

# Poisson Rates

*Samantha Bothwell*

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## Definition

The Poisson Distribution is discrete and used for count data, where you determine the probability that a given number of events occur within a time period. This distribution takes one parameter, often denoted with the symbol  $\lambda$  which is a known fixed rate which represents the number of events expected to occur in a certain time interval. We would say

$$N \sim \text{Poisson}(\lambda)$$

## Probability Mass Function

The probability of observing  $k$  events occur within a time interval is given by the equation:

$$P(k \text{ events occur}) = \frac{e^{-\lambda} \lambda^k}{k!}$$

where  $k = 0, 1, 2, \dots$

This pmf can be adapted so that the parameter is a time rate,  $\lambda = rt$ . In this case,  $r$  is 1 unit per time period. The equation then changes to the following:

$$P(k \text{ events occur in time } t) = \frac{e^{-rt} (rt)^k}{k!}$$

## Poisson vs Binomial Distribution

Use the Poisson distribution when you are trying to determine the probability of a certain number of defined events happening in a given time period.

Use the Binomial distribution when you are trying to determine the probability of a certain number of defined events happening out of a determined number of trials.

The binomial distribution converges to the poisson for a large sample size. (i.e.  $np \rightarrow \lambda$  for large  $n$ )

## Poisson Process

To explain the poisson process, I will first define the following variables:

- $N_t$  = count of the number of events that are registered at time  $t$
- $v$  = the mean number of events per unit time (*intensity*)
- $vt$  = mean number of events in  $t$  time units

Here,  $N_t \sim \text{Poisson}(\lambda)$  where  $\lambda = vt$

## Definition

The collection  $N_t : t \geq 0$  is called the poisson process if it satisfies the following:

1.  $N_0 = 0$
2. If  $s < t$ ,  $N_s$  and  $N_t - N_s$  are independent (*memoryless*)
3.  $N_{s+t} - N_t$  and  $N_s$  have the same distribution (*stationary*)
4.  $\lim_{t \rightarrow 0} \frac{P(N_t=1)}{t} = \lambda$
5.  $\lim_{t \rightarrow 0} \frac{P(N_t \geq 1)}{t} = 0$  (can't have two events happen at the same time)

## Descriptive Statistics

Based on the pmf, let's derive the mean and variance of the poisson distribution:

$$\begin{aligned}\mu = E[N] &= \sum_{k=0}^{\infty} k \frac{e^{-\lambda} \lambda^k}{k!} \\&= \sum_{k=0}^{\infty} \frac{e^{-\lambda} \lambda^k}{(k-1)!} \\&= \lambda \sum_{k=1}^{\infty} \frac{e^{-\lambda} \lambda^{k-1}}{(k-1)!} \quad \text{Let } m = k - 1 \\&= \lambda \sum_{m=0}^{\infty} \frac{e^{-\lambda} \lambda^m}{m!} \\&= \lambda \cdot 1 \quad \text{Since } \sum_{k=0}^{\infty} \frac{e^{-\lambda} \lambda^k}{k!} = 1 \\&= \lambda\end{aligned}$$

We will calculate variance using the formula  $\sigma^2 = E[N(N-1)] + E[N] - E[N]^2$ .

We will first derive  $E[N(N-1)]$

$$\begin{aligned}E[N(N-1)] &= \sum_{k=0}^{\infty} k(k-1) \frac{e^{-\lambda} \lambda^k}{k!} \\&= \sum_{k=0}^{\infty} \frac{e^{-\lambda} \lambda^k}{(k-2)!} \\&= \lambda^2 \sum_{k=2}^{\infty} \frac{e^{-\lambda} \lambda^{k-2}}{(k-2)!} \quad \text{Let } m = k - 2 \\&= \lambda^2 \sum_{m=0}^{\infty} \frac{e^{-\lambda} \lambda^m}{m!} \\&= \lambda^2 \cdot 1 \quad \text{Since } \sum_{k=0}^{\infty} \frac{e^{-\lambda} \lambda^k}{k!} = 1 \\&= \lambda^2\end{aligned}$$

Now, we can solve for the variance:

$$\sigma^2 = Var[N] = E[N(N-1)] + E[N] - E[N]^2 = \lambda^2 + \lambda - \lambda^2 = \lambda$$

So, for the poisson distribution,  $\mu = \sigma^2 = \lambda$

### **Example**

An example of using the poisson process to model the number of lightening strikes can be found in the following article:

Stochastic Modeling of Lightning Occurrence by Nonhomogeneous Poisson Process