

2003 Spring Comp

$$Q1: y_i = \beta' x_i + \gamma' w_i + u_i \quad (i=1, \dots, n)$$

$$E[x_i u_i] = 0$$

$$E[w_i u_i] \neq 0$$

a) Can we estimate β consistently using least squares?

No. Endogeneity even affects the exogenous regressor.

$$\hat{\beta}_{OLS} = (\mathbf{X}' M_w \mathbf{X})^{-1} \mathbf{X}' M_w \mathbf{Y}$$

$$= (\mathbf{X}' M_w \mathbf{X})^{-1} \mathbf{X}' M_w (\mathbf{X} \beta + \mathbf{W} \gamma + \mathbf{u})$$

$$= \beta + (\mathbf{X}' M_w \mathbf{X})^{-1} \mathbf{X}' M_w \mathbf{u}$$

$$= \beta + (\mathbf{X}' \mathbf{X} - \mathbf{X}' \mathbf{W} (\mathbf{W}' \mathbf{W})^{-1} \mathbf{W}' \mathbf{X})^{-1} [\mathbf{X}' \mathbf{u} - \mathbf{X}' \mathbf{W} (\mathbf{W}' \mathbf{W})^{-1} \mathbf{W}' \mathbf{u}]$$

$$= \beta + \left[\sum_{i=1}^n x_i x_i' - \left(\sum_{i=1}^n w_i x_i' \right)' \left(\sum_{i=1}^n w_i w_i' \right)^{-1} \left(\sum_{i=1}^n w_i x_i' \right) \right]^{-1}$$

$$\cdot \left[\sum_{i=1}^n x_i u_i - \left(\sum_{i=1}^n w_i x_i' \right)' \left(\sum_{i=1}^n w_i w_i' \right)^{-1} \left(\sum_{i=1}^n w_i u_i \right) \right]$$

$$\rightarrow \beta - \left[\Sigma_{XX} - \Sigma_{WX} \Sigma_{WW}^{-1} \Sigma_{WX}' \right]^{-1} \left[\Sigma_{WX} \Sigma_{WW}^{-1} E[w_i u_i] \right]$$

$$= \beta \text{ iff } \Sigma_{WX} = 0.$$

b) Suppose $\text{cov}(x_i, w_i') = 0$ and $\mathbf{X}' \mathbf{W} = 0$ ($\Leftrightarrow E[x_i w_i'] = 0$ if $E[x_i] = 0$ or $E[w_i'] = 0$)Let $z_i = \begin{bmatrix} z_{i1} \\ \vdots \\ z_{i2} \end{bmatrix}$ be an instrument for w_i If $E[w_i u_i] = 0$ were true, we could estimate

$$\hat{\gamma}_{OLS} = (\mathbf{W}' M_X \mathbf{W})^{-1} \mathbf{W}' M_X \mathbf{Y}$$

$$= (\hat{\mathbf{W}}' \hat{\mathbf{W}})^{-1} \hat{\mathbf{W}}' \hat{\mathbf{Y}}$$

$$= (\hat{\mathbf{W}}' \hat{\mathbf{W}})^{-1} \hat{\mathbf{W}}' \hat{\mathbf{Y}}$$

where $\hat{\mathbf{W}} = M_X \mathbf{W}$ (i)where $\hat{\mathbf{Y}} = M_X \mathbf{Y}$ (ii)

(i) Residual regression

- Regress W on X , get residual \tilde{W}
- Regress Y on W .

(ii) Double-Residual regression

- Regress W on X , Y on X , get residuals \tilde{W}, \tilde{Y}
- Regress \tilde{Y} on \tilde{W} .

Since $X'W=0$, we will have

$$\begin{aligned}\hat{\gamma}_{OLS} &= (\tilde{W}'\tilde{W})^{-1} \tilde{W}'\tilde{Y} \\ &= (W'(I - P_X)W)^{-1} W'(I - P_X)Y \\ &= (W'W - \underbrace{W'P_X W}_{=0})^{-1} [WY - \underbrace{W'P_X Y}_{=0}] \\ &= (W'W)^{-1} W'Y \\ &= (W'W)^{-1} W'\hat{Y}\end{aligned}$$

But $\hat{\gamma}_{OLS} \xrightarrow{P} \gamma$ by endogeneity. Thus, we will use 2SLS:

$$\begin{aligned}\hat{\gamma}_{2SLS} &= (\hat{W}'M_X\hat{W})^{-1} \hat{W}'M_X Y \quad \text{where } \hat{W} = P_Z W \\ &= (W'P_Z' M_X P_Z W)^{-1} W'P_Z' M_X Y \\ &= (W'P_Z' P_Z W - W'P_Z' P_Z P_Z W)^{-1} W'P_Z M_X Y \\ &= (W'P_Z W - W'P_Z P_Z P_Z W)^{-1} W'P_Z M_X Y\end{aligned}$$

↳ projection onto instruments.

c) i) Regress Y on $X \Rightarrow \hat{\beta} = (X'X)^{-1} X'Y$

$$\Rightarrow \hat{Y} = M_X Y$$

ii) Regress \hat{Y} on W using Z as instrument.

↳ 2SLS: Regress \hat{Y} on $\hat{W} = P_Z W$.

$$\hat{\gamma}_{2SLS}^{(c)} = (W'W)^{-1} W'Y$$

$$= (W'P_Z W)^{-1} W'P_Z M_X Y$$

d) Here, we see that $\hat{\gamma}_{2SLS}^{(b)} = \hat{\gamma}_{2SLS}^{(c)}$ only if $W'P_Z P_X P_Z W = 0$, but this need not be true if Z and X are correlated.

$$\hat{\gamma}_{2SLS}^{(b)} \xrightarrow{p} \hat{\gamma}_{2SLS}^{(c)} \text{ if } E[z_i x_i'] = 0 \text{ or}$$

$$\hat{\gamma}_{2SLS}^{(b)} = \hat{\gamma}_{2SLS}^{(c)} \text{ if } Z'X = 0.$$

Claim: If $E[X_i v_i] = 0$ where $w_i' = z_i' \pi + v_i$, then

$$E[X_i z_i'] = 0$$

$$\text{Pf: } \underbrace{E[X_i w_i']}_{=0} = E[X_i z_i'] \pi + \underbrace{E[X_i v_i]}_{=0}$$

$$\Rightarrow E[X_i z_i'] = 0 \quad (\text{since } \pi \neq 0 \text{ by assumption that } Z \text{ is valid instr. for } X.)$$

Thus, if we assume $E[X_i v_i] = 0$, we have that

$$\hat{\gamma}_{2SLS}^{(b)} \xrightarrow{p} \hat{\gamma}_{2SLS}^{(c)}$$