

1) Sea $A(x_1, x_2, \dots, x_n)$ donde $A \sim N(\mu, \sigma)$
muestre que los estimadores de máxima verosimilitud son:

$$\hat{\mu} = \frac{1}{n} \sum_{i=1}^n x_i$$

$$\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n x_i^2$$

sol:

$$L = \prod_{i=1}^n \frac{e^{-\frac{1}{2\sigma^2}(x_i - \mu)^2}}{\sqrt{2\pi\sigma^2}} = \frac{1}{\sqrt{2\pi\sigma^2}^n} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2}$$

Ahora

$$\ln(L(x, \mu, \sigma)) = \ln \left[\frac{1}{\sqrt{2\pi\sigma^2}^n} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2} \right] = \sum_{i=1}^n \ln \left[\frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2} (x_i - \mu)^2} \right]$$
$$= \sum_{i=1}^n \left[\ln \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right) + \ln \left(e^{-\frac{1}{2\sigma^2} (x_i - \mu)^2} \right) \right] = \sum_{i=1}^n \left[\ln \frac{1}{\sqrt{2\pi\sigma^2}} - \frac{1}{2\sigma^2} (x_i - \mu)^2 \right]$$

derivar parcialmente respecto a μ

$$\frac{d}{d\mu} \ln(L) = -\frac{1}{2\sigma^2} \sum_{i=1}^n 2(x_i - \mu) = 0$$

$$\sum_{i=1}^n x_i - n\mu = 0 \rightarrow \sum_{i=1}^n x_i = n\mu$$

$$\therefore \mu = \frac{1}{n} \sum_{i=1}^n x_i$$

derivar respecto σ

$$\frac{d(\ln(L))}{d\sigma} = \sum_{i=1}^n \frac{d}{d\sigma} \left[\ln \frac{1}{\sqrt{2\pi}\sigma} - \frac{1}{2} \frac{(x_i - \mu)^2}{\sigma^2} \right]$$

$$= \sum_{i=1}^n \left[-\frac{d}{d\sigma} (\ln(\sqrt{2\pi}\sigma)) - \frac{(x_i - \mu)^2}{2} \frac{d}{d\sigma} \left(\frac{1}{\sigma^2} \right) \right]$$

$$= \sum_{i=1}^n \left[-\frac{1}{\sigma} + \frac{(x_i - \mu)^2}{\sigma^3} \right] = 0$$

$$\sum_{i=1}^n \frac{1}{\sigma} = \sum_{i=1}^n \frac{(x_i - \mu)^2}{\sigma^3}$$

$$n\sigma^2 = \sum_{i=1}^n (x_i - \mu)^2$$

Por definición $\mu = \bar{x}$

$$\sigma^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \bar{x})^2$$