

# ESEMPI DI DISTRIBUZIONI A.C.

Note Title

23/10/2017

## DISTRIBUZIONE UNIFORME SU UN INTERVALLO $[a, b]$

Si indica  $U([a, b])$  ed è la dist. continua A.C. associata alla densità

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

$$f(x) \geq 0 \quad \forall x \in \mathbb{R}; \quad \int_{\mathbb{R}} f(x) dx = \int_a^b \frac{1}{b-a} dx = 1.$$

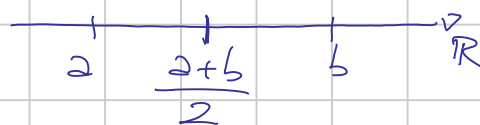
$$U([a, b])(A) = \frac{\mathcal{L}^1(A)}{b-a} \quad \forall A \in \mathcal{B}([a, b])$$

Si è  $X$  v.o. r.c.  $\mathbb{P}_X = U([a, b])$

$$\mathbb{P}(X \in [a, b]) = 1$$

$$\mathbb{E}[X] = \int_{\mathbb{R}} x \mathbb{P}_X(dx) = \int_{\mathbb{R}} x f(x) dx = \int_a^b \frac{x}{b-a} dx =$$

$$= \frac{1}{2(b-a)} x^2 \Big|_{x=a}^{x=b} = \frac{b^2 - a^2}{2(b-a)} = \frac{a+b}{2}$$



$$\mathbb{E}[X^2] = \int_{\mathbb{R}} x^2 \mathbb{P}_X(dx) = \int_{\mathbb{R}} x^2 f(x) dx = \int_a^b \frac{x^2}{b-a} dx =$$

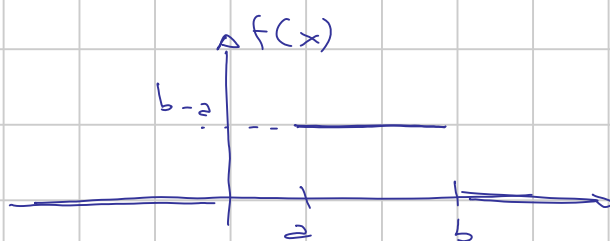
$$= \frac{1}{3(b-a)} x^3 \Big|_{x=a}^{x=b} = \frac{b^3 - a^3}{3(b-a)} = \frac{a^2 + ab + b^2}{3}$$

$a^2 + ab + b^2$

$$\text{Var}[X] = \mathbb{E}[X^2] - (\mathbb{E}[X])^2 = \frac{a^2 + ab + b^2}{3} - \frac{(a+b)^2}{4} =$$

$$= \frac{a^2 - 2ab + b^2}{12} = \frac{(b-a)^2}{12}$$

$$F_X(t) = \int_{-\infty}^t f(x) dx$$



$$t \leq a \quad F_X(t) = \int_{-\infty}^t 0 \, dx = 0$$

$$t \in [a, b) \quad F_X(t) = \int_{-\infty}^t f(x) \, dx = \int_{-\infty}^a 0 \, dx + \int_a^t \frac{1}{b-a} \, dx = \\ = \int_a^t \frac{1}{b-a} \, dx = \frac{t-a}{b-a}$$

$$t \geq b \quad F_X(t) = 1$$



DISTRIBUZIONE GAUSSIANA O NORMALE DI PARAMETRI  
 $m \in \mathbb{R}, \sigma^2 > 0$

Si indica col simbolo  $N(m, \sigma^2)$  ed è la distribuzione  
 A.S. associata alle densità

$$f(x) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left(-\frac{(x-m)^2}{2\sigma^2}\right) \quad \forall x \in \mathbb{R}$$

$$f(x) > 0 \quad \forall x \in \mathbb{R}$$

$$\int_{\mathbb{R}} f(x) \, dx = 1 \quad ?$$

INTEGRALE NOTEVOLE

$$\int_{\mathbb{R}} \exp(-t^2) \, dt = \sqrt{\pi}$$

$$\int_{\mathbb{R}} f(x) \, dx = \frac{1}{\sqrt{2\pi\sigma^2}} \int_{\mathbb{R}} \exp\left(-\frac{(x-m)^2}{2\sigma^2}\right) \, dx$$

$$t = \frac{x-m}{\sigma\sqrt{2}} \quad x = m + \underbrace{t\sigma\sqrt{2}}_{>0} \quad dx = \sigma\sqrt{2} \, dt$$

$$\frac{1}{\sqrt{2\pi\sigma^2}} \int_{\mathbb{R}} \exp(-t^2) \sigma\sqrt{2} \, dx = \frac{1}{\sqrt{2\pi\sigma^2}} \sigma\sqrt{2} \sqrt{\pi} = 1$$

Se  $m=0$   $\sigma=1$ , la distribuzione gaussiana di param:  
 $t$ :  $m=0$  e  $\sigma=1$  si dice GAUSSIANA STANDARD  
 NORMALE STANDARD

e si indica  $N(0, 1)$   
sio  $X_0$  v.e. i.e.  $\mathbb{P}_{X_0} = N(0, 1)$

$$\mathbb{E}[X_0] = \int_{\mathbb{R}} x f(x) dx = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} x \exp\left(-\frac{x^2}{2}\right) dx$$

$$\text{perch\u00e9 } \mathbb{E}[|X_0|] = \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} |x| \exp\left(-\frac{x^2}{2}\right) dx < +\infty$$

$$\Rightarrow \mathbb{E}[X_0] = 0$$

$$\text{Var}[X_0] = \mathbb{E}[X_0^2] = \int_{\mathbb{R}} x^2 \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx =$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} (-x) \left( \underbrace{-x \exp\left(-\frac{x^2}{2}\right)}_{\frac{d}{dx} \exp\left(-\frac{x^2}{2}\right)} \right) dx =$$

$$= \frac{1}{\sqrt{2\pi}} \left\{ (-x) \exp\left(-\frac{x^2}{2}\right) \Big|_{x=0}^{x \rightarrow \pm\infty} + \int_{\mathbb{R}} \exp\left(-\frac{x^2}{2}\right) dx \right\}$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} \exp\left(-\frac{x^2}{2}\right) dx \quad t = \frac{x}{\sqrt{2}} \quad x = t\sqrt{2} \quad dx = \sqrt{2} dt$$

$$= \frac{1}{\sqrt{2\pi}} \int_{\mathbb{R}} \exp(-t^2) \sqrt{2} dt = \frac{1}{\sqrt{2\pi}} \sqrt{2} \sqrt{\pi} = 1$$

$$\mathbb{P}_{X_0} = N(0, 1) \Rightarrow \mathbb{E}[X_0] = 0 \quad \text{Var}[X_0] = 1$$

sia  $X_0$  i.e.  $\mathbb{P}_{X_0} = N(0, 1)$ , siano  $m \in \mathbb{R}$  e  $\sigma > 0$   
e considero la v.e.  $X = m + \sigma X_0$

$$\mathbb{P}_{X_0} = f_0(x) dx \quad f_0(x) = \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right)$$

$$\Rightarrow \mathbb{P}_X = f(x) dx \quad \text{con } f(x) = \frac{1}{|\sigma|} f_0\left(\frac{x-m}{\sigma}\right) =$$

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

$$\Rightarrow \mathbb{P}_X = \mathcal{N}(m, \sigma^2)$$

$$\text{Se } Y \text{ è r.c. } \mathbb{P}_Y = \mathcal{N}(m, \sigma^2) \Rightarrow \mathbb{E}[Y] = \mathbb{E}[m + \sigma X_0]$$

dove  $X_0$  è una qualsiasi v.c. r.c.  $\mathbb{P}_{X_0} = \mathcal{N}(0, 1)$

$$\text{e lo stesso vale per la varianza } \text{Var}[Y] = \text{Var}[m + \sigma X_0]$$

$$\mathbb{E}[Y] = \mathbb{E}[m + \sigma X_0] = \mathbb{E}[m] + \sigma \mathbb{E}[X_0] = m + \sigma \cdot 0 = m$$

$$\text{Var}[Y] = \text{Var}[m + \sigma X_0] = \sigma^2 \text{Var}[X_0] = \sigma^2 \cdot 1 = \sigma^2$$

$$F_Y(t) = \mathbb{P}(Y \leq t) = \mathbb{P}(m + \sigma X_0 \leq t) = \mathbb{P}\left(X_0 \leq \frac{t-m}{\sigma}\right)$$

$$\longrightarrow \triangleright F_Y(t) = F_{X_0}\left(\frac{t-m}{\sigma}\right) \quad \forall t \in \mathbb{R}$$

$F_{X_0}$ , legge associata a  $\mathcal{N}(0, 1)$  si indica  $\underline{\Phi}$

$$\underline{\Phi}(t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{x^2}{2}\right) dx$$

$$\text{PROPRIETÀ } \underline{\Phi}(t) + \underline{\Phi}(-t) = 1 \quad \forall t \in \mathbb{R}$$

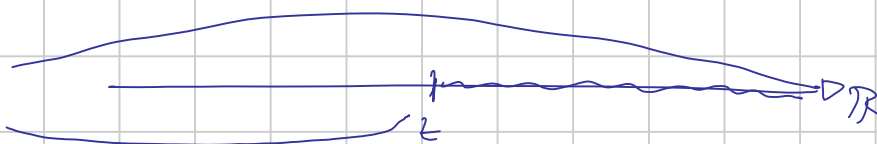
Oss È vero per ogni legge di distribuzione A.c. in cui la densità associata ha una funzione pari.

$$\underline{\Phi}(-t) = \int_{-\infty}^{-t} f(x) dx \quad y = -x \quad x = -y \quad dx = -1 \cdot dy$$

$$= \int_{+\infty}^t f(-y) (-1) dy = \int_t^{+\infty} f(-y) dy = \int_t^{+\infty} f(y) dy =$$

$$= \int_{\mathbb{R}} f(y) dy - \int_{-\infty}^t f(y) dy = 1 - \underline{\Phi}(t)$$

$$\int_{\mathbb{R}} = \int_{-\infty}^t + \int_t^{+\infty}$$



# DISTRIBUZIONE ESPONENZIALE DI PARAMETRO $\lambda > 0$

Si indica  $\exp(\lambda)$  ed è la distribuzione A.S. associata alle deviate

$$f(x) = \begin{cases} 0 & x \leq 0, \\ \lambda e^{-\lambda x} & x > 0. \end{cases}$$

Se è una v.o. T.c.  $\mathbb{P}_X = \exp(\lambda) = 0 \quad X \geq 0$  p.c.

$f(x) > 0 \quad \forall x > 0$

$$\int_{\mathbb{R}} f(x) dx = \int_0^{+\infty} \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_{x=0}^{x \rightarrow +\infty} = 0 - (-1) = 1$$

Se  $\mathbb{P}_X = \exp(\lambda)$ , voglio calcolare  $\mathbb{E}[X] = \text{Var}[X]$

$$\begin{aligned} \mathbb{E}[X] &= \int_{\mathbb{R}} x f(x) dx = \int_0^{+\infty} x \lambda e^{-\lambda x} dx = \int_0^{+\infty} (-x) \underbrace{(-\lambda e^{-\lambda x})}_{= \frac{d}{dx} e^{-\lambda x}} dx \\ &= (-x) e^{-\lambda x} \Big|_{x=0}^{x \rightarrow +\infty} + \int_0^{+\infty} e^{-\lambda x} dx = \end{aligned}$$

$$= -\frac{1}{\lambda} e^{-\lambda x} \Big|_{x=0}^{x \rightarrow +\infty} = 0 + \frac{1}{\lambda} \cdot 1 = \frac{1}{\lambda}$$

$$\begin{aligned} \mathbb{E}[X^2] &= \int_{\mathbb{R}} x^2 f(x) dx = \int_0^{+\infty} x^2 \lambda e^{-\lambda x} dx = \int_0^{+\infty} (-x^2) \underbrace{(-\lambda e^{-\lambda x})}_{= \frac{d}{dx} e^{-\lambda x}} dx \\ &= (-x^2) e^{-\lambda x} \Big|_{x=0}^{x \rightarrow +\infty} + 2 \int_0^{+\infty} x e^{-\lambda x} dx = \frac{2}{\lambda} \int_0^{+\infty} \lambda x e^{-\lambda x} dx \end{aligned}$$

$$\mathbb{E}[X^2] = \frac{2}{\lambda^2} \quad \text{Var}[X] = \frac{2}{\lambda^2} - \left(\frac{1}{\lambda}\right)^2 = \frac{1}{\lambda^2} \quad \mathbb{E}[X] = \frac{1}{\lambda}$$

$$F_X(t) = \int_{-\infty}^t f(x) dx$$

$$F_X(t) = 0 \quad t \leq 0$$

$$\begin{aligned} t > 0 \quad F_X(t) &= \int_{-\infty}^t f(x) dx = \int_0^t \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_{x=0}^{x=t} \\ &= -e^{-\lambda t} + 1 \end{aligned}$$

$$F_X(t) = \begin{cases} 0 & t < 0, \\ 1 - e^{-\lambda t} & t \geq 0 \end{cases}$$

$$t, s \in \mathbb{R} \quad t, s \geq 0$$

$$\mathbb{P}(X \leq t+s \mid X > s) = \mathbb{P}(X \leq t)$$

PROPRIETÀ DI  
MANCANZA DI  
MEMORIA

$$\mathbb{P}(X \leq t+s \mid X > s) = \frac{\mathbb{P}(X \leq t+s, X > s)}{\mathbb{P}(X > s)} =$$

$$\{X \leq t+s, X > s\} = \{X \leq t+s\} \cap \{X > s\} =$$

$$= \{X \leq t+s\} \setminus \{X \leq s\}$$

$$\{X \leq s\} \subseteq \{X \leq t+s\}$$

$$\mathbb{P}(X \leq t+s, X > s) = \mathbb{P}(X \leq t+s) - \mathbb{P}(X \leq s)$$

$$\mathbb{P}(X > s) = 1 - \mathbb{P}(X \leq s)$$

$$\mathbb{P}(X \leq t+s \mid X > s) = \frac{\mathbb{P}(X \leq t+s) - \mathbb{P}(X \leq s)}{1 - \mathbb{P}(X \leq s)} = \frac{F_X(t+s) - F_X(s)}{1 - F_X(s)}$$

$$= \frac{(1 - e^{-\lambda(t+s)}) - (1 - e^{-\lambda s})}{1 - (1 - e^{-\lambda s})} = \frac{e^{-\lambda s} - e^{-\lambda t - \lambda s}}{e^{-\lambda s}}$$

$$= 1 - e^{-\lambda t} = F_X(t) = \mathbb{P}(X \leq t)$$

Si può dimostrare che la distribuzione esponenziale è l'unica distribuzione concentrata su  $[0, +\infty)$  ma non unicamente in  $\mathbb{Q}$ . I.c. le v.e. aventi tale distribuzione godono delle proprietà di mancanza di memoria.

# ES 4 foglio 2

$$X := \# \text{Terze} - \# \text{croci}$$

$$\Omega = \{0, 1\}^n = \{w = (w_1, \dots, w_n) : w_i \in \{0, 1\}\}$$

0 -> croce

1 -> Terza

$Y(w)$  = v. e. che conta il numero delle Terze usate

$$Y(w) = \sum_{i=1}^n w_i = \# \text{Terze usate}$$

$$\Rightarrow \# \text{croci} = n - Y(w) = n - \sum_{i=1}^n w_i$$

$$X(w) = \sum_{i=1}^n w_i - \left( n - \sum_{i=1}^n w_i \right) = 2 \sum_{i=1}^n w_i - n = 2Y(w) - n$$

$$X(\Omega) = \{-n, -n+2, -n+4, \dots, n-2, n\}$$

$$K \in \{-n, -n+2, \dots, n-2, n\}$$

$$\mathbb{P}(X=k) \quad \{X=k\} = \{Y=?\}$$

$$X = 2Y - n \Rightarrow \{X=k\} = \{2Y - n = k\} = \left\{ Y = \frac{n+k}{2} \right\}$$

$$\mathbb{P}(X=k) = \mathbb{P}\left(Y = \frac{n+k}{2}\right) = \mathbb{B}_{1n}(n, p) \left( \left\{ \frac{n+k}{2} \right\} \right) =$$

$$\begin{cases} \binom{n}{\frac{n+k}{2}} p^{\frac{n+k}{2}} (1-p)^{n-\frac{n+k}{2}} & \frac{n+k}{2} = 0, \dots, n \\ 0 & \text{altrimenti.} \end{cases}$$

n.B.  $\frac{n+k}{2} = 0, 1, \dots, n-1, n$

$$n+k = 0, 2, \dots, 2n-2, 2n$$

$$k = -n, -n+2, \dots, n-2, n \quad \underline{ok}$$

## Es 5 foglio 2

$X_1 := \#$  estratto dalla 1<sup>a</sup> urna  $1, \dots, n$

$X_2 := \#$  estratto dalla 2<sup>a</sup> urna  $1, \dots, n$

$X := \max\{X_1, X_2\} \quad \Rightarrow \quad X(\Omega) = \{1, \dots, n\}$

$\mathbb{P}_{X_1} = \mathbb{P}_{X_2} =$  distribuzioni uniformi su  $\{1, 2, \dots, n\}$  cioè

$$\mathbb{P}(X_1 = j) = \mathbb{P}(X_2 = j) = \frac{1}{n} \quad \forall j = 1, \dots, n$$

$$\mathbb{P}(X = j) = ? \quad \rightarrow \quad \{X = j\} = \{X_1 = j, X_2 \leq j\} \cup \{X_1 < j, X_2 = j\}$$

$$\mathbb{P}(X = j) = \mathbb{P}(X_1 = j, X_2 \leq j) + \mathbb{P}(X_1 < j, X_2 = j) = \textcircled{A}$$

$\{X_1 \in A\}, \{X_2 \in B\}$  sono event. indipendenti.

$$\textcircled{A} = \mathbb{P}(X_1 = j) \mathbb{P}(X_2 \leq j) + \mathbb{P}(X_1 < j) \mathbb{P}(X_2 = j) = \textcircled{B}$$

$$\{X_2 \leq j\} = \bigcup_{k=1}^j \{X_2 = k\} \quad \mathbb{P}(X_2 \leq j) = \sum_{k=1}^j \mathbb{P}(X_2 = k) = \sum_{k=1}^j \frac{1}{n} = \frac{j}{n}$$

$$\{X_1 < j\} = \bigcup_{k=1}^{j-1} \{X_1 = k\} \quad \mathbb{P}(X_1 < j) = \sum_{k=1}^{j-1} \mathbb{P}(X_1 = k) = \sum_{k=1}^{j-1} \frac{1}{n} = \frac{j-1}{n}$$

$$\textcircled{B} = \frac{1}{n} \cdot \frac{j}{n} + \frac{j-1}{n} \cdot \frac{1}{n} = \frac{2j-1}{n^2}$$

$$\mathbb{P}(X = j) = \frac{2j-1}{n^2} \quad \forall j = 1, \dots, n$$

$$\mathbb{E}[X] = \sum_{j=1}^n j \mathbb{P}(X = j) = \sum_{j=1}^n \frac{j(2j-1)}{n^2} =$$

$$= \frac{2}{n^2} \sum_{j=1}^n j^2 - \frac{1}{n^2} \sum_{j=1}^n j$$



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

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

$$E[X] = \frac{2}{n^2} \cdot \frac{n(n+1)(2n+1)}{6} - \frac{1}{n^2} \cdot \frac{n(n+1)}{2} =$$

$$= \frac{n+1}{6n} (4n+2-3) = \frac{(n+1)(4n-1)}{6n}$$

— 0 —

$$X(\Omega) = \{1, \dots, n\}$$

$$X = \max\{X_1, X_2\}$$

$$\{X=j\} = \{X \leq j\} \setminus \{X \leq j-1\}$$

$$P(X=j) = P(X \leq j) - P(X \leq j-1)$$

$$= P(\max\{X_1, X_2\} \leq j) - P(\max\{X_1, X_2\} \leq j-1)$$

$$= P(X_1 \leq j, X_2 \leq j) - P(X_1 \leq j-1, X_2 \leq j-1)$$

$$= P(X_1 \leq j)P(X_2 \leq j) - P(X_1 \leq j-1)P(X_2 \leq j-1)$$

$$= \frac{j}{n} \cdot \frac{j}{n} - \frac{j-1}{n} \cdot \frac{j-1}{n} = \frac{j^2 - (j^2 - 2j + 1)}{n^2} = \frac{2j-1}{n^2}$$

— 0 —

## DISTRIBUZIONE CONGIUNTA

Siano  $X$  e  $Y$  v.e. su  $(\Omega, \mathcal{E}, \mathbb{P})$  e supponiamo che siano entrambe discrete

$$X(\Omega) = \{x_i\}_{i \in \mathcal{I}} \quad Y(\Omega) = \{y_j\}_{j \in \mathcal{J}}$$

$$(X, Y): \omega \in \Omega \mapsto (X(\omega), Y(\omega)) \in \mathbb{R}^2$$

$$(X, Y)(\Omega) \subseteq X(\Omega) \times Y(\Omega) = \{(x_i, y_j) : i \in \mathcal{I}, j \in \mathcal{J}\}$$

$\Rightarrow (X, Y)$  è una funzione vettoriale la cui immagine è un insieme finito o numerabile

$$P_{ij} := \mathbb{P}(X=x_i, Y=y_j) \\ (i, j) \in \mathcal{I} \times \mathcal{J}$$

$$P_i = \mathbb{P}(X=x_i) \\ q_j = \mathbb{P}(Y=y_j)$$

PROP  $\forall i \in \mathcal{I} \quad P_i = \sum_{j \in \mathcal{J}} P_{ij}$

$$\forall j \in \mathcal{J} \quad q_j = \sum_{i \in \mathcal{I}} P_{ij}$$

DM  $P_i = \mathbb{P}(X=x_i) \quad \{X=x_i\} = \bigcup_{j \in \mathcal{J}} \{X=x_i, Y=y_j\}$

$$P_i = \mathbb{P}(X=x_i) = \mathbb{P}\left(\bigcup_{j \in \mathcal{J}} \{X=x_i, Y=y_j\}\right) =$$

$$= \sum_{j \in \mathcal{J}} \underbrace{\mathbb{P}(X=x_i, Y=y_j)}_{P_{ij}} = \sum_{j \in \mathcal{J}} P_{ij}$$

DENSITÀ DISCRETA CONGIUNTA DI  $X$  E  $Y$   
NEL PUNTO  $(x_i, y_j)$