = \sum_{n=1}^{\infty} P^n(y, x) < \infty.$$

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So by test for divergence,  $\lim_{n \rightarrow \infty} P^n(y, x) = 0$ .

(c). Since  $\lim_{n \rightarrow \infty} (\pi P^n)(x) = \pi(x)$ , we have

$$\lim_{n \rightarrow \infty} \sum_{y \in X} \pi(y) P^n(y, x) = \sum_y \lim_{n \rightarrow \infty} \pi(y) P^n(y, x)$$

$$= \sum_y \pi(y) \lim_{n \rightarrow \infty} P^n(y, x)$$

$$= \sum_y (\pi(y) \cdot 0)$$

$$= 0.$$

$$= \pi(x).$$

□.

3. Notice that it's an irreducible Markov Chain with finite state space, we can apply the "main convergence theorem" such that  $\pi(0) = \frac{1}{\mathbb{E}_0(T_0)}$

Starting from state 0, it will return to 0 in 3 steps ( $0 \rightarrow -1 \rightarrow -2 \rightarrow 0$ ) or 4 steps ( $0 \rightarrow 1 \rightarrow 2 \rightarrow 3 \rightarrow 0$ ), each w/ probability  $\frac{1}{2}$ . Hence

$$\mathbb{E}_0(T_0) = \frac{1}{2} \cdot 3 + \frac{1}{2} \cdot 4 = \frac{7}{2}$$

Therefore  $\pi(0) = \frac{1}{\frac{7}{2}} = \frac{2}{7}$ .

□.