Applied Stochastic Processes

Solution sheet 10

Solution 10.1

- (a) 1. True, by Theorem 6.6 of the lecture notes.
 - 2. False. Theorem 6.6 guarantees uniqueness of solutions to the renewal equation, so g + 10 cannot also be a solution.
 - 3. False. Again, by Theorem 6.6, the solution is of the form

 $h + 10 + (h + 10) \star m = g + 10 + 10 \star m = g + 10 + 10F.$

Thus, g + 10 is not a solution.

- 4. True, as shown in the previous question.
- 5. False. If it were true, we would have

$$g + F = h + F + g \star F = g + F,$$

implying m(t) = F(t) for all t, which is never the case.

- (b) 1. False. For instance, take $F(t) = \mathbb{1}_{\{t \ge 2\}}$ and $h(t) = \mathbb{1}_{\{t \in [0,1]\}}$. Then it is easy to check that the solution g to the renewal equation oscillates between 0 and 1.
 - 2. False. This would be the conclusion of the Smith key renewal theorem if F were nonlattice. However, for lattice distributions, the conclusion fails — see the counterexample in (1).
 - 3. True, by the uniqueness of solutions to the renewal equation and the Smith key renewal theorem.
 - 4. False. See the counterexample from (1).
 - 5. True. This follows directly from applying the renewal equation at t + s.

Solution 10.2

We have:

$$m(s+t) = \mathbb{E}[N_{s+t}] = \mathbb{E}\left[\sum_{k\geq 1} \mathbb{1}_{\{S_k\in[0,s+t]\}}\right] = \mathbb{E}\left[\sum_{k\geq 1} \left(\mathbb{1}_{\{S_k\in[0,t]\}} + \mathbb{1}_{\{S_k\in(t,s+t]\}}\right)\right]$$
$$= \mathbb{E}\left[\sum_{k\geq 1} \mathbb{1}_{\{S_k\in[0,t]\}}\right] + \mathbb{E}\left[\sum_{k\geq 1} \mathbb{1}_{\{S_k\in(t,s+t]\}}\right]$$
$$= m(t) + \mathbb{E}\left[\sum_{k\geq 1} \mathbb{1}_{\{S_k\in(t,s+t]\}}\right] \le m(t) + \mathbb{E}\left[\sum_{k\geq 1} \mathbb{1}_{\{S_k\in[t,s+t]\}}\right]$$

We now aim to show that the second term is bounded above by 1 + m(s). The reason it is not exactly equal to m(s) is that we lack control over where the interval [t, s + t] starts relative to the last arrival time. We write:

$$\mathbb{E}\left[\sum_{k\geq 1}\mathbbm{1}_{\{S_k\in[t,s+t]\}}\right] = \mathbb{E}\left[\mathbbm{1}_{\{S_{N_t+1}\in[t,s+t]\}} + \sum_{k\geq N_t+2}\mathbbm{1}_{\{S_k\in[S_{N_t+1},s+t]\}}\right].$$

Here we have separated the first arrival after time t from the remaining arrivals that occur in the interval [t, t + s]. We can now bound the first term by 1, and shift the second term to start from zero:

$$\mathbb{E}\left[\mathbbm{1}_{\{S_{N_{t}+1}\in[t,s+t]\}} + \sum_{k\geq N_{t}+2} \mathbbm{1}_{\{S_{k}\in[S_{N_{t}+1},s+t]\}}\right] \leq 1 + \mathbb{E}\left[\sum_{k\geq 1} \mathbbm{1}_{\{S_{k}\in[0,s+t-S_{N_{t}+1}]\}}\right]$$
$$\leq 1 + \mathbb{E}\left[\sum_{k\geq 1} \mathbbm{1}_{\{S_{k}\in[0,s]\}}\right]$$
$$= 1 + m(s),$$

where the last inequality follows from the fact that $s + t - S_{N_t+1} \leq s$ almost surely.

Solution 10.3

(a) Let us define $h(t) = \mathbb{1}_{\{t \le x\}}(1 - F(t))$ for $t \ge 0$. Note that $h \ge 0$, it is measurable, continuous a.e. and bounded by 1. Also it vanishes outside the compact interval [0, x]. This implies that h is directly Riemann integrable. Since F is non-lattice, by Smith's key renewal theorem it follows that

$$\lim_{t \to \infty} a_x(t) = \frac{1}{\mathbb{E}[T_1]} \int_0^\infty h(t) dt = \frac{1}{\mu} \int_0^x (1 - F(t)) dt =: G(x).$$

(b) To see that G is a distribution function, we note that

$$\lim_{x \to \infty} G(x) = \frac{1}{\mu} \int_0^\infty \mathbb{P}[T_1 > t] dt = \frac{\mathbb{E}[T_1]}{\mu} = 1.$$

This means that A_t converges in distribution to a random variable with distribution G.

Solution 10.4

(a) Note that $h \ge 0$ and it is a non increasing function. Also

$$\int_0^\infty h(t)dt = \int_0^\infty \mathbf{P}[U_1 > t]dt = \mathbf{E}[U_1] < \infty,$$

which means that h is directly Riemann integrable. Since F is non-lattice and g is solution of the equation g = h + g * F, we know by Smith's key renewal theorem that

$$\lim_{t \to \infty} g(t) = \frac{1}{\mathbf{E}[T_1]} \mathbf{E}[U_1] = \frac{\mathbf{E}[U_1]}{\mathbf{E}[U_1] + \mathbf{E}[V_1]}$$

Solution 10.5 Ear fixed k > 0

For fixed
$$\kappa \geq 0$$
,

$$\mathbf{P}[X_n = k] = \frac{n!}{k!(n-k)!} p_n^k (1-p_n)^{n-k}$$
(1)

$$=\underbrace{\frac{n\cdot(n-1)\cdot\ldots\cdot(n-k+1)}{n\cdot n\cdot\ldots\cdot n}}_{\stackrel{n\to\infty}{\underbrace{}} \to 1}\cdot\frac{1}{k!}\cdot\underbrace{(p_n\cdot n)^k}_{\stackrel{n\to\infty}{\underbrace{}} \lambda^k}\left(1-\frac{p_n\cdot n}{n}\right)^{n-k},\tag{2}$$

2/3

and since $\frac{p_n \cdot n}{n} \cdot (n-k) \to \lambda$, one has $(1 - \frac{p_n \cdot n}{n})^{n-k} \to e^{-\lambda}$ as $n \to \infty$. Hence,

$$\mathbf{P}[X_n = k] \longrightarrow \frac{e^{-\lambda}}{k!} \lambda^k = \mathbf{P}[X = k], \tag{3}$$

and for $y \in \mathbb{R}$,

$$F_{X_n}(y) = \mathbf{P}[X_n \le y] = \sum_{k \le y} \mathbf{P}[X_n = k] \xrightarrow{n \to \infty} \sum_{k \le y} \mathbf{P}[X = k] = \mathbf{P}[X \le y] = F_X(y).$$
(4)