



PROBABILITIES II
CONTINUOUS PROBABILITY DISTRIBUTIONS
(FR: LOIS DE PROBABILITE CONTINUES)

1. Definitions

Position of the problem :

Up to now, we have only considered probability distributions on finite sample spaces. What if it's not the case anymore ?

Eg : You choose randomly a number between 0.2 and 0.6. What is the probability that it is in-between 0.3 and 0.35 ?

Definition (probability distribution on a bounded interval) :

Let $I = [a; b]$ be a bounded interval and f a continuous and positive function on I .

A probability distribution P on I has for density f if :

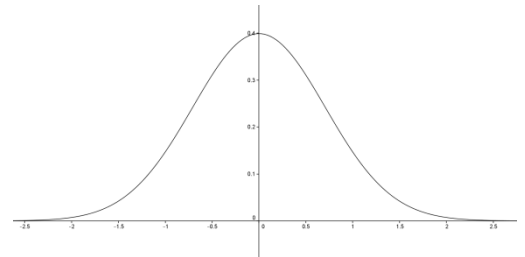
$$P(I) = \int_a^b f(x) dx = 1$$

$$P([c; d]) = \int_c^d f(x) dx$$

Notes :

1. f is often denoted p.d.f (probability density function) in English (fr: fonction de densité)
2. **Graphical Interpretation :**

$P([c; d])$ represents the area below the graph of the density function between c and d and $P(I)$ (area below the graph of the density function between a and b) is taken as the unit area.



Note : make the link between $P(I) = \int_a^b f(x) dx = 1$ and $\sum P_i x_i = 1$ for discrete probability distributions

3. If a random variable X follows a distribution P , we can write $P(c \leq X \leq d) = \int_c^d f(x) dx$

This definition can be widened to a non-bounded interval :

Definition (probability distribution on an interval in the form $I = [a; +\infty[$) :

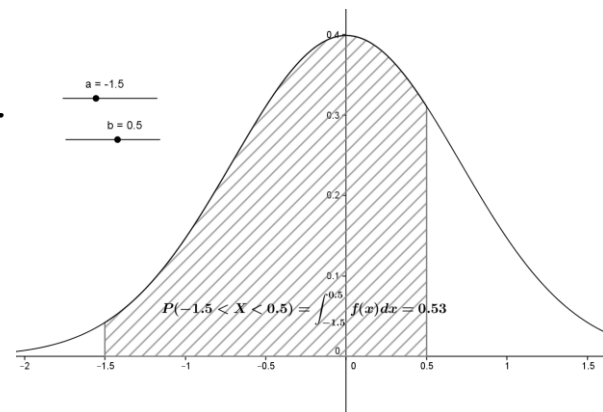
Let I be an interval in the form $I = [a; +\infty[$ and f continuous and positive function on I .

A probability distribution P on I has for density f if :

- $P(I) = \lim_{t \rightarrow +\infty} \left(\int_a^t f(x) dx \right) = 1$

- $P([c; d]) = \int_c^d f(x) dx$

- $P([c; +\infty[) = 1 - \int_a^c f(x) dx$



Note :

- $P(\{\alpha\}) = 0$ for any α in I . (since $P(\{\alpha\}) = \int_{\alpha}^{\alpha} f(x) dx = 0$)
- we hence have $P([c; d]) = P([c; d[) = P(]c; d]) = P(]c; d[)$

Properties :

- **The expected value of a continuous random variable X is :** $E(X) = \int_a^b xf(x) dx$
- **If K and J are two intervals contained in I such that $P(J) \neq 0$. We can define the conditional probability $P_J(K) = \frac{P(J \cap K)}{P(J)}$ of K given J .**
- **If K and J are a partition of I , then $P(J \cup K) = P(J) + P(K) = 1$.**

Notes :

1. Be careful using $P(J \cup K) = P(J) + P(K) - P(J \cap K)$, $P(J \cap K)$ can be zero (if $J \cap K$ is finite) without J and K being disjoint.
2. We sometime use the cumulative distribution function (CDF) (fr: fonction de répartition) defined as :

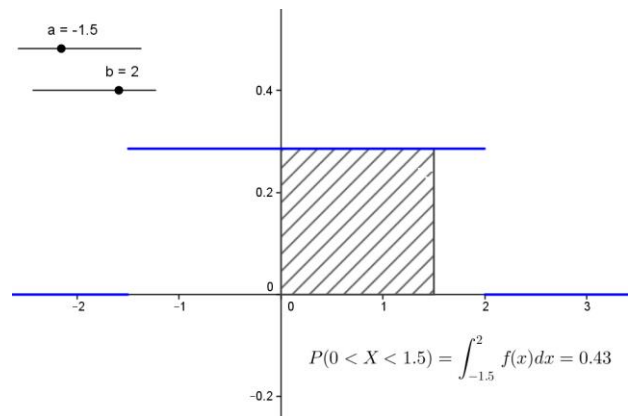
$$F(x) = P(X \leq x) = \int_{-\infty}^x f(x) dx$$

2. The uniform (or rectangular) distribution on $[a ; b]$

Definition :

The uniform distribution on $[a; b]$, denoted $\mathcal{U}_{[a; b]}$ is defined by having for density function

f the function equal to :
$$\begin{cases} \frac{1}{b-a} & \text{if } a \leq x \leq b \\ 0 & \text{elsewhere} \end{cases}$$



Note : - It is the one involved when choosing randomly a number between a and b .

- For any c and d in $[a; b]$, $P(X \in [c; d]) = \int_c^d \frac{1}{b-a} dx = \frac{d-c}{b-a}$

Examples :

- Calculate $P_{[0.2; 0.6]}([0.5; 0.55])$
- You choose randomly a number between 0 and 1 . Given it is between 0.6 and 0.7 , what is the probability for it to be bigger than 0.68 ?

Property:

The expected value of a continuous random variable $X \sim \mathcal{U}_{[a; b]}$ is : $\frac{a+b}{2}$

Proof : $E(X) = \int_a^b \frac{x}{b-a} dx = \dots$

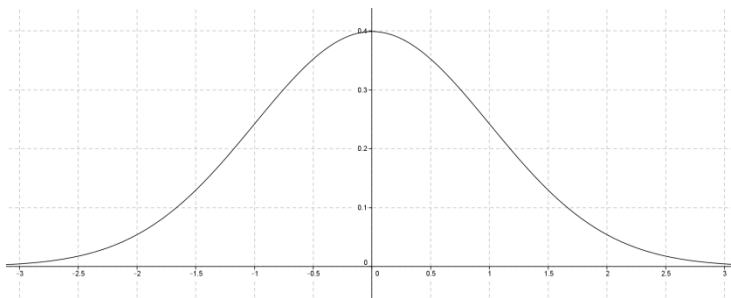
3. The standard normal (or Gaussian) distribution (fr: loi normale centrée réduite)

Definition :

A random variable X follows a standard normal distribution when its density function is defined on \mathbb{R} by :

$$f(x) = \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}}$$

We write $X \sim \mathcal{N}(0;1)$



Notes :

- This graph is known as the **Gaussian curve** or the bell-shaped graph.

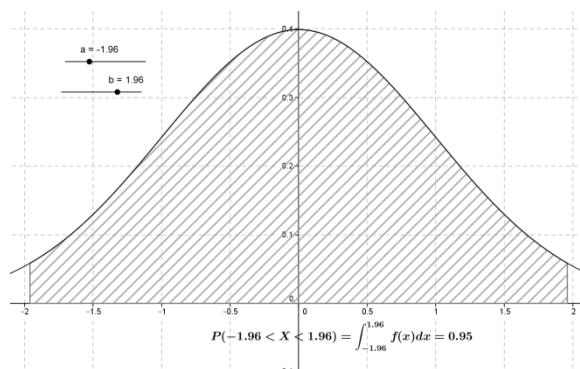
Properties:

If $X \sim \mathcal{N}(0;1)$ then :

$$P(-1,96 \leq X \leq 1,96) \approx 0.95$$

The expected value of X is $\mu = 0$

The standard deviation of X is $\sigma = 1$



4. The normal (or Gaussian) distribution (fr: loi normale)

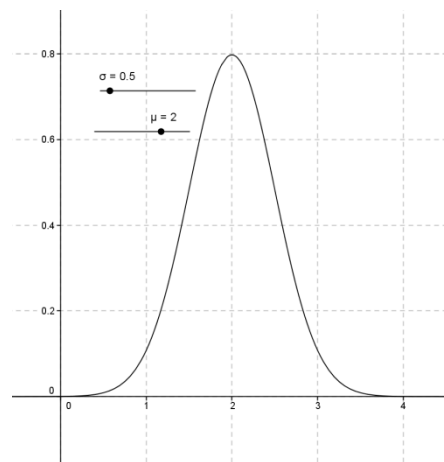
Definition :

A random variable X follows a normal distribution with expected value μ and standard deviation σ when the random variable $\frac{X - \mu}{\sigma}$ follows a standard normal distribution.

We write $X \sim \mathcal{N}(\mu; \sigma^2)$

Notes :

- The density function is then : $f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$
- μ picks out the peak of the graph and σ its spread (small values lead to tall and narrow graphs, larger values give short, fat graphs).



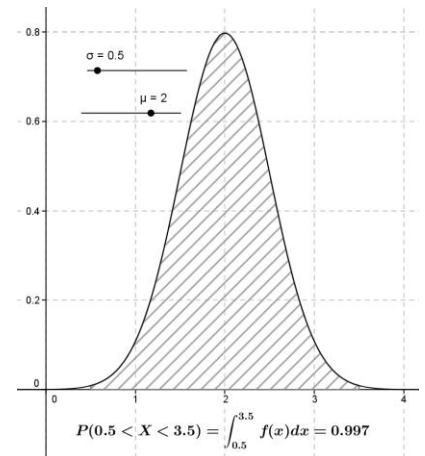
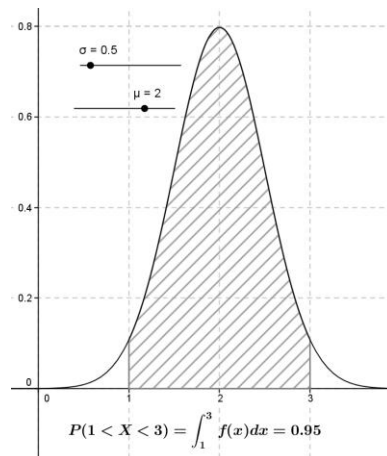
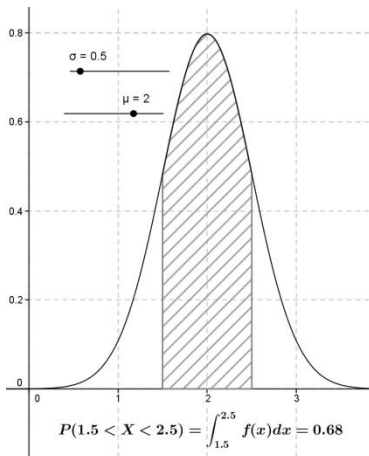
Properties:

If $X \sim \mathcal{N}(\mu, \sigma^2)$ then :

$$P(\mu - \sigma \leq X \leq \mu + \sigma) \approx 0.68$$

$$P(\mu - 2\sigma \leq X \leq \mu + 2\sigma) \approx 0.95$$

$$P(\mu - 3\sigma \leq X \leq \mu + 3\sigma) \approx 0.99$$



Note :

1. Since we can't calculate directly (we don't know a primitive) , the calculator is very helpful to calculate approximate values of $P(a \leq X \leq b)$ with the function **normalFrép** .



Ex : if $X \sim \mathcal{N}(100; 25)$, then $P(70 \leq X \leq 95) = \text{normalFrép}(70, 95, 100, 5)$

2. With a spreadsheet : $P(X \leq a) = \text{LOI.NORMALE.N}(a; \mu ; \sigma ; \text{VRAI})$ and $\text{LOI.NORMALE.INVERSE.N}(k; \mu ; \sigma)$ gives the number a such that $P(X \leq a) = k$.

Bullet points of the chapter

- ✓ Dealing with the uniform and normal probability distributions
- ✓ Knowing the graphs of their density functions
- ✓ Knowing that If $X \sim \mathcal{N}(0; 1)$ then $P(-1.96 \leq X \leq 1.96) \approx 0.95$
- ✓ Knowing that If $X \sim \mathcal{N}(\mu; \sigma^2)$ then $P(\mu - \sigma \leq X \leq \mu + \sigma) \approx 0.68$
- ✓ Knowing that If $X \sim \mathcal{N}(\mu; \sigma^2)$ then $P(\mu - 2\sigma \leq X \leq \mu + 2\sigma) \approx 0.95$
- ✓ Knowing that If $X \sim \mathcal{N}(\mu; \sigma^2)$ then $P(\mu - 3\sigma \leq X \leq \mu + 3\sigma) \approx 0.99$
- ✓ Using a calculator or a spreadsheet to get $P(a \leq X \leq b)$