# Introduction to statistics TODO subtitle

Caio Volpato (caioau)

 ${\sf caioau.keybase.pub} \to {\sf caioauheyuuiavlc.onion}$ 

210B C5A4 14FD 9274 6B6A 250E EFF5 B2E1 80F2 94CE

All Copylefts are beautiful: licensed under CC BY-SA 4.0

# Summary:

- Probability concepts
- Discrete distributions.
- Continuous distributions.
- Calculations on the Normal distribution.
- Convergence
- Inference

### Motivation



Figure 1: Dados apontam  $\dots$  (data shows  $\dots$ )

motivation exemple cholate

### Sample space $\Omega$

It's the set of all the possible outcomes of a experiment, denoted by S or  $\boldsymbol{\Omega}$ 

#### **Event**

It's a subset of the sample space.

## Probability (Definition):

Given a experiment with a sample space  $\Omega$  and a class of events  $\mathcal{A}$ , the probability denoted by  $\mathbb{P}$  is a function which has  $\mathcal{A}$  as domain and associate a numerical value between [0,1] as image.

### Probability properties:

- lacksquare  $\mathbb{P}(\Omega)=1$  and  $\mathbb{P}(\emptyset)=0$
- $0 \leq \mathbb{P}(A) \leq 1$ , for every event A
- § For any sequence of mutually exclusive events  $A_1, A_2, ...$  that's events that  $A_i \cap A_j$  when  $i \neq j$  we have that:

$$\mathbb{P}\left(\bigcup_{i=1}^{\infty}A_i\right)=\sum_{i=1}^{\infty}\mathbb{P}(A_i)$$

#### Event independence:

Two events are independent when the occurrence of the first does not affect the probability of ocurrence of the second.

Two events A and B are independent if:

$$\mathbb{P}(A \cap B) = \mathbb{P}(A)\mathbb{P}(B)$$

#### Conditional Events:

The probability of a event A to occur given that the event B occurred is:

$$\mathbb{P}(A|B) = \frac{\mathbb{P}(A \cap B)}{\mathbb{P}(B)}$$

### Bayes theorem:

$$\mathbb{P}(A|B) = \frac{\mathbb{P}(B|A)\mathbb{P}(A)}{\mathbb{P}(B)}$$

General case:

$$\mathbb{P}(A_i|B) = \frac{\mathbb{P}(B|A_i)\mathbb{P}(A_i)}{\sum_{j=1}^n \mathbb{P}(B|A_j)\mathbb{P}(A_j)}$$

## Bayes example (from Veritasium):

You are felling sick, so you go to the doctor, there you run a battery of tests. After getting the results you tested positive for a rare disease (affects 0.1% of the population), the test will correctly identify that you have it 99% of the times.

What's the chances that you actually have the disease? 99%?

### Bayes example Solution

Let's denote the event of you have the disease H (stands for hypothesis, the prior) and the test been positive denoted by E (stands for evidence), so we have:  $\mathbb{P}(H) = 0.001$  and  $\mathbb{P}(E|H) = 0.99$ 

$$\mathbb{P}(H|E) = \frac{\mathbb{P}(E|H)\mathbb{P}(H)}{\mathbb{P}(E)} = \frac{\mathbb{P}(E|H)\mathbb{P}(H)}{\mathbb{P}(H)\mathbb{P}(E|H) + \mathbb{P}(H^C)\mathbb{P}(E|H^C)} = \frac{0.99 \cdot 0.001}{0.001 \cdot 0.99 + 0.999 \cdot 0.01} = 0.09 = 9\%$$

What if you test again and it's also positive? You can just take the posterior probability we just calculated and use as a prior:

$$= \frac{0.99 \cdot 0.09}{0.09 \cdot 0.99 + 0.91 \cdot 0.01} = 0.907 \approx 91\%$$

Awesome video: A visual guide to Bayesian thinking

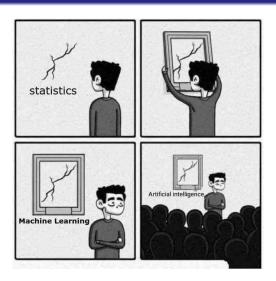


Figure 2: Credits: sandserifcomics

# Random Variable (RV)

Consider a experiment with a sample space  $\Omega$  associated with it. A function that maps each element  $\omega \in \Omega$  to a Real number such that  $[w \leq X]$  it's called random variable (RV)  $(X : \Omega \to \mathbb{R})$ 

• Example: Imagine a experiment that consist of 3 consecutive fair coin tosses, so the sample space of this experiment is:

$$S = \{(H,H,H), (H,H,T), \dots (T,T,T)\}$$
. Now we want to create a random variable X that counts the number of heads in each outcome, so  $X((H,H,H)) = 3$  and  $X((H,H,T)) = 2$ .

# Random Variable:

# Probability Mass Function (PMF):

$$f_X(x) = \mathbb{P}[X = x] = \mathbb{P}[\{\omega \in \Omega : X(\omega) = x\}]$$

# Probability Density Function (PDF)

$$\mathbb{P}[a \le X \le b] = \int_a^b f(x) dx$$

## Cumulative Distribution Function (CDF)

$$F_X(x) = \mathbb{P}[X \le x]$$

#### Expectation:

- Discrete :  $\mathbb{E}[X] = \sum x \mathbb{P}(X = x)$ 
  - Continuous:  $\mathbb{E}[X] = \int_{-\infty}^{\infty} xf(x)dx$

# Variance:

$$\mathbb{V}[X] = \sigma_X^2 = \mathbb{E}[X^2] - \mathbb{E}^2[X]$$

# Sample mean:

$$\overline{X_n} = \frac{1}{n} \sum_{i=1}^n X_i$$

$$s^{2} = \frac{1}{n-1} \sum_{i=1}^{n} (X_{i} - \overline{X})^{2}$$

# Discrete distributions

#### Bernoulli:

Consider a experiment with has two possible outcomes: success (X=1, with probability p) or failure (X=0), this random variable is called Bernoulli, the PMF is:

$$\mathbb{P}(X=k)=p^k(1-p)^{1-k}$$

#### Binomial:

Now consider a Bernoulli experiment conducted n times, let X be the random variable that represents the number of successes, X is called Binomial, the PMF is:

$$\mathbb{P}(X=k) = \binom{n}{k} p^k (1-p)^{n-k}$$

# Discrete distributions

#### Geometric:

Again consider a Bernoulli experiment conducted n times, but the first n-1 are failures and the last nth is a success. Let X be number of tries, which is called Geometric, the PMF is:

$$\mathbb{P}(X=k)=(1-p)^kp$$

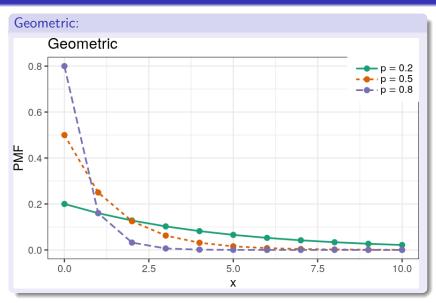
• A important property is that Geometric distribution is the only discrete distribution that is **memoryless**.

#### Poisson:

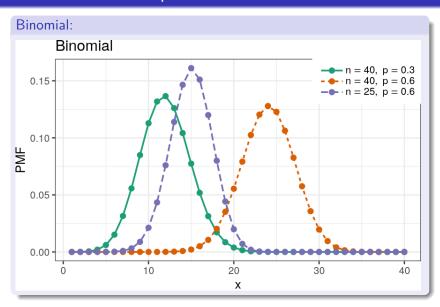
A random variable which value can assume 0,1,2 . . . is called Poisson with  $\lambda > 0$  parameter if your PMF is:

$$\mathbb{P}(X=k) = \frac{e^{-\lambda}\lambda^k}{k!}$$

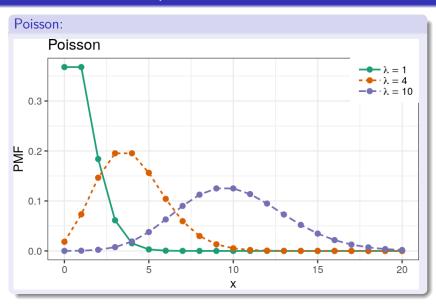
# Discrete distributions plots



# Discrete distributions plots



# Discrete distributions plots



# Continuous distributions

### Normal (or Gaussian, bell curve):

A continuous real random variable is called Normal with  $\sigma^2 > 0$  (squared scale),  $\mu \in \mathbb{R}$  (location) parameters if your PDF is:

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} \exp\left(-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2\right)$$

## Continuous distributions

### Exponential

A continuous **positive** random variable is called Exponential with  $\lambda > 0$  (rate or inverse scale) parameter if your PDF is:

$$f(x) = \lambda e^{-\lambda x}$$

Important property: Exponential and Geometric (discrete) distribution are the only distributions that are **memoryless**.

### Memoryless property:

$$\mathbb{P}[X > x + y \mid X > y] = \mathbb{P}[X > x]$$

So no matter how much time has passed it's like the process is starting from beginning.

### Continuous distributions

#### Pareto

A continuous  $x \in [x_m, \infty)$  random variable is called pareto with  $x_m > 0$  (scale) and  $\alpha > 0$  (shape) parameters if your PDF is:

$$f(x) = \frac{\alpha x_m^{\alpha}}{x^{\alpha+1}}$$

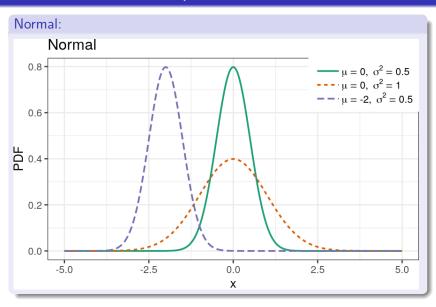
Zipf is the discrete distribution of pareto Pareto is a **heavy tailed** distribution: It means it goes to zero slower (than exponential).

### Pareto principle (80-20 law):

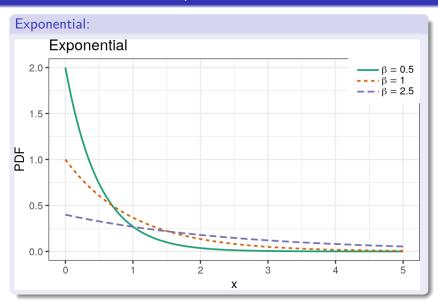
The pareto principle states that 80% of results is caused by 20% of the effects, for example wealth distribution, software bugs and so on

It's a particular pareto distributed values when lpha pprox 1.161

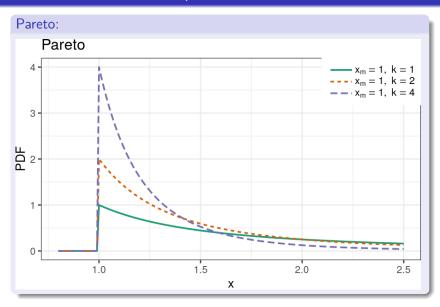
# Continuous distributions plots

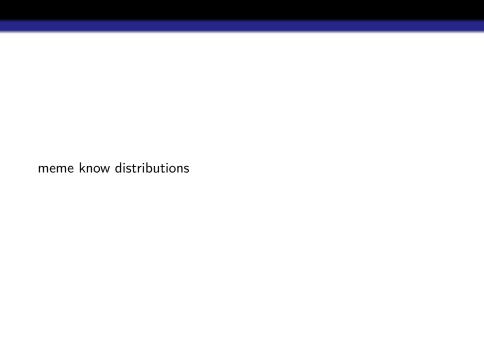


# Continuous distributions plots



# Continuous distributions plots





#### Calculations on the Normal distribution

Given a Normal distributed values, how to calculate the probability on it?

With normal distribution we usually use a standard normal (where  $\mu=0,\sigma=1$ ) cumulative table and standardize the values.

• How to standardize the values: Given  $X \sim N(\mu, \sigma^2)$ 

$$z = \frac{x - \mu}{\sigma}$$
 or  $z = \frac{x - \overline{x}}{s}$ 

z is called z score and is standard normal distributed.

• Standard cumulative  $\Phi(x)$ :  $\Phi(x) = \mathbb{P}(z \le x)$  also  $\Phi(-x) = 1 - \Phi(x)$ 

 $\Phi(x)$  values can we found in a table or using NORMSDIST function in Excel or in Python using stats.norm.cdf function from SciPy.

#### 68-95-99.7 rule:

The 68-95-99.7 rule also know as the empirical rule is a shorthand to remember the percentage of Normal distributed values that lie within arround the mean with a width of 1,2,3 standard deviations.

$$\mathbb{P}(\mu - 1\sigma \le X \le \mu + 1\sigma) \approx 0.6827$$

$$\mathbb{P}(\mu - 2\sigma \le X \le \mu + 2\sigma) \approx 0.9545$$

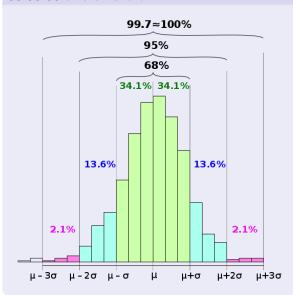
$$\mathbb{P}(\mu - 3\sigma \le X \le \mu + 3\sigma) \approx 0.9973$$

We don't have to memorize this values, we can calculate it:

$$\mathbb{P}(\mu - 1\sigma \le X \le \mu + 1\sigma) = \mathbb{P}(-1\sigma \le X - \mu \le 1\sigma) =$$

$$\mathbb{P}\left(-1 \leq \frac{X - \mu}{\sigma} \leq 1\right) = \mathbb{P}(-1 \leq z \leq 1) = \Phi(1) - \Phi(-1) \approx 0.6827$$

#### 68-95-99.7 rule: Chart



Calculations on the Normal distribution: Example:

tabela e calcular python, excel, normalizar (z score), regra 65,95,99

#### Assumptions on distribution choice

antes se eh discreta ou continua descrever a natureza da normal, exp, pareto Discretas:

Bernouli: resultado dicotonimo, exemplo: moeda, homem ou mulher, sim ou nao. Voto em 20 turno
Binom: quantidade de sucesso dado um numero fixo de experimento independente; Dados 20 dispositivos independentes, depois de

muitas horas, qual a prob de 15 apresentarem defeito.

Geometrica: Numero de falhas ate primeiro sucesso. Exemplo loteria: Dado p ser 1 em 1 milhao qual a prob de ganhar depois de 3 tentativas?

Poisson: Contagem de pessoas inscritas em algum programa que desistem.

Continuas

Normal : Sem restricao de valores (pode ser positivo ou negativo).

Exemplo: Altura de criancas do mesmo sexo e idade (funciona bem pra qui quadrado).

# order statitics

defs, min, max, median, q1,q3, IQR, pq? estat robusta, boxplot

# Convergence

defs, lei dos grandes numeros, teorema do limite central

#### Inference

metodo da maxima verosimilhanca, e grafico qxq (scipy.stats.probplot) (colocar exemplo com distribuicao errada: t student e fitar com a normal)

max veross: find the most likely parameter value, given data. That is, given a prob description of data, find the optimum value for that data (derivatives).

### Further reading:

- Portal action (pt)
- stat cookbook
- havard youtube
- (https://www.youtube.com/playlist?list=PL2SOU6wwxB0uwwH80KTross, barry james, meyer
- khan academy
- http://www.randomservices.org/random/
- divulgacao: pizza de dados, senhora toma cha, andar do bebado, /r/dataisbeautiful