

Esercizio a casa Costruire gli stimatori puntuali (con i due metodi) per il parametro ϑ nel caso di X_1, \dots, X_n campione statistico di v.e. di Poisson (λ).

Esempi Supponiamo che X_1, \dots, X_n siano distribuite con densità

$$f_{\vartheta}(x) = \begin{cases} (\vartheta+1) x^{\vartheta}, & 0 < x < 1 \\ 0, & \text{alt.} \end{cases}$$

Supponiamo $\vartheta \geq 0$.

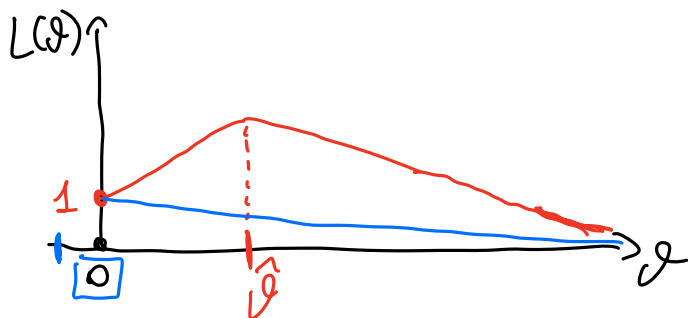
① Metodo verosimiglianza

$$[0, +\infty) \ni \vartheta \mapsto L(\vartheta; x_1, \dots, x_n) = \prod_{i=1}^n f_{\vartheta}(x_i) =$$

$$= \begin{cases} \prod_{i=1}^n (\vartheta+1) x_i^{\vartheta}, & \text{se } x_i \in (0,1) \forall i \\ 0, & \text{se } \exists x_i \notin (0,1) \end{cases}$$

il metodo non funziona

Supponiamo che $x_i \in (0,1) \forall i=1, \dots, n$.



$$L(\vartheta) = (\vartheta+1)^n (x_1 \dots x_n)^{\vartheta}$$

$$L'(\vartheta) = 0 \iff \frac{d}{d\vartheta} \left[n \log(\vartheta+1) + \vartheta \log \left(\prod_{i=1}^n x_i \right) \right] = 0$$

$$\iff \frac{n}{\vartheta+1} + \log \left(\prod_{i=1}^n x_i \right) = 0 \iff \vartheta = \frac{n}{\log \left(\frac{1}{\prod_{i=1}^n x_i} \right)} - 1$$

È una buona stima? $\vartheta \geq 0$?

$$\frac{n}{\log \left(\frac{1}{\prod_{i=1}^n x_i} \right)} - 1 \geq 0 \iff \prod_{i=1}^n x_i \geq e^{-n}$$

$$\hat{\vartheta} = \begin{cases} \frac{n}{\log \left(\frac{1}{\prod_{i=1}^n x_i} \right)} - 1 & , \text{ se } \prod_{i=1}^n x_i \geq e^{-n} \\ 0 & , \text{ se } \prod_{i=1}^n x_i < e^{-n} \end{cases}$$

$$(2) \quad E_{\vartheta} [X^k] = \frac{1}{n} \sum_{i=1}^n x_i^k$$

$k=1$

$$E_{\vartheta} [X] = \bar{x}$$

$$\begin{aligned} E_{\vartheta} [X] &= \int_{-\infty}^{+\infty} x f_{\vartheta}(x) dx = \int_0^1 x \cdot (\vartheta+1) x^{\vartheta} dx = \\ &= \frac{\vartheta+1}{\vartheta+2} x^{\vartheta+2} \Big|_0^1 = \frac{\vartheta+1}{\vartheta+2} \end{aligned}$$

$$\frac{\vartheta+1}{\vartheta+2} = \bar{x} \iff \vartheta+1 = \bar{x}(\vartheta+2) \iff$$

$$(1-\bar{x})J = 2\bar{x} - 1 \iff J = \frac{2\bar{x} - 1}{1-\bar{x}}$$

\uparrow
 $x_i \in (0,1) \forall i$
 $0 < \bar{x} < 1$

È una buona stima? $J \geq 0$?

$$\frac{2\bar{x} - 1}{1-\bar{x}} \geq 0 \iff \bar{x} \geq \frac{1}{2}$$

$$\tilde{J} = \frac{2\bar{x} - 1}{1-\bar{x}} \quad \text{se } \bar{x} \geq \frac{1}{2}$$

Se $\bar{x} < \frac{1}{2}$, consideriamo $\boxed{k=2}$.

$$E_J[X^2] = \int_{-\infty}^{+\infty} x^2 f_J(x) dx = \int_0^1 x^2 (J+1) x^J dx =$$

$$= \frac{J+1}{J+3} x^{J+3} \Big|_0^1 = \frac{J+1}{J+3}$$

$$\frac{J+1}{J+3} = \sum_{i=1}^n \frac{x_i^2}{n} = \bar{m}_2 \iff J+1 = \bar{m}_2 (J+3)$$

$$\iff (1 - \bar{m}_2)J = 3\bar{m}_2 - 1$$

$$\iff J = \frac{3\bar{m}_2 - 1}{1 - \bar{m}_2}$$

È una buona stima se $\bar{m}_2 \geq \frac{1}{3}$.

Se $\bar{m}_2 < \frac{1}{3}$, si continua con $\boxed{k=3}$.

ESEMPI

Supponiamo che le v.e. del campione statistico abbiano densità

$$f_{\theta}(x) = \begin{cases} \frac{1}{2\theta} & , \quad x \in [-\theta, \theta] \\ 0 & , \quad \text{alt.} \end{cases}$$

Stimare $\theta > 0$.

①

$$(0, +\infty) \ni \theta \longmapsto L(\theta; x_1, \dots, x_n) = \prod_{i=1}^n f_{\theta}(x_i) = \begin{cases} \left(\frac{1}{2\theta}\right)^n & , \quad x_i \in [-\theta, \theta] \quad \forall i \\ 0 & , \quad \exists x_i \notin [-\theta, \theta] \end{cases}$$

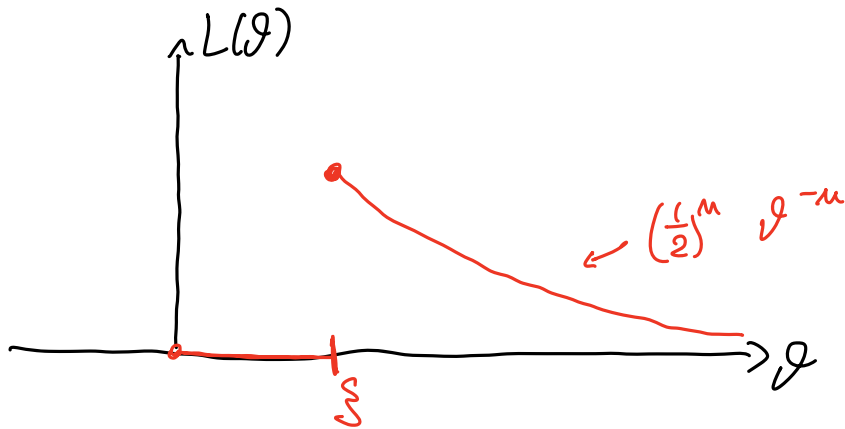
$$x_i \in [-\theta, \theta] \quad \forall i \iff -\theta \leq x_i \leq \theta \quad \forall i \iff -\theta \leq x_i \quad \text{e} \quad x_i \leq \theta \quad \forall i \iff -x_i \leq \theta \quad \text{e} \quad x_i \leq \theta \quad \forall i$$

$$\iff \theta \geq \max\{x_i, -x_i\} \quad \forall i \iff$$

$$\iff \theta \geq |x_i| \quad \forall i \iff \theta \geq \max\{|x_1|, |x_2|, \dots, |x_n|\}$$

$$\text{Sia } \hat{\theta} := \max\{|x_1|, \dots, |x_n|\}$$

$$(0, +\infty) \ni \vartheta \mapsto L(\vartheta; x_1, \dots, x_n) = \begin{cases} \left(\frac{1}{2\vartheta}\right)^n, & \vartheta \geq \xi \\ 0, & \vartheta < \xi \end{cases}$$



Quindi $\hat{\vartheta} = \max \{ |x_1|, \dots, |x_n| \}$

oss Abbiamo costruito la statistica campionaria

$$g(X_1, \dots, X_n) = \max \{ |X_1|, \dots, |X_n| \}$$

come stima del parametro ϑ .

$$\textcircled{2} E_{\vartheta} [X^k] = \frac{1}{n} \sum_{i=1}^n x_i^k, \quad k=1, 2, \dots$$

$$\boxed{k=1} E_{\vartheta} [X] = \int_{-\infty}^{+\infty} x f_{\vartheta}(x) dx = \int_{-\vartheta}^{\vartheta} x \cdot \frac{1}{2\vartheta} dx = 0$$

$0 = \bar{x}$ NON TROVO SOLUZIONI

$$\begin{aligned} \boxed{k=2} E_{\vartheta} [X^2] &= \int_{-\infty}^{+\infty} x^2 f_{\vartheta}(x) dx = \int_{-\vartheta}^{\vartheta} x^2 \cdot \frac{1}{2\vartheta} dx = \\ &= \frac{1}{2\vartheta} \cdot \frac{x^3}{3} \Big|_{-\vartheta}^{\vartheta} = \frac{\vartheta^3}{3\vartheta} = \frac{\vartheta^2}{3} \end{aligned}$$

$$\frac{\mathcal{J}^2}{3} = \frac{1}{n} \sum_{i=1}^n x_i^2 \quad \Leftrightarrow \quad \mathcal{J} = \left(\frac{3}{n} \sum_{i=1}^n x_i^2 \right)^{1/2}$$

$$\tilde{\mathcal{J}} = \left(\frac{3}{n} \sum_{i=1}^n x_i^2 \right)^{1/2}$$

OSS

1. $E[g(x_1, \dots, x_n)] = \mathcal{J}$?

La statistica campionaria come stima del parametro \mathcal{J} è corretta?

2. $E[(g(x_1, \dots, x_n) - \mathcal{J})^2]$ deve essere il più piccolo possibile

$$E[(g(x_1, \dots, x_n) - \mathcal{J})^2] = \text{Var}(g(x_1, \dots, x_n)) + \underbrace{(E[g(x_1, \dots, x_n)] - \mathcal{J})^2}_{\text{bias}}$$

\bar{X} media campionaria S^2 varianza campionaria

Teorema Sia X_1, \dots, X_n campione statistico di v.e. gaussiane di tipo $\mathcal{N}(\mu, \sigma^2)$. Allora:

(i) \bar{X} e S^2 sono indipendenti ;

(ii) $\bar{X} \sim \mathcal{N}\left(\mu, \frac{\sigma^2}{n}\right)$ e

$$\frac{(n-1)S^2}{\sigma^2} \sim \chi^2_{(n-1)} \left[\begin{array}{l} \text{ha densit\`e chi-quadro} \\ \text{e } (n-1) \text{ gradi di} \\ \text{libert\`e} \end{array} \right]$$

$$(iii) T = \frac{\sqrt{n}(\bar{X} - \mu)}{S} \sim T_{n-1}$$

[ha densit\`e di Student a $(n-1)$ gradi di libert\`e]

$$\underline{\text{OSS}} \quad \frac{\sqrt{n}(\bar{X} - \mu)}{S} = \frac{X_1 + \dots + X_n - n \cdot \mu}{\sqrt{n} \cdot S^2}$$

Def Si chiama densit\`e Gamma di parametri

ν e λ ($\nu > 0, \lambda > 0$) la funzione

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

dove $\Gamma(\nu) := \int_0^{+\infty} t^{\nu-1} e^{-t} dt$.

$$\underline{\text{OSS}} \quad \Gamma(\nu+1) = \nu \Gamma(\nu) \quad \forall \nu > 0,$$

$$\Gamma(n) = (n-1)! \quad \forall n \in \mathbb{N}$$

Prop f \u00e8 una buona densit\`e

dim • $f(x) \geq 0 \quad \forall x \in \mathbb{R}$

• f integrabile

• $\int_{-\infty}^{+\infty} f(x) dx = 1$.

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

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

$$= \frac{1}{\Gamma(r)} \lambda \int_0^{+\infty} (\lambda x)^{r-1} e^{-\lambda x} dx = \frac{\lambda}{\Gamma(r)} \int_0^{+\infty} t^{r-1} e^{-t} \frac{1}{\lambda} dt$$

$$= \frac{1}{\Gamma(r)} \int_0^{+\infty} t^{r-1} e^{-t} dt = 1. \quad \square$$

Una v.a. con densità Gamma di parametri r e λ
e indice con

$$\Gamma(r, \lambda).$$

oss $\Gamma(1, \lambda) \sim$ esponenziale (λ)

$$f(x) = \begin{cases} \frac{1}{\Gamma(1)} \lambda^1 \cdot x^{1-1} e^{-\lambda x} = \lambda e^{-\lambda x}, & x > 0 \\ 0 & , \text{ alt.} \end{cases}$$

Prop • $X \sim \Gamma(\alpha, \lambda)$ ammette tutti i momenti finiti,
 e $E[X^\beta] = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha) \lambda^\beta} \quad \forall \beta > 0$

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

$\text{Var}(X) = \frac{\alpha}{\lambda^2}$

• $X \sim \Gamma(\alpha, \lambda)$ ha funzione generatrice dei momenti

$$\varphi_X(t) = \left(\frac{\lambda}{\lambda - t} \right)^\alpha \quad \forall t \in (-\infty, \lambda)$$

dim $\varphi_X(t) = E[e^{tX}] = \int_{-\infty}^{+\infty} e^{tx} f(x) dx =$

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

$$= \int_0^{+\infty} \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{x(t-\lambda)} dx = \quad (\text{per } t < \lambda \Rightarrow t - \lambda < 0)$$

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

$$= \frac{\lambda^\alpha}{\Gamma(\alpha) (\lambda - t)^{\alpha-1}} \int_0^{+\infty} [(\lambda - t)x]^{\alpha-1} e^{-(\lambda - t)x} dx \stackrel{y = (\lambda - t)x}{=} \downarrow$$

$$= \frac{\lambda^r}{\Gamma(r)(\lambda-t)^{r-1}} \int_0^{+\infty} y^{r-1} e^{-y} \frac{dy}{\lambda-t} = \frac{\lambda^r}{(\lambda-t)^r} \quad \square$$

Prop Se $X \sim \Gamma(r, \lambda)$ e $Y \sim \Gamma(s, \lambda)$, e sono indipendenti, allora $X+Y \sim \Gamma(r+s, \lambda)$.

dim X, Y indep $\Rightarrow P_{X+Y}(t) = P_X(t) \cdot P_Y(t)$

Quindi $P_{X+Y}(t) = \left(\frac{\lambda}{\lambda-t}\right)^r \cdot \left(\frac{\lambda}{\lambda-t}\right)^s = \left(\frac{\lambda}{\lambda-t}\right)^{r+s}$

$\Rightarrow X+Y$ ha distribuzione $\Gamma(r+s, \lambda)$. \square