MATH4240 Tutorial 10 Notes

1 Basics

- A random variable X (with value in [0,∞)) is memoryless if P (X > t + s | X > t) = P (X > s).
 If X is continuous, it must be exponentially distributed.
 If X is discrete, it must be geometrically distributed.
- If $\xi_i \sim \text{Exp}(\lambda_i)$ are independent, then $\min(\xi_1, \ldots, \xi_n) \sim \text{Exp}(\sum \lambda_j)$ and $P(\min(\xi_1, \ldots, \xi_n) = \xi_i) = \frac{\lambda_i}{\sum \lambda_j}$ (see Tutorial 1)
- For a Markov jump process, the transition function $P_{xy}(t)$ satisfies Chapman-Kolmogorov equation

$$P(t+s) = P(t)P(s)$$

that is, $P_{t+s}(x,y) = \sum_{z} P_t(x,z)P_s(z,y)$

- The rate matrix / generator $D = [q_{xy}] = P'(0) = \lim_{t \to 0^+} \frac{P(t) I}{t}$ of a Markov jump process satisfies
 - $-q_{xy} \ge 0$ for $x \ne y$ and $-q_x = q_{xx} \le 0$
 - $-\sum_{y} q_{xy} = 0$, or $q_x = \sum_{y \neq x} q_{xy}$
 - (forward equation) P' = PD, that is, $P'_{xy}(t) = \sum_z P_{xz}(t)D_{zy}$. In particular, $\pi' = \pi D$
 - (backward equation) P' = DP, that is, $P'_{xy}(t) = \sum_z D_{xz} P_{zy}(t)$

In particular, $P(t) = \exp(tD)$.

(Compare with discrete-time case where D = P - I)

• If x is not an absorbing state, then the transition probability Q_{xy} of the embedded Markov chain / jump chain for $x \neq y$ is

$$Q_{xy} = q_{xy}/q_x = \frac{q_{xy}}{\sum_{z \neq x} q_{xz}}$$

And if x is absorbing, $Q_{xy} = 0$ for $x \neq y$ and $Q_{xx} = 1$.

• Recall that in last session we talked about an alternative definition of Poisson process:

Definition 1. A monotone increasing random process X_t with natural number value is a Poisson process with rate λ if it has independent increment, $X_0 = 0$, and

$$P(X_{t+h} - X_t = 0) = 1 - \lambda h + o(h)$$

$$P(X_{t+h} - X_t = 1) = \lambda h + o(h)$$

Taking t = 0 we can see that

$$\frac{1}{h} (P(X_{t+h} = X_t) - 1) = -\lambda + o(1)$$
$$\frac{1}{h} (P(X_{t+h} = X_t + 1) - 0) = \lambda + o(1)$$

which means that the rate matrix and the jump matrix are of the form

$$D = \begin{pmatrix} -\lambda & \lambda & 0 & 0 & \dots \\ 0 & -\lambda & \lambda & 0 & \dots \\ \vdots & \ddots & \ddots & \ddots & \ddots \end{pmatrix}, \qquad Q = \begin{pmatrix} 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & 1 & 0 & \dots \\ \vdots & \ddots & \ddots & \ddots & \ddots \end{pmatrix}$$

So to specify a Markov jump process, it suffices to specify the rate matrix.

Example 1. Suppose at time t > 0 a Poisson process X_t with rate λ has exactly n arrivals $X_t = n$. Let us compute the distribution of the arrival times T_1, \ldots, T_n , that is the pdf $f(t_1, \ldots, t_n \mid X_t = n)$. Easy to see that

$$\{ T_1 = t_1, \dots, T_n = t_n, X_t = n \} = \{ \xi_1 = t_1, \xi_n = t_n - t_{n-1}, \xi_{n+1} > t - t_n \}$$

and by assumption, the waiting times are independent. These imply

$$f_{T_1,...,T_n}(t_1,...,t_n \mid X_t = n) = f_{\xi_1}(t_1) \dots f_{\xi_n}(t_n - t_{n-1}) P(X_{n+1} > t - t_n) / P(X_t = n)$$
$$= \frac{\lambda e^{-\lambda t_1} \dots \lambda e^{-\lambda (t_n - t_{n-1})} e^{-\lambda (t - \lambda_n)}}{e^{-\lambda t} (\lambda t)^n / n!}$$
$$= n! / t^n$$

for $0 \le t_1 < \ldots < t_n \le t$. In particular, $(T_1, \ldots, T_n \mid X_t = n)$ is uniformly distributed on $\{ 0 \le t_1 < \ldots < t_n \le t \}$. In the specific case n = 1, the arrival is uniformly distributed: $(T_1 \mid X_t = 1) \sim \text{Unif}(0, t)$.

2 Thinning

Recall that Poisson process satisfies the following two properties:

• (Superposition) The sum $\sum_i X_i(t)$ of independent Poisson process $X_i(t) \operatorname{Poi}(\lambda_i t)$ is a Poisson process with rate being the sum of all individual rates:

$$\sum X_i(t) \sim \operatorname{Poi}\left((\sum \lambda_i)t\right)$$

• (Thinning) If each arrival of a Poisson process $X(t) \sim \text{Poi}(\lambda t)$ independently has probability p_i of being type i, the arrival processes $Y_1(t), \ldots, Y_n(t)$ of individual types are independent Poisson processes with weighted rate:

$$Y_i(t) \sim \operatorname{Poi}(p_i \lambda t)$$

Example 2. Suppose we have two independent Poisson processes X_t, Y_t with rate λ, μ respectively. What is the probability that we observe k arrivals of X_t at the nth arrival of Y_t ?

Let us consider the total process $Z_t = X_t + Y_t$, which is also a Poisson process with rate $\lambda + \mu$. Then X_t, Y_t can be considered as the thinning of Z_t , that is, each arrival of Z_t has probability $p = \frac{\lambda}{\lambda + \mu}$ of being an arrival for X_t (and $1 - p = \frac{\mu}{\lambda + \mu}$ for Y_t).

The event that we see k arrivals of X_t at the nth arrival of Y_t is the same as having k-1 successes (for being arrival of X_t) in the first k+n-1 arrivals, and the (k+n)th trial is a success (to be of X_t). So the probability is

$$\binom{k+n-1}{k-1}p^{k-1}(1-p)^n \cdot p = \binom{k+n-1}{n}\left(\frac{\mu}{\lambda+\mu}\right)^n\left(\frac{\lambda}{\lambda+\mu}\right)^k$$

In another word, the number of arrivals from Y given that we see k arrivals from X obeys negative binomial distribution $N \sim \text{NegBin}(k, \frac{\lambda}{\lambda+\mu})$.

Example 3. In a bank there are two tellers, Teller 1 and Teller 2, with exponentially distributed serving time with mean 3 and 6 minutes respectively. Suppose three people, A, B and C, enter the bank, and A, B are served by the tellers first while C waits for the first teller available.

- 1. What is the expected total amount of time for C to complete the service?
- 2. What is the expected time until the last one leave?

3. What is the probability that C is the last one to leave?

For simplicity, let us assume A is served by Teller 1, and B is served by Teller 2. It is possible to consider the service time T_A, T_B, T_C of each person individually and compute respectively

- 1. $E(\min(T_A, T_B) + T_C)$
- 2. $E(\max(T_A, T_B, \min(T_A, T_B) + T_C))$
- 3. $P(\max(T_A, T_B, \min(T_A, T_B) + T_C) = \min(T_A, T_B) + T_C)$

which you can handle by using the min property of exponential distribution, the identity $\max(x, y) = x + y - \min(x, y)$, and some algebra. (You should try to solve it this way.)

Here, we will use another approach. Instead of 3 people waiting, let us imagine there are infinitely many people waiting to be served, and A, B, C are the first three in the queue.

Let X_t, Y_t be the number of customers departed from the tellers (after being served) respectively. Then they must be Poisson processes $X_t \sim \text{Poi}(\frac{1}{3}t)$ and $Y_t \sim \text{Poi}(\frac{1}{6}t)$, so the total departure $Z_t = X_t + Y_t \sim \text{Poi}((\frac{1}{3} + \frac{1}{6})t) = \text{Poi}(\frac{1}{2}t)$ is also Poisson. This implies

1. On T_C being the service time of C, and τ_1 being the first arrival time of Z, we have

E (time C complete service) $= E (\tau_1 + T_C)$ $= E (\tau_1) + E (T_C | \text{ first departure is from teller 1)} P \text{ (first departure is from teller 1)}$ $+ E (T_C | \text{ first departure is from teller 2)} P \text{ (first departure is from teller 2)}$ $= 2 + 3 \cdot \frac{1/3}{1/2} + 6 \cdot \frac{1/6}{1/2} = 6$

2. Let τ_2 be the time of the second departure (from Z), and T be the time between the second departure and the last departure of A, B, C. Then E (time last one leaves) = $E(\tau_2 + T)$.

By the memoryless property, we may assume that the last one to leave starts being served *exactly when the* second departure happens. Easy to see that the last one must be served by a different teller from the second one, so

E (time last one leaves) $= E (\tau_2 + T)$ $= 2 \cdot 2 + E \text{ (teller 2 service time)} P \text{ (second departure is from teller 1)}$ + E (teller 1 service time) P (second departure is from teller 2) $\frac{1/2}{1/6} = \frac{1/6}{100}$

$$= 2 \cdot 2 + 6 \cdot \frac{1/3}{1/2} + 3 \cdot \frac{1/6}{1/2} = 9$$

3. C is the last one to leave iff the first two departures are from different tellers. So

P (C is last to leave) = P (first is from teller 1) P (second is from teller 2) + P (first is from teller 2) P (second is from teller 1) $= 2 \cdot \frac{1/3}{1/2} \frac{1/6}{1/2} = 4/9$