

~~ВВХ~~

ВЗ 2

① $f(x; \lambda) = \frac{\lambda}{2} e^{-\lambda(|x|-1)}, \lambda > 0$

а) $L(\lambda, x) = \prod_{i=1}^n f_{x_i}(X_i; \lambda) = \prod_{i=1}^n \frac{\lambda}{2} e^{-\lambda(|x_i|-1)}$

$$\ln L(\lambda, x) = \sum_{i=1}^n \ln \left[\frac{\lambda}{2} e^{-\lambda(|x_i|-1)} \right] = \sum \left[\ln \frac{\lambda}{2} - \lambda(|x_i|-1) \right] =$$
$$= n \ln \frac{\lambda}{2} - \lambda \sum (|x_i|-1)$$

$$\frac{\partial}{\partial \lambda} \ln L(\lambda, x) = n \frac{1}{\lambda} - \sum (|x_i|-1) = 0 \Rightarrow \hat{\lambda} = \frac{n}{\sum (|x_i|-1)}$$

б) $E \hat{\lambda} = E \frac{n}{\sum (|x_i|-1)} = \frac{n}{E \sum (|x_i|-1)} = \frac{n}{n \frac{1}{\lambda}} = \lambda$

т.е. $E(|x_i|-1) = \frac{1}{\lambda}$ по распредел. оценка эффективна

в) $\frac{1}{n} \sum_{i=1}^n (|x_i|-1) \xrightarrow{P} E[|x|-1] = \frac{1}{\lambda} \quad (ЗБЧ)$

$$\hat{\lambda} = \frac{1}{\frac{1}{n} \sum (|x_i|-1)} \xrightarrow{P} \frac{1}{\frac{1}{\lambda}} = \lambda \Rightarrow \hat{\lambda} \text{ состоятельная}$$

г) ~~$E(|x_i|-1) = \frac{1}{\lambda}$, $E|x| = \frac{1}{\lambda} + 1 \Rightarrow \hat{\lambda} = \frac{1}{\frac{1}{\lambda} + 1}$~~

~~$E[X] = 0$~~ $E[X] = 0$, $E[X^2] = \int_{-\infty}^{+\infty} x^2 \frac{\lambda}{2} e^{-\lambda(|x|-1)} dx =$

$$= 2 \frac{\lambda}{2} \int_0^{\infty} x^2 e^{-\lambda(x-1)} dx = \lambda \int_0^{\infty} x^2 e^{-\lambda(x-1)} dx = \lambda e^{\lambda} \int_0^{\infty} x^2 e^{-\lambda x} dx$$

$$= \frac{2}{\lambda^3} \cdot \lambda e^{\lambda} = \frac{2e^{\lambda}}{\lambda^2} ?$$

Скорее всего в условии опечатка, без заданной функции $f(x, \lambda)$ не совб. условию нормировки

При условии $f(x, \lambda) = \frac{\lambda}{2} e^{-\lambda(|x|-1)}$:

$$E[X^2] = \int_{-\infty}^{+\infty} x^2 \frac{\lambda}{2} e^{-\lambda(|x|-1)} dx = \frac{\lambda}{2} \int_0^{\infty} x^2 e^{-\lambda x} dx = \frac{\lambda}{2} \frac{2}{\lambda^3} = \frac{1}{\lambda^2}$$

$$\overline{X^2} = \frac{1}{\lambda^2} \Rightarrow \hat{\lambda}^2 = \frac{1}{\overline{X^2}}, \lambda > 0 \Rightarrow \hat{\lambda} = \frac{1}{\sqrt{\overline{X^2}}}$$

$$e) \Sigma(\rho) = E[S^2(\lambda, x)] = E\left(\frac{1}{\lambda} - (|x|-1)\right)^2 = E\{Y = |x|-1\}$$

$$\approx EY = \frac{1}{\lambda} \quad DY = \frac{1}{\lambda^2} \quad E\left(\frac{1}{\lambda} - Y\right) = 0$$

$$\Sigma(\rho) = E\left(\frac{1}{\lambda} + 1 - |x|\right)^2 = E\left(\left(\frac{1}{\lambda} + 1\right)^2 - 2\left(\frac{1}{\lambda} + 1\right)|x| + |x|^2\right) =$$

$$= \left(\frac{1}{\lambda} + 1\right)^2 - 2\left(\frac{1}{\lambda} + 1\right) = E\left(\frac{1}{\lambda^2} - 2\frac{1}{\lambda}(|x|-1) + (|x|-1)^2\right) =$$

$$= \frac{1}{\lambda^2} - \frac{2}{\lambda} \cdot \frac{1}{\lambda} + E(|x|-1)^2 = -\frac{1}{\lambda^2} + E|x|^2 - 2E|x| + 1 =$$

$$= -\frac{1}{\lambda^2} + \frac{2e}{\lambda^2} - (2E|x| - 2) - 1 = -\frac{1}{\lambda^2} + \frac{2e}{\lambda^2} - \frac{2}{\lambda} - 1$$

$$E[X] = \frac{2e}{\lambda^2}$$

если строго по условию

$$V2) f_{X|a}(x) = \frac{1}{\Gamma(a)} \theta^a x^{a-1} e^{-x\theta} \quad 1, x > 0$$

$$a) L(\theta, x) = \prod f_{X|a}(x_i, \theta) \quad \ln L(\theta, x) = -\ln \Gamma(a) \cdot 2 \ln \theta + (2-1) \ln x - \frac{x}{\theta}$$

$$\Sigma_n(\theta) = E(S'(\theta, x)), \quad \hat{\theta} = E\left(\frac{\partial}{\partial \theta} \ln L(\theta, x)\right)$$

$$\frac{\partial}{\partial \theta} \ln L(\theta, x) = -\frac{1}{\theta} + \frac{x}{\theta^2} \quad E(X) = 2\theta$$

$$E\left(\frac{\partial}{\partial \theta} \ln L(\theta, x)\right) = \frac{1}{\theta^2} - \frac{2}{\theta^2} E(X) = \frac{1}{\theta^2} - \frac{2}{\theta^2} 2\theta = -\frac{1}{\theta^2}$$

$$\hat{\theta} = \frac{1}{\theta^2} \quad \Sigma_n(\theta) = \frac{n}{\theta^2}$$

$$b) \frac{\partial}{\partial \theta} \ln L(\theta) = -\frac{n}{\theta} + \frac{1}{\theta^2} \sum X_i = 0$$

$$\hat{\theta} = \frac{1}{n} \sum X_i = \frac{\bar{X}}{2}$$

$$E \hat{\theta} = E \frac{\bar{X}}{2} = \frac{1}{2} 2\theta = \theta \Rightarrow \text{несмещенная}$$

$$D(\hat{\theta}) = D\left(\frac{\bar{X}}{2}\right) = \frac{1}{4} D\bar{X} = \frac{1}{4} \cdot \frac{D\bar{X}}{n} = \frac{1}{4} \cdot \frac{2\theta^2}{n} = \frac{\theta^2}{2n}$$

$$\text{по теореме Рао-Крамера: } D(\hat{\theta}) \geq \frac{1}{\Sigma_n(\theta)} = \frac{\theta^2}{2n}$$

получается, что $D(\hat{\theta})$ — мин из возможных $D(\hat{\theta})$
 \Rightarrow оценка эффективная

$$c) \sqrt{n}(\hat{\theta} - \theta) \xrightarrow{d} \mathcal{N}\left(0, \frac{1}{\Sigma_n(\theta)}\right) = \mathcal{N}\left(0, \frac{\theta^2}{2}\right)$$

(13) $T \sim \exp(1)$, $F(t) = 1 - e^{-t}$

a) $p_1 = P(T \leq 1) = 1 - e^{-1}$

$p_2 = P(1 < T \leq 2) = e^{-1} - e^{-2}$

$p_3 = P(T > 2) = e^{-2}$

$L(\lambda, n) = \prod p_i^{n_i} = p_1^{n_1} p_2^{n_2} p_3^{n_3}$, $n_3 = n - n_1 - n_2$

$\ln L(\lambda, n) = n_1 \ln(1 - e^{-\lambda}) + n_2 \ln(e^{-\lambda} - e^{-2\lambda}) + n_3 \ln(e^{-2\lambda}) =$

$= (n_1 + n_2) \ln(1 - e^{-\lambda}) - \lambda n_2 - 2\lambda n_3 + c$

$\frac{\partial}{\partial \lambda} L(\lambda, n) = (n_1 + n_2) \frac{e^{-\lambda}}{1 - e^{-\lambda}} - n_2 - 2n_3 = 0$

$\frac{n_1 + n_2}{e^{\hat{\lambda}} - 1} = n_2 + 2n_3$

$\hat{\lambda} = \ln \left(\frac{n_1 + 2n_2 + 2n_3}{n_2 + 2n_3} \right) = \ln \left(\frac{2n - n_1}{2n - 2n_1 - n_2} \right)$

b) $\hat{\lambda} = \ln \left(\frac{2 - \frac{n_1}{n}}{2 - 2\frac{n_1}{n} - \frac{n_2}{n}} \right)$

$n_1 \sim \text{Bin}(n, p_1)$

$n_2 \sim \text{Bin}(n, p_2)$

$E\left(\frac{n_1}{n}\right) = p_1$, $D\left(\frac{n_1}{n}\right) = \frac{p_1(1-p_1)}{n}$

$P\left(\left|\frac{n_1}{n} - p_1\right| \geq \varepsilon\right) \leq \frac{p_1(1-p_1)}{n\varepsilon^2} \xrightarrow{n \rightarrow \infty} 0$

где $\frac{n_2}{n}$ аналогично

$\frac{2 - \frac{n_1}{n}}{2 - 2\frac{n_1}{n} - \frac{n_2}{n}} \xrightarrow{P} \frac{2 - p_1}{2 - 2p_1 - p_2} = \frac{2 - (1 - e^{-1})}{2 - 2(1 - e^{-1}) - (e^{-1} - e^{-2})} =$

$= \frac{1 + e^{-1}}{e^{-1} + e^{-2}} = e^{\lambda}$

$\Rightarrow \hat{\lambda} = \ln \left(\frac{2 - \frac{n_1}{n}}{2 - 2\frac{n_1}{n} - \frac{n_2}{n}} \right) \xrightarrow{P} \ln(e^{\lambda}) = \lambda$

сходим составление

$$(14) \quad \xi \in U[0; \theta] \quad X_1 = e^{\xi}$$

$$\begin{aligned} a) \quad F_X(x) &= P(X_1 \leq x) = P(e^{\xi} \leq x) = P(\xi \leq \ln x) = \\ &= F_{\xi}(\ln x) = \begin{cases} F_{\xi}(t) = \frac{t}{\theta} & t \in [0; \theta] \\ 0 & \text{else} \end{cases} = \frac{\ln x}{\theta} \end{aligned}$$

$$F_X(x) = \begin{cases} 0 & x < 1 \\ \frac{\ln x}{\theta} & 1 \leq x \leq e^{\theta} \\ 1 & x > e^{\theta} \end{cases} \quad f(x) = F'_X(x) = \begin{cases} \frac{1}{\theta x} & 1 \leq x \leq e^{\theta} \\ 0 & \text{else} \end{cases}$$

$$\begin{aligned} L(\theta, x) &= \prod_{i=1}^n f_{X_i}(X_i) = \prod_{i=1}^n \frac{1}{\theta X_i} \cdot \mathbb{1}(1 \leq X_i \leq e^{\theta}) = \\ &= \frac{1}{\theta^n \cdot \prod X_i} \cdot \mathbb{1}(\text{все } 1 \leq X_i \leq e^{\theta}) \end{aligned}$$

$$\mathbb{1}(\text{все } 1 \leq X_i \leq e^{\theta}) \Leftrightarrow \theta \geq \ln(X_{(n)})$$

$$L(\theta, x) \rightarrow \text{max при } \theta \rightarrow \text{min, но } \theta \geq \ln(X_{(n)})$$

тогда $\hat{\theta} = \ln(X_{(n)}) = \ln(e^{\xi_{(n)}}) = \xi_{(n)}$

$$b) \quad F_{\hat{\theta}_n}(t) = P(\hat{\theta} < t) = P(\xi_{(n)} \leq t) = (P(\xi \leq t))^n = \left(\frac{t}{\theta}\right)^n$$

$$\begin{aligned} \forall \varepsilon > 0 \quad &\hookrightarrow P(|\hat{\theta}_n - \theta| > \varepsilon) = P(\hat{\theta}_n < \theta - \varepsilon) + P(\hat{\theta}_n > \theta) = \\ &= P(\hat{\theta} < \theta - \varepsilon) = P F_{\hat{\theta}}(\theta - \varepsilon) = \left(\frac{\theta - \varepsilon}{\theta}\right)^n = \left(1 - \frac{\varepsilon}{\theta}\right)^n \rightarrow 0 \end{aligned}$$

$$\lim_{n \rightarrow \infty} \left(1 - \frac{\varepsilon}{\theta}\right)^n = 0 \quad \Rightarrow \quad P(|\hat{\theta} - \theta| > \varepsilon) \xrightarrow{n \rightarrow \infty} 0$$

оценка состоятельна