

## • CÁLCULO DE LA ESPERANZA

$$X \sim \text{GaI}(a, \lambda)$$

$$f_X(x; a, \lambda) = \frac{\lambda^a}{\Gamma(a)} \frac{1}{x^{a+1}} e^{-\lambda/x}, \quad \begin{matrix} a > 0 \\ \lambda > 0 \\ x > 0 \end{matrix}$$

$$E(X) = \int_0^{+\infty} x \cdot \frac{\lambda^a}{\Gamma(a)} \cdot \frac{1}{x^{a+1}} e^{-\lambda/x} dx$$

$$= \frac{\lambda^a}{\Gamma(a)} \cdot \int_0^{+\infty} \underbrace{\frac{1}{x^a}}_{x^{a-1+1}} e^{-\lambda/x} dx \quad ; \quad a-1 > 0$$

$$= \frac{\lambda^a}{\Gamma(a)} \cdot \frac{\Gamma(a-1)}{\lambda^{a-1}} = \frac{\lambda}{a-1}$$

$$\text{var}(X) = E(X^2) - (E(X))^2$$

$$E(X^2) = \int_0^{\infty} x^2 \cdot \frac{\lambda^a}{\Gamma(a)} \cdot \frac{1}{x^{a+1}} e^{-\lambda/x} dx$$

## • DISTRIBUCION NGaI(m, c; a, b)

$$(X, Y) \sim \text{NGaI}(m, c; a, b) \left. \begin{array}{l} \downarrow \quad \downarrow \\ \text{N} \quad \text{GaI} \end{array} \right\}$$

$$\begin{aligned} f_{(X, Y)}(x, y) &= f_Y(y) \cdot f_{X|Y=y}(x) \\ &\propto \frac{1}{y^{a+1}} e^{-\lambda/y} \cdot \frac{1}{y^{1/2}} e^{-\frac{(x-m)^2}{2yc}} \end{aligned}$$

## • FUNCIÓN DE VEROSIMILITUD

Sea  $x_1, \dots, x_n \sim N(\mu, \sigma^2)$  con ambos parámetros desconocidos.

$$\begin{aligned} l(\mu, \sigma^2 | \mathbf{x}) &= \prod_{i=1}^n f_{x_i}(\mu, \sigma^2) \\ &= \prod_{i=1}^n \frac{1}{\sqrt{2\pi} \sigma} e^{-\frac{(x_i - \mu)^2}{2\sigma^2}} \end{aligned}$$

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

Se simplifica el sumatorio:

$$\begin{aligned} \sum_{i=1}^n (x_i - \mu)^2 &= \sum_{i=1}^n (x_i - \bar{x} + \bar{x} - \mu)^2 \\ &= \sum_{i=1}^n \left[ (x_i - \bar{x})^2 + (\bar{x} - \mu)^2 + 2(x_i - \bar{x})(\bar{x} - \mu) \right] \end{aligned}$$

$$= \underbrace{\sum_{i=1}^n (x_i - \bar{x})^2}_{n s^2} + \underbrace{\sum_{i=1}^n (\bar{x} - \mu)^2}_{n(\bar{x} - \mu)^2} + \underbrace{\sum_{i=1}^n 2(x_i - \bar{x})(\bar{x} - \mu)}_{2(\bar{x} - \mu) \cdot \sum_{i=1}^n (x_i - \bar{x})}$$

$$\begin{aligned} &= n s^2 + n(\bar{x} - \mu)^2 + 2(\bar{x} - \mu) \cdot \underbrace{\sum_{i=1}^n (x_i - \bar{x})}_{\substack{\sum_{i=1}^n x_i - n\bar{x} \\ \parallel \\ n\bar{x} - n\bar{x} \\ \parallel \\ 0}} \end{aligned}$$

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

$$\sum_{i=1}^n (x_i - \mu)^2 = n s^2 + n(\bar{x} - \mu)^2$$

$$\ell(\mu, \sigma^2 | \mathbf{x}) = (2\pi)^{-\frac{n}{2}} (\sigma^2)^{-n/2} e^{-\frac{1}{2\sigma^2} [n s^2 + n(\bar{x} - \mu)^2]}$$

## DISTRIBUCION A POSTERIORI

### CASO 1: A priori no informativa

$$\pi(\theta, \phi) \propto \frac{1}{\phi}$$
$$l(\theta, \phi | x) \propto \frac{1}{\phi^{n/2}} \cdot e^{-\frac{1}{2\phi} [n s^2 + n(\bar{x} - \mu)^2]}$$

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$$\pi(\theta, \phi | x) \propto \underbrace{\frac{1}{\phi} \cdot \frac{1}{\phi^{n/2}} \cdot e^{-\frac{n s^2}{2\phi}}}_{\text{formato; Ga I}} \cdot \underbrace{e^{-\frac{n(\bar{x} - \theta)^2}{2\phi}}}_{\text{Normal}}$$

$\Downarrow$

$$\underbrace{\frac{1}{\phi^{1/2}} \cdot \frac{1}{\phi^{n/2}} e^{-\frac{n s^2}{2\phi}}}_{\text{GaI}\left(\frac{n-1}{2}, \frac{n s^2}{2}\right)} \cdot \underbrace{\frac{1}{\phi^{1/2}} e^{-\frac{(\bar{x} - \theta)^2}{2\phi/n}}}_{\mathcal{N}(\bar{x}; \phi/n)}$$

$$\frac{1}{\phi^{1/2}} \cdot \frac{1}{\phi^{n/2}} = \frac{1}{\phi^{\frac{n+1}{2}}} \equiv \frac{1}{\phi^{a+1}} \equiv \frac{1}{\phi^{\frac{n-1}{2}+1}}$$

$$(\theta, \phi | \mathbf{x}) \sim \text{NGaI} \left( \bar{x}, \frac{1}{n} ; \frac{n-1}{2}, \frac{n s^2}{2} \right)$$

CASO 2: A priori conjugada

$$(\theta, \phi) \sim \text{NGaI}(m, c; a, b)$$

$$\begin{aligned} \pi(\theta, \phi) &\propto \frac{1}{\phi^{a+1}} e^{-b/\phi} \cdot \frac{1}{\phi^{1/2}} e^{-\frac{(\theta-m)^2}{2c\phi}} \\ \ell(\theta, \phi | \mathbf{x}) &\propto \frac{1}{\phi^{n/2}} e^{-\frac{n s^2}{2\phi}} \cdot e^{-\frac{n(\theta-\bar{x})^2}{2\phi}} \end{aligned}$$


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$$\pi(\theta, \phi | \mathbf{x}) \propto \frac{1}{\phi^{a+1 + \frac{n}{2} + \frac{1}{2}}} e^{-\left[ b + \frac{n s^2}{2} + \frac{1}{2} \left( \frac{m^2 + n c \bar{x}^2}{1 + n c} \right) \right] / \phi} \cdot e^{-\frac{(\theta - \mu^*)}{2 \sigma^{*2}}}$$

Simplificación de la parte exponencial:

$$\frac{(\theta - m)^2}{c\phi} + \frac{(\theta - \bar{x})^2}{\phi/n} = \underbrace{\left(\frac{1}{c\phi} + \frac{n}{\phi}\right)}_A \theta^2 - 2\theta \underbrace{\left(\frac{m}{c\phi} + \frac{n\bar{x}}{\phi}\right)}_B + \underbrace{\left(\frac{m^2}{c\phi} + \frac{n\bar{x}^2}{\phi}\right)}_C$$

$$\sigma^{*2} = \frac{1}{\frac{1}{c\phi} + \frac{n}{\phi}} = \frac{c\phi}{1+nc} = \frac{c}{1+nc} \phi$$

$$\mu^* = \left(\frac{m}{c\phi} + \frac{n\bar{x}}{\phi}\right) \cdot \frac{c\phi}{1+nc} = \frac{m + nc\bar{x}}{1+nc}$$

$$= \underbrace{\frac{1}{1+nc}}_{w_1} m + \underbrace{\frac{nc}{1+nc}}_{w_2} \bar{x}$$

$$A\theta^2 - 2\theta B + C = A\left(\theta - \frac{B}{A}\right)^2 + \left(C - \frac{B^2}{A}\right)$$