


9.7  Sea \mathbf{X}_n una muestra aleatoria de tamaño n de la distribución uniforme sobre el intervalo $[0, \theta]$.

(a) Hallar un estadístico suficiente para θ basado en \mathbf{X}_n .

(b) Hallar el estimador de máxima verosimilitud de θ basado en \mathbf{X}_n .

(c) Sea $\hat{\theta}_n$ el estimador de máxima verosimilitud de θ hallado en el inciso anterior.

Mostrar que $\mathbf{E}_\theta[\hat{\theta}_n] = \frac{n}{n+1}\theta$ y $\text{var}_\theta(\hat{\theta}_n) = \frac{n\theta^2}{(n+1)^2(n+2)}$ para concluir que $\hat{\theta}_n$ converge en media cuadrática a θ cuando $n \rightarrow \infty$.

(2)

$$\underline{X} = (X_1, \dots, X_n) \sim \mathcal{U}(0, \theta), f_\theta(x_i) = \frac{1}{\theta} \mathbb{1}\{0 < x_i < \theta\}$$

$$f_\theta(\underline{x}) = \prod_{i=1}^n f_\theta(x_i) = \left(\frac{1}{\theta}\right)^n \cdot \underbrace{\prod_{i=1}^n \mathbb{1}\{0 < x_i < \theta\}}_{\mathbb{1}\{\max(x_1, \dots, x_n) < \theta\}} \mathbb{1}\{\min(x_1, \dots, x_n) > 0\}$$

$$L(\theta) = f_\theta(\underline{x}) = \underbrace{\frac{1}{\theta^n} \mathbb{1}\{\max(x_1, \dots, x_n) < \theta\}}_{g(\max(x_1, \dots, x_n), \theta)} \underbrace{\mathbb{1}\{\min(x_1, \dots, x_n) > 0\}}_{h(\underline{x})}$$

Por lo tanto, por teorema de factorización

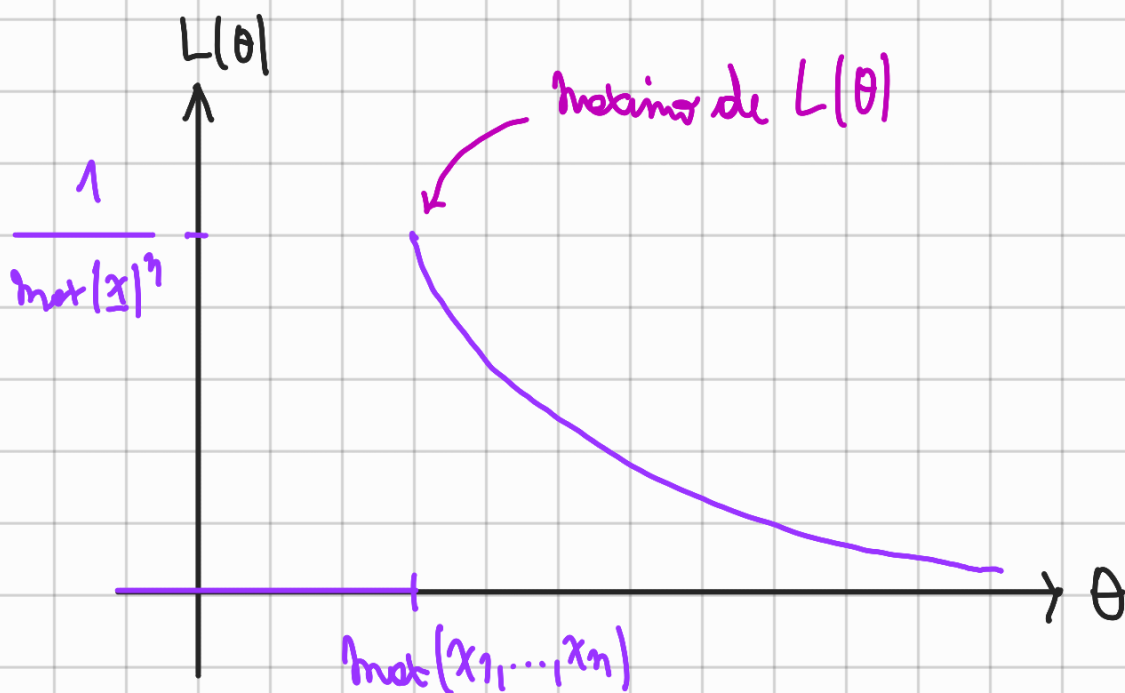
$$\Rightarrow \boxed{T = \max(X_1, \dots, X_n)}$$

(6)

$$\Rightarrow L(\theta) = \frac{1}{\theta^n} \mathbb{1}\{\max(x_1, \dots, x_n) < \theta\} \mathbb{1}\{\min(x_1, \dots, x_n) > 0\}, \quad \theta > 0$$

No puedo derivar (como derivar una $\mathbb{1}$) ni opción \ln (de $\mathbb{1}$ de 0 o 1)

Miro el gráfico



$$\Rightarrow \hat{\theta} = \max(x_1, \dots, x_n)$$

(c) Common distribution $\hat{\theta}$?

$$\begin{aligned}F_{\hat{\theta}}(t) &= \mathbb{P}(\hat{\theta} \leq t) = \mathbb{P}(\max\{X_1, \dots, X_n\} \leq t) \\&= \mathbb{P}(X_1 \leq t) \cdot \mathbb{P}(X_2 \leq t) \dots \mathbb{P}(X_n \leq t) \\&= \mathbb{P}(X_1 \leq t)^n \\&= F_{X_1}(t)^n \\&= \left(\frac{1}{\theta} t\right)^n \mathbb{1}\{t < \theta\} + 1 \mathbb{1}\{t \geq \theta\}\end{aligned}$$

$$\begin{aligned}f_{\hat{\theta}}(t) &= \frac{d}{dt} F_{\hat{\theta}} = n \left(\frac{1}{\theta} t\right)^{n-1} \cdot \frac{1}{\theta} \mathbb{1}\{0 < t < \theta\} \\&= \frac{n}{\theta^n} t^{n-1} \mathbb{1}\{0 < t < \theta\} \rightarrow \int_0^{\theta} \frac{n}{\theta^n} t^{n-1} dt = 1 \quad \checkmark\end{aligned}$$

$$\begin{aligned}E[\hat{\theta}] &= \int_0^{\theta} t \frac{n}{\theta^n} t^{n-1} dt = \frac{n}{\theta^n} \int_0^{\theta} t^n dt = \frac{n}{\theta^n} \left(\frac{t^{n+1}}{n+1} \Big|_0^{\theta} \right) \\&= \frac{n}{\theta^n} \frac{\theta^{n+1}}{n+1}\end{aligned}$$

$$\Rightarrow E[\hat{\theta}] = \frac{n}{n+1} \theta$$

$$V_n(\hat{\theta}) = E(\theta^2) - E(\theta)^2$$

$$E(\theta^2) = \int_0^{\theta} t^2 \frac{n}{\theta^n} t^{n-1} dt = \frac{n}{\theta^n} \int_0^{\theta} t^{n+1} dt = \frac{n}{\theta^n} \frac{\theta^{n+2}}{n+2}$$

$$= \frac{n}{n+2} \theta^2$$

$$E(\theta)^2 = \left(\frac{n}{n+1} \theta \right)^2 = \frac{n^2}{(n+1)^2} \theta^2$$

$$V_n(\hat{\theta}) = \frac{n}{n+2} \theta^2 - \frac{n^2}{(n+1)^2} \theta^2 = \frac{n(n+1)^2 - n^2(n+2)}{(n+2)(n+1)^2} \theta^2$$

$$= \frac{n^3 + 2n^2 + n - n^3 - 2n^2}{(n+2)(n+1)^2} \theta^2$$

$$\Rightarrow V_m(\hat{\theta}) = \frac{n\theta^2}{(n+2)(n+1)^2}$$

Convergencia en media cuadrática:

$$\lim_{n \rightarrow \infty} ECM(\hat{\theta}) = 0, \forall \theta \in \Theta$$

Error cuadrático medio: $ECM(\hat{\theta}) = B(\hat{\theta})^2 + V(\hat{\theta})$

Desvío: $B(\hat{\theta}) = E[\hat{\theta}] - \theta$

$$B(\hat{\theta}) = \frac{n}{n+1} \theta - \theta = \left(\frac{n}{n+1} - 1 \right) \theta = \frac{-1}{n+1} \theta$$

$$ECM(\hat{\theta}) = \frac{1}{(n+1)^2} \theta^2 + \frac{n\theta^2}{(n+2)(n+1)^2}$$

$$\lim_{n \rightarrow \infty} ECM(\hat{\theta}) = \lim_{n \rightarrow \infty} \overbrace{\frac{1}{(n+1)^2} \theta^2}^{\rightarrow 0} + \overbrace{\frac{n\theta^2}{(n+2)(n+1)^2}}^{\rightarrow 0} = 0$$

$$\Rightarrow \hat{\theta} \text{ converge en media cuadrática a } \theta \text{ cuando } n \rightarrow \infty$$