

ESERCITAZIONE 2

- 1) Data $X \sim N(\theta, 1)$ θ è un parametro di posizione?
Come verifico se θ è parametro di posizione?

1. Considero $f_X(x, \theta) = \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{1}{2}(x-\theta)^2\right\}$

2. Pongo $\theta = 0$ e calcolo $f_X(x, \theta=0) = \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{1}{2}x^2\right\} = f_X(x)$

3. Calcolo $g_P(x, \theta) = f(x-\theta) = \frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{1}{2}(x-\theta)^2\right\}$

4. Nel punto 3 ho riattento la densità del punto 1 $\Rightarrow \theta$ è parametro di posizione

Data $X \sim \text{Exp}(\theta)$ θ è un parametro di scala?

1. $f_X(x, \theta) = \frac{1}{\theta} e^{-x/\theta}$

2. Pongo $\theta = 1 \Rightarrow f_X(x, \theta=1) = e^{-x} = f_X(x)$ $\rightarrow$ f-densità standard

3. Calcolo $g_S(x, \theta) = \frac{1}{\theta} f\left(\frac{x}{\theta}\right) = \frac{1}{\theta} e^{-x/\theta}$

4. Facile ho riattento la densità del punto 1 $\Rightarrow \theta$ è parametro di scala.

- 2) Dato $\underline{X}_n = (X_1, \dots, X_n)$ t.c. X_i iid $i=1, \dots, n$ e con $f_X(x, \theta) = \frac{\theta e^{-\theta x}}{x^{\theta+1}}$, $x > 0$, $\theta > 0$

a) Scrivere $L(\theta, \underline{x}_n)$

b) Verifica che $\hat{\theta}_{ML} = \frac{n}{\sum_{i=1}^n \ln x_i - n}$

$$\begin{aligned} \text{a) } L(\theta, \underline{x}_n) &= f_n(\underline{x}_n, \theta) = \prod_{i=1}^n f_X(x_i, \theta) = \prod_{i=1}^n \left(\frac{\theta e^{-\theta x_i}}{x_i^{\theta+1}} \mathbb{1}_{(0, \infty)}(x_i) \right) \\ &= \frac{\theta^m e^{m\theta}}{\left(\prod_{i=1}^m x_i\right)^{\theta+1}} \prod_{i=1}^m \mathbb{1}_{(0, \infty)}(x_i) = \underbrace{\prod_{i=1}^m \left(\frac{\mathbb{1}_{(0, \infty)}(x_i)}{x_i} \right)}_{h(\underline{x}_n)} \underbrace{\frac{\theta^m e^{m\theta}}{\left(\prod_{i=1}^m x_i\right)^{\theta}}}_{g(\theta, \underline{x}_n)} \propto \frac{\theta^m e^{m\theta}}{\left(\prod_{i=1}^m x_i\right)^{\theta}} \end{aligned}$$

b) Osserviamo che $\mathcal{X} = (e, +\infty)$ che non dipende da $\theta \Rightarrow$ problema di stima regolare.

$$\ell(\theta, \underline{x}) = \log L(\theta, \underline{x}) = \log(h(\underline{x}) g(\theta, \underline{x})) = \log(h(\underline{x})) + \log(g(\theta, \underline{x}))$$

$$\Rightarrow \frac{d}{d\theta} \ell(\theta, \underline{x}) = \frac{d}{d\theta} (\log(h(\underline{x})) + \log(g(\theta, \underline{x}))) = \frac{d}{d\theta} \log(g(\theta, \underline{x})) \quad \text{perché}$$

$$\ell(\theta, \underline{x}) = \log \left(\frac{\theta^m e^{m\theta}}{(\prod_{i=1}^m x_i)^{\theta+1}} \prod_{i=1}^m \frac{1(x_i)}{(e, +\infty)} \right) = \log(\theta^m) + m\theta - (\theta+1) \log \prod_{i=1}^m x_i + \log \prod_{i=1}^m 1(x_i)$$

$$= m \log(\theta) + m\theta - (\theta+1) \sum_{i=1}^m \log x_i + \log \prod_{i=1}^m 1(x_i)$$

$$\frac{d}{d\theta} \ell(\theta, \underline{x}) = \frac{m}{\theta} + m - \sum_{i=1}^m \log x_i = 0$$

$$\Rightarrow m + m\theta - \theta \sum_{i=1}^m \log x_i = 0 \Rightarrow \theta \left(\sum_{i=1}^m \log x_i - m \right) = m$$

$$\Rightarrow \hat{\theta} = \frac{m}{\sum_{i=1}^m \log x_i - m} \quad \rightarrow \text{candidato}$$

$$\frac{d^2}{d\theta^2} \ell(\theta, \underline{x}) = -\frac{m}{\theta^2} < 0 \quad \forall \theta \in \Theta \Rightarrow \left. \frac{d^2}{d\theta^2} \ell(\theta, \underline{x}) \right|_{\theta=\hat{\theta}} < 0$$

$\Rightarrow \hat{\theta} = \hat{\theta}_{MV}$ è stimatore di max verosimiglianza.

3) Dato $\underline{X}_n = (X_1, \dots, X_n)$ con X_i iid $i=1, \dots, n$ e $f_X(x, \theta) = \frac{2x}{\theta^2} e^{-\frac{x^2}{\theta^2}}$ con $x > 0$ e $\theta > 0$.

a) Modello statistico per $\underline{X}_n = (X_1, \dots, X_n)$

$(\mathcal{X}^n = \mathbb{R}^+, f_n(\underline{x}_n, \theta), \Theta = \mathbb{R}^+)$

$$f_n(\underline{x}_n, \theta) = \prod_{i=1}^n f_X(x_i, \theta) = \prod_{i=1}^n \left(\frac{2x_i}{\theta^2} e^{-\frac{x_i^2}{\theta^2}} \right) = \left(\prod_{i=1}^n x_i \right) \left(\frac{2}{\theta^2} \right)^n e^{-\frac{\sum_{i=1}^n x_i^2}{\theta^2}}$$

b) Determina $L(\theta, \underline{x}_n)$, il suo nucleo e $\hat{\theta}_{MV}$ per un generico campione osservato $\underline{x}_n$.

$$L(\theta, \underline{x}_n) = \prod_{i=1}^m f_n(x_i, \theta) = \left(\prod_{i=1}^m x_i \right) \left(\frac{2}{\theta^2} \right)^m e^{-\frac{\sum_{i=1}^m x_i^2}{\theta^2}} = \underbrace{\left(\prod_{i=1}^m 2x_i \right)}_{h(\underline{x}_n)} \underbrace{\theta^{-2m} e^{-\frac{\sum_{i=1}^m x_i^2}{\theta^2}}}_{g(\theta, \underline{x}_n) \text{ nucleo.}}$$

Per calcolare $\hat{\theta}_{MV}$ posso concentrarmi solo sul nucleo della f.d.v. $\Rightarrow$

$$\begin{aligned} \ell(\theta, \underline{x}_n) &= \log L(\theta, \underline{x}_n) \propto \log \left(\theta^{-2m} e^{-\frac{\sum_{i=1}^m x_i^2}{\theta^2}} \right) = \\ &= -2m \log \theta - \frac{\sum_{i=1}^m x_i^2}{\theta^2} \end{aligned}$$

$$\frac{d}{d\theta} \ell(\theta, \underline{x}_n) = -\frac{2m}{\theta} + 2 \frac{\sum_{i=1}^m x_i^2}{\theta^3} = 0 \Rightarrow \hat{\theta} = \sqrt{\frac{\sum_{i=1}^m x_i^2}{m}}$$

$$\left. \frac{d^2}{d\theta^2} \ell(\theta, \underline{x}_n) \right|_{\theta=\hat{\theta}} = \left(\frac{2m}{\theta^2} - \frac{6 \sum_{i=1}^m x_i^2}{\theta^4} \right) \Big|_{\theta=\hat{\theta}} = \left(\frac{2m^2}{\sum_{i=1}^m x_i^2} - \frac{6m^2}{\sum_{i=1}^m x_i^2} \right) < 0$$

quindi $\hat{\theta} = \hat{\theta}_{MV} = \sqrt{\frac{\sum_{i=1}^m x_i^2}{m}}$

c) Calcolare l'informazione di Fisher $I_m^{on}(\hat{\theta}_{MV})$

$$I_m^{on}(\hat{\theta}_{MV}) = -\frac{d^2}{d\theta^2} \ell(\hat{\theta}_{MV}, \underline{x}_n) = \frac{4m^2}{\sum_{i=1}^m x_i^2}$$

d) Scrivere l'approssimazione normale della f.d.v.

$$\tilde{L}(\theta, \underline{x}_n) = \exp \left\{ -\frac{1}{2} (\theta - \hat{\theta}_{MV})^2 I_n(\hat{\theta}_{MV}) \right\} = \exp \left\{ -\frac{1}{2} \left(\theta - \sqrt{\frac{\sum_{i=1}^m x_i^2}{m}} \right)^2 \frac{4m^2}{\sum_{i=1}^m x_i^2} \right\}$$

4) Dato $\underline{X}_n = (X_1, \dots, X_n)$ con $X_i \ i=1, \dots, n$ iid

$$f_X(x, \theta) = 2\theta x e^{-\theta x^2} \mathbb{1}_{(0, +\infty)}(x) \quad \text{con } \theta > 0$$

a) Scrivere la f.d.v. $L(\theta, \underline{x}_n)$

$$L(\theta, \underline{x}_n) = \prod_{i=1}^n \left(2\theta x_i e^{-\theta x_i^2} \mathbb{1}_{(0, +\infty)}(x_i) \right) = \left(\prod_{i=1}^n x_i \mathbb{1}_{(0, +\infty)}(x_i) \right) (2\theta)^n e^{-\theta \sum_{i=1}^n x_i^2}$$

b) Individua il nucleo di $L(\theta, \underline{x}_n)$ e una statistica sufficiente per il modello

$$L(\theta, \underline{x}_n) = \underbrace{\left(\prod_{i=1}^n x_i \mathbb{1}_{(0, +\infty)}(x_i) \right)}_{h(\underline{x}_n)} \underbrace{\theta^n e^{-\theta \sum_{i=1}^n x_i^2}}_{g(\theta, \underline{x}_n)}$$

Per criterio di fattorizzazione di Fisher $T(\underline{x}_n) = \sum_{i=1}^n x_i^2$ è statistica sufficiente

c) Calcola $\hat{\theta}_{MV}$

$$\ell(\theta, \underline{x}_n) \propto \log(g(\theta, \underline{x}_n)) = n \log \theta - \theta \sum_{i=1}^n x_i^2$$

$$\frac{d}{d\theta} \ell(\theta, \underline{x}_n) = \frac{n}{\theta} - \sum_{i=1}^n x_i^2 = 0 \Rightarrow \hat{\theta} = \frac{n}{\sum_{i=1}^n x_i^2}$$

$$\frac{d^2}{d\theta^2} \ell(\theta, \underline{x}_n) = -\frac{n}{\theta^2} < 0 \quad \forall \theta \in \mathbb{D} \Rightarrow \hat{\theta}_{MV} = \hat{\theta} = \frac{n}{\sum_{i=1}^n x_i^2}$$

5) $X_n = (X_1, \dots, X_m)$ con X_i iid $i=1, \dots, m$

$$f_X(x, \theta) = \frac{2x}{\theta^2} \mathbb{1}_{(0, \theta)} \quad \text{con } \theta > 0$$

a) Calcolare $\mathbb{E}_\theta[\bar{X}_n]$ da $\bar{X}_n = \frac{1}{m} \sum_{i=1}^m X_i$

$$\mathbb{E}_\theta[\bar{X}_n] = \mathbb{E}_\theta\left[\frac{1}{m} \sum_{i=1}^m X_i\right] = \frac{1}{m} \sum_{i=1}^m \mathbb{E}[X_i] = \mathbb{E}[X_1]$$

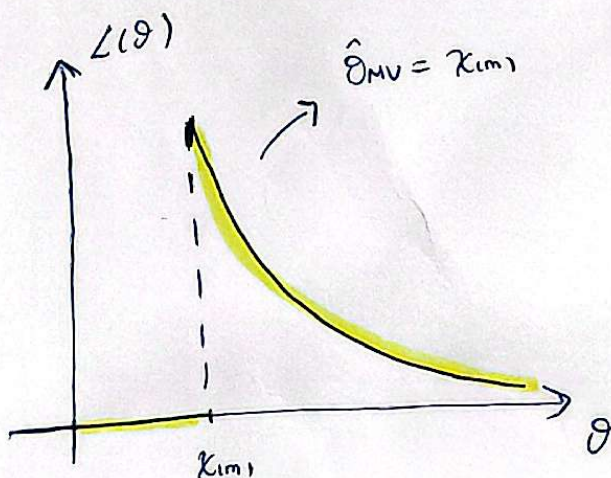
$$= \int_0^\theta \frac{2x^2}{\theta^2} dx = \frac{2}{\theta^2} \left[\frac{x^3}{3} \right]_0^\theta = \frac{2}{\theta^2} \frac{\theta^3}{3} = \frac{2}{3} \theta$$

b) Calcolare $L(\theta, x_n)$ e $\hat{\theta}_{MV}$

$$L(\theta, x_n) = \prod_{i=1}^m \left(\frac{2x_i}{\theta^2} \mathbb{1}_{(0, \theta)} \right) = \left(\prod_{i=1}^m 2x_i \right) \theta^{-2m} \prod_{i=1}^m \mathbb{1}_{(0, \theta)} \propto \theta^{-2m} \mathbb{1}_{(x_{(m)}, +\infty)}$$

$$\prod_{i=1}^m \mathbb{1}_{(0, \theta)} = \begin{cases} 1 & \text{se } 0 < x_i < \theta \quad \forall i=1, \dots, m \\ 0 & \text{altrimenti} \end{cases} = \begin{cases} 1 & \text{se } 0 < x_{(1)} < \dots < x_{(m)} < \theta \\ 0 & \text{altrimenti} \end{cases}$$

$$= \mathbb{1}_{(x_{(m)}, +\infty)}$$



6) Data $f_X(x, \theta) = a \frac{x^2}{\theta^3} \mathbb{1}_{[0, 2\theta]}^{(x)}$ con $\theta > 0$

i) Determina a in modo che f_X sia una densità

Deve valere che $\int_X f_X(x, \theta) dx = 1 \implies \int_X a \frac{x^2}{\theta^3} \mathbb{1}_{[0, 2\theta]}^{(x)}$

$$= \frac{a}{\theta^3} \int_0^{2\theta} x^2 dx = \frac{a}{\theta^3} \left[\frac{x^3}{3} \right]_0^{2\theta} = \frac{a}{\theta^3} \left[\frac{8\theta^3}{3} \right] = \frac{8}{3} a = 1$$

impongo l'uguaglianza

$$\implies a = 3/8$$

ii) Calcola $\hat{\theta}_{MV}$ data $f_X(x, \theta) = \frac{3}{8} \frac{x^2}{\theta^3} \mathbb{1}_{[0, 2\theta]}^{(x)}$

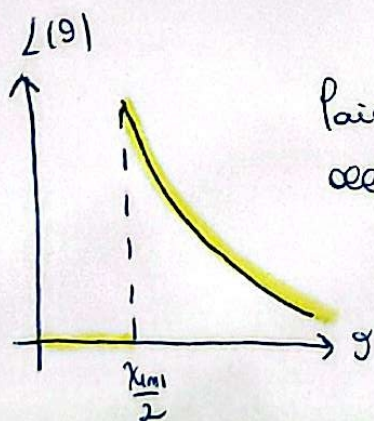
$$L(\theta, \underline{x}_n) = \prod_{i=1}^n \left(\frac{3}{8} \frac{x_i^2}{\theta^3} \mathbb{1}_{[0, 2\theta]}^{(x_i)} \right) = \left(\frac{3}{8} \right)^n \frac{\prod_{i=1}^n x_i^2}{\theta^{3n}} \prod_{i=1}^n \mathbb{1}_{[0, 2\theta]}^{(x_i)}$$

Si tratta di un problema di stima non regolare (supporto $[0, 2\theta]$ dipende da θ)

$$\prod_{i=1}^n \mathbb{1}_{[0, 2\theta]}^{(x_i)} = \begin{cases} 1 & 0 \leq x_i \leq 2\theta \quad \forall i=1, \dots, n \\ 0 & \text{altrimenti} \end{cases} = \begin{cases} 1 & \theta \geq x_{(n)}/2 \quad \forall i \\ 0 & \text{altrimenti} \end{cases}$$

$$= \begin{cases} 1 & \frac{x_{(n)}}{2} \leq \theta \\ 0 & \text{altrimenti} \end{cases} = \mathbb{1}_{\left[\frac{x_{(n)}}{2}, +\infty \right)}^{(\theta)}$$

$$L(\theta, \underline{x}_n) = \left(\frac{3}{8} \right)^n \frac{\prod_{i=1}^n x_i^2}{\theta^{3n}} \mathbb{1}_{\left[\frac{x_{(n)}}{2}, +\infty \right)}^{(\theta)} \propto \theta^{-3n} \mathbb{1}_{\left[\frac{x_{(n)}}{2}, +\infty \right)}^{(\theta)}$$



Poiché la $L(\theta)$ è decrescente di θ nell'insieme $\left[\frac{x_{(n)}}{2}, +\infty \right)$ allora $\hat{\theta}_{MV}(\underline{x}_n) = \frac{x_{(n)}}{2}$