

Lezione 5 -

Recap

↳ We have introduced random variables, absolutely continuous, and discrete random variables.

Definition of Discrete random variables

↳ A random variable which assumes values in a finite subset of \mathbb{R} called S

$$P(X \in S) =$$

Discrete density \rightarrow probability X is equal to any x

$$- p(x) = 0 \quad \forall x \notin S$$

$$- p_x(x) > 0 \text{ for all } x \in S, \text{ and } \sum_{x_i} p(x_i) = 1$$

Theorem \rightarrow density \rightarrow distribution function.

$$\text{1. } F_x(x) = P(X \leq x) = \sum_{i: x_i \leq x} p_x(x_i), \quad x \in \mathbb{R}$$

2. if $x_1 \leq x_2 \leq x_3 \leq \dots$ Distribution function \rightarrow Probability density.

$$p_x(x_k) = F_x(x_k) - F_x(x_{k-1}) \quad \text{if } k \geq 2 \quad p_x(x_1) = F_x(x_1)$$

Example : $X = \text{# of bits received wrong for 4 transmitted}$

transmission $n = 4$ bits each $\begin{cases} \text{on} \\ \text{away} \end{cases}$

$$X \in S = \{0, 1, 2, 3, 4\}$$

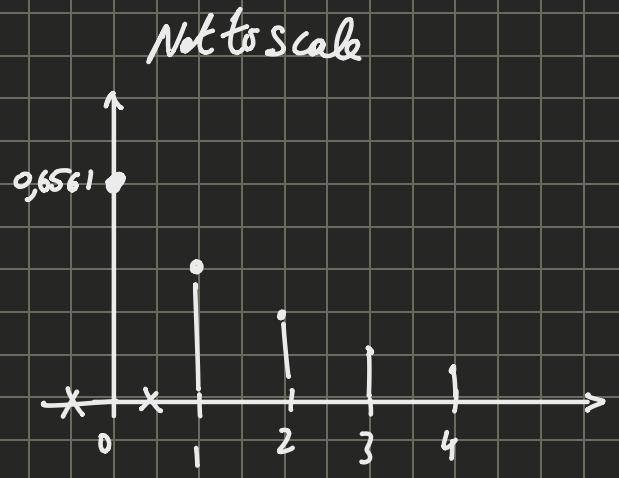
$$p_x(0) = P(X=0) = 0,6561$$

$$p_x(1) = P(X=1) = .2916$$

$$p_x(2) = = .0496$$

$$p_x(3) = = 0,0036$$

$$p_x(4) = = 0,0001$$



$$F_x(x) = \sum_{\{x_i \leq x\}} p_x(x_i)$$

$$\text{if } x < 0 \quad F_x(x) = 0$$

$$\text{if } 0 \leq x < 1 \quad F_x(x) = P(X \leq x) = p_x(0) = .6561$$

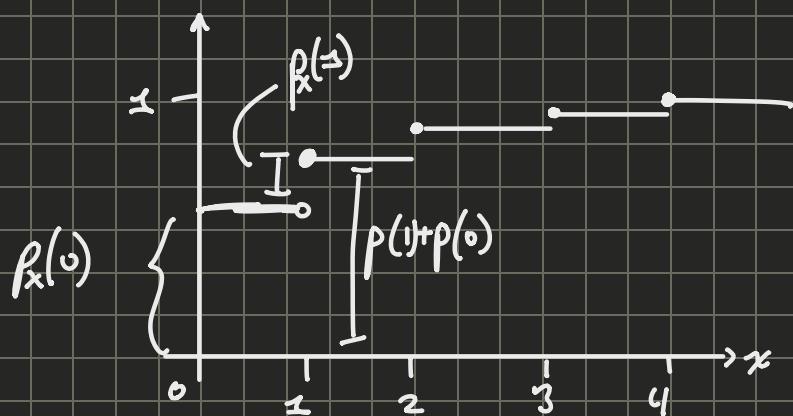
\downarrow

$$= F_x(0)$$

$$\text{if } 1 \leq x < 2 \quad F_x(x) = P(X \leq x) = p_x(0) + p_x(1) = .9477$$

$$\text{if } x \geq 2 \quad F_x(x) = P(X \leq x) = p_x(0) + p_x(1) + p_x(2) + p_x(3) + p_x(4) = 1$$

\Rightarrow Related distribution function



The step/jump
if the density is
at the point.

The distribution of a discrete random variable is a step function. The jumps are at the support points.

The jump in the steps are the probabilities of the individual support points.

We can recover the density by looking at all the steps.

$$p_x(x_k) = F_x(x_k) - F_x(x_{k-1})$$

Theorem

To assign a discrete random variable it's sufficient to assign its density.

Mean of a discrete random variable

Discrete random variable with density p_x and $S = \{x_1, x_2, \dots\}$

$$E(x) = \sum_{i=1}^{\infty} x_i p_x(x_i)$$

The variance is:

$$\text{Var}(x) = \sum_{i=1}^{\infty} (x_i - E(x)) \cdot p_x(x_i)$$

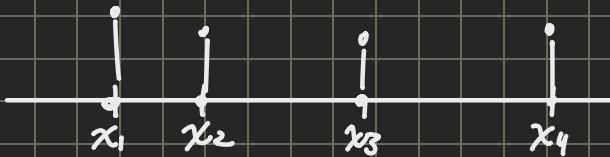
The standard deviation:

$$\sigma(x) = \sqrt{\text{Var}(x)}$$

$$P(X \in B) = \sum_{i, x_i \in B} p_x(x_i)$$

$B \subset \mathbb{R}$

$S \subset \mathbb{R}$



\bar{x} = the center of gravity of the random variable

$$\text{Var}(x) = \sum_{i=1}^{\infty} (x_i - \bar{x})^2 p_x(x_i) \rightarrow \text{indicates the variability around the mean.}$$

Properties:

$$\text{Var}(x) \geq 0$$

If $\text{Var}(x) = 0 \Leftrightarrow x = \text{constant almost surely.}$

$$\Rightarrow P(x=c) = 1 \quad c = \text{some constant}$$

$$\{x_1, x_2, \dots\}$$

Let X be a discrete random variable with S_x and p_x

let's consider $g: \mathbb{R} \rightarrow \mathbb{R}$ such that $Y = g(x)$

e.g. $Y = \log(x)$, $Y = x^2$, $Y = \sqrt{x}$

$$E[g(x)] = \sum_{i=1}^{\infty} g(x_i) \cdot p_x(x_i)$$

$$E[X^2] = \sum_{i=1}^{\infty} x_i^2 p_x(x_i)$$

\hookrightarrow still second moment

$$E[X^n] = n^{\text{th}} \text{ moment of } X$$

Properties of E and Var (Analogs for discrete)

$$\hookrightarrow E[aX+b] = aEX + b$$

$$\hookrightarrow \text{Var}(aX+b) = a^2 \text{Var}(X)$$

$$\hookrightarrow \text{Var}(X) = E(X^2) - E(X)^2$$

$$\hookrightarrow \text{Var}(X) = 0 \text{ only if } X = c \text{ almost surely.}$$

some constant

Degenerate Random Variable

\hookrightarrow Random variable $X = c$ almost surely, where c is a real-valued constant

Bernoulli Random Variable.

$X \sim \text{Be}(p)$ if:

$$X = \begin{cases} 1 & \text{with probability } p \\ 0 & \text{with probability } 1-p \end{cases}$$

$$p \in (0, 1)$$

$$\hookrightarrow EX = 1 \cdot p_X(1) + 0 \cdot p_X(0) = p \cdot 1 + 0 \cdot (1-p) = p$$

$$\text{Var}(X) = E(X^2) - E(X)^2 = p - p^2 = p(1-p)$$

$$E(X^2) = p \quad E(X)^2 = p^2$$

X is a uniform discrete random if $S = \{x_1, x_2, \dots, x_n\}$

$$\text{and } p_{X_i}(x_i) = p = \frac{1}{n} \quad \forall i$$

\hookrightarrow # of elements in S

$$\hookrightarrow 1^2 \cdot p_{X_i}(1) + 0^2 \cdot p_{X_i}(0) = 1 \cdot p + 0 \cdot (1-p) = p$$

Sequence of n Bernoulli Trials

Let's introduce a sequence of Bernoulli trials.

Hypothesis ① The outcome of single trial can

only be
 SUCCESS - same result
 FAILURE - it's complement

Hypothesis ② The probability of "success" is constant, i.e. does not change with the trial.
 $p \in (0, 1)$

Hypothesis ③ The outcomes are independent, the outcome of one trial does not affect the outcome of another.

Let us consider $X = \text{number of successful result in } n \text{ Bernoulli trials.}$

$p = \text{probability of success in a trial } p \in (0, 1)$

Then

$X \sim \text{Bin}(n, p)$ Binomial with parameters n and p .

$$P_X(x) = P(X=x) = \binom{n}{x} p^x (1-p)^{n-x}, x=0, 1, \dots, n \quad S_x = \{0, 1, \dots, n\}$$

$\binom{n}{x}$ choose p $\hookrightarrow x \text{ number of success in } n$
 \hookrightarrow Bernoulli trials.

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

$\binom{n}{0} := 1$, come la availability all'insieme può essere necessario.

$$\mathbb{E}X = n \cdot p$$

$$\text{Var}(X) = np(1-p)$$

$$X \sim \text{Be}(p) = X \sim \text{Bin}(1, p)$$

A sequence of n Bernoulli is a sequence of n independent trials with only two possible results.

Exercise:

Several microstructure, each has probability 5% to develop a crack

1) what is the probability that at least one crack micro structure over $n=20$ tested

$X_{20} = n$ of cracked microstructure over the 20 tested

$$X \sim \text{Bin}(20, 0,05)$$

"Success" = crack in micro structure

$$\begin{aligned} P(X_{20} \geq 1) &= 1 - P(X_{20} < 1) = 1 - P(X_{20} = 0) = 1 - \binom{20}{0} p^0 (1-p)^{20} \\ &= 1 - (1-p)^{20} \\ &= 0,6415 \end{aligned}$$

Exercise 2

↳ Probability that there will be at most 3 [✓] cracked microstructures over $n=100$ examined

$$n=100 \quad p=0,05$$

X_{100} = n° of cracked microstructures over 100 tested

$$X_{100} \sim \text{Bin}(100, 0,05)$$

$$\begin{aligned} P(X_{100} \leq 3) &= P(X_{100}=0) + P(X_{100}=1) + P(X_{100}=2) + P(X_{100}=3) \\ &= \underbrace{\binom{100}{0} p^0 (1-p)^{100} + \binom{100}{1} p^1 (1-p)^{99} + \binom{100}{2} p^2 (1-p)^{98} + \binom{100}{3} p^3 (1-p)^{97}}_{= 0,2578} \end{aligned}$$

Example:

Shooting a target, Share probability of hitting it is $p=0,2$

1) $n=2$, what is the probability of hitting the target at least once?

X_2 = number of times target hit over 2 trials $\sim \text{Bin}(2, 0,2)$

$$P(X_2 \geq 1) = 1 - P(X_2=0) = 1 - \binom{2}{0} p^0 (1-p)^2 \approx 0.36$$

2) n to have more than 50% probability of hitting the target at least once

$$X_n \sim \text{Bin}(n, p) \rightarrow P(X_n \geq 1)$$

(so what we want > 0.5)

$$0,5 < P(X_n \geq 1) = 1 - P(X_n = 0)$$

$$\Rightarrow P(X_n = 0) < 0,5$$

$$\cancel{\left(\frac{n}{\cancel{p}}\right)}^n p^{\cancel{n}} (1-p)^n \Rightarrow (1-p)^n < \frac{1}{2}$$

$$n \ln(0.8) < -\ln(2) \Rightarrow n = \frac{-\log(2)}{\log(0.8)} = 3,2$$

$\Rightarrow n \geq 4$ to have probability of hitting target at least once

Poisson Distribution

Taking $X \sim \text{Bin}(n, p)$ but with $n \rightarrow \infty$, and p decreases such that $np = \lambda = \text{const.}$

Then

$$\text{if } k=0, \dots, n, \quad P(X=k) = \binom{n}{k} p^k (1-p)^{n-k} \rightarrow e^{-\lambda} \frac{\lambda^k}{k!} \quad k \geq 0, 1.$$

