

# Probability distributions

### What is a probability distribution?

Think of a probability distribution as a function that maps an elementary event (i.e. every time an observation happens, one and only one of these events will happen) to the probability that that event happens.

## Terms###

### Probability mass function (PMF)

probability that a discrete random variable is exactly equal to some value (i.e. the definition of probability distribution that I gave above)

### Cumulative density function (CDF)

Probability that a random variable ( $X$ ) (or distribution function) will take a value less than or equal to  $x$ . This should always approach 1. This is the area under the **PMF** from  $-\infty$  to  $x$  ( $\int_{-\infty}^x \text{PMF}$ ).

### Probability density

For continuous variables, it would not make sense to talk about the probability that  $X$  takes on a certain value since that assumes an infinite precision- so the probability of a continuous random variable attaining a specific value is 0. Instead, we use the probability density function- where you take the integral to find the probability that the observation will be within a range of values.

## Binomial

Describes discrete events that end in success or failure.

### Examples

- Roll a dice, what is the probability that it's 1? (success: dice=1, failure: dice  $\neq$  1)
- flip a coin, what is the probability of getting heads? (success: head, failure: tail)

Binomial distributions have 2 parameters,  $N$  (or  $n$ ) and  $\theta$  (or  $p$ ).  $N$  is the number of trials,  $\theta$  is the probability of success.

### Formula

$$P(X|\theta, N) = \frac{N!}{X!(N-X)!} \theta^X (1-\theta)^{N-X}$$

## Properties

- expected value:  $E[X] = N\theta$
- variance:  $\text{Var}(X) = N\theta(1 - \theta)$

## Normal

Bell curve distribution.

**Parameters:** mean ( $\mu$ ) and standard deviation ( $\sigma$ )

### Formula

$$p(X|\mu, \sigma) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(X - \mu)^2}{2\sigma^2}\right)$$

## Properties

- expected value:  $E[X] = \mu$
- variance:  $\text{Var}(X) = \sigma^2$

## t

Similar to normal distribution - use this when you think the data is normally distributed but don't know the mean or standard deviation

## Chi-Square ( $\chi^2$ )

Often found in categorical data analysis. This is the result of taking a sum of squares of normally distributed variables

## F

Pops up when comparing two different  $\chi^2$  distributions

## Poisson distribution

Discrete distribution which describes how many times an event (specifically a poisson process) is likely to occur within some period of time

**poisson point process** - a process that generates discrete events where the average time between events is known.

requirements:

- events are independent

- average rate is constant
- two events cannot occur at the same time

### Formula

$$P(X = x|\lambda) = \frac{e^{-\lambda}\lambda^x}{x!}$$

Interpret this as the probability that  $x$  events will happen in the time period where  $\lambda$  events happen per time period on average.

### properties

- expected value:  $E[X] = \lambda$
- variance:  $V(X) = \lambda$

## Exponential distribution

Continuous probability distribution modeling the time between events in a poisson point process

$$p(x|\lambda) = \begin{cases} \lambda e^{-\lambda x} & x \geq 0 \\ 0 & x < 0 \end{cases}$$

### properties

- expected value:  $E[X] = 1/\lambda$
- variance:  $V(X) = 1/\lambda^2$

## References

- Learning statistic with R chapter 9
- [https://en.wikipedia.org/wiki/Binomial\\_distribution](https://en.wikipedia.org/wiki/Binomial_distribution)
- <https://byjus.com/maths/poisson-distribution/>
- <https://towardsdatascience.com/the-poisson-distribution-and-poisson-process-explained-4e2cb17d459>
- [https://en.wikipedia.org/wiki/Exponential\\_distribution](https://en.wikipedia.org/wiki/Exponential_distribution)