

# Solution proposal STK2130-sp19

## Problem 1

a) A class is defined as a set of states which communicate.

One class is  $\{2, 4\}$  since the paths  $2, 4, 2$ , and  $4, 2, 4$  are possible so 2 and 4 communicate

The second class is  $\{1, 5\}$  since the paths  $1, 5, 1$ , and  $5, 1, 5$  are possible so 1 and 5 communicate. The state  $\{3\}$  is absorbing and therefore is a class of its own.

If the chain enters  $\{1, 5\}$ , it stays there so there is an infinite number of visits to this class and the class is recurrent. If the initial state is 2 the path  $2, 4, 5$  is possible and the chain does not return to 2. Hence the class  $\{2, 4\}$  is transient. The state  $\{3\}$  is absorbing, and hence is recurrent since the chain stays in 1 always, i.e. an infinite number of times.

b) From the Chapman-Kolmogorov equations

$$P_{45}^2 = \sum_{k=1}^5 P_{4k}P_{k5} = (0, 1/4, 1/4, 1/4, 1/4) \begin{pmatrix} 1/2 \\ 0 \\ 0 \\ 1/4 \\ 1/2 \end{pmatrix} = 1/16 + 2/16 = 3/16.$$

c) The transition matrix for the chain moving outside  $\{1, 3, 5\}$  and being absorbed in  $\{1, 3, 5\}$  is when the states are  $\{2, 4, A\}$

$$Q = \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/4 & 1/4 & 1/2 \\ 0 & 0 & 1 \end{pmatrix}.$$

Then  $P(X_3 = 2, X_k \notin \{1, 3, 5\}, k = 1, 2 | X_0 = 4) = Q_{42}^3$  But

$$\begin{aligned} Q^3 &= \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/4 & 1/4 & 1/2 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/4 & 1/4 & 1/2 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/4 & 1/4 & 1/2 \\ 0 & 0 & 1 \end{pmatrix} \\ &= \begin{pmatrix} 0 & 1/2 & 1/2 \\ 1/4 & 1/4 & 1/2 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1/8 & 1/8 & 3/4 \\ 1/16 & 3/16 & 6/8 \\ 0 & 0 & 1 \end{pmatrix} \end{aligned}$$

so

$$Q_{42}^3 = \sum_{k=1}^5 Q_{4k}Q_{k2}^2 = (1/4, 1/4, 1/2) \begin{pmatrix} 1/8 \\ 1/16 \\ 0 \end{pmatrix} = (1/4)(1/8) + (1/4)(1/16) = 3/64.$$

Remark that there are only two possible paths:  $4, 4, 4, 2$  and  $4, 2, 4, 2$  with probabilities  $(1/4)(1/4)(1/4) = 1/64$  and  $(1/4)(1/2)(1/4) = 1/32$ .

d)

$$\begin{aligned}
& P(X_3 = 1, X_k \notin \{1, 3, 5\}, k = 1, 2 | X_0 = 4). \\
= & P(X_3 = 1, X_2 = 2, X_1 \notin \{1, 3, 5\} | X_0 = 4) \\
+ & P(X_3 = 1, X_2 = 4, X_1 \notin \{1, 3, 5\} | X_0 = 4) \\
= & P(X_3 = 1 | X_2 = 2, X_1 \notin \{1, 3, 5\}, X_0 = 4) P(X_2 = 2, X_1 \notin \{1, 3, 5\}, | X_0 = 4) \\
+ & P(X_3 = 1 | X_2 = 4, X_1 \notin \{1, 3, 5\}, X_0 = 4) P(X_2 = 4, X_1 \notin \{1, 3, 5\} | X_0 = 4) \\
= & Q_{42}^2 P_{21} + Q_{44}^2 P_{41}
\end{aligned}$$

where  $P(X_3 = 1 | X_2 = 2, X_1 \notin \{1, 3, 5\}, X_0 = 4) = P_{21} = 1/2$  and  $P(X_3 = 1 | X_2 = 4, X_1 \notin \{1, 3, 5\}, X_0 = 4) = P_{41} = 0$  follow by the Markov property. Thus  $Q_{42}^2 P_{21} + Q_{44}^2 P_{41} = (1/16)(1/2) = 1/32$  since  $Q_{42}^2 = 1/16$

Here there is only one possible path: 4,4,2,1 with probability  $(1/4)(1/4)(1/2) = 1/32$ .

e) If  $X_0 \in \{1, 3, 5\}$ ,  $T = 0$  so  $\mu_i = E[T | X_0 = i] = 0, k \in \{1, 3, 5\}$ . Also

$$\begin{aligned}
\mu_2 &= \sum_{i=1}^5 E[T, X_1 = i | X_0 = 2] \\
&= \sum_{i=1}^5 E[T | X_1 = i, X_0 = 2] P(X_1 = i | X_0 = 2) \\
&= \sum_{i=1}^5 1 + E[T | X_0 = i] P(X_1 = i | X_0 = 2)
\end{aligned}$$

since by the Markov property  $E[T | X_k = i, X_{k-1} = j] = 1 + E[T | X_{k-1} = i, X_{k-2} = j] = 1 + E[T | X_{k-1} = i]$ . Thus  $\mu_2 = 1 + \mu_1 P_{22} + \mu_4 P_{24} = 1 + \mu_4/2$ . Similarly  $\mu_4 = 1 + \mu_2 P_{42} + \mu_4 P_{44} = 1 + \mu_2/4 + \mu_4/4$ . The equations

$$\begin{aligned}
\mu_2 &= 1 + \mu_4/2 \\
\mu_4 &= 1 + \mu_2/4 + \mu_4/4
\end{aligned}$$

have solutions  $\mu_2 = 2$  and  $\mu_4 = 2$ .

5) The class  $\{1, 5\}$  is a closed class so once the chain enters the class it stays there. Hence  $\pi_5$  is the limit of the proportion of time the chain is in state 5. Similarly  $\pi_1 = \lim_{n \rightarrow \infty} P(X_n = 1 | X_0 = 1)$  is the limit of the proportion of time the chain is in state 1.  $(\pi_1, \pi_5)$  is the solution of the equations

$$(\pi_1, \pi_5) = (\pi_1, \pi_5) \begin{pmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{pmatrix}$$

and  $\pi_1 + \pi_5 = 1$  so  $(\pi_1, \pi_5) = (1/2, 1/2)$ .

## Problem 2

a) A birth and death process is a continuous time Markov chain with state space  $0, 1, 2, \dots$ . When the chain is in state  $i$  the times until the next change of state are independent exponentially distributed with mean  $1/v_i$  where  $v_0 = \lambda_0$  and  $v_i = \lambda_i + \mu_i, i = 1, 2, \dots$ . The move to the next state is described by a binary random variable which is independent of how long the chain is in state  $i$  and has a distribution where the probability that the change is to  $i + 1$  is  $\lambda_i/(\lambda_i + \mu_i)$  the probability that the change is to  $i - 1$  is  $\mu_i/(\lambda_i + \mu_i)$   $i = 1, 2, \dots$  and the probability that the move is to 1 if  $i = 0$  is 1.

b) The state is the number of customers so the state space is  $0, 1, \dots, s$ . The chain moves from  $i$  to  $i+1$   $i = 0, 1, \dots, s - 1$  when a new customer arrives so  $\lambda_i = \lambda$   $i = 0, \dots, s - 1$ . If no server is free so the state is  $i = s$  the new customer leaves so  $\lambda_s = 0$ . If  $i$  servers are busy the chain moves from  $i$  to  $i - 1, i = 1, \dots, s$  when the first server is free. This variable is the minimum of  $i$  independent exponentially distributed variable, which is an exponentially distributed variable with expectation  $1/i\mu$

Hence  $v_0 = \lambda, v_i = \lambda + i\mu, i = 1, \dots, s - 1$  and  $v_s = s\mu$ . The transition matrix of the jumps has elements 0 except  $P_{0,1}$  and  $P_{i,i+1} = \lambda/(\lambda + i\mu), P_{i,i-1} = \mu/(\lambda + i\mu), i = 1, \dots, s - 1$  and  $P_{s,s-1} = 1$ .

The instantaneous transition rates are therefore  $q_{01} = \lambda, q_{i,i+1} = \lambda, i = 1, \dots, s - 1$   $q_{i,i-1} = i\mu, i = 1, \dots, s - 1$  and  $q_{s,s-1} = s\mu$ .

c) The Kolmogorov backward equations have the form

$$P'_{ij}(t) = \sum_{k \neq i} q_{ik} P_{kj}(t) - v_i P_{ij}(t).$$

With the results from part b)

$$\begin{aligned} P'_{0j}(t) &= \lambda P_{1j}(t) - \lambda P_{0j}(t) \\ P'_{i,j}(t) &= \lambda P_{i,i+1}(t) + i\mu P_{i-1,j}(t) - (\lambda + i\mu) P_{i,j}(t), \quad i = 1, \dots, s - 1 \\ P'_{s,j}(t) &= s\mu P_{s-1,j}(t) - s\mu P_{s,j}(t). \end{aligned}$$

d) The balance equations are

$$v_j P_j = \sum_{k \neq j} q_{kj} P_k$$

where  $P_j$  are the limiting probabilities. In this case

$$\begin{aligned} v_0 P_0 &= q_{10} P_1 & \text{i.e. } \lambda P_0 &= \mu P_1 \\ v_i P_i &= q_{i-1,i} P_{i-1} + q_{i+1,i} P_{i+1} & \text{i.e. } (\lambda + i\mu) P_i &= \lambda P_{i-1} + (i + 1)\mu P_{i+1}, \quad i = 1, \dots, s - 1 \\ v_s P_s &= q_{s-1,s} P_{s-1} & \text{i.e. } s\mu P_s &= \lambda P_{s-1}. \end{aligned}$$

- e) First  $P_1 = \frac{\lambda}{\mu}P_0$ . Inserting in the next equation  $(\lambda + \mu)P_1 = \lambda P_0 + 2\mu P_2$ , i.e.  $(\lambda + \mu)P_1 = \mu P_1 + 2\mu P_2$  or  $P_2 = \frac{\lambda}{2\mu}P_1$ . If  $P_i = \frac{\lambda}{i\mu}P_{i-1}$ ,  $(\lambda + i\mu)P_i = (i+1)\mu P_{i+1} + \lambda P_{i-1} = (i+1)\mu P_{i+1} + i\mu P_i$ , so  $P_{i+1} = \frac{\lambda}{(i+1)\mu}P_i$  for  $i = 1, \dots, s-1$ . Also  $P_s = \frac{\lambda}{s\mu}P_{s-1}$ .

$$\text{Hence, } P_i = \prod_{j=0}^i \frac{\lambda}{\mu} \cdots \frac{\lambda}{i\mu} = \left(\frac{\lambda}{\mu}\right)^i \frac{1}{i!} P_0 \text{ and } P_i = \frac{\left(\frac{\lambda}{\mu}\right)^i \frac{1}{i!}}{\sum_{j=0}^s \left(\frac{\lambda}{\mu}\right)^j \frac{1}{j!}}.$$

### Problem 3

- a) It is impossible for the chain to return to 0 after an odd number of steps since it moves one step up with probability  $p$  or one step down with probability  $1-p$ . If the chain returns to 0 after  $2n$  steps, there must have been  $n$  steps up and  $n$  steps down. There are  $\binom{2n}{n}$  possible locations of the  $n$  steps upward and each of them occur with probability  $p$ . Hence  $P_{00}^{2n} = \binom{2n}{n} p^n (1-p)^n$
- b) From Stirling's approximation

$$\begin{aligned} (2n)! &= \frac{(2n)^{2n} \sqrt{(2\pi 2n)}}{e^{2n}} = \frac{2^{2n+1} \sqrt{\pi} n^{2n+1/2}}{e^{2n}} \\ (n!)^2 &= \frac{n^{2n} 2\pi n}{e^{2n}} = \frac{2\pi n^{2n+1}}{e^{2n}}. \end{aligned}$$

Hence  $\frac{(2n)!}{(n!)^2} = \frac{2^{2n} \sqrt{\pi}}{\pi \sqrt{n}} = \frac{4^n}{\sqrt{\pi n}}$  and  $P_{00}^{2n} = \binom{2n}{n} p^n (1-p)^n$  is approximately  $\frac{[4p(1-p)]^n}{\sqrt{\pi n}}$ .

Therefore  $\sum_{n=1}^{\infty} P_{00}^{2n} = \infty$  if and only if  $p = 1/2$  and the chain is recurrent if  $p = 1/2$  and transient if  $p \neq 1/2$ .

- c) By decomposing the event  $T_1 < \infty$  according to whether  $X_1 = 1$  or  $X_1 = -1$   $f = P(T_1 < \infty | X_0 = 0) = P(T_1 < \infty, X_1 = 1 | X_0 = 0) + P(T_1 < \infty, X_1 = -1 | X_0 = 0)$ . But

$$\begin{aligned} &P(T_1 < \infty, X_1 = 1 | X_0 = 0) \\ &= P(T_1 < \infty, X_1 = 1, X_0 = 0) / P(X_0 = 0) \\ &= P(T_1 < \infty | X_1 = 1, X_0 = 0) P(X_1 = 1, X_0 = 0) / P(X_0 = 0) \\ &= P(T_1 < \infty | X_1 = 1) P(X_1 = 1 | X_0 = 0) \end{aligned}$$

where we have used the Markov property. But  $P(T_1 < \infty | X_1 = 1) = 1$  and  $P(X_1 = 1 | X_0 = 0) = p$ , so  $P(T_1 < \infty, X_1 = 1 | X_0 = 0) = p$ .

Next consider  $P(T_1 < \infty, X_1 = -1 | X_0 = 0)$  which by the same calculations equals  $P(T_1 < \infty | X_1 = -1) P(X_1 = -1 | X_0 = 0)$ . But to reach 1 starting in  $-1$  the chain must first reach 0, i.e one unit about the starting value which has probability  $P(T_0 < \infty, | X_0 = -1)$  and is equal to  $P(T_1 < \infty | X_0 = 0)$ . Then having reach 0 the chain must reach 1, which also has probability  $f = P(T_1 < \infty | X_0 = 0)$ . Thus  $P(T_1 < \infty | X_1 = -1) = f^2$  since by the Markov property the two events must be independent. Knowing that  $X_n = 0$

how the chain has reach this state is independent of the future behavior, so  $P(T_1 < \infty | X_1 = -1)P(X_1 = -1 | X_0 = 0) = f^2q$  and

$$f = p + qf^2.$$

The fact that  $p < q$  means that the chain has a drift downward. The probability  $P(T_1 < \infty | X_0 = 0)$  is the same as the probability for absorption if the state 1 has been defined as an absorbing state. That this event should have probability 1 when there is a downward drift is not reasonable. Hence  $f = p/q$  is the natural choice.

- d) Decomposing By decomposing the event  $T_0 < \infty$  according to whether  $X_1 = 1$  or  $X_1 = -1$  and arguing as in part c)  $P(T_0 < \infty | X_0 = 0) = P(T_0 < \infty | X_1 = 1)P(X_1 = 1 | X_0 = 0) + P(T_0 < \infty | X_1 = -1)P(X_1 = -1 | X_0 = 0)$  But from part c)  $P(T_0 < \infty | X_1 = -1) = p/q$  so  $(T_0 < \infty | X_1 = -1)P(X_1 = -1 | X_0 = 0) = (p/q)q = p$ .

Write  $X_n = \sum Z_i$  where  $Z_i = 1$  if the  $i$ th step is up and  $X_i = -1$  if the  $i$ th step is down and  $Z_1, Z_2, \dots$  are independent.  $E[Z_i] = p - q = 2p - 1 < 0$ . By the strong law of large numbers  $\frac{1}{n} \sum_{i=1}^n Z_i \rightarrow 2p - 1 < 0$  with probability 1. Hence  $X_n \rightarrow -\infty$  with probability 1 so  $P(T_0 < \infty | X_1 = 1) = 1$  and  $P(T_0 < \infty | X_1 = 1)P(X_1 = 1 | X_0 = 0) = p$ .

Therefore  $P(T_0 < \infty | X_0 = 0) = 2p$ .

Using the results from part d) the heuristic argument for choosing the solution  $p/q$  in part c) can be made rigorous. If the solution had been 1, the probability  $P(T_0 < \infty | X_0 = 0)$  would have been  $1 \times p + 1 \times q = 1$ . Thus there would have been an infinite number of returns to 0, which contradicts that the chain is transient when  $p < 1/2$ .