

Solutions to tutorial exercises for stochastic processes

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

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

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

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

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

$$\begin{aligned} p_t(0, 1) &= \mathbb{P}^0(X_t = 1) \\ &= \sum_{k=0}^{n-1} \mathbb{P}^0 \left(X_t = 1, t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right) \\ &= \sum_{k=0}^{n-1} \mathbb{P}^0 \left(X_t = 1 \mid t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right) \mathbb{P}^0 \left(t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right). \quad (1) \end{aligned}$$

Since

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

we have by the tower property

$$\begin{aligned}\mathbb{P}^0 \left(X_t = 1 \mid t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right) &= \mathbb{E}^0 \left[\mathbb{1}_{\{X_t=1\}} \mid t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right] \\ &= \mathbb{E}^0 \left[\mathbb{E}^0 \left[\mathbb{1}_{\{X_t=1\}} \mid \mathfrak{F}_{t^* \wedge \frac{k+1}{n}t} \right] \mid t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right].\end{aligned}$$

We now use the strong Markov property to get

$$\begin{aligned}\mathbb{P}^0 \left(X_t = 1 \mid t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right) &= \mathbb{E}^0 \left[\mathbb{P}^{X_{t^* \wedge \frac{k+1}{n}t}} \left(X_{t-t^* \wedge \frac{k+1}{n}t} = 1 \right) \mid t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right] \\ &= \mathbb{E}^0 \left[\mathbb{P}^1 (X_{t-t^*} = 1) \mid t^* \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right] \right] \\ &\leq \sup_{s \in \left[\frac{k}{n}t, \frac{k+1}{n}t \right]} \mathbb{P}^1 (X_{t-s} = 1),\end{aligned}$$

and similarly

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

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

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

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

$$\begin{aligned}q(0, 1) &= \left. \frac{d}{dt} p_t(0, 1) \right|_{t=0} \\ &= \left. \frac{d}{dt} \int_0^t \beta e^{-\beta s} p_{t-s}(1, 1) ds \right|_{t=0} \\ &= \left. \beta e^{-\beta t} p_0(1, 1) \right|_{t=0} \\ &= \beta.\end{aligned}$$

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, \dots, τ_n are independent exponential random variables. Therefore, using the moment generating function of the exponential distribution:

$$\mathbb{E} \left[\exp \left(-\lambda \sum_{k=0}^n \tau_k \right) \middle| Z_0, Z_1, \dots \right] = \prod_{k=0}^n \frac{c(Z_k)}{c(Z_k) + \lambda}$$

We now take expectations on both sides and let $n \rightarrow \infty$ 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 $\lambda \downarrow 0$:

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