

$A(x_1, x_2, \dots, x_n)$, donde $A \sim N(\mu, \sigma^2)$

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

La función de verosimilitud está dada por:

$$\textcircled{1} l((\mu, \sigma^2); \mathbf{x}) = \prod_{i=1}^n \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2}$$

y la de log-verosimilitud:

$$\textcircled{2} L((\mu, \sigma^2); \mathbf{x}) = -\frac{n}{2} \log(2\pi\sigma^2) - \frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2$$

Iguálamos las expresiones:

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

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

$$\Rightarrow l(u) = \ln L(u) = \ln(2\pi\sigma^2)^{-n/2} + \ln(e^{-\frac{1}{2\sigma^2} \sum_{i=1}^n (x_i - \mu)^2})$$

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

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Derivamos e igualamos a 0

$$\Rightarrow \frac{\partial l(\mu)}{\partial \mu} = 0 = \frac{1}{2\sigma^2} \cdot 2 \sum_{i=1}^n (x_i - \mu) (-1)$$

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

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

$$\Rightarrow \sum_{i=1}^n (x_i - \mu) = 0$$

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

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

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

$$\bullet \frac{\partial L}{\partial \sigma} = -\frac{n}{2(2\sigma^2)} + \frac{1}{2\sigma^2} \sum$$

$$0 = -\frac{n}{2} + \frac{1}{2\sigma^2} \sum$$

$$\frac{n}{2\sigma} = \frac{1}{2\sigma^2}$$

$$\frac{2n}{2\sigma} = \frac{1}{\sigma^2} \Rightarrow$$

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

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