

49) $X \sim \text{alt}(p)$

i) $L(p) = \prod p^{x_i} (1-p)^{1-x_i} = p^{\sum x_i} (1-p)^{m-\sum x_i}$

$l(p) = \sum x_i \log p + (m - \sum x_i) \log(1-p)$

$l'(p) = \sum x_i / p + (m - \sum x_i) / (1-p) \stackrel{!}{=} 0$

$(1-p) \sum x_i - (m - \sum x_i) p = 0$

$p(-\sum x_i - m + \sum x_i) = -\sum x_i$

$\hat{p} = \bar{X}$

$l''(p) = -\frac{\sum x_i}{p^2} - \frac{(m - \sum x_i)}{(1-p)^2}$

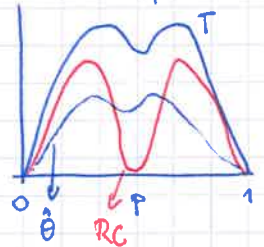
$J_m(p) = \frac{m}{p} + \frac{m}{1-p} = \frac{m}{p(1-p)}$

$\Gamma_m(\bar{X} - p) \xrightarrow{D} N(0, \frac{p(1-p)}{m})$

ii) invariance: $\theta = p(1-p)$ $\hat{\theta} = \hat{p}(1-\hat{p}) = \bar{X}(1-\bar{X})$ $g(t) = t(1-t)$ $g'(t) = (1-t) - t = 1-2t$

$\Gamma_m(\bar{X}(1-\bar{X}) - \theta) \xrightarrow{D} N(0, \frac{(1-2p)^2 p(1-p)}{m})$ $p \neq 1/2$

iii) \bar{X} je NNO, NNO θ je $T = \frac{m}{m-1} \bar{X}(1-\bar{X})$, $\hat{\theta}$ ni je nezahvisni
je efektivni, ni je efektivni



50) $X \sim \text{Po}(\lambda)$ $L(\lambda) = e^{-m\lambda} \lambda^{\sum x_i} / \prod x_i!$ $l(\lambda) = -m\lambda + \sum x_i \log \lambda + c$

i) $l'(\lambda) = -m + \sum x_i / \lambda = 0$ $\hat{\lambda} = \bar{X}$

$l''(\lambda) = -\sum x_i / \lambda^2$

$J_m(\lambda) = m/\lambda$

$\Gamma_m(\bar{X} - \lambda) \xrightarrow{D} N(0, \lambda)$

ii) $\theta = e^{-\lambda}$ $g(t) = e^{-t}$ $g'(t) = -e^{-t}$

$\Gamma_m(e^{-\bar{X}} - e^{-\lambda}) \xrightarrow{D} N(0, e^{-2\lambda} \lambda)$

iii) \bar{X} je NNO, NNO θ je $T = (1 - \frac{1}{m})^{\sum x_i}$
je efektivni, ni je efektivni

51) $X \sim \text{Exp}(\lambda)$ $L(\lambda) = \lambda^m e^{-\lambda \sum x_i}$ $l(\lambda) = m \log \lambda - \lambda \sum x_i$ $l'(\lambda) = m/\lambda - \sum x_i = 0$

i) $\hat{\lambda} = 1/\bar{X}$ $l''(\lambda) = -m/\lambda^2$ $J_m(\lambda) = m/\lambda^2$

ii) $\Gamma_m(1/\bar{X} - \lambda) \xrightarrow{D} N(0, \lambda^2)$

iii) $\frac{m-1}{\sum x_i}$ je NNO λ ale ni je efektivni

52) $X \sim \text{Ge}(p)$ $L(p) = p^m (1-p)^{\sum x_i}$ $l(p) = m \log p + \sum x_i \log(1-p)$

i) $l'(p) = m/p - \sum x_i / (1-p) = 0$

$l''(p) = -m/p^2 - \sum x_i / (1-p)^2$

$m - mp - p \sum x_i = 0$

$J_m(p) = m/p^2 - \frac{\sum x_i}{p} \frac{m}{(1-p)^2} = m(\frac{1}{p^2} - \frac{1}{p(1-p)})$

$p(m + \sum x_i) = m$

$\hat{p} = \frac{m}{m + \sum x_i} = \frac{1}{1 + \bar{X}}$

$\Gamma_m(\frac{1}{1+\bar{X}} - p) \xrightarrow{D} N(0, \frac{1-2p}{p^2(1-p)})$

ii) $\theta = p(1-p)$ $\hat{\theta} = \frac{1}{1+\bar{X}} (1 - \frac{1}{1+\bar{X}}) = \frac{1-\bar{X}}{(1+\bar{X})^2}$

$g(t) = t(1-t)$ $g'(t) = 1-2t$ $\Gamma_m(\hat{\theta} - \theta) \xrightarrow{D} N(0, \frac{(1-2p)^2 \cdot (1-2p)}{p(1-p)})$

53) $X \sim R(\theta - 1/2, \theta + 1/2)$ $L(\theta) = \prod I[\theta - 1/2 < x_i < \theta + 1/2] = I[\theta - 1/2 < \min x_i] I[\max x_i < \theta + 1/2]$

= $\begin{cases} 1 & \text{al } \theta < \min x_i + 1/2 \text{ a } \theta > \max x_i - 1/2 \\ 0 & \text{inad} \end{cases} \Rightarrow \hat{\theta}$ je zvezda a hodnot $[\max x_i - 1/2, \min x_i + 1/2]$

ii) a MSE: $\min x_i \xrightarrow{P} \theta - 1/2$ $\Rightarrow \max x_i - 1/2 \xrightarrow{P} \theta$ \Rightarrow zvezd' velhod je
 $\max x_i \xrightarrow{P} \theta + 1/2$ $\Rightarrow \min x_i + 1/2 \xrightarrow{P} \theta$ \Rightarrow dalje zvezd' velhod

54) $X \sim R\{1..M\}$ $L(M) = \prod \frac{1}{M} I[x_i \in \{1..M\}] = \frac{1}{M^m} I[1 \leq \min x_i \leq \max x_i \leq M] =$

$\begin{cases} 1/M^m & \text{al } M \geq \max x_i \\ 0 & \text{inad} \end{cases} \Rightarrow \hat{M} = \max x_i$

ii) $P(\hat{M} \leq z) = P(X_i \leq z)^m = (\frac{z}{M})^m$ $f_{\hat{M}}(z) = m z^{m-1} / M^m = P(\hat{M} \leq M - \varepsilon)$

$P(|\hat{M} - M| > \varepsilon) = P(\hat{M} < M - \varepsilon) \leq P(\hat{M} \leq M - \varepsilon) \stackrel{\uparrow}{\leq} (\frac{M - \varepsilon}{M})^m = (1 - \frac{\varepsilon}{M})^m \xrightarrow{m \rightarrow \infty} 0$
al $\varepsilon \in \mathbb{R}$

55) $Y_i \sim N(\theta x_i, 1)$ $L(\theta) = c \exp\{-\frac{1}{2} \sum (y_i - \theta x_i)^2\}$ $l(\theta) = c - \frac{1}{2} \sum (y_i - \theta x_i)^2$ iii)

i) $l'(\theta) = \sum (y_i - \theta x_i) x_i = 0$

$\sum y_i x_i = \theta \sum x_i^2$

$\frac{\sum y_i x_i}{\sum x_i^2} = \hat{\theta}$

ii) $E\hat{\theta} = \frac{1}{\sum x_i^2} \sum x_i EY_i = \theta$ nezahvisni

$l'' = -\sum x_i^2$

$RC_m = 1/\sum x_i^2$

$mn\hat{\theta} = \frac{\sum x_i^2 \cdot 1}{(\sum x_i^2)^2} = \frac{1}{\sum x_i^2} = RC$

$$56) P(X=k) = \frac{1}{1-(1-p)^m} \binom{m}{k} p^k (1-p)^{m-k} \quad k=1,2,\dots,m$$

$$i) L(p) = \frac{1}{(1-(1-p)^m)^m} \prod_{i=1}^m \binom{m}{x_i} p^{\sum x_i} (1-p)^{\sum (m-x_i)}$$

$$l(p) = -m \log(1-(1-p)^m) + c + \sum x_i \log p + \sum (m-x_i) \log(1-p)$$

$$l'(p) = \frac{-m}{1-(1-p)^m} \cdot m(1-p)^{m-1} + \frac{\sum x_i}{p} - \frac{\sum (m-x_i)}{1-p} = 0$$

$$l''(p) = \text{mathematica script}$$

$$57) X \sim \theta x^{\theta-1} e^{-x^\theta} \mathbb{I}(x>0) \quad L(\theta) = \theta^m (\prod x_i)^{\theta-1} e^{-\sum x_i^\theta} \mathbb{I}(\min x_i > 0)$$

$$i) l(\theta) = m \log \theta + (\theta-1) \log(\prod x_i) - \sum x_i^\theta$$

$$l'(\theta) = \frac{m}{\theta} + \sum \log x_i - \sum x_i^\theta \log x_i = 0$$

$$l''(\theta) = -\frac{m}{\theta^2} - \sum x_i^\theta (\log x_i)^2$$

spojiti iterativna fca, $l'(0) = \infty$ jedini
 $l'(\infty) = -\infty$ rjesenje

$$ii) EX = \int_0^\infty \theta x^\theta e^{-x^\theta} dx \stackrel{x^\theta = t}{=} \int_0^\infty t^{1/\theta} e^{-t} dt = \Gamma(\frac{1}{\theta} + 1)$$

$$E X^\theta (\log X)^2 = \text{mathematica} = c'$$

$$\Gamma_m(\hat{\theta} - \theta) \rightarrow N(0, (\frac{1}{\theta^2} + c')^{-1})$$

$$58) f(x) = \frac{e^{-(x-\theta)}}{(1+e^{-(x-\theta)})^2} \quad L(\theta) = e^{-\sum x_i + m\theta} / \prod (1+e^{-(x_i-\theta)})^2$$

$$l(\theta) = -\sum x_i + m\theta - 2 \sum \log(1+e^{-(x_i-\theta)})$$

$$i) l'(\theta) = m - 2 \sum \frac{1 \cdot e^{-(x_i-\theta)}}{1+e^{-(x_i-\theta)}} = m - 2 \sum \frac{e^{-x_i}}{e^{-\theta} + e^{-x_i}} = 0$$

$$l''(\theta) = +2 \sum \frac{e^{-x_i} (-e^{-\theta})}{(e^{-\theta} + e^{-x_i})^2}$$

spojiti iter. fca jedini
 $l'(-\infty) = m$ rjesenje
 $l'(\infty) = -m$

$$ii) J_m(\theta) = \frac{m}{3}$$

$$E \frac{2e^{-x} e^{-\theta}}{(e^{-\theta} + e^{-x})^2} = \int_{\mathbb{R}} \frac{2e^{-x} e^{-\theta}}{(e^{-\theta} + e^{-x})^2} \frac{e^{-x} e^{\theta}}{(1+e^{-(x-\theta)})^2} dx = \frac{2e^{-2\theta}}{e^{-2\theta}} \int_1^\infty \frac{(t-1)}{t^4} = 2 \left[\frac{-1}{3t^3} + \frac{1}{2t^2} \right]_1^\infty = \frac{1}{3}$$

$$\begin{aligned} 1+e^{-(x-\theta)} = t & \quad t-1 = e^{-x} e^{\theta} \\ (-1) \cdot e^{-(x-\theta)} dx = dt & \quad (t-1)e^{-\theta} = e^{-x} \\ -e^\theta e^{-x} dx = dt & \end{aligned}$$

$$\Gamma_m(\hat{\theta} - \theta) \rightarrow N(0, 3)$$

$$59) X \sim N(\theta, \theta^2) \quad L(\theta) = c \cdot \theta^{-m} \exp\left\{-\frac{1}{2\theta^2} \sum (x_i - \theta)^2\right\} \quad l(\theta) = -m \log \theta + c - \frac{1}{2\theta^2} \sum (x_i - \theta)^2$$

$$i) l'(\theta) = -\frac{m}{\theta} + \frac{1}{\theta^3} \sum (x_i - \theta) + \frac{1}{\theta^5} \sum (x_i - \theta)^2 = 0$$

$$-m + \frac{\sum x_i - m\theta}{\theta} + \frac{\sum x_i^2 - 2\theta \sum x_i + m\theta^2}{\theta^2} = 0$$

$$-m + \frac{\sum x_i}{\theta} - m + \frac{\sum x_i^2}{\theta^2} - \frac{2\sum x_i}{\theta} + m = 0 +$$

$$-m\theta^2 + \theta \sum x_i - 2\theta \sum x_i + \sum x_i^2 = 0 \quad \frac{-\sum x_i \pm \sqrt{(\sum x_i)^2 + 4 \sum x_i^2 m}}{2m} = \hat{\theta}$$

$$ii) \hat{\theta} = \frac{1}{2} \left[-\frac{1}{2} \bar{x} + \frac{1}{2} \sqrt{(\bar{x})^2 + 4 \frac{1}{m} \sum x_i^2} \right] \xrightarrow{m \rightarrow \infty} -\frac{\theta}{2} + \frac{1}{2} \sqrt{\theta^2 + 4 \cdot 2\theta^2} = -\frac{\theta}{2} + \frac{1}{2} \sqrt{9\theta^2} = \theta$$

$$iii) l''(\theta) = \frac{m}{\theta^2} + \left(-\frac{2}{\theta^3} \sum (x_i - \theta) + \frac{1(-m)}{\theta^2} \right) + \left(\frac{-3}{\theta^4} \sum (x_i - \theta)^2 - \frac{2 \sum (x_i - \theta)}{\theta^3} \right)$$

$$J_m(\theta) = -\frac{m}{\theta^2} + \frac{m}{\theta^2} + \frac{3m}{\theta^2} = \frac{3m}{\theta^2}$$

$$\Gamma_m(\hat{\theta} - \theta) \xrightarrow{D} N(0, \frac{\theta^2}{3})$$

60) $P(Y=y|X) = \frac{\lambda(x)^y e^{-\lambda(x)}}{y!}$ $\lambda(x) = e^{\beta x}$, X unabhängig von β

i) $L(\beta) = \prod P(Y_i = y_i | X = x_i) \cdot P(X = x_i) = \prod P(Y_i = y_i | X_i = x_i)$
 $= \prod_{i=1}^m \frac{(e^{\beta x_i})^{y_i} e^{-e^{\beta x_i}}}{y_i!} P(X = x_i)$

$l(\beta) = \sum (y_i x_i \beta - e^{\beta x_i} + \log(y_i!)) + \log P(X = x_i)$

$= \sum y_i x_i \cdot \beta - \sum e^{\beta x_i} + c$

$l'(\beta) = \sum x_i y_i - \sum e^{\beta x_i} \cdot x_i \stackrel{!}{=} 0$

$l''(\beta) = - \sum e^{\beta x_i} x_i^2$ $J_m(\beta) = m E e^{\beta X} \cdot X^2$ $\Gamma_m(\hat{\beta} - \beta) \xrightarrow{D} N(0, \frac{1}{E e^{\beta X} X^2})$

61) $P(X=x) = \begin{cases} p & x \in \{-1, 1\} \\ 1-2p & x=0 \end{cases}$ $d_{ij} : y_j = \# \{X_i = 0\}$

$L(p) = (1-2p)^d p^{m-d}$ $l(p) = d \log(1-2p) + (m-d) \log p$

$l'(p) = \frac{-2d}{1-2p} + \frac{m-d}{p} = 0$

$-2pd + m - 2mp - d + 2pd = 0$

$\hat{p} = (d - m) / (-2m) = (m - d) / 2m = \sum I[X_i \neq 0] / 2m$

$l''(p) = \frac{-2d \cdot 2}{(1-2p)^2} - \frac{m-d}{p^2}$

$EY = E \sum I[X=0] = m P(X=0) = m(1-2p)$

$J_m(p) = \frac{4m(1-2p)}{(1-2p)^2} + \frac{m - m(1-2p)}{p^2} = \frac{4m}{1-2p} + \frac{2m}{p}$
 $= \frac{4mp + 2m - 4mp}{(1-2p)p} = \frac{m \cdot 2}{p(1-2p)}$

$\Gamma_m(\hat{p} - p) \xrightarrow{D} N(0, p(1-2p)/2)$

62) $L(\theta) = \prod \frac{1}{2} \exp(-|x_i - \theta|) = \frac{1}{2^m} \exp\{-\sum |x_i - \theta|\}$

$l(\theta) = c - \sum |x_i - \theta| \Rightarrow$ maximalisiert $-\sum |x_i - \theta|$ wrt θ

minimiert $\sum |x_i - \theta|$ Median $\{x_i\}$. (siehe online) **Lemma 2.4. MSI**

MLE: Vollständig parameter

63) $X \sim N(\mu, \sigma^2)$

i) $L(\mu, \sigma^2) = c \cdot (\sigma^2)^{-m} \exp\{-\frac{1}{2\sigma^2} \sum (x_i - \mu)^2\}$

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

$\nabla l = \left(\frac{1}{\sigma^2} \sum (x_i - \mu) \quad ; \quad -\frac{m}{2\sigma^2} + \frac{1}{2(\sigma^2)^2} \sum (x_i - \mu)^2 \right)$

$\frac{\partial l}{\partial \mu} = 0 \Rightarrow \sum x_i = m\mu$
 $\hat{\mu} = \bar{x}$

$\frac{\partial l}{\partial \sigma^2} = 0 \Rightarrow m = \frac{\sum (x_i - \mu)^2}{\sigma^2}$, $\hat{\sigma}^2 = \frac{1}{m} \sum (x_i - \mu)^2$ $\mu = \hat{\mu}$

ii) $H_{\mu, \sigma^2} = \begin{pmatrix} -\frac{m}{\sigma^2} & -\frac{\sum (x_i - \mu)}{\sigma^4} \\ -\frac{\sum (x_i - \mu)}{\sigma^4} & \frac{m}{2\sigma^4} - \frac{\sum (x_i - \mu)^2}{(\sigma^2)^3} \end{pmatrix}$

$J_m(\mu, \sigma^2) = \begin{pmatrix} m/\sigma^2 & 0 \\ 0 & m/2\sigma^4 \end{pmatrix}$ $\Gamma_m \left(\begin{pmatrix} \bar{x} \\ \frac{1}{m-1} S^2 \end{pmatrix} - \begin{pmatrix} \mu \\ \sigma^2 \end{pmatrix} \right) \xrightarrow{D} N_2 \left(\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} \sigma^2 & 0 \\ 0 & 2\sigma^4 \end{pmatrix} \right)$