

ESERCITAZIONE 4

1) $X_1 \sim N(0,1)$
 $X_2 \sim N(0,1)$ indipendenti tra loro. Determina le distribuzioni delle seguenti v.a.

i) $U = (X_1 - X_2) \sim ?$

$$E[U] = E[X_1 - X_2] = E[X_1] - E[X_2] = 0 - 0 = 0 \Rightarrow U \sim N(0,2)$$

$$V[U] = V(X_1 - X_2) = V(X_1) + V(X_2) = 1 + 1 = 2$$

ii) $Z = \frac{U}{\sqrt{2}} \sim ?$

$$E[Z] = E[U/\sqrt{2}] = \frac{1}{\sqrt{2}} E[U] = \frac{1}{\sqrt{2}} \cdot 0 = 0 \Rightarrow Z = \frac{U}{\sqrt{2}} \sim N(0,1)$$

$$V[Z] = V[U/\sqrt{2}] = \frac{1}{2} V[U] = \frac{1}{2} \cdot 2 = 1$$

iii) $Z^2 = \frac{U^2}{2} = ? \rightarrow Z^2 = N(0,1)^2 = \chi^2_1$
 $= \frac{U^2}{2}$

2) $X_n = (X_1, \dots, X_n)$ t.c. $X_i | \mu, \sigma^2 \sim N(\mu, \sigma^2) \quad \forall i=1, \dots, n$

Determinare $E[H_n]$ e $V[H_n]$ essendo $H_n = a S_n^2 + b \bar{X}_n^2$ con $a, b \in \mathbb{R}$

$$E[H_n] = E[a S_n^2 + b \bar{X}_n^2] = a E[S_n^2] + b E[\bar{X}_n^2] = a V[X] + b E[\bar{X}_n^2]$$

$$= a \sigma^2 + b (V[\bar{X}_n] + E[\bar{X}_n]^2) \quad \text{poiché } V[X] = E[X^2] - E[X]^2$$

$$= a \sigma^2 + b \left(\frac{V[X]}{n} + E[X]^2 \right) \quad E[X^2] = V[X] + E[X]^2$$

$$= a \sigma^2 + b \left(\frac{\sigma^2}{n} + \mu^2 \right)$$

→ questo poiché per il caso normale S_n^2 e $\bar{X}_n$ sono indipendenti.

$$V[H_n] = V[a S_n^2 + b \bar{X}_n^2] = a^2 V[S_n^2] + b^2 V[\bar{X}_n^2]$$

$$\frac{(n-1)}{\sigma^2} S_n^2 = \frac{1}{\sigma^2} \sum_{i=1}^n (X_i - \bar{X}_n)^2 \sim \chi^2_{n-1} = \text{Gamma}\left(\frac{n-1}{2}, \frac{1}{2}\right)$$

$$\Rightarrow V\left[\frac{(n-1) S_n^2}{\sigma^2}\right] = \frac{n-1}{2} \quad \text{poiché } V[\chi^2_k] = 2k \Rightarrow \frac{(n-1)^2}{\sigma^4} V[S_n^2] = 2(n-1)$$

→ Varianza dello $\chi^2\left(\frac{n-1}{2}, \frac{1}{2}\right)$

$$V[S_n^2] = \frac{2\sigma^4}{n-1}$$

Si suppone che $\mu = 0 \Rightarrow \bar{X}_n \sim N(\mu = 0, \sigma^2/m) = N(0, \sigma^2/m)$

$$\bar{X}_n^2 = \left(\frac{\bar{X}_n \sqrt{m}}{\sigma} \right)^2 = \frac{\sigma^2}{m} \left(\frac{\bar{X}_n \sqrt{m}}{\sigma} \right)^2 = \frac{\sigma^2}{m} Z^2 \quad \text{con } Z = \frac{\bar{X}_n \sqrt{m}}{\sigma} \sim N(0,1)$$

quindi $\bar{X}_n^2 \sim \frac{\sigma^2}{m} Z^2 = \frac{\sigma^2}{m} \chi_1^2 \sim \text{Gamma}\left(\frac{1}{2}, \frac{m}{2\sigma^2}\right)$

quindi $Z^2 \sim \chi_1^2 = \text{Gamma}\left(\frac{1}{2}, \frac{1}{2}\right)$

$$V(\bar{X}_n^2) = \frac{1}{2} \left(\frac{2\sigma^2}{m} \right)^2 = \frac{2\sigma^4}{m^2}$$

otteniamo quindi $V(\bar{X}_n) = a^2 \frac{2\sigma^4}{m-1} + b^2 \frac{2\sigma^4}{m^2}$

3) $X: \theta$ iid $i=1, \dots, n$ $f_X(x, \theta) = 2\theta^2 x \mathbb{1}_{[0, 1/\theta]}$ con $\theta > 0$. Determina $\mathbb{E}_\theta[X]$ e lo stimatore dei momenti di θ .

$$\mathbb{E}_\theta[X] = \int_0^{1/\theta} 2\theta^2 x^2 dx = 2\theta^2 \left[\frac{x^3}{3} \right]_0^{1/\theta} = 2\theta^2 \frac{1}{3\theta^3} = \frac{2}{3} \frac{1}{\theta}$$

$$\mu_1(\theta) = m_1(\bar{X}_n) \Rightarrow \mathbb{E}_\theta[X] = \bar{X}_n \Rightarrow \frac{2}{3} \frac{1}{\theta} = \bar{X}_n \Rightarrow \hat{\theta}_n = \frac{2}{3\bar{X}_n}$$

4) $X: \theta \sim \text{Unif}[-\theta, \theta]$ con $\theta > 0$. Determina stimatore dei momenti di θ .

$$f(x, \theta) = \frac{1}{2\theta} \mathbb{1}_{[-\theta, \theta]} \Rightarrow \mathbb{E}[X] = \frac{(\theta - \theta)^2}{2} = 0 \Rightarrow \mathbb{E}[X] = \bar{X}_n = 0 = \bar{X}_n$$

utilizziamo il momento secondo $\mathbb{E}[X^2] = \frac{1}{n} \sum X_i^2$

$$\mathbb{E}[X^2] = V[X] + \mathbb{E}[X]^2 = \frac{4\theta^2}{12} + 0 = \frac{4\theta^2}{12} = \frac{1}{3} \sum X_i^2 \Rightarrow \hat{\theta}_{\text{mom}} = \sqrt{\frac{3}{n} \sum X_i^2}$$

5) $X: \theta \sim \text{Ga}^R(\theta, 2)$ $\theta > 0$ calcola $\hat{\theta}_{\text{mom}}$ e il suo valore atteso e varianza e MSE.

i) $\mu_1(\theta) = \mathbb{E}_\theta[X] = \frac{\theta}{2}$; $m_1(\bar{X}_n) = \frac{1}{n} \sum X_i = \bar{X}_n \Rightarrow \frac{\theta}{2} = \bar{X}_n \Rightarrow \hat{\theta}_{\text{mom}} = 2\bar{X}_n$

ii) $\mathbb{E}[\hat{\theta}_{\text{mom}}] = \mathbb{E}[2\bar{X}_n] = 2\mathbb{E}[\bar{X}_n] = 2\mathbb{E}[X] = 2 \frac{\theta}{2} = \theta \quad \forall \theta$

$$V[\hat{\theta}_{\text{mom}}] = V[2\bar{X}_n] = 4V[\bar{X}_n] = 4 \frac{V[X]}{n} = 4 \frac{\theta}{4n} = \frac{\theta}{n}$$

$$\text{MSE}(\hat{\theta}_{\text{mom}}) = V_0[\hat{\theta}_{\text{mom}}] + B_0^2(\hat{\theta}_{\text{mom}}) = V_0[\hat{\theta}_{\text{mom}}] = \theta/n \quad \forall \theta$$

6) $X_1, \dots, X_n \sim \text{Pois}(\theta)$ iid

Calcolare lo snv di $g(\theta) = P(X=0; \theta)$ quando $\sum x_i = 6$ e $n=2$

$$f_X(x; \theta) = \frac{e^{-\theta} \theta^x}{x!}$$

$$l(\theta, x_1) \propto e^{-n\theta} \theta^{\sum x_i}$$

$$l(\theta, x_1) \propto -n\theta + \sum x_i \log \theta$$

$$\frac{d}{d\theta} l(\theta) = -n + \frac{\sum x_i}{\theta} = 0 \Rightarrow \hat{\theta}_{nv} = \frac{\sum x_i}{n} = \bar{x}_n$$

Per la proprietà di equivanza dello snv $\lambda = g(\theta) \Rightarrow \hat{\lambda}_{nv} = g(\hat{\theta}_{nv})$

quindi $\hat{\lambda}_{nv} = g(\hat{\theta}_{nv}) = g(\bar{x}_n) = P(X=0; \bar{x}_n)$

$$= \frac{e^{-\hat{\theta}_{nv}} \hat{\theta}_{nv}^0}{0!} = e^{-\hat{\theta}_{nv}} = e^{-\bar{x}_n} = e^{-\frac{6}{2}} = e^{-3}$$

7) Sia X v.v. che assume valori $(0, 1, 2)$ con probabilità $(1-\frac{\theta}{2}, \frac{\theta}{4}, \frac{\theta}{4})$ determina $E[X]$ e lo stimatore dei momenti di θ .

$$E[X] = 0 \cdot (1-\frac{\theta}{2}) + 1 \cdot \frac{\theta}{4} + 2 \cdot \frac{\theta}{4} = \frac{\theta}{4} + 2 \frac{\theta}{4} = \frac{3}{4} \theta$$

$$E[X] = \frac{3}{4} \theta = \bar{X}_n \Rightarrow \hat{\theta}_{mom} = \frac{4}{3} \bar{X}_n$$