

Lecture 16

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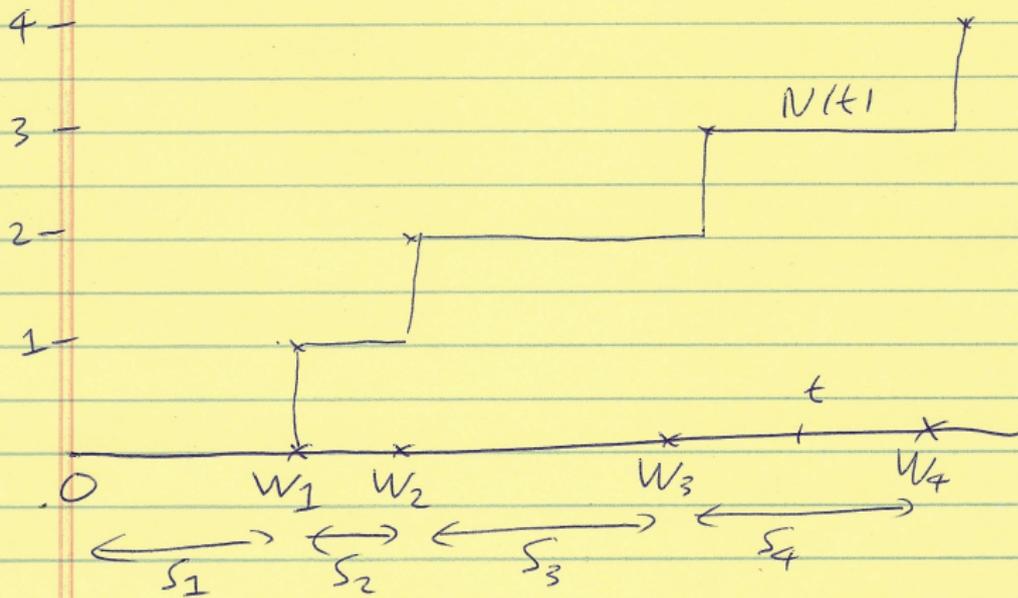
Notation for studying random times (of events) over $0 \leq t < \infty$. Cannot have two events at the same time.

- W_k = time at which k 'th event occurs ($W_0 = 0$).
- $S_k = W_k - W_{k-1}$ is the time between successive events.
- $N(t)$ = number of events during time $[0, t]$.
- $N(s, t) = N(t) - N(s)$ = number of events during $(s, t]$.

Note that the event $\{W_n \leq t\}$ is the same as the event $\{N(t) \geq n\}$. So, regardless of the probability model,

$$\mathbb{P}(W_n \leq t) = \mathbb{P}(N(t) \geq n).$$

We will study the mathematically simplest probability model. Fix a parameter $0 < \lambda < \infty$.



Poisson point process (PPP) of rate λ .

This process is defined by the properties

- (a) $N(s, t)$ has $\text{Poisson}(\lambda(t - s))$ distribution.
- (b) For disjoint intervals $(s_1, t_1), (s_2, t_2), \dots, (s_k, t_k)$ the random variables $N(s_1, t_1), N(s_2, t_2), \dots, N(s_k, t_k)$ are independent.

Why does such a process exist? Consider large M , and suppose events could only happen at times $\frac{1}{M}, \frac{2}{M}, \frac{3}{M}, \dots$, independently with probability λ/M . Then the number of events during (s, t) is almost $\text{Binomial}(M(t - s), \lambda/M)$. We picture the PPP as the $M \rightarrow \infty$ limit, and the Poisson distribution arises as limit of Binomials.

A more informal description in terms of infinitesimal intervals is

- (a') $\mathbb{P}(\text{event during } [t, t + dt]) = \lambda dt$
- (b') What happens in disjoint time intervals is independent.

Conceptual point; “Poisson” is not an arbitrary assumption, but instead arises automatically from (a', b'). Also λ is a “rate”: the mean number of events per unit time.

The PPP is used as an over-simplified model for events that occur at “completely random” times

- accidents
- coincidences
- start of phone calls (from phone company viewpoint)
- customer joining supermarket checkout line (from the supermarket viewpoint)
- earthquakes
- murders

For many of these examples we know that in fact the rates vary with time. But we can adapt the model to allow a time-varying rate function $\lambda(t)$.

The PPP with rate function $\lambda(t)$.

Start with the informal description in terms of infinitesimal intervals:

(a') $\mathbb{P}(\text{event during } [t, t + dt]) = \lambda(t) dt$

(b') What happens in disjoint time intervals is independent.

We can then deduce the other description. Write $\Lambda(t) = \int_0^t \lambda(u) du$.

(a) $N(s, t)$ has $\text{Poisson}(\Lambda(t) - \Lambda(s))$ distribution.

(b) For disjoint intervals (s_1, t_1) , (s_2, t_2) , \dots , (s_k, t_k) the random variables $N(s_1, t_1)$, $N(s_2, t_2)$, \dots , $N(s_k, t_k)$ are independent.

We will see details later. For now, the **mathematical** point is that we can deduce results for the “rate function $\lambda(t)$ ” case from results for the “constant rate λ ” case, so we can set up theory in the constant rate case. A **modeling** point is that any process of random events has some mean rate function $\lambda(t)$ – what is special about the Poisson process is the independence property. Is this approximately true in a given example?

Poisson point process (PPP) of rate λ defined by the properties

(a) $N(s, t)$ has $\text{Poisson}(\lambda(t - s))$ distribution.

(b) For disjoint intervals $(s_1, t_1), (s_2, t_2), \dots, (s_k, t_k)$ the random variables $N(s_1, t_1), N(s_2, t_2), \dots, N(s_k, t_k)$ are independent.

In this model we will consider

- $W_k =$ time at which k 'th event occurs ($W_0 = 0$).
- $S_k = W_k - W_{k-1}$ is the time between successive events.

The first results are [board]

- W_1 has $\text{Exponential}(\lambda)$ distribution
- W_k has $\text{Gamma}(\lambda, k)$ distribution:

$$f_{W_k}(t) = \frac{\lambda^k t^{k-1} e^{-\lambda t}}{(k-1)!}.$$

- S_1, S_2, S_3, \dots are IID $\text{Exponential}(\lambda)$.

Note that we can use the fact

$$S_1, S_2, S_3, \dots \text{ are IID Exponential}(\lambda)$$

as a (mathematical) construction of the PPP, or as an easy way to simulate the process.

Here is the next result.

Theorem

Fix $t > 0$ and $k \geq 1$. Conditional on $\{N(t) = k\}$ the times (W_1, W_2, \dots, W_k) of events in the PPP are distributed as the order statistics of k IID $\text{Uniform}(0, t)$ random variables.

[board]

The same result holds if instead we condition on $\{W_{k+1} = t\}$.