

# The Network of Foreign Direct Investment Flows: Theory and Empirical Analysis<sup>1</sup>

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- Motivation
  - Explain FDI networks
  - Violation of Independence Assumptions
  - Theoretical Importance of Dependence Terms
- FDI as a Network
  - Reciprocity
  - Transitivity
- Simultaneously test exogenous variables

- Dyad-level Covariates

- Gravity +
- Contiguity +
- Common Language +
- Four Types of Defense Treaties +
- Colonial Relationships +
- PTA depth +

- Country-level Covariates (s/r)

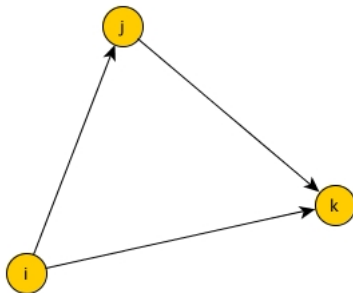
- GDP per capita +/-
- GDP Growth Rate +/-
- Polity IV +/-
- Political Violence -/-
- Trade Openness +/-

- Reciprocity



- Standard practice to resolve political opposition from competing firms
- Example: Chinese firms' mergers (Tingley et. al. 2015)

- Transitivity



- MNC expansion and supply-chain fragmentation
- PTA networks
- Example: Volkswagen and EU

# The Count Exponential Random Graph Model (ERGM)

The probability (likelihood function) of observing the network is:

$$\Pr_{\theta;h,g}(Y = y) = \frac{h(y)\exp(\theta \cdot g(y))}{\kappa_{h,g}(\theta)}$$

Decomposition:

$\underbrace{h(y)}$	$\underbrace{\theta}$	$\underbrace{g(y)}$	$\underbrace{\kappa_{h,g}(\theta)}$
<i>Distribution</i>	<i>Effects</i>	<i>Net Stats</i>	<i>Normalizer</i>

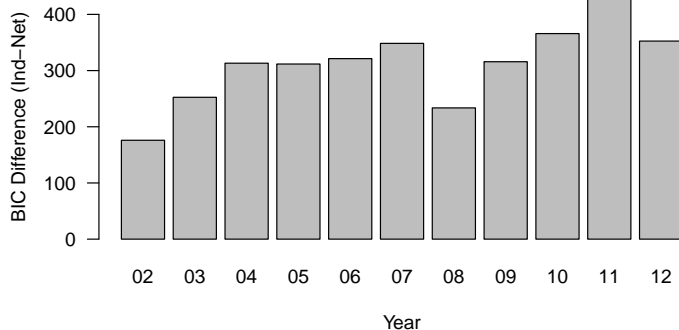
Constants:

- Sum,  $\text{Sum}^{(1/2)}$ , and Nonzero

$$\text{Reciprocity : } g(\mathbf{y}) = \sum_{(i,j) \in \mathbb{Y}} \min(\mathbf{y}_{i,j}, \mathbf{y}_{j,i})$$

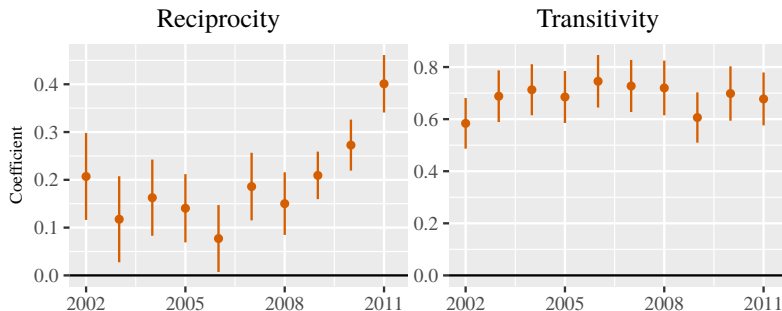
$$\text{Transitive Weights : } g(\mathbf{y}) = \sum_{(i,j) \in \mathbb{Y}} \min \left( \mathbf{y}_{i,j}, \max_{k \in N} \left( \min(\mathbf{y}_{i,k}, \mathbf{y}_{k,j}) \right) \right),$$

## BIC Difference between Models

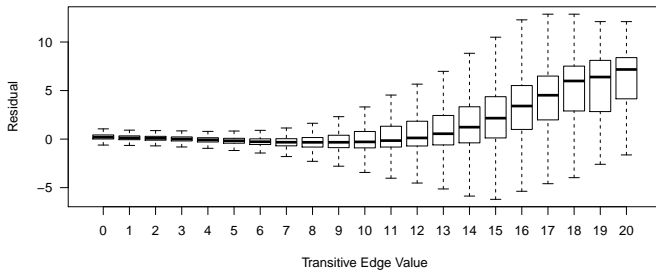
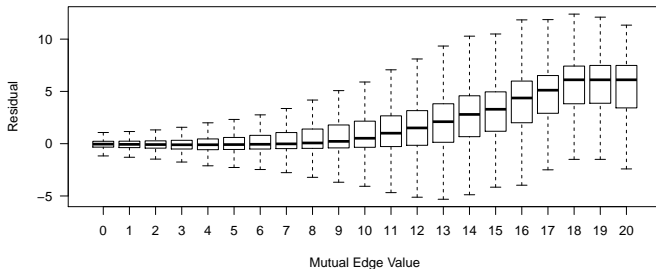




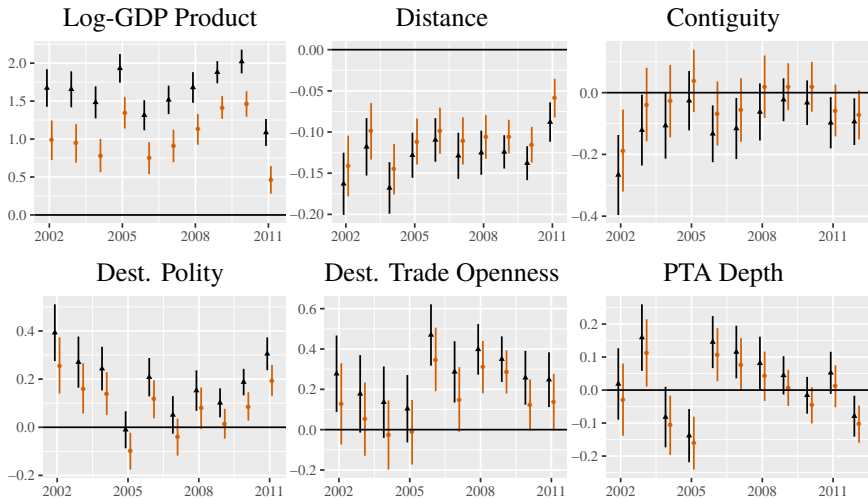
# Count Model and Network Dependencies



# Network Statistics



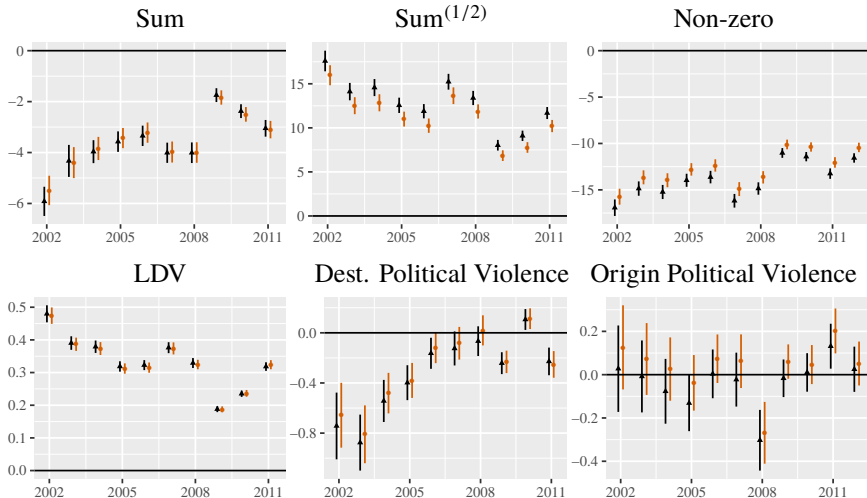
# Covariate Results



# Conclusion and Future Research

- Network terms are substantively important
- Network terms need to be modeled instead of being assumed away
- Network dynamics and methodological constraints

# Additional Covariates



$$\text{Sum : } g(\mathbf{y}) = \sum_{(i,j) \in \mathbb{Y}} y_{i,j}$$

$$\text{Sum, Fractional Moment : } g(\mathbf{y}) = \sum_{(i,j) \in \mathbb{Y}} y_{i,j}^{1/2}$$

$$\text{Non-Zero : } g_k = \sum_{(i,j) \in \mathbb{Y}} \mathbb{I}(y_{i,j} \neq 0)$$

$$\text{Dyadic Covariate : } g(\mathbf{y}, \mathbf{x}) = \sum_{(i,j)} \mathbf{y}_{i,j} x_{i,j}$$

$$\text{Sender Covariate : } g(\mathbf{y}, \mathbf{x}) = \sum_i x_i \sum_j \mathbf{y}_{i,j}$$

$$\text{Receiver Covariate : } g(\mathbf{y}, \mathbf{x}) = \sum_j x_j \sum_i \mathbf{y}_{i,j}$$