

95) $X \sim N(\mu, \sigma^2)$ $H_0: \mu = \mu_0$

$\tau = \mu, \eta = \sigma^2$

P. 64

$L(\mu, \sigma^2) = \frac{1}{(\sqrt{2\pi}\sigma^2)^m} \exp\left\{-\frac{1}{2\sigma^2} \sum (x_i - \mu)^2\right\}$

$\ell(\mu, \sigma^2) = c - \frac{m}{2} \log \sigma^2 - \frac{1}{2\sigma^2} \sum (x_i - \mu)^2$

$U(\mu, \sigma^2) = \left(\underbrace{\frac{\sum (x_i - \mu)}{\sigma^2}}_{U_1}, -\frac{m}{2\sigma^2} + \frac{\sum (x_i - \mu)^2}{2\sigma^4} \right)$

$\hat{\mu} = \bar{X} \quad \hat{\sigma}^2 = \frac{1}{m} \sum (x_i - \bar{X})^2$

$\hat{\theta} = (\hat{\mu}, \hat{\sigma}^2)$

• per $\tau = c_0$: $\tilde{\sigma}^2 = \frac{1}{m} \sum (x_i - \mu_0)^2$

$\tilde{\theta} = (\mu_0, \tilde{\sigma}^2)$

$-E \frac{\partial U}{\partial \theta^T}(\mu, \sigma^2) = \begin{pmatrix} +\frac{m}{\sigma^2} & 0 \\ 0 & \frac{m}{2\sigma^4} \end{pmatrix} = J_m(\theta)$

$J(\theta) = \begin{pmatrix} \frac{1}{\sigma^2} & 0 \\ 0 & \frac{1}{2\sigma^4} \end{pmatrix}$

$J^{-1}(\theta) = \begin{pmatrix} \sigma^2 & 0 \\ 0 & 2\sigma^4 \end{pmatrix}$

$J''(\theta) = \sigma^2$

• $LR := \lambda \cdot \left[\cancel{\ell} - \frac{m}{2} \log \hat{\sigma}^2 - \frac{1}{2\hat{\sigma}^2} \sum (x_i - \bar{X})^2 - \cancel{\ell} + \frac{m}{2} \log \tilde{\sigma}^2 + \frac{1}{2\tilde{\sigma}^2} \sum (x_i - \mu_0)^2 \right]$
 $= m \log \frac{\tilde{\sigma}^2}{\hat{\sigma}^2} = m \log \frac{\sum (x_i - \mu_0)^2}{\sum (x_i - \bar{X})^2}$

• $W = m (\bar{X} - \mu_0)^2 (\hat{\sigma}^2)^{-1}$

• $R = \frac{1}{m} \left(\frac{\sum (x_i - \mu_0)}{\tilde{\sigma}^2} \right)^2 \tilde{\sigma}^2 = m \frac{(\bar{X} - \mu_0)^2}{\tilde{\sigma}^2}$

Kritik: Test annimmt $T_m > \chi^2_{1-\alpha}(1,2)$

96) $X \sim \log N(\mu, \sigma^2)$ $H_0: \mu = \mu_0$

$\tau = \mu, \eta = \sigma^2$

P. 5

$\ell(\mu, \sigma^2) = c - \frac{m}{2} \log \sigma^2 - \frac{1}{2\sigma^2} \sum (\log x_i - \mu)^2$

$U = \left(\frac{\sum (\log x_i - \mu)}{\sigma^2}, -\frac{m}{2\sigma^2} + \frac{\sum (\log x_i - \mu)^2}{2\sigma^4} \right)$

$\hat{\mu} = \frac{1}{m} \sum \log x_i \quad \hat{\sigma}^2 = \frac{1}{2} \sum (\log x_i - \hat{\mu})^2$

$\tilde{\sigma}^2 = \frac{1}{m} \sum (\log x_i - \mu_0)^2$

$J(\theta) = \begin{pmatrix} 1/\sigma^2 & 0 \\ 0 & 1/2\sigma^4 \end{pmatrix} \Rightarrow J''(\theta) = \sigma^2$

• $LR = \dots = m \log \left[\frac{\sum (\log x_i - \mu_0)^2}{\sum (\log x_i - \hat{\mu})^2} \right]$

• $W = m (\hat{\mu} - \mu_0)^2 (\hat{\sigma}^2)^{-1}$

• $R = \frac{1}{m} \left[\frac{\sum (\log x_i - \mu_0)}{\tilde{\sigma}^2} \right]^2 \tilde{\sigma}^2 = m \frac{(\hat{\mu} - \mu_0)^2}{\tilde{\sigma}^2}$

annimmt $T_m > \chi^2_{1-\alpha}(1,2)$

97) $Y|N \sim \text{Bi}(\tilde{m}, p)$ $p \sim \tilde{m} = N$, $N \sim P_0(\lambda)$ $H_0: p = p_0$ $\tau = p$, $\eta = \lambda$ **Gr 70**

$$L(p, \lambda) = \prod \binom{m_i}{y_i} p^{y_i} (1-p)^{m_i - y_i} \frac{\lambda^{m_i} e^{-\lambda}}{m_i!} = c \cdot p^{\sum y_i} (1-p)^{\sum m_i - \sum y_i} \lambda^{\sum m_i} e^{-m\lambda}$$

$$\ell(p, \lambda) = c + \sum y_i \log p + (\sum m_i - \sum y_i) \log(1-p) + \sum m_i \log \lambda - m\lambda$$

$$U(p, \lambda) = \left(\frac{\sum y_i}{p} - \frac{\sum m_i - \sum y_i}{1-p}, \frac{\sum m_i}{\lambda} - m \right) \quad \hat{\lambda} = \frac{\sum m_i}{m} \quad \hat{p} = \frac{\sum y_i}{\sum m_i}$$

$$\hat{\lambda} \text{ maximizes } m\lambda \Rightarrow \tilde{\lambda} = \hat{\lambda}$$

$$J(p, \lambda) = \begin{pmatrix} \frac{1}{p} + \frac{\lambda}{1-p} & 0 \\ 0 & \frac{1}{\lambda} \end{pmatrix} \quad J''(\theta) = \frac{1/1}{(1-p)\lambda + p\lambda} = \frac{p(1-p)}{\lambda}$$

$$\begin{aligned} \bullet LR &= 2 \left[\cancel{\ell} + \sum y_i \log \hat{p} + \sum (N_i - y_i) \log(1-\hat{p}) + \cancel{\sum N_i \log \hat{\lambda} - m\hat{\lambda}} - \cancel{\ell} - \sum y_i \log p_0 - \sum (N_i - y_i) \log(1-p_0) - \cancel{\sum N_i \log \hat{\lambda} + m\hat{\lambda}} \right] \\ &= 2 \left[\sum y_i \log \hat{p}/p_0 + \sum (N_i - y_i) \log[(1-\hat{p})/(1-p_0)] \right] \end{aligned}$$

$$\bullet W = m (\hat{p} - p_0)^2 \left[\frac{\hat{p}(1-\hat{p})}{\lambda} \right]^{-1} = m \frac{(\hat{p} - p_0)^2}{\hat{p}(1-\hat{p})} \cdot \hat{\lambda}$$

$$\begin{aligned} \bullet R &= \frac{1}{m} \left(\frac{\sum y_i}{p_0} - \frac{\sum (N_i - y_i)}{1-p_0} \right)^2 \frac{p_0(1-p_0)}{\hat{\lambda}} = \frac{1}{m} \left(\frac{\sum y_i - p_0 \sum y_i - p_0 \sum N_i + p_0 \sum y_i}{p_0(1-p_0)} \right)^2 \frac{p_0(1-p_0)}{\hat{\lambda}} \\ &= \frac{1}{m} \left(\sum N_i \left(\frac{\sum y_i}{\sum N_i} - p_0 \right) \right)^2 \frac{1}{\hat{\lambda} p_0(1-p_0)} \end{aligned}$$

annäherung an $T_m > \chi^2_{1, (1-\alpha)}$

98) $X \sim M(1, p_1, p_2, p_3, p_4)$ **R 68**

→ • standardisierte Parameterisierung $(p_1, p_2, p_3, p_4)^T = p$ $y_j = \sum_{i=1}^m x_{ij}$ (počet j v rídkach)

$$1) L(p) = \prod p_j^{y_j} \quad \ell(p) = \sum y_j \log p_j \quad p_1 = \tau \quad (p_2, p_3, p_4) = \psi \quad H_0: p_1 = 1/4$$

$$\hat{p}_j = y_j/m \quad \text{pe } \tilde{p} \text{ maximalizuj } \sum y_j \log p_j \quad \text{na podm } p_1 = 1/4, \sum_{j=2}^4 p_j = 3/4$$

$$2) f(p_2, p_3, p_4, \lambda) = y_1 \log 1/4 + y_2 \log p_2 + y_3 \log p_3 + y_4 \log p_4 + \lambda \left(\frac{3}{4} - \sum p_j \right)$$

$$\frac{\partial}{\partial p_j} f = y_j/p_j - \lambda = 0$$

$$\frac{\partial}{\partial \lambda} f = 3/4 - \sum p_j = 0$$

$$\Rightarrow p_j = y_j/\lambda \quad \Rightarrow$$

$$3/4 - 1/\lambda \sum y_j = 0 \quad \Rightarrow \quad \lambda = 4/3 \sum y_j$$

$$\Rightarrow \tilde{p}_j = \frac{y_j}{\sum_{j=2}^4 y_j} \cdot \frac{3}{4} \quad j=2,3,4$$

$$\bullet LR = 2 \left(\sum_{j=1}^4 y_j \log(y_j/m) - y_1 \log 1/4 - \sum_{j=2}^4 y_j \log \left(\frac{y_j \cdot 3/4}{\sum_{j=2}^4 y_j} \right) \right)$$

• Fisher. inf. momentum počítá jeho momenty ale robust'ě to není podm. $\sum_{j=1}^4 p_j = 1$

→ • parameterizácia $(p_1, p_2, p_3, 1-p_1-p_2-p_3)$ $p = (p_1, p_2, p_3)^T$

$$L(p) = \prod_{j=1}^3 p_j^{y_j} \cdot (1-p_1-p_2-p_3)^{m-y_1-y_2-y_3}$$

$$\hat{p}_j = y_j/m$$

$$\tilde{p} = \left(1/4, \frac{y_2}{m-y_1}, \frac{y_3}{m-y_1} \right)^T \quad \Rightarrow$$

$$1 - \sum_{j=1}^3 \hat{p}_j = \frac{y_4}{m} \quad 1 - \sum_{j=1}^3 \tilde{p}_j = \frac{y_4}{m-y_1}$$

$$\bullet LR = 2 \left(\sum_{j=1}^3 y_j \log(y_j/m) + y_4 \log(y_4/m) - y_1 \log 1/4 - \sum_{j=2}^3 y_j \log \left(\frac{y_j \cdot 3/4}{m-y_1} \right) - y_4 \log \left(\frac{y_4}{m-y_1} \right) \right)$$

99) α **R 98** a **R skript** distribucije $\hat{p} = (0,240; 0,258; 0,264; 0,235)$

a) $LR = 60,08$ $u_0 \cdot p_1 = 1/4$ $\chi^2_1(1/4) = 3,84$ $p\text{-val} = 9 \cdot 10^{-11} \Rightarrow$ **numistom**

b) $H_1: p_1 = p_2$ $LR = 40,41$ $p\text{-val} = 2 \cdot 10^{-18} \Rightarrow$ **numistom**

c) $H_1: p_3 = 1,1 p_1$ $LR = 1,39$ $p\text{-val} = 0,239 \Rightarrow$ **numistom**

100) $(x_i, y_i)^T \sim$ **mult. giban zla** $y_i | x_i \sim N(\beta_0 + \beta_1 x_i, \sigma^2)$ x_i **nezav. na** β_0, σ^2
 pe $\sigma^2 = 1$ **one variable** **R 92**

$H_0: \beta_1 = 0$ $H_1: \beta_1 \neq 0$

$L(\beta, \sigma^2) = c \cdot \prod \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{1}{2\sigma^2} (y_i - \beta_0 - \beta_1 x_i)^2\right\} = c(\sigma^2)^{-m/2} \exp\left\{-\frac{1}{2\sigma^2} \sum (y_i - \beta_0 - \beta_1 x_i)^2\right\}$

$\ell(\beta, \sigma^2) = c - \frac{m}{2} \log \sigma^2 - \frac{1}{2\sigma^2} \sum (y_i - \beta_0 - \beta_1 x_i)^2$

$\frac{\partial \ell}{\partial \sigma^2} = -\frac{m}{2\sigma^2} + \frac{1}{2\sigma^4} \sum (y_i - \beta_0 - \beta_1 x_i)^2 \stackrel{!}{=} 0 \Rightarrow \hat{\sigma}^2 = \frac{1}{m} \sum (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2$

$\hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x}$ $\hat{\beta}_1 = \frac{\sum (y_i - \bar{y})(x_i - \bar{x})}{\sum (x_i - \bar{x})^2}$ **pe** $\hat{\beta}_0$ **a** $\hat{\beta}_1$ **u** **R 92**
 (Dugic, Jureković) **Vah 4.1**

na $H_0: \ell(\beta_0, \sigma^2) = c - \frac{m}{2} \log \sigma^2 - \frac{1}{2\sigma^2} \sum (y_i - \beta_0)^2$

$\Rightarrow$ **standardizirano normalno model** $\Rightarrow \tilde{\beta}_0 = \bar{y}$ $\tilde{\sigma}^2 = \frac{1}{m} \sum (y_i - \tilde{\beta}_0)^2$ $\tilde{\beta}_1 = 0$

$LR = 2 \cdot \left[\ell - \frac{m}{2} \log \hat{\sigma}^2 - \frac{1}{2\hat{\sigma}^2} \sum (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2 - \ell + \frac{m}{2} \log \tilde{\sigma}^2 + \frac{1}{2\tilde{\sigma}^2} \sum (y_i - \tilde{\beta}_0)^2 \right]$
 $= \frac{2}{\sigma^2} \left[\log \left[\frac{\tilde{\sigma}^2}{\hat{\sigma}^2} \right] \right]$
 $= m \log \frac{\sum (y_i - \tilde{\beta}_0)^2}{\sum (y_i - \hat{\beta}_0 - \hat{\beta}_1 x_i)^2}$ **numistom** **u** **R** $LR > \chi^2_1(1-\alpha)$

101) **mult. giban** (x_i, y_i) **u** **R 42** **zla**

$P(Y=1|X=x) = \frac{e^{\alpha + \beta x}}{1 + e^{\alpha + \beta x}}$

$P(Y=0|X=x) = 1 - P(Y=1|X=x)$

$\ell(\alpha, \beta) = \sum y_i (\alpha + \beta x_i) - \sum \log(1 + e^{\alpha + \beta x_i})$

α, β **numericky** **u** **R skriptu**

$U(\alpha, \beta) = \begin{pmatrix} \sum y_i - \sum e^{\alpha + \beta x_i} / (1 + e^{\alpha + \beta x_i}) \\ \sum x_i y_i - \sum x_i e^{\alpha + \beta x_i} / (1 + e^{\alpha + \beta x_i}) \end{pmatrix} \} U_2$

na $H_0: \beta = 0$ $\ell(\alpha) = \sum y_i \alpha - \sum \log(1 + e^{\alpha})$
 $U(\alpha) = \sum y_i - \sum \frac{e^{\alpha}}{1 + e^{\alpha}} \stackrel{!}{=} 0 \Rightarrow \bar{y} = \frac{e^{\alpha}}{1 + e^{\alpha}} \Rightarrow \tilde{\alpha} = \log \frac{\bar{y}}{1 - \bar{y}}$
 $\tilde{\beta} = 0$

$LR = 2 \left[\sum y_i (\hat{\alpha} + \hat{\beta} x_i) - \sum \log(1 + e^{\hat{\alpha} + \hat{\beta} x_i}) - \sum y_i \tilde{\alpha} + \sum \log(1 + e^{\tilde{\alpha}}) \right]$

Informacioni matrica: **u** **R 42** **minim** **pe** $X = \begin{pmatrix} 1 & x_1 \\ \vdots & \vdots \\ 1 & x_m \end{pmatrix}$ **a** $W = \text{diag} \left(\frac{e^{\alpha + \beta x_i}}{(1 + e^{\alpha + \beta x_i})^2} \right)$
 w_i

u $J_m(\alpha, \beta) = X' W X = \begin{pmatrix} \sum w_i & \sum w_i x_i \\ \sum w_i x_i & \sum w_i x_i^2 \end{pmatrix}$

$\tau = \beta$, $H_0: \beta = 0$ **pe** J'' **prvob** (42) **matrica** $[J_m(\alpha, \beta) / m]^{-1}$ **zavisi**

$R = \frac{1}{m} [U_2(\tilde{\alpha}, 0)]^2 J''(\tilde{\alpha}, 0)$ $W = m (\hat{\alpha}, \hat{\beta}) - (\tilde{\alpha}, 0)$ $\sim \chi^2_1(1-\alpha)$ **power**