

Teorie odhadů

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Je odhad $\hat{\sigma}^2 = \frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2$ nestranný pokud $X_i \sim N(\mu, \sigma^2)$?

a) μ je známe

$$E[\hat{\sigma}^2] = E\left[\frac{1}{n} \sum_{i=1}^n (x_i - \mu)^2\right] = \frac{1}{n} \sum_{i=1}^n E[(x_i - \mu)^2] = \frac{1}{n} \sum_{i=1}^n \sigma^2 = \sigma^2 \Rightarrow \text{nestranný}$$

b) μ neznáme $\Rightarrow \hat{\mu} = \frac{1}{n} \sum_{i=1}^n x_i$

$$\begin{aligned} E[\hat{\sigma}^2] &= E\left[\frac{1}{n} \sum_{i=1}^n \left(x_i - \frac{1}{n} \sum_{j=1}^n x_j\right)^2\right] = \frac{1}{n} E\left[\sum_{i=1}^n \left(x_i^2 - 2x_i \frac{1}{n} \sum_{j=1}^n x_j + \frac{1}{n^2} \sum_{j=1}^n x_j \sum_{k=1}^n x_k\right)\right] = \\ &= \frac{1}{n} \sum_{i=1}^n E[x_i^2] - \frac{2}{n^2} \sum_{i=1}^n \sum_{j=1}^n E[x_i x_j] + \frac{n}{n^2} \sum_{i=1}^n \sum_{k=1}^n E[x_i x_k] = \frac{1}{n} \sum_{i=1}^n E[x_i^2] - \frac{1}{n^2} \sum_{i=1}^n \sum_{j=1}^n E[x_i x_j] = \\ &= \frac{1}{n} \sum_{i=1}^n (\sigma^2 + \mu^2) - \frac{1}{n^2} \left(\underbrace{(n^2 - n)}_{i \neq j} \cdot \mu^2 + \underbrace{n}_{i=j} (\sigma^2 + \mu^2) \right) = \sigma^2 + \mu^2 - \frac{1}{n^2} (n^2 \mu^2 - n \mu^2 + n \sigma^2 + n \mu^2) = \\ &= \sigma^2 - \frac{1}{n} \sigma^2 = \frac{n-1}{n} \sigma^2 \Rightarrow \text{není nestranný} \end{aligned}$$

2 definice rozptylu $\sigma^2 = E[(x_i - \mu)^2] = E[x_i^2] - 2\mu \overbrace{E[x_i]}^{\mu} + \mu^2 = E[x_i^2] - \mu^2$

$$E[x_i x_j] = \begin{cases} i=j & E[x_i^2] = \sigma^2 + \mu^2 \\ i \neq j & E[x_i] \cdot E[x_j] = \mu^2 \end{cases}$$

$$\ast E[x_i^2] = E[\underbrace{x_i - \mu}_a + \underbrace{\mu}_b]^2 = E[(x_i - \mu)^2] + 2E[(x_i - \mu) \cdot \mu] + E[\mu^2] =$$

$$= \sigma^2 + 2\mu \underbrace{E[x_i - \mu]}_{\substack{\downarrow E[x_i] = \mu \\ \Rightarrow \\ 2\mu E[\mu - \mu] = 0}} + \mu^2 = \underline{\underline{\sigma^2 + \mu^2}}$$