

Assumptions of the LS estimator

Non-linearity:

Slide 23

a) In variables:

$$y_i = \beta_1 + \beta_2 \underline{x_{2i}^2} + u_i$$

↓ linearization

$$x_{2i}^2 = x_{2i}^1$$



b) In parameters:

$$y_i = \beta_1 + \underline{\beta_2^2} x_{2i} + u_i$$

unknown!



Fixed explanatory variables:

$$E(X) = X$$

$$E[(X^T X)^{-1} X^T] = (X^T X)^{-1} X^T$$

etc.

Slide 24

Heteroscedasticity:

Slide 27

$$\text{Var}(u_i) = \sigma_i^2$$

Autocorrelation:

Slide 29

Is about covariance or correlation between stochastic variable u and lagged stochastic variable u , i.e.:

t	u_t ^{i}	u_{t-1} ^{j}
1	u_1	—
2	u_2	u_1
3	u_3	u_2
4	u_4	u_3
5	u_5	u_4

Assumption 8, example:

Slide 33

$$y_i = b_1 + b_2 x_{2i} + e_i, \quad \text{var}(x_2) = 0$$

Then $x_{2i} = \bar{x}_2, \forall i$:

$$b_2 = \frac{\sum_i (x_{2i} - \bar{x}_2)(y_i - \bar{y})}{\sum_i \underbrace{(x_{2i} - \bar{x}_2)^2}_{=0}}$$

not defined

and

$$b_1 = \bar{y} - \underbrace{b_2 \bar{x}_2}_{\text{not defined}}$$

not defined.

Assumption 10:

Slide 37

Consequences of perfect (multi)collinearity:

1) $\det(X^T X) = 0$;

2) matrix $X^T X$ is singular and thus non-invertible;

3) $b = (X^T X)^{-1} X^T y$ is not defined.

Sample properties of the LS estimator

Unbiasedness:

Slide 43

$$b = (X^T X)^{-1} X^T y = (X^T X)^{-1} X^T (X\beta + u)$$

$$b = \underbrace{(X^T X)^{-1} X^T X}_{I} \beta + (X^T X)^{-1} X^T u$$

$$b = \beta + (X^T X)^{-1} X^T u$$

$$E(b) = E(\beta) + E[(X^T X)^{-1} X^T u]$$

$$E(b) = \beta + \underbrace{(X^T X)^{-1} X^T}_{\text{fixed regressors}} \underbrace{E(u)}_{=0} = \underline{\underline{\beta}}$$

Efficiency:

Slide 44

$$b - \beta = (X^T X)^{-1} X^T u$$

$$(b - \beta)^T = u^T X (X^T X)^{-1}$$

symmetry

$$\begin{aligned} \text{Var} - \text{Cov}(b) &= E[(b - \beta)(b - \beta)^T] = \\ &= E[(X^T X)^{-1} X^T u \cdot u^T X (X^T X)^{-1}] = \\ &= \underbrace{(X^T X)^{-1} X^T}_{\text{fixed regressors}} \underbrace{E(uu^T)}_{\sigma^2 \cdot I} X (X^T X)^{-1} = \end{aligned}$$

$$= \sigma^2 \cdot (X^T X)^{-1} \underbrace{X^T X}_{=I} (X^T X)^{-1} =$$

$$= \underline{\underline{\sigma^2 \cdot (X^T X)^{-1}}}$$

Slide 46:

Assume a tri variate regression model:

$$\underline{y_i} = b_1 + \underbrace{b_2}_{\text{circled}} \underline{x_{2i}} + b_3 x_{3i} + e_i \quad (\text{SRM})$$

Slide 47:

BLUE or MVUE

"best"

(most efficient)

"minimum
variance"