

14)  $X_1 \dots X_m \sim Po(\lambda)$

i)  $T_m = \frac{L_m(\lambda_1)}{L_m(\lambda_0)} = \frac{e^{-m\lambda_1} \lambda_1^{\sum x_i}}{e^{-m\lambda_0} \lambda_0^{\sum x_i}} = e^{-m(\lambda_1 - \lambda_0)} \left(\frac{\lambda_1}{\lambda_0}\right)^{\sum x_i} \geq c$

$-m(\lambda_1 - \lambda_0) + \sum x_i \log(\lambda_1/\lambda_0) \geq \log c$   $\lambda_1 > \lambda_0$   
 $\log \lambda_1/\lambda_0 > 0$   
 $\sum x_i \geq [\log c + m(\lambda_1 - \lambda_0)] / \log(\lambda_1/\lambda_0) =: x$

$x: \text{og } \alpha \stackrel{!}{=} P_{H_0}(\sum X_i \geq x) \Rightarrow x = \text{quantil } (1-\alpha) Po(m\lambda_0)$   
 $\sum X_i \stackrel{H_0}{\sim} Po(m\lambda_0)$

ii)  $\sum X_i \leq x \Rightarrow x = \alpha\text{-quantil } Po(m\lambda_0)$  pursi dest.

iii)  $\hat{\lambda} = \bar{X}$  (G.S. 91)  $\sqrt{m}(\bar{X} - \lambda) \xrightarrow{d} N(0, \lambda)$

$\sqrt{m} \frac{\bar{X} - \lambda_0}{\sqrt{\lambda_0}} \stackrel{H_0}{\sim} N(0, 1)$  test:  $\sqrt{m} \frac{|\bar{X} - \lambda_0|}{\sqrt{\lambda_0}} \leq \mu_{1-\alpha/2}$

resp  $\sqrt{m} \frac{|\bar{X} - \lambda_0|}{\sqrt{\bar{X}}} \geq \mu_{1-\alpha/2}$  resp  $\frac{m(\bar{X} - \lambda_0)^2}{\bar{X}} \geq \chi_{1-\alpha}^2$  asympt. dest.

iv)  $LR_m \Rightarrow \frac{L_m(\bar{X})}{L_m(\lambda_0)} \Rightarrow LR_m = 2[-m(\bar{X} - \lambda_0) + \sum X_i \log(\bar{X}/\lambda_0)] \geq \chi_{1-\alpha}^2$

17)  $X_1 \dots X_m \sim N(\mu_x, \sigma_x^2)$

i) MV:  $\hat{\mu} = \bar{X}, \hat{\sigma}^2 = \frac{1}{m} \sum (X_i - \bar{X})^2$   
 $l_m(\hat{\mu}, \hat{\sigma}^2) = c - \frac{m}{2} \log \hat{\sigma}^2 - \frac{1}{2\hat{\sigma}^2} \sum (X_i - \bar{X})^2 = c - \frac{m}{2} \log \hat{\sigma}^2 - \frac{m}{2}$

$m H_0: \mu_x = \mu_0$

$\tilde{l}_m(\sigma^2) = c - \frac{m}{2} \log \sigma^2 - \frac{1}{2\sigma^2} \sum (X_i - \mu_0)^2$

$\tilde{U}_m(\sigma^2) = \frac{m}{2\sigma^2} + \frac{\sum (X_i - \mu_0)^2}{2\sigma^4} = 0 \Rightarrow \tilde{\sigma}^2 = \frac{1}{m} \sum (X_i - \mu_0)^2$

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

$LR_m = 2[-\frac{m}{2} \log \hat{\sigma}^2 + \frac{m}{2} \log \tilde{\sigma}^2] = m \log \left[ \frac{\tilde{\sigma}^2}{\hat{\sigma}^2} \right] \geq \chi_{1-\alpha}^2$

t-test:  $\sqrt{m}(\bar{X} - \mu) \xrightarrow{d} N(0, \sigma^2)$   $\frac{\sqrt{m}(\bar{X} - \mu_0)}{\sqrt{S^2}} \stackrel{H_0}{\sim} t_{m-1}$

ii)  $m H_0: \sigma_x^2 = \sigma_0^2$

$\tilde{l}_m(\mu) = c - \frac{m}{2} \log \sigma_0^2 - \frac{1}{2\sigma_0^2} \sum (X_i - \mu)^2$

$\tilde{U}_m(\mu) = \frac{1}{\sigma_0^2} \sum (X_i - \mu) = 0 \Rightarrow \tilde{\mu} = \frac{1}{m} \sum X_i$

$\tilde{l}_m(\tilde{\mu}) = c - \frac{m}{2} \log \sigma_0^2 - \frac{1}{2\sigma_0^2} \sum (X_i - \bar{X})^2$

$LR_m = 2[-\frac{m}{2} \log \hat{\sigma}^2 + \frac{m}{2} \log \sigma_0^2 - \frac{m}{2} + \frac{1}{2\sigma_0^2} \sum (X_i - \bar{X})^2] = m \log \left[ \frac{\sigma_0^2}{\hat{\sigma}^2} \right] + m \left[ \frac{\hat{\sigma}^2}{\sigma_0^2} - 1 \right] \geq \chi_{1-\alpha}^2$

oproti tomu  $\frac{(m-1)S_m^2}{\sigma_0^2} \stackrel{H_0}{\sim} \chi_{m-1}^2$        $\frac{(m-1)S_m^2}{\sigma_0^2} = \frac{\sum (X_i - \bar{X})^2}{\sigma_0^2} = \frac{m \hat{\sigma}^2}{\sigma_0^2}$

ii) na  $H_0$ :  $l_m(\hat{\mu}, \hat{\sigma}^2) = c - \frac{m}{2} \log \sigma_0^2 - \frac{1}{2\sigma_0^2} \sum (X_i - \mu_0)^2$   
 $LR_m = 2 \left[ -\frac{m}{2} (\log \hat{\sigma}^2 - \log \sigma_0^2) - \frac{m}{2} + \frac{1}{2\sigma_0^2} \sum (X_i - \mu_0)^2 \right] = m \log \left[ \frac{\sigma_0^2}{\hat{\sigma}^2} \right] - m + \frac{\sum (X_i - \mu_0)^2}{\sigma_0^2}$   
 $\geq \chi_2^2(1-\alpha)$

~~ii) 2 Pr 63:  $\Gamma_m \left( \begin{pmatrix} \bar{X} \\ \hat{\sigma}^2 \end{pmatrix} - \begin{pmatrix} \mu \\ \sigma^2 \end{pmatrix} \right) \xrightarrow{d} N_2 \left( \begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma^2 & 0 \\ 0 & 2\sigma^4 \end{pmatrix} \right)$   
Analiz Vzh 4.15:  $X \sim N_2(\mu, V)$ ,  $h(V) = 2 \Rightarrow (X-\mu)' V^{-1} (X-\mu) \sim \chi_2^2$   
 obecně  $X \sim N_m(\mu, V)$ ,  $h(V) = r \Rightarrow (X-\mu)' V^{-1} (X-\mu) \sim \chi_r^2$~~

~~$V = J^{-1}(\mu, \sigma^2) / m$        $V^{-1} = m J(\mu, \sigma^2)$   
 $m \left( \begin{pmatrix} \bar{X} \\ \hat{\sigma}^2 \end{pmatrix} - \begin{pmatrix} \mu_0 \\ \sigma_0^2 \end{pmatrix} \right)' J(\mu_0, \sigma_0^2) \left( \begin{pmatrix} \bar{X} \\ \hat{\sigma}^2 \end{pmatrix} - \begin{pmatrix} \mu_0 \\ \sigma_0^2 \end{pmatrix} \right) \stackrel{H_0}{\sim} \chi_2^2$   
 odhad  $J(\hat{\mu}, \hat{\sigma}^2) \Rightarrow W_m = m \left[ \frac{(\bar{X} - \mu_0)^2}{\hat{\sigma}^2} + \frac{(\hat{\sigma}^2 - \sigma_0^2)^2}{2\hat{\sigma}^4} \right] \geq \chi_2^2(1-\alpha)$~~

78)  $X \sim \text{Mult}(m; p_1, p_2, p_3, p_4)$        $H_0: p_1 = p_2 = p_3 = p_4 = 1/4$        $H_1: \neq H_0$   
 $L_m(p) = c p_1^{x_1} p_2^{x_2} p_3^{x_3} p_4^{x_4}$       MV: P. 68       $p_j = \frac{\sum_{i=1}^m X_{ij}}{m}$        $p = \frac{\sum X_i / m}{x}$   
 $L_m(p_0) = c \left( \frac{1}{4} \right)^m$   
 $LR_m = 2 \left[ \sum_{j=1}^4 x_j \log \left( \frac{x_j}{m} \right) - m \log \frac{1}{4} \right] = 2 \sum_{j=1}^4 x_j \log \left[ \frac{4x_j}{m} \right] \stackrel{H_0}{\sim} \chi_3^2(1-\alpha)$   
 $= 324,15 > \chi_3^2(1-\alpha) = 7,8$       namíche  $H_0$

80)  $L_m(\alpha, \beta) = \prod_{i=1}^m \frac{\lambda(x_i)^{\alpha_i} e^{-\lambda(x_i)}}{\alpha_i!}$        $f_X(x_i) = \frac{\pi \lambda(x_i)^{\alpha_i} e^{-\sum \lambda(x_i)}}{\pi \alpha_i!}$        $\frac{\partial \lambda}{\partial \alpha} = \lambda$        $\frac{\partial \lambda}{\partial \beta} = \lambda x$   
 $L_m(\alpha, \beta) = \sum_{i=1}^m \alpha_i \log \lambda(x_i) - \sum_{i=1}^m \lambda(x_i) + c$        $\lambda(x) = \alpha + \beta x$        $e^{\alpha + \beta x}$   
 $\frac{\partial L_m}{\partial \alpha} = \sum \frac{\alpha_i x_i}{\lambda(x_i)} - \sum \alpha_i = 0$        $\frac{\partial L}{\partial \alpha} = \sum \alpha_i - m \bar{\alpha} = 0$   
 $\frac{\partial L_m}{\partial \beta} = \sum \frac{\alpha_i x_i}{\lambda(x_i)} x_i - \sum \alpha_i x_i = 0$        $\frac{\partial L}{\partial \beta} = \sum \alpha_i x_i - m \bar{\alpha} \bar{x} = 0$   
 $\bar{\alpha} = \log \bar{y}$   
 $LR_m = 2 \left[ \sum \alpha_i \log(\hat{\alpha} + \hat{\beta} x_i) - \sum \exp(\hat{\alpha} + \hat{\beta} x_i) - \sum \alpha_i \log(\bar{\alpha}) + m \bar{\alpha} \right] \geq \chi_{qr}^2(1-\alpha)$   
 $\hat{\alpha} = -1,185$        $\hat{\beta}_1 = 0,51$        $\hat{\beta}_2 = 0,33$   
 $\hat{\alpha} = -0,52$        $\hat{\beta}_1 = -0,51$        $\hat{\beta}_2 = 1,87$   
 $LR_m = 11,81 > \chi_2^2(0,95) = 5,99$       namíche  $H_0: (\beta_1, \beta_2)' = (0, 0)$

$\chi_2^2(1-\alpha)$