

Linear Assumption

①

Ans: The bias of an estimator measures how far its average value is from the true parameter.

Mathematically $\text{Bias}(\hat{\beta}) = E[\hat{\beta}] - \beta$ if it is equal to 0, the estimator is unbiased.

② for OLS unbiased meaning $E[\hat{\beta}] = \beta$

$$\hat{\beta} = (x'x)^{-1} x'y \quad y = x\beta + \varepsilon$$

$$\hat{\beta} = (x'x)^{-1} x'x\beta + (x'x)^{-1} x'\varepsilon \Rightarrow$$

$$\hat{\beta} = \beta + (x'x)^{-1} x'\varepsilon$$

$$E[\hat{\beta}|x] = E[\beta + (x'x)^{-1} x'\varepsilon|x] =$$

$$= \beta + (x'x)^{-1} x' E[\varepsilon|x]$$

OLS assumes the errors have zero ~~mean~~ mean

Given $x \quad E[\varepsilon|x] = 0 \Rightarrow \boxed{E[\hat{\beta}|x] = \beta}$

④ Homoskedasticity

Mathematical form: $\text{Var}(\epsilon_i | x) = \sigma^2 + i$

This means every error term ϵ_i has the same Variance - the spread of the errors is constant across all levels of x .

$$d) E[\epsilon|x] = 0$$

$$\text{Var}(\epsilon|x) = \sigma^2 I_n$$

X has full column (so $X'X$ is invertible)

DLS

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

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

$$\text{Var}(\hat{\beta}|x) = A \text{Var}(\epsilon|x) A'$$

$$\text{Var}(\hat{\beta}|x) = \text{Var}((X'X)^{-1} X' \epsilon | x) = (X'X)^{-1} X' \text{Var}(\epsilon|x) X(X')^{-1}$$

$$\text{Var}(\hat{\beta}|x) = (X'X)^{-1} X' (\sigma^2 I_n) X(X')^{-1} =$$

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

$$e) E(\varepsilon|x) = 0 \quad \text{Var}(\varepsilon|x) = \sigma^2 I_n$$

() ↴

The avg res. is 0

they're centred around
the regression line

↳ the spread of res
is the same for all obs

$$\text{Var}(\varepsilon|x) = \sigma^2 (I - H)$$

$$H = X(X^T X)^{-1} X^T$$

() If $\varepsilon \sim N(0, \sigma^2 I)$ then the res are normally distributed as well: $\hat{\varepsilon} \sim N(0, \sigma^2 (I - H))$



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

$$\hat{\varepsilon} = y - \hat{y} = y - Hy = (I - H)y$$

$$\hat{\varepsilon} = (I - H)(X\beta + \varepsilon)$$

$(I - H)(X\beta) = 0$ (a key prop of the hat matrix)

$$(I - X(X^T X)^{-1} X^T) X =$$

$$= X - X \underbrace{(X^T X)^{-1} X^T}_{I} X = X - X = 0.$$

$$\text{so } \hat{\varepsilon} = (I - H)\varepsilon$$

① D

⑥ C

⑪ D

⑤ A

⑩ b

⑦ B

② D

⑪ c

⑯ d

③ B

⑧ A

⑬ A

⑫ b

④ D

⑩ A

⑭ B

⑮ c

⑤ D