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II The discrete r.v.

Def: let X be a discrete r.v.

The probability function

that characterizes X is

given by: - $X(\Omega) = \{ \text{all the possible issues} \}$

- $\forall i \in X(\Omega), P(X=i)$

Properties :

$$\sum_{i \in X(\Omega)} P(X=i) = 1$$

$$\forall i \in X(\Omega), P(X=i) \geq 0$$

When we consider a r.v:

- discrete
- continuous

mean = expectation

Expectation and Variance

def: expectation - (probability) mean

If $\sum_{i \in X(\Omega)} ixP(X=i)$ exists

we say that X has an expectation

$$\Rightarrow E[X] = \sum_{i \in X(\Omega)} ixP(X=i)$$

Example

let X be the issue when you throw a dice

$$- X(\Omega) = \{1, 2, 3, 4, 5, 6\} = [1, 6]$$

$$- \forall i \in X(\Omega), P(X=i) = \frac{1}{6}$$

Assumption:

x_1 is associated to X_1

\vdots
 x_n is associated to X_n

X_1, \dots, X_n are independant and identically distributed (iid)

Example:

$$\begin{aligned} E[X] &= \sum_{i=1}^6 i \times \frac{1}{6} \\ &= \frac{1}{6} \times \frac{6 \times 7}{2} = 3.5 \end{aligned}$$

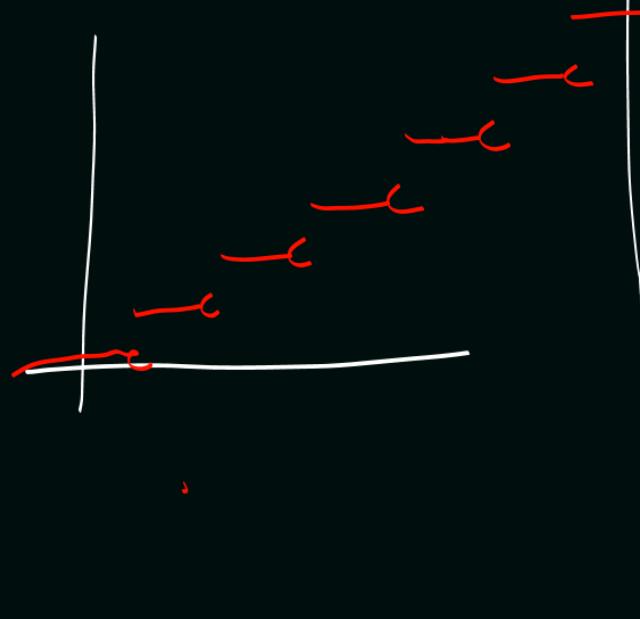
Chapter 1 : Introduction to probability

Notations :

X : random variable (r.v.)

x_1, \dots, x_n : n observations of X

Example

$$F_X(t) = \begin{cases} 0 & \text{if } t < 1 \\ \frac{1}{6} & \text{if } t \in [1; 2[\\ \frac{2}{6} & \text{if } t \in [2; 3[\\ \frac{3}{6} & \text{if } t \in [3; 4[\\ \frac{5}{6} & \text{if } t \in [4; 5[\\ \frac{5}{6} & \text{if } t \in [5; 6[\\ 1 & \text{if } t \geq 6 \end{cases}$$


if $t < 1$
if $t \in [1; 2[$
if $t \in [2; 3[$
if $t \in [3; 4[$
if $t \in [4; 5[$
if $t \in [5; 6[$
if $t \geq 6$

III

Common discrete r.v

- uniform
- bernoulli
- binomial
- geometric
- Poisson

def: Variance

If $\sum_{i \in X(\Omega)} i^2 P(X=i)$ exists

then the variance of X exists and

$$V[X] = \sum_{i \in X(\Omega)} (i - E[X])^2 P(X=i)$$

distribution function:

Let X be a discrete r.v:

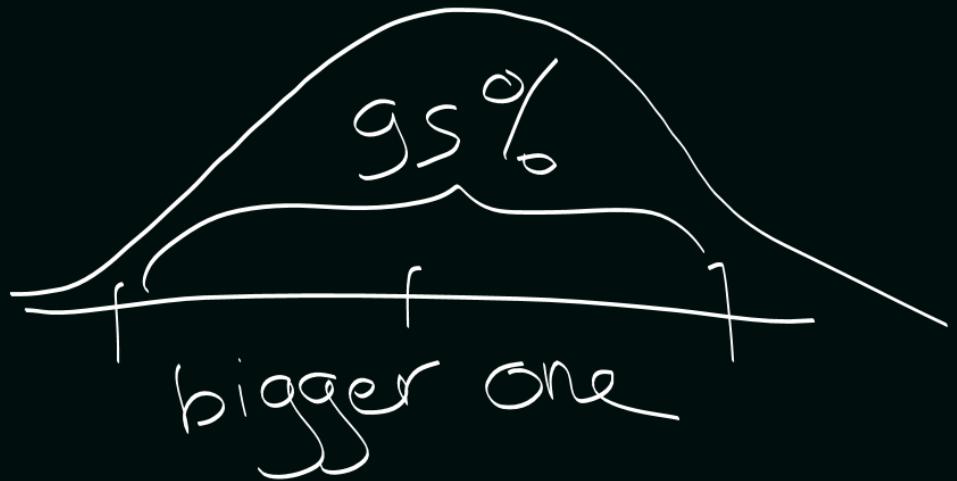
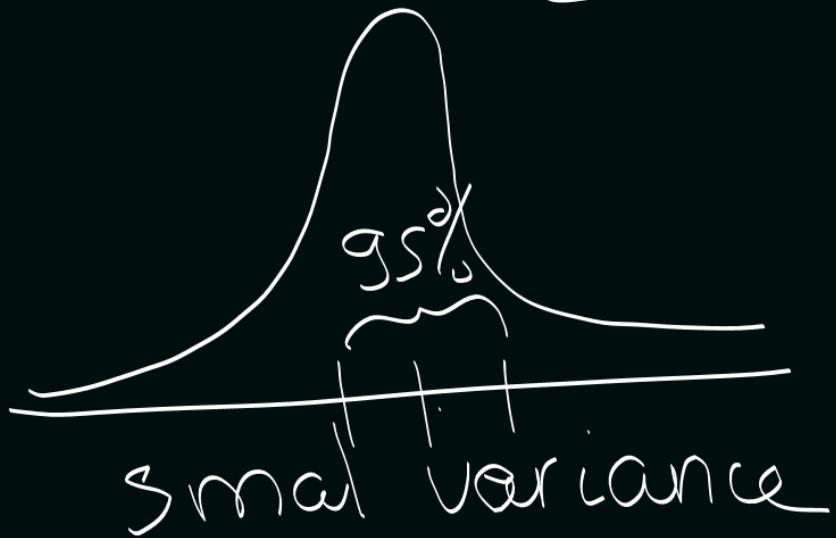
$$\forall t \in \mathbb{R} \quad F_X(t) = P(X \leq t)$$
$$= \sum_{\substack{i \in X(\mathbb{R}) \\ i \leq t}} P(X = i)$$

sometimes

$$F_X(t) = P(X < t)$$

Rq: In general

$$V[X] = E[(X - E[X])^2]$$



Ex :

let F be the function

$$F(t) = \begin{cases} 0 & \text{if } t < 3 \\ \frac{1}{2} & \text{if } t \in [3, 5[\\ \frac{3}{4} & \text{if } t \in [5, 6[\\ 1 & \text{if } t \in [6, +\infty[\end{cases}$$

Let X be the associated r.v.

Then $X(\Omega) = \{3, 5, 6\}$

$$P(X=3) = \frac{1}{2}$$

$$P(X=5) = \frac{1}{4} = \left(\frac{3}{4} - \frac{1}{2}\right)$$

$$P(X=6) = \frac{1}{4} = \left(1 - \frac{3}{4}\right)$$

Properties:

- constant piecewise function
- increasing function
- $\lim_{x \rightarrow -\infty} F_X(x) = 0$
- $\lim_{x \rightarrow +\infty} F_X(x) = 1$

Rk: $V[X] = E[X^2] - E^2[X]$

Property: $V[X] \geq 0$

standard deviation: $\sigma(X) = \sqrt{V[X]}$

c) Binomial

$$X \sim \text{Bin}(n; p)$$

$$n \in \mathbb{N}^*$$

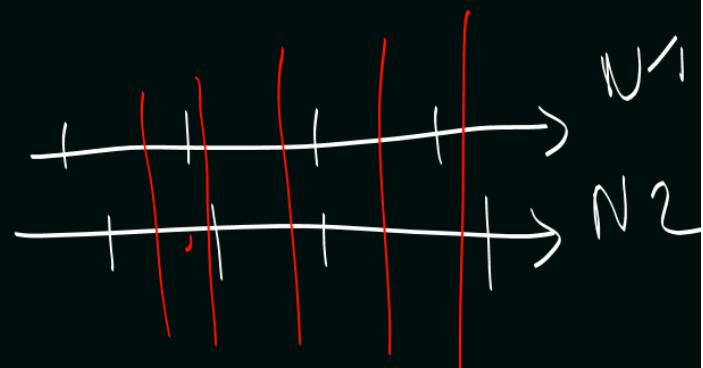
$$p \in [0; 1]$$

$$- X(\Omega) = \{0; 1; \dots; n\}$$

$$\begin{aligned} - \forall k \in X(\Omega), P(X=k) &= C_n^k p^k (1-p)^{n-k} \\ &= \binom{n}{k} p^k (1-p)^{n-k} \end{aligned}$$

Example:

You throw a coin n times
and I count the number
of time I get P
 $X \sim \mathcal{B}(n; 0.5)$



Simulation



a) Uniform

let $X \sim U(\{1, \dots, K\})$ K an integer

- $X(\omega) = \{1, \dots, K\}$

- $\forall i \in X(\omega)$, $P(X=i) = \frac{1}{K} = \frac{1}{\#X(\omega)}$

In practice :

$$X \sim \mathcal{B}(n; P)$$

$$X = Y_1 + Y_2 + \dots + Y_n$$

$$\text{with } Y_i \sim \mathcal{B}(P)$$

and Y_1, Y_2, \dots, Y_n are independent

$$\binom{n}{k} = \frac{n!}{k!(n-k)!}$$

: n^b of combinations
of k elements
among n
(without order)

Geometric

$$X \sim G(P) \quad P \in [0; 1]$$

$$X(\omega) = [1, +\infty[= \{1, 2, \dots, +\infty\}_{k=1}^{\infty}$$

$$\forall k \in X(\omega), P(X=k) = P((1-P)^{k-1}P)$$

a real interval $[1, 3]$
an integer interval $[1, 3] = \{1, 2, 3\}$

b) Bernoulli

$$X \sim \mathcal{B}(P) \quad P \in [0; 1]$$

$$\begin{cases} X(\omega) = \{0, 1\} \\ P(X=1) = P \\ P(X=0) = 1 - P = q \end{cases}$$

(The important thing is that we have 2 issues)

$$\frac{1}{n} \sum_{k=1}^n k^2 = \frac{(n+1)(2n+1)}{6}$$

$$1^2 + 2^2 + \dots + n^2$$

$$2) X \sim \mathcal{B}(p)$$

$$\begin{aligned} E[X] &= 0 \times P(X=0) + 1 \times P(X=1) \\ &= 1 \times P(X=1) \end{aligned}$$

$$E[X] = p$$

$$\begin{aligned} V[X] &= (0^2 \times P(X=0) + 1^2 \times P(X=1)) - p^2 \\ V[X] &= p \times (1-p) = pq \end{aligned}$$

$$V[x] = \frac{(k+1)(2k+1)}{6} - \frac{(k+1)^2}{4}$$

$$V[x] = \frac{k^2 - 1}{12}$$

$$\begin{aligned}
 \text{Var}[X] &= E[X^2] - E[X]^2 \\
 E[X^2] &= \sum_{i=1}^K i^2 \times P(X=i) \\
 &= \frac{1}{K} \sum_{i=1}^K i^2 = \frac{1}{K} \times \frac{K(K+1)(2K+1)}{6}
 \end{aligned}$$

Rk:

$$+ \sum_{k=1}^n R_k = \frac{n(n+1)}{2}$$

idea:

$$\begin{array}{ccccccc} & | & 2 & 3 & \dots & n \\ & n & n-1 & n-2 & & & 1 \\ \hline & (n+1) & (n+1) & (n+1) & & & (n+1) \end{array}$$

$$)\quad X \sim U(\{1, \dots, K\})$$

$$\mathbb{E}[X] = \sum_{i=1}^K i \times P(X=i)$$

$$= \sum_{i=1}^{K+1} \frac{1}{K} \times i = \frac{1}{K} \sum_{i=1}^K i + \frac{1}{K} \times (K+1) = \frac{1}{K} \frac{K(K+1)}{2}$$

$$\Rightarrow \boxed{\mathbb{E}[X] = \frac{K+1}{2}}$$

Rk:

let X and Y two r.v

let F_X the distribution fct for X

$$F_Y$$

—————
 Y

If $\forall t \in \mathbb{R}, F_X(t) = F_Y(t)$

then X and Y have the same law

$X \sim \text{Exp}(P)$: the first time of
success when we
perform bernoulli
experimentations.

? Can you compute the expectation
and the variance of X when :

$$1) X \sim U\{1, \dots, K\}$$

$$2) X \sim \mathcal{B}(p)$$

$$3) X \sim \mathcal{B}(n; p)$$

$$\begin{aligned}V[X] &= V[Y_1 + \dots + Y_n] \\&= V[Y_1] + \dots + V[Y_n]\end{aligned}$$

the variables
are
independ-
ent

$$V[X] = n \times p \times q$$

Property:

$$X \sim \mathcal{G}(p)$$

$$E[X] = 1/p$$

$$V[X] = \frac{1-p}{p^2}$$

$$\underline{\text{Berechnung:}} \quad E[X] = \sum \binom{n}{k} p^k (1-p)^{n-k}$$

$$= \sum_{k=0}^n \frac{n!}{k!(n-k)!} \times k \times p^k (1-p)^{n-k}$$

$$(a+b)^n = \sum_{k=0}^n \binom{n}{k} a^k b^{n-k}$$

$$3) X \sim \mathcal{B}(n; p)$$

$$X = Y_1 + Y_2 + \dots + Y_n \quad \text{with } Y_i \sim \mathcal{B}(p)$$

$$\begin{aligned} E[X] &= E[Y_1 + \dots + Y_n] \\ &= E[Y_1] + E[Y_2] + \dots + E[Y_n] \end{aligned}$$

Y_1, \dots, Y_n : independant

$$\boxed{E[X] = nP}$$

Poisson :

$$X \sim P(\lambda) \quad \lambda > 0$$

$$\begin{cases} X(2) = N \\ \forall k \in \mathbb{N}, P(X=k) = \frac{\lambda^k e^{-\lambda}}{k!} \end{cases}$$

Rk:

Let X and Y two discrete r.v.

$$\begin{aligned} \forall i \in X(\mathcal{S}) \quad & P(X=i, Y=j) \\ \forall j \in Y(\mathcal{S}) \quad & = P(X=i) \times P(Y=j) \end{aligned}$$

Properties:

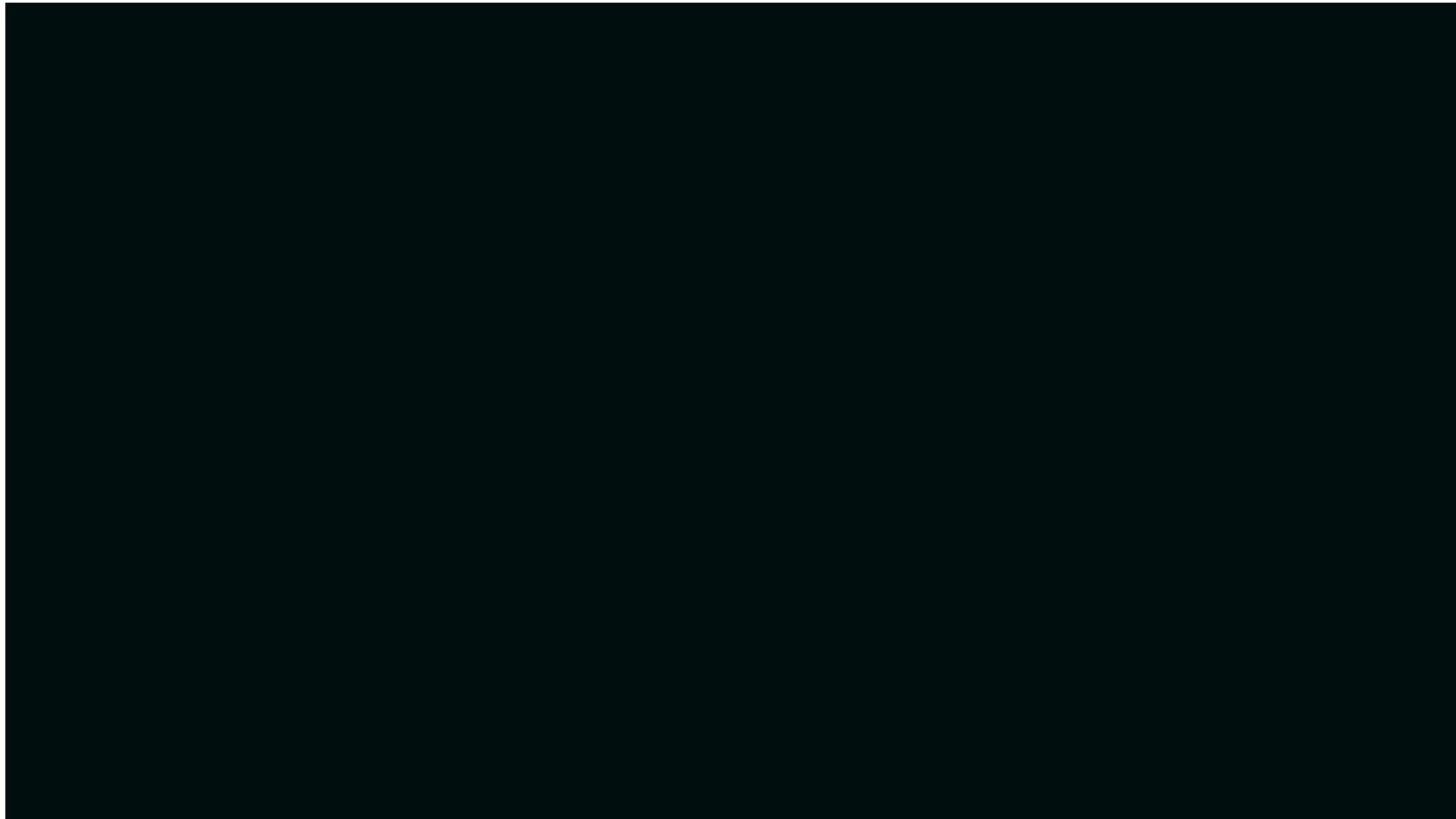
Let X and Y be two r.v

let $\lambda \in \mathbb{R}$.

$$\times E[X + \lambda Y] = E[X] + \lambda E[Y]$$

$$\times V[\lambda X] = \lambda^2 V[X]$$

$$\times \text{ if } X \text{ and } Y \text{ are independant } V[X + Y] = V[X] + V[Y]$$



$$\underline{RB} : \mathcal{B}(P) = \mathcal{B}(1, P)$$

Google : CRAN

Comprehensive R Archive
Network

IV Continuous r.v.

Rk: Let X be a continuous r.v.

$$P(X=a) = 0$$

→ density function

$$\begin{cases} f: \mathbb{R} \rightarrow \mathbb{R} \\ - \forall x \in \mathbb{R}, f(x) \geq 0 \\ - \int_{-\infty}^{+\infty} f(x) dx = 1 \end{cases}$$

Distribution Function:

Let X be a continuous r.v.

$$\begin{aligned}F_X(t) &= P(X \leq t) = P(X < t) \\&= \int_{-\infty}^t f(x) dx\end{aligned}$$

Let $a < b$

$$P(X \in [a, b]) = \int_a^b f(t) dt$$

where f is the density function
associated to X .

$$\begin{aligned} \text{R.R.: } P(X \in [a, b]) &= P(X \in [a, b[) \\ &= P(X \in]a, b]) \\ &= P(X \in]a, b[) \end{aligned}$$

Expectation and Variance:

let X be a continuous r.v.

If $\int_{-\infty}^{+\infty} |x| f(x) dx$ exists

then the expectation of X exists and

$$E[X] = \int_{-\infty}^{+\infty} x f(x) dx$$

If $\int_{-\infty}^{+\infty} x f(x) dx$ exists

then the variance of X exists and

$$\begin{aligned} V(X) &= \int_{-\infty}^{+\infty} (x - E[X])^2 f(x) dx \\ &= \int_{-\infty}^{+\infty} x^2 f(x) dx - E[X]^2 \end{aligned}$$

Properties

$$X \sim P(\lambda)$$

$$E[X] = \lambda$$

$$\sqrt{V[X]} = \sqrt{\lambda}$$

Properties:

The distribution f^d :

- is a increasing f^d
- $\lim_{x \rightarrow -\infty} F_X(x) = 0$, $\lim_{x \rightarrow +\infty} F_X(x) = 1$
- $x \rightarrow -\infty$ is a continuous f^d

Bk: Let X and Y be 2 continuous

f.v
 X and Y are independant if
 $f_{(X,Y)}(x,y) = f_X(x) \times f_Y(y)$

For Friday:

Proof of $E[X]$

for $X \sim E[A]$

Exponential

$X \sim \mathcal{E}(\lambda) \lambda > 0$

density $f(x) = \lambda e^{-\lambda x}$ if $x \geq 0$

$$f(x) = \begin{cases} \lambda e^{-\lambda x} & \text{if } x \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

Proof:

$$\begin{aligned}
 E[X] &= \int_a^b \frac{1}{b-a} \times t \, dt \\
 &= \frac{1}{b-a} \int_a^b t \, dt \\
 &= \frac{1}{b-a} \times \left[\frac{t^2}{2} \right]_a^b = \frac{b^2 - a^2}{(b-a) \times 2} \\
 &= \frac{b+a}{2}
 \end{aligned}$$

Gaussian

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

expectation

variance

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

Properties:

Let $X \sim U(a, b)$

$$- E[X] = \frac{a+b}{2}$$

$$- V[X] = \frac{(b-a)^2}{12}$$

$$- F_X(t) = \begin{cases} 0 & \text{if } t < a \\ \frac{t-a}{b-a} & \text{if } t \in [a, b] \\ 1 & \text{if } t > b \end{cases}$$

IV

Famous r.v

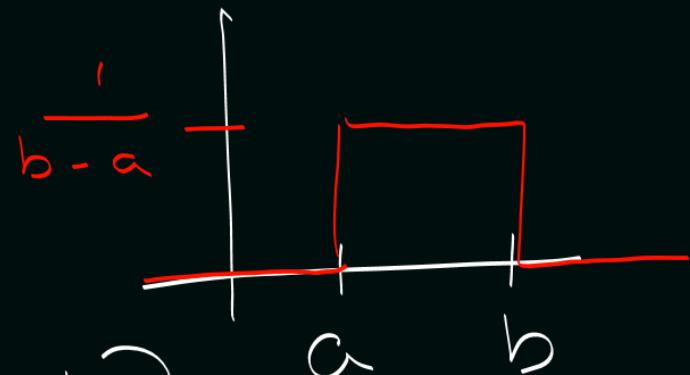
- uniform
- exponential
- gaussian = normal

a) Uniform

$$X \sim U([a, b])$$

density fct:

$$f(x) = \begin{cases} \frac{1}{b-a} & \text{on } [a, b] \\ 0 & \text{otherwise} \end{cases}$$



Properties:

$$E[X] = \frac{1}{\lambda}$$

$$\sqrt{[X]} = \frac{1}{\sqrt{\lambda}}$$

Rk:

On the table, we get

$F_X(u)$ for $u \geq 0$

and $X \sim N(0, 1)$

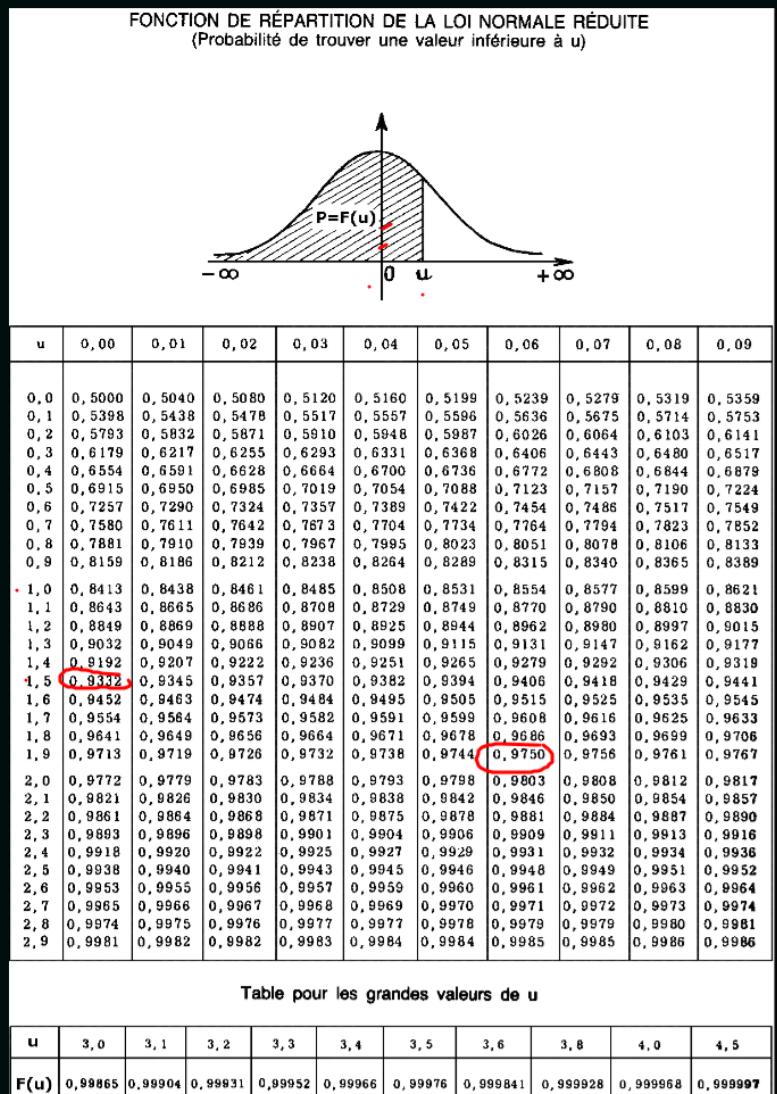
How to compute $P(X \in [a, b])$?

Properties:

Let X be a $\mathcal{N}(\mu, \sigma^2)$

$$Y = \frac{X - \mu}{\sigma} \sim \mathcal{N}(0, 1)$$

standard gaussian



Exercise:

Compute for $X \sim \mathcal{P}(0; 1)$

$$- P(X \in [2; 2.4])$$

$$- P(X \in [-1; 1])$$

$$- P(X \geq -2)$$

$$\begin{aligned} P(X \in [a, b]) &= P(X \leq b) \\ &\quad - P(X < a) \\ &= F(b) - F(a) \end{aligned}$$

? How to compute $F(u)$ for $u < 0$?

Let $Y \sim \mathcal{N}(0, 1)$

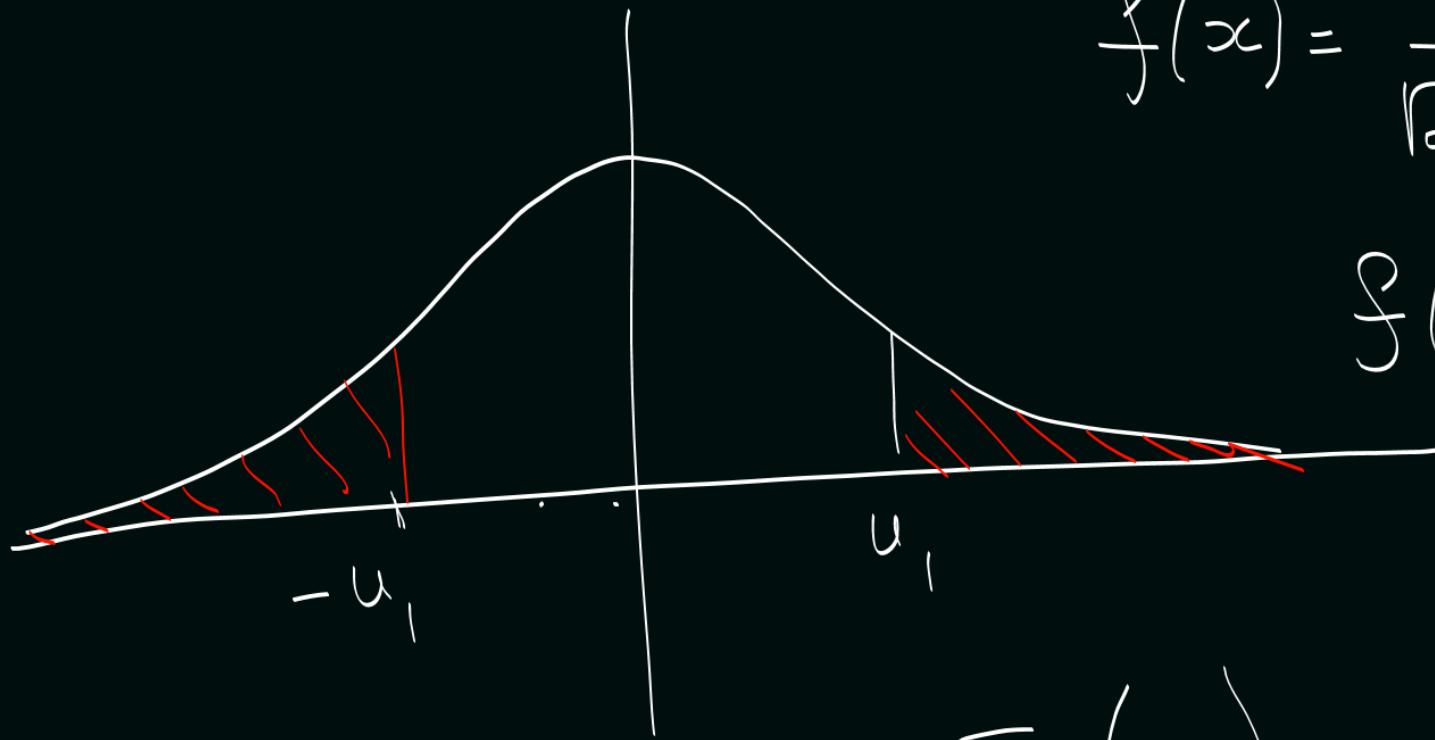
let $\tau > 0$ and $\mu \in \mathbb{R}$

$$X = \mu + \tau Y$$

$$\Rightarrow X \sim \mathcal{N}(\mu, \tau^2)$$

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

$$f(x) = f(-x)$$



$$F_X(-\omega) = 1 - F_X(\omega)$$

We are not able to compute

$$\int_a^b \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{1}{2}x^2\right) dx$$

- ⇒ This quantity can be approximated
- ⇒ the approximations for a standard Gaussian are in a table

How to make the computations
with the R software?

for uniform: `runiform`, `puniform`, `duniform`

Simulations
of $U(0; 1)$

$$\begin{aligned}\mathcal{P}(X \geq -2) &= 1 - \mathcal{P}(X \leq -2) \\&= 1 - F_X(-2) \\&= 1 - \left(1 - F_X(2)\right) \\&= F_X(2) \\&= 0.972\end{aligned}$$

$$\cdot P(X \in [2, 2.4]) \\ = F_X(2.4) - F_X(2)$$

$$= 0.9918 - 0.9772$$

$$= 0.0146$$

$$P(X \in [-1, 1])$$

$$= F_X(1) - F_X(-1)$$

$$= F_X(1) - (1 - F_X(1))$$

$$= 2 \cdot F_X(1) - 1$$

$$= 2 \times 0.8413 - 1 = 0.6826$$

Compute for $X \sim \mathcal{U}(2, 4)$

$$- P(X < 2) = 0.5$$

$$- P(X \in [2, 4])$$

$$X \sim \mathcal{U}(2, 4) \Rightarrow Y = \frac{X-2}{2} \sim \mathcal{U}(0, 1)$$

$$P(X \in [2, 4]) = P(Y \in [0, 1]) = F_Y(1) - F_Y(0) \\ = 0.8413 - 0.5 = 0.3413$$