

$$X \quad P_X \quad E[X] = \sum_{x \in \mathbb{R}} x P_X(x)$$
$$E[|X|^k] < +\infty \quad \longrightarrow \quad E[X^k] = \sum_{x \in \mathbb{R}} x^k P_X(x)$$

X ha momento k -esimo finito

$$E[\phi(X)] = \sum_{x \in \mathbb{R}} \phi(x) P_X(x)$$

$$E[|X - E[X]|^k]$$

Def: X v. a. discreta, se ha momento secondo finito, posso considerare

$$E[(X - E[X])^2] = \text{Var}(X) \leftarrow \text{Varianza}$$

X P_X

$$\text{Var}(X) = \sum_{x \in R} \underbrace{(x - E[X])^2}_{P_X(x)}$$

$$\text{Var}(X) \geq 0 \longrightarrow \sqrt{\text{Var}(X)} \leftarrow \text{deviazione standard.}$$

PROPRIETA' DELLA VARIANZA

$$\begin{aligned} \bullet \text{Var}(X) &= E[(X - E[X])^2] = E[X^2 + (E[X])^2 - 2XE[X]] = \\ &= E[X^2] + E[(E[X])^2] + E[-2XE[X]] = \\ &= E[X^2] + (E[X])^2 - 2(E[X])^2 = \\ &= E[X^2] - (E[X])^2 \end{aligned}$$

$$\begin{aligned} \cdot \text{Var}(cX) &= E[(cX - E[cX])^2] = \leftarrow \text{def} \\ &= E[(cX)^2] - (E[cX])^2 = \leftarrow \text{formula alternative} \\ &= E[c^2 X^2] - (c E[X])^2 = \\ &= c^2 E[X^2] - c^2 (E[X])^2 = \\ &= c^2 (E[X^2] - (E[X])^2) = c^2 \text{Var}(X) \end{aligned}$$

$$\text{Var}(-X) = \text{Var}(X)$$

- $$\begin{aligned} \text{Var}(X+c) &= E[(X+c - E[X+c])^2] = \\ &= E[(X+c - E[X]-c)^2] = \\ &= E[(X - E[X])^2] = \text{Var}(X) \end{aligned}$$
- $$\text{Var}(X) = 0 \iff X = \text{cst}$$

$$\iff X = \text{cst} \quad X = c \quad E[X] = c \quad E[(X - E[X])^2] =$$

$$= E[(c - c)^2] = 0$$

$$\Rightarrow \text{Var}(X) = 0 \Rightarrow X = \text{cst}$$

$$\text{Var}(X) = \sum_{x \in \mathbb{R}} (x - E[X])^2 p(x) = 0$$

$$\updownarrow$$
$$(x - E[X])^2 p(x) = 0$$

$$\forall x \in \mathbb{R} \quad x = E[X]$$

$$\Downarrow$$
$$X = \text{cst} = E[X]$$

Teo Tchebyscheff: X v.2. con varianza finita, allora

$\forall \varepsilon > 0$ vale

$$P(|X - E[X]| > \varepsilon) \leq \frac{\text{Var}(X)}{\varepsilon^2}$$

Teorema di Markov: Y v.2. con speranza finita,
allora $\forall \varepsilon > 0$

$$P(|Y| > \varepsilon) \leq \frac{E[|Y|]}{\varepsilon} \Leftrightarrow$$

$$E[|Y|] \geq \varepsilon P(|Y| > \varepsilon)$$

$$\begin{aligned} \mathcal{P}(|X - E[X]| > \varepsilon) &= \mathcal{P}(\underbrace{(X - E[X])^2 > \varepsilon^2}) \leq \\ &\leq \frac{E[(X - E[X])^2]}{\varepsilon^2} = \frac{\text{Var}(X)}{\varepsilon^2} \end{aligned}$$

$$\mathcal{P}(|Y| > 3)$$

$$\boxed{Y^2 > 9} \iff \boxed{|Y| > 3}$$

$$y^2 > 9 \iff |y| > 3$$

$B(1, p)$ Bernoulliana

$$X \begin{cases} 1 \\ 0 \end{cases}$$

$$p = P(X=1)$$

$$1-p = P(X=0)$$

$$E[X] = p$$

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

$$E[X^2] = p$$

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

$X^2 \begin{cases} 1 \\ 0 \end{cases}$ $P(X^2=1) = P(X=1) = p$ $P(X^2=0) = P(X=0) = 1-p$
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$$X \sim \mathcal{P}(\lambda) \quad E[X] = \lambda \quad \text{Var}(X) = E[X^2] - (E[X])^2$$

$$E[X^2] = \sum_{k=0}^{\infty} k^2 e^{-\lambda} \frac{\lambda^k}{k!} = e^{-\lambda} \lambda \sum_{k=1}^{\infty} k \frac{\lambda^{k-1}}{(k-1)!} \quad k-1 = h$$

$$= e^{-\lambda} \lambda \sum_{h=0}^{\infty} (h+1) \frac{\lambda^h}{h!} \Rightarrow \lambda \left(\sum_{h=0}^{\infty} (h+1) \frac{e^{-\lambda} \lambda^h}{h!} \right) =$$

$$\text{Var}(X) = \cancel{\lambda^2} + \lambda - \cancel{\lambda^2} = \lambda \quad \Bigg| \quad = \lambda E[X+1] = \lambda (E[X] + 1) = \lambda^2 + \lambda$$

Geometric

$$\text{Var}(X) = E[X^2] - (E[X])^2 = \sum_{k=1}^{\infty} k^2 p(1-p)^{k-1} - \frac{1}{p^2} = \frac{2-p}{p^2} - \frac{1}{p^2} = \frac{1-p}{p^2}$$

$$X \sim \mathcal{B}(n, p) \quad X = \sum_{i=1}^n X_i \quad X_i \sim \mathcal{B}(1, p)$$

independent:

$$\text{Var}(X) = \text{Var}\left(\sum_{i=1}^n X_i\right)$$

COVARIANZA.

$$E[XY] = E[X]E[Y] \leftarrow \text{vero se } X, Y \text{ sono indipendenti}$$

Def X, Y v.2.

$$\text{Cov}(X, Y) = E[(X - E[X]) \cdot (Y - E[Y])]$$

$$\text{Cov}(X, Y) \leq \sqrt{\text{Var}[X]} \cdot \sqrt{\text{Var}[Y]}$$

$$P_{xy} \quad \text{Cov}(X, Y) = \sum_{x, y} (x - E[X]) (y - E[Y]) \cdot P_{xy}(x, y)$$

$$\begin{aligned}
 \text{Cov}(X, Y) &= E[(X - E[X])(Y - E[Y])] = \\
 &= E[XY - XE[Y] - YE[X] + E[X]E[Y]] \\
 &= E[XY] - E[Y]E[X] - \cancel{E[Y]E[X]} + \cancel{E[X]E[Y]}
 \end{aligned}$$

Def: diremo che due variabili aleatorie X, Y sono non correlate se la $\text{Cov}(X, Y) = 0$

Se X, Y sono ind. $\Rightarrow E[XY] = E[X]E[Y] \Rightarrow \text{Cov}(X, Y) = 0$
 $\Rightarrow X, Y$ sono non correlate.

$$X \begin{cases} -1 & \frac{1}{3} \\ 0 & \frac{1}{3} \\ 1 & \frac{1}{3} \end{cases}$$

$$Y = X^2$$

$$E[X] = -1 \cdot \frac{1}{3} + 0 \cdot \frac{1}{3} + 1 \cdot \frac{1}{3} = 0$$

$$X^2 \begin{cases} 1 & \frac{2}{3} \\ 0 & \frac{1}{3} \end{cases}$$

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

$$\text{Cov}(X, Y) = E[X \cdot X^2] - \underbrace{E[X] \cdot E[X^2]}_0 = E[X^3] = (-1)^3 \cdot \frac{1}{3} + 0^3 \cdot \frac{1}{3} + 1^3 \cdot \frac{1}{3} = 0$$

$$P(X = 1, Y = 0) = P(X = 1, X^2 = 0) = 0$$

$$P(X = 1) \cdot P(X^2 = 0) = \frac{1}{3} \cdot \frac{1}{3} = \frac{1}{9}$$

X Y

$$\begin{aligned}
 \text{Var}(X+Y) &= E[(X+Y - E[X+Y])^2] = \\
 &= E[((X - E[X]) + (Y - E[Y]))^2] = \\
 &= E[(X - E[X])^2 + (Y - E[Y])^2 + 2(X - E[X])(Y - E[Y])] \\
 &= \text{Var}(X) + \text{Var}(Y) + \underline{\underline{2 \text{Cov}(X, Y)}}
 \end{aligned}$$

$$\text{Var}[X+Y] = \text{Var}[X] + \text{Var}[Y] \Rightarrow X, Y \text{ sono non correlate}$$

$$X = \sum_{i=1}^n X_i \quad X_i \text{ independent.}$$

$$\text{Var } X = \text{Var} \left(\sum_{i=1}^n X_i \right) = \sum_{i=1}^n \text{Var}(X_i) = \sum_{i=1}^n p(1-p) = np(1-p)$$

2 telefoni

Mediamente al primo telefono arrivano 5 telefonate
al secondo 3 telefonate.

Qual è la probabilità che in tutto arrivino 10 telefonate

$$X \sim P(5) \quad Y \sim P(3) \quad X+Y$$

$$P(X+Y=10) = P\left(\bigcup_{i=0}^{10} (X=i, Y=10-i)\right) = \sum_{i=0}^{10} P(X=i, Y=10-i)$$

$$= \sum_{i=0}^{10} P(X=i) P(Y=10-i) = \sum_{i=0}^{10} e^{-5} \frac{5^i}{i!} \cdot e^{-3} \frac{3^{10-i}}{(10-i)!} =$$

$$= \frac{e^{-(5+3)}}{10!} \left(\sum_{i=0}^{10} \frac{10!}{i!(10-i)!} 5^i 3^{10-i} \right) = \frac{e^{-(5+3)}}{10!} \sum_{i=0}^{10} \binom{10}{i} 5^i 3^{10-i}$$

$$= \frac{e^{-8}}{10!} (5+3)^{10} = \frac{e^{-8}}{10!} 8^{10}$$

$$X+Y \sim P(3+5)$$

