

# Stochastic Processes

## Homework 5

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### Problem 1

- $X_n$  is our Markov chain with state space  $\mathbb{S} = \{(S,S), (S,C), (C,S), (C,C)\}$
- $X_0$  corresponds to day 0 and 1 where it is given that  $X_0 = (S,S)$
- $X_1$  corresponds to day 2 and 3, and  $X_2$  corresponds to day 4 and 5
- We are interested in day 5 being sunny which means day 4 can be sunny or cloudy, therefore by using the law of total probability, this state of day 5 being sunny can be computed as:

$$\Pr(X_2 = (S,S) \mid X_0 = (S,S)) + \Pr(X_2 = (C,S) \mid X_0 = (S,S))$$

- Using n-step transition probability for a Markov chain, we can rewrite the above probability as:

$$P_{(S,S),(S,S)}^{(2)} + P_{(S,S),(C,S)}^{(2)} ; \text{ where } P \text{ is the given transition probability matrix}$$

- $P^{(2)}$  is computed by  $P * P$ :

$$P^{(2)} = \begin{matrix} & \begin{matrix} (S,S) & (S,C) & (C,S) & (C,C) \end{matrix} \\ \begin{matrix} (S,S) \\ (S,C) \\ (C,S) \\ (C,C) \end{matrix} & \begin{bmatrix} 0.49 & 0.21 & 0.12 & 0.18 \\ 0.20 & 0.20 & 0.12 & 0.48 \\ 0.35 & 0.15 & 0.20 & 0.30 \\ 0.10 & 0.10 & 0.16 & 0.64 \end{bmatrix} \end{matrix}$$

- Finally, the probability that it is sunny on day 5 is:

$$P_{(S,S),(S,S)}^{(2)} + P_{(S,S),(C,S)}^{(2)} = 0.49 + 0.12 = 0.61$$

- The long run fraction of sunny days is computed by the limiting distribution for sunny days, where sunny days correspond to the states  $(S,S)$  and  $(C,S)$
- So, we are looking for  $\pi_{(S,S)} + \pi_{(C,S)}$ , where  $\pi_j$  is the initial probability of state  $j \in \mathbb{S}$
- We can compute the entire limiting distribution of  $X_n$  by solving the following system of equations:

$$\begin{aligned} \pi_j &= \sum_{k \in \mathbb{S}} \pi_k P_{kj} \quad \forall j \in \mathbb{S} \\ 1 &= \sum_{k \in \mathbb{S}} \pi_k \end{aligned}$$

- Applying the above equations, we have the following system of equations:

$$\begin{aligned} \pi_{(S,S)} &= 0.7\pi_{(S,S)} + 0.0\pi_{(S,C)} + 0.5\pi_{(C,S)} + 0.0\pi_{(C,C)} \\ \pi_{(S,C)} &= 0.3\pi_{(S,S)} + 0.0\pi_{(S,C)} + 0.5\pi_{(C,S)} + 0.0\pi_{(C,C)} \\ \pi_{(C,S)} &= 0.0\pi_{(S,S)} + 0.4\pi_{(S,C)} + 0.0\pi_{(C,S)} + 0.2\pi_{(C,C)} \\ \pi_{(C,C)} &= 0.0\pi_{(S,S)} + 0.6\pi_{(S,C)} + 0.0\pi_{(C,S)} + 0.8\pi_{(C,C)} \\ 1 &= \pi_{(S,S)} + \pi_{(S,C)} + \pi_{(C,S)} + \pi_{(C,C)} \end{aligned}$$

- Solving the above system will yield:  $(\pi_{(S,S)}, \pi_{(S,C)}, \pi_{(C,S)}, \pi_{(C,C)}) = (0.25, 0.15, 0.15, 0.45)$
- Therefore, the long run fraction of sunny days is  $\pi_{(S,S)} + \pi_{(C,S)} = 0.25 + 0.15 = 0.40$

## Problem 2

- $X_n$  is our Markov chain with state space  $\mathbb{S} = \{\text{exam 1, exam 2, exam 3}\} = \{1, 2, 3\}$  where  $X_n$  represents the type of the  $n^{\text{th}}$  exam.
  - $X_n$  is a Markov chain because the probability of  $X_n$  taking a value in  $\mathbb{S}$  is dependent only on what the previous exam was,  $X_{n-1}$ , and how the class performed on  $X_{n-1}$
- We can build the transition probability matrix  $P$  by using conditional probability:

$$P_{ij} = \Pr(X_n = j \mid X_{n-1} = i) \quad \forall i, j \in \mathbb{S}$$

- By using the law of total probability and conditioning on the class' performance, we can rewrite  $P_{ij}$ :

$$\begin{aligned} P_{ij} &= \Pr(X_n = j \mid X_{n-1} = i) \\ &= \Pr(X_n = j \mid \text{class did well on } X_{n-1} = i) + \Pr(X_n = j \mid \text{class did bad on } X_{n-1} = i) \\ &= \Pr(X_n = j \mid \text{class did well on } X_{n-1} = i) * \Pr(\text{class did well on } X_{n-1} = i) + \\ &\quad \Pr(X_n = j \mid \text{class did bad on } X_{n-1} = i) * \Pr(\text{class did bad on } X_{n-1} = i) \end{aligned}$$

$$\begin{aligned} P_{11} &= \Pr(X_n = 1 \mid \text{class did well on } X_{n-1} = 1) * \Pr(\text{class did well on } X_{n-1} = 1) + \\ &\quad \Pr(X_n = 1 \mid \text{class did bad on } X_{n-1} = 1) * \Pr(\text{class did bad on } X_{n-1} = 1) \\ &= (1/3)*(0.3) + (1)*(1 - 0.3) = 0.8 \end{aligned}$$

$$\begin{aligned} P_{12} &= \Pr(X_n = 2 \mid \text{class did well on } X_{n-1} = 1) * \Pr(\text{class did well on } X_{n-1} = 1) + \\ &\quad \Pr(X_n = 2 \mid \text{class did bad on } X_{n-1} = 1) * \Pr(\text{class did bad on } X_{n-1} = 1) \\ &= (1/3)*(0.3) + (0)*(1 - 0.3) = 0.1 \end{aligned}$$

$$\begin{aligned} P_{13} &= \Pr(X_n = 3 \mid \text{class did well on } X_{n-1} = 1) * \Pr(\text{class did well on } X_{n-1} = 1) + \\ &\quad \Pr(X_n = 3 \mid \text{class did bad on } X_{n-1} = 1) * \Pr(\text{class did bad on } X_{n-1} = 1) \\ &= (1/3)*(0.3) + (0)*(1 - 0.3) = 0.1 \end{aligned}$$

$$\begin{aligned} P_{21} &= \Pr(X_n = 1 \mid \text{class did well on } X_{n-1} = 2) * \Pr(\text{class did well on } X_{n-1} = 2) + \\ &\quad \Pr(X_n = 1 \mid \text{class did bad on } X_{n-1} = 2) * \Pr(\text{class did bad on } X_{n-1} = 2) \\ &= (1/3)*(0.6) + (1)*(1 - 0.6) = 0.6 \end{aligned}$$

$$\begin{aligned} P_{22} &= \Pr(X_n = 2 \mid \text{class did well on } X_{n-1} = 2) * \Pr(\text{class did well on } X_{n-1} = 2) + \\ &\quad \Pr(X_n = 2 \mid \text{class did bad on } X_{n-1} = 2) * \Pr(\text{class did bad on } X_{n-1} = 2) \\ &= (1/3)*(0.6) + (0)*(1 - 0.6) = 0.2 \end{aligned}$$

$$\begin{aligned} P_{23} &= \Pr(X_n = 3 \mid \text{class did well on } X_{n-1} = 2) * \Pr(\text{class did well on } X_{n-1} = 2) + \\ &\quad \Pr(X_n = 3 \mid \text{class did bad on } X_{n-1} = 2) * \Pr(\text{class did bad on } X_{n-1} = 2) \\ &= (1/3)*(0.6) + (0)*(1 - 0.6) = 0.2 \end{aligned}$$

$$\begin{aligned} P_{31} &= \Pr(X_n = 1 \mid \text{class did well on } X_{n-1} = 3) * \Pr(\text{class did well on } X_{n-1} = 3) + \\ &\quad \Pr(X_n = 1 \mid \text{class did bad on } X_{n-1} = 3) * \Pr(\text{class did bad on } X_{n-1} = 3) \\ &= (1/3)*(0.9) + (1)*(1 - 0.9) = 0.4 \end{aligned}$$

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$$\begin{aligned} P_{32} &= \Pr(X_n = 2 \mid \text{class did well on } X_{n-1} = 3) * \Pr(\text{class did well on } X_{n-1} = 3) + \\ &\quad \Pr(X_n = 2 \mid \text{class did bad on } X_{n-1} = 3) * \Pr(\text{class did bad on } X_{n-1} = 3) \\ &= (1/3)*(0.9) + (0)*(1 - 0.9) = 0.3 \end{aligned}$$

$$\begin{aligned} P_{33} &= \Pr(X_n = 3 \mid \text{class did well on } X_{n-1} = 3) * \Pr(\text{class did well on } X_{n-1} = 3) + \\ &\quad \Pr(X_n = 3 \mid \text{class did bad on } X_{n-1} = 3) * \Pr(\text{class did bad on } X_{n-1} = 3) \\ &= (1/3)*(0.9) + (0)*(1 - 0.9) = 0.3 \end{aligned}$$

$$P = \begin{matrix} & \begin{matrix} 1 & 2 & 3 \end{matrix} \\ \begin{matrix} 1 \\ 2 \\ 3 \end{matrix} & \begin{bmatrix} 0.8 & 0.1 & 0.1 \\ 0.6 & 0.2 & 0.2 \\ 0.4 & 0.3 & 0.3 \end{bmatrix} \end{matrix}$$

- The proportion of exams being type  $i$  in the long run, where  $i \in \mathbb{S}$ , is represented by the entire limiting distribution of  $X_n$
- We can compute the entire limiting distribution of  $X_n$  by solving the following system of equations:

$$\begin{aligned} \pi_j &= \sum_{k \in \mathbb{S}} \pi_k P_{kj} \quad \forall j \in \mathbb{S}; \text{ where } \pi_j \text{ is the initial probability of state } j \in \mathbb{S} \\ 1 &= \sum_{k \in \mathbb{S}} \pi_k \end{aligned}$$

- Applying the above equations, we have the following system of equations:

$$\begin{aligned} \pi_1 &= 0.8\pi_1 + 0.6\pi_2 + 0.4\pi_3 \\ \pi_2 &= 0.1\pi_1 + 0.2\pi_2 + 0.3\pi_3 \\ \pi_3 &= 0.1\pi_1 + 0.2\pi_2 + 0.3\pi_3 \\ 1 &= \pi_1 + \pi_2 + \pi_3 \end{aligned}$$

- Solving the above system will yield:  $(\pi_1, \pi_2, \pi_3) = (5/7, 1/7, 1/7)$
- Therefore, the proportion of exams being type 1, 2, and 3 in the long run is  $5/7$ ,  $1/7$ , and  $1/7$  respectively

### Problem 3

- $X_n$  is our Markov chain with state space  $\mathbb{S} = \{\text{sunny, cloudy, rainy}\} = \{0, 1, 2\}$  where  $X_n$  represents the weather on day  $n$ .
  - $X_n$  is a Markov chain because the probability of  $X_n$  taking a value in  $\mathbb{S}$  is dependent only on the value that was taken by  $X_{n-1}$ , the weather of the previous day.
- We can build the transition probability matrix  $P$  by using conditional probability:

$$P_{ij} = \Pr(X_n = j \mid X_{n-1} = i) \quad \forall i, j \in \mathbb{S}$$

$$P_{00} = \Pr(X_n = 0 \mid X_{n-1} = 0) = 0$$

$$P_{01} = \Pr(X_n = 1 \mid X_{n-1} = 0) = 1/2$$

$$P_{02} = \Pr(X_n = 2 \mid X_{n-1} = 0) = 1/2$$

$$P_{10} = \Pr(X_n = 0 \mid X_{n-1} = 1) = 1/4$$

$$P_{11} = \Pr(X_n = 1 \mid X_{n-1} = 1) = 1/2$$

$$P_{12} = \Pr(X_n = 2 \mid X_{n-1} = 1) = 1/4$$

$$P_{20} = \Pr(X_n = 0 \mid X_{n-1} = 2) = 1/4$$

$$P_{21} = \Pr(X_n = 1 \mid X_{n-1} = 2) = 1/4$$

$$P_{22} = \Pr(X_n = 2 \mid X_{n-1} = 2) = 1/2$$

$$P = \begin{matrix} & \begin{matrix} 0 & 1 & 2 \end{matrix} \\ \begin{matrix} 0 \\ 1 \\ 2 \end{matrix} & \begin{bmatrix} 0 & 1/2 & 1/2 \\ 1/4 & 1/2 & 1/4 \\ 1/4 & 1/4 & 1/2 \end{bmatrix} \end{matrix}$$

- The proportion of days being sunny in the long run, and the proportion of days being cloudy in the long run can both be found in entire limiting distribution of  $X_n$
- We are interested in the values of  $\pi_0$  and  $\pi_1$  to find the proportions of interest for sunny days and cloudy days respectively, where  $\pi_j$  is the initial probability of state  $j \in \mathbb{S}$
- We can compute the entire limiting distribution of  $X_n$  by solving the following system of equations:

$$\pi_j = \sum_{k \in \mathbb{S}} \pi_k P_{kj} \quad \forall j \in \mathbb{S}$$

$$1 = \sum_{k \in \mathbb{S}} \pi_k$$

- Applying the above equations, we have the following system of equations:

$$\pi_0 = 0\pi_0 + \frac{1}{4}\pi_1 + \frac{1}{4}\pi_2$$

$$\pi_1 = \frac{1}{2}\pi_0 + \frac{1}{2}\pi_1 + \frac{1}{4}\pi_2$$

$$\pi_2 = \frac{1}{2}\pi_0 + \frac{1}{4}\pi_1 + \frac{1}{2}\pi_2$$

$$1 = \pi_0 + \pi_1 + \pi_2$$

- Solving the above system will yield:  $(\pi_0, \pi_1, \pi_2) = (1/5, 2/5, 2/5)$
- Therefore, the proportion of days being sunny and cloudy in the long run is  $1/5$  and  $2/5$  respectively

#### Problem 4

- Using conditional probability, we can rewrite the given limit as:

$$\lim_{n \rightarrow \infty} \Pr(X_{n-1} = 2 \mid X_n = 1) = \lim_{n \rightarrow \infty} \frac{\Pr(X_n=1 \mid X_{n-1}=2) * \Pr(X_{n-1}=2)}{\Pr(X_n=1)}$$

- Since we are taking the limit of  $n \rightarrow \infty$  this means we want to know the long run value of those probabilities, where:

$$\lim_{n \rightarrow \infty} \Pr(X_n = j \mid X_{n-1} = i) = P_{ij} ; \text{ where } P \text{ is the given transition probability matrix}$$

$$\lim_{n \rightarrow \infty} \Pr(X_n = j) = \pi_j ; \text{ where } \pi_j \text{ is the initial probability of state } j \in \mathbb{S} = \{0, 1, 2\}$$

- The value of  $\pi_j \forall j \in \mathbb{S}$  is the entire limiting distribution of the Markov Chain,  $X_n$
- We can compute the entire limiting distribution of  $X_n$  by solving the following system of equations:

$$\pi_j = \sum_{k \in \mathbb{S}} \pi_k P_{kj} \quad \forall j \in \mathbb{S}$$

$$1 = \sum_{k \in \mathbb{S}} \pi_k$$

- Applying the above equations, we have the following system of equations:

$$\pi_0 = 0.4\pi_0 + 0.6\pi_1 + 0.4\pi_2$$

$$\pi_1 = 0.4\pi_0 + 0.2\pi_1 + 0.2\pi_2$$

$$\pi_2 = 0.2\pi_0 + 0.2\pi_1 + 0.4\pi_2$$

$$1 = \pi_1 + \pi_2 + \pi_3$$

- Solving the above system will yield:  $(\pi_0, \pi_1, \pi_2) = (11/24, 7/24, 1/4)$
- Using the given matrix  $P$  and the computed values of  $(\pi_0, \pi_1, \pi_2)$  we can compute the given limit:

$$\lim_{n \rightarrow \infty} \Pr(X_{n-1} = 2 \mid X_n = 1) = \lim_{n \rightarrow \infty} \frac{\Pr(X_n=1 \mid X_{n-1}=2) * \Pr(X_{n-1}=2)}{\Pr(X_n=1)} = \frac{P_{21} * \pi_2}{\pi_1} = \frac{(0.2)*(1/4)}{(7/24)} = 6/35$$

## Problem 5

- When the transition probability matrix has row sums and column sum both equal to one, then this is known as a doubly stochastic transition probability matrix. This means that the given  $P$  satisfies the following conditions:

$$P_{ij} \geq 0 \quad \forall i, j \in \mathbb{S} = \{1, 2, \dots, M\}$$

$$\sum_{k \in \mathbb{S}} P_{ik} = \sum_{k \in \mathbb{S}} P_{kj} = 1 \quad \forall i, j \in \mathbb{S} = \{1, 2, \dots, M\}$$

- When a transition probability matrix  $P$  satisfies the above conditions (ie.  $P$  is doubly stochastic), then when computing the entire limiting distribution of  $X_n$  there is only one solution to the following system of equations:

$$\pi_j = \sum_{k \in \mathbb{S}} \pi_k P_{kj} \quad \forall j \in \mathbb{S}$$

$$1 = \sum_{k \in \mathbb{S}} \pi_k$$

- The solution to the above system when  $P$  is doubly stochastic is the following:

$$\pi_j = \sum_{k \in \mathbb{S}} \frac{1}{M} P_{kj} = \frac{1}{M} \quad \forall j \in \mathbb{S}; \text{ where } M \text{ is the size of the state space}$$

- Therefore, the limiting probability vector  $\underline{\pi} = (\pi_1, \pi_2, \dots, \pi_M) = (1/M, 1/M, \dots, 1/M)$