

## Applied Stochastic Models (SS 08)

### Problem Set 4

#### Problem 1

Find the mean and variance of the Weibull distribution with hazard rate (or failure rate)

$$r(t) = \frac{f(t)}{1 - F(t)} = \lambda\alpha(\lambda t)^{\alpha-1}, \lambda > 0, \alpha > 0, t \geq 0.$$

**Solution:** The hazard rate function  $\lambda(t) = \frac{f(t)}{1-F(t)}$ ,  $t > 0$ , of some random lifetime  $T$  uniquely determines the distribution of  $T$ , since necessarily

$$F^T(t) = 1 - e^{-\int_0^t \lambda(u) du}, \quad f^T(t) = \lambda(t) e^{-\int_0^t \lambda(u) du}, \quad t > 0.$$

A random life time  $T$  with the above failure rate

$$r(t) = \lambda\alpha(\lambda t)^{\alpha-1}, \lambda > 0, \alpha > 0, t \geq 0,$$

thus has the density

$$f^T(t) = r(t) e^{-\int_0^t r(u) du} = \lambda\alpha(\lambda t)^{\alpha-1} e^{-(\lambda t)^\alpha}, \lambda > 0, \alpha > 0, t \geq 0.$$

A direct calculation yields

$$\begin{aligned} \mathbb{E}T &= \int_0^\infty t \lambda\alpha(\lambda t)^{\alpha-1} e^{-(\lambda t)^\alpha} dt \\ &= \alpha \int_0^\infty (\lambda t)^\alpha e^{-(\lambda t)^\alpha} dt. \end{aligned}$$

Substituting  $u := (\lambda t)^\alpha$  yields

$$\mathbb{E}T = \frac{1}{\lambda} \int_0^\infty u^{\frac{1}{\alpha}} e^{-u} du = \frac{1}{\lambda} \Gamma\left(\frac{1}{\alpha} + 1\right).$$

A similar calculation yields for the second moment

$$\mathbb{E}T^2 = \int_0^\infty t^2 \lambda\alpha(\lambda t)^{\alpha-1} e^{-(\lambda t)^\alpha} dt,$$

where the same substitution as above yields

$$\mathbb{E}T^2 = \frac{1}{\lambda^2} \int_0^\infty u^{\frac{2}{\alpha}} e^{-u} du = \frac{1}{\lambda^2} \Gamma\left(\frac{2}{\alpha} + 1\right).$$

This means that the variance of  $T$  is given by

$$\text{Var}(T) = \mathbb{E}T^2 - (\mathbb{E}T)^2 = \frac{1}{\lambda^2}(\Gamma(\frac{2}{\alpha} + 1) - \Gamma(\frac{1}{\alpha} + 1)^2).$$

**Problem 2**

Let  $N(t)$  be a Poisson process, and let  $Y_1, Y_2, Y_3, \dots$  be independent and identically distributed random variables. Find the mean and variance of

$$\sum_{i=1}^{N(t)} Y_i.$$

**Solution:**

Conditioning by  $N(t)$  yields for the expectation

$$\mathbb{E} \sum_{i=1}^{N(t)} Y_i = \mathbb{E} \left[ \mathbb{E} \left[ \sum_{i=1}^{N(t)} Y_i | N(t) \right] \right] = \mathbb{E}[N(t)\mathbb{E}Y_1] = \mathbb{E}[N(t)]\mathbb{E}[Y_1] = \left( \int_0^t \lambda(u)du \right) \mathbb{E}[Y_1],$$

while the same procedure yields for the second moment

$$\begin{aligned} \mathbb{E} \left( \sum_{i=1}^{N(t)} Y_i \right)^2 &= \mathbb{E} \left[ \mathbb{E} \left[ \left( \sum_{i=1}^{N(t)} Y_i \right)^2 | N(t) \right] \right] \\ &= \mathbb{E} \left[ \mathbb{E} \left[ \sum_{i,j=1}^{N(t)} Y_i Y_j | N(t) \right] \right] = \mathbb{E} \left[ \mathbb{E} \left[ \sum_{i=1}^{N(t)} Y_i^2 + \sum_{i \neq j} Y_i Y_j | N(t) \right] \right] \\ &\stackrel{Y_i \text{ indep}}{=} \mathbb{E}[N(t)\mathbb{E}[Y_1^2] + \mathbb{E}[N(t)^2 - N(t)](\mathbb{E}Y_1)^2] = \mathbb{E}N(t)\text{Var}Y_1 + \mathbb{E}N(t)^2(\mathbb{E}Y_1)^2. \end{aligned}$$

Hence the variance is given by

$$\begin{aligned} \text{Var} \left( \sum_{i=1}^{N(t)} Y_i \right) &= \mathbb{E}N(t)\text{Var}Y_1 + \text{Var}N(t)(\mathbb{E}Y_1)^2 \\ &= \left( \int_0^t \lambda(u)du \right) (\text{Var}Y_1 + (\mathbb{E}Y_1)^2) = \left( \int_0^t \lambda(u)du \right) \mathbb{E}Y_1^2. \end{aligned}$$

**Problem 3**

Consider a homogeneous Poisson process  $N(t)$  with random (and with respect to time constant) intensity  $\lambda$  which takes two values  $\lambda_1, \lambda_2$  with equal probabilities. (This means that first, someone rolls the intensity according to a fair coinflip, and then generates the Poisson process with this intensity.) Find the probability generating function of  $N(t)$ .

**Solution:**

If  $N(t)$  is a homogeneous Poisson process with deterministic rate  $\lambda$ , then its pgf is given by

$$\phi(z) = \mathbb{E}[z^{N(t)}] = \sum_{k \geq 0} P(N(t) = k)z^k = \sum_{k \geq 0} e^{-\lambda t} \frac{(\lambda t)^k}{k!} z^k = e^{-\lambda t(1-z)}.$$

If the intensity is given by a Bernoulli experiment, we may derive by conditioning that

$$\phi(z) = \mathbb{E}[z^{N(t)}] = \mathbb{E}[\mathbb{E}[z^{N(t)}|\lambda]] = \frac{1}{2} (e^{-\lambda_1 t(1-z)} + e^{-\lambda_2 t(1-z)}).$$

#### Problem 4

Let the times between the events of a renewal process  $N(t)$  be uniformly distributed on  $(0, 1)$ . Find the mean and variance of  $N(t)$  for  $0 < t < 1$ .

#### Solution:

Let  $0 < t < 1$  and consider the pgf of  $N(t)$ :

$$G(z, t) := \mathbb{E}[z^{N(t)}], \quad z \in \mathbb{R}.$$

Let  $T_1, T_2, \dots$  denote the times between the events of the renewal process, which are iid and  $U(0, 1)$ -distributed. Then with  $S_n := T_1 + T_2 + \dots + T_n$  we may write

$$N(\omega, t) = \sum_{n \geq 1} \mathbb{1}\{S_n \leq t\}.$$

We derive a differential equation for  $G(z, t)$  as follows:

$$\begin{aligned} G(z, t) &= \mathbb{E}[z^{N(t)}] = \int z^{N(\omega, t)} \mathbb{P}(d\omega) = \int z^{N(\omega, t)} \mathbb{1}\{T_1 > t\} \mathbb{P}(d\omega) + \int z^{N(\omega, t)} \mathbb{1}\{T_1 < t\} \mathbb{P}(d\omega) \\ &= \int z^0 \mathbb{1}\{T_1 > t\} \mathbb{P}(d\omega) + \int z^{\sum_{n \geq 1} \mathbb{1}\{S_n \leq t\}} \mathbb{1}\{T_1 < t\} \mathbb{P}(d\omega) \\ &= (1 - t) + \int z^{\sum_{n \geq 1} \mathbb{1}\{t_1 + \dots + t_n \leq t\}} \mathbb{1}\{t_1 < t\} \mathbb{P}^{(T_1, T_2, \dots)}(d(t_1, t_2, \dots)) \\ &= (1 - t) + \iint z^{\sum_{n \geq 1} \mathbb{1}\{t_1 + \dots + t_n \leq t\}} \mathbb{1}\{t_1 < t\} \mathbb{P}^{T_1}(dt_1) \mathbb{P}^{(T_2, T_3, \dots)}(d(t_2, t_3, \dots)) \\ &= (1 - t) + \int_0^t \int_0^t z^{1 + \sum_{n \geq 2} \mathbb{1}\{t_2 + \dots + t_n \leq t - t_1\}} \mathbb{P}^{T_1}(dt_1) \mathbb{P}^{(T_2, T_3, \dots)}(d(t_2, t_3, \dots)) \\ &\stackrel{\text{Fubini}}{=} (1 - t) + \int_0^t \int_0^t z^{1 + \sum_{n \geq 2} \mathbb{1}\{t_2 + \dots + t_n \leq t - t_1\}} \mathbb{P}^{(T_2, T_3, \dots)}(d(t_2, t_3, \dots)) \mathbb{P}^{T_1}(dt_1) \\ &= (1 - t) + \int_0^t \int z z^{\sum_{n \geq 2} \mathbb{1}\{T_2 + \dots + T_n \leq t - t_1\}} \mathbb{P}(d\omega) \mathbb{P}^{T_1}(dt_1). \end{aligned}$$

Since  $(T_2, T_3, \dots) \stackrel{d}{=} (T_1, T_2, \dots)$ , this means that

$$G(z, t) = (1 - t) + z \int_0^t G(z, t - t_1) dt_1 = (1 - t) + z \int_0^t G(z, u) du.$$

Differentiating with respect to  $t$  yields

$$\frac{d}{dt}G(z, t) = -1 + zG(z, t).$$

With respect to  $t$  this is a first order linear differential equation with constant coefficients. The homogeneous solutions are obviously given by  $const \cdot e^{zt}$ , while a particular solution is the constant function  $t \mapsto \frac{1}{z}$ . Since necessarily  $G(z, 0) = 1$ , this yields

$$(1) \quad G(z, t) = \frac{1 - (1 - z)e^{zt}}{z}.$$

To find the mean and variance of  $N(t)$ , we need to investigate the first and second derivative of its pgf  $G$  with respect to  $z$ . Equation (1) is equivalent to

$$zG(z, t) = 1 - (1 - z)e^{zt},$$

and differentiating this twice yields (' denotes differentiation with respect to  $z$ )

$$(2) \quad zG'(z, t) + G(z, t) = (1 - (1 - z)t)e^{zt},$$

and

$$(3) \quad zG''(z, t) + 2G'(z, t) = t(1 - (1 - z)t)e^{zt} + te^{zt}.$$

Evaluating (2) at  $z = 1$  yields

$$\mathbb{E}[N(t)] = e^t - 1, \quad 0 < t < 1,$$

while evaluating (3) at  $z = 1$  yields

$$\mathbb{E}[N(t)(N(t) - 1)] + 2\mathbb{E}[N(t)] = 2te^t, \quad 0 < t < 1,$$

which means for the variance that

$$\text{Var}(N(t)) = \mathbb{E}[N(t)(N(t) - 1)] + E[N(t)] - (E[N(t)])^2 = 2te^t - (e^t - 1) - (e^t - 1)^2, \quad 0 < t < 1.$$

For  $t > 1$  the situation becomes pretty messy, since the differential equation for  $G(z, t)$  takes a different form, and its solution seems to require a recursive consideration of the intervals  $(k, k + 1)$ ,  $k \in \mathbb{N}$  where at each step the solution on  $(k, k + 1)$  becomes more and more complicated. Another approach is to use the renewal equation and its Laplace transform, which yields a nice closed form of  $\hat{m}(s)$ . But then again, the inversion of the transform is difficult.