## Solutions to tutorial exercises for stochastic processes

T1. If  $\beta + \delta = 0$  then by the Kolmogorov backward-equations:

$$\frac{\mathrm{d}}{\mathrm{d}t}p_t(x,y) = 0,$$

so that  $p_t(x,y)$  is constant. Since  $p_0(x,x) = 1$ ,  $p_t(x,y) = \mathbb{1}_{\{x=y\}}$ . Now suppose that  $\beta + \delta > 0$ . The Kolmogorov backward-equations are solved uniquely by  $P_t = \exp(tQ)$ . The matrix Q has eigenvalues 0 and  $-\beta - \delta$ . We can find a matrix U such that

$$Q = U \begin{pmatrix} 0 & 0 \\ 0 & -\beta - \delta \end{pmatrix} U^{-1}.$$

This is solved by

$$U = \begin{pmatrix} 1 & -\frac{\beta}{\beta + \delta} \\ 1 & \frac{\delta}{\beta + \delta} \end{pmatrix}.$$

We conclude

$$P_{t} = U \begin{pmatrix} 1 & 0 \\ 0 & \exp(-(\beta + \delta)t) \end{pmatrix} U^{-1}$$
$$= \begin{pmatrix} \frac{\delta}{\beta + \delta} + \frac{\beta}{\beta + \delta} e^{-t(\beta + \delta)} & \frac{\beta}{\beta + \delta} \left(1 - e^{-t(\beta + \delta)}\right) \\ \frac{\delta}{\beta + \delta} \left(1 - e^{-t(\beta + \delta)}\right) & \frac{\beta}{\beta + \delta} + \frac{\delta}{\beta + \delta} e^{-t(\beta + \delta)} \end{pmatrix}.$$

T2. (a) Let  $t^*$  be the time of the first jump. If  $X_0 = 0$  then this  $t^*$  is exponentially distributed with parameter  $\beta$ . We now have

$$p_{t}(0,1) = \mathbb{P}^{0}(X_{t} = 1)$$

$$= \sum_{k=0}^{n-1} \mathbb{P}^{0} \left( X_{t} = 1, t^{*} \in \left[ \frac{k}{n} t, \frac{k+1}{n} t \right] \right)$$

$$= \sum_{k=0}^{n-1} \mathbb{P}^{0} \left( X_{t} = 1 \mid t^{*} \in \left[ \frac{k}{n} t, \frac{k+1}{n} t \right] \right) \mathbb{P}^{0} \left( t^{*} \in \left[ \frac{k}{n} t, \frac{k+1}{n} t \right] \right). \tag{1}$$

Since

$$\sigma\left(t^* \in \left\lceil \frac{k}{n}t, \frac{k+1}{n}t \right\rceil\right) \subseteq \mathfrak{F}_{t^* \wedge \frac{k+1}{n}t},$$

we have by the tower property

$$\mathbb{P}^{0}\left(X_{t}=1 \mid t^{*} \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]\right) = \mathbb{E}^{0}\left[\mathbb{1}_{\left\{X_{t}=1\right\}} \mid t^{*} \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]\right]$$

$$= \mathbb{E}^{0}\left[\mathbb{E}^{0}\left[\mathbb{1}_{\left\{X_{t}=1\right\}} \mid \mathfrak{F}_{t^{*} \wedge \frac{k+1}{n}t}\right] \mid t^{*} \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]\right].$$

We now use the strong Markov property to get

$$\mathbb{P}^{0}\left(X_{t}=1 \mid t^{*} \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]\right) = \mathbb{E}^{0}\left[\mathbb{P}^{X_{t^{*} \wedge \frac{k+1}{n}t}}\left(X_{t-t^{*} \wedge \frac{k+1}{n}t}=1\right) \mid t^{*} \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]\right]$$

$$= \mathbb{E}^{0}\left[\mathbb{P}^{1}\left(X_{t-t^{*}}=1\right) \mid t^{*} \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]\right]$$

$$\leq \sup_{s \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]}\mathbb{P}^{1}\left(X_{t-s}=1\right),$$

and similarly

$$\mathbb{P}^{0}\left(X_{t}=1 \mid t^{*} \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]\right) \geq \inf_{s \in \left[\frac{k}{n}t, \frac{k+1}{n}t\right]} \mathbb{P}^{1}\left(X_{t-s}=1\right).$$

So the sum in (1) converges to the Riemann-Stieltjes integral as  $n \to \infty$ :

$$p_t(0,1) = \int_0^t \mathbb{P}^1(X_{t-s} = 1) dF(s)$$
$$= \int_0^t \beta e^{-\beta s} p_{t-s}(1,1) ds,$$

where  $F(s) = 1 - \exp(-\beta s)$ . By using the same argumentation we find

$$p_t(1,0) = \int_0^t \delta e^{-\delta s} p_{t-s}(0,0) ds$$

(b) By definition of the Q-matrix we have

$$q(0,1) = \frac{\mathrm{d}}{\mathrm{d}t} p_t(0,1) \Big|_{t=0}$$

$$= \frac{\mathrm{d}}{\mathrm{d}t} \int_0^t \beta e^{-\beta s} p_{t-s}(1,1) \mathrm{d}s \Big|_{t=0}$$

$$= \beta e^{-\beta t} p_0(1,1) \Big|_{t=0}$$

$$= \beta.$$

Similarly we can find  $q(1,0) = \delta$ . Since the row sums of the Q matrix are zero we conclude  $q(0,0) = -\beta$  and  $q(1,1) = -\delta$ .

T3. Items (a) and (b) are equivalent since

$$\{N(t) < \infty\} = \left\{\sum_{n=0}^{\infty} \tau_n > t\right\}.$$

We will now show the equivalence of (b) and (c). Let  $\lambda > 0$ . Conditioned on  $Z_0, Z_1, \ldots$ , the random variables  $\tau_0, \ldots, \tau_n$  are independent exponential random variables. Therefore, using the moment generating function of the exponential distribution:

$$\mathbb{E}\left[\exp\left(-\lambda\sum_{k=0}^{n}\tau_{k}\right) \mid Z_{0}, Z_{1}, \dots\right] = \prod_{k=0}^{n} \frac{c(Z_{k})}{c(Z_{k}) + \lambda}$$

We now take expectations on both sides and let  $n \to \infty$  to find

$$\mathbb{E}\left[\exp\left(-\lambda\sum_{k=0}^{\infty}\tau_{k}\right)\right] = \mathbb{E}\left[\prod_{k=0}^{\infty}\frac{c(Z_{k})}{c(Z_{k})+\lambda}\right],$$

where we use dominated convergence to switch limit and expectation. The product  $\prod_{k=0}^{\infty} \frac{c(Z_k)}{c(Z_k) + \lambda} > 0$  if and only if  $\prod_{k=0}^{\infty} \left(1 + \frac{\lambda}{c(Z_k)}\right) < \infty$  and this is equivalent to  $\sum_{k=0}^{\infty} \frac{\lambda}{c(Z_k)} < \infty$ . We conclude by taking  $\lambda \downarrow 0$ :

$$\mathbb{P}\left(\sum_{k=0}^{\infty} \tau_k < \infty\right) = \mathbb{P}\left(\sum_{k=0}^{\infty} \frac{1}{c(Z_k)} < \infty\right).$$