Fixed-effects regressions

Data Science and Causal Inference Workshop 2025

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Forms of endogeneity: Why a regression may fail



Reverse causality

Example: Does bank performance drive CEO compensation, or does CEO compensation drive bank performance?

2. Selection bias

Example: Firms choosing to issue bonds vs taking bank loans based on unobservable characteristics

3. Omitted variable bias

Example: Unobservable managerial quality affecting both firm performance and capital structure decisions

The fixed effects approach



- Entity fixed effects control for time-invariant unobservable characteristics by:
 - Removing all between-variation (differences across entities)
 - Focusing only on within-variation (changes within entities over time)
 - Controlling for unobservable characteristics that don't change over time
- Banking example (entity and time fixed effects):
 - Bank fixed effects control for time-invariant bank characteristics like charter value, organizational culture and geographic location
 - Time fixed effects control for common shocks across all banks in each year/quarter (e.g., financial crisis, Covid-19)

Implementation with panel data



Two equivalent approaches:

1. De-meaning method: subtract entity-specific means from all variables

$$Y_{it} - \overline{Y}_i = \beta(X_{it} - \overline{X}_i) + (\varepsilon_{it} - \overline{\varepsilon}_i)$$

2. Least Squares Dummy Variable (LSDV): include dummy variables for each entity

$$Y_{it} = \beta X_{it} + \alpha_i + \varepsilon_{it}$$

Both approaches generalise to settings with high-dimensional fixed effects (e.g., one possibility for corporate loans: *firm* + *bank* + *time* + *location*)

Limitations of fixed effects



- Key considerations:
 - Cannot control for time-varying omitted variables
 - Cannot estimate the effect of time-invariant variables (e.g., bank headquarters location)
 - Requires sufficient within-variation in variables of interest
 - May have lower statistical power due to removed variation
- Important note:
 - Fixed effects are not a panacea they only address specific sources of endogeneity

Clustering standard errors



- Why cluster at entity level?
 - Observations within each entity (e.g., bank) are likely correlated over time
 - Bank's risk-taking in 2023 is not independent from its risk-taking in 2022
 - Standard errors will be understated without clustering
- Serial correlation in panel data:
 - Firm characteristics persist over time
 - Shocks often affect the same entity for multiple periods
 - Corporate policies show strong autocorrelation
- Panel data best practice: cluster at least at the entity level



Illustration with R, Python and Stata

Fixed-effects regressions with "fixest" (R), "pyfixest" (Python) and "reghdfe" (Stata)



Positron IDE (by the makers of RStudio)
Seamless switching between R and Python
https://positron.posit.co