

Panel Data Regression Models

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Covered Topics

- What is Panel Data?
- Pooled Regression
- Fixed Effect Models
- Random Effect Models
- Other Panel Data Models

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What is Panel Data? (1)

- Multiple dimensioned
- Dimensions, e.g.,
 - cross-section and time
 - node-to-node

What is Panel Data? (2)

Node-to-Node Example

Y_{ij} = flow from node i to node j

$X2_{ij}$ = unit cost between node i
and node j

$X3_{ij}$ = capacity between node i
and node j

What is Panel Data?(3)

with cross-section(i) and time(t) indices

i	t	Y_{it}	$X1_{it}$	XK_{it}
1	1	Y_{11}	$X1_{11}$...	XK_{11}
:	:	:	:	...	:
1	T_1	Y_{1T_1}	$X1_{1T_1}$...	XK_{1T_1}
:	:	:	:	:
N	1	Y_{N1}	$X1_{N1}$...	XK_{N1}
:	:	:	:	...	:
N	T_N	Y_{NTN}	$X1_{NTN}$	XK_{NTN}

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Balanced Panel Data

- Each cross-sections has equal number of time periods

$$\text{or } T_1 = T_2 = \dots = T_N$$

- Simple Data Structure
- Less complicated computation

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Linear Model for Panel Data (1)

$$Y_{it} = \beta_{1it} X_{1it} + \beta_{2it} X_{2it} + \dots + \beta_{Kit} X_{Kit} + \varepsilon_{it}$$

- beta coefficients could be

time - invariant : $\beta_{kit} = \beta_{ki}$ for $\forall t$

section - invariant : $\beta_{kit} = \beta_{kt}$ for $\forall i$

both : $\beta_{kit} = \beta_k$ for $\forall i, t$

Linear Model for Panel Data (2)

- variance of error terms could be

time - invariant : $V(\varepsilon_{it}) = \sigma_i^2$ for $\forall t$

section - invariant : $V(\varepsilon_{it}) = \sigma_t^2$ for $\forall i$

both : $V(\varepsilon_{it}) = \sigma^2$ for $\forall i, t$

Pooled Regression (1)

General Assumption:

$$Y_{it} = \beta_1 X_{1it} + \beta_2 X_{2it} + \dots + \beta_K X_{Kit} + \varepsilon_{it}$$

where

- 1) all the β coefficients are both time and cross-sectional invariant

Pooled Regression (2)

- 2) homoscedastic error terms

$$V(\varepsilon_{it}) = \sigma^2 \text{ for all } i,t$$

or it is also time-invariant and section-invariant

\Rightarrow OLS applies.

Pooled Regression (2)

i	t	Y_{it}	$X1_{it}$...	XK_{it}
1	1	Y_{11}	$X1_{11}$...	XK_{11}
1	2	Y_{12}	$X1_{12}$...	XK_{12}
:	:	:	:	...	:
1	T	Y_{1T}	$X1_{1T}$...	XK_{1T}
:	:	\dot{Y}	:	X	:
N	1	Y_{N1}	$X1_{N1}$...	XK_{N1}
N	2	Y_{N2}	$X1_{N2}$...	XK_{N2}
:	:	:	:	...	:
N	T	Y_{NT}	$X1_{NT}$...	XK_{NT}

Fixed Effect Models (1)

Also called LS Dummy Variable (LSDV) Model. X_1 is constant.

$$Y_{it} = \beta_{1i} + \beta_2 X_{2it} + \dots + \beta_K X_{Kit} + \varepsilon_{it}$$

- 1) all the β coefficients are time-invariant and cross-sectional invariant except that β_1 is section-variant but time-invariant

Fixed Effect Models (2)

2) homoscedastic error terms

$$V(\varepsilon_{it}) = \sigma^2 \text{ for all } i,t$$

or it is time-invariant and section-invariant

\Rightarrow OLS applies if dummy variables are introduced.

Fixed Effect Models (3)

Equivalent LSDV

$$Y_{it} = \beta_{11}D_{1it} + \beta_{12}D_{2it} + \dots + \beta_{1N}D_{Nit} \\ + \beta_2 X_{2it} + \dots + \beta_K X_{Kit} + \varepsilon_{it}$$

where

$$D_{jit} = 1 \text{ if } i=j, j=1,\dots,N \\ = 0 \text{ otherwise}$$

Note that

$$D_{1it} + D_{2it} + \dots + D_{Nit} = 1 \text{ for all } i,t$$

Fixed Effect Models (4)

i	t	Y_{it}	D_{lit}	D_{Nit}
1	1	Y ₁₁	1	...	0
:	:	:	:	...	:
1	T ₁	Y _{1T1}	1	...	0
:	:	:	:	:
N	1	Y _{N1}	0	...	1
:	:	:	:	...	:
N	T _N	Y _{NTN}	0	1

Fixed Effect Models (5)

Alternative form of FEM

$$Y_{it} = \beta_{1t} + \beta_2 X_{2it} + \dots + \beta_K X_{Kit} + \varepsilon_{it}$$

Note that β_1 is time-variant but section-invariant, instead

Equivalent LSDV for this FEM is as follows:

Fixed Effect Models (6)

Equivalent LSDV

$$Y_{it} = \beta_{11}D_{1it} + \beta_{11}D_{2it} + \dots + \beta_{1T}D_{Tit} \\ + \beta_2X_{2it} + \dots + \beta_KX_{Kit} + \varepsilon_{it}$$

where

$$D_{jit} = 1 \text{ if } t=j, j=1, \dots, T \\ = 0 \text{ otherwise}$$

Note that

$$D_{1it} + D_{2it} + \dots + D_{Tit} = 1 \text{ for all } i, t$$

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Fixed Effect Models (7)

Common Problems

- many dummy variables required
- multi-collinearity problem likely
- interpretation of variant coefficients
- What if error terms are heteroscedastic?

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Fixed Effect Models (8)

Heteroscedasticity

-cross-section weight

$$V(\varepsilon_{it}) = \sigma_i^2 \text{ for } \forall t$$

-time weight

$$V(\varepsilon_{it}) = \sigma_t^2 \text{ for } \forall i$$

=>WLS applies

Fixed Effect Models (9)

Heteroscedasticity

-cross-section covariance

$$\text{COV}(\varepsilon_{it}, \varepsilon_{jt}) = \sigma_{ij} = \sigma_{ji} \text{ for } \forall t$$

-auto-correlation

$$\text{COV}(\varepsilon_{is}, \varepsilon_{it}) = \sigma_{st} = \sigma_{ts} \text{ for } \forall i$$

=>FGLS applies

Random Effect Models (1)

or REM for short. Also known as Error Component Models (ECM). X_1 is also constant.

$$Y_{it} = \beta_{1i} + \beta_2 X_{2it} + \dots + \beta_K X_{Kit} + \varepsilon_{it}$$

Similar to FEM except that

$$\beta_{1i} = \beta_1 + \xi_i$$

where ξ_i is cross-sectional variation

Random Effect Models (2)

Case 1 $V(\xi_i)$ is section-invariant
and $C(\varepsilon_{it}, \xi_i)$ is invariant

$$Y_{it} = \beta_1 + \beta_2 X_{2it} + \dots + \beta_K X_{Kit} + \varepsilon'_{it}$$

where $\varepsilon'_{it} = \varepsilon_{it} + \xi_i$

Random Effect Models (3)

Note that

$$V(\varepsilon'_{it}) = V(\varepsilon_{it}) + V(\xi_i) + 2C(\varepsilon_{it}, \xi_i)$$

$V(\varepsilon'_{it})$ is invariant.

\Rightarrow OLS applies. Trivial.

Random Effect Models (4)

Case 2 $V(\xi_i)$ is section-variant

and/or $C(\varepsilon_{it}, \xi_i)$ is section-variant

Note that $V(\varepsilon'_{it})$ is section - variant.

Error term is heteroscedastic

\Rightarrow FGLS applies. How?

$$V(\varepsilon') = \begin{bmatrix} \sigma_1^2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & \sigma_1^2 & 0 & 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & \ddots & 0 & 0 & 0 & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & 0 & 0 & \ddots & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 0 & \sigma_N^2 & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \sigma_N^2 \end{bmatrix}$$

Random Effect Models (5)

Case 3 $V(\xi_i)$ is section-variant

and/or $C(\varepsilon_{it}, \xi_i)$ is section-variant

but $\varepsilon_{it} = \varepsilon_i$

Note that $V(\varepsilon'_{it})$ is section - variant.

Error term is general

\Rightarrow FGLS applies. How?

$$V(\varepsilon') = \begin{bmatrix} \sigma_1^2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & \sigma_1^2 & 0 & 0 & 0 & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & \ddots & 0 & 0 & 0 & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & 0 & 0 & \ddots & 0 & 0 & 0 \\ \hline 0 & 0 & 0 & 0 & 0 & 0 & \sigma_N^2 & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \sigma_N^2 \end{bmatrix}$$

FEM vs REM (1)

They are substitute if no theoretical preference. Note that

- 1) FEM is preferred when T is large and N is small.
- 2) T is small but N is large. Degree of freedom for FEM is small. REM is more efficient.

FEM vs REM (2)

3) For T is small and N is large, FEM is preferred if cross-sectional variation (ξ_i) is non-random. Otherwise, REM is preferred.

FEM vs REM (3)

4) ξ_i and X_{kit} are correlated. FEM yields unbiased estimator but REM yields biased estimator

Cross-sectional Heteroscedasticity (1)

$$Y_{it} = \beta_{1i} + \beta_{2i} X_{2it} + \dots + \beta_{Ki} X_{Kit} + \varepsilon_{it}$$

Assume time-invariant variance-covariance for error terms.

In addition to possibility of different cross-section weights (variances), covariances between errors of cross sections could be non-zero.

$$V(\varepsilon) = \begin{bmatrix} \begin{array}{ccc|ccc|ccc} \text{Cross-section\#1} & & & & & & \text{Cross-section\#N} & & \\ \hline \sigma_1^2 & 0 & 0 & \sigma_{12} & 0 & 0 & \sigma_{1N} & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & \sigma_1^2 & 0 & 0 & \sigma_{12} & 0 & 0 & \sigma_{1N} \\ \hline \sigma_{12} & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & \sigma_{12} & 0 & 0 & \ddots & 0 & 0 & \ddots \\ \hline \sigma_{1N} & 0 & 0 & \ddots & 0 & 0 & \sigma_N^2 & 0 & 0 \\ 0 & \ddots & 0 & 0 & \ddots & 0 & 0 & \ddots & 0 \\ 0 & 0 & \sigma_{1N} & 0 & 0 & \ddots & 0 & 0 & \sigma_N^2 \end{array} \end{bmatrix}$$

Cross-sectional Heteroscedasticity (2)

$$V(\boldsymbol{\varepsilon}_{it}) = \boldsymbol{\sigma}_j^2$$

$$\text{Cov}(\boldsymbol{\varepsilon}_{it}, \boldsymbol{\varepsilon}_{jt}) = \boldsymbol{\sigma}_{jj} \text{ for all } t$$

or they are time-invariant but section-variant

=> WLS will not apply as there are non-zero covariances between observations. Need GLS or FGLS.