Exercise 1. Let $\{X(t) \mid t \in \mathbb{R}^+\}$ be a Markov process with n states $\{1, 2, \dots, n\}$ and generator:

$$Q = \begin{pmatrix} -\lambda_1 & \lambda_1 & 0 & 0 & \cdots & 0 & 0 \\ 0 & -\lambda_2 & \lambda_2 & 0 & \cdots & 0 & 0 \\ 0 & 0 & -\lambda_3 & \lambda_3 & \cdots & 0 & 0 \\ 0 & 0 & 0 & -\lambda_4 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & 0 & \cdots & -\lambda_{n-1} & \lambda_{n-1} \\ \lambda_n & 0 & 0 & 0 & \cdots & 0 & -\lambda_n \end{pmatrix}.$$

- (i). Find the stationary distribution of the process
- (ii). Give an intuitive explanation of the result.

Solution. Starting in a given state i, we see that we have a cyclic path, we go deterministically to i+1 if $i \neq n$ and in 1 if i=n, spending a exponentially distributed time with parameter λ_i in each position.

Intuitively, the stationary distribution in i is the mean proportion of time spent in i during a full cycle, that is

$$\pi_i = \frac{\frac{1}{\lambda_i}}{\sum_{j=1}^n \frac{1}{\lambda_k}}.$$

We check now that it is really a stationary distribution. Clearly,

$$\sum_{j=1}^{n} \pi_j = 1.$$

Note that $\pi Q = 0$, since

$$\begin{cases} \lambda_i \pi_i = \lambda_{i+1} \pi_{i+1}, si \ i \neq n \\ \lambda_n \pi_n = \lambda_1 \pi_1 \end{cases}$$

Note that unicity is obtained by a direct resolution of the system.

Exercise 2. Let $\{X(t) \mid t \in \mathbb{R}^+\}$ be a Markov process with n states $\{1, 2, \dots, n\}$ and generator:

$$Q = \begin{pmatrix} -\lambda_1 & \lambda_1 & 0 & 0 & \cdots & 0 & 0 \\ \lambda_2\mu_2 & -\lambda_2 & \lambda_2(1-\mu_2) & 0 & \cdots & 0 & 0 \\ \lambda_3\mu_3 & 0 & -\lambda_3 & \lambda_3(1-\mu_3) & \cdots & 0 & 0 \\ \lambda_4\mu_4 & 0 & 0 & -\lambda_4 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \lambda_{n-1}\mu_{n-1} & 0 & 0 & 0 & \cdots & -\lambda_{n-1} & \lambda_{n-1}(1-\mu_{n-1}) \\ \lambda_n & 0 & 0 & \cdots & 0 & -\lambda_n \end{pmatrix}.$$

- (i). Guess the expression $\pi_1 = \lim_{t \to \infty} P[X(t) = 1]$.
- (ii). Find the stationary distribution of the process.

Solution. At every state i except 1 and n, we jump to the next state with probability $1 - \mu_i$, or we return to 1 with probability μ_i , we deterministically return to 1 in position n and we jump in 2 when we are in 1. As before, we infer that the stationary distribution in 1 is equal to the mean time spent in 1, divided by the duration of a return in 1—duration of a cycle. Let us denote C the duration of a cycle, then

$$E[C] = \sum_{k=2}^{n} E[C|\text{the last state visited before 1 is } k] \mathbb{P}\{\text{the last state visited before 1 is } k\}$$

$$= \sum_{k=2}^{n} \left(\frac{1}{\lambda_1} + \dots + \frac{1}{\lambda_k}\right) \times (1 - \mu_2) \cdots (1 - \mu_{k-1}) \mu_k,$$

and so

$$\pi_1 = \frac{\frac{1}{\lambda_1}}{\mathrm{E}[C]}.$$

With the same methodology, we find π_i , $2 \le i \le n$. Let τ_i be the time spent in i during a cycle, then

$$E[\tau_i] = E[\tau_i|i \text{ hit during the cycle}] \mathbb{P}\{i \text{ hit during the cycle}\}$$
$$= \frac{1}{\lambda_i}(1-\mu_2)\cdots(1-\mu_{i-1}),$$

we used that

 $E [\tau_i | i \text{ is not hit during the cycle}] = 0.$

Thus,

$$\pi_i = \frac{\mathrm{E}[\tau_i]}{\mathrm{E}[C]}.$$

Exercise 3. Let $\{X(t) \mid t \in \mathbb{R}^+\}$ be an irreducible Markov process on a finite space of n states, with generator Q. Let us take λ so that $\lambda > \max_i \{-Q_{ii}\}$ and define the matrix:

$$P = I + \frac{1}{\lambda}Q$$
 (where I is the identity matrix).

- (i). Show that P is a transition matrix, and that its stationary distribution is identical to the one of Q.
- (ii). Let $\{N(t) \mid t \in \mathbb{R}^+\}$ be a Poisson process with parameter λ and let $\{Y_k \mid k \in \mathbb{N}\}$ be a Markov chain with matrix P, independent from $\{N(t)\}$. Let $T_0 = 0$, and T_1, T_2, \cdots denote the arrival times in the Poisson process.

We define the process $\{Z(t) \mid t \in \mathbb{R}^+\}$ as follows:

$$Z(t) = Y_n \qquad \forall t \in [T_n, T_{n+1}].$$

Show that $\{Z(t)\}$ is a Markov process with generator Q.

Solution. (i). Let π be the stationary distribution of Q, then

$$\pi P = \pi + \underbrace{\frac{1}{\lambda}\pi Q}_{=0} = \pi.$$

As there exists a unique solution (to a multiplicative constant) satisfying xP = x, π is the stationary distribution.

(ii). $\{Z(t)\}$ is a Markov process, since,

- $\mathbb{P}\{Z(t+s)=j|Z(u), 0 \leq u \leq s\} = \mathbb{P}\{Z(t+s)=j|Z(s)\}$, since the Poisson process N is Markovian, its arrivals between s and s+t do not depend on the history of the process before s. Moreover, $\{Y_n\}$ satisfy the Markov property and thus the future transitions depend only the the current one s.
- $$\begin{split} \bullet \ \mathbb{P}\{Z(t+s) = j | Z(s) = i\} &= \mathbb{P}\{Z(t) = j | Z(0) = i\} \colon \\ \mathbb{P}\{Z(t+s) = j | Z(s) = i\} &= \sum_{k} \mathbb{P}\{Z(t+s) = j | Z(s) = i, \ k \ arrivals \ from \ N \ on \ [s,s+t]\} \\ &\times \underbrace{\mathbb{P}\{N(t+s) N(s) = k | Z(s) = i\}}_{=\mathbb{P}\{N(t) = k | Z(0) = i\}} \\ &= \sum_{k} (P^k)_{ij} \mathbb{P}\{N(t) = k | Z(0) = i\} \\ &= \mathbb{P}\{Z(t) = j | Z(0) = i\}. \end{split}$$

Let us show now that it is a Markov process with generator Q. We show that the transition matrix of $\{Z(t)\}$ can be expressed as the exponential of Q,

$$L_{ij}(t) := \mathbb{P}\{Z(t) = j | Z(0) = i\}$$

$$= \sum_{k=0}^{\infty} \mathbb{P}\{Z(t) = j | Z(0) = i, k \text{ arrivals of } N \text{ on } [0, t]\}$$

$$\times \mathbb{P}\{N(t) = k | Z(0) = i\}$$

$$= \sum_{k=0}^{\infty} (P^k)_{ij} \exp\{-\lambda t\} \frac{(\lambda t)^k}{k!}.$$

So that,

$$L(t) = \exp\{-\lambda It\} \sum_{k=0}^{\infty} \frac{(\lambda t P)^k}{k!} = \exp\{(\lambda P - \lambda I)t\},\,$$

and $\lambda P - \lambda I = \lambda (I + \frac{1}{\lambda}Q) - \lambda I = Q$, thus

$$L(t) = \exp\{Qt\}.$$

Exercise 4. Show that the renewal function R(t) satisfy a renewal equation, and specify the corresponding function g(t).

Solution. We could get a renewal equation for R(t) by conditioning on the first time S_1 after S_0 . However, here, since $R(t) = \sum_{n \geq 0} F^{(n)}(t)$, we directly get that

$$(R * F)(t) = \sum_{n \ge 0} (F^{(n)} * F)(t) = \sum_{n \ge 0} F^{(n+1)}(t),$$

$$= \sum_{n \ge 1} F^{(n)}(t) = \sum_{n \ge 0} F^{(n)}(t) - F^{(0)}(t),$$

and so

$$R(t) = F^{(0)}(t) + (R * F)(t).$$

with, $g(t) = F^{(0)}(t) = 1_{\{t \ge 0\}}$.

This also shows that $t \ge 0$, (R * F)(t) = R(t) - 1, which was used during the lecture in the proof of the theorem giving the distribution of the duration of life L of a transitive renewal process.

Exercise 5. Let S_1, S_2, \ldots be the successive times at which cars cross a certain fixed position on the highway. We assume that the intervals of time W_1, W_2, \ldots between each renewal are i.i.d. with cumulative distribution $F(\cdot)$. Suppose that at time t = 0, a pedestrian arrives at this fixed position, and wants to cross the road. Assume that he needs τ units of time to cross it. Let L be the time that the pedestrian has to wait before starting to cross the road.

- (a) Find the distribution of L and its expectation.
- (b) Same questions if we assume that the arrivals of cars follow a Poisson process with parameter λ .

Solution. (a). The pedestrian starts to cross the road at $L = S_n$ if and only if $W_1 \le \tau, \ldots, W_n \le \tau$ and $W_{n+1} > \tau$. So that, L is the duration of life of a renewal $\{\tilde{S}_n\}$ having its n^{th} interval of time given by

$$\tilde{W}_n = \begin{cases} W_n, & si \ W_n \le \tau, \\ +\infty, & otherwise. \end{cases}$$

Thus, the distribution of \tilde{W}_n is given by

$$\tilde{F}(t) = \begin{cases} F(t), & \text{if } t \leq \tau, \\ F(\tau), & \text{if } t > \tau. \end{cases}$$

We deduce that

$$\mathbb{P}\{L \le t\} = \{1 - \tilde{F}(\infty)\}\tilde{R}(t) = \{1 - F(\tau)\}\tilde{R}(t),$$

where $\tilde{R}(t) = \sum_{n\geq 0} \tilde{F}^{(n)}(t)$.

(b). The mean waiting time of the pedestrian is

$$E[L] = \frac{1}{1 - F(\tau)} \int_0^{\tau} \{F(\tau) - F(t)\} dt.$$

More precisely, if the arrivals of the cars follow a Poisson process with parameter λ , then

$$E[L] = \frac{1}{\lambda} (\exp{\{\lambda \tau\}} - 1) - \tau.$$