

### Question 1: Markov Chain

(a) Transition matrix  $Q$   
states =  $\{1, 2, 3, 4\}$

Transition probabilities:

From state 1:  $P(1 \rightarrow 1) = 0.5$

$P(1 \rightarrow 2) = 0.5$

From state 2:  $P(2 \rightarrow 1) = 0.25$

$P(2 \rightarrow 3) = 0.75$

From state 3:  $P(3 \rightarrow 3) = 0.75$

$P(3 \rightarrow 4) = 0.25$

From state 4:  $P(4 \rightarrow 3) = 0.25$

$P(4 \rightarrow 4) = 0.75$

Transition Matrix  $Q$ :

$$Q = \begin{pmatrix} 0.5 & 0.5 & 0 & 0 \\ 0.25 & 0 & 0.75 & 0 \\ 0 & 0 & 0.75 & 0.25 \\ 0 & 0 & 0.25 & 0.75 \end{pmatrix}$$

(b) Recurrent and transient states:

1. State 1 and 2

•  $1 \leftrightarrow 2$  (they communicate with each other)

• From state 1, you can return to state 1

• From state 2, you can return to state 2

• Thus state 1 and 2 form a recurrent class.

2. State 3 and 4

•  $3 \leftrightarrow 4$  (they can communicate)

• From state 3, you can return to state 3

• From state 4, you can return to state 4

Thus state 3 and 4 are recurrent.

### (c) Stationary Distributions

stationary distribution satisfies  $\pi Q = \pi$  and  $\sum \pi_i = 1$

1. For recurrent class  $\{1, 2\}$

• Let  $\pi_1, \pi_2$  be probabilities for state 1 and 2.

• Equations:

$$\pi_1 = 0.5\pi_1 + 0.25\pi_2$$

$$\pi_2 = 0.5\pi_1 + 0\pi_2$$

• Second eq<sup>n</sup>:  $\pi_2 = 0.5\pi_1$

• Substitute into the first eq<sup>n</sup>:

$$\pi_1 = 0.5\pi_1 + 0.25 \times 0.5\pi_1 \Rightarrow \pi_1 = 0.625\pi_1$$

This shows  $\pi_1 = 0$ , which is not valid.

$$\pi_1 = 0.5\pi_1 + 0.25\pi_2$$

$$\pi_2 = 0.5\pi_1$$

Substituting in first eq<sup>n</sup>:

$$\pi_1 = 0.5\pi_1 + 0.25 \times 0.5\pi_1 = 0.5\pi_1 + 0.125\pi_1 = 0.625\pi_1$$

$$\pi_2 = 0.5\pi_1$$

$$\Rightarrow \pi_1 = \frac{2}{3}, \pi_2 = \frac{1}{3}$$

• Thus one stationary distribution is:

$$\pi^1 = \left( \frac{2}{3}, \frac{1}{3}, 0, 0 \right)$$

• Similarly for recurrent class  $\{3, 4\}$

$$\text{Eq<sup>n</sup> 1: } \pi_3 = 0.75\pi_3 + 0.25\pi_4$$

$$\pi_4 = 0.25\pi_3 + 0.75\pi_4$$

$$\rightarrow 0.5 \times 0.5 = 1$$

$$\rightarrow \bar{X}_2 = \bar{X}_1 = 0.5$$

$$X^{(2)} = (0, 0.5, 0.5)$$

find answers:  $X^{(1)} = (3/2, 1/2, 0, 0)$   
 $X^{(2)} = (0, 0, 1/2, 1/2)$

Question 2:

(a) Let  $X_1 = 0.5$ ,  $L \rightarrow \text{loss}$

for iterations

$$P(W=0) = 0.8$$

$$P(L=0) = 0.2$$

$$P(L=0.5) = 0.8$$

$$P(L=1) = 0.2$$

$$P(L=2) = 0.2$$

$$\text{Say } P = (0.8, 0.2)$$

we find stationary distribution  $\pi = (\pi_0, \pi_1, \pi_2)$

subject to  $\pi = P^T \pi$ ,  $\pi_0 + \pi_1 + \pi_2 = 1$

$$\pi_0 = 0.8\pi_0 + 0.2\pi_1$$

$$\pi_1 = 0.2\pi_0 + 0.8\pi_1$$

$$\pi_2 = 0.2\pi_1 + 0.8\pi_2$$

$$\pi_0 = 0.5, \pi_1 = 0.5, \pi_2 = 0$$

or  $\pi_0 = 0.5, \pi_1 = 0.5$  (long run proportion of being in  $\pi_0$ )

(b) the dinner probabilities, after a win = 0.7, after a loss = 0.2

the expected proportion of dinners,  $\pi_0 = 0.7 + 0.2$

$\neq 0.5$  (long-run proportion of dinners with dinner)

(c) The expected waiting time until a dinner is the inverse of the probability of dinner in a state given expected time of dinner =  $1/0.5 = 2$



Question 3:

(a) Cat chain

- cat moves 2 rooms (Room 1 and Room 2)
- It moves to other room with probab 0.8, so it stays in the same room with probab 0.2

$\pi_c(1)$  = long-run probab; Cat is in Room 1

$$\pi_c(2) = 1 - \pi_c(1)$$

$$\text{Transition matrix } P_c = \begin{bmatrix} 0.2 & 0.8 \\ 0.8 & 0.2 \end{bmatrix}$$

$$\text{Stationary distribution } \pi_c = [\pi_1, \pi_2]$$

$$\pi_c \cdot P_c = \pi_c, \pi_1 + \pi_2 = 1$$

$$\begin{aligned} \text{From eqn: } \pi_1 &= 0.2\pi_1 + 0.8\pi_2 \\ \Rightarrow 0.8\pi_1 &= 0.8\pi_2 \Rightarrow \pi_1 = \pi_2 \Rightarrow 0.5 \end{aligned}$$

$$\pi_c = [0.5, 0.5]$$

Mouse distribution:

$$\pi_m = [\pi_1, \pi_2]$$

$$P_m = \begin{bmatrix} 0.7 & 0.3 \\ 0.6 & 0.4 \end{bmatrix}$$

$$\pi_1 = 0.7\pi_1 + 0.6\pi_2$$

$$\pi_1 + \pi_2 = 1$$

$$\Rightarrow \pi_1 = \frac{2}{3}, \pi_2 = \frac{1}{3}$$

$$\pi_m = \left[ \frac{2}{3}, \frac{1}{3} \right]$$

(b) 4 possible joint states,

$$(1,1), (1,2), (2,1), (2,2)$$

let  $Z_n$  be the joint state of (cat room, mouse room)

$\Rightarrow Z_n$  is a Markov Chain with 4 states, as transition to next state depends on previous one.