

$$y = \tilde{X}\beta + \epsilon$$

FGLS $\hat{\Sigma} \hat{\Lambda}$ GLS Σ

$$\text{plim } \sqrt{T} (\hat{\beta}(\hat{\Sigma}) - \hat{\beta}(\Sigma)) = 0$$

$$\text{Var}(\hat{\beta}_{OLS}) = (\tilde{X}'\tilde{X})^{-1} \tilde{X}'\Omega\tilde{X}(\tilde{X}'\tilde{X})^{-1} \geq \text{Var}(\hat{\beta}_{GLS})$$

$$\hat{\beta}_{OLS} \equiv \hat{\beta}(\hat{\Sigma}) = \hat{\beta}(\Sigma) \quad \text{if } \tilde{X} = I \otimes X$$

$$\text{plim} \begin{bmatrix} \hat{\beta}_1 - \beta_1 \\ \hat{\beta}_2 - \beta_2 \end{bmatrix} = \left\{ \frac{1}{T} \begin{bmatrix} X_1' & 0 \\ 0 & X_2' \end{bmatrix} \begin{bmatrix} \hat{\sigma}^{11} I_T & \hat{\sigma}^{12} I_T \\ \hat{\sigma}^{21} I_T & \hat{\sigma}^{22} I_T \end{bmatrix} \begin{bmatrix} X_1 & 0 \\ 0 & X_2 \end{bmatrix} \right\}^{-1} \cdot \left\{ \frac{1}{T} \begin{bmatrix} X_1' & 0 \\ 0 & X_2' \end{bmatrix} \begin{bmatrix} \hat{\sigma}^{11} I_T & \hat{\sigma}^{12} I_T \\ \hat{\sigma}^{21} I_T & \hat{\sigma}^{22} I_T \end{bmatrix} \begin{bmatrix} \epsilon_1 \\ \epsilon_2 \end{bmatrix} \right\} \quad (1)$$

$(\hat{\Sigma} \otimes I_T)^{-1}$

If some equation is misspecified, $\text{plim } \frac{1}{T} X_i' \epsilon_i \neq 0$

$$(1) = M^{-1} \left\{ \text{plim} \frac{1}{T} \begin{bmatrix} \hat{\sigma}^{11} \overbrace{X_1' \epsilon_1}^{\neq 0} + \hat{\sigma}^{12} X_1' \epsilon_2 \\ \hat{\sigma}^{21} \underbrace{X_2' \epsilon_1}_{\neq 0} + \hat{\sigma}^{22} \underbrace{X_2' \epsilon_2}_{=0} \end{bmatrix} \right\}$$

if $\text{plim} \frac{X_1' \epsilon_1}{T} \neq 0 \Rightarrow$ get inconsistent estimators for both equations

How do you test if this is going to be consistent?

$$\begin{aligned} y_1 &= X_1 \beta_{11} + \tilde{X}_2 \beta_{12} + u_2 \\ y_2 &= \tilde{X}_1 \beta_2 + X_2 \beta_{22} + u_2 \end{aligned} \quad \left. \begin{array}{l} \leftarrow \text{assumed } \beta_{12} = 0 \\ \leftarrow \beta_2 = 0 \end{array} \right\} \text{This is seemingly unrelated regression}$$

$$H_0: \beta_{12} = 0 \\ \beta_{21} = 0$$

Wald test:

Estimate $\hat{\pi}$. Unrestricted form

$$\hat{\pi} = [I \otimes (Z'Z)^{-1} Z'] y$$

OLS
eqn by eqn.

$$\sqrt{T}(\hat{\pi} - \pi) \xrightarrow{d} N(0, \Omega \otimes (Z'Z)^{-1})$$

$$y_1 = \frac{\gamma_1}{\Delta} z_1 + \frac{\beta_1 \gamma_2}{\Delta} z_2 + \frac{\beta_1 \gamma_3}{\Delta} z_3 + v_1$$

$$y_2 = \frac{\beta_1 \gamma_1}{\Delta} z_1 + \frac{\gamma_2}{\Delta} z_2 + \frac{\gamma_3}{\Delta} z_3 + v_2$$

overidentified, but restrictions are all nonlinear.

$$H_0: h(\pi) = 0 \quad (\text{all three of } \frac{\beta_1 \gamma_1}{\Delta} / \frac{\gamma_1}{\Delta}, \frac{\beta_1 \gamma_2}{\Delta} / \frac{\gamma_2}{\Delta}, \frac{\beta_1 \gamma_3}{\Delta} / \frac{\gamma_3}{\Delta} \text{ are equal})$$

$$h(\hat{\pi}) \approx \overbrace{h(\pi)}^{=0} + \overbrace{\frac{\partial h}{\partial \pi}}^{\text{matrix}} (\hat{\pi} - \pi) + R$$

$$\Rightarrow h(\hat{\pi}) \approx \frac{\partial h}{\partial \pi} (\hat{\pi} - \pi) = H'(\hat{\pi} - \pi)$$

$$\Rightarrow \sqrt{T} (h(\hat{\pi}) - \underbrace{h(\pi)}_0) = H' \sqrt{T}(\hat{\pi} - \pi) \xrightarrow{d} N(0, H' \Psi H')$$

should be transpose
↓

$$\Rightarrow W = h(\hat{\pi}) \{ H' [\hat{\Omega} \otimes (Z'Z)^{-1}] H' \}^{-1} h(\hat{\pi})^A \sim \chi^2 \text{ (* of over identification)}$$

$$LR = -2 \left[\log L(\hat{\Pi}_{OLS}, \hat{\Sigma}) - \log L(\hat{\Pi}_{ML}, \hat{\Sigma}_{ML}) \right] \overset{A}{\sim} \chi^2 (* OI D)$$

$$\text{where } L = \text{constant} + \frac{1}{2} \log \det(\Omega^{-1}) - \frac{1}{2} \text{tr}(\Omega^{-1} V' V)$$

$$= \text{constant} + \frac{1}{2} \log \det(\Omega^{-1} \otimes I) - \frac{1}{2} [(y - (I \otimes Z) \pi)']$$

$$(\Omega^{-1} \otimes I) (y - (I \otimes Z) \pi)]$$

$$= \text{constant} - \frac{nT}{2} \log \det \Omega$$

concentrating the likelihood form

$$\underline{\underline{3SLS = 2SLS + SUR}}$$

LI \equiv FI if all eqns are just identified

$$y_i = \bar{X}_i \delta_i + u_i$$

$$\begin{bmatrix} y_{11} \\ \vdots \\ y_{1T} \\ \vdots \\ y_{m1} \\ \vdots \\ y_{mT} \end{bmatrix} = \begin{bmatrix} \bar{X}_1 & 0 \\ \vdots & \vdots \\ 0 & \bar{X}_M \end{bmatrix} \begin{bmatrix} \delta_{11} \\ \vdots \\ \delta_{1, r1s1} \\ \vdots \\ \delta_{m1} \\ \vdots \\ \delta_{m, r1s_m} \end{bmatrix} + \begin{bmatrix} u_{11} \\ \vdots \\ u_{1T} \\ \vdots \\ u_{m1} \\ \vdots \\ u_{mT} \end{bmatrix}$$

Can't use OLS due to endogeneity

Looking for:

$$\hat{\delta}_{IV} = (\tilde{W}' \tilde{X})^{-1} \tilde{W}' y$$

where

$$\tilde{W} = \begin{bmatrix} W_1 & 0 \\ \vdots & \vdots \\ 0 & W_M \end{bmatrix}$$

If $W_i = Z (Z' Z)^{-1} Z' \bar{X}_i = P_Z \bar{X}_i$, then this is just 2SLS eqn by equation.

$$\Rightarrow \tilde{w}' = \hat{X}' (\mathbf{I} \otimes P_z)$$

$$\Rightarrow \hat{\delta}_{2SLS} = (\hat{X}' (\mathbf{I} \otimes P_z) \hat{X})^{-1} \hat{X}' (\mathbf{I} \otimes P_z) y$$

But we also know $v(u) = \Sigma \otimes \mathbf{I}_T$

$$\text{Let } \hat{\delta}_{3SLS} = (\hat{X}' (\underbrace{\hat{\Sigma}^{-1}}_{\text{2SLS first to estimate this}} \otimes P_z) \hat{X})^{-1} \hat{X}' (\hat{\Sigma}^{-1} \otimes P_z) y$$

can be shown that $\text{plim}(\hat{\delta}_{3SLS} - \hat{\delta}_{FIML}) \xrightarrow{d} 0$

Still need to make sure we have correct specification

$$\text{FGLS: } \hat{\beta}_{OLS} \rightarrow \hat{\Sigma}^1 \rightarrow \hat{\beta}_{FGLS}(\hat{\Sigma}^1) \rightarrow \hat{\Sigma}^2 \rightarrow \hat{\beta}_{FGLS}(\hat{\Sigma}^2) \rightarrow \hat{\Sigma}^3 \rightarrow \dots \\ \rightarrow \hat{\beta}_{FIML} \quad (\text{Hellwig})$$

If you iterate 3SLS, does this converge to $\hat{\delta}_{FIML}$?

• No. $W_i = P_z X_i = (Z \Pi_i Z_i)$

Why is 3SLS better than 2SLS?

$$\text{Var}(\hat{\delta}_{2SLS}) = \text{plim} \left[(\hat{X}' X)^{-1} \hat{X}' (\Sigma \otimes \mathbf{I}) \hat{X} (\hat{X}' X)^{-1} \right]$$

$$\begin{bmatrix} (\hat{X}_1' X_1)^{-1} & 0 \\ 0 & (\hat{X}_2' X_2)^{-1} \end{bmatrix} \begin{bmatrix} \sigma_{11} (\hat{X}_1' \hat{X}_1) & \sigma_{12} (\hat{X}_1' \hat{X}_2) \\ & \sigma_{22} (\hat{X}_2' \hat{X}_2) \end{bmatrix} \begin{bmatrix} (\hat{X}_1' X_1)^{-1} & 0 \\ 0 & (\hat{X}_2' X_2)^{-1} \end{bmatrix}$$

$$= \begin{bmatrix} \sigma_{11} (\hat{X}_1' \hat{X}_1)^{-1} & \sigma_{12} (\hat{X}_1' \hat{X}_1)^{-1} (\hat{X}_1' \hat{X}_2) (\hat{X}_2' X_2)^{-1} \\ & \sigma_{22} (\hat{X}_2' \hat{X}_2)^{-1} \end{bmatrix}$$

$$\text{Var}(\hat{\beta}_{3SLS}) = [\hat{X}'(\Sigma \otimes I)^{-1} \hat{X}]^{-1}$$

$$\text{Var}(\hat{\beta}_{3SLS}) - \text{Var}(\hat{\beta}_{2SLS})$$

$$g' \{ \hat{X}'(\Sigma \otimes I)^{-1} \hat{X} - \{ [\hat{X}' \hat{X}] [\hat{X}'(\Sigma \otimes I) \hat{X}]^{-1} [\hat{X}' \hat{X}] \} \} g$$

$$\text{let } \Sigma^{-1} \otimes I = P'P \Rightarrow \Sigma \otimes I = P^{-1}(P^{-1})'$$

$$\Rightarrow g' \{ \hat{X}' P' [I - \underbrace{(P^{-1})^{-1} \hat{X} [\hat{X}' P^{-1} (P^{-1})' \hat{X}]^{-1} \hat{X}' P^{-1}}_B] P \hat{X} \} g$$

$$= h' [I - B [B' B]^{-1} B] h \geq 0$$

$$\hat{\beta}_{3SLS} = (\tilde{W}' \tilde{W})^{-1} \tilde{W}' y \quad \tilde{W} = I \otimes Z$$

$$\Rightarrow \hat{\beta}_{3SLS} = \hat{\beta}_{2SLS}$$

What if $T \leq K$? Can't do 2SLS or 3SLS, since cannot estimate reduced form.

$\hat{\beta}_{OIV}$ is more efficient than $\hat{\beta}_{3SLS}$

$$\hat{\beta}_{2SLS(1)} = (W_1' X_1)^{-1} W_1' y_1$$

can choose subset of instruments and still get consistent estimate

$$\Rightarrow \begin{matrix} \text{IV for} \\ \text{each} \\ \text{eqn} \end{matrix} \begin{bmatrix} \hat{\beta}_1 \\ \vdots \\ \hat{\beta}_m \end{bmatrix} \rightarrow \hat{\beta}, \hat{\Gamma} \rightarrow \hat{\Pi} = -\hat{\Gamma} \hat{\beta}^{-1} \rightarrow w_i = [z_i \hat{\Pi}_i \ z_i]$$

where $\hat{\Pi}_i \rightarrow \Pi_i$ and $\hat{\Gamma}_i \rightarrow \Gamma_i$ asymptotically unless we have any misspecification. equivalent to $\hat{\beta}_{3SLS}$