

Omitted Variable Bias

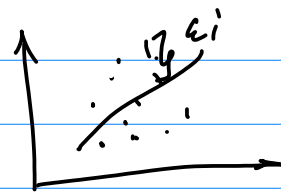
$$\bar{x} \xrightarrow{P} E(X_i)$$

$$\hat{Cov}(X_i, Y) \xrightarrow{P} Cov(X_i, Y)$$

Deterministic Regressors:

True: $y_i = \beta_1 + \beta_2 \cdot x_i + \beta_3 w_i + \varepsilon_i$

Est: $y_i = \hat{\beta}_1 + \hat{\beta}_2 x_i + u_i$



$$E(\varepsilon_i) = 0, E(\varepsilon_i^2) = \sigma^2,$$

$$E(\varepsilon_i, \varepsilon_j) = 0$$

$$u_i = \beta_3 w_i + \varepsilon_i$$

$$E(u_i) = \beta_3 w_i$$

$\Rightarrow \hat{\beta}$ - BLUE

$$\hat{\beta}_2 = \frac{\hat{Cov}(X, Y)}{\hat{Var}(X)} = \frac{\hat{Cov}(X, \beta_1 + \beta_2 \cdot X + \beta_3 W + \varepsilon)}{\hat{Var}(X)} =$$

$$= \beta_2 + \beta_3 \frac{\hat{Cov}(X, W)}{\hat{Var}(X)} + \frac{\hat{Cov}(X, \varepsilon)}{\hat{Var}(X)} \neq 0$$

$\bar{\varepsilon} = 0$
 $E(\varepsilon) = 0$
 $\bar{\varepsilon} \neq 0$

$$E(\hat{\beta}_2) = \beta_2 + \beta_3 E\left(\frac{\hat{Cov}(X, W)}{\hat{Var}(X)}\right) + E\left(\frac{\hat{Cov}(X, \varepsilon)}{\hat{Var}(X)}\right)$$

$$= \beta_2 + \beta_3 \frac{\hat{Cov}(X, W)}{\hat{Var}(X)} + \frac{E(\hat{Cov}(X, \varepsilon))}{\hat{Var}(X)} = 0$$

$$E\left(\frac{1}{n-1} \sum (x_i - \bar{x})(\varepsilon_i - \bar{\varepsilon})\right) = \sum \frac{1}{n-1} (x_i - \bar{x}) \left(\underbrace{E(\varepsilon_i)}_0 - \underbrace{E(\bar{\varepsilon})}_0 \right)$$

$$E(\hat{\beta}_2) = \beta_2 + \beta_3 \cdot \frac{\hat{Cov}(X, W)}{\hat{Var}(X)}$$

if $\beta_3 = 0$ or
 $X \perp W$
 $\sum x_i w_i = 0$
 \Rightarrow unbiased

Stochastic Regressors:

True: $y_i = \beta_1 + \beta_2 x_i + \beta_3 w_i + \epsilon_i$

Est: $y_i = \hat{\beta}_1 + \hat{\beta}_2 x_i + u_i$

$\epsilon \perp x \Rightarrow E(\epsilon_i | X) = 0 \Rightarrow \text{cov}(\epsilon_i, x_i) = 0$ (1) Deterministic (+)

$$\left[\begin{array}{l} E(\epsilon_i^2 | X) = 0 \\ E(\epsilon_i \epsilon_j | X) = 0 \end{array} \right]$$

$\text{cov}(\epsilon_i, x_i) = 0$
 $\Rightarrow x_i$ - exogenous

$\text{cov}(\epsilon_i, x_i) \neq 0$
 $\Rightarrow x_i$ - endogenous

$y_t = \beta_1 + \beta_2 t + \epsilon_t$

(2) Stochastic (\mathcal{T}_{t-1})

$[\pi_t = \beta_1 + \beta_2 \pi_{t-1} + \epsilon_t]$

$\pi_{t-1} = \beta_1 + \beta_2 \pi_{t-2} + \epsilon_{t-1}$

True: $y_i = \beta_1 + \beta_2 x_i + \beta_3 w_i + \epsilon_i$

Est: $y_i = \hat{\beta}_1 + \hat{\beta}_2 x_i + u_i$

$$p \lim_{n \rightarrow \infty} \hat{\beta}_2 = p \lim_{n \rightarrow \infty} \frac{\hat{\text{cov}}(x, y)}{\hat{\text{var}}(x)} = \frac{\text{cov}(x, y)}{\text{var}(x)} = \sum \frac{1}{n-1} (x_i - \bar{x})(y_i - \bar{y})$$

$$= \frac{\text{cov}(x, \beta_1 + \beta_2 x + \beta_3 w + \epsilon)}{\text{var}(x)} =$$

$$= \beta_2 + \beta_3 \frac{\text{cov}(x, w)}{\text{var}(x)} + \frac{\text{cov}(x, \epsilon)}{\text{var}(x)} \quad \begin{array}{l} \text{cov}(x, x) = \\ \text{var}(x) \end{array}$$

$$\star \quad [E(\varepsilon|X)] = 0 \quad \Rightarrow \quad \text{Cov}(\varepsilon, X) = 0$$

$$\text{Cov}(\varepsilon, X) = E(\varepsilon \cdot X) - E(\varepsilon)E(X) = 0$$

$$\parallel$$

$$E(E(\varepsilon|X))$$

$$E(\varepsilon \cdot X) = E(E(\varepsilon \cdot X|X)) \overset{0}{=} E(X \cdot E(\varepsilon|X)) \overset{p}{\parallel}$$

Deterministic regressors:

True: $y_i = \beta_1 + \beta_2 x_i + \varepsilon_i$

Est: $y_i = \hat{\beta}_1 + \hat{\beta}_2 \cdot x_i + \hat{\beta}_3 w_i + \hat{\varepsilon}_i$

$$\bullet \quad \text{Var}(\hat{\beta}_k) = \frac{\sigma_\varepsilon^2}{\text{TSS}_k (1 - R_k^2)}$$

R_k^2 : R^2 of $x_i | x_{-i}$

$$\text{Var}(\hat{\beta}_2) = \frac{\sigma_\varepsilon^2}{\text{TSS}_2 \cdot (1 - \hat{\rho}_{x_1 y}^2)}$$

\bullet $d \downarrow$ t -test: $t \sim t_{n-k}$