## Problem 1

Let N(t) be a Poisson process of rate  $\lambda$ . For fixed  $t, u \in [0, t]$ , and n = 1, 2, ..., find the conditional distribution of N(u) given N(t) = n, i.e. find a formula for  $\mathbb{P}(N(u) = k | N(t) = n)$  for k = 0, 1, 2, ..., n.

Solution:

Assume that t > 0. By Bayes' rule we have

$$\begin{split} P(N(u) = k | N(t) = n) &= \frac{P(N(u) = k, N(t) = n)}{P(N(t) = n)} \\ &= \frac{P(N(t) = n | N(u) = k) P(N(u) = k)}{P(N(t) = n)}. \end{split}$$

Then, by the memoryless property we have

$$P(N(t) = n | N(u) = k) = P(N(t - u) = n - k),$$

and hence we obtain

$$P(N(u) = k | N(t) = n) = \frac{P(N(t - u) = n - k)P(N(u) = k)}{P(N(t) = n)}$$

$$= \frac{e^{-\lambda(t - u)} \frac{(\lambda(t - u))^{n - k}}{(n - k)!} e^{-\lambda u} \frac{(\lambda u)^k}{k!}}{e^{-\lambda t} \frac{(\lambda t)^n}{n!}}$$

$$= e^{-\lambda(t - u + u - t)} \frac{n!}{(n - k)!k!} \frac{(t - u)^{n - k} u^k}{t^n}$$

$$= e^{-\lambda(0)} \binom{n}{k} \frac{(t - u)^{n - k} u^k}{t^n}$$

$$= \binom{n}{k} \frac{(t - u)^{n - k} u^k}{t^n}$$

$$= \binom{n}{k} \left(\frac{u}{t}\right)^k \left(\frac{t - u}{t}\right)^{n - k}$$

which we may recognise as the pmf of the Binomial distribution with parameters n and  $\frac{t}{u}$ . Rather than solving directly as we did above, we may solve this question more quickly using superposition and the independent increments property of the Poisson process:

Note that we may write

$$P(N(u) = k|N(t) = n) = P(N(u) = k|(N(t) - N(u)) + N(u) = n).$$

Since (N(t) - N(u)) and N(u) are independent by the independent increments property, by superposition we have

$$(N(t) - N(u)) + N(u) \sim \text{Poisson}(\lambda(t - u + u)) = \text{Poisson}(\lambda t).$$

Hence, as we saw for this general set-up in class, the conditional distribution of N(u) given (N(t) - N(u)) + N(u) = n is Binomial $(n, \frac{u}{t})$ , and we obtain the same formula for the conditional probability as above.

See the class notes for another alternative way to solve/understand this question which involves the arrival times of the Poisson process.

## Problem 2

Let N(t) be a Poisson process of rate  $\lambda$ . Given that N(t) = 3, determine the conditional distributions of the first three arrival times  $S_1, S_2, S_3$ . Solution:

In class we showed the result that the conditional joint distribution of  $(S_1, S_2, S_3)$  is given by the joint distribution of a point process of ordered uniform r.v. on [0, t]. In particular, we have that the pdf of the conditional joint distribution is given by

$$f(s_1, s_2, s_3) = \frac{3!}{t^3}$$

for  $0 < s_1 < s_2 < s_3 < t$ . We may now find the conditional distributions for each of the three arrival times by marginalizing over the other two. We obtain:

$$f_{S_1}(s_1) = \int_{s_1}^t \int_{s_1}^{s_3} \frac{3!}{t^3} ds_2 ds_3$$
$$= \frac{3!}{t^3} (\frac{1}{2}t^2 - s_1t + \frac{1}{2}s_1^2)$$

$$f_{S_2}(s_2) = \int_{s_2}^t \int_0^{s_2} \frac{3!}{t^3} ds_1 ds_3$$
$$= \frac{3!}{t^3} s_2(t - s_2)$$

$$f_{S_3}(s_3) = \int_0^{s_3} \int_0^{s_2} \frac{3!}{t^3} ds_1 ds_2$$
$$= \frac{3!}{t^3} \frac{1}{2} s_3^2$$
$$= \frac{3s_3^2}{t^3}$$

## Problem 3

Customers arrive at a theme park according to a Poisson process N(t) of rate  $\lambda$ . Each customer pays \$1 on arrival. At time t, the discounted value of the total sum collected so far is

$$D_t = \sum_{i=1}^{N(t)} e^{-\beta S_i},$$

where  $S_i$  is the *i*th arrival time, and  $\beta > 0$  is the discount rate. Compute  $\mathbb{E}D_t$ .

Solution:

By the law of total probability,

$$\mathbb{E}[D_t] = \mathbb{E}\left[\sum_{i=1}^{N(t)} e^{-\beta S_i}\right],$$

$$= \sum_{n=0}^{\infty} \mathbb{E}\left[\sum_{i=1}^{N(t)} e^{-\beta S_i} | N(t) = n\right] P(N(t) = n)$$

Let  $U_1, U_2, \ldots, U_n$  be i.i.d Uniform((0, t]) random variables. Then since the conditional distribution of the first n arrival times is the same as the distribution of  $(U_1, U_2, \ldots, U_n)$ , we have

$$\mathbb{E}\left[\sum_{i=1}^{N(t)} e^{-\beta S_i} | N(t) = n\right] = \mathbb{E}\left[\sum_{i=1}^n e^{-\beta U_i}\right]$$
$$= \sum_{i=1}^n \mathbb{E}[e^{-\beta U_i}]$$
$$= \sum_{i=1}^n \int_0^t e^{-\beta u} \frac{1}{t} dt$$
$$= \frac{n}{\beta t} (1 - e^{-\beta t}).$$

Hence we obtain that the expectation of D(t) is given by

$$\mathbb{E}[D_t] = \sum_{n=0}^{\infty} \frac{n}{\beta t} (1 - e^{-\beta t}) e^{-\lambda t} \frac{(\lambda t)^n}{n!}$$

$$= (1 - e^{-\beta t}) e^{-\lambda t} \frac{\lambda}{\beta} \sum_{n=1}^{\infty} \frac{(\lambda t)^{n-1}}{(n-1)!}$$

$$= (1 - e^{-\beta t}) e^{-\lambda t} \frac{\lambda}{\beta} e^{\lambda t}$$

$$= \frac{\lambda}{\beta} (1 - e^{-\beta t})$$

## Problem 4

Alpha particles are emitted by a radioactive source according to a Poisson process of rate  $\lambda$ . Each alpha particle independently survives for a random amount of time and then is annihilated. The lifetimes  $Y_1, Y_2, \ldots$  of the particles have common distribution function  $G(y) = \mathbb{P}(Y_k \leq y)$ . Let M(t) denote the number of alpha particles in existence at time t.

a. Determine the distribution of M(t).

Solution:

Let us denote the number of particles created by the emission process by time t as  $N(t) \sim Poisson(\lambda t)$  where the distribution is determined by the Poisson process assumption. Let  $0 \leq S_1, S_2, \ldots, S_n \leq t$  be the emission times (i.e. arrival times) of the first n. Then a particle k exists at time t if and only if

$$S_k + Y_k \ge t$$
.

With this observation we see that we may express M(t) as:

$$M(t) = \sum_{k=0}^{N(t)} \mathbf{1}_{S_k + Y_k \ge t}.$$

Additionally, by the law of total probability we have, for any  $m \in \mathbb{N}$ ,

$$P(M(t) = m) = \sum_{n=0}^{\infty} P(M(t) = m|N(t) = n)P(N(t) = n).$$

Let  $U_1, U_2, \ldots, U_n$  be i.i.d Uniform((0, t]) random variables. Then since the conditional distribution of the first n arrival times is the same as the distribution of the ordered uniform random variables  $(U_{(1)}, U_{(2)}, \ldots, U_{(n)})$ , where  $U_{(i)}$  denotes the ith smallest value out of  $U_1, U_2, \ldots, U_n$ , we obtain the following result for the conditional probability of M(t) given N(t) = n

$$P(M(t) = m | N(t) = n) = P(\sum_{k=0}^{N(t)} \mathbf{1}_{S_k + Y_k \ge t} = m | N(t) = n)$$

$$= P(\sum_{k=0}^{n} \mathbf{1}_{U_k + Y_k \ge t} = m)$$

$$= \binom{n}{m} P(U_k + Y_k \ge t)^m (1 - P(U_k + Y_k \ge t))^{n-m}$$

where we recognize the binomial distribution with parameter  $p = P(U_k + Y_k \ge t)$ . This probability p is given by

$$p = \int_0^t P(Y_k \ge t - U_k | U_k = u) P(U_k = u) du$$

$$= \frac{1}{t} \int_0^t P(Y_k \ge t - u) du$$

$$= \frac{1}{t} \int_0^t (1 - P(Y_k \le t - u)) du$$

$$= \frac{1}{t} \int_0^t (1 - G(t - u)) du$$

$$= \frac{1}{t} \int_0^t (1 - G(y)) dy$$

Recognizing the pdf for the binomial, we have found  $M(t)|N(t) = n \sim \text{Binomial}(n, \frac{\mu}{t})$ . Returning to the distribution of M(t) we now have

$$\begin{split} P(M(t) = m) &= \sum_{n=0}^{\infty} P(M(t) = m | N(t) = n) P(N(t) = n) \\ &= \sum_{n=0}^{\infty} \binom{n}{m} p^m (1 - p)^{n-m} e^{-\lambda t} \frac{(\lambda t)^n}{n!} \\ &= e^{-\lambda t} p^m \sum_{n=0}^{\infty} \frac{n!}{(n - m)! m!} (1 - p)^{n - m} \frac{(\lambda t)^n}{n!} \\ &= e^{-\lambda t} p^m \frac{(\lambda t)^m}{m!} \sum_{n=0}^{\infty} \frac{1}{(n - m)!} (\lambda t)^{n - m} (1 - p)^{n - m} \\ &= e^{-\lambda t} \frac{(p \lambda t)^m}{m!} e^{\lambda t (1 - p)} \\ &= e^{-p \lambda t} \frac{(p \lambda t)^m}{m!}, \end{split}$$

i.e. the number of alpha particles in existence at time t follows a Poisson distribution:

$$M(t) \sim \text{Poisson}(p\lambda t)$$
.

b. Show that as  $t \to \infty$ , the distribution you found in part a converges to Poisson $(\lambda \mu)$ , where  $\mu = \mathbb{E}Y$  is the mean lifetime of an alpha particle.

Solution:

Recall that we found in (a) that  $M(t) \sim \text{Poisson}(p\lambda t)$  where

$$p = \frac{1}{t} \int_0^t (1 - G(y)) dy.$$

Then as  $t \to \infty$  we have that the parameter of the distribution converges to  $\lambda \mu$  since

$$\lim_{t \to \infty} \lambda pt = \lambda \int_0^\infty (1 - G(y)) dy = \lambda \mu$$

where we recognise the integral as the mean lifetime of an alpha particle  $\mu = \mathbb{E}Y$ . Hence the distribution converges to Poisson( $\lambda\mu$ ).