

# Consultatie

7.02.2023

① Tre  $X_1, X_2, \dots, X_n$  ezantre de wol n di  
prop  $\text{Poi}(\theta)$   
$$P(X=k) = \frac{\theta^k}{(1+\theta)^{k+1}} \quad \left( \frac{A \theta^k}{(1+\theta)^k} \right)$$

$\theta > 0$

Def. estimatori os. pri met momenten  
o met ver. maxime g' studentij prop.  
acutro.

Sol: Met momenten:

$$\bar{X}_n = E_0[X_1] \rightarrow g(\theta)$$

$$\tilde{\theta}_n = h(\bar{X}_n)$$

$$E_0[X_1] = \sum_{k \geq 0} k P_0(X_1 = k)$$

$$\sum_k k q^{k-1} \quad (q^k)'$$

$$= \sum_{k \geq 0} \textcircled{k} \frac{\theta^k}{(1+\theta)^{k+1}} = \frac{\theta}{(1+\theta)^2} \sum_{k \geq 1} k \left( \frac{\theta}{1+\theta} \right)^{k-1}$$

$$= \frac{\theta}{(1+\theta)^2} \sum_{k \geq 1} k q^{k-1} \quad \text{unde } q = \frac{\theta}{1+\theta}$$

$$E_{\theta}[X_i] = \frac{\theta}{(1+\theta)^2} \sum_{k \geq 0} (k+1) 2^k$$

$$= \frac{\theta}{(1+\theta)^2} \sum_{k \geq 0} (2^{k+1})' \rightarrow \frac{d}{dz} 2^{k+1}$$

$$= \frac{\theta}{(1+\theta)^2} \frac{d}{dz} \left( \sum_{k \geq 0} 2^{k+1} \right)$$

$$\left( \sum_{k \geq 0} 2^{k+1} \right)' = \frac{d}{dz} \left( \frac{1}{1-2} - 1 \right)$$

$$f'(g) \cdot g'$$

$$= \frac{\theta}{(1+\theta)^2} \frac{d}{dz} \left( \frac{2}{1-2} \right) = \frac{\theta}{(1+\theta)^2} \frac{1}{(1-2)^2}$$

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

$$\bar{X}_n = \theta \Rightarrow \left[ \hat{\theta}_n = \bar{X}_n \right] \text{ est. estimant pi mesure}$$

Théor. ver. maximale :

$$L_n(\theta; x) = \prod_{i=1}^n f_{\theta}(x_i) = \prod_{i=1}^n \frac{\theta^{x_i}}{(1+\theta)^{x_i+1}}$$

$$= \frac{\theta^{\sum x_i}}{(1+\theta)^{n+\sum x_i}}$$

$$\begin{aligned} \ell_n(\theta; x) &= \log L_n(\theta; x) \\ &= \left( \sum_{i=1}^n x_i \right) \log(\theta) - \left( n + \sum_{i=1}^n x_i \right) \log(1+\theta) \end{aligned}$$

Ex. derivăm:

$$\frac{\partial \ell_n(\theta; x)}{\partial \theta} = \frac{\sum_{i=1}^n x_i}{\theta} - \frac{n + \sum_{i=1}^n x_i}{1+\theta} = 0$$

$$(1) \quad \frac{\sum x_i}{\theta} = \frac{n + \sum x_i}{1+\theta} \quad (2) \quad \theta = \frac{\sum x_i}{n}$$

$\hat{\theta}_n = \bar{X}_n$  est. derivăm în pct. maxim

Prop: 1) lipsirea

$$\mathbb{E}_\theta[X_1] = \mathbb{E}_\theta[X_i] = \theta \Rightarrow \hat{\theta}_n \text{ este un estimator nedegradat}$$

2) constanta

$$LNM: \quad \bar{X}_n \xrightarrow{a.s.} \mathbb{E}[X_1] = \theta \Rightarrow \hat{\theta}_n \text{ este constant}$$

3) normalitatea asimptotică

$$TLC: \quad \sqrt{n}(\bar{X}_n - \theta) \xrightarrow{d} N(0, \underline{\text{Var}}_2(X_1))$$

$$\text{Var}_2(X_1) = \underbrace{\mathbb{E}[X_1^2]}_{?} - \underbrace{\mathbb{E}[X_1]^2}_{=\theta^2}$$

$$\underline{\mathbb{E}_\theta[X_1^2]} = \sum_{k \geq 0} k^2 \mathbb{P}_\theta(X_1 = k) = \sum_{k \geq 0} k^2 \frac{\theta^k}{(1+\theta)^{k+1}}$$

$$= \sum_{k \geq 1} k^2 \frac{\theta^k}{(1+\theta)^{k+1}} = \frac{\theta}{(1+\theta)^2} \sum_{k \geq 1} k^2 \left(\frac{\theta}{1+\theta}\right)^{k-1}$$

$$= \frac{\theta}{(1+\theta)^2} \sum_{k \geq 0} (k+1)^2 \underset{\downarrow q}{q^k} \quad \left( \underbrace{(k+1)(k+1) - (k+1)}_k \right)$$

$$= \frac{\theta}{(1+\theta)^2} \sum_{k \geq 0} (k^2 + 2k + 1) q^k$$

$$= \frac{\theta}{(1+\theta)^2} \left[ \sum_{k \geq 0} k^2 q^k + 2 \sum_{k \geq 0} k q^k + \sum_{k \geq 0} q^k \right]$$

$$= \frac{\theta}{(1+\theta)^2} \left[ \underbrace{\sum_{k \geq 0} k^2 q^k}_{(\theta+1)\mathbb{E}_\theta[X_1^2]} + 2 \underbrace{\sum_{k \geq 0} (k+1) q^k}_{\left(\sum_{k \geq 0} q^{k+1}\right)'} - \underbrace{\sum_{k \geq 0} q^k}_{\frac{q}{1-q}} \right]$$

$$= \frac{\theta}{(1+\theta)^2} \left[ (\theta+1) \mathbb{E}_\theta[X_1^2] + 2 \left( \frac{q}{1-q} \right)' - \frac{1}{1-q} \right]$$

$$= \frac{\theta}{(1+\theta)^2} \left[ (\theta+1) \underline{\mathbb{E}_\theta[X_1^2]} + 2(1+\theta)^2 - (\theta+1) \right]$$

$$E[X^2] = \frac{\theta}{1+\theta} E[X^2] + 2\theta - \frac{\theta}{1+\theta}$$

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

$$\Rightarrow E[X^2] = 2\theta^2 + \theta \Rightarrow \text{Var}(X) = \theta^2 + \theta = \theta(\theta+1)$$

$$\sqrt{n}(\hat{\theta}_n - \theta) \xrightarrow{d} N(0, \theta(\theta+1))$$

$$\begin{aligned} \text{MSE}(\hat{\theta}_n) &= \text{Var}(\hat{\theta}_n) + \underbrace{\text{Bias}(\hat{\theta}_n)^2}_{=0} = \text{Var}(\hat{\theta}_n) \\ &= \frac{\theta(1+\theta)}{n} \xrightarrow{n \rightarrow \infty} 0 \end{aligned}$$

Ex 2)  $X_1, \dots, X_n \sim \text{Exp}(\theta), \theta > 0$

1) Est. obținut prin metode momentelor

2) EVM în proporție

3)  $T_n = \frac{n-1}{n} \hat{\theta}_n$ , unde  $\hat{\theta}_n$  este est. de ver. max

Comparăm cei 3 estimatori

$$1) \quad \mathbb{E}_{\theta}(\theta)$$

$$f_{\theta}(x) = \theta x^{-\theta-1}, \quad \theta > 0, \quad x > 0$$

$$\bar{X}_n = \mathbb{E}_{\theta}(X_1) \quad \left. \begin{array}{l} \text{oder } \mathbb{E}_{\theta}(X_1) = 1/\theta \end{array} \right\} \Rightarrow \boxed{\tilde{\theta}_n = \frac{1}{\bar{X}_n}} \quad \text{est. oft mit} \\ \text{mitteln}$$

$$\text{Ist lineal funkt } \tilde{\theta}_n? \quad \mathbb{P}_{\theta}(\bar{X}_n > 0) = ?$$

$$\mathbb{P}_{\theta}(\bar{X}_n > 0) = \mathbb{P}_{\theta}(X_1 > 0) = 1 \quad \text{pt ca } \mathbb{P}_{\theta}(X_1 > 0) = 1$$

$$\mathbb{P}_{\theta}(\bar{X}_n > 0) = 1 \quad \text{dies ist lineal funkt.}$$

2) EVM

$$L_n(\theta; x) = \prod_{i=1}^n \theta e^{-\theta x_i} = \theta^n e^{-\theta \sum x_i}$$

$$l_n(\theta; x) = n \log \theta - \theta \sum x_i \Rightarrow \frac{\partial l}{\partial \theta} = 0 \quad \Rightarrow \quad \boxed{\tilde{\theta}_n = \frac{1}{\bar{X}_n}}$$

Prop: a) konsistent

$$\text{LNM: } \bar{X}_n \xrightarrow{\text{a.s.}} \mathbb{E}(X) = \frac{1}{\theta}$$

$$\hat{\theta}_n = g(\bar{X}_n), \quad g(x) = \frac{1}{x}$$

Th. Appl.  
Cont

$$g(\bar{X}_n) \xrightarrow{\text{a.s.}} g(1/\theta) = \theta \Rightarrow \underline{\tilde{\theta}_n \xrightarrow{\text{a.s.}} \theta}$$

6) Normalisierung asymptotisch

$$\text{ZCL: } \sqrt{n} \left( \bar{X}_n - \frac{1}{\theta} \right) \xrightarrow{d} N(0, \text{Var}_{\theta}(X_1))$$

$$\sqrt{n}(\bar{X}_n - 1/\theta) \xrightarrow{d} N(0, \frac{1}{\theta^2})$$

$$\hat{\theta}_n = \frac{1}{\bar{X}_n} \approx g(\bar{X}_n)$$

Méthode Delta ( $g$  dér.)

$$\sqrt{n}(g(\bar{X}_n) - g(1/\theta)) \xrightarrow{d} g'(1/\theta) N(0, \frac{1}{\theta^2})$$

$$\sqrt{n}(\hat{\theta}_n - \theta) \xrightarrow{d} N(0, g'(1/\theta)^2 \frac{1}{\theta^2})$$

$$g'(1/\theta) = -\theta^2 \Rightarrow \sqrt{n}(\hat{\theta}_n - \theta) \xrightarrow{d} N(0, \frac{\theta^4}{\theta^2})$$

$= N(0, \theta^2)$

3) Détermination

$$E[\hat{\theta}_n] = E\left[\frac{1}{\bar{X}_n}\right] = E\left[\frac{1}{\frac{1}{n} \sum_{i=1}^n X_i}\right] = n E\left[\frac{1}{\sum_{i=1}^n X_i}\right]$$

$X_1, X_2, \dots, X_n \sim \text{Exp}(\theta)$  indep

et  $X_1, \dots, X_n \sim \Gamma(n, \theta)$

$$h(x) = \frac{\theta^n x^{n-1} e^{-\theta x}}{(n-1)!}$$

Obs:  $X \perp Y$ ,  $Z \subset X + Y$

$$f_Z(z) = \int f_X(z-x) f_Y(x) dx$$

$X \perp Y$  indep descn,  $Z = X + Y$

$$P(Z=k) = P(X+Y=k)$$

$$= \sum_x P(X+Y=k | X=x) P(X=x)$$

$$= \sum_x \underbrace{P(Y=k-x | X=x)}_{\text{indep}} P(X=x)$$

$$= \sum_x P(Y=k-x) P(X=x)$$

$$f_Z(z) = \int f_Y(z-t) f_X(t) dt$$

$$E\left[\frac{1}{\sum X_i}\right] = \int \frac{1}{x} h_n(x) dx$$

$$= \int_0^\infty \frac{1}{x} \frac{\theta^n x^{n-1} e^{-\theta x}}{(n-1)!} dx$$

$$= \frac{\theta}{n-1} \int_0^\infty h_{n-1}(x) dx = \frac{\theta}{n-1}$$

$$\boxed{E[\hat{\theta}_n] = \frac{n\theta}{n-1} \quad \text{debiased}}$$



$$\sum X_i \sim \Gamma(n, \theta)$$

densitatea lui  $\Gamma(n, \theta)$  notăm cu  $h_n(x)$

$$\underline{h_n(x)} = \theta^n \frac{x^{n-1}}{(n-1)!} e^{-\theta x}, x \geq 0$$

$$\int \frac{1}{x} h_n(x) dx = \int_0^{\infty} \frac{1}{x} \theta^n \frac{x^{n-1}}{(n-1)!} e^{-\theta x} dx$$

$$= \int_0^{\infty} \theta^n \left( \frac{x^{n-2}}{(n-1)!} e^{-\theta x} \right) dx$$

$$= \int_0^{\infty} \underbrace{\left( \frac{\theta}{n-1} \right)}_{\text{constant}} \underbrace{\left( \frac{\theta^{n-1} x^{n-2}}{(n-2)!} e^{-\theta x} \right)}_{h_{n-1}(x)} dx$$

$$= \frac{\theta}{n-1} \int_0^{\infty} h_{n-1}(x) dx = \frac{\theta}{n-1} \quad \underline{\quad \quad \quad = 1}$$

Putem să calculăm și  $MSE_{\theta}(\hat{\theta}_n)$

$$MSE_{\theta}(\hat{\theta}_n) = Var_{\theta}(\hat{\theta}_n) + b_{\theta}(\hat{\theta}_n)^2$$

$$b_{\theta}(\hat{\theta}_n) = E[\hat{\theta}_n] - \theta = \frac{n}{n-1} \theta - \theta = \frac{\theta}{n-1}$$

$$Var(\hat{\theta}_n) = E[\hat{\theta}_n^2] - \underbrace{E[\hat{\theta}_n]^2}_{\left(\frac{n}{n-1}\theta\right)^2}$$

$$E[\hat{\theta}_n^2] = n^2 E\left[\left(\frac{1}{\sum X_i}\right)^2\right]$$

$$= n^2 \int \frac{1}{x^2} f_{\sum}(x) dx = \frac{n^2 \theta^2}{(n-1)(n-2)}$$

$$Var(\hat{\theta}_n) = \frac{n^2 \theta^2}{(n-1)(n-2)} - \frac{n^2 \theta^2}{(n-1)^2} = \frac{n \theta^2}{(n-1)^2(n-2)}$$

$$\bar{T}_n = \frac{n-1}{n} \hat{\theta}_n \quad \text{und biased}$$

$$\left. \begin{array}{l} \hat{\theta}_n \xrightarrow{a.s.} \theta \\ \left(\frac{n-1}{n}\right) \rightarrow 1 \end{array} \right\} \Rightarrow \frac{n-1}{n} \hat{\theta}_n \xrightarrow{a.s.} \theta \quad \text{consist.}$$

$$\begin{aligned} \sqrt{n}(\bar{T}_n - \theta) &= \sqrt{n}\left(\frac{n-1}{n} \hat{\theta}_n - \theta\right) \\ &= \sqrt{n}\left(\hat{\theta}_n - \theta - \frac{1}{n} \hat{\theta}_n\right) \\ &= \underbrace{\sqrt{n}(\hat{\theta}_n - \theta)}_{\substack{\downarrow d \\ N(0, \theta^2)}} - \underbrace{\frac{1}{\sqrt{n}} \hat{\theta}_n}_{\substack{\downarrow P \\ 0}} \end{aligned}$$

$$\sqrt{n}(T_n - \theta) = \sqrt{n}(\hat{\theta}_n - \theta) - \frac{1}{\sqrt{n}} \hat{\theta}_n \quad \text{Th. Slutsky}$$

$\downarrow \text{Id}$   
 $N(0, \sigma^2)$

$\downarrow \theta$   
 $0$

$$\left. \begin{array}{l} \frac{1}{\sqrt{n}} \rightarrow 0 \\ \hat{\theta}_n \xrightarrow{P} \theta \end{array} \right\} \Rightarrow \frac{1}{\sqrt{n}} \hat{\theta}_n \xrightarrow{P} 0$$

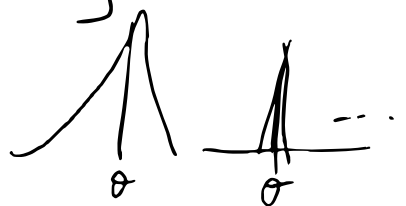
$\begin{array}{c} \sqrt{n}(T_n - \theta) \\ \downarrow \text{Id} \\ N(0, \sigma^2) \end{array}$

$\hat{\theta}_n, T_n$  ? per core & aligen.

$$\begin{aligned} \text{MSE}_\theta(T_n) &= \text{Var}_\theta(T_n) \\ &= \frac{(n-1)^2 \text{Var}_\theta(\hat{\theta}_n)}{(n-1)^2} \leq \text{MSE}_\theta(\hat{\theta}_n) \\ &= \frac{(n-1)^2}{n^2} \frac{n\sigma^2}{(n-1)^2(n-2)} = \frac{\sigma^2}{n-2} \end{aligned}$$

$$\begin{aligned} \text{MSE}_\theta(\hat{\theta}_n) &= \frac{n^2\sigma^2}{(n-1)^2(n-2)} + \frac{\sigma^2}{(n-1)^2} \\ &= \frac{\sigma^2}{(n-1)^2} \left[ \frac{n^2}{n-2} + 1 \right] \end{aligned}$$

$$\hat{\theta}_n \simeq N(\theta, \frac{\sigma^2}{n})$$



Cat este ind. în Fisher?

$$I_1(\theta) = E \left[ \left( \frac{\partial}{\partial \theta} \log f_{\theta}(X_1) \right)^2 \right] \\ = -E \left[ \frac{\partial^2}{\partial \theta^2} \log f_{\theta}(X_1) \right]$$

$$\log f_{\theta}(x) = \log(\theta e^{-\theta x}) = \log \theta - \theta x$$

$$\frac{\partial}{\partial \theta} \log f_{\theta}(x) = \frac{1}{\theta} - x$$

$$\frac{\partial^2}{\partial \theta^2} \log f_{\theta}(x) = -\frac{1}{\theta^2}$$

$$I_1(\theta) = \frac{1}{\theta^2} \Rightarrow I_n(\theta) = n I_1(\theta) = \frac{n}{\theta^2}$$

$$\text{MIRC} = \frac{1}{I_n(\theta)} = \frac{\theta^2}{n}$$

$$V_{\theta}(T_n) = \frac{\theta^2}{n-2} > \frac{\theta^2}{n} = \text{MIRC}$$

decî  $T_n$  nu este  
eficient

$$e_{\theta}^{\text{eff}}(\hat{\theta}_n) = \frac{\text{MIRC}}{V_{\theta}(\hat{\theta}_n)} = \frac{n-2}{n} \xrightarrow{n \rightarrow \infty} 1$$

$$\textcircled{\Sigma_x} \quad X \sim N(0, \sigma^2)$$

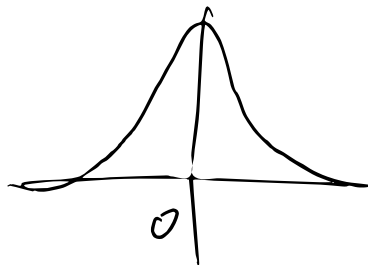
$x_1, \dots, x_n$  din pop  $X + \theta$

$$\boxed{F_\theta(x) = F(x - \theta)} \quad , F \text{ fct. din rep } N(0, \sigma^2)$$

Cat este mediana?

$$F_\theta^{-1}(1/2) = ?$$

$$F_\theta(0) = F(0) = 1/2$$



$$\textcircled{F_\theta^{-1}}\left(\frac{1}{2}\right) = 0 \quad \Rightarrow \quad x_{1/2} = 0$$

Med. empirică:

$$\hat{x}_n(1/2) = \hat{F}_n^{-1}(1/2)$$

$$\left. \begin{aligned} f_\theta(0) &= f(0) \\ &= \frac{1}{\sqrt{2\pi}\sigma} \end{aligned} \right\}$$

$$\sqrt{n}(\hat{x}_n(1/2) - x_{1/2}) = \sqrt{n}(\hat{x}_n(1/2) - 0)$$

$$\xrightarrow{d} N\left(0, \frac{\frac{1}{2}(1-\frac{1}{2})}{f_\theta(x_{1/2})^2}\right) \quad \left| \quad \begin{array}{l} \sqrt{n}(\hat{x}_{np} - x_p) \\ \downarrow d \\ N\left(0, \frac{p(1-p)}{f(x_p)^2}\right) \end{array} \right.$$

$$\sqrt{n}(\hat{x}_n(1/2) - \theta) \xrightarrow{d} N\left(0, \frac{\frac{1}{4}}{\frac{1}{2\pi\sigma^2}}\right) \\ = N\left(0, \frac{\pi\sigma^2}{2}\right)$$

$$\underline{\bar{X}_n} \leftarrow \text{ERM}$$

$$\sqrt{n}(\bar{X}_n - \theta) \xrightarrow{d} N(0, \sigma^2)$$

$$\text{Cum } \frac{\pi\sigma^2}{2} > \sigma^2 \text{ due to bias } \bar{X}_n$$