

ESE 520 Probability and Stochastic Processes

Lecture 17

”The Poisson process continued.”

What can be said about the relation of a Poisson process $(N_t), t \geq 0$ and the sequence of random variables $\{T_k\}, k = 1, 2, \dots$ from the point of view of their probability distributions?

First, we can write

$$F_1(t) = P(T_1 \leq t) = 1 - P(T_1 > t) = 1 - P(N_t = 0) = 1 - P_0(t) = 1 - e^{-\lambda t}$$

so that the pdf of T_1 is given by

$$f(t) = \begin{cases} \lambda e^{-\lambda t}, & t > 0 \\ 0, & t \leq 0. \end{cases}$$

Therefore, T_1 has the *exponential distribution* with parameter λ .

For any $k = 1, 2, \dots$, we have that

$$\begin{aligned} F_k(t) &= P(T_k \leq t) = P(N_t > k - 1) = \\ &1 - P(N_t \leq k - 1) = 1 - \sum_{j=0}^{k-1} P(N_t = j) = 1 - \sum_{j=0}^{k-1} \frac{(\lambda t)^j}{j!} e^{-\lambda t}. \end{aligned}$$

The density of T_k is then found as

$$\begin{aligned} f_k(t) &= F'_k(t) = -\left(\sum_{j=0}^{k-1} j \lambda \frac{(\lambda t)^{j-1}}{j!}\right) e^{-\lambda t} + \lambda \sum_{j=0}^{k-1} \frac{(\lambda t)^j}{j!} e^{-\lambda t} = \\ &\lambda e^{-\lambda t} \left[-\sum_{j=1}^{k-1} \frac{(\lambda t)^{j-1}}{(j-1)!} + \sum_{j=0}^{k-1} \frac{(\lambda t)^j}{j!} \right] = \\ &\lambda e^{-\lambda t} \left[-\sum_{i=0}^{k-2} \frac{(\lambda t)^i}{(i)!} + \sum_{i=0}^{k-1} \frac{(\lambda t)^i}{i!} \right] = \\ &\lambda e^{-\lambda t} \frac{(\lambda t)^{k-1}}{(k-1)!}, \quad k = 1, 2, \dots \end{aligned} \tag{1}$$

The pdf $f_k(t)$ found in (1) is called the *Erlang's probability distribution*, with notation Erlang (k, λ) , $k = 1, 2, \dots$

We see that Erlang (k, λ) is nothing but the Gamma probability distribution $\Gamma(\alpha, \beta)$ (in notation we used before), where the first parameter $\alpha = k$ is an integer and the second parameter $\beta = \frac{1}{\lambda}$.

Using general properties of Gamma distribution, we have that

$$E(T_k) = \frac{k}{\lambda},$$

and

$$Var(T_k) = \frac{k}{\lambda^2}.$$

Moreover, for any $t > 0$, we can formally calculate that

$$\begin{aligned} P(T_k > s + t | T_{k-1} = s) &= P(N_{s+t} = k - 1 | N_s = k - 1) = \\ P(N_{s+t} - N_s = 0 | N_s = k - 1) &= P(N_t = 0) = e^{-\lambda t} \end{aligned}$$

since N has independent increments.

It implies that

$$P(T_k - T_{k-1} > t | T_{k-1} = s) = e^{-\lambda t}$$

meaning that *random variable $T_k - T_{k-1}$ is independent of earlier arrivals and has the same distribution as T_1 .*

Thus, by letting

$$\tau_k := T_k - T_{k-1},$$

we have shown that **random variables $\tau_k, k = 1, 2, \dots$ are i.i.d. random variables each of which has exponential distribution with parameter $\lambda > 0$.**

Also, it holds

$$T_k = \sum_{i=1}^k \tau_i \text{ with } \tau_0 = 0.$$

Example 1. Let $(N_t), t \geq 0$ be a Poisson process with parameter $\lambda > 0$. Show that

$$\lim_{t \rightarrow \infty} \frac{N_t}{t} = \lambda \text{ a.e.}$$

Solution: Choose $n \in \mathbb{N}$ and write

$$N_n = N_1 + (N_2 - N_1) + \dots + (N_n - N_{n-1}) = \sum_{i=1}^n X_i,$$

where $X_i := N_i - N_{i-1}$.

By independence of increments, all X_1, \dots, X_n are independent random variables with

$$E(X_i) = \lambda, i = 1, 2, \dots, n.$$

By strong law of large numbers,

$$\frac{1}{n} N_n \rightarrow \lambda \text{ as } n \rightarrow \infty \text{ a.e.}$$

Then, for any $t > 1$, we find $n \in \mathbb{N}$ such that $n \leq t \leq n+1$ and

$$N_n \leq N_t \leq N_{n+1}$$

implying

$$\frac{n}{n+1} \frac{N_n}{n} \leq \frac{N_n}{n+1} \leq \frac{N_t}{n+1} \leq \frac{N_{n+1}}{n+1}.$$

Since $\lim_{n \rightarrow \infty} \frac{N_n}{n} = \lim_{n \rightarrow \infty} \frac{N_{n+1}}{n+1}$, it follows that $\lim_{t \rightarrow \infty} \frac{N_t}{t} = \lambda$ a.e.

Alternative characterization of Poisson process $(N_t), t \geq 0$:

Let τ_1, τ_2, \dots be a sequence of independent random variables each of which having the same exponential distribution with parameter $\lambda > 0$.

For example, τ_i could be times between the emissions of radioactive particles or times between calls made at a telephone tower.

Define

$$T_n := \tau_1 + \dots + \tau_n \text{ with } \tau_0 = 0, n = 0, 1, \dots$$

T_n can be interpreted as time of n-th emission (the n-th call).

Let

$$N_t := \max_n \{n : T_n \leq t\}, t \geq 0.$$

N_t is then the number of calls made up to time t (on the interval $[0, t]$).

Alternatively,

$$N_t = \sum_{k=1}^{\infty} \mathbf{1}_{\{T_k \leq t\}}.$$

Theorem. It holds that

$$P(N_t = n) = e^{-\lambda t} \frac{(\lambda t)^n}{n!}, n = 0, 1, 2, \dots$$

Proof. We observe that

$$\{N_t < n\} = \{T_n > t\}.$$

Also, since (N_t) is an integer-valued process, we can write

$$\begin{aligned} P(N_t = n) &= P(N_t < n + 1) - P(N_t < n) = \\ &= P(T_{n+1} > t) - P(T_n > t). \end{aligned}$$

It is then enough to prove that

$$P(T_n > t) = e^{-\lambda t} \sum_{k=0}^{n-1} \frac{(\lambda t)^k}{k!}, n = 1, 2, \dots \quad (2)$$

We prove (2) by induction.

For $n = 1$:

$$P(T_1 > t) = P(\tau_1 > t) = e^{-\lambda t}.$$

Assume that (2) is true for some n . Then

$$\begin{aligned} P(T_{n+1} > t) &= P(T_n + \tau_{n+1} > t) = \\ P(\tau_{n+1} > t) + P(T_n > t - \tau_{n+1}, t \geq \tau_{n+1} > 0) &= \\ e^{-\lambda t} + \int_0^t P(T_n > t - s) f_{\tau_{n+1}}(s) ds &\stackrel{\sim}{=} \end{aligned}$$

where the last formula is the result of using the *conditioning formula* on the variable τ_{n+1} .

Remark 1. (the conditioning formula)

a) If X and Y are discrete random variables, then we can write

$$P(X = x) = P(X = x, Y \in \mathbb{R}) = \sum_y P(X = x, Y = y) =$$

$$\sum_y P(X = x|Y = y)P(Y = y) = \sum_y P(X = x|Y = y)f_Y(y).$$

b) Similarly, for continuous random variables X and Y :

$$P(X \in B) = \dots = \int P(X \in B|Y = y)f_Y(y)dy.$$

We continue with the calculations:

$$\begin{aligned} &\stackrel{\sim}{=} e^{-\lambda t} + \int_0^t e^{-\lambda(t-s)} \sum_{k=0}^{n-1} \frac{[\lambda(t-s)]^k}{k!} \lambda e^{-\lambda s} ds = \\ &e^{-\lambda t} + e^{-\lambda t} \sum_{k=0}^{n-1} \frac{\lambda^{k+1}}{k!} \int_0^t (t-s)^k ds = \\ &e^{-\lambda t} + e^{-\lambda t} \sum_{k=0}^{n-1} \frac{\lambda^{k+1}}{k!} \frac{t^{k+1}}{k+1} = \\ &e^{-\lambda t} + e^{-\lambda t} \sum_{k=0}^{n-1} \frac{(\lambda t)^{k+1}}{(k+1)!} = \\ &e^{-\lambda t} \sum_{k=0}^n \frac{(\lambda t)^k}{(k)!}. \end{aligned}$$

It is also not hard to see that N will be a process with stationary and independent increments. \square

Example 2. Let (N_t) be a Poisson process with $\lambda = 2$.

a) Find $P(N_5 = 10)$. By definition of N ,

$$P(N_5 = 10) = e^{-\lambda t} \frac{(\lambda t)^{10}}{10!} = [\lambda = 2, t = 5] = e^{-10} \frac{10^{10}}{10!}.$$

b) Find the probability that $N_i - N_{i-1} = 2$ for all $i = 1, \dots, 5$.

We have:

$$P(N_i - N_{i-1} = 2, 1 = 1, \dots, 5) = P(\cap_{i=1}^5 \{N_i - N_{i-1} = 2\}) = \\ \prod_{i=1}^5 \left(\frac{2^2}{2!} e^{-2} \right) = \left(\frac{2^2}{2!} e^{-2} \right)^5.$$

Example 3. Customers arrive at a store according to a Poisson process of rate λ . What is the expected time until the nth customer arrives? What is the expected time between customers?

Solution: We use the property of the Poisson process following from its construction:

$$T_n = \tau_1 + \dots + \tau_n,$$

where T_n is then the time of arrival of nth customer and τ_i is the time between arrivals of two corresponding customers.

We know that

$$E(\tau_i) = \frac{1}{\lambda}$$

and

$$E(T_n) = nE(\tau_i) = \frac{n}{\lambda}.$$

We also know that, in fact, T_n has Erlang (n, λ) distribution.