

Sur 17

1. #6.2.7 a)  $f(x; \theta) = \frac{1}{\Gamma(x) \theta^\alpha} x^{\alpha-1} e^{-x/\theta}$   $x > 0, \theta > 0, \alpha > 0$

(Sur 6)

$$l(\theta) = \ln f(x; \theta) = (\alpha-1) \ln x - \frac{x}{\theta} - \ln \Gamma(x) - \alpha \ln \theta$$

$$l'(\theta) = \frac{x}{\theta^2} - \frac{\alpha}{\theta} \Rightarrow \text{Pour un échantillon } l'(\theta) = 0 \Rightarrow \hat{\theta} = \frac{\sum x_i}{n \alpha}$$

$$l''(\theta) = -\frac{2x}{\theta^3} + \frac{\alpha}{\theta^2}$$

$$I(\theta) = -E l''(\theta) = \frac{2EX}{\theta^3} - \frac{\alpha}{\theta^2} = \frac{2\alpha\theta}{\theta^3} - \frac{\alpha}{\theta^2} = \frac{\alpha}{\theta^2} \quad (\text{p.158})$$

$$b) \text{Var } \hat{\theta} = \frac{\text{Var } X}{n \alpha^2} = \frac{\alpha \theta^2}{n \alpha^2} = \frac{\theta^2}{n \alpha} \quad (\text{p.158})$$

Borne inférieure Cramer Rao =  $\frac{1}{n I(\theta)} = \frac{\theta^2}{n \alpha} \Rightarrow \hat{\theta}$  est efficace.

$$c) \sqrt{n}(\hat{\theta} - \theta) \xrightarrow{D} N(0, I^{-1}(\theta)) = N(0, \frac{\theta^2}{\alpha})$$

2. #6.2.9  $E(X) = \int_0^{\infty} x \frac{3\theta^2}{(x+\theta)^4} dx = \frac{\theta}{2}$  (intégration en parties)

(Sur 4)

$$EY = 2 E\bar{X} = 2 EX = 2 \cdot \frac{\theta}{2} = \theta \Rightarrow \text{non biaisé}$$

$$l(\theta) = \ln f(x; \theta) = 3 \ln(3\theta) - 4 \ln(x+\theta)$$

$$l'(\theta) = \frac{3}{\theta} - \frac{4}{x+\theta}$$

$$l''(\theta) = -\frac{3}{\theta^2} + \frac{4}{(x+\theta)^2}$$

$$I(\theta) = -E l''(\theta) = \frac{3}{\theta^2} - 4 E \frac{1}{(x+\theta)^2} = \frac{3}{5\theta^2}$$

Borne inférieure de Fisher Rao =  $\frac{1}{n I(\theta)} = \frac{5\theta^2}{3n}$

Par contre  $Var(2\bar{X}) = 4 Var \bar{X} = \frac{4 Var X}{n}$

$$E(X+\theta)^2 = 3\theta \int_0^\infty \frac{(x+\theta)^2 dx}{(x+\theta)^4} = 3\theta^3 \int_\theta^\infty \frac{1}{u^2} du = 3\theta^2$$

Mais  $E(X+\theta)^2 = EX^2 + 2\theta EX + \theta^2$   
 $= EX^2 + \theta^2 + \theta^2 = 3\theta^2 \Rightarrow EX^2 = 3\theta^2 - 2\theta^2$

$$Var X = EX^2 - (EX)^2 = 3\theta^2 - 2\theta^2 - \left(\frac{\theta}{2}\right)^2 = \frac{3\theta^2}{4}$$

$$\Rightarrow Var(2\bar{X}) = 4 \cdot \frac{3\theta^2}{4n} = \frac{3\theta^2}{n}$$

Efficacité de  $(2\bar{X}) = \frac{\text{Information}(2\bar{X})}{\text{Information(Fisher)}} = \frac{n/3\theta^2}{3n/5\theta^2} = \frac{5}{3(3)} = \frac{5}{9}$

3. # 6.4.3

$$L(\theta) = \prod f(x_i; \theta), \quad \theta = (\theta_1, \theta_2)$$

(sur 3)

$$= \frac{1}{\theta_2^n} e^{-\sum (x_i - \theta_1) / \theta_2} \quad \theta_1 \leq \min x_i$$

$$-\infty < \theta_2 < \infty$$

$\Rightarrow \hat{\theta}_1 = \min x_i$   
 aussi

$$l(\theta) = \ln L(\theta)$$

$$\frac{\partial l(\theta)}{\partial \theta_2} = -\frac{n}{\theta_2} + \frac{\sum (x_i - \theta_1)}{\theta_2^2} = 0$$

$$\Rightarrow \hat{\theta}_2 = \frac{\sum (x_i - \hat{\theta}_1)}{n}$$

4. #6.4.6

On sait que le max. de vraisemblance d'une fonction  $V(\theta)$  est  $V(\hat{\theta})$ .

(sur 4)

$$a) P(X \leq b) = \Phi\left(\frac{b - \mu}{\sigma}\right) = 0.90$$

les estimés de  $\mu, \sigma$  sont  $\bar{x}$  et  $s$ . Puisque

$$\frac{b - \mu}{\sigma} = \Phi^{-1}(0.90) = 1.282$$

$$\begin{aligned}\hat{b} &= \hat{\sigma} \Phi^{-1}(0.90) + \hat{\mu} \\ &= s \cdot \Phi^{-1}(0.90) + \bar{x}\end{aligned}$$

$$b). P(X \leq c) = \Phi\left(\frac{c - \mu}{\sigma}\right)$$

$\Rightarrow$  L'estimé mle est  $\Phi\left(\frac{c - \bar{x}}{s}\right)$ .