

## Homework 7

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### Chapter 6: Continuous-Time Markov Chains

2. Suppose that a one-celled organism can be in one of two states - either  $A$  or  $B$ . An individual in state  $A$  will change to state  $B$  at an exponential rate  $\alpha$ ; an individual in state  $B$  divides into two new individuals of type  $A$  at an exponential rate  $\beta$ . Define an appropriate continuous-time Markov chain for a population of such organisms and determine the appropriate parameters for this model.

*Solution.* Let  $N_A(t)$  and  $N_B(t)$  represent the number of individuals in states  $A$  and  $B$ , respectively. Then  $\{(N_A(t), N_B(t))\}$  is a Markov process. If there are  $a$  in state  $A$  and  $b$  in state  $B$ , then the total transition rate is  $v_{(a,b)} = a\alpha + b\beta$ . The individual transition probabilities are

$$P_{(a,b),(a-1,b+1)} = \frac{a\alpha}{a\alpha + b\beta}$$

$$P_{(a,b),(a+2,b-1)} = \frac{b\beta}{a\alpha + b\beta}$$

□

5. There are  $N$  individuals in a population, some of whom have a certain infection that spreads as follows. Contacts between two members of this population occur in accordance with a Poisson process having rate  $\lambda$ . When a contact occurs, it is equally likely to involve any of the  $\binom{N}{2}$  pairs of individuals in the population. If a contact involves an infected and non-infected individual, then with probability  $p$  the non-infected individual becomes infected. Once infected, an individual remains infected throughout. Let  $X(t)$  denote the number of infected members of the population at time  $t$ .

- (a) Is  $\{X(t), t \geq 0\}$  a continuous-time Markov chain?

**Answer.** If the states are  $\{0, 1, \dots, N\}$  representing the number of infected people, then since contacts are a Poisson process, it follows that the transition rates are exponentially distributed, so the process is memory-less. Thus, it is a Markov process.

- (b) Specify its type.

**Answer.** This is a pure birth process, since individuals cannot become non-infected.

- (c) Starting with a single infected individual, what is the expected time until all members are infected?

*Solution.* If there are  $i$  infected individuals, there are  $N - i$  non-infected individuals, so  $i(N - i)$  contacts between them. There are a total of  $\binom{N}{2}$  possible contacts, so the probability of a contact between infected and non-infected is  $i(N - i)/\binom{N}{2}$ , and the probability of infection is  $pi(N - i)/\binom{N}{2}$ , so the birth rates are

$$\lambda_i = \frac{\lambda pi(N - i)}{\binom{N}{2}}, \quad i = 1, 2, \dots, N - 1$$

If  $T_i$  represents the time to transition from  $i$  infected individuals to  $i + 1$ , then we seek

$$E\left[\sum_{i=1}^{N-1} T_i\right] = \sum_{i=1}^{N-1} E[T_i] = \sum_{i=1}^{N-1} \frac{1}{\lambda_i} = \sum_{i=1}^{N-1} \frac{\binom{N}{2}}{\lambda pi(N - i)} = \frac{N(N - 1)}{2\lambda p} \sum_{i=1}^{N-1} \frac{1}{i(N - i)}$$

□

6. Consider a birth and death process with birth rates  $\lambda_i = (i+1)\lambda, i \geq 0$ , and death rates  $\mu_i = i\mu, i \geq 0$ .

(a) Determine the expected time to go from state 0 to state 4.

*Solution.* We use the recursive formula

$$E[T_i] = \frac{1}{\lambda_i} + \frac{\mu_i}{\lambda_i} E[T_{i-1}] = \frac{1}{(i+1)\lambda} (1 + i\mu E[T_{i-1}])$$

Starting with  $E[T_0] = 1/\lambda_0 = 1/\lambda$ , we have

$$\begin{aligned} E[T_1] &= \frac{1}{2\lambda} \left( 1 + \mu \cdot \frac{1}{\lambda} \right) = \frac{\lambda + \mu}{2\lambda^2} \\ E[T_2] &= \frac{1}{3\lambda} \left( 1 + 2\mu \cdot \frac{\lambda + \mu}{2\lambda^2} \right) = \frac{\lambda^2 + \lambda\mu + \mu^2}{3\lambda^3} \\ E[T_3] &= \frac{1}{4\lambda} \left( 1 + 3\mu \cdot \frac{\lambda^2 + \lambda\mu + \mu^2}{3\lambda^3} \right) = \frac{\lambda^3 + \lambda^2\mu + \lambda\mu^2 + \mu^3}{4\lambda^4} \end{aligned}$$

The expected time to go from state 0 to state 4 is thus

$$\begin{aligned} E[T_0 + T_1 + T_2 + T_3] &= E[T_0] + E[T_1] + E[T_2] + E[T_3] \\ &= \frac{1}{\lambda} + \frac{\lambda + \mu}{2\lambda^2} + \frac{\lambda^2 + \lambda\mu + \mu^2}{3\lambda^3} + \frac{\lambda^3 + \lambda^2\mu + \lambda\mu^2 + \mu^3}{4\lambda^4} \end{aligned}$$

□

(b) Determine the expected time to go from state 2 to state 5.

*Solution.* Using the same recursive formula, we have

$$E[T_4] = \frac{1}{5\lambda} \left( 1 + 4\mu \cdot \frac{\lambda^3 + \lambda^2\mu + \lambda\mu^2 + \mu^3}{4\lambda^4} \right) = \frac{\lambda^4 + \lambda^3\mu + \lambda^2\mu^2 + \lambda\mu^3 + \mu^4}{5\lambda^5}$$

so the expected time to go from state 2 to state 5 is

$$\begin{aligned} E[T_2 + T_3 + T_4] &= E[T_2] + E[T_3] + E[T_4] \\ &= \frac{\lambda^2 + \lambda\mu + \mu^2}{3\lambda^3} + \frac{\lambda^3 + \lambda^2\mu + \lambda\mu^2 + \mu^3}{4\lambda^4} + \frac{\lambda^4 + \lambda^3\mu + \lambda^2\mu^2 + \lambda\mu^3 + \mu^4}{5\lambda^5} \end{aligned}$$

□

(c) Determine the variances in parts (a) and (b).

*Solution.* We use the recursive formula

$$\begin{aligned} \text{Var}(T_i) &= \frac{1}{\lambda_i(\lambda_i + \mu_i)} + \frac{\mu_i}{\lambda_i} \text{Var}(T_{i-1}) + \frac{\mu_i}{\lambda_i + \mu_i} (E[T_{i-1}] + E[T_i])^2 \\ &= \frac{1}{(i+1)\lambda[(i+1)\lambda + i\mu]} + \frac{i\mu}{(i+1)\lambda} \text{Var}(T_{i-1}) + \frac{i\mu}{(i+1)\lambda + i\mu} (E[T_{i-1}] + E[T_i])^2 \end{aligned}$$

Starting with  $\text{Var}(T_0) = 1/\lambda_0^2 = 1/\lambda^2$ , we have

$$\text{Var}(T_1) = \frac{1}{2\lambda(2\lambda + \mu)} + \frac{\mu}{2\lambda} \cdot \frac{1}{\lambda^2} + \frac{\mu}{2\lambda + \mu} \left( \frac{1}{\lambda} + \frac{\lambda + \mu}{2\lambda^2} \right)^2$$

and etc... The algebra is pretty ugly and I think unnecessary, but the variances of parts (a) and (b), respectively, are

$$\begin{aligned} \text{Var}(T_0 + T_1 + T_2 + T_3) &= \text{Var}(T_0) + \text{Var}(T_1) + \text{Var}(T_2) + \text{Var}(T_3) \\ \text{Var}(T_2 + T_3 + T_4) &= \text{Var}(T_2) + \text{Var}(T_3) + \text{Var}(T_4) \end{aligned}$$

□

9. The birth and death process with parameters  $\lambda_n = 0$  and  $\mu_n = \mu, n > 0$  is called a pure death process. Find  $P_{ij}(t)$ .

*Solution.* The death rate is constant, so since the deaths arrive according to a Poisson process with rate  $\mu$ , in any interval of length  $t$ , the probability we go from state  $i$  to state  $j > 0$  ( $i - j$  death arrivals) is

$$P_{ij}(t) = e^{-\mu t} \frac{(\mu t)^{i-j}}{(i-j)!}, \quad 0 < j \leq i$$

Then for  $P_{i0}(t)$ , this is the complement of the probability that there is still some positive number of individuals remaining in the population (i.e. there are fewer than  $i$  arrivals in a period of time  $t$ ). Thus

$$P_{i0}(t) = 1 - \sum_{k=0}^{i-1} e^{-\mu t} \frac{(\mu t)^k}{k!} = \sum_{k=i}^{\infty} e^{-\mu t} \frac{(\mu t)^k}{k!}$$

□