

### Exercises with solutions on Markov chain and applications

- Recall the definition of a Markov chain (do it at home)
- Give the form of the transition matrix of a general Markov chain and give an example (do it at home)
- Give the form of the transition matrix of an homogeneous Markov chain and give an example (do it at home)
- Consider a Markov chain with  $E = \{0, 1, 2\}$  and

$$P(6, 7) = \begin{pmatrix} 0.2 & 0.3 & 0.5 \\ 0.3 & 0.5 & 0.2 \\ 0.1 & 0.1 & 0.8 \end{pmatrix}$$

$$\mathbf{P} = \begin{pmatrix} \frac{1}{2} & 0 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 1 \\ \frac{2}{3} & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}$$

- (a) Find: (i)  $P(X_7 = j | X_6 = 0)$ ,  $j = 0, 1, 2$ . (ii)  $P(X_7 = 1 | X_6 = j)$ ,  $j = 0, 1, 2$ .  
 (b) Which of the following are legitimate one-step transition probability matrices

$$\text{i) } P_0 = \begin{pmatrix} -1 & 2 \\ 0.5 & 0.5 \end{pmatrix}, \quad \text{ii) } Q_0 = \begin{pmatrix} 0.3 & 0.7 \\ 0.4 & 0.6 \end{pmatrix} \quad \text{and} \quad R_0 = \begin{pmatrix} 0.3 & 0.7 \\ 0.5 & 0.6 \end{pmatrix}$$

**Solution:** a. (i) We have that  $P(X_7 = 0 | X_6 = 0) = 0.2$ ,  $P(X_7 = 1 | X_6 = 0) = 0.3$  and  $P(X_7 = 2 | X_6 = 0) = 0.5$ .

(ii) We have that  $P(X_7 = 1 | X_6 = 0) = 0.3$ ,  $P(X_7 = 1 | X_6 = 1) = 0.5$  and  $P(X_7 = 1 | X_6 = 2) = 0.1$ .

b. (i) The matrix is not a legitimate transition probability because the entry  $(1, 1)$  is negative.

ii) Yes  $Q_0$  is a transition probability matrix.

iii) The matrix is not a legitimate transition probability because the elements of the second row do not add to one.

- A computer system can operate in two different modes. Every hour, it remains in the same mode or switches to a different mode according to the transition probability matrix

$$P = \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix}$$

- (a) Compute the 2-step transition probability matrix.

**Solution:** Since the Markov chain is homogeneous 2-step transition probability matrix is given by

$$P^{(2)} = P^2 = \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix} \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix} = \begin{pmatrix} 0.52 & 0.48 \\ 0.48 & 0.52 \end{pmatrix}$$

- (b) If the system is in Mode I at 5:30 pm, what is the probability that it will be in Mode I at 8:30 pm on the same day?

**Solution:** We have

$$P(X_{8:30} = j \mid X_{5:30} = i) = P(X_3 = j \mid X_0 = i) = P_{i,j}^{(3)}$$

therefore

$$P^{(3)} = P^3 = \begin{pmatrix} 0.52 & 0.48 \\ 0.48 & 0.52 \end{pmatrix} \begin{pmatrix} 0.4 & 0.6 \\ 0.6 & 0.4 \end{pmatrix} = \begin{pmatrix} 0.496 & 0.504 \\ 0.504 & 0.496 \end{pmatrix}.$$

So  $P_{i,i}^{(3)} = 0.496$  for  $i = 1, 2$ .

6. Consider a Markov chain with  $E = \{0, 1, 2\}$ ,  $\alpha_0 = (0.3, 0.4, 0.3)$ ,

$$P(0, 1) = \begin{pmatrix} 0.2 & 0.3 & 0.5 \\ 0.3 & 0.5 & 0.2 \\ 0.1 & 0.1 & 0.8 \end{pmatrix} \quad \text{and} \quad P(1, 2) = \begin{pmatrix} 0.1 & 0.1 & 0.8 \\ 0.3 & 0.5 & 0.2 \\ 0.2 & 0.5 & 0.3 \end{pmatrix}$$

- (a) Calculate  $P(X_0 = 1)$ ,  $P(X_1 = 1)$ ,  $P(X_2 = 1)$ ,

**Solution:** We have  $\alpha_0 = (P(X_0 = 0), P(X_0 = 1), P(X_0 = 2))$  then  $P(X_0 = 1) = 0.4$ .

We know that  $P(X_1 = 1) = \alpha_1(1)$  where

$$\alpha_1 = \alpha_0 P(0, 1) = (0.3, 0.4, 0.3) \begin{pmatrix} 0.2 & 0.3 & 0.5 \\ 0.3 & 0.5 & 0.2 \\ 0.1 & 0.1 & 0.8 \end{pmatrix}$$

$$\alpha_1 = (0.21, 0.32, 0.47) \quad \text{and} \quad P(X_1 = 1) = 0.32$$

Also  $P(X_2 = 1) = \alpha_2(1)$  where

$$\alpha_2 = \alpha_1 P(1, 2) = (0.21, 0.32, 0.47) \begin{pmatrix} 0.1 & 0.1 & 0.8 \\ 0.3 & 0.5 & 0.2 \\ 0.2 & 0.5 & 0.3 \end{pmatrix}$$

$$\alpha_2 = (0.211, 0.416, 0.373) \quad \text{and} \quad P(X_2 = 1) = 0.416$$

- (b) Calculate  $P(X_0 = 1, X_1 = 2)$ ,  $P(X_0 = 1, X_1 = 0)$ ,  $P(X_0 = 0, X_2 = 2)$ .

**Solution:** We have

$$\begin{aligned} P(X_0 = 1, X_1 = 2) &= P(X_1 = 2 \mid X_0 = 1) P(X_0 = 1) \\ &= P_{1,2}(0, 1) \alpha_0(1) = 0.2 \times 0.4 = 0.08 \end{aligned}$$

and

$$\begin{aligned} P(X_0 = 1, X_1 = 0) &= P(X_1 = 0 \mid X_0 = 1) P(X_0 = 1) \\ &= P_{1,0}(0, 1) \alpha_0(1) = 0.3 \times 0.4 = 0.12 \end{aligned}$$

$$\begin{aligned} P(X_0 = 0, X_2 = 2) &= P(X_2 = 2 \mid X_0 = 0) P(X_0 = 0) \\ &= P_{0,2}(0, 2) \alpha_0(0) \end{aligned}$$

and

$$\begin{aligned} P(0, 2) &= P(0, 1) P(1, 2) \\ &= \begin{pmatrix} 0.2 & 0.3 & 0.5 \\ 0.3 & 0.5 & 0.2 \\ 0.1 & 0.1 & 0.8 \end{pmatrix} \begin{pmatrix} 0.1 & 0.1 & 0.8 \\ 0.3 & 0.5 & 0.2 \\ 0.2 & 0.5 & 0.3 \end{pmatrix} \\ &= \begin{pmatrix} 0.21 & 0.42 & 0.37 \\ 0.22 & 0.38 & 0.4 \\ 0.2 & 0.46 & 0.34 \end{pmatrix} \end{aligned}$$

then  $P_{0,2}(0, 2) \alpha_0(0) = 0.37 \times 0.3 = 0.111$

(c) Calculate  $P(X_0 = 1, X_1 = 2, X_2 = 2)$ ,  $P(X_0 = 2, X_1 = 1, X_2 = 0)$ .

**Solution:** We can write

$$\begin{aligned} P(X_0 = 1, X_1 = 2, X_2 = 2) &= P(X_2 = 2 \mid X_0 = 1, X_1 = 2) P(X_0 = 1, X_1 = 2) \\ &= P(X_2 = 2 \mid X_1 = 2) \times 0.08 \\ &= P_{2,2}(1, 2) \times 0.08 = 0.3 \times 0.08. \end{aligned}$$

And

$$\begin{aligned} P(X_0 = 2, X_1 = 1, X_2 = 0) &= P(X_2 = 0 \mid X_0 = 2, X_1 = 1) P(X_1 = 1 \mid X_0 = 2) P(X_0 = 2) \\ &= P(X_2 = 0 \mid X_1 = 1) P(X_1 = 1 \mid X_0 = 2) \alpha_0(2) \\ &= P_{1,0}(1, 2) \times P_{2,1}(0, 1) \times \alpha_0(2) = 0.1 \times 0.1 \times 0.3 = 0.003. \end{aligned}$$

7. Consider a Markov chain with  $E = \{1, 2\}$ ,

$$P(0, 1) = \begin{pmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{pmatrix}, \quad P(1, 2) = \begin{pmatrix} 0.6 & 0.4 \\ 0.2 & 0.8 \end{pmatrix}$$

Suppose that  $X_0 = 1$ . Find the probability that at stage 2 the chain is in state 2.

**Solution:** We should find  $P(X_2 = 2) = \alpha_2(2)$  then we need  $\alpha_2$  which is given by  $\alpha_2 = \alpha_1 P(1, 2) = \alpha_0 P(0, 1) P(1, 2)$ . But  $\alpha_0 = (P(X_0 = 1), P(X_0 = 2)) = (1, 0)$

$$(1, 0) \begin{pmatrix} 0.7 & 0.3 \\ 0.5 & 0.5 \end{pmatrix} \begin{pmatrix} 0.6 & 0.4 \\ 0.2 & 0.8 \end{pmatrix} = (0.48, 0.52)$$

Then  $P(X_2 = 2) = 0.52$ .

8. Consider a Markov chain with  $E = \{1, 2\}$ ,  $\alpha_3 = (0.2, 0.8)$ ,

$$P(3, 4) = \begin{pmatrix} 0.3 & 0.7 \\ 0.6 & 0.4 \end{pmatrix}, \quad P(4, 5) = \begin{pmatrix} 0.5 & 0.5 \\ 0.8 & 0.2 \end{pmatrix}$$

(a) Find  $P(X_3 = 2, X_4 = 1)$ ,  $P(X_3 = 1, X_4 = 1, X_5 = 2)$ ,  $P(X_5 = 2)$ .

**Solution:** We have

$$P(X_3 = 2, X_4 = 1) = P(X_4 = 1 \mid X_3 = 2) P(X_3 = 2) = P_{2,1}(3, 4) \alpha_3(2) = 0.6 \times 0.8.$$

We can write by definition.

$$\begin{aligned} P(X_3 = 1, X_4 = 1, X_5 = 2) &= P(X_5 = 2 \mid X_4 = 1, X_3 = 1) P(X_4 = 1, X_3 = 1) \\ &= P(X_5 = 2 \mid X_4 = 1) P(X_4 = 1 \mid X_3 = 1) P(X_3 = 1) \\ &= P_{1,2}(4, 5) P_{1,1}(3, 4) \alpha_3(1) = 0.5 \times 0.3 \times 0.2. \end{aligned}$$

To compute  $P(X_5 = 2) = \alpha_5(2)$  we need  $\alpha_5$  which is given by

$$\alpha_5 = \alpha_4 P(4, 5) = \alpha_3 P(3, 4) P(4, 5)$$

Then

$$\alpha_5 = (0.2, 0.8) \begin{pmatrix} 0.3 & 0.7 \\ 0.6 & 0.4 \end{pmatrix} \begin{pmatrix} 0.5 & 0.5 \\ 0.8 & 0.2 \end{pmatrix} = (0.638, 0.362)$$

and  $\alpha_5(2) = 0.362$ .

9. Consider a Markov chain with  $E = \{1, 2\}$ ,

$$P(3, 4) = \begin{pmatrix} 0.3 & 0.7 \\ 0.6 & 0.4 \end{pmatrix}, \quad P(4, 5) = \begin{pmatrix} 0.5 & 0.5 \\ 0.8 & 0.2 \end{pmatrix}$$

(a) Find  $P(X_4 = 2 \mid X_3 = 1)$ ,  $P(X_4 = 2, X_5 = 1 \mid X_3 = 1)$ ,  $P(X_5 = 1 \mid X_3 = 1)$ .

**Solution:** We have

$$P(X_4 = 2 \mid X_3 = 1) = P_{1,2}(3, 4) = 0.7$$

and

$$\begin{aligned} P(X_4 = 2, X_5 = 1 \mid X_3 = 1) &= P(X_5 = 1 \mid X_4 = 2, X_3 = 1) P(X_4 = 2 \mid X_3 = 1) \\ &= P(X_5 = 1 \mid X_4 = 2) P(X_4 = 2 \mid X_3 = 1) \\ &= P_{2,1}(4, 5) P_{1,2}(3, 4) = 0.8 \times 0.7 = 0.56 \end{aligned}$$

and

$$P(X_5 = 1 \mid X_3 = 1) = P_{1,1}(3, 5)$$

so we need the matrix  $P(3, 5)$  which is given by

$$\begin{aligned} P(3, 5) &= P(3, 4)P(4, 5) \\ &= \begin{pmatrix} 0.3 & 0.7 \\ 0.6 & 0.4 \end{pmatrix} \begin{pmatrix} 0.5 & 0.5 \\ 0.8 & 0.2 \end{pmatrix} \\ &= \begin{pmatrix} 0.71 & 0.29 \\ 0.62 & 0.38 \end{pmatrix} \end{aligned}$$

then  $P(X_5 = 1 \mid X_3 = 1) = 0.71$ .

(b) Set

$$P(7, 8) = \begin{pmatrix} 0.3 & 0.7 \\ 0.2 & 0.8 \end{pmatrix}, \quad P(8, 9) = \begin{pmatrix} 0.6 & 0.4 \\ 0.5 & 0.5 \end{pmatrix}$$

Find  $P(X_9 = 1 \mid X_5 = 2, X_7 = 1)$ .

**Solution:** From the definition of the Markov chain we can write

$$P(X_9 = 1 \mid X_5 = 2, X_7 = 1) = P(X_9 = 1 \mid X_7 = 1) = P_{1,1}(7, 9)$$

so we need the matrix  $P(7, 9)$  which is given by

$$\begin{aligned} P(7, 9) &= P(7, 8)P(8, 9) \\ &= \begin{pmatrix} 0.3 & 0.7 \\ 0.2 & 0.8 \end{pmatrix} \begin{pmatrix} 0.6 & 0.4 \\ 0.5 & 0.5 \end{pmatrix} \\ &= \begin{pmatrix} 0.53 & 0.47 \\ 0.52 & 0.48 \end{pmatrix} \end{aligned}$$

then  $P(X_9 = 1 \mid X_5 = 2, X_7 = 1) = 0.53$ .

10. Suppose that an homogeneous Markov chain has state space  $E = \{1, 2, 3\}$ , transition matrix

$$P = \begin{pmatrix} \frac{1}{3} & \frac{2}{3} & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \end{pmatrix}$$

and initial distribution  $\alpha_0 = (\frac{1}{2}, \frac{1}{3}, \frac{1}{6})$ .

(a) Find:  $P(X_2 = 2)$ ,  $P(X_0 = 1, X_3 = 3)$ ,

**Solution:**  $P(X_2 = 2) = \alpha_2(2)$  which can be calculated easily from the formula

$$\alpha_2 = \alpha_1 P(1, 2) = \alpha_0 P(0, 1) P(1, 2) = \alpha_0 P^2$$

then

$$\begin{pmatrix} \frac{1}{3} & \frac{2}{3} & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \end{pmatrix}^2 = \begin{pmatrix} \frac{4}{9} & \frac{2}{7} & \frac{1}{3} \\ \frac{24}{7} & \frac{11}{24} & \frac{1}{4} \\ \frac{1}{3} & \frac{24}{7} & \frac{3}{8} \end{pmatrix}$$

and

$$\alpha_2 = \left( \frac{1}{2}, \frac{1}{3}, \frac{1}{6} \right) \begin{pmatrix} \frac{4}{9} & \frac{2}{7} & \frac{1}{3} \\ \frac{24}{7} & \frac{11}{24} & \frac{1}{4} \\ \frac{1}{3} & \frac{24}{7} & \frac{3}{8} \end{pmatrix} = \left( \frac{3}{8}, \frac{5}{16}, \frac{5}{16} \right)$$

then :  $\alpha_2(2) = \frac{5}{16}$ .

$$P(X_0 = 1, X_3 = 3) = P(X_3 = 3 | X_0 = 1) P(X_0 = 1) = \alpha_0(1) P_{1,3}(0, 3)$$

hence we need to compute  $P(0, 3) = P(0, 1)P(1, 2)P(2, 3) = P^3$ . Now we have

$$\begin{pmatrix} \frac{1}{3} & \frac{2}{3} & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \end{pmatrix}^3 = \begin{pmatrix} \frac{37}{108} & \frac{41}{108} & \frac{5}{18} \\ \frac{7}{37} & \frac{37}{108} & \frac{17}{17} \\ \frac{18}{288} & \frac{144}{288} & \frac{48}{3} \end{pmatrix}$$

hence  $P(X_0 = 1, X_3 = 3) = \frac{1}{2} \times \frac{5}{18} = \frac{5}{36}$ .

(b) Find  $P(X_1 = 2, X_2 = 3, X_3 = 1 | X_0 = 1)$ ,  $P(X_2 = 3 | X_3 = 1)$

**Solution:**

$$\begin{aligned} & P(X_1 = 2, X_2 = 3, X_3 = 1 | X_0 = 1) \\ &= \frac{P(X_1 = 2, X_2 = 3, X_3 = 1, X_0 = 1)}{P(X_0 = 1)} \\ &= \frac{P(X_3 = 1 | X_0 = 1, X_1 = 2, X_2 = 3) P(X_2 = 3 | X_0 = 1, X_1 = 2) P(X_0 = 1, X_1 = 2)}{P(X_0 = 1)} \\ &= P(X_3 = 1 | X_2 = 3) P(X_2 = 3 | X_1 = 2) P(X_1 = 2 | X_0 = 1) \\ &= P_{3,1} P_{2,3} P_{1,2} = \frac{1}{4} \times \frac{1}{2} \times \frac{2}{3} = \frac{1}{12} \end{aligned}$$

From the definition of conditional probability we have

$$\begin{aligned} P(X_2 = 3 | X_3 = 1) &= \frac{P(X_2 = 3, X_3 = 1)}{P(X_3 = 1)} = \frac{P(X_3 = 1 | X_2 = 3) P(X_2 = 3)}{P(X_3 = 1)} \\ &= \frac{1}{\alpha_3(1)} P_{3,1}(2, 3) \alpha_2(3). \end{aligned}$$

Moreover

$$\alpha_3 = \alpha_2 P = \left( \frac{3}{8}, \frac{5}{16}, \frac{5}{16} \right) \begin{pmatrix} \frac{1}{3} & \frac{2}{3} & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \end{pmatrix} = \left( \frac{23}{64}, \frac{21}{64}, \frac{5}{16} \right)$$

thence

$$P(X_2 = 3 | X_3 = 1) = \frac{1}{\frac{23}{64}} \frac{1}{4} \frac{5}{16} = \frac{5}{23}$$

- (c) Find  $P(X_{12} = 1 \mid X_5 = 3, X_{10} = 1)$ ,  $P(X_3 = 3, X_5 = 1 \mid X_0 = 1)$ ,  $P(X_3 = 3 \mid X_0 = 1)$

**Solution:** The the Markov property of the chain leads to

$$\begin{aligned} P(X_{12} = 1 \mid X_5 = 3, X_{10} = 1) &= P(X_{12} = 1 \mid X_{10} = 1) \\ &= P_{1,1}(10, 12). \end{aligned}$$

This the (1,1) entry of the 2-step transition probability matrix  $P(10, 12) = P(10, 11)P(11, 12) = P^2$  which was computed in a. then  $P_{1,1}(10, 12) = P_{1,1}^2 = \frac{37}{108}$ .

From the properties of the conditional probability we can write

$$\begin{aligned} P(X_3 = 3, X_5 = 1 \mid X_0 = 1) &= P(X_5 = 1 \mid X_3 = 3, X_0 = 1) P(X_3 = 3 \mid X_0 = 1) \\ &= P(X_5 = 1 \mid X_3 = 3) P(X_3 = 3 \mid X_0 = 1) \\ &= P_{3,1}(3, 5)P_{1,3}(0, 3) \end{aligned}$$

So we need the two matrix  $P(3, 5)$  and  $P(0, 3)$ , but we know that  $P(3, 5) = P(3, 4)P(4, 5) = P^2$  and  $P(0, 3) = P(0, 1)P(1, 2)P(2, 3) = P^3$  since the chain is homogeneous, therefore  $P_{3,1}(3, 5) = P_{3,1}^2 = \frac{1}{3}$  and  $P_{1,3}(0, 3) = P_{1,3}^3 = \frac{5}{18}$ , finally  $P(X_3 = 3, X_5 = 1 \mid X_0 = 1) = \frac{1}{3} \cdot \frac{5}{18} = \frac{5}{54}$ .

$$P(X_3 = 3 \mid X_0 = 1) = P_{1,3}(0, 3) = P_{1,3}^3 = \frac{5}{18}.$$

11. A Markov chain  $\{X_n, n \geq 0\}$  with states 0, 1, 2, has the transition probability matrix

$$P = \begin{pmatrix} \frac{1}{2} & \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} \end{pmatrix}$$

- (a) Draw a state transition diagram of the Markov chain  
 (b) If  $P(X_0 = 0) = P(X_0 = 1) = \frac{1}{4}$ , find  $E[X_3]$ .

**Solution:** We have

$$\begin{aligned} E[X_3] &= \sum_{i=0}^2 iP(X_3 = i) \\ &= P(X_3 = 1) + 2P(X_3 = 2) \\ &= \alpha_3(1) + 2\alpha_3(2) \end{aligned}$$

$$P^3 = \begin{pmatrix} \frac{1}{2} & \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} \end{pmatrix}^3 = \begin{pmatrix} \frac{13}{36} & \frac{11}{54} & \frac{47}{108} \\ \frac{4}{9} & \frac{27}{9} & \frac{11}{27} \\ \frac{5}{12} & \frac{2}{9} & \frac{13}{36} \end{pmatrix}$$

then  $\alpha_3 = \alpha_0 P^3$

$$\begin{pmatrix} 1 & 1 & 1 \\ \frac{1}{4} & \frac{1}{4} & \frac{1}{2} \end{pmatrix} \begin{pmatrix} \frac{13}{36} & \frac{11}{54} & \frac{47}{108} \\ \frac{4}{9} & \frac{27}{9} & \frac{11}{27} \\ \frac{5}{12} & \frac{2}{9} & \frac{13}{36} \end{pmatrix} = \begin{pmatrix} 59 & 43 & 169 \\ 144 & 216 & 432 \end{pmatrix}$$

then  $\alpha_3(1) = \frac{43}{216}$  and  $\alpha_3(2) = \frac{169}{432}$ , then  $E[X_3] = \frac{43}{216} + 2\frac{169}{432} = \frac{373}{144}$ .

12. A system has three possible states, 0, 1 and 2. Every hour it makes a transition to a different state, which is determined by a coin flip. For example, from state 0, it makes a transition to state 1 or state 2 with probabilities 0.5 and 0.5.

- (a) Find the transition probability matrix.

**Solution:** If the coin is fair the transition matrix is given by

$$P = \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix}$$

If the coin is not fair the transition matrix is given by

$$P = \begin{pmatrix} 0 & p & 1-p \\ 1-p & 0 & p \\ p & 1-p & 0 \end{pmatrix}$$

- (b) Find the three-step transition probability matrix.

**Solution:** If the coin is fair three-step transition probability matrix is given by

$$P^3 = \left( \begin{pmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix} \right)^3 = \begin{pmatrix} \frac{1}{4} & \frac{3}{8} & \frac{3}{8} \\ \frac{3}{8} & \frac{1}{4} & \frac{3}{8} \\ \frac{3}{8} & \frac{3}{8} & \frac{1}{4} \end{pmatrix}$$

If the coin is not fair three-step transition probability matrix

$$P^3 = \left( \begin{pmatrix} 0 & p & 1-p \\ 1-p & 0 & p \\ p & 1-p & 0 \end{pmatrix} \right)^3 = \begin{pmatrix} p^3 - (p-1)^3 & -3p^2(p-1) & 3p(p-1)^2 \\ 3p(p-1)^2 & p^3 - (p-1)^3 & -3p^2(p-1) \\ -3p^2(p-1) & 3p(p-1)^2 & p^3 - (p-1)^3 \end{pmatrix}$$

13. Let
- $\{X_n, n \geq 0\}$
- be a sequence of random variables taking values in
- $\{1, 2, 3\}$
- . Let
- $X_0 = 2$
- and let
- $Q$
- be a matrix given by

$$Q = \begin{pmatrix} \frac{1}{2} & \frac{1}{3} & \frac{1}{6} \\ 0 & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} \end{pmatrix}$$

- (a) Can you deduce
- $P(X_n = 1 | X_{n-1} = 2)$
- and
- $P(X_n = 3 | X_{n-1} = 3)$
- from the matrix
- $Q$
- ?
- 
- (b) Find if possible
- $P(X_5 = 1 | X_4 = 3)$
- and
- $P(X_2 = 2 | X_1 = 2)$
- ?

**Solution:** There is nothing saying that  $\{X_n, n \geq 0\}$  is a Markov chain and  $Q$  is its transition probability matrix.

14. Consider the Markov chain on states 1, 2, 3 with transition probability matrix

$$P = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 \\ \frac{1}{2} & \frac{1}{2} & \frac{1}{4} \\ 0 & \frac{1}{4} & \frac{2}{3} \end{pmatrix}$$

- (a) Find the communicating classes of this M.C.

**Solution:** We have  $1 \rightarrow 2 \rightarrow 3 \rightarrow 2 \rightarrow 1$  hence  $\{1, 2, 3\}$  is the communicating class.

- (b) Is it irreducible?

**Solution:** There is only one communicating class of this Markov chain. Therefore the chain is recurrent and irreducible.

15. Consider the Markov chain on states 1, 2, 3, 4 with transition probability matrix

$$P = \begin{pmatrix} \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 \\ 0 & 0 & \frac{1}{4} & \frac{3}{4} \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

- (a) Find the communicating classes of this M.C.

**Solution:** We have  $1 \rightarrow 2 \rightarrow 1$  and  $3 \rightarrow 4$  hence  $\{1, 2\}$ ,  $\{3\}$  and  $\{4\}$  are the three communicating classes of the Markov chain. The class  $\{1, 2\}$  is aperiodic and the classes  $\{1, 2\}$ ,  $\{4\}$  recurrent, the state 4 is absorbing, and  $\{3\}$  is transient.

- (b) Is it irreducible ?

**Solution:** The Markov chain is not irreducible since it has three communicating classes.

16. Let  $\{X_n, n \geq 0\}$  be a Markov chain taking values in  $\{1, 2, 3\}$ . Consider the following matrix

$$P = \begin{pmatrix} a & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & b & \frac{1}{2} \\ \frac{1}{2} & \frac{1}{2} & c \end{pmatrix}$$

- (a) Find  $a$ ,  $b$  and  $c$  such that  $P$  is a transition probability matrix  $\{X_n, n \geq 0\}$ .

**Solution:** We should have  $a = b = c = 0$ .

- (b) Specify communicating classes of the Markov chain  $\{X_n, n \geq 0\}$ .

**Solution:** We have  $1 \rightarrow 2 \rightarrow 3 \rightarrow 1$  hence the chain has one communicating class  $\{1, 2, 3\}$ , then is irreducible.

- (c) Determine whether these classes are transient or recurrent

**Solution:** The class  $\{1, 2, 3\}$  is recurrent.

- (d) Do the same for the following matrices corresponding to some Markov chains

$$\mathbf{P} = \begin{pmatrix} \frac{1}{2} & 0 & 0 & \frac{1}{2} \\ 0 & 0 & 0 & 1 \\ \frac{2}{3} & \frac{1}{3} & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix}, \quad \mathbf{Q} = \begin{pmatrix} \frac{1}{4} & \frac{3}{4} & 0 & 0 & 0 \\ \frac{1}{2} & \frac{1}{2} & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{3} & \frac{2}{3} & 0 \\ 1 & 0 & 0 & 0 & 0 \end{pmatrix}, \quad \mathbf{R} = \begin{pmatrix} \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} & 0 & 0 \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & 0 \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 0 & 0 & \frac{1}{2} & \frac{1}{2} \end{pmatrix}$$

**For P:** Assume states are 1, 2, 3, 4. We have  $1 \rightarrow 4 \rightarrow 3 \rightarrow 2 \rightarrow 4 \rightarrow 3 \rightarrow 1$  there is only one communicating class  $\{1, 2, 3, 4\}$ . The Markov chain is irreducible

**For Q:** Assume states are 1, 2, 3, 4, 5. We have  $5 \rightarrow 1 \rightarrow 2 \rightarrow 1$  and  $4 \rightarrow 3$  the classes are  $\{1, 2\}$ ,  $\{3\}$ ,  $\{4\}$ ,  $\{5\}$ . The class  $\{1, 2\}$  is recurrent and aperiodic  $\{4\}$  and  $\{5\}$  are transient  $\{3\}$  is recurrent (absorbing), so the Markov chain is not irreducible

**For R:** Assume states are 1, 2, 3, 4, 5. We have  $1 \rightarrow 3 \rightarrow 1$  and  $4 \rightarrow 5 \rightarrow 4$  the communicating classes are  $\{1, 3\}$ ,  $\{2\}$ ,  $\{4, 5\}$ . So the chain is not irreducible. The two classes recurrent  $\{1, 3\}$  and  $\{4, 5\}$  are aperiodic and recurrent and the class  $\{2\}$  is transient.