

# Generative Dynamics of Supreme Court Citations

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August 31, 2018

This work was supported in part by NIH, R01 AI36664-01; and NSF grants SES- 1558661, SES-1637089, SES-1619644, and CISE-1320219.



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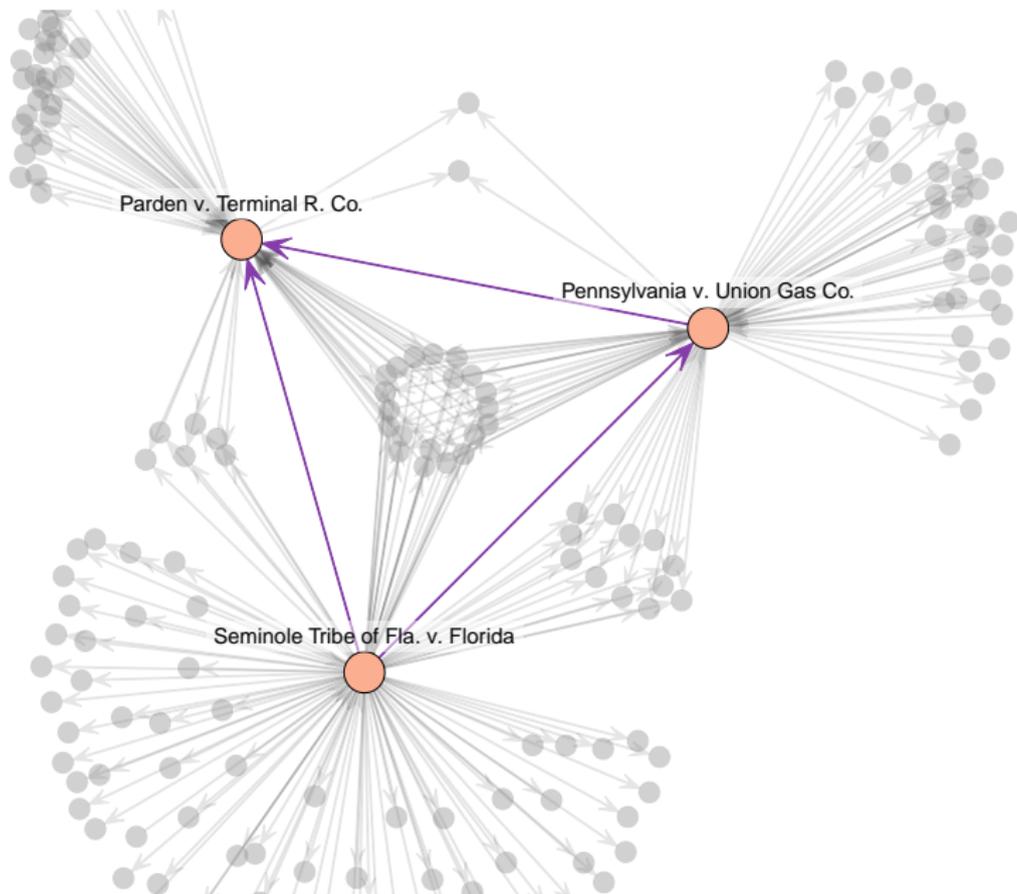
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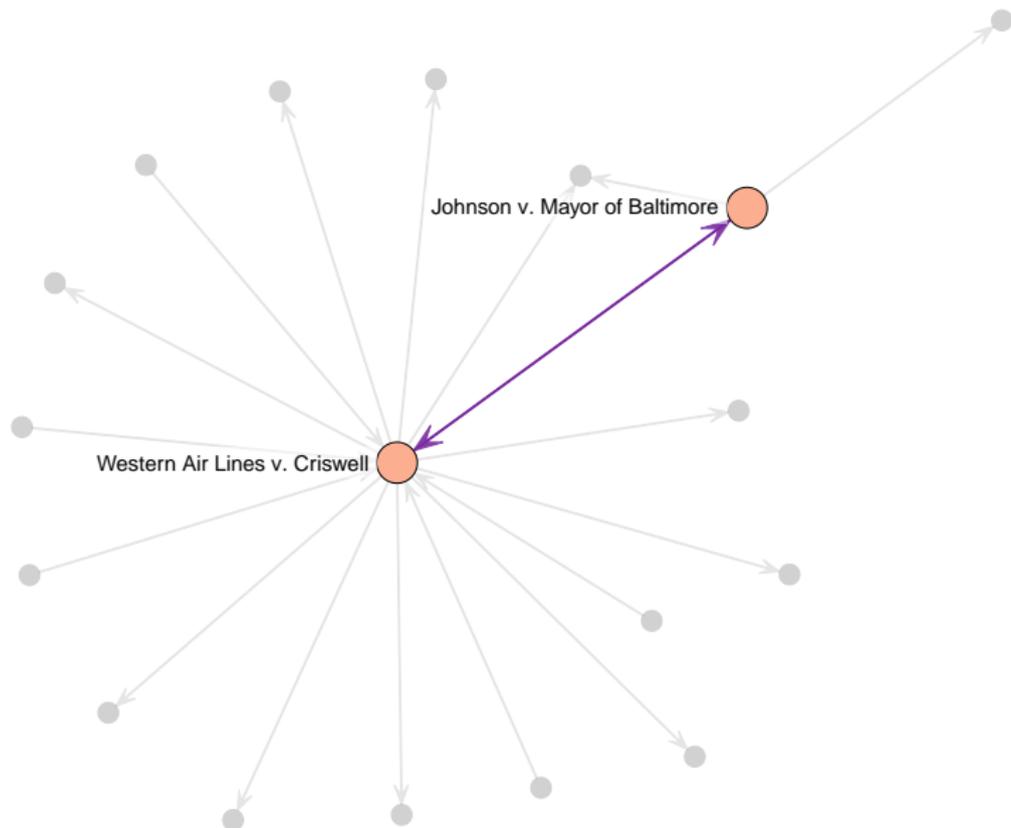
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We propose a dyadic, network-analytic approach.

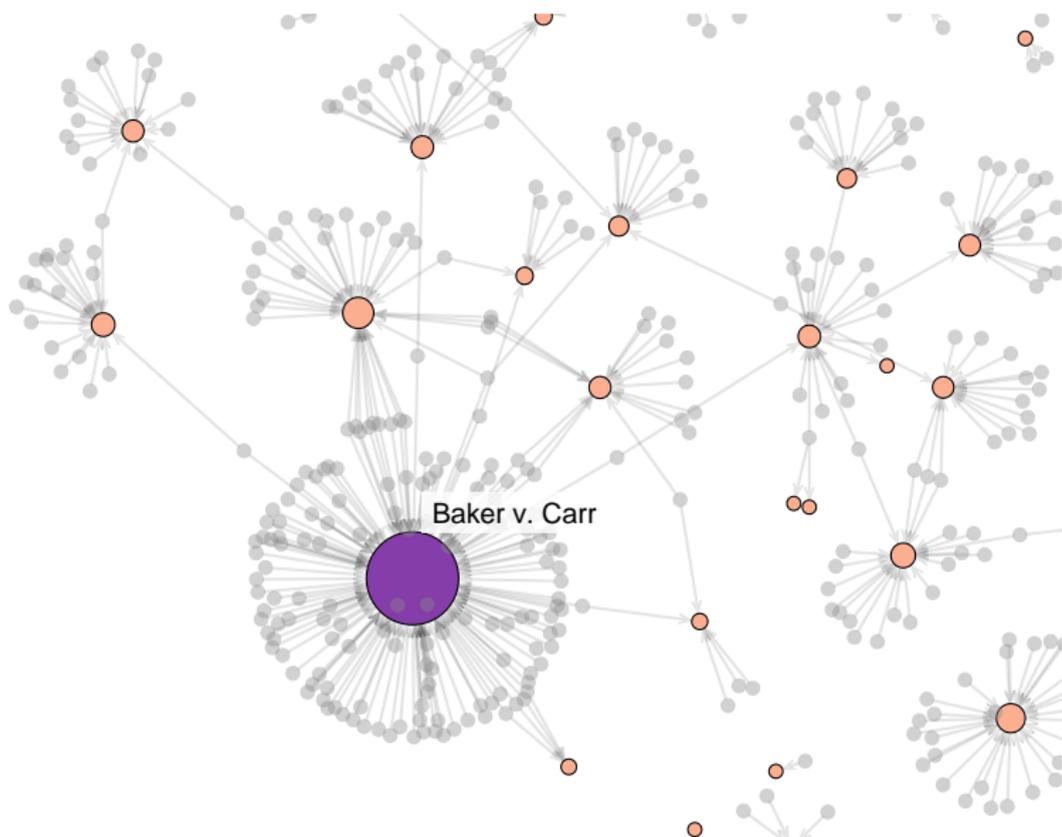
# Transitivity



# Reciprocity



# Popularity and Activity



# Characteristics of the Supreme Court Citation Network

- ▶ Acyclic, but only partially.

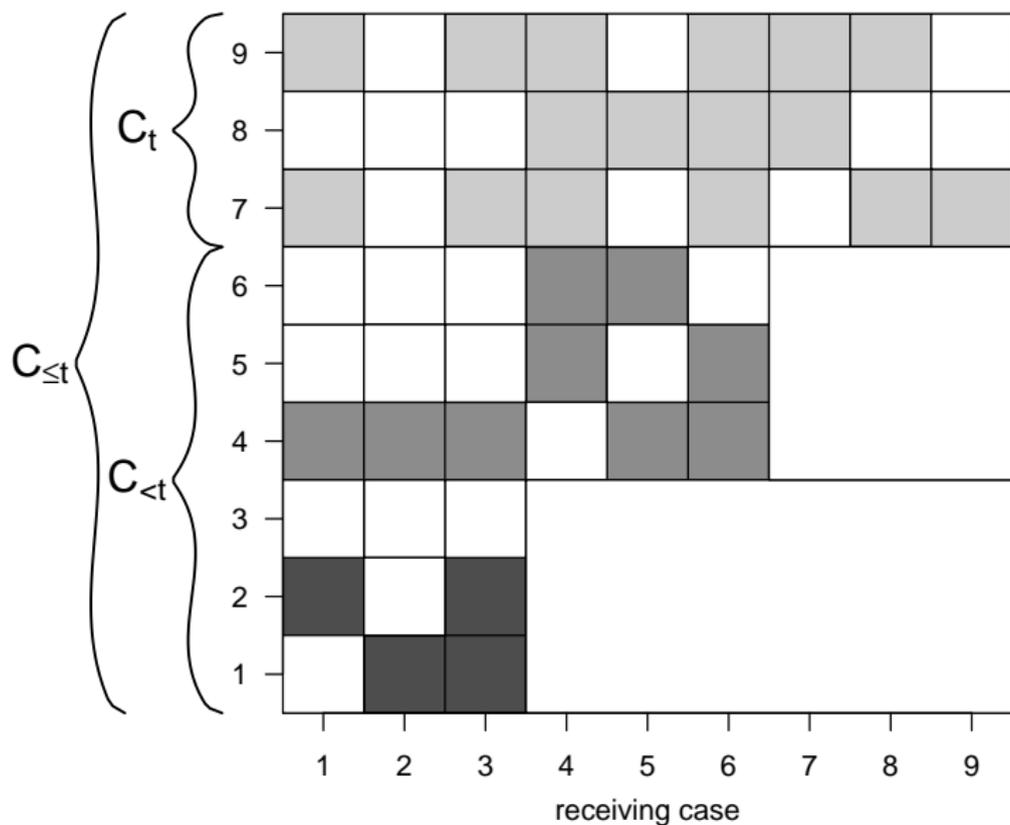
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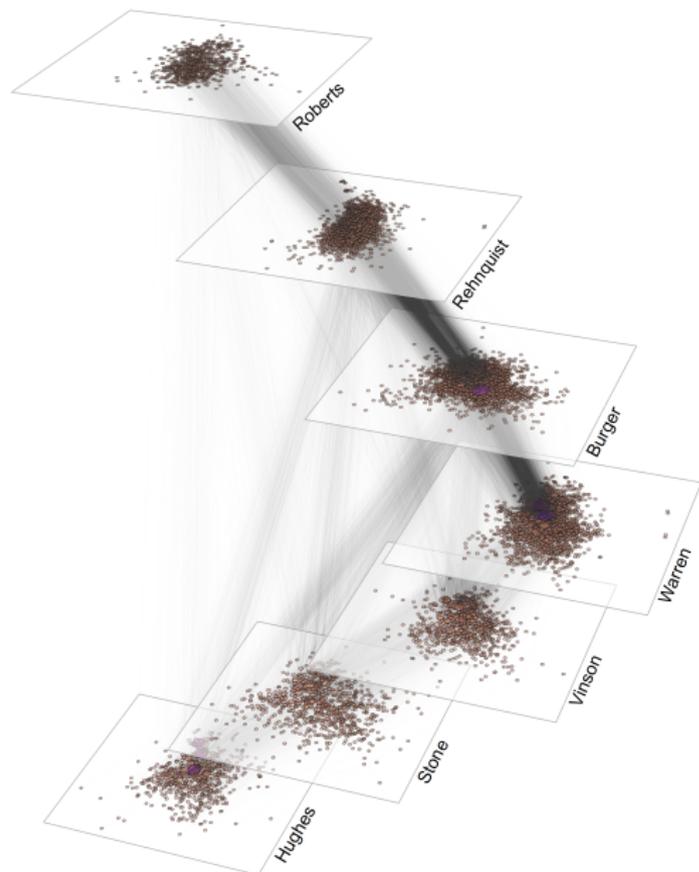
- ▶ Acyclic, but only partially.
- ▶ Edges can be formed but never destroyed.
- ▶ Edges can only be formed with nodes entry.

# The Citation-TERGM



# Data

	Terms	Cases
CE Hughes	1937 - 1941	629
HF Stone	1942 - 1946	766
FM Vinson	1946 - 1953	788
E Warren	1954 - 1969	2159
WE Burger	1970 - 1986	2805
W Rehnquist	1987 - 2005	2022
J Roberts	2006 - 2015	796



# c-TERGM Specification

## Dependence Effects

- ▶ # Transitive Ties
- ▶ # Out-two-stars
- ▶ # Mutual Dyads
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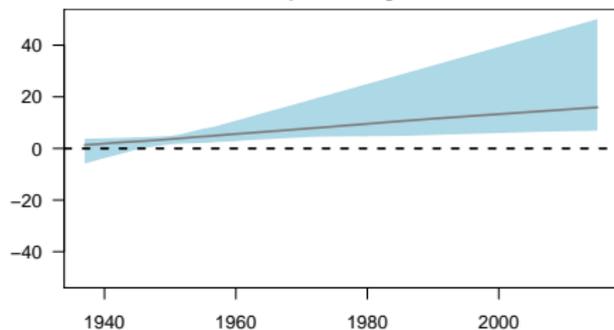
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## Exogenous Effects

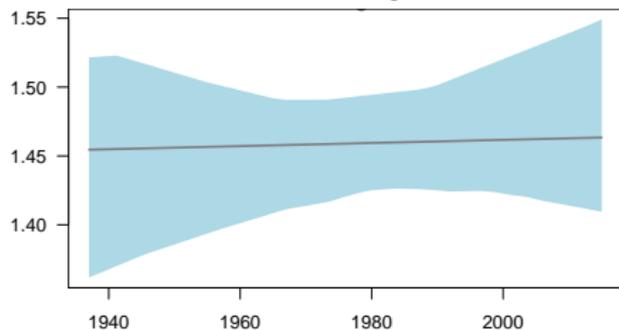
- ▶ Ideological Distance (dyad)
- ▶ Issue Area (sender; receiver; mixing matrix)
- ▶ Justice (sender; receiver; homophily)
- ▶ Age (receiver)
- ▶ Overruled Case (receiver)
- ▶ Majority size (receiver)
- ▶ Ideological Breadth (receiver)

# Results: Dependence Effects

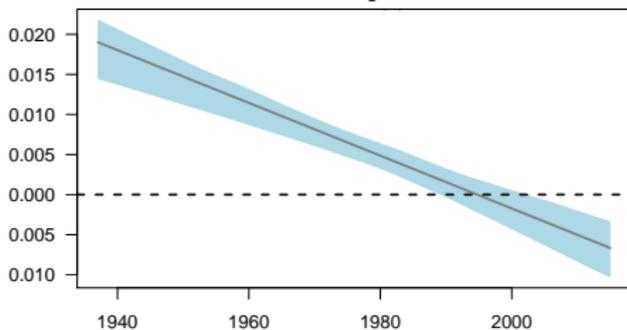
## Reciprocity



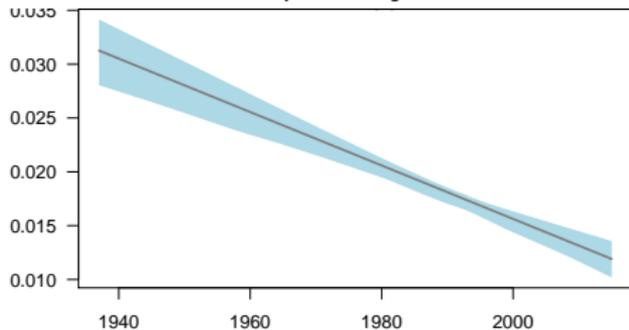
## Transitivity



## Activity



## Popularity

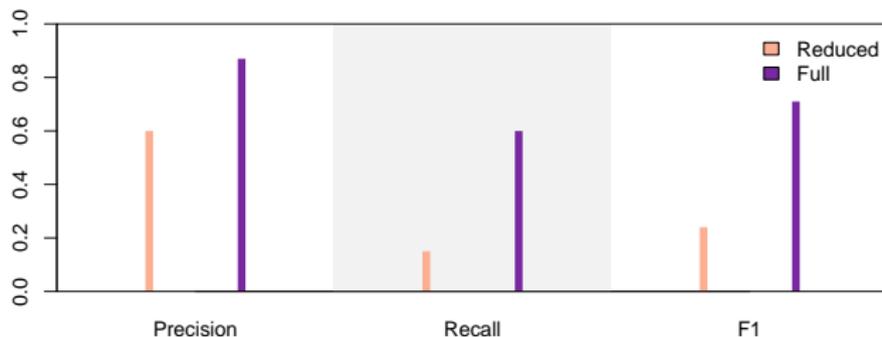


## Results: Predictive Performance

- ▶ Randomly split dyads 80/20 into training and test sets
- ▶ Estimate models using training set, then predict test set
- ▶ This process was repeat ten times

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## **Key Findings**

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## **Limitations**

1. Do not consider the signs of the citations
2. We miss recent terms
3. Case emergence considered exogenous

Questions or Comments?

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A copy of the slides will be available at my website: <https://tedhchen.com>  
or our project github page:  
[https://github.com/desmarais-lab/Supreme\\_Court\\_Citation\\_Network](https://github.com/desmarais-lab/Supreme_Court_Citation_Network).

# Citation Temporal Exponential Random Graph Model

The likelihood of the c-TERGM is given by

$$l(\theta, C_{\leq T}) = \prod_{t=1}^T \frac{\exp[\theta' \mathbf{h}(C_t, C_{<t})]}{\sum_{C_t^* \in \mathcal{C}_t} \exp[\theta' \mathbf{h}(C_t^*, C_{<t})]}. \quad (1)$$

Decomposition:

$$\underbrace{\mathbf{h}(C_t, C_{<t})}_{\text{Net Stats}} \quad \underbrace{\theta}_{\text{Effects}} \quad \underbrace{\exp[\theta' \mathbf{h}(C_t^*, C_{<t})]}_{+ \text{ Weight}} \quad \underbrace{\sum_{C_t^* \in \mathcal{C}_t} \exp[\theta' \mathbf{h}(C_t^*, C_{<t})]}_{\text{Normalizer}}$$

$\mathbf{h}$  can capture virtually any form of interdependence among the edges  
+ covariates

# Issue Areas

1	Criminal Procedure	8	Economic Activity
2	Civil Rights	9	Judicial Power
3	First Amendment	10	Federalism
4	Due Process	11	Interstate Relations
5	Privacy	12	Federal Taxation
6	Attorneys	13	Miscellaneous
7	Unions	14	Private Action