

$$\textcircled{2} \quad y_i = \beta + u_i$$

$$\text{MKK: } \hat{y} = \hat{\beta}$$

$$RSS = \sum_{i=1}^n (y_i - \hat{y}_i)^2 = \sum_{i=1}^n (y_i - \hat{\beta})^2 \rightarrow \min_{\hat{\beta}}$$

$$f(x) \sum_{i=1}^n (y_i - \hat{\beta}) = 0$$

$$\hat{\beta} = \frac{\sum_{i=1}^n y_i}{n} = \bar{y}$$

$$RSS = \sum_{i=1}^n (y_i - \bar{y})^2 \quad TSS = ESS + RSS$$

|| ||
n 0

$$TSS = \sum_{i=1}^n (y_i - \bar{y})^2$$

$$ESS = \sum_{i=1}^n (\hat{y}_i - \bar{y})^2 = \sum_{i=1}^n (\bar{y} - \bar{y})^2 = 0$$

$$R^2 = \frac{ESS}{TSS} = 0$$

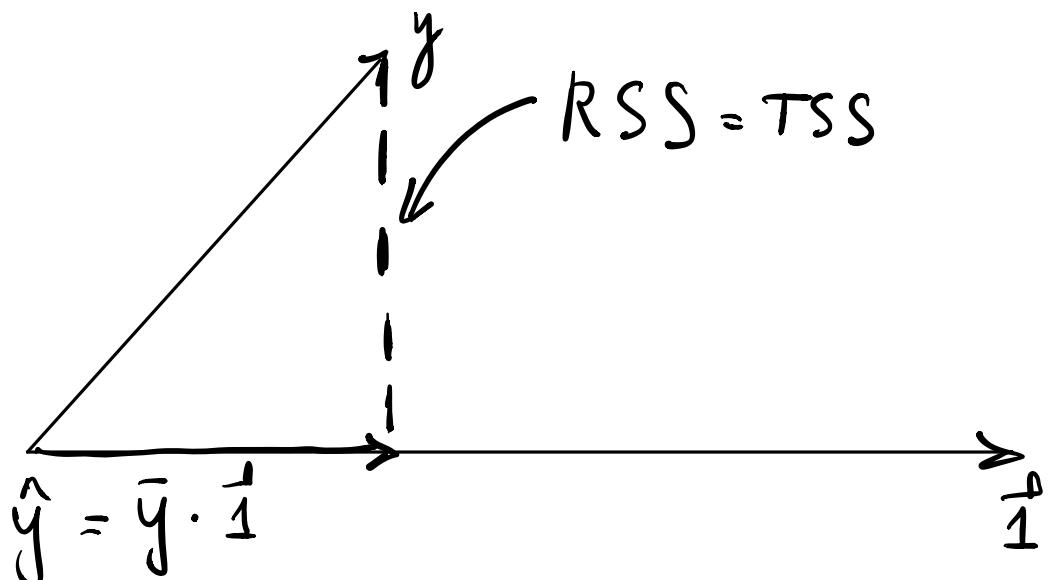
$$R^2 = \underbrace{\text{scorr}^2(y, \hat{y})}_{=} = 0 = \text{scov}^2(y - \hat{y})$$

$$\text{scov}(y, x) = \sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})$$

$$\text{scorr}(y, x) = \frac{\sum_{i=1}^n (y_i - \bar{y})(x_i - \bar{x})}{\sqrt{\sum_{i=1}^n (y_i - \bar{y})^2} \sqrt{\sum_{i=1}^n (x_i - \bar{x})^2}}$$

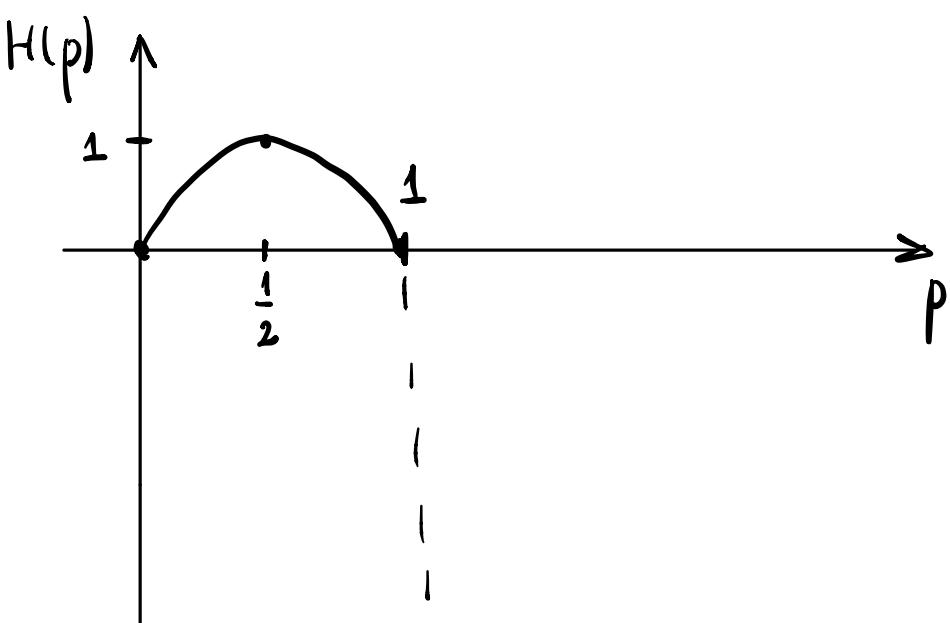
$$y = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

$$x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$



④ $X = \begin{cases} 1, & p \\ 0, & 1-p \end{cases}$

$$H(p) = p \log_2 p + (1-p) \log_2 (1-p)$$



⑤

$$Z = \{1, 2, 3\}$$

γ_1 - magn. $\in [1 \text{ km.}]$

γ_2 - 2 km.

$1 - \gamma_1 - \gamma_2$ - 3 km.

E₁:

$$p(z_i=1 | x_i, \Theta_{old}) = \overbrace{\gamma_1 p(x_i | z_i=1, \Theta_{old})}^{g_1}$$

$$\begin{aligned} & \gamma_1 p(x_i | z_i=1, \Theta_{old}) + \underbrace{\gamma_2 p(x_i | z_i=2, \Theta_{old})}_{+} + \\ & + (1 - \gamma_1 - \gamma_2) p(x_i | z_i=3, \Theta_{old}) \end{aligned}$$

$$g_2 = p(z_i=2 | x_i, \Theta_{old}) = \frac{\gamma_2 p(x_i | z_i=2, \Theta_{old})}{\gamma_1 + \gamma_2}$$

$E_2, M:$

$$\begin{aligned} Q &= \sum_{i=1}^n g_1 (\ln p(x_i | \theta) + \ln \gamma_1) + \\ &\quad + g_2 (\ln p(x_i | \theta) + \ln \gamma_2) + \\ &\quad + (1 - g_1 - g_2) (\ln p(x_i | \theta) + \ln (1 - \gamma_1 - \gamma_2)) \\ &\rightarrow \max_{\theta, \gamma_1, \gamma_2} \end{aligned}$$

⑥

$$X \sim \chi^2_5$$

$$X \perp\!\!\!\perp Y$$

$$Y \sim \chi^2_{10}$$

$$F_{k,m} = \frac{\chi^2_k / k}{\chi^2_m / m}$$

$$Z = \frac{X+Y}{X}$$

$$\underline{P\{Z > z^*\} = 0.05}$$

$$P\left\{ 1 + \frac{Y}{X} > z^* \right\} = 1 - P\left\{ 1 + \frac{Y}{X} \leq z^* \right\} =$$

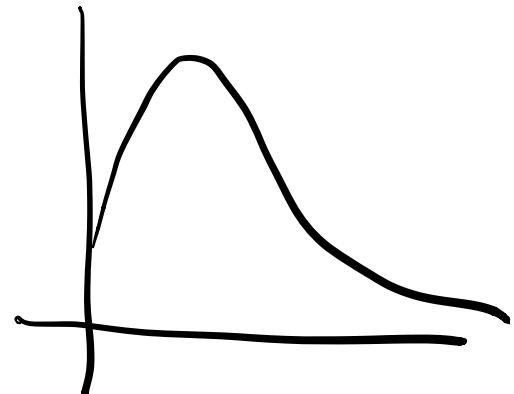
$$= 1 - P \left\{ \frac{Y}{X} \leq z^* - 1 \right\} =$$

$$= 1 - P \left\{ \underbrace{\frac{Y/10}{X/5}}_{Q} \leq (z^* - 1) \frac{5}{10} \right\}$$

$$Q \sim F_{10,5}$$

$$P \{ Z > z^* \} = 0.05$$

$$1 - P \left\{ Q \leq (z^* - 1) \frac{5}{10} \right\} = 0.05$$



$$P \left\{ Q \leq (z^* - 1) \frac{5}{10} \right\} = 0.95$$

$$(z^* - 1) \frac{5}{10} = 4.74$$

$$z^* = 10.48$$

(7) \mathbb{R}^3 $W = \left\{ (x_1, x_2, x_3) \mid \underbrace{x_1 + 2x_2 + x_3 = 0}_{\uparrow} \right\}$

$$V = \text{lin} \left[(1, 2, 3)^T \right]$$

$$\dim V = 1 - \|\hat{u}\|^2 \sim \chi_1^2 \quad \mathbb{R}^n$$

$$\dim W = 2 - \|\hat{u}\|^2 \sim \chi_2^2 \quad V, \dim V = k$$

$$\dim(V \cap W) = 0 \quad \dim V^\perp = n-k$$

$$\dim V^\perp = 3-1 = 2 - \|\hat{u}\|^2 \sim \chi_2^2$$

$$\dim W^\perp = 3-2 = 1$$

T. \hat{u} имеет норм. стандартн. распр.
 неприм. в ма V_k — \hat{u} -прост.

$$\|\hat{u}\|^2 \sim \chi_k^2$$

$$2) \text{ М-я проекция: } \underbrace{X(X^T X)^{-1} X^T}_{\text{М-я проекция}} \cdot u = \hat{u}$$

М-я проекция
 вектора u на
 подпр-о, орт. ст-иции
 X

$$u \rightarrow V, \dim V = 1$$

$$\text{lin} \begin{pmatrix} " \\ 1 \\ 2 \\ 3 \end{pmatrix}$$

$$X = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}$$

$$\hat{u} = \underbrace{X(X^T X)^{-1} X^T u}_{(X^T X)^{-1} X^T y}$$

$$\begin{aligned} y &= X\beta + u \\ \hat{y} &= X\hat{\beta} \\ &\quad (X^T X)^{-1} X^T y \end{aligned}$$

$$\hat{u} = \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}_{3 \times 1} \left[(1 \ 2 \ 3) \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix}_{1 \times 3} \right]^{-1} (1 \ 2 \ 3)_{1 \times 3} \begin{pmatrix} u_1 \\ u_2 \\ u_3 \end{pmatrix}_{3 \times 1}$$

$$W: \underline{x_1 + 2x_2 + x_3 = 0}$$

$$z = -x_1 - 2x_2$$

$$\begin{pmatrix} 1 \\ 0 \\ -1 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \\ -2 \end{pmatrix} - \text{сним из базиса } W$$

$$X = \begin{pmatrix} 1 & 0 \\ 0 & 1 \\ -1 & -2 \end{pmatrix}$$

③ $u \sim N(0, I)$
 $\underbrace{Au}_{\sim} \sim N(0, I)$

$$\text{Var}(Au) = I$$

$$A \underbrace{\text{Var}(u)}_{I} A^T = I$$

$$AA^T = I \Rightarrow A^T = A^{-1} \Rightarrow A - \text{ортогональная.}$$

$$2) \det(A) = \pm 1$$

u - ср. вектор
A - матрица const
$E(A \cdot u) = A E(u)$
$\text{Var}(Au) = A \cdot \text{Var}(u) \cdot A^T$

$$3) \langle c_1, c_2 \rangle = c_1^\top \cdot c_2 = 0$$

$$4) \text{Расп} \|u\|^2 \sim \chi_n^2$$

↑
размерност
простр-а, в к-ии лежит u .

① І-ви груп:

$$\begin{matrix} \delta_1 & \dots & \delta_{10} \\ \downarrow \\ \left\{ \begin{array}{l} \text{форм}, p \\ \text{крайн}, 1-p \end{array} \right. \end{matrix}$$

$$1) X_i \sim \text{Bin}(10, p)$$

2) $x_1 \dots x_n$ - тозиорка

$$\begin{aligned} L &= P\{X_1=x_1, \dots, X_n=x_n\} = \underbrace{P\{X_1=x_1\}}_{\substack{\text{C}_{10}^{x_1} p^{x_1} (1-p)^{10-x_1} \\ P\{X_1=x_1\}}} \cdot \dots \cdot \underbrace{P\{X_n=x_n\}}_{\substack{\text{C}_{10}^{x_n} p^{x_n} (1-p)^{10-x_n} \\ \dots}} \end{aligned}$$

$$\cdot \dots \cdot \binom{x_n}{10} p^{x_n} (1-p)^{10-x_n} = \\ = \cancel{\binom{x_1}{10} \cancel{\binom{x_2}{10}} \dots \cancel{\binom{x_n}{10}}} p^{\sum_{i=1}^n x_i} (1-p)^{\sum_{i=1}^n (10-x_i)}$$

$$\ell = \sum_{i=1}^n x_i \ln p + \sum_{i=1}^n (10-x_i) \ln (1-p)$$

$$\ell'_p = \frac{\sum x_i}{p} - \frac{\sum (10-x_i)}{1-p}$$

$$\frac{\sum x_i}{\hat{p}} - \frac{10n - \sum x_i}{1-\hat{p}} = 0$$

$$3) \quad \hat{p}_{ml} = \frac{\sum_{i=1}^n x_i}{10n}$$

$$4) \quad \hat{p} \stackrel{ac.}{\sim} N(p, I^{-1}(p))$$

$$\ell''_p = - \frac{\sum x_i}{p^2} - \frac{\sum (10-x_i)}{(1-p)^2}$$

$$I = \mathbb{E}(-H) = \mathbb{E}\left(\left(\frac{\sum x_i}{p^2}\right) + \frac{\sum (10 - x_i)}{(1-p)^2}\right) =$$

$$= \frac{np}{p^2} + \frac{10n - np}{(1-p)^2} = \frac{n}{p} + \frac{10n - np}{(1-p)^2}$$

$$\hat{I} = \frac{n}{\hat{p}} + \frac{10n - n\hat{p}}{(1-\hat{p})^2}$$

$$\text{Var}(\hat{p}) = \hat{I}^{-1} = \frac{1}{\hat{I}}$$

$$p \in [\hat{p} - 1.96 \sqrt{\text{Var}(\hat{p})}; \hat{p} + 1.96 \sqrt{\text{Var}(\hat{p})}]$$

5) $\widehat{\mathbb{E}(X_i)}$ $\text{Var}(X_i) = 10p(1-p)$

$$\mathbb{E}(X_i) = 10p \quad \text{Var}(\hat{X}_i) = 10\hat{p}(1-\hat{p})$$

$$\widehat{\mathbb{E}(X_i)} = 10\hat{p}$$

6) $\text{Var}(10\hat{p}) = 100 \text{Var}(\hat{p})$

$$\text{Var} (10\hat{p}(1-\hat{p})) = 100 \text{Var}(\hat{p} - \hat{p}^2) =$$

$$= 100 \left[\underbrace{\text{Var}(\hat{p})}_{\text{zweiseitig}} + \text{Var}(\hat{p}^2) - 2\text{cov}(\hat{p}, \hat{p}^2) \right]$$

$$\hat{p} \sim N(p, \underbrace{I^{-1}(p)}_{\sigma^2})$$

$$\begin{aligned} \text{Var}(\hat{p}^2) &= \underbrace{\mathbb{E}(\hat{p}^4)}_{p^4 + 6p^2\sigma^2 + 3\sigma^4} - \left[\mathbb{E}(\hat{p}^2) \right]^2 = \\ &= p^4 + 6p^2\sigma^2 + 3\sigma^4 - (p^2 + \sigma^2)^2 \end{aligned}$$

$$\begin{aligned} \text{cov}(\hat{p}, \hat{p}^2) &= \mathbb{E}(\hat{p}^3) - \mathbb{E}(\hat{p}) \mathbb{E}(\hat{p}^2) = \\ &= p^3 + 3p\sigma^2 - p \cdot (p^2 + \sigma^2) \\ &= p^3 + 3p\sigma^2 - p^3 - p\sigma^2 = 2p\sigma^2. \end{aligned}$$