

HW 0421

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(a)  $E(x) = \gamma_1 + \theta_1$ ,  $E(z)$  From  $x = \gamma_1 + \theta_1 z + v$  to  
obtain  $x - E(x) = \theta_1 [z - E(z)] + v$

Multiply by  $[z - E(z)]$

$$\Rightarrow [z - E(z)][x - E(x)] = \theta_1 [z - E(z)]^2 + [z - E(z)]v$$

$$\Rightarrow E[(z - E(z))(x - E(x))] = \theta_1 E[(z - E(z))^2] + E(z - E(z))v$$

Assuming  $E(z - E(z))v = 0$

$$\Rightarrow \theta_1 = \frac{E[(z - E(z))(x - E(x))]}{E[(z - E(z))^2]} = \frac{\text{Cov}(z, x)}{\text{Var}(z)}$$

$$\Rightarrow x = \gamma_1 + \theta_1 z + v$$

b. Subtract  $E(y) = \pi_0 + \pi_1 E(z)$  from  $y = \pi_0 + \pi_1 z + u$

$$\Rightarrow y - E(y) = \pi_1 (z - E(z)) + u$$

$$\Leftrightarrow (z - E(z))(y - E(y)) = \pi_1 (z - E(z))^2 + (z - E(z))u$$

Assuming  $E(z - E(z))u = 0$ ,

$$\Leftrightarrow E[(z - E(z))(y - E(y))] = \pi_1 E[(z - E(z))^2]$$

Solving for  $\pi_1$ , we have  $\pi_1 = \frac{E[(z - E(z))(y - E(y))]}{E[(z - E(z))^2]} = \frac{\text{Cov}(z, y)}{\text{Var}(z)}$

$$C. \quad y = \beta_1 + \beta_2 + e = \beta_1 + \beta_2 (\gamma_1 + \theta_{12} + v) + e \\ \rightarrow (\beta_1 + \beta_2 \gamma_1) + \beta_2 \theta_{12} + \beta_2 v + e$$

$$\pi_0 = \beta_1 + \beta_2 \gamma_1, \quad \pi_1 = \beta_2 \theta_{12}, \quad u = \beta_2 v + e$$

$$d. \quad \textcircled{c} \quad \pi_1 = \beta_2 \theta_{12}$$

$$\rightarrow \text{solving } \pi_1 = \beta_2 \theta_{12} \text{ for } \beta_2, \text{ we have } \beta_2 = \frac{\pi_1}{\theta_{12}}$$

$$e) \text{ from (a), } \hat{\theta}_{12} = \frac{\widehat{\text{Cov}(z, x)}}{\widehat{\text{Var}(z)}} = \frac{\sum (z_i - \bar{z})(x_i - \bar{x}) / N}{\sum (z_i - \bar{z})^2 / N}$$

$$\hat{\pi}_1 = \frac{\widehat{\text{Cov}(z, y)}}{\widehat{\text{Var}(z)}} = \frac{\sum (z_i - \bar{z})(y_i - \bar{y}) / N}{\sum (z_i - \bar{z})^2 / N}$$

A consistent estimator if  $z$  is uncorrelated with  $u$

$$\rightarrow \hat{\beta}_2 = \hat{\pi}_1 / \hat{\theta}_{12} = \frac{\sum (z_i - \bar{z})(y_i - \bar{y})}{\sum (z_i - \bar{z})(x_i - \bar{x})} = \frac{\widehat{\text{Cov}(z, y)}}{\widehat{\text{Cov}(z, x)}}$$

$$\begin{aligned} \widehat{\text{Cov}(z, y)} &\xrightarrow{P} \text{Cov}(z, y) \quad \widehat{\text{Cov}(z, x)} \xrightarrow{P} \text{Cov}(z, x) \\ \rightarrow \hat{\beta}_2 = \hat{\pi}_1 / \hat{\theta}_{12} &= \frac{\widehat{\text{Cov}(z, y)}}{\widehat{\text{Cov}(z, x)}} \xrightarrow{P} \frac{\text{Cov}(z, y)}{\text{Cov}(z, x)} = \beta_2 \end{aligned}$$

C10&02

$\beta_2 (\text{WAGE}) (+)$  :  $\nearrow$  wage  $\rightarrow$  incentive for women to supply more labour hours

$\beta_3 (\text{EDUC}) (+)$   $\rightarrow$  opportunities  $\rightarrow$  more labour supply.

$\beta_4 (\text{AGE})$  : Uncertain

$\beta_5 (\text{KIDSL6}) (-)$  taking care children  $\rightarrow$  less time

$\beta_6$  (NWFEDIC) (-) other sources  $\rightarrow$   $\downarrow$  supply labor.

b. Explain: Endogeneity issues arising from

① Simultaneity:  $WAGE$  may be influence labour supply.

$\rightarrow$  Violate OLS assumption.

② Omitted variable.

\* measurement error.

c.  $EXPER$  &  $EXPER^2$  Can be valid instruments.

1. Relevance: both are expected to be positively corr with  $WAGE$ .

2. Exogeneity. Neither  $EXPER$  nor  $EXPER^2$  should have a effect on hours beyond their influence on  $WAGE$ .

3. Independence.

d. yes, the supply equation is identified

e. First-stage  $WAGE = \beta_0 + \beta_1 EXPER + \beta_2 EXPER^2 + \dots$

- Obtain fitted values

- Second stage regression.

$$HOURS = \beta_0 + \beta_1 \widehat{WAGE} + \beta_2 EDUC + \dots$$

- ~~Test~~ model fit

- Interpret results.