

## SOLUTIONS 7

Saliba, April 17, 2019

**Exercise 1 (Countable exponential races).** Let  $I$  be a countable space and let  $T_k, k \in I$ , be independent exponential random variables with  $T_k \sim \text{Exp}(q_k)$  with  $0 < q := \sum_{k \in I} q_k < \infty$ . Set  $T = \inf_k T_k$ . Let  $K$  be the random variable with values in  $I$  that is equal to  $k$  whenever  $T = T_k$  and  $T_j > T_k$  for  $j \neq k$ . Show that  $T$  and  $K$  are independent with  $T \sim \text{Exp}(q)$  and  $\mathbb{P}(K = k) = q_k/q$ . Deduce that  $\mathbb{P}(K = k \text{ for some } k) = 1$ .

**Solution.** We have  $K = k$  if  $T_k < T_j$  for all  $j \neq k$ . By the total probability formula, we have

$$\begin{aligned} \mathbb{P}(K = k \text{ and } T \geq t) &= \mathbb{P}(T_k \geq t \text{ and } T_j > T_k \text{ for all } j \neq k) \\ &= \int_t^\infty q_k e^{-q_k s} \mathbb{P}(T_j > s \text{ for all } j \neq k) ds \\ &= \int_t^\infty q_k e^{-q_k s} \prod_{j \neq k} e^{-q_j s} ds \\ &= \int_t^\infty q_k e^{-qs} ds = \frac{q_k}{q} e^{-qt}. \end{aligned}$$

Hence we have that  $\mathbb{P}(K = k \text{ for some } k) = 1$  and  $T$  and  $K$  have the claimed joint distribution.

**Exercise 2 (General construction of Markov processes).** Let us consider a countable state space  $E$  and an array of positive numbers  $(\lambda_{i,j})_{i,j \in E; i \neq j}$  with  $\sum_{j \in E; j \neq i} \lambda_{i,j} < \infty$  for all  $i \in E$ . We recursively define a continuous time stochastic process  $(X(t))_{t \geq 0}$  on  $E$  starting at  $i_0 \in E$  as follows:

- (i). Define  $T_0 = 0$  and set  $X(T_0) = i_0 \in E$ ;
- (ii). For  $n \in \mathbb{N}$ : suppose we know  $T_{n-1}$  and  $X(T_{n-1}) = i_{n-1}$ . Independently of the previous steps, generate independent exponential random variables  $E_1, E_2, \dots$  with  $E_j \sim \text{Exp}(\lambda_{i_{n-1},j})$ . Define  $T_n = T_{n-1} + \inf_{j \in \mathbb{N}} E_j$  and  $i_n = \text{argmin}_{j \in E} E_j$ , that is, the (random) index of the exponential variable that is the smallest. Then put

$$X(t) = \begin{cases} i_{n-1} & \text{for } t \in [T_{n-1}, T_n) \\ i_n & \text{for } t = T_n. \end{cases}$$

- a) What is the distribution of the time between the jumps of the process  $(X(t))_{t \geq 0}$ ?
- b) Let  $\hat{P}_{ij}$  be the probability

$$\hat{P}_{ij} = \mathbb{P}(X(T_n) = j \mid X(T_{n-1}) = i).$$

Find the matrix  $\hat{P} = (\hat{P}_{ij})_{i,j \in E}$ .

- c) Show that  $(X(t))_{t \geq 0}$  is a homogeneous Markov process.

**Solution.** a) We are looking for the distribution of the waiting time between two jumps, i.e. the distribution of  $S_n = T_n - T_{n-1}$ , by definition this is defined as

$$S_n = \inf_{j \in \mathbb{N}} E_j.$$

According to exercise 1, we have that  $S_n \sim \text{Exp}(\sum_{j=1}^\infty \lambda_{i_{n-1},j})$ .

- b) We know by a) that the waiting time between two jumps of the process is a n exponential random variable arising as the minimum of an countable exponential race. The first exercise gives us additionally that

$$\hat{P}_{ij} = \mathbb{P}(X(T_n) = j \mid X(T_{n-1}) = i) = \frac{\lambda_{i,j}}{\sum_{k \neq i} \lambda_{k,j}}.$$

- c) We have to show that  $(X(t))_{t \geq 0}$  is a Markov process, that is

$$\mathbb{P}(X_t = j \mid \{X_r, r \leq s, X_s = i\}) = \mathbb{P}(X_t = j \mid X_s = i).$$

Since we condition on  $\{X_r, r \leq s, X_s = i\}$ , there exists a time  $m$  (depending on  $\omega$ ) such that  $\{X_r, r \leq s, X_s = i\} = \{T_{m-1} < s < T_m \text{ and } X_s = i\}$ . First, note that by construction the process before time  $T_{m-1}$  is irrelevant for determining this probability:

$$\begin{aligned} & \mathbb{P}(X_t = j \mid \{X_r, r \leq s \text{ and } T_{m-1} < s < T_m \text{ and } X_s = i\}) \\ &= \mathbb{P}(X_t = j \mid \{X_r, T_{m-1} \leq r \leq s \text{ and } T_m > s \text{ and } X_s = i\}). \end{aligned}$$

Then memorylessness property of the exponential random variables implies that for  $S_m = T_m - T_{m-1}$

$$S_m \sim S_m - (s - T_{m-1}) \sim \text{Exp}\left(\sum_{j=1}^{\infty} \lambda_{i,j}\right),$$

i.e. knowing that the exponential rate exceeds  $s - T_{m-1}$  is irrelevant for determining the current transitions probabilities. Moreover, since  $X_{T_{m-1}} = X_s$  by definition of  $T_{m-1}$  and  $T_m$ ,  $\{X_s = i\}$  is the only relevant information for the next evolution of the process based on information contained in  $\{X_r, T_{m-1} \leq r \leq s \text{ and } T_m > s \text{ and } X_s = i\}$ . Thus

$$\mathbb{P}(X_t = j \mid \{X_r, T_{m-1} \leq r \leq s \text{ and } T_m > s \text{ and } X_s = i\}) = \mathbb{P}(X_t = j \mid X_s = i)$$

this finishes the proof.

**Definition (The  $Q$ -matrix).** One way of thinking about the evolution of the Markov process  $(X(t))_{t \geq 0}$  is in terms of its  $Q$ -matrix, which is known as the generator of the process. A matrix  $Q = (q_{ij})_{i,j \in E}$  is a  $Q$ -matrix if it satisfies

- (i).  $-\infty < q_{ii} \leq 0$  for all  $i \in E$ ;
- (ii).  $0 \leq q_{ij} < \infty$  for all  $i \neq j$ ;
- (iii).  $\sum_{j \in E} q_{ij} = 0$  for all  $i \in E$ .

The  $Q$ -matrix of the Markov process  $(X(t))_{t \geq 0}$  as constructed above is given by  $q_{ii} = -\sum_{j \neq i} \lambda_{i,j}$  for  $i \in E$ , and  $q_{ij} = \lambda_{ij}$  for  $j \neq i$ .

**Exercise 3.** In a population of size  $N$ , a rumor is begun by a single individual who tells it to everyone he meets; they in turn pass the rumor to everyone they meet, once a person has passed the rumor to somebody he exits the system. Assume that each individual meets another randomly with exponential rate  $1/N$ . Let  $X(t)$ ,  $t \geq 0$  be the number in  $E = \{1, \dots, N\}$  of people who know the rumor at time  $t$ .

- a) Draw a graph to visualize the chain. Write down the  $Q$ -matrix of the chain.
- b) How long does it take in average until everyone knows the rumor if  $X(0) = 1$ ?

**Solution.** a) The  $Q$ -matrix has the form

$$\begin{pmatrix} -\frac{N-1}{N} & \frac{N-1}{N} & 0 & \cdots & 0 \\ 0 & -\frac{N-2}{N} & \frac{N-2}{N} & \cdots & 0 \\ \vdots & \vdots & \vdots & \ddots & 0 \\ 0 & 0 & 0 & -\frac{1}{N} & \frac{1}{N} \\ 0 & 0 & 0 & 0 & 0 \end{pmatrix}.$$

- b) We need to compute  $\mathbb{E}_1(T_N)$  where  $T_N = \inf\{t : X(t) = N\}$ . Remark that  $T_N$  is just a sum of exponential random variables

$$T_N = \sum_{i=1}^{N-1} E_i,$$

where  $E_i \sim \text{Exp}(\frac{N-i}{N})$ , So that

$$\mathbb{E}_1(T_N) = \sum_{i=1}^{N-1} \frac{N}{N-i} \approx N \log N.$$

You could notice that this is exactly the continuous time version of the coupon's collector model.

**Exercise 4.** For  $i \in \mathbb{N}$ , let  $E_i$  be independent copies of an exponential random variable of parameter  $\lambda$ . We let  $T_n := E_1 + \cdots + E_n$  and

$$N(t) := \sum_{n=1}^{\infty} \mathbb{1}_{\{T_n \leq t\}}, \quad t \geq 0.$$

The process  $(N(t))_{t \geq 0}$  is called a homogeneous Poisson process with intensity  $\lambda$ . Let  $T_0 = 0$  and we say that  $T_1, T_2, T_3, \dots$  are the successive arrival times of the Poisson process, and  $E_n$  the intervals  $T_n - T_{n-1}$ .

- (i). Show that  $T_n$  follows an Erlang law with parameters  $n$  and  $\lambda$  having density:

$$f_{T_n}(t) = \frac{\lambda^n}{(n-1)!} t^{n-1} e^{-\lambda t} \mathbb{1}_{\{t > 0\}}.$$

- (ii). Show that,  $\forall t > 0$ ,  $N(t)$  follows a Poisson law with parameter  $\lambda t$ , i.e.

$$\mathbb{P}(N(t) = k) = e^{-\lambda t} \frac{(\lambda t)^k}{k!}, \quad k = 0, 1, 2, \dots$$

**Solution.** (i). We proceed by induction on  $n$ . For  $n = 1$ ,  $T_1$  follows an exponential law with parameter  $\lambda$ , which is equivalent to an Erlang law of parameter 1 and  $\lambda$ . Suppose that  $T_n \sim \text{Erlang}(n, \lambda)$ . Remark that  $E_{n+1} \sim \text{Exp}(\lambda)$  is independent of  $T_n$ . For  $t > 0$ , we have

$$\begin{aligned} F_{T_{n+1}}(t) &= \mathbb{P}(T_{n+1} \leq t) = \mathbb{P}(T_n + E_{n+1} \leq t) = \int_0^\infty \mathbb{P}(T_n + E_{n+1} \leq t \mid T_n = u) f_{T_n}(u) du \\ &= \int_0^t \mathbb{P}(E_{n+1} \leq t - u) f_{T_n}(u) du = \int_0^t F_{E_{n+1}}(t - u) f_{T_n}(u) du. \end{aligned}$$

Implying that

$$\begin{aligned} f_{T_{n+1}}(t) &= \frac{d}{dt} \int_0^t (1 - e^{-\lambda(t-u)}) f_{T_n}(u) du \\ &= \int_0^t \left( (1 - e^{-\lambda(t-u)}) f_{T_n}(u) \right)' du + (1 - e^{-\lambda(t-t)}) f_{T_n}(t) \\ &= \int_0^t \lambda e^{-\lambda(t-u)} \frac{\lambda^n}{(n-1)!} u^{n-1} e^{-\lambda u} du \\ &= \frac{\lambda^{n+1}}{(n-1)!} e^{-\lambda t} \int_0^t u^{n-1} du \\ &= \frac{\lambda^{n+1}}{n!} t^n e^{-\lambda t}. \end{aligned}$$

(ii). By definition of  $(N(t))_{t \geq 0}$  and of the arrival times  $T_i$ , we know that

$$\mathbb{P}(N(t) \geq n) = \mathbb{P}(T_n \leq t) = F_{T_n}(t).$$

By (i), we have

$$\begin{aligned} F_{T_{n+1}}(t) &= \int_0^t (1 - e^{-\lambda(t-u)}) f_{T_n}(u) du = F_{T_n}(t) - \int_0^t e^{-\lambda(t-u)} \cdot \frac{\lambda^n}{(n-1)!} u^{n-1} e^{-\lambda u} du \\ &= F_{T_n}(t) - \frac{\lambda^n}{n!} t^n e^{-\lambda t} \end{aligned}$$

It is a recursive relation between  $F_{T_{n+1}}(t)$  and  $F_{T_n}(t)$ . As  $F_{T_1}(t) = 1 - e^{-\lambda t}$ , we get

$$F_{T_{n+1}}(t) = 1 - \sum_{k=0}^n e^{-\lambda t} \frac{(\lambda t)^k}{k!}, \quad n \in \mathbb{N}.$$

Using this result, we obtain the distribution of  $N(t)$  for a fixed  $t$

$$\begin{aligned} \mathbb{P}(N(t) = n) &= \mathbb{P}(N(t) \geq n) - \mathbb{P}(N(t) \geq n+1) \\ &= \sum_{k=0}^n e^{-\lambda t} \frac{(\lambda t)^k}{k!} - \sum_{k=0}^{n-1} e^{-\lambda t} \frac{(\lambda t)^k}{k!} = e^{-\lambda t} \frac{(\lambda t)^n}{n!}. \end{aligned}$$

that is  $N(t) \sim \text{Poi}(\lambda t)$ .