Section 02 - Poisson distribution

MCB 111

September 16, 2022

The number of independent events in a time interval

An example from physical science: you are watching a meteor shower from Charles River. Suppose meteors arrive independently from each other, and during a very small time interval δt minute, the probability of seeing a meteor is proportional to the length of the time interval:

$$\mathbb{P}(\text{seeing a meteor during } \delta t \text{ minute}) \approx \lambda \delta t \tag{1}$$

How many stars (X) can I see for the entire meteor shower that lasts T minutes?

What is the probability that I see no meteor at all?

$$\mathbb{P}(X = 0) \approx (1 - \lambda \delta t)^{\frac{T}{\delta t}}$$

$$\approx \exp\left(-\lambda \delta t \frac{T}{\delta t}\right)$$

$$= \exp\left(-\lambda T\right)$$
(2)

What is the probability that I see exactly one meteor?

$$\mathbb{P}(X=1) \approx \binom{T/\delta t}{1} (\lambda \delta t)^{1} (1 - \lambda \delta t)^{\frac{T}{\delta t} - 1}$$

$$\approx \frac{T}{\delta t} (\lambda \delta t) \exp\left(-\lambda \delta t \frac{T}{\delta t} - \lambda \delta t\right)$$

$$\approx \lambda T \exp\left(-\lambda T\right)$$
(3)

What is the probability that I see exactly two meteors?

$$\mathbb{P}(X=1) \approx \begin{pmatrix} T/\delta t \\ 2 \end{pmatrix} (\lambda \delta t)^{2} (1 - \lambda \delta t)^{\frac{T}{\delta t} - 2}$$

$$\approx \frac{\frac{T}{\delta t} \left(\frac{T}{\delta t} - 1\right)}{2!} (\lambda \delta t)^{2} \exp\left(-\lambda \delta t \frac{T}{\delta t} - 2\lambda \delta t\right)$$

$$\approx \frac{(\lambda T)^{2}}{2!} \exp\left(-\lambda T\right)$$
(4)

What is the probability that I see exactly *n* meteors?

$$\mathbb{P}(X = n) \approx \binom{T/\delta t}{n} (\lambda \delta t)^{n} (1 - \lambda \delta t)^{\frac{T}{\delta t} - n}$$

$$\approx \frac{\frac{T}{\delta t} \cdots \left(\frac{T}{\delta t} - n + 1\right)}{n!} (\lambda \delta t)^{n} \exp\left(-\lambda \delta t \frac{T}{\delta t} - n\lambda \delta t\right)$$

$$\approx \frac{(\lambda T)^{n}}{n!} \exp\left(-\lambda T\right)$$
(5)

This formula gives us the Poisson distribution with parameter " λT "

$$\mathbb{P}(X=n) = \frac{\lambda^n}{n!} \exp(-\lambda), \quad n = 0, 1, 2, \cdots$$
 (6)

Very useful when you are counting something that occurs independently in space/time with a fixed rate

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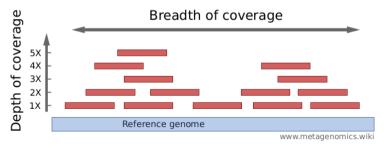
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Genome mutations

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• Shotgun sequencing: the number of reads covering a site in the genome



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• The number of bird poops on the floor



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- The number of fruitflies in BioLabs (Maybe, but correlation induced by experiments?)

Nice properties of the Poisson distribution

Parameter $\lambda = Mean = Variance$

$$\mathbb{E}(X) = 0 + \sum_{n=1}^{\infty} n \exp(-\lambda) \frac{\lambda^n}{n!}$$

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

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"Variance= λ " is left for you to do

The sum of independent Poisson variables is also Poisson

$$egin{aligned} X_1 &\sim \mathsf{Poisson}(\lambda_1) \ X_2 &\sim \mathsf{Poisson}(\lambda_2) \ &\downarrow \ X_1 + X_2 &\sim \mathsf{Poisson}(\lambda_1 + \lambda_2) \end{aligned} \tag{8}$$

Why?

Intuitive answer:

In the meteor shower example, if λ_1 is the rate of seeing red meteors, and λ_2 is the rate of seeing green meteors, then in a short interval δ_t , the probability of seeing either a green or a red meteor is

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 $\mathbb{P}(\text{seeing some meteor in } \delta t \text{ minutes})$ $= 1 - \mathbb{P}(\text{seeing nothing})$ $= 1 - (1 - \lambda_1 \delta t)(1 - \lambda_2 \delta t)$ $= (\lambda_1 + \lambda_2)\delta t - \lambda_1 \lambda_2 (\delta t)^2$ $\approx (\lambda_1 + \lambda_2)\delta t$ (9)

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$$\begin{split} &\mathbb{P}(\text{seeing some meteor in } \delta t \text{ minutes}) \\ &= 1 - \mathbb{P}(\text{seeing nothing}) \\ &= 1 - (1 - \lambda_1 \delta t)(1 - \lambda_2 \delta t) \\ &= (\lambda_1 + \lambda_2)\delta t - \lambda_1 \lambda_2 (\delta t)^2 \\ &\approx (\lambda_1 + \lambda_2)\delta t \end{split} \tag{9}$$

So you can define the rate of seeing either red or green meteors as $\lambda_1 + \lambda_2$, which defines another Poisson process (as well as a Poisson distribution).

This summation property is very useful in biology for developing null hypotheses where sub-processes of interest are independently Poisson. Some examples:

- The spike trains of some independent neurons
- Mutations at independent genomic regions
- DNA double-strand breaks across a group of cells

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In general, in a Poisson process with n types of outcomes, the probability that the next outcome is of type i is

$$\frac{\lambda_i}{\lambda_1 + \lambda_2 + \dots + \lambda_n} \tag{10}$$