

Legge Gamma

Funzione Gamma di Eulero

$$\Gamma(\alpha) = \int_0^{+\infty} \underbrace{x^{\alpha-1} e^{-x}}_{\text{integrand}} dx, \quad \alpha > 0$$

Per $\alpha < 1$ $x^{\alpha-1} \rightarrow +\infty$ per $x \rightarrow 0^+$

$\frac{x^{\alpha-1}}{e^{-x}}$ è integrabile per $\alpha > 0$ nell'intorno di 0
è limitato nell'intorno di 0

$$\int_0^1 \frac{1}{x^\beta} dx < \infty \quad \underline{\beta < 1}$$

$$\frac{x^{-\beta+1}}{-\beta+1} \Big|_0^1 =$$

$$\int_0^\infty \underbrace{x^{\alpha-1}}_x \underbrace{e^{-x}}_e dx$$

$$\beta = 1 - \alpha < 1$$

$$\boxed{\alpha > 0}$$

$$\Gamma(\alpha) = \int_0^{\infty} x^{\alpha-1} e^{-x} dx$$

$$\Gamma(1) = \int_0^{+\infty} e^{-x} dx = 1$$

$$\Gamma\left(\frac{1}{2}\right) = \sqrt{\pi}$$

$$\int_0^{+\infty} \frac{1}{\sqrt{x}} e^{-x} dx$$

Profiere

$$\Gamma(\alpha+1) = \alpha \cdot \Gamma(\alpha)$$

Demostr.

$$\begin{aligned} \Gamma(\alpha+1) &= \int_0^{+\infty} x^\alpha \underbrace{e^{-x}} dx && \text{fu fash} \\ &= \int_0^{+\infty} (-e^{-x}) \alpha x^{\alpha-1} dx && = \int_0^{+\infty} x^{\alpha-1} e^{-x} dx \\ &= \alpha \cdot \int_0^{+\infty} x^{\alpha-1} e^{-x} dx && = \Gamma(\alpha) \\ &= \alpha \Gamma(\alpha) \end{aligned}$$

Applicazione

$$\Gamma(2) = \Gamma(1+1) = 1 \cdot \Gamma(1) = 1$$

$$\Gamma(3) = \Gamma(2+1) = 2 \cdot \Gamma(2) = 2 \cdot 1 = 2$$

$$\Gamma(4) = \Gamma(3+1) = 3 \Gamma(3) = 3 \cdot 2 \cdot 1 = 6$$

⋮

$$\Gamma(n) = (n-1)! \quad (\text{per induzione})$$

$$\Gamma\left(\frac{3}{2}\right) = \Gamma\left(\frac{1}{2} + 1\right) = \frac{1}{2} \Gamma\left(\frac{1}{2}\right) = \frac{1}{2} \sqrt{\pi}$$

Densität

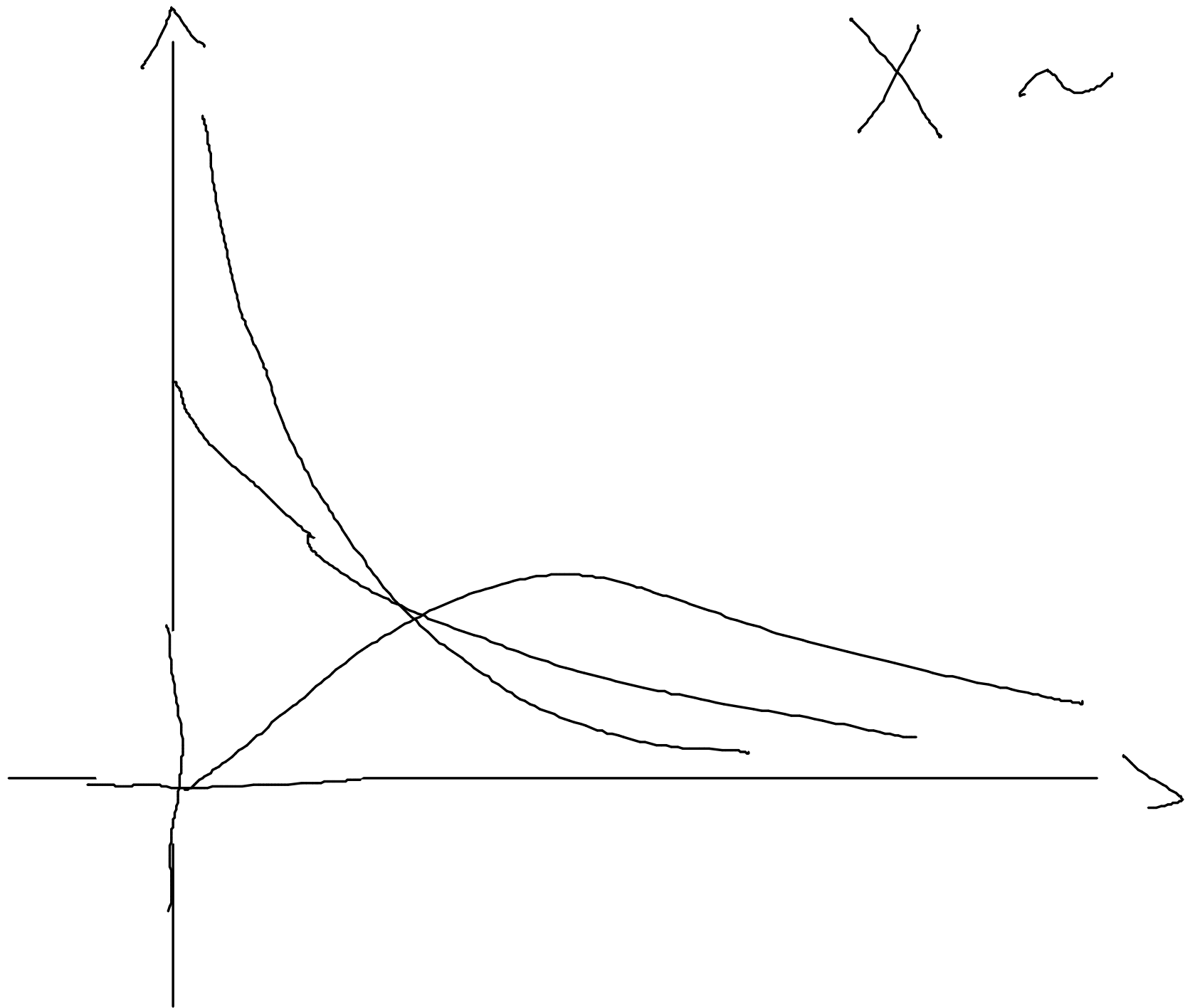
$$\Gamma(\alpha, \lambda)$$

$$\left. \begin{array}{l} \alpha > 0 \\ \lambda > 0 \end{array} \right\} x > 0$$

$$f(x) = \begin{cases} \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} & x > 0 \\ 0 & x \leq 0 \end{cases}$$

$$x \leq 0$$

$$\left. \begin{array}{l} \alpha = 1 \\ \lambda e^{-\lambda x} \\ 0 \end{array} \right\}$$



$X \sim f$

Sia X con densità $\Gamma(\alpha, \lambda)$

Calcoliamo $E[X^k]$

momento di
ordine k

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

$$= \int_0^{+\infty} x^k \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} dx$$

$$= \frac{\Gamma(\alpha+k)}{\lambda^{k+\alpha}} \frac{\lambda^\alpha}{\Gamma(\alpha)} \int_0^{+\infty} \frac{\lambda^{k+\alpha}}{\Gamma(\alpha+k)} x^{(k+\alpha)-1} e^{-\lambda x} dx =$$

= 1

k intero

$$= \frac{\Gamma(\alpha+k)}{\lambda^k \cdot \Gamma(\alpha)} = \frac{(\alpha+k-1) \Gamma(\alpha+k-1)}{\lambda^k \Gamma(\alpha)} = \dots$$

$$E[X^2] = \frac{\Gamma(\alpha+2)}{\lambda^2 \Gamma(\alpha)} = \frac{(\alpha+1) \Gamma(\alpha+1)}{\lambda^2 \Gamma(\alpha)} = \frac{(\alpha+1)\alpha \Gamma(\alpha)}{\lambda^2 \Gamma(\alpha)} =$$

$$E[X] = \frac{\Gamma(\alpha+1)}{\lambda \Gamma(\alpha)} = \frac{\alpha \Gamma(\alpha)}{\lambda \Gamma(\alpha)} = \frac{\alpha}{\lambda}$$

$$\text{Var } X = E[X^2] - E^2[X] = \frac{\alpha(\alpha+1)}{\lambda^2} - \frac{\alpha^2}{\lambda^2} =$$

In fact, for $\alpha=1$ $\text{Var } X = \frac{1}{\lambda^2}$
 (caso dell'esponenziale)
 allora $\frac{\alpha}{\lambda^2}$

Teorema Se X e Y s. a. independentes

ou $X \sim \Gamma(\alpha_1, \lambda)$, $Y \sim \Gamma(\alpha_2, \lambda)$

Allora la s. a. $Z = X + Y \sim \Gamma(\alpha_1 + \alpha_2, \lambda)$

1)
$$\left. \begin{array}{l} X \sim \Gamma(\alpha_1, \lambda) \\ Y \sim \Gamma(\alpha_2, \lambda) \\ Z \sim \Gamma(\alpha_3, \lambda) \end{array} \right\} \text{independentes}$$

$$U = X + Y + Z \sim \Gamma(\alpha_1 + \alpha_2 + \alpha_3, \lambda)$$

$$U = \underbrace{(X + Y)}_{\Gamma(\alpha_1 + \alpha_2, \lambda)} + Z \sim \Gamma(\alpha_3, \lambda)$$

$X + Y$

e Z sono indipendenti?

2) Senza l'indipendenza il risultato
non vale

$$X \sim \Gamma(\alpha, \lambda)$$

$$Y = X \sim \Gamma(\alpha, \lambda)$$

se valesse il teorema,

$X + Y$ dovrebbe avere

legge $\Gamma(2\alpha, \lambda)$

non è vero!

Densità normale (o Gaussiana)

Si chiama densità normale di
parametri μ e σ^2 ($\mu \in \mathbb{R}$, $\sigma > 0$)
la densità definita da

$$f(x) = \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}} \quad x \in \mathbb{R}$$

$$\int_{-\infty}^{+\infty} e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx = \sqrt{2\pi} \sigma$$

Se $X \sim f$ \mathcal{N} allora che

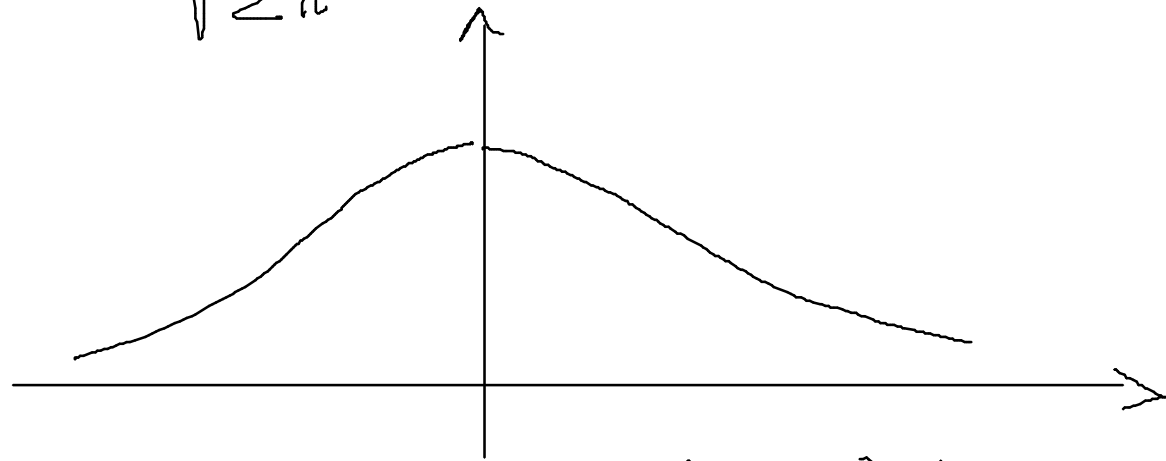
X ha densità normale di par. μ e σ^2

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

\mathcal{L}_e $N(0,1)$ H ~~da~~ χ^2 $\mu=0$
 densità normale standard $\sigma^2=1$

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

$x \in \mathbb{R}$



Calcoli $E[X] = e$ $Var X$

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

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

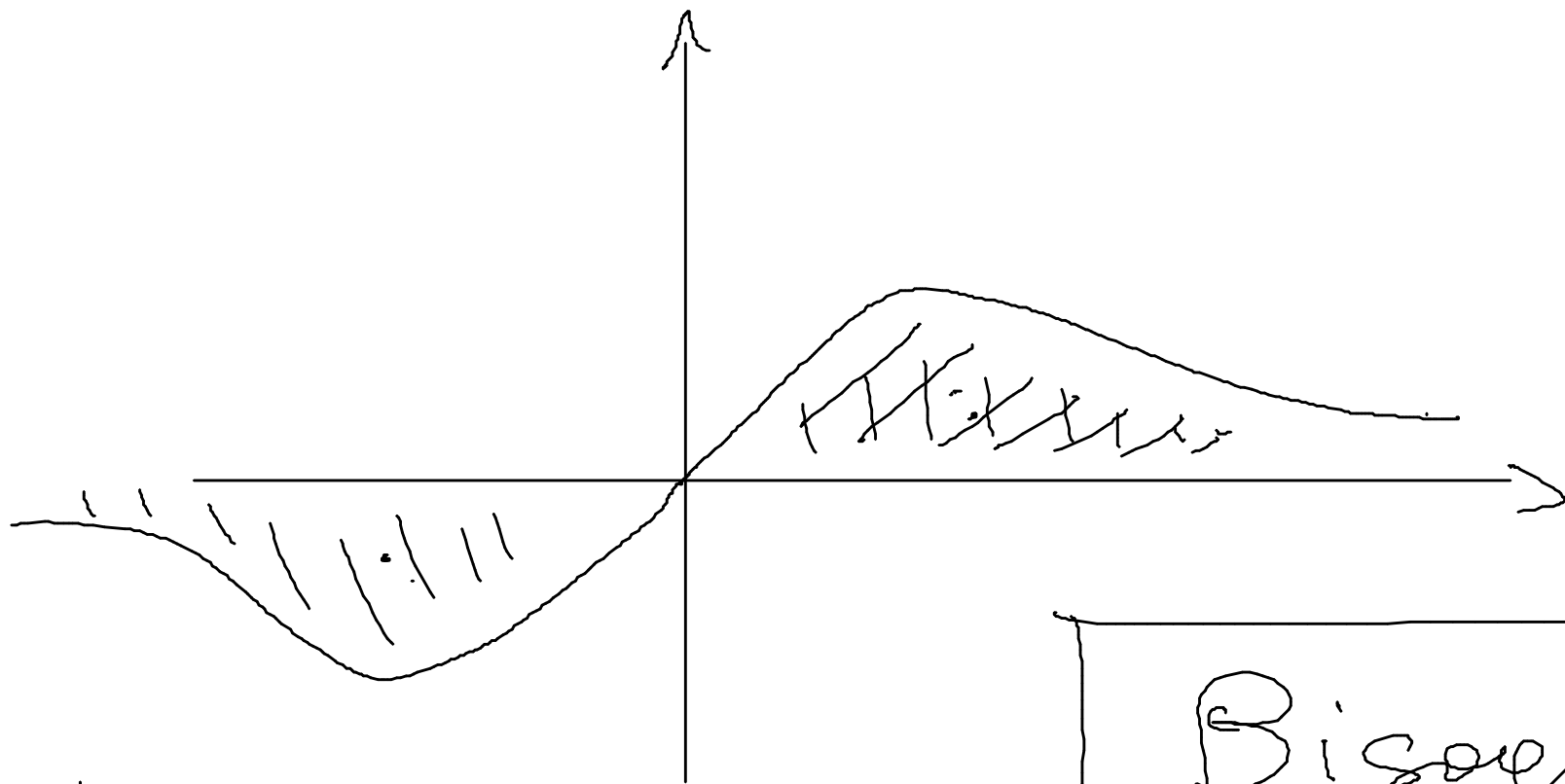
$$= \int_{-\infty}^{+\infty} x \left(\frac{1}{\sqrt{2\pi}} \right) e^{-x^2/2} dx$$

$$= \frac{1}{\sqrt{2\pi}} \left[-e^{-\frac{x^2}{2}} \right]_{-\infty}^{+\infty} = 0$$

$$\frac{1}{\sqrt{2\pi}} x e^{-x^2/2}$$

$$g(x) = x e^{-\frac{x^2}{2}}$$

$$g(-x) = -g(x)$$



$$\int_{-\infty}^{+\infty} g(x) dx = 0$$

Bisogna verificare
che $\int_0^{+\infty} g(x) dx < \infty$

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

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

$$= \int_{-\infty}^{+\infty} x^2 \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} \underbrace{x^2}_{(x)} \cdot \underbrace{e^{-x^2/2}}_{(x e^{-x^2/2})} dx$$

$$= \frac{1}{\sqrt{2\pi}} \left[\cancel{\left[-e^{-\frac{x^2}{2}} \cdot (x) \right]_{-\infty}^{+\infty}} - \int_{-\infty}^{+\infty} -e^{-x^2/2} dx \right] =$$

$$= \int_{-\infty}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = 1$$

$$E[X] = 0$$

$$\text{Var } X = 1$$

$$N(0, 1)$$

$$N\left(\underset{\uparrow}{\mu}, \underset{\uparrow}{\sigma^2}\right)$$

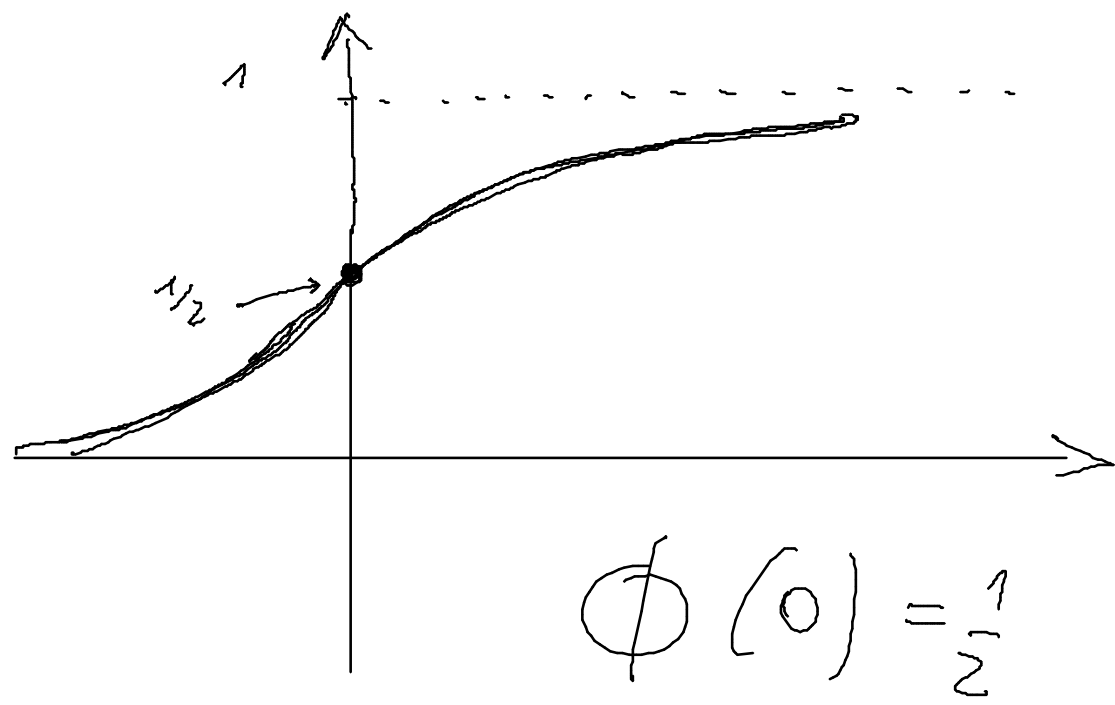
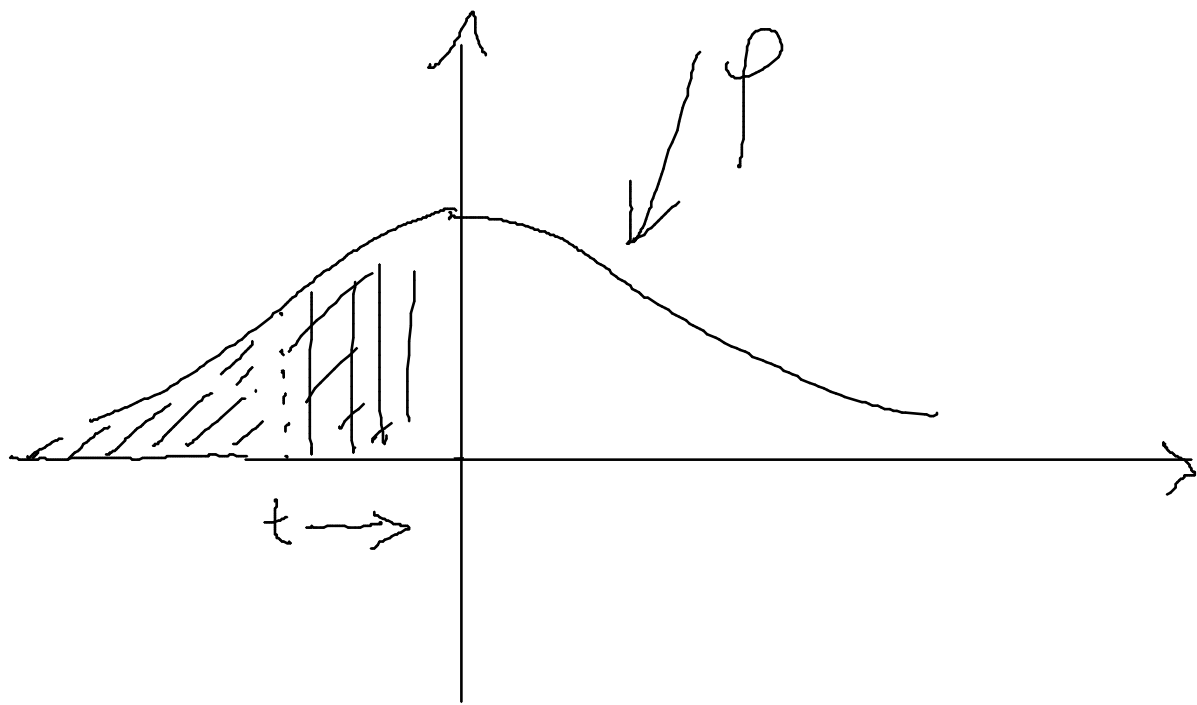
Funzione di ripartizione della $N(0, 1)$

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

~ fratelli

Tavole statistiche

$$\Phi(t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$



$$\Phi(0) = \frac{1}{2}$$

$$\Phi(t) = \int_{-\infty}^t \varphi(x) dx$$

$$\Phi(t) + \Phi(-t) = 1$$

↑

Definizione. Sia X una v. a.

Si dice che X è simmetrica se
 X e $-X$ hanno la stessa legge.

cioè $\forall t$

$$P(X \leq t) = P(-X \leq t)$$

Supponiamo che X sia una v. a. continua

$$P(X = t) = 0 \quad \forall t$$

$$\begin{aligned} \rightarrow P(X \leq t) &= P(-X \leq t) = \\ &= P(X \geq -t) = 1 - P(X < -t) \\ &= 1 - P(X \leq -t) \\ &\quad \underbrace{\hspace{10em}} \\ &\quad P(X = -t) \end{aligned}$$

Quindi, se X è continua e
simmetrica, con f.d.r. F , la
formula precedente diventa

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

$$F(t) = 1 - F(-t)$$

$$F(t) + F(-t) = 1$$

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

Proposizione - Sia $X \sim N(\mu, \sigma^2)$ -

siano $a \neq 0$, $b \in \mathbb{R}$ due costanti.

Allora la v.a.

$$Y = aX + b \sim N(a\mu + b, a^2\sigma^2)$$

Dimostrare.

(i) $a > 0$

$$G(t) = \underline{P(Y \leq t)} = P(aX + b \leq t) =$$

$$= P(aX \leq t - b) \stackrel{\downarrow}{=} P\left(X \leq \frac{t - b}{a}\right) =$$

$$= F_X\left(\frac{t - b}{a}\right) = \int_{-\infty}^{\frac{t - b}{a}} \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x - \mu)^2}{2\sigma^2}} dx$$

$$(ii) \quad a < 0$$

$$G(t) = P\left(X \geq \frac{t-b}{a}\right) =$$

$$= 1 - P\left(X < \frac{t-b}{a}\right) =$$

$$= 1 - \int_{-\infty}^{\frac{t-b}{a}} \frac{1}{\sqrt{2\pi}\sigma} e^{-\dots} \dots$$

$$\frac{(t - (a\mu + b))^2}{2(a\sigma)^2}$$

$$G'(t) = \frac{1}{\sqrt{2\pi} (a\sigma)} e^{-\dots}$$

$$\frac{1}{\sqrt{2\pi} |a|\sigma} e^{-\dots}$$

Corollario

siano $\mu \in \mathbb{R}$, $\sigma > 0$

(i) Se $X \sim \mathcal{N}(0, 1)$, allora

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

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

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

Dim.

$$(ii) \quad Y = \frac{X - \mu}{\sigma} = \underbrace{\left(\frac{1}{\sigma}\right)}_a X + \underbrace{\left(\frac{-\mu}{\sigma}\right)}_b$$

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

$$\sim N\left(\underbrace{\left(\frac{1}{\sigma}\right)\mu + \left(\frac{-\mu}{\sigma}\right)}_{a\mu + b}, \underbrace{\left(\frac{1}{\sigma}\right)^2 \sigma^2}_{a^2 \sigma^2}\right) = N(0, 1)$$

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

Calcolare $E[X]$ e $\text{Var } X$

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

$$X = \sigma Y + \mu$$

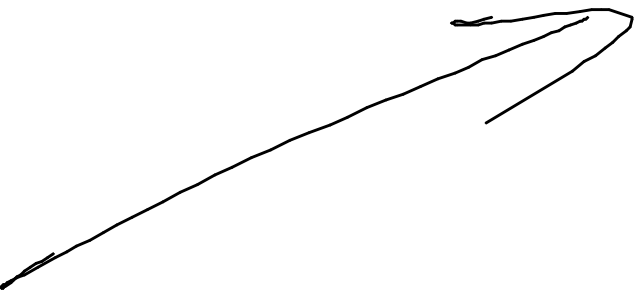
$$E[X] = E[\sigma Y + \mu] = \sigma \underbrace{E[Y]}_{=0} + \mu = \mu$$

$$\begin{aligned} \text{Var } X &= \text{Var}(\sigma Y + \mu) = \text{Var}(\sigma Y) = \\ &= \sigma^2 \underbrace{\text{Var } Y}_{=1} = \sigma^2 \end{aligned}$$

$$\xi \rightarrow X \sim N(\mu, \sigma^2)$$

$$P(X \leq t) = P\left(\underbrace{\frac{X - \mu}{\sigma}}_{N(0,1)} \leq \frac{t - \mu}{\sigma}\right)$$

$$= \Phi\left(\frac{t - \mu}{\sigma}\right)$$



$$G'(t) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{\left(\frac{t-b}{a} - \mu\right)^2}{2\sigma^2}} \cdot \frac{1}{a} =$$

$$= \frac{1}{\sqrt{2\pi}(\sigma a)} e^{-\frac{(t - \overbrace{(a\mu + b)})^2}{2(\underbrace{a\sigma})^2}}$$

$$N(a\mu + b, a^2\sigma^2)$$

Sia ora $X \sim N(0, 1)$

basta vedere che X è simmetrica

$$P(X \leq t) = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx$$

$$P(-X \leq t) = P(X \geq -t) = \left(f(x) = f(-x) \right)$$

$$= \int_{-t}^{+\infty} \frac{1}{\sqrt{2\pi}} e^{-x^2/2} dx = \begin{matrix} x = -y \\ dx = -dy \end{matrix}$$

$$= \int_t^{\infty} \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy = \int_{-\infty}^t \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dy =$$

$$= P(X \leq t)$$

$$\int_{-\infty}^{+\infty} x^n e^{-x^2} dx$$

$$\int x e^{-x^2} dx$$

$$\int e^{-x^2} dx$$

$$\int_{-\infty}^{+\infty} f(x) dx = \int_0^{+\infty} \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} dx = \dots$$

$$\lambda x = y \quad \longrightarrow \quad x = \frac{y}{\lambda} \quad ; \quad dx = \frac{dy}{\lambda}$$

$$= \int_0^{+\infty} \frac{\lambda^\alpha}{\Gamma(\alpha)} \frac{y^{\alpha-1}}{\lambda^{\alpha-1}} e^{-y} \frac{dy}{\lambda} = \frac{1}{\Gamma(\alpha)} \underbrace{\int_0^{+\infty} y^{\alpha-1} e^{-y} dy}_{= \Gamma(\alpha)} = 1$$

Es. Sia $X \sim f$.

$$Z = X + Y = \underline{2X}$$

$$\frac{d}{dt} \int_a^{(t)} f(x) dx = f(t)$$

Calcolare la legge di $2X$

$$G(t) = P(Z \leq t) = P(2X \leq t) =$$

$$= P\left(X \leq \frac{t}{2}\right) = \int_{-\infty}^{\frac{t}{2}} f(x) dx$$

$$G'(t) = \frac{1}{2} f\left(\frac{t}{2}\right)$$

$$G'(t) = \frac{1}{2} f\left(\frac{t}{2}\right) =$$

$$\begin{cases} \frac{\lambda^\alpha}{\Gamma(\alpha)} \left(\frac{t}{2}\right)^{\alpha-1} \frac{1}{2} e^{-\lambda \frac{t}{2}} & t > 0 \\ 0 & t \leq 0 \end{cases} = \Gamma\left(\alpha, \frac{\lambda}{2}\right) \quad \begin{matrix} \lambda > 0 \\ \lambda < 0 \end{matrix}$$

$$\frac{\left(\frac{\lambda}{2}\right)^\alpha}{\Gamma(\alpha)}$$

$$f(t) = \begin{cases} \frac{\mu^\beta}{\Gamma(\beta)} \frac{t^{\beta-1}}{t} e^{-\mu t} & t > 0 \\ 0 & t \leq 0 \end{cases} = \Gamma(\beta, \mu)$$

$$\begin{matrix} \lambda = \mu \\ \alpha = \beta \end{matrix}$$