

# System Equations Estimation Methods

## System Equations Models

Seemingly Unrelated (SUR) Model

Simultaneous Equations Model

- Limited Information Estimation Methods  
(Single Equation Estimation Methods)
  - ILS, 2SLS, & (LIML)
- Full Information Estimation Methods  
(System Equation Estimation Methods)
  - 3SLS, I3SLS, & (FIML)

# Seemingly Unrelated Regression (SUR) Models

The models:

$$\begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_m \end{pmatrix} = \begin{pmatrix} X_1 & 0 & \cdots & 0 \\ 0 & X_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & X_m \end{pmatrix} \begin{pmatrix} \beta_1 \\ \beta_2 \\ \vdots \\ \beta_m \end{pmatrix} + \begin{pmatrix} \varepsilon_1 \\ \varepsilon_2 \\ \vdots \\ \varepsilon_m \end{pmatrix}$$

or  $y = X\beta + \varepsilon$

where

$$E[\varepsilon] = 0, \text{var}(\varepsilon) = V = \begin{pmatrix} \sigma_{11}I & \sigma_{12}I & \cdots & \sigma_{1m}I \\ \sigma_{21}I & \sigma_{22}I & \cdots & \sigma_{2m}I \\ \vdots & \vdots & \ddots & \vdots \\ \sigma_{m1}I & \sigma_{m2}I & \cdots & \sigma_{mm}I \end{pmatrix}$$

# Seemingly Unrelated Regression (SUR) Models

Example: System of two equations:

$$y_{1t} = \beta_{10} + \beta_{11}x_{1t} + \beta_{12}x_{2t} + u_{1t}$$

$$y_{2t} = \beta_{20} + \beta_{22}x_{2t} + \beta_{23}x_{3t} + u_{2t}$$

If there exists relationship of the error terms across equations, i.e.  $E(u_{1i}, u_{2i}) \neq 0$ .

Ignorance of this problem by using single equation estimation method (or separate OLS) will lead to less efficient estimators.

# Estimation of SUR Models

Two-step system estimation or FGLS

1. Estimate Var-Cov matrix  $\hat{V}$
2. Estimate all parameters  $\beta_j$  using FGLS.

$$\hat{\beta}_{SUR} = \hat{\beta}_{FGLS} = \left( X' \hat{V}^{-1} X \right)^{-1} X' \hat{V}^{-1} y$$

Then, system equation estimation method will provide a more asymptotically efficient estimators.

However, if there exists specification error problem in any equations, the problem will spread through out the whole system.

# Simultaneous Equations Model

More than one equation.

Jointly dependent or endogenous variables.

General Structural Form model can be stated as:

$$\begin{aligned}\gamma_{11}y_{t1} + \gamma_{21}y_{t2} + \cdots + \gamma_{m1}y_{tm} + \beta_{11}x_{t1} + \cdots + \beta_{k1}x_{tk} &= \varepsilon_{t1} \\ \gamma_{12}y_{t1} + \gamma_{22}y_{t2} + \cdots + \gamma_{m2}y_{tm} + \beta_{12}x_{t1} + \cdots + \beta_{k2}x_{tk} &= \varepsilon_{t2} \\ &\vdots \\ \gamma_{1m}y_{t1} + \gamma_{2m}y_{t2} + \cdots + \gamma_{mm}y_{tm} + \beta_{1m}x_{t1} + \cdots + \beta_{km}x_{tk} &= \varepsilon_{tm}\end{aligned}$$

# Simultaneous Equations Model

Structural Form model:

$$\begin{bmatrix} y_1 & y_2 & \cdots & y_m \end{bmatrix}_t \begin{bmatrix} \gamma_{11} & \gamma_{12} & \cdots & \gamma_{1m} \\ \gamma_{21} & \gamma_{22} & \cdots & \gamma_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ \gamma_{m1} & \gamma_{m2} & \cdots & \gamma_{mm} \end{bmatrix} +$$
$$\begin{bmatrix} x_1 & x_2 & \cdots & x_m \end{bmatrix}_t \begin{bmatrix} \beta_{11} & \beta_{12} & \cdots & \beta_{1m} \\ \beta_{21} & \beta_{22} & \cdots & \beta_{2m} \\ \vdots & \vdots & \ddots & \vdots \\ \beta_{m1} & \beta_{m2} & \cdots & \beta_{mm} \end{bmatrix} = \begin{bmatrix} \varepsilon_{1t} & \varepsilon_{2t} & \cdots & \varepsilon_{mt} \end{bmatrix}$$

or  $y'_t \Gamma + x'_t B = \varepsilon'_t$



# Simultaneous Equations Model

Two endogenous variables models:

$$y_{1i} = \beta_{10} + \beta_{12}y_{2i} + \gamma_{11}x_{1i} + u_{1i}$$

$$y_{2i} = \beta_{20} + \beta_{21}y_{1i} + \gamma_{21}x_{1i} + u_{2i}$$

## Example

### Keynesian Income Determination Model

Consumption Function  $C_t = \beta_0 + \beta_1 Y_t + u_{1t}$

Income Identity  $Y_t = C_t + I_t$

# System Equations Model

Simultaneous Bias occurs when  $Cov(Y_t, u_t) \neq 0$

OLS estimators will be biased, inconsistent, and inefficient.

$$E(\hat{\beta}) \neq \beta \quad \text{Biased}$$

$$p \lim(\hat{\beta}) \neq \beta \quad \text{Inconsistent}$$



# System Equations Model

## Structural Form Model

shows structural relationship between endogenous variables and exogenous variables & lagged endogenous variables.

$$C_t = \beta_0 + \beta_1 Y_t + u_{1t}$$

$$Y_t = C_t + I_t$$

## Reduced Form Model

shows relationship only between endogenous variables and exogenous variables.

$$C_t = \pi_0 + \pi_1 I_t + w_{1t}$$

$$Y_t = \pi_2 + \pi_3 I_t + w_{2t}$$

where

$$\pi_0 = \frac{\beta_0}{1 - \beta_1} \quad \pi_1 = \frac{\beta_1}{1 - \beta_1}$$
$$\pi_2 = \frac{\beta_0}{1 - \beta_1} \quad \pi_3 = \frac{1}{1 - \beta_1}$$
$$w_{1t} = \frac{u_{1t}}{1 - \beta_1} \quad w_{2t} = \frac{u_{2t}}{1 - \beta_1}$$

# Single Equation Estimation Methods

## Indirect Least Squares (ILS)

- Estimate Reduced form model using OLS
- Solve for estimated  $\beta$

Structural Form Model:

$$Y\Gamma + XB = E$$

Reduced Form Model:

$$Y = X\Pi + V$$

ILS: 1. Estimate Reduced Form  $\hat{\Pi} = (X'X)^{-1} X'Y$   
2. Solve for B  $\hat{B}_{ILS} = -\hat{\Pi}\Gamma$

# Single Equation Estimation Methods

## Two-Stage Least Squares (2SLS)

1. Estimate Reduced form model using OLS

Predict endogenous variable ( $\hat{Y}$ )

2. Use  $\hat{Y}$  as Instrumental Variables (IV) for lagged endogenous variables instead of using actual  $Y$ , then, obtain matrix  $\hat{Z}$ , then estimate model using IV technique

$$\hat{B} = (\hat{Z}'\hat{Z})^{-1} \hat{Z}'Y$$

# System Equations Estimation Methods

## Three-Stage Least Squares (3SLS)

If there exists relationship of the error terms across equations, system equation estimation will lead to more asymptotically efficient estimators.

2SLS + construct estimated  $V$  matrix

3-stage perform FGLS using estimated  $S$  matrix

$$\hat{B} = \left[ \hat{Z}' (\hat{V}^{-1} \otimes I) \hat{Z} \right]^{-1} \hat{Z}' (\hat{V}^{-1} \otimes I) y$$

# OLS, ILS, 2SLS, 3SLS

## Single Equation Estimation Methods

OLS:    **biased & inefficient**

ILS:    **biased & Asymptotically efficient**

2SLS:   **biased & Asymptotically efficient**

## System Equation Estimation Methods

3SLS:   **biased & More Asymptotically efficient**

**but if there exists specification error in one or any equation(s), the specification error problem will spread through all equations in the system.**