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Correction: exam Stat Mai 2018

Ex 1:

$$p/ L(x_1, \dots, x_n, \alpha) = e^{-n\alpha} \frac{\sum_{i=1}^n x_i}{\prod_{i=1}^n x_i!}$$

$$\log L(x_1, \dots, x_n, \alpha) = -n\alpha + \sum_{i=1}^n x_i \log \alpha + k.$$

$$\frac{\partial}{\partial \alpha} \log L(x_1, \dots, x_n, \alpha) = -n + \frac{\sum_{i=1}^n x_i}{\alpha} = 0$$

$$\Leftrightarrow \alpha = \frac{1}{n} \sum_{i=1}^n x_i$$

$$\frac{\partial^2}{\partial \alpha^2} \log L(x_1, \dots, x_n, \alpha) = - \frac{\sum_{i=1}^n x_i}{\alpha^2} < 0$$

$$\hat{\alpha} = \bar{X}_n = \frac{1}{n} \sum_{i=1}^n x_i$$

$$2/ E(\hat{\alpha}) = E(\bar{X}_n) = \frac{1}{n} \sum_{i=1}^n E(x_i)$$

$$\text{on } E(x_i) = \sum_{n=0}^{+\infty} n P(X=n) = \sum_{n=0}^{+\infty} n e^{-\alpha} \frac{\alpha^n}{n!}$$

$$= \alpha e^{-\alpha} \sum_{n=1}^{+\infty} \frac{\alpha^{n-1}}{(n-1)!} = \alpha$$

donc $\hat{\alpha}$ est un estimateur sans biais

② $(X_i)_{i=1, \dots, n}$ sont iid d'après la loi de grands nombres

$$\hat{\alpha} = \bar{X}_n \text{ converge vers } \underline{\mathbb{E}(X)} = \alpha$$

3°) D'après le théorème TCL on a:

$$\frac{\bar{X}_n - \mathbb{E}(\bar{X}_n)}{\sqrt{V(\bar{X}_n)}} \xrightarrow{L.i.} \mathcal{N}(0,1)$$

cherchons z_α tel $\mathbb{P}\left(\left|\frac{\bar{X}_n - \mathbb{E}(\bar{X}_n)}{\sqrt{V(\bar{X}_n)}}\right| \leq z_\alpha\right) = \frac{95}{100}$

$$2\Phi(z_\alpha) - 1 = 0,95 \Rightarrow \Phi(z_\alpha) = 0,975$$

d'où $z_\alpha = 1,96$.

$$V(\bar{X}_n) = \frac{1}{n^2} \sum_{i=1}^n V(X_i) = \frac{1}{n} V(X)$$

$$\mathbb{E}(X^2) = \sum_{n=2}^{+\infty} n^2$$

$$\mathbb{E}(X(X-1)) = \sum_{n=0}^{+\infty} n(n-1) \mathbb{P}(X=n)$$

$$= e^{-\alpha} \sum_{n=2}^{+\infty} n(n-1) \frac{\alpha^n}{n!} = e^{-\alpha} \sum_{n=2}^{+\infty} \frac{\alpha^n}{(n-2)!}$$

$$= \alpha^2$$

(3)

$$P(\alpha \in [\bar{X}_n - z_\alpha \sqrt{V(\bar{X}_n)}, \bar{X}_n + z_\alpha \sqrt{V(\bar{X}_n})]) = \dots$$

$$V(\bar{X}_n) = \frac{1}{n} V(x) = \frac{1}{n} (\alpha^2 + \alpha - \alpha^2) = \frac{\alpha}{n}$$

On aime α par \bar{X}_n

on aura

$$I_\alpha = \left[\bar{X}_n - z_\alpha \sqrt{\frac{\alpha}{n}}, \bar{X}_n + z_\alpha \sqrt{\frac{\alpha}{n}} \right]$$

Exercice 2:

$$\int_{\mathbb{R}} f(x) dx = \int_0^{+\infty} \frac{\lambda}{\Gamma(k)} x^{k-1} e^{-x/\theta} dx = \frac{\lambda}{\theta^k} \int_0^{+\infty} (\theta y)^{k-1} e^{-y} \theta dy$$

$$= \frac{\lambda}{\theta^k} \int_0^{+\infty} y^{k-1} e^{-y} dy = \theta^{-k} \Gamma(k) = \theta^{-k} \lambda \Gamma(k) = 1$$

$$\text{donc } \lambda = \frac{1}{(k-1)!}$$

$$2/ \quad L(x_1, \dots, x_n, \theta) = \left(\frac{n}{\theta k}\right)^n \prod_{i=1}^n x_i^{k-1} e^{-\frac{1}{\theta} \sum x_i} \quad \frac{1}{(\int_0, \infty) C)^n} \quad (4)$$

from $x_1, \dots, x_n \in]0, \infty[$

$$\text{Log } L(x_1, \dots, x_n, \theta) = n \text{Log } \frac{n}{\theta k} - n \text{Log}(\theta) + \sum_{i=1}^n (k-1) \text{Log } x_i - \frac{1}{\theta} \sum_{i=1}^n x_i$$

$$\frac{\partial \text{Log } L(x_1, \dots, x_n, \theta)}{\partial \theta} = -\frac{nk}{\theta} + \frac{1}{\theta^2} \sum_{i=1}^n x_i = 0$$

$$\Leftrightarrow \frac{nk}{\theta} = \frac{1}{\theta^2} \sum_{i=1}^n x_i \Leftrightarrow \theta = \frac{1}{nk} \sum_{i=1}^n x_i$$

$$\frac{\partial^2 \text{Log } L(x_1, \dots, x_n, \theta)}{\partial \theta^2} = +\frac{nk}{\theta^2} - \frac{2}{\theta^3} \sum_{i=1}^n x_i = \frac{1}{\theta^2} \left[nk - \frac{2}{\theta} \sum_{i=1}^n x_i \right]$$

$$\frac{\partial^2 \text{Log } L}{\partial \theta^2} / \frac{\partial \text{Log } L}{\partial \theta} = -\frac{nk}{\theta^2} < 0$$

$$\text{Dmn } \hat{\theta} = \frac{\sum_{i=1}^n x_i}{nk}$$

$$3/ E(\hat{\theta}) = \frac{1}{nk} \sum_{i=1}^n E(x_i) = \frac{E(x)}{k} = \hat{\theta} \quad (5)$$

$$\text{Var}(\hat{\theta} | x) = \int_0^{+\infty} x \frac{\lambda}{\theta^k} x^{k-1} e^{-\frac{x}{\theta}} dx, \quad y = \frac{x}{\theta}$$

$$= \int_0^{+\infty} \theta y \frac{\lambda}{\theta^k} (\theta y)^{k-1} e^{-y} dy \theta$$

$$= \int_0^{+\infty} \theta \lambda y^k e^{-y} dy = \theta \Gamma(k+1) \lambda \theta^k$$

$$= \underline{\underline{k\theta}}$$

$$E(x^2) = \int_0^{+\infty} \theta^2 \lambda y^{k+1} e^{-y} dy = \Gamma(k+2) \theta^2 \lambda$$

$$= (k+1)! \theta^2 \lambda$$

$$V(x) = k(k+1)\theta^2 - k\theta^2 = k\theta^2$$

$$V(\bar{x}_n) = \frac{1}{n^2} \sum_{i=1}^n V(x_i) = \frac{k\theta^2}{n} \Rightarrow V(\hat{\theta}) = V\left(\frac{\bar{x}_n}{k}\right)$$

$$= \frac{\theta^2}{kn}$$

4/ $E(\hat{\theta}) = \theta \Rightarrow \hat{\theta}$ est un estimateur sans biais (6)

et d'après le th^m de grands nombres

$$\hat{\theta} = \frac{\sum_{i=1}^n X_i}{n} \longrightarrow \frac{E(X)}{k} = \theta \Rightarrow \hat{\theta} \text{ sans biais}$$

est un estimateur convergent (L.G.N.).

$$5/ I_n(\theta) = V\left(\frac{\partial}{\partial \theta} \ln L\right) = V\left(\frac{-nk}{\theta} + \frac{1}{\theta^2} \sum_{i=1}^n X_i\right)$$

$$= \frac{1}{\theta^4} \sum_{i=1}^n V(X_i) = \frac{1}{\theta^4} nk\theta^2 = \frac{nk}{\theta^2}$$

$$\Rightarrow I_n(\theta) = \frac{1}{V(\hat{\theta})} \Rightarrow \hat{\theta} \text{ optimal (efficace).}$$

et sans biais

Exercice 3

1) (X_i) iid $\xrightarrow{\text{par T.C.L}}$ $\frac{\bar{X}_n - E(\bar{X}_n)}{\sqrt{V(\bar{X}_n)}} \xrightarrow{n \rightarrow \infty} \mathcal{N}(0, 1)$

$\Rightarrow \bar{X}_n \sim \mathcal{N}(E(\bar{X}_n), V(\bar{X}_n))$

$E(\bar{X}_n) = \frac{1}{n} \sum_{i=1}^n \underbrace{E(X_i)}_p = p$

$V(\bar{X}_n) = \frac{V(X_i)}{n} = \frac{p(1-p)}{n}$

2) $\left[\bar{X}_n - t_\alpha \sqrt{\frac{S_n^2}{n}}, \bar{X}_n + t_\alpha \sqrt{\frac{S_n^2}{n}} \right]$,

avec $S_n^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X}_n)^2$

$t_\alpha = F_{St(n-1)}^{-1} \left(1 - \frac{\alpha}{2} \right)$
Student

$\approx F_{\mathcal{N}(0,1)}^{-1} \left(1 - \frac{\alpha}{2} \right)$

$= F_{\mathcal{N}(0,1)}^{-1} \left(1 - \frac{0,98}{2} \right)$

$= F_{\mathcal{N}(0,1)}^{-1} (0,51) \stackrel{\text{Table}}{=} 0,02 \text{ (ou } 0,03)$

(et si on regard $F_{St(99)}^{-1}(0,51) \rightarrow$ la plus proche dans la table Student est 0,126.)