

$X_1, \dots, X_n \sim \text{iid } N(\mu, \sigma^2)$  • identically distributed  
 • independent  
 Метод матем. мат.:  
 $E[X] = \mu = \bar{X} \Rightarrow \hat{\mu} = \bar{X}$   
 $E[X^2] = \sigma^2 + \mu^2 = \bar{X}^2 \Rightarrow \hat{\sigma}^2 = \bar{X}^2 - (\bar{X})^2$

1) Нечувствительность

$\mu$   $E[\hat{\mu}] = E[\bar{X}] = E\left[\frac{1}{n}(X_1 + \dots + X_n)\right] =$   
 $= \frac{1}{n} \cdot (\underbrace{E[X_1]}_{\mu} + \dots + \underbrace{E[X_n]}_{\mu}) = \frac{1}{n} \cdot n\mu = \mu$

$\sigma^2$   $\Rightarrow \hat{\mu}$  — нечувствительна  
 оумена  $\mu$

$E[\hat{\sigma}^2] = E[\bar{X}^2 - (\bar{X})^2] = E[\bar{X}^2] - E[(\bar{X})^2]$

$E\left[\frac{1}{n} \sum_{i=1}^n X_i^2\right] = \frac{1}{n} \sum_{i=1}^n E[X_i^2] = \frac{1}{n} (\sigma^2 + \mu^2) n = \sigma^2 + \mu^2$

$E[\bar{X}^2] = E\left[\left(\frac{1}{n} \sum_{i=1}^n X_i\right)^2\right] = \frac{1}{n^2} E[(X_1 + X_2 + \dots + X_n)^2] =$   
 $= \frac{1}{n^2} E[X_1^2 + X_2^2 + \dots + X_n^2 + 2\underbrace{X_1 X_2}_{=0} + 2\underbrace{X_1 X_3}_{=0} + \dots + \underbrace{X_{n-1} X_n}_{=0}] =$   
 $= \frac{1}{n^2} E\left[\sum_{i=1}^n X_i^2 + (2) \sum_{i < j} X_i X_j\right]$

$E[X_i \cdot X_j] = E[X_i] \cdot E[X_j] = \mu^2$   
 т.к.  $X_i$  и  $X_j$  независимы

$= \frac{1}{n^2} (n(\sigma^2 + \mu^2) + 2 \cdot \frac{n(n-1)}{2} \cdot \mu^2) \quad \text{Ⓢ}$

$C_n^2 = \frac{n!}{2! (n-2)!} = \frac{n(n-1)}{2}$

$$\begin{aligned} \textcircled{=}& \frac{1}{n} (\sigma^2 + \cancel{\mu^2} + \mu^2 \cdot n - \cancel{\mu^2}) = \frac{1}{n} (\sigma^2 + \mu n) = \\ &= \frac{\sigma^2}{n} + \mu^2 \end{aligned}$$

$$\begin{aligned} E[\hat{\sigma}^2] &= E[\bar{X}^2] - E[(\bar{X})^2] = \\ &= \sigma^2 + \cancel{\mu^2} - \left( \frac{\sigma^2}{n} + \cancel{\mu^2} \right) = \sigma^2 - \frac{\sigma^2}{n} = \sigma^2 \left( 1 - \frac{1}{n} \right) = \\ &= \sigma^2 \cdot \left( \frac{n-1}{n} \right) \neq \sigma^2 \end{aligned}$$

$\Rightarrow \hat{\sigma}_{MM}^2$  - смещенная оценка дисперсии

$$\hat{\sigma}^2 = \bar{X}^2 - (\bar{X})^2 = \frac{1}{n} \sum_{i=1}^n (X_i - \bar{X})^2 \quad \downarrow$$

$$\left( \frac{1}{\cancel{n}} \sum_{i=1}^n (X_i - \bar{X})^2 \right) \cdot \frac{\cancel{n}}{n-1} =$$

$$= \frac{1}{\cancel{n} \cdot 1} \sum_{i=1}^n (X_i - \bar{X})^2 = \hat{S}^2$$

несмещенная оценка дисперсии

df = delta degrees of freedom = 1

$$E[\hat{\sigma}^2] = \sigma^2 \cdot \frac{n-1}{n} \xrightarrow{n \rightarrow \infty} \sigma^2$$

$\hat{\sigma}^2$  - асимптотическая несмещенная

$$E[\hat{\sigma}^2 + 5] \underset{n \rightarrow \infty}{=} \sigma^2 + \textcircled{5}$$

↑ принцип сходимости всегда

2) состоятельность

если  $P(|\hat{\theta} - \theta| > \varepsilon) \xrightarrow[n \rightarrow \infty]{} 0$ , то говорят

$\hat{\theta} \xrightarrow[n \rightarrow \infty]{\mathbb{P}} \theta$  ( $\hat{\theta}$  сходится по вероятности к  $\theta$ )

$$P(|\hat{\theta} - \theta| > \varepsilon) \leq \frac{\text{var}(\hat{\theta})}{\varepsilon^2}$$

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

$$\begin{aligned} \text{var}(\hat{\mu}) &= \text{var}(\bar{X}) = \text{var}\left(\frac{1}{n} \sum_{i=1}^n X_i\right) = \frac{1}{n^2} \text{var}\left(\sum_{i=1}^n X_i\right) = \\ &= \frac{1}{n^2} \sum_{i=1}^n \text{var}(X_i) = \frac{1}{n^2} \cdot n\sigma^2 = \frac{\sigma^2}{n} \end{aligned}$$

$$P(|\bar{X} - \mu| > \varepsilon) \leq \frac{\sigma^2}{n \cdot \varepsilon^2} \xrightarrow[n \rightarrow \infty]{} 0$$

$\Rightarrow \hat{\mu} = \bar{X}$  — состоятельная оценка для  $\mu$

3)  $\exists$  эффективное

$$J(\theta) = \mathbb{E} \left[ \left( \frac{\partial \ln f(X, \theta)}{\partial \theta} \right)^2 \right] = \mathbb{E} \left[ \frac{\partial^2 \ln f(X, \theta)}{\partial \theta^2} \right]$$

$$f(x, \mu) = \frac{1}{\sqrt{2\pi\sigma^2}} \cdot \exp\left(-\frac{1}{2} \cdot \left(\frac{x-\mu}{\sigma}\right)^2\right)$$

$$\ln f(x, \mu) = -\frac{1}{2} \ln(2\pi) - \frac{1}{2} \ln(\sigma^2) - \frac{1}{2\sigma^2} (x-\mu)^2$$

$$\frac{\partial \ln f(x, \mu)}{\partial \mu} = + \frac{1}{\sigma^2} (x-\mu)^1 \cdot 2 = \frac{x-\mu}{\sigma^2}$$

$$\frac{\partial^2 \ln f(x, \mu)}{\partial \mu^2} = \left( \frac{x-\mu}{\sigma^2} \right)'_{\mu} = \frac{0-1}{\sigma^2} = -\frac{1}{\sigma^2}$$

$$J(\mu) = \mathbb{E} \left[ \frac{\partial^2 \ln f(x, \mu)}{\partial \mu^2} \right] = \mathbb{E} \left[ -\frac{1}{\sigma^2} \right] = \frac{1}{\sigma^2}$$

теор-ба Рао-Крамера-Фреше:

$$\text{var}(\hat{\mu}) \geq \frac{1}{n \cdot J(\mu)} = \frac{\sigma^2}{n}$$

$\Rightarrow \hat{\mu}$  - эффективная

в классе всех несмещенных

(у  $\hat{\mu}$  самая малая дисперсия)