

1 $P = \{\Gamma(\alpha, \beta) : \alpha > 0, \beta > 0\}$. Асимпт. макс. роб. для ОМН $(\hat{\alpha}, \hat{\beta})$ -?

$$L(\alpha, \beta) = \prod_{i=1}^n \frac{\alpha^\beta x_i^{\beta-1}}{\Gamma(\beta)} e^{-\alpha x_i} = \frac{\alpha^{n\beta}}{\Gamma(\beta)^n} \left(\prod_{i=1}^n x_i \right)^{\beta-1} e^{-\alpha \sum_{i=1}^n x_i}$$

$$\ell(\alpha, \beta) = \ln L(\alpha, \beta) = n\beta \ln \alpha - n \ln \Gamma(\beta) + (\beta-1) \sum_{i=1}^n \ln x_i - \alpha \sum_{i=1}^n x_i$$

$$\frac{\partial \ell}{\partial \alpha} = \frac{n\beta}{\alpha} - \sum_{i=1}^n x_i = 0 \Rightarrow \frac{\beta}{\alpha} = \bar{x}, \quad \frac{\partial \ell}{\partial \beta} = n \ln \alpha - \frac{n}{\Gamma(\beta)} \Gamma'(\beta) + \sum_{i=1}^n \ln x_i =$$

$$= n(\ln \alpha - \psi(\beta) + \bar{\ln x}) = 0 \Rightarrow \psi(\beta) - \ln \alpha = \bar{\ln x} \quad (\psi - \text{психометрический}) ; \quad \frac{\partial^2 \ell}{\partial \alpha^2} = -\frac{n\beta}{\alpha^2}, \quad \frac{\partial^2 \ell}{\partial \alpha \partial \beta} =$$

$$= \frac{n}{\alpha}, \quad \frac{\partial^2 \ell}{\partial \beta^2} = -n \frac{\Gamma''(\beta)}{\Gamma(\beta)} + n \left(\frac{\Gamma'(\beta)}{\Gamma(\beta)} \right)^2 \Rightarrow I(\alpha, \beta) = -E \frac{\partial^2 \ell}{\partial \alpha \partial \beta} = -E \begin{pmatrix} -\frac{n\beta}{\alpha^2} & \frac{n}{\alpha} \\ \frac{n}{\alpha} & -n \frac{\Gamma''(\beta)}{\Gamma(\beta)} + n \left(\frac{\Gamma'(\beta)}{\Gamma(\beta)} \right)^2 \end{pmatrix} = n \begin{pmatrix} \frac{\beta}{\alpha^2} & -\frac{1}{\alpha} \\ -\frac{1}{\alpha} & \frac{\Gamma''(\beta)}{\Gamma(\beta)} - \left(\frac{\Gamma'(\beta)}{\Gamma(\beta)} \right)^2 \end{pmatrix}$$

Асимпт. макс. роб. $\text{cov}(\hat{\alpha}, \hat{\beta}) = I^{-1} = \frac{1}{n} \begin{pmatrix} \frac{\alpha^2}{\beta} & \alpha \\ \alpha & \frac{1}{\beta(\frac{\Gamma''(\beta)}{\Gamma(\beta)} - (\frac{\Gamma'(\beta)}{\Gamma(\beta)})^2)} \end{pmatrix}$

$$= \frac{1}{n \left(\beta \left(\frac{\Gamma''(\beta)}{\Gamma(\beta)} - \left(\frac{\Gamma'(\beta)}{\Gamma(\beta)} \right)^2 \right) - 1 \right)} \begin{pmatrix} 1 & \alpha \\ \alpha & \beta \end{pmatrix}$$

2 $X_1, \dots, X_n \sim \text{Poi}(\theta)$; асимпт. макс. роб. для ОМН $\hat{\theta}$ -?

$$L(\theta) = \prod_{i=1}^n \frac{e^{-\theta} \theta^{x_i}}{x_i!} = \frac{e^{-n\theta} \theta^{\sum x_i}}{n!}$$

$$\ell(\theta) = \ln L(\theta) = -n\theta + \sum_{i=1}^n x_i \ln \theta - \sum_{i=1}^n \ln(x_i!)$$

$$\frac{\partial \ell}{\partial \theta} = -n + \frac{\sum x_i}{\theta} = 0 \Rightarrow \hat{\theta} = \frac{\sum x_i}{n} = \bar{x}$$

$$I = -E \frac{\partial^2 \ell}{\partial \theta^2} = -E \left(-\frac{\sum x_i}{\theta^2} \right) = \frac{n\theta}{\theta^2} = \frac{n}{\theta}, \quad I^{-1} = \frac{\theta}{n}, \quad \hat{\theta} = I^{-1} \frac{\partial \ell}{\partial \theta} = \frac{\theta}{n} \left(-n + \frac{\sum x_i}{\theta} \right) = \frac{\sum x_i}{n} - \theta = \bar{x} - \theta = \hat{\theta}$$

\bar{x} : не является оценкой, так как всегда равно 0; $\hat{\theta}$: максимальная правдоподобная оценка, так как всегда больше 0; метод Ньютона: опирается на функцию, которую всегда можно считать, но вычислительно затратен

3 $X_1, \dots, X_n \sim \text{Bern}(\theta)$, $\hat{\theta} = X_1$, $T(x) = \sum_{i=1}^n x_i$ - м.т.; оценить θ с помощью ТН или макс.-правдоподобия

$$T(X) \sim \text{Bin}(n, \theta) \quad (\text{число успехов}), \quad E T(X) = n\theta ; \quad \hat{\theta}_T = \frac{T(X)}{n}, \quad E \hat{\theta}_T = \frac{E T(X)}{n} = \frac{n\theta}{n} = \theta \Rightarrow \text{не смещенная}$$

$$D \hat{\theta} = D X_1 = \theta(1-\theta), \quad D \hat{\theta}_T = D \left(\frac{T(X)}{n} \right) = \frac{1}{n^2} D T(X) = \frac{1}{n^2} n \theta(1-\theta) = \frac{\theta(1-\theta)}{n} < D \hat{\theta} \quad (\text{лучше})$$