

MATH350 - Exemplo 3.3

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Exemplo 3.3

$$X \sim \Gamma(\alpha, \beta)$$

$$\theta = (\alpha, \beta) \quad \text{e} \quad \Omega = (0, \infty) \times (0, \infty)$$

$$f(x, \theta) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{\beta x}$$

onde

$$\Gamma(\alpha) = \int_0^\infty s^{\alpha-1} e^{-\beta s} ds \quad \text{e} \quad E(X) = \frac{\alpha}{\beta}$$

$$\begin{aligned}
 L(\theta) &= \prod_{i=1}^n \left\{ \frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{\beta x} \right\} \\
 &= \left(\frac{\beta^\alpha}{\Gamma(\alpha)} \right)^n e^{-\beta \sum x} \prod_{i=1}^n x^{\alpha-1}
 \end{aligned}$$

$$I(\theta) = n\alpha \log(\beta) - n \log[\Gamma(\alpha)] - \beta \sum_{i=1}^n x_i + (\alpha - 1) \sum_{i=1}^n \log(x_i)$$

considerando

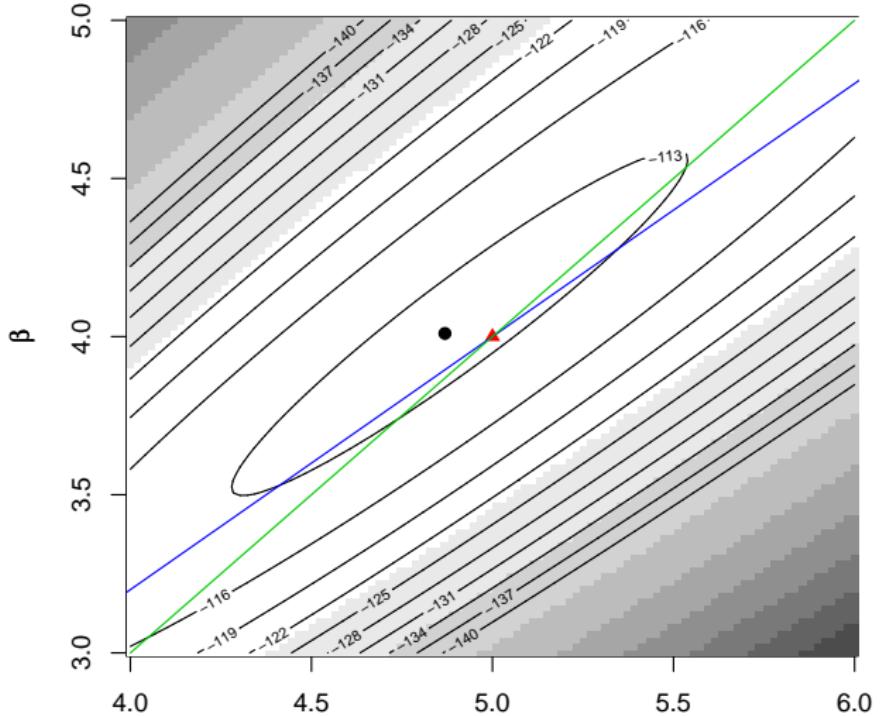
$$n = 25, \alpha = 4 \text{ e } \beta = 5$$

temos

$$E(X) = \frac{\alpha}{\beta} = \frac{4}{5}$$

Essa esperança pode ser demonstrada pela definição.

$$E(X) \int^D x.f(x)dx$$



Exemplo 3.3 (cont) Fazendo a derivada segunda

$$\frac{\delta I(\theta)}{\delta \alpha} = n \log(\beta) + \sum_{i=1}^n \log(x_i) - n \frac{\Gamma'(\alpha)}{\Gamma(\alpha)}$$

$$\frac{\delta I(\theta)}{\delta \beta} = \frac{n\alpha}{\beta} - \sum_{i=1}^n x_i$$

$$\frac{n\hat{\alpha}}{\hat{\beta}} - \sum_{i=1}^n x_i = 0 \rightarrow \hat{\beta} = \frac{\hat{\alpha}}{\bar{X}}$$

substituindo em $\frac{\delta I(\theta)}{\delta \beta}$

$$n \log(\hat{\alpha}) - n \log(\bar{X}) + \sum_{i=1}^n \log(x_i) - n \frac{\Gamma'(\alpha)}{\Gamma(\alpha)}$$

que requer solução numérica

segunda derivadas

$$\frac{\delta^2 I(\hat{\theta})}{\delta \alpha^2} = -n \left[\frac{\Gamma''(a\hat{\theta})}{\Gamma(a\hat{\theta})} - \left(\frac{\Gamma'(a\hat{\theta})}{\Gamma(a\hat{\theta})} \right)^2 \right]$$

$$\frac{\delta^2 I(\hat{\theta})}{\delta \beta} = -\frac{n\hat{\alpha}}{\beta^2}$$

$$\frac{\delta^2 I(\hat{\theta})}{\delta \alpha \delta \beta} = \frac{n}{\hat{\beta}}$$

Exemplo 3.3 (cont)

i.e. $X_i \sim \Gamma(\alpha, \beta)$.

$$U(\theta) = \left\{ n \log \beta + \sum_{i=1}^n \log X_i - n \frac{\Gamma'(\alpha)}{\Gamma(\alpha)}, \frac{n\alpha}{\beta} - \sum_{i=1}^n x_i \right\}^T$$

e

$$I_O(\theta) = I_E(\theta) = \begin{pmatrix} n[\Gamma''(\alpha)/\Gamma(\alpha) - (\Gamma(\alpha)/\Gamma(\alpha))^2] & -n/\beta \\ -n/\beta & n\alpha/\beta \end{pmatrix}$$

i.e. series expandida multi-variavel de Taylor

$$I(\theta) = I(\hat{\theta}) + (\theta - \hat{\theta})^T U(\hat{\theta}) - \frac{1}{2}(\theta - \hat{\theta})^T I_O(\theta^*)(\theta - \hat{\theta})$$

para

$$\|\theta^* - \theta\| \leq \|\hat{\theta} - \theta\| \quad (4.1)$$

$$U(\theta) = U(\hat{\theta}) - (\theta - \hat{\theta})^T I_O(\theta^+)$$

para

$$\|\theta^+ - \theta\| \leq \|\hat{\theta} - \theta\| \quad (4.2)$$

Para Deviance aplicando equaçao (4.1)

$$\begin{aligned} D(\theta) &= 2 \left[I(\hat{\theta}) - \{ I(\hat{\theta}) + (\theta - \hat{\theta})^T U(\hat{\theta}) + \frac{1}{2}(\theta - \hat{\theta})^T I_O(\theta^*)(\theta - \hat{\theta}) \} \right] \\ &= -(\theta - \hat{\theta})^T I_O(\theta^*)(\theta - \hat{\theta}) \\ &\approx -(\theta - \hat{\theta})^T I_O(\hat{\theta})(\theta - \hat{\theta}) \end{aligned}$$