

1) Equation (theoretical) $y_i = \alpha + \beta x_i + \varepsilon_i$

2) Assumptions : - deterministic / stochastic

$$E(\varepsilon_i) \quad E(\varepsilon_i | X)$$

- classic linear reg. assumptions

$$\left[\begin{array}{ccc} E(\varepsilon_i) = 0 & E(\varepsilon_i^2) = \sigma^2 & E(\varepsilon_i \varepsilon_j) = 0 \\ & \text{"} & \text{"} \end{array} \right]$$

$$\text{Var}(\varepsilon_i^2) = E(\varepsilon_i^2) - E^2(\varepsilon_i) \quad \text{Cov}(\varepsilon_i, \varepsilon_j)$$

$$\text{"} \quad \text{"} \\ E(\varepsilon_i \varepsilon_j) - E(\varepsilon_i) \cdot E(\varepsilon_j)$$

3) Method : OLS (WLS, IV, MLE)

4) Properties

$$\beta \sum (x_i - \bar{x})^2 + \sum (x_i - \bar{x}) \varepsilon_i - \bar{\varepsilon} \sum (x_i - \bar{x})$$

"0"

$$\sum (x_i - \bar{x}) (\beta x_i + \varepsilon_i - \beta \bar{x} - \bar{\varepsilon})$$

$$\hat{\beta} = \frac{\text{Cov}(x, y)}{\text{Var}(x)} = \frac{\sum (x_i - \bar{x})(y_i - \bar{y})}{\sum (x_i - \bar{x})^2} = \beta + \frac{\sum (x_i - \bar{x}) \varepsilon_i}{\sum (x_i - \bar{x})^2} =$$

$$a_i = \frac{x_i - \bar{x}}{\sum (x_i - \bar{x})^2} = \beta \sum \left(\frac{x_i - \bar{x}}{\sum (x_i - \bar{x})^2} \right) \varepsilon_i =$$

$$= \beta + \sum a_i \varepsilon_i$$

Problem 1. Properties of a_i

a) $\sum a_i = 0$ d.h.

b) $\sum a_i^2 = \frac{1}{\sum (x_i - \bar{x})^2}$

$$\sum \left(\frac{x_i - \bar{x}}{\sum (x_i - \bar{x})^2} \right)^2 = \frac{\sum (x_i - \bar{x})^2}{\left(\sum (x_i - \bar{x})^2 \right)^2} = \frac{1}{\sum (x_i - \bar{x})^2}$$

c) $\sum a_i \cdot x_i = 1$

$$\sum \frac{(x_i - \bar{x}) x_i}{\sum (x_i - \bar{x})^2} = \frac{1}{\sum (x_i - \bar{x})^2} \left(\sum (x_i - \bar{x}) x_i - \sum (x_i - \bar{x}) \bar{x} \right) = 1$$

"0"



Problem 2. $\hat{\beta} = \beta + \sum a_i \varepsilon_i$

$$E(\hat{\beta}) = \beta + E\left(\sum a_i \varepsilon_i\right) =$$

$$\sum a_i E(\varepsilon_i) = 0$$

Problem 3.

$$\hat{\beta} = \frac{\text{Cov}(X, Y)}{\text{Var}(X)} = \frac{\text{Cov}(X, \alpha + \beta X + \varepsilon)}{\text{Var}(X)} =$$

$$\frac{\text{Cov}(X, \alpha) + \text{Cov}(X, \beta X) + \text{Cov}(X, \varepsilon)}{\text{Var}(X)} = \beta + \frac{\text{Cov}(X, \varepsilon)}{\text{Var}(X)}$$

$$E(\hat{\beta}) = \beta + E\left(\frac{\text{Cov}(X, \varepsilon)}{\text{Var}(X)}\right) =$$

$$\beta + \frac{\text{Cov}(X, E\varepsilon)}{\text{Var}(X)} = \beta$$

Problem 3.

$$y_i = \alpha + \beta_1 x_{1i} + \beta_2 x_{2i} + \epsilon_i$$

$$\|b - Ax\|_2^2$$

\Downarrow

$$A^T A x = A^T b \quad - \text{normal equation}$$

$\hat{\epsilon}$ normal to $\text{span}(A)$
orthogonal

$$\|y - X\hat{\beta}\|_2^2$$

$$X^T X \beta = X^T y \quad - \text{normal equation}$$

$$y_i = \alpha + \beta_1 x_{1i} + \beta_2 x_{2i} + \epsilon_i$$

$$y = \hat{\alpha} + \hat{\beta}_1 x_1 + \hat{\beta}_2 x_2$$

$$\text{Cov}(x_1, -) ; \text{Cov}(x_2, -) :$$

$$\begin{cases} \hat{\beta}_1 \text{Var}(x_1) + \hat{\beta}_2 \text{Cov}(x_1, x_2) = \text{Cov}(x_1, y) \\ \hat{\beta}_1 \text{Cov}(x_1, x_2) + \hat{\beta}_2 \text{Var}(x_2) = \text{Cov}(x_2, y) \end{cases}$$

$$\hat{\beta}_1 \text{Var}(x_1) + \hat{\beta}_2 \text{Cov}(x_1, x_2) = \text{Cov}(x_1, y)$$

$$\hat{\beta}_1 \text{Cov}(x_1, x_2) + \hat{\beta}_2 \text{Var}(x_2) = \text{Cov}(x_2, y)$$

$$\begin{bmatrix} \text{Var}(x_1) & \text{Cov}(x_1, x_2) \\ \text{Cov}(x_1, x_2) & \text{Var}(x_2) \end{bmatrix} \begin{bmatrix} \hat{\beta}_1 \\ \hat{\beta}_2 \end{bmatrix} = \begin{bmatrix} \text{Cov}(x_1, y) \\ \text{Cov}(x_2, y) \end{bmatrix}$$

$$\hat{\beta}_1 = \frac{\Delta_1}{\Delta} = \frac{\text{Cov}(x_1, y) \text{Var}(x_2) - \text{Cov}(x_1, x_2) \text{Cov}(x_2, y)}{\text{Var}(x_1) \text{Var}(x_2) - [\text{Cov}(x_1, x_2)]^2}$$

$$\hat{\beta}_2 = \frac{\Delta_2}{\Delta}$$

$$\hat{\beta} = (X'X)^{-1} X'y$$

F-tests for linear restrictions

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

$$H_0: \beta_1 = \beta_2 = \beta_3 = 0 \Rightarrow y_R$$

$$1 - \frac{RSS}{TSS}$$

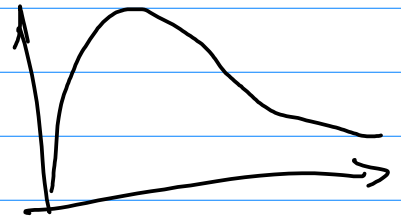
H_0 : one of restrictions isn't valid $\Rightarrow y_{ur}$

$$F = \frac{(RSS_R - RSS_{ur}) / q}{RSS_{ur} / (n - k)}$$

\leftarrow # of restrictions

$\sim F(q, n-k)$

\downarrow
est. coefficient



$$F = \frac{(R^2_{ur} - R^2_R) / q}{(1 - R^2_{ur}) / (n - k)}$$

$$RSS_R \overset{?}{>} RSS_{ur}$$

F-tests for linear restrictions $R^2 = \frac{ESS}{TSS} =$

$$H_0: \beta_1 = \beta_2 = \beta_3 = 0 \Rightarrow y_k \quad 1 - \frac{RSS}{TSS}$$

H_0 : one of restrictions isn't valid $\Rightarrow y_{ur}$

$$F = \frac{(RSS_k - RSS_{ur}) / q}{RSS_{ur} / (n-k)} \quad \begin{matrix} \leftarrow \# \text{ of restrictions} \\ \sim H_0 \end{matrix} F(q, n-k)$$

F-test
Goodness-of-fit

$$H_0: \beta_2 = \dots = \beta_k = 0 \quad y_k = \beta_1 + \epsilon_i$$

$$H_a: \exists i: \beta_i \neq 0$$

$$F = \frac{(TSS - RSS) / k-1}{RSS / n-k} = \frac{ESS / k-1}{RSS / n-k}$$

$$k=1 \Rightarrow F = t^2$$

$$R^2 = 1 - \frac{RSS}{TSS}$$

$$R^2_{adj} = \bar{R}^2 = 1 - \frac{RSS/(n-k)}{TSS/(n-1)}$$