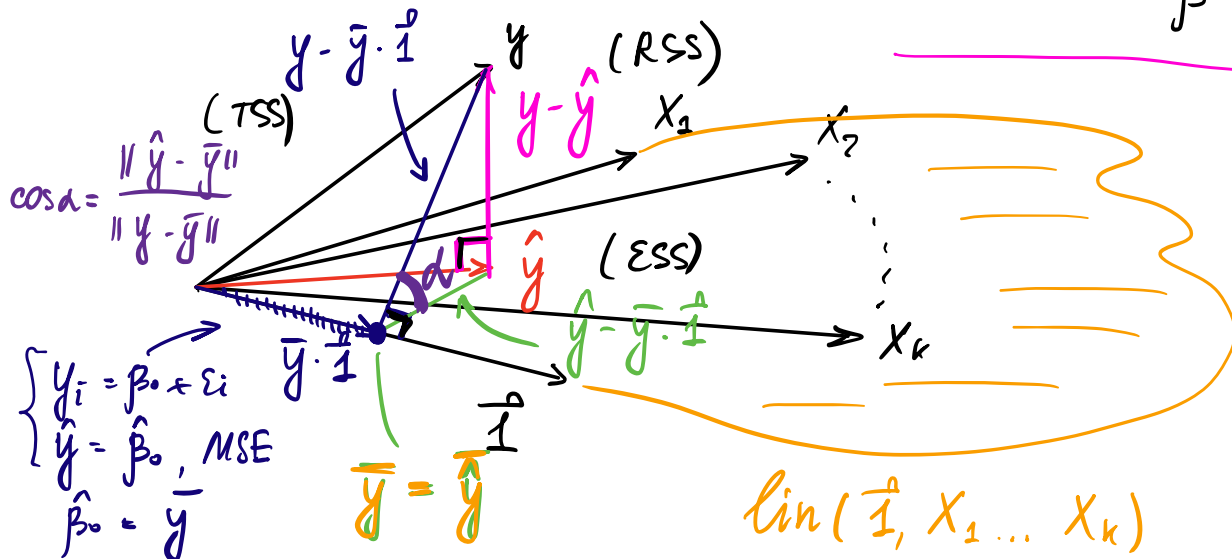


$$\underset{N \times 1}{y} = \underset{N \times (K+1)}{X} \underset{(K+1) \times 1}{\beta} + \underset{N \times 1}{\varepsilon}$$

$$X = \begin{bmatrix} | & | & | & \dots & | \\ 1 & x_1 & x_2 & \dots & x_k \\ | & | & | & \dots & | \end{bmatrix}$$

$$\hat{y} = X\hat{\beta} \quad \text{non nom. MNRK: } \text{MSE} = \|y - \hat{y}\|_2^2 \rightarrow \min_{\hat{\beta}}$$



$$\hat{y} = X\hat{\beta} = \hat{\beta}_0 \begin{bmatrix} 1 \\ | \end{bmatrix} + \hat{\beta}_1 \begin{bmatrix} | \\ x_1 \\ | \end{bmatrix} + \hat{\beta}_2 \begin{bmatrix} | \\ x_2 \\ | \end{bmatrix} + \dots + \hat{\beta}_k \begin{bmatrix} | \\ x_k \\ | \end{bmatrix}$$

$$\Rightarrow \hat{y} \in \text{lin}(\mathbf{1}, x_1, \dots, x_k)$$

$$\text{RSS} = \sum_i (y_i - \hat{y}_i)^2 = \|y - \hat{y}\|_2^2$$

↑ residual sum of squares

$$\text{TSS} = \sum_i (y_i - \bar{y})^2 = \|y - \bar{y} \cdot \mathbf{1}\|_2^2$$

\ total sum of squares

$$ESS = \sum_i (\hat{y}_i - \bar{\hat{y}})^2 = \|\hat{y} - \bar{y} \cdot \mathbf{1}\|_2^2$$

↑  
estimated sum of squares

$$TSS = ESS + RSS$$

$$R^2 = \frac{\sum_i (\hat{y}_i - \bar{\hat{y}})^2}{\sum_i (y_i - \bar{y})^2} = \frac{ESS}{TSS} = 1 - \frac{RSS}{TSS} =$$

$$= \cos^2 \alpha \quad \left( \begin{matrix} TSS, ESS \end{matrix} \right)$$

$$y_i = \beta_0 + \beta_1 x_{1i} + \dots + \beta_k x_{ki} + \varepsilon_i \quad UR \text{ (unrestricted model)}$$

+ oparur.  $\rightarrow R$  (restricted model)

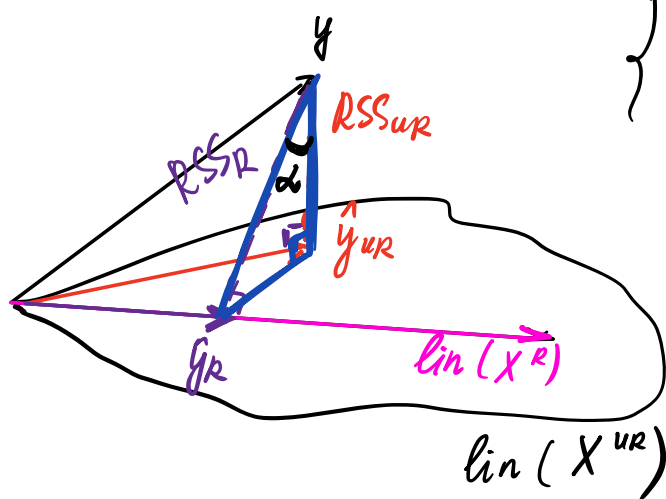
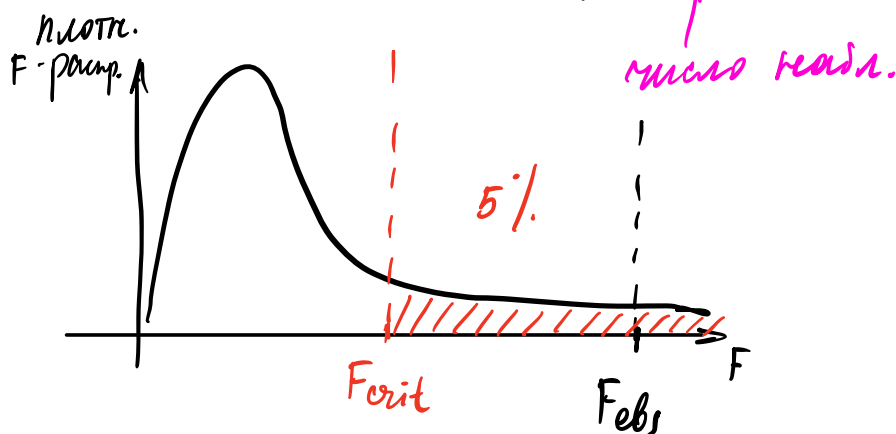
$$R: y_i = \beta_0 + \beta_1 x_{1i} + \varepsilon_i \quad \text{Oparur: } \beta_2 = \dots = \beta_k = 0$$

$$\left\{ \begin{array}{l} H_0: \beta_2 = \dots = \beta_k = 0 \\ H_1: \beta_2^2 + \dots + \beta_k^2 > 0 \end{array} \right\} \Rightarrow \begin{array}{l} \text{Тест.} \text{ } \text{un. } 0 \text{ } \text{на } \text{exp.} \\ UR \text{ } \overline{\text{vs}} \text{ } R \end{array}$$

- 2)  $\begin{cases} H_0: \text{бегрн у UR, у R нугенб} \\ H_1: \text{бегрн UR, а R нугенб кеберн} \end{cases}$

$$F = \frac{(RSS_R - RSS_{UR}) / (k_{UR} - k_R)}{RSS_{UR} / (N - k_{UR})} \stackrel{H_0}{\sim} F_{(k_{UR} - k_R; N - k_{UR})}$$

↑  
micro perp. б UR      micro perp. б R



- $\begin{cases} H_0: \text{UR бегрн, R бегр.} \\ H_1: \text{UR бегрн, R кеберн.} \end{cases}$

$$\tan^2 \alpha = \frac{\| \hat{y}_{UR} - \hat{y}_R \|^2}{\| y - \hat{y}_{UR} \|^2} = \frac{RSS_R - RSS^{UR}}{RSS_{UR}}$$

Пример:

$$UR: y_i = \beta_0 + \beta_1 x_{1i} + \beta_2 x_{2i} + \varepsilon_i$$

$$\varepsilon_i \sim N(0, \sigma^2 I)$$

$$X = \begin{pmatrix} 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 1 & 1 \end{pmatrix} \quad y = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{pmatrix}$$

$$\hat{\beta} = (X^T X)^{-1} X^T y = \begin{pmatrix} 2 \\ 2 \\ 1 \end{pmatrix} \quad \hat{y} = X \hat{\beta} = \begin{pmatrix} 2 \\ 2 \\ 2 \\ 4 \\ 5 \end{pmatrix}$$

$$RSS_{UR} = \sum_i (y_i - \hat{y}_i)^2 = 1 + 0 + 1 + 0 + 0 = 2$$

Тест. пер. на знар. в целом

$$\begin{cases} H_0: \beta_1 = \beta_2 = 0 \\ H_1: \beta_1^2 + \beta_2^2 > 0 \end{cases}$$

$$R: y_i = \beta_0 + \varepsilon_i$$

$$\hat{\beta}_0 = \bar{y} = \frac{15}{5} = 3$$

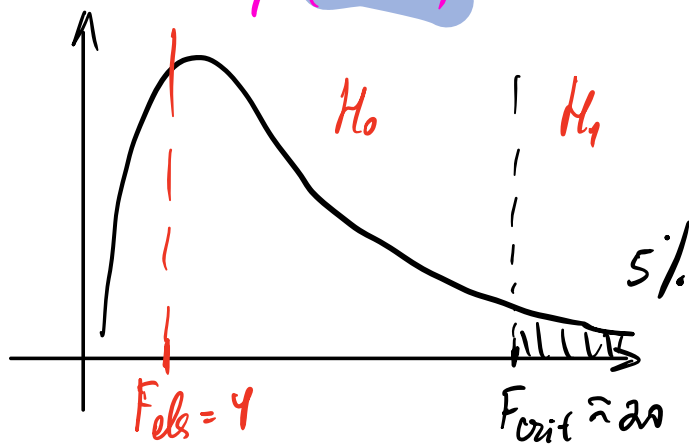
$$\hat{y}_R = \begin{pmatrix} 3 \\ 3 \\ 3 \\ 3 \\ 3 \end{pmatrix}$$

$$y = \begin{pmatrix} 1 \\ 2 \\ 3 \\ 4 \\ 5 \end{pmatrix}$$

$$RSS_R = \sum_i (y_i - \hat{y}_R)^2 = 4 + 1 + 1 + 4 + 4 = 10$$

$$F = \frac{(RSS_R - RSS_{UR}) / (k_{UR} - k_R)}{RSS_{UR} / (N - k_{UR})} =$$

$$= \frac{(10 - 2) / (3 - 1)}{2 / (5 - 3)} = \frac{8 / 2}{2 / 2} = 4 \stackrel{H_0}{\sim} F_{(2,4)}$$



→ 5%  $H_0$  не отвергн.