

# Classical statistical inference

Part 1

Associated notebook:

[03-Basic statistics and proba concepts/Basic-statistics 01.ipynb](#)

# Why some statistics ?

- Python for **data** (observation / numerical simulations) manipulation
- Data most often contain a **stochastic** component: observational device, numerical noise, simulation of stochastic process, ...

$\Rightarrow$  **Data**  $\approx$  Random variable (RV)

- Statistics is the tool needed to manipulate **RV**
- *Goals for 2<sup>nd</sup> part of the lecture:*
  - **Uncertainty** calculation (no, this is not black magic)
  - Make **prediction based on data modelling** (first step towards machine learning)

# Definitions and notations recap



$\Omega$

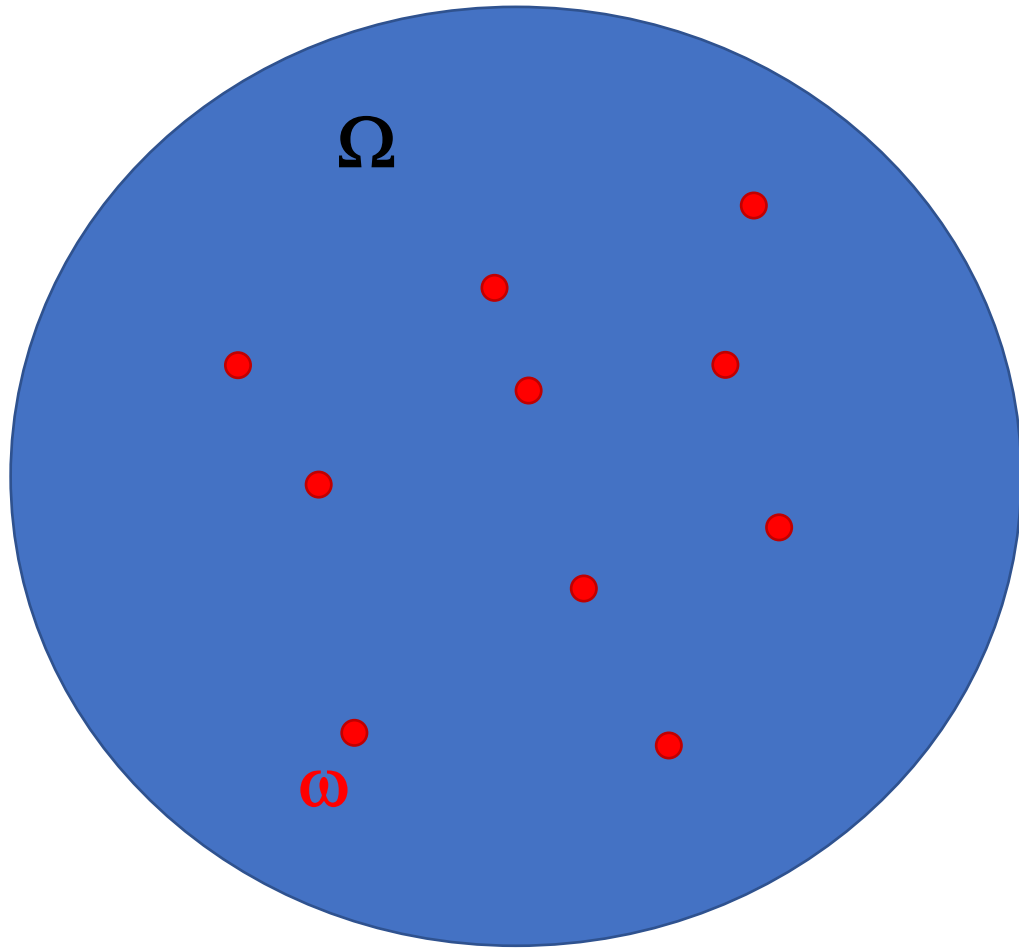
$\Omega$ : Sample space  $\equiv$  all possible outcome of an experiment

e.g. of experiment

- I measure the magnitude of a star (in a binary system, for a transit, ...)
- I count galaxies for different L at a given z
- I obtain the spectrum of a candidate SN
- I measure a GW signal
- ...

This is an abstract space. For the mag of a star,  $\Omega \equiv \mathbb{R}$

# Definitions and notations recap

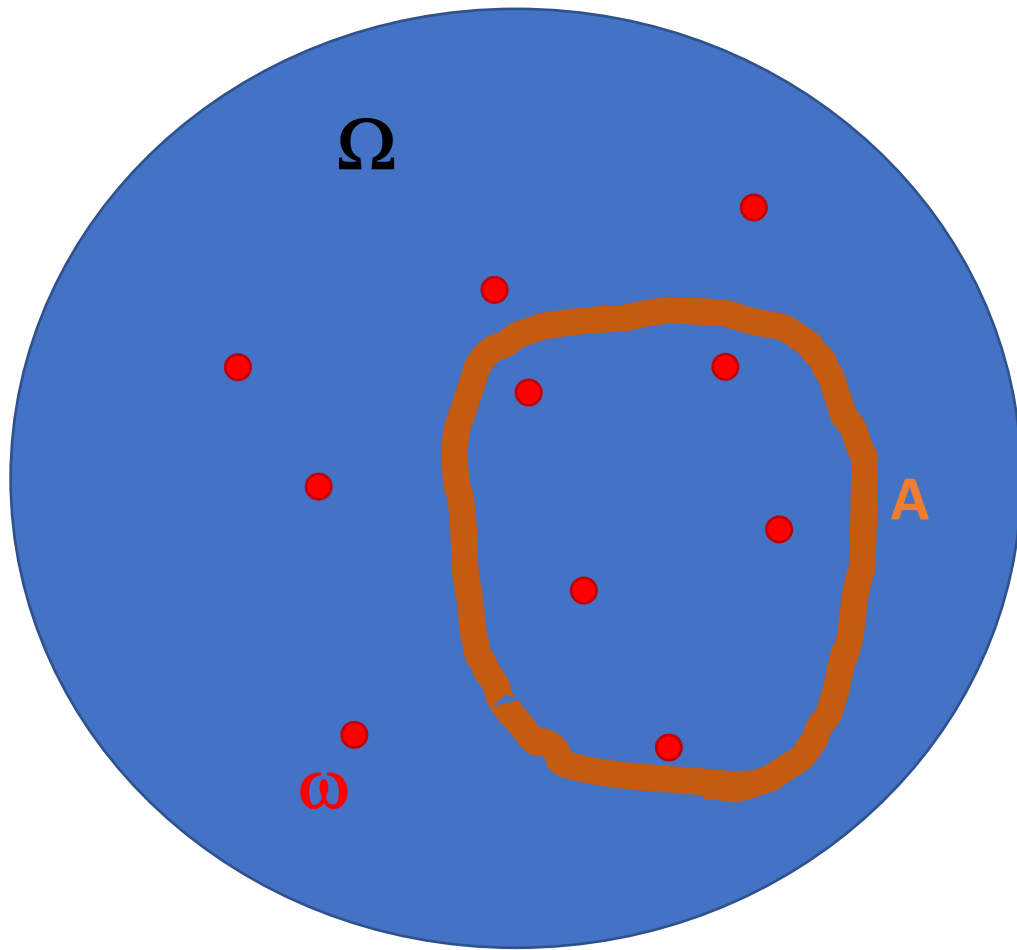


$\Omega$ : Sample space  $\equiv$  all possible outcome of an experiment

$\omega$ : Realisations of the experiment

E.g. There have been 10 measurements of the magnitude of a star.  
Each measurement is a different **realisation**

# Definitions and notations recap



$\Omega$ : Sample space  $\equiv$  all possible outcome of an experiment

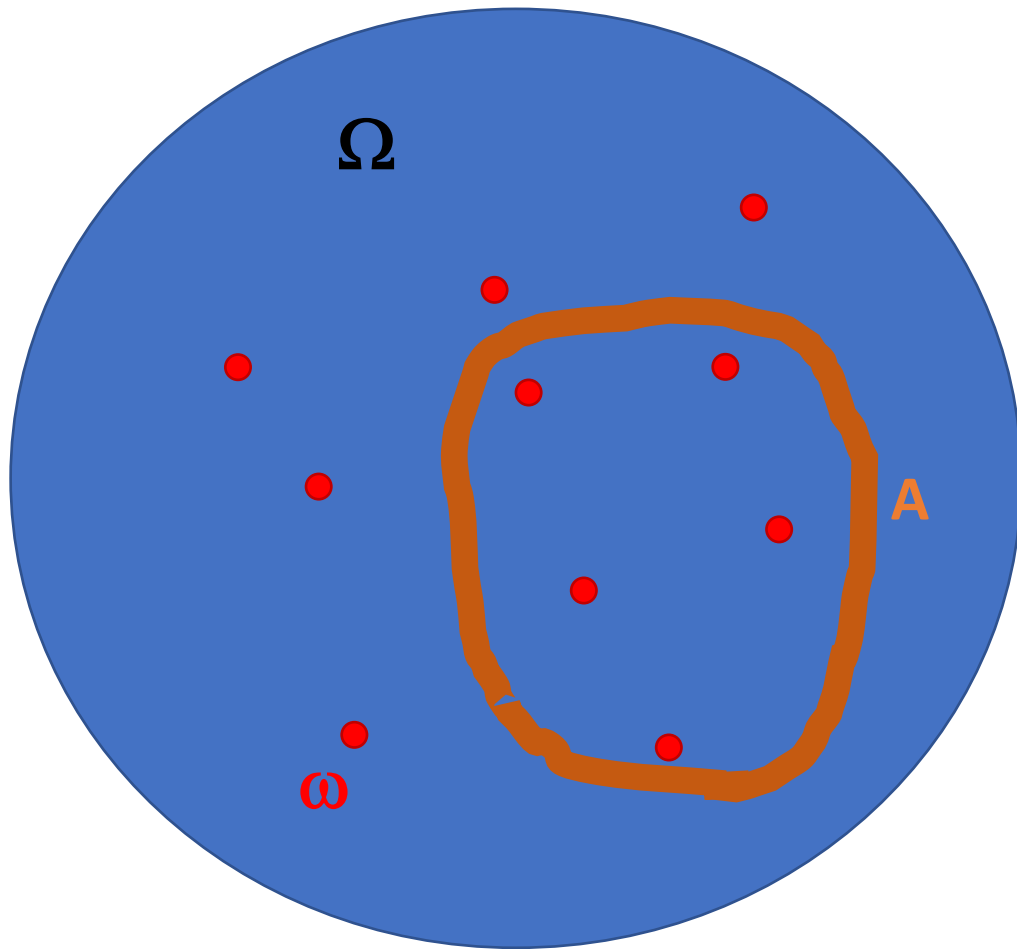
$\omega$ : Realisations of the experiment

$A$ : Event  $\equiv$  a subsample of  $\omega \cong$  Your data set

E.g. You have obtained and are working on 5 measurements of the magnitude of the star.

But an event can be a bit more convoluted quantity, e.g. all measurements you've done that have  $m < 15$  mag

# Definitions and notations recap



$\Omega$ : Sample space  $\equiv$  all possible outcome of an experiment

$\omega$ : Realisations of the experiment

A: Event  $\equiv$  a subsample of  $\omega \cong$  Your data set

$p(A)$ : Probability of an event / value to be in  $[x-dx, x+dx]$

e.g. probability that  $m < 15$  mag

What means  $p(A)$  in frequentist/classical inference ?

Relative frequency of an event  
if experiment is repeated an infinite number of times

# Random variable

A random variable is a variable whose value results from the measurement of a quantity that is subject to random variations

In Python:

`np.random`

- **`np.random.choice(array)`**: choice at random in an array
- **`np.random.seed(value)`**: sets the seed of the rnd generator
- **`np.random.rand(shape)`**: random floats drawn from uniform distribution
- **`np.random.randint(low, high, shape)`**: rnd integers btw low and high

Go to: Sect. I.2. of the notebook

# Conditional probability $p(A \mid B)$

$$p(A \mid B) = \frac{p(A \cap B)}{p(B)} = \text{fraction of times that } A \text{ occurs given that } B \text{ occurred}$$

Reads "Probability of  $A$  given  $B$ "

- The calculation of  $p(A \mid B)$  follows **Bayes** theorem

$$p(A \mid B) = \frac{p(B \mid A) p(A)}{p(B)}$$

- The probability to have a flu given that you have fever is different from the probability to have fever given that you have a flu

$$p(A \mid B) \neq p(B \mid A)$$



# Bayes theorem

$$p(A | B) = \frac{p(B | A) p(A)}{p(B)}$$

## Question:

**A:** rare disease that affects 0.1 % of the population.

**B:** test that is efficient at 99 % (i.e. 1 % False positive rate).

If you have a positive test (B), what is the probability for you to be affected by this disease (A) ?

NB: Efficiency is NOT sensitivity (sensitivity generally means fraction of true positive).

# Bayes theorem

$$p(A | B) = \frac{p(B | A) p(A)}{p(B)}$$

## Question

A: **rare disease** that affects **0.1 %** of the population.

B: **test** that is efficient at **99 %** (i.e. **1 % False positive** rate).

If you have a positive test (B), what is the probability for you to be affected by this disease (A) ?

**Solution:** (See Sect. I.3. of the notebook )

Among 1000 persons, 1 has the disease (it touches **0.1 %** of the population =  $p(A)$  ).

The test has 99% efficiency ( $=p(B | A)$  ). Which means that 1% of the people will be tested positive while not being sick.

=> 10 people will be positive while healthy. You should also have  $\approx 1$  being positive while being effectively sick.  $p(B)=0.01 + 0.001 = 0.011$

=>  $p(\text{disease} | +) \approx 1/11 = 9 \%$

**BEWARE**

RARE events common in astronomy  
Conditional probabilities are often implicit

# Probability density / mass function

Coin Toss (Bernoulli PMF): The PDF is the **normalised** histogram we had obtained

$$\text{Ber}(k | p) = p^k (1 - p)^{1-k}$$

$k$  in  $\{0, 1\} \equiv \{\text{failure, success}\}$

parameter (success rate)

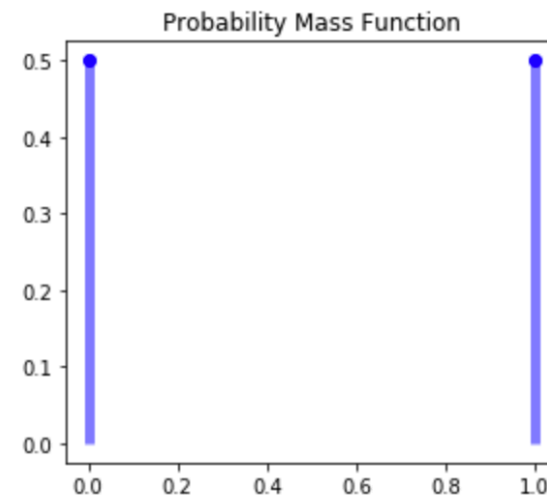
Uniform PDF:

$$\begin{aligned} h(x) &= \frac{1}{b-a} \text{ if } a \leq x \leq b \\ h(x) &= 0 \text{ otherwise} \end{aligned}$$

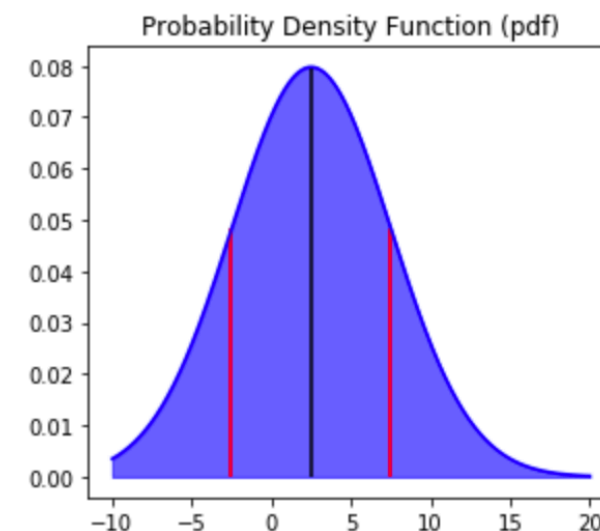
Gaussian PDF:

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

PDF In Python: **go to** Sect. 1.4 of the notebook



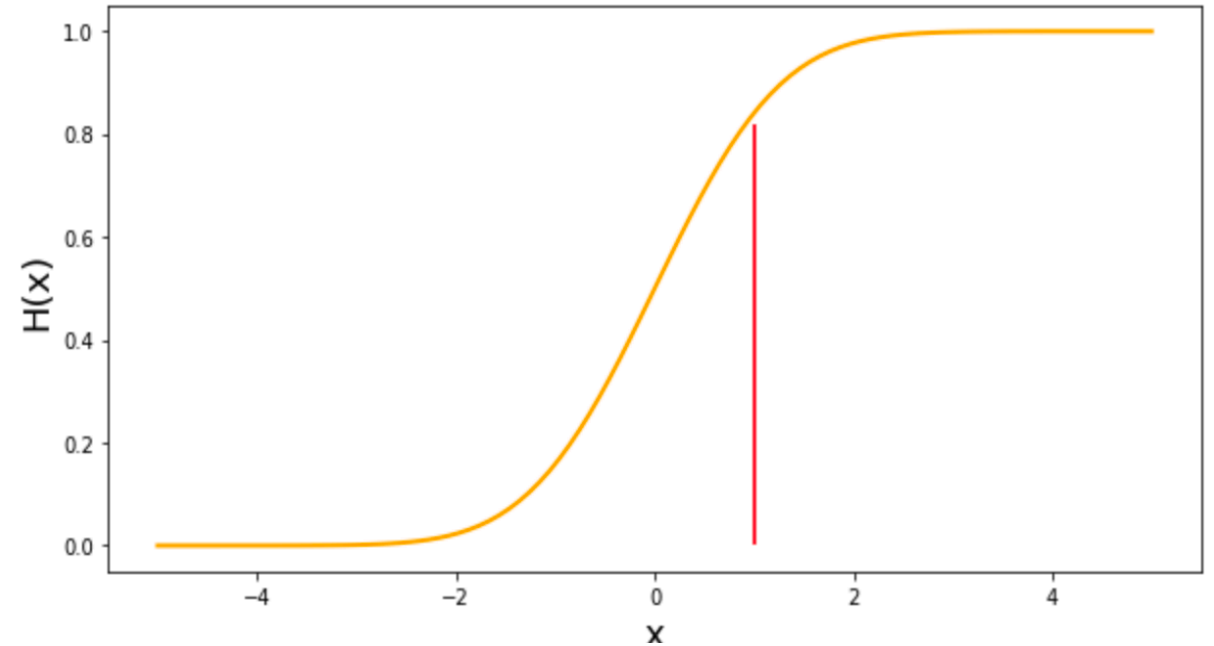
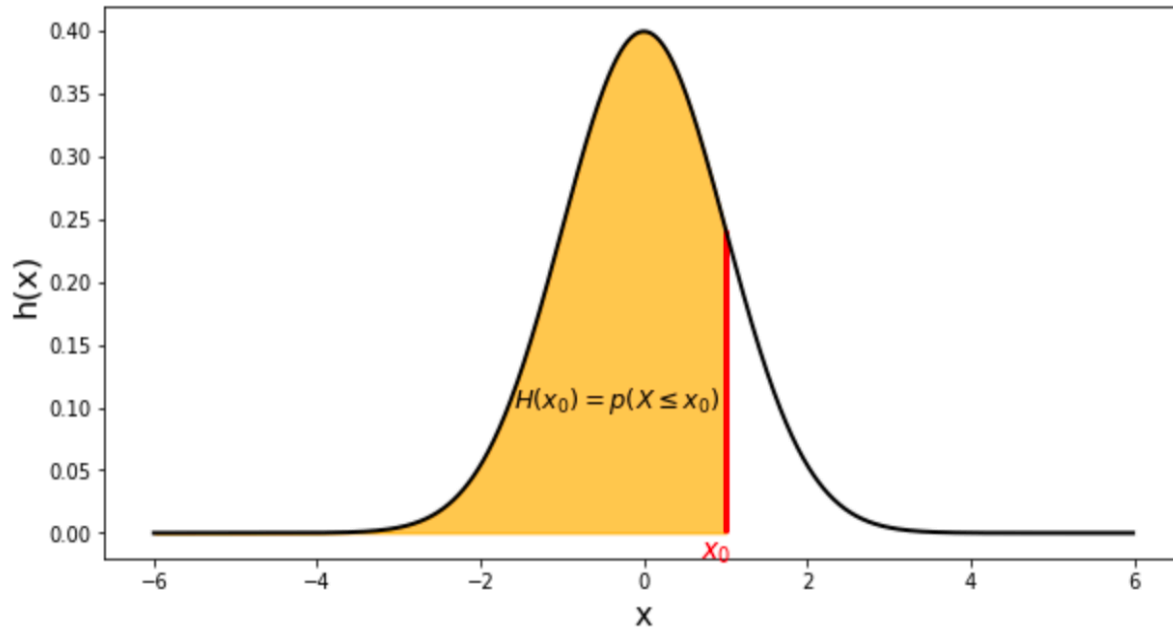
$$\int p(x) dx = 1$$



# Cumulative density function

This is the **integral** of the PDF:

$$p(X \leq x) = H(x) = \int_{-\infty}^x h(x') dx'$$



**CDF In Python:** **go to** Sect. I.5 of the notebook

$$H(x | \mu, \sigma) = \frac{1}{\sqrt{2\pi} \sigma} \int_{-\infty}^x \exp\left(-\frac{1}{2} \frac{(x' - \mu)^2}{\sigma^2}\right) dx'.$$

# Probability enclosed between 1-2-3 $\sigma$ for $N(\mu, \sigma)$

See the last exercise of Sect. I.5 of the notebook

