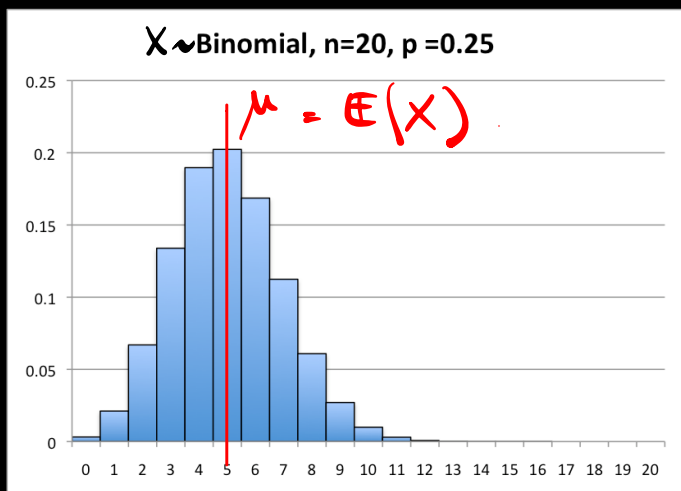


Richiamo: Consideriamo un esperimento con risultati

$$X \in \{0, 1, 2, \dots, m\}$$



fdm $P(X=k)$

$$\Rightarrow P(X \in B) = \sum_{k \in B} P(X=k)$$

$$\bullet E(X) = \sum_{k \in \text{range}(X)} k P(X=k)$$

$$\bullet E(X+Y) = E(X) + E(Y)$$

$$\bullet E(aX) = a E(X)$$

$$\bullet I_A = \begin{cases} 1 & \text{se } A \text{ avviene} \\ 0 & \text{altrimenti} \end{cases}$$

$$P(I_A = 1) = p$$

$$E(I_A) = P(I_A = 1) = p$$

$$X = \sum_{j=1}^n I_{A_j} \Rightarrow E(X) = E\left(\sum_{j=1}^n I_{A_j}\right) = \sum_{j=1}^n E(I_{A_j}) = \sum_{j=1}^n P(I_{A_j} = 1)$$

E: $X \sim \text{Bin}(n, p)$ $E(X) = ?$

$$X = I_{A_1} + I_{A_2} + \dots + I_{A_n} \quad A_j = \{\text{successo alla prova } j\}.$$

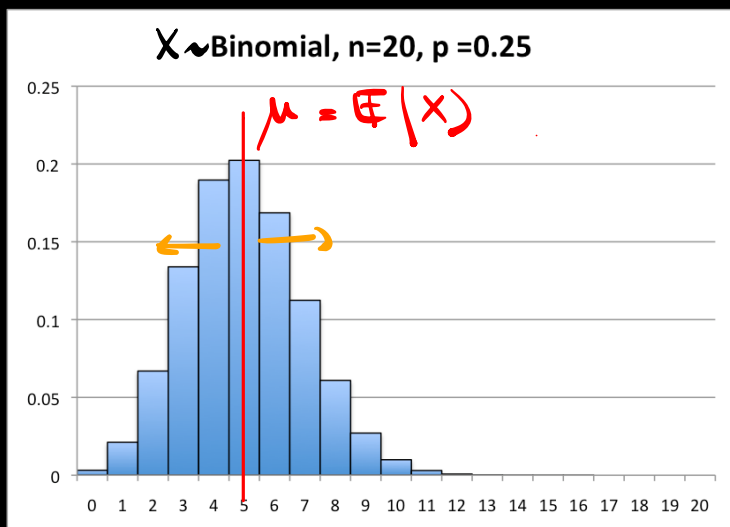
$$P(I_{A_j} = 1) = p$$

$$\begin{aligned} E(X) &= E\left(\sum_{j=1}^n I_{A_j}\right) = \sum_{j=1}^n E(I_{A_j}) \\ &= \sum_{j=1}^n P(I_{A_j} = 1) = \sum_{j=1}^n p = n \cdot p. \end{aligned}$$

Se viene applicato la definizione di valore atteso

$$E(X) = \sum_{k=0}^n k \cdot P(X=k) = \sum_{k=0}^n k \binom{n}{k} p^k (1-p)^{n-k}$$

$$= \underbrace{n}_{\sim} = \dots = n \cdot p$$



$$Var(X) = \sum_{k \in \text{range}} (k - \mu)^2 P(X=k)$$

$$\mu = E(X) \quad E((X - \mu)^2)$$

$$Var(X) = E(X^2) - E(X)^2 \geq 0$$

$$SD(X) = \sqrt{Var(X)}$$

Somme di VA indipendenti

Richiamo: $E(X+Y) = E(X) + E(Y)$

Cosa succede nel caso della varianza?

Fatto: Se X, Y sono indipendenti

$$Var(X+Y) = Var(X) + Var(Y)$$

Richiamo

• Se $X \perp Y$

$$E(XY) = E(X)E(Y)$$

$$Var(X) = E((X-\mu)^2)$$

Dim: $Var(X+Y) = E((X+Y - E(X+Y))^2) =$
 $= E((\underbrace{X - E(X)}_{\hat{X}}) + (\underbrace{Y - E(Y)}_{\hat{Y}}))^2) = E(\hat{X} + \hat{Y})^2$
 $= E(\hat{X}^2 + 2\hat{X}\hat{Y} + \hat{Y}^2) = E(\hat{X}^2) + E(2\hat{X}\hat{Y}) + E(\hat{Y}^2)$
 $= \underbrace{E((X - E(X))^2)}_{Var(X)} + 2 \underbrace{E(X - E(X))E(Y - E(Y))}_{E(X) - E(X) = 0} + \underbrace{E((Y - E(Y))^2)}_{Var(Y)}$

Ez: Lancio due volte una moneta $I_1, I_2 \in \{0, 1\}$

$$Var(I_1) = \frac{1}{2} \left(1 - \frac{1}{2}\right) = \frac{1}{4} \Rightarrow Var(I_1 + I_2) = Var(I_1) + Var(I_2) = \frac{1}{4} + \frac{1}{4} = \frac{1}{2}$$

In generale Siano $X_1, \dots, X_n$ indipendenti allora

$$Var(X_1 + X_2 + \dots + X_n) = Var(X_1) + \dots + Var(X_n)$$

Se $X_1, \dots, X_n$ hanno la stessa distribuzione e indipend.
(Indipendenti e Identicamente Distribuite IID)

$$\Rightarrow E(X_i) = \mu, \quad \text{Var}(X_i) = \sigma^2 \quad \text{per ogni } i$$

Definiamo $S_n = X_1 + \dots + X_n$

- $E(S_n) = E(X_1 + \dots + X_n) = \sum_{i=1}^n E(X_i) = n E(X_1)$

$$\Rightarrow$$

- $\text{Var}(S_n) = \text{Var}(X_1 + \dots + X_n) = \sum \text{Var}(X_i) = n \text{Var}(X_1)$
- $SD(S_n) = \sqrt{\text{Var}(S_n)} = \sqrt{n \cdot \text{Var}(X_1)} = \sqrt{n} SD(X_1)$

Esempio: Lancio 50 dadi con risultato X_j $E(X_j) = 3.5$
 S_{50} = somma dei risultati = $X_1 + \dots + X_{50}$ $\text{Var}(X_j) = 2.91$

$$E(S_{50}) = 50 \cdot 3.5 = 175$$

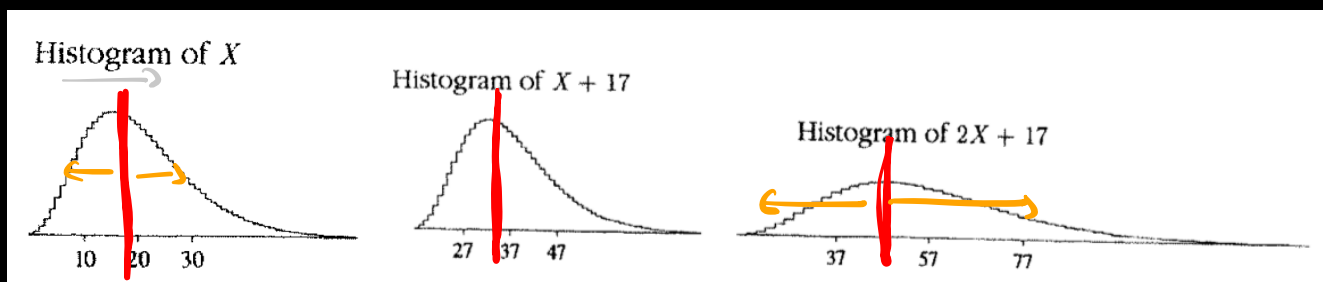
$$\text{Var}(S_n) = 50 \text{Var}(X_1) = 50 \cdot 2.91 = 145$$

Traslationi e scaling

Consideriamo $Y = a \cdot X + b$ → scaling.
→ translation. per a, b fissi

$$\text{Allora } E(Y) = E(aX + b) = E(aX) + E(b) = a E(X) + b$$

$$\begin{aligned} \text{Var}(Y) &= E((Y - E(Y))^2) = E((aX + b) - \underbrace{E(aX + b)}_{aE(X) + b})^2) \\ &= E(a^2(X - E(X))^2) = a^2 E((X - E(X))^2) = \\ &= a^2 \text{Var}(X). \end{aligned}$$



traslare la VA non influenza la forma della sua f.d.m., e quindi varianza, ma solo il valore medio (centro della distribuzione) invece scalare la VA influenza anche la varianza.

Ex: $\text{Var}(X_j) = 7$ $\text{Var}(u X_j) = u^2 \text{Var}(X_j)$.

$= u^2 7$

Δ questo non è uguale a

$$\text{Var}(S_n) = \text{Var}(X_1 + \dots + X_n) = n \text{Var}(X_j) = n 7$$

Consideriamo: $\bar{X}_n = \frac{1}{n} (X_1 + X_2 + \dots + X_n)$ (media empirica)

$$\begin{aligned} \bullet \text{ } E(\bar{X}_n) &= E\left(\frac{1}{n} (X_1 + \dots + X_n)\right) = \frac{1}{n} E(X_1 + \dots + X_n) = \frac{1}{n} \sum_{j=1}^n E(X_j) \\ &= \frac{1}{n} \cdot n E(X_1) = E(X_1) \end{aligned}$$

$$\begin{aligned} \bullet \text{ } \text{Var}(\bar{X}_n) &= \text{Var}\left(\frac{1}{n} (X_1 + \dots + X_n)\right) = \frac{1}{n^2} \cdot \text{Var}(X_1 + \dots + X_n) = \frac{1}{n^2} \sum_{j=1}^n \text{Var}(X_j) \\ &= \frac{1}{n^2} \cdot n \text{Var}(X_1) = \frac{1}{n} \text{Var}(X_1) \end{aligned}$$

Esempio: $X \sim \text{Bin}(n, p)$ $X = I_{A_1} + I_{A_2} + \dots + I_{A_n}$

$$\mathbb{E}(X) = n \cdot p.$$

$$\text{Var}(X) = \text{Var}(I_{A_1} + \dots + I_{A_n}) = n \cdot \text{Var}(I_{A_1}) = n p (1-p)$$

$$\bar{X}_n = \frac{1}{n} (I_{A_1} + \dots + I_{A_n}) = \frac{1}{n} X$$

$$\mathbb{E}(\bar{X}_n) = \mathbb{E}(I_{A_1}) = \mathbb{P}(I_{A_1} = 1) = p$$

$$\text{Var}(\bar{X}_n) = \frac{1}{n} \text{Var}(I_{A_1}) = \frac{1}{n} p (1-p)$$

Esempio dell'esempio $p = \frac{1}{2}$

$$\mathbb{E}(\bar{X}_n) = \frac{1}{2}$$

$$\text{Var}(\bar{X}_n) = \frac{1}{n} \frac{1}{4} \xrightarrow{n \rightarrow \infty} 0 \quad \rightarrow \text{dispersione} \rightarrow 0$$

$$\mathbb{P}\left(\left|\bar{X}_n - \frac{1}{2}\right| > \varepsilon\right) \xrightarrow{n \rightarrow \infty} 0 \quad \forall \varepsilon > 0$$