

# Topic 6: Fixed Effects, Difference-in-differences, and Factor Models

*ECON 5783 – University of Arkansas*

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# Imputation Estimator review

The last set of slides, we introduced an “imputation estimator” for panel data treatment effects:

1. Estimate model for  $y_{it}(\infty)$  using observations with  $d_{it} = 0$  and get fitted values for full sample,  $\hat{y}_{it}(\infty)$
2. Regress  $y_{it} - \hat{y}_{it}(\infty)$  on  $d_{it}$  or event-study indicators to estimate treatment effects  
→ Estimating the overall effect, ATT, or dynamic effects of being treated for  $\ell$  periods,  $ATT^\ell$  respectively

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This topic will extend this procedure to “factor models” that will allow more general trending behavior

## Factor Models

# Factor Model

Untreated potential outcomes are given by a factor model:

$$y_{it}(0) = \sum_{r=1}^p f_{t,r} * \gamma_{i,r} + u_{it}$$

- $f_{t,r}$  is the  $r$ -th **factor** (macroeconomic shock) at time  $t$ .
- $\gamma_{i,r}$  is unit  $i$ 's **factor loading** (exposure) to the  $r$ -th factor.

# Factor Model Example

$$y_{it}(0) = \sum_{r=1}^p f_{t,r} * \gamma_{i,r} + u_{it}$$

If we are thinking about housing prices,  $y_{it}$ :

- $\gamma_i$  are characteristics of neighborhood / house
- $f_t$  are demand shocks in each period

# Factor Model Example

$$y_{it}(0) = \sum_{r=1}^p f_{t,r} * \gamma_{i,r} + u_{it}$$

If we are thinking about wages,  $y_{it}$ :

- $\gamma_i$  are worker's latent skills (e.g. computer skills)
- $f_t$  reflect changing firm's demand for skills

# Factor Model Example

$$y_{it}(0) = \sum_{r=1}^p f_{t,r} * \gamma_{i,r} + u_{it}$$

If we are thinking about county employment,  $y_{it}$ :

- $\gamma_i$  are characteristics of a county (e.g. their manufacturing share)
- $f_t$  reflect national shocks to the economy (e.g. the “China shock”)



## Two-way Fixed Effect vs. Factor Model

The factor model is a generalization of the TWFE model. If  $\mathbf{f}_t = (\lambda_t, 1)'$  and  $\boldsymbol{\gamma}_i = (1, \mu_i)'$ , then our factor model becomes the TWFE model:

$$y_{it}(0) = \mathbf{f}_t' \boldsymbol{\gamma}_i + u_{it} = \lambda_t + \mu_i + u_{it}$$

We can add unit and/or time fixed-effects as 'known' factors if we want

# Factor Model and Parallel Trends

Say you have a single treatment timing and two periods. Let  $D_i$  be out treated group indicator. Then

$$\begin{aligned}\mathbb{E}[\Delta y_i \mid D_i = d] &= \mathbb{E}[y_{i1} - y_{i0} \mid D_i = d] \\ &= \Delta \mathbf{f} \mathbb{E}[\boldsymbol{\gamma}_i \mid D_i = d]\end{aligned}$$

Under a factor model, the average change in  $y_{it}$  for group  $D_i = d$  is the change in factor shocks  $\mathbf{f}$  times the average exposure to those shocks

# Factor Model and Parallel Trends

$$\mathbb{E}[\Delta y_i \mid D_i = d] = \Delta \mathbf{f} \mathbb{E}[\gamma_i \mid D_i = d]$$

Say the treated group has higher exposure to a shock than the control group

- $\implies$  the trends differ by treatment status

# Factor Model and Parallel Trends

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- $\implies$  the trends differ by treatment status

That is, a factor model allows for “non-parallel trends” based on difference in exposures to shocks

# Example

$$y_{it}(0) = \sum_{r=1}^p f_{t,r} * \gamma_{i,r} + u_{it}$$

Say we are thinking about neighborhood housing prices,  $y_{nt}$ . We are interested in some treatment  $D_n$ , e.g. access to subways.

- Say  $\gamma_n$  is the walkability of the neighborhood
- $f_t$  are demand shocks for walkable neighborhoods

If new subways are built in more walkable neighborhoods, then we do not believe parallel trends hold in this setting