

Inferring Latent Influence Diffusion Networks in The United States Senate

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Influence In The U.S. Congress



Figure: John McCain [R] and Charles Schumer [D].

Legislative Influence

- ▶ Want to measure the influence of legislators on each other.
- ▶ Measurement: seniority? position? money? **personal relationships?**
- ▶ Use temporal patterns in bill cosponsorship as evidence.

The Data – Bill Cosponsorship Times

Example Bill: S.2715, emergency unempl. benefits for 9/11 victims

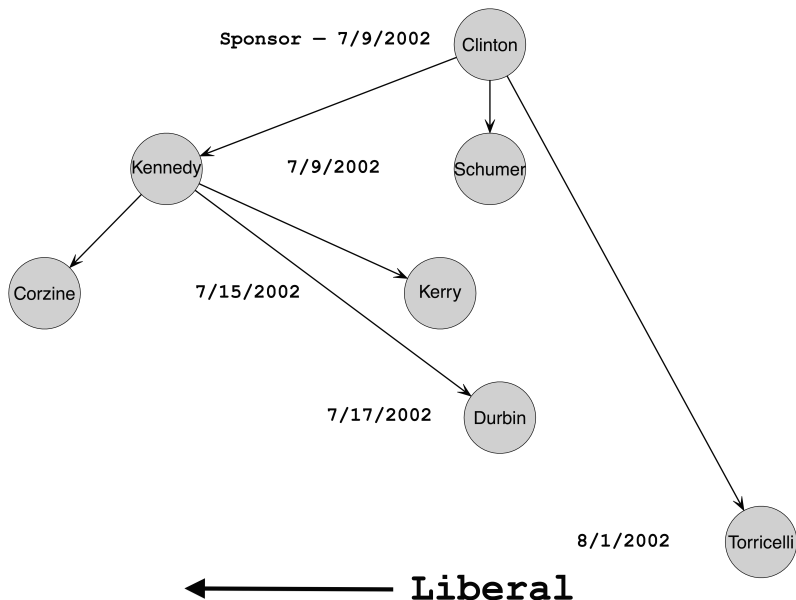
Sponsor

- ▶ Clinton – 7/9/02

Cosponsors

- ▶ Kennedy – 7/9/02
- ▶ Schumer – 7/9/02
- ▶ Corzine – 7/15/02
- ▶ Kerry – 7/15/02
- ▶ Durbin – 7/17/02
- ▶ Torricelli – 8/1/02

Want To Infer: Influence Relationships



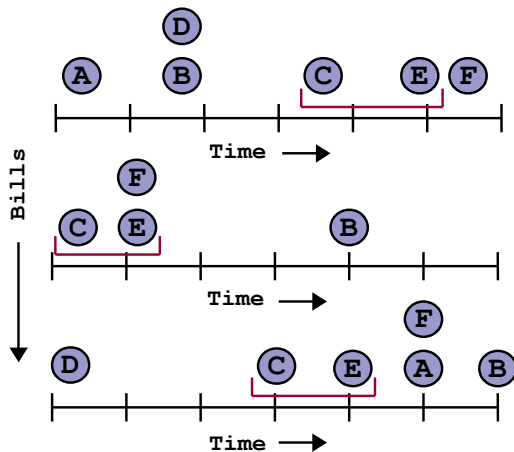
Inferring Influence

Inference Strategy

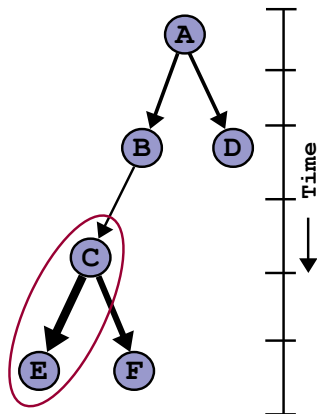
- ▶ Assume cosponsorship patterns arise from underlying cascade of influence.
- ▶ Aggregate over many bills to infer underlying influence network.
- ▶ Gomez Rodriguez, M., Leskovec, J., & Krause, A. (2010). "**Inferring networks of diffusion and influence**". *KDD*

Traces of Influence

Bill Cosponsorship Delay



Temporal Cascades



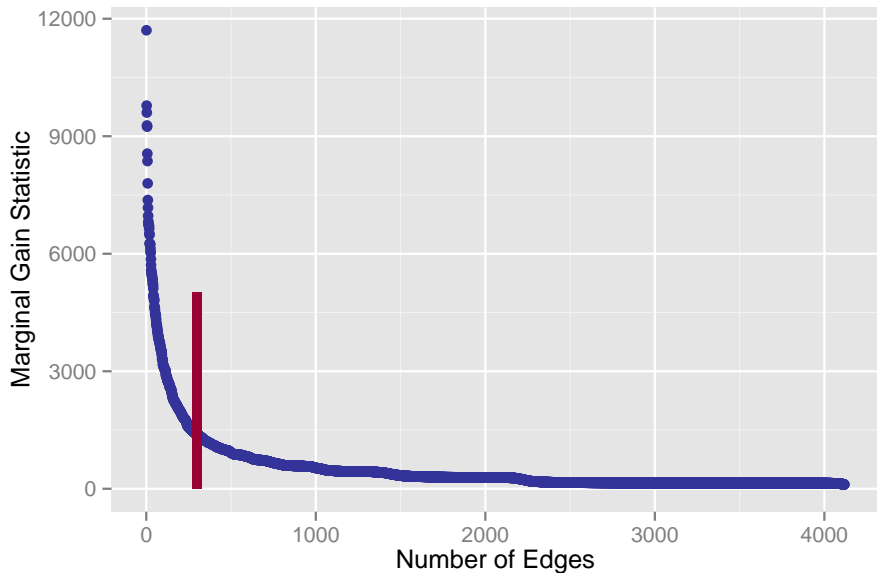
Diffusion Model

- ▶ Want to infer best network to explain observed cascades.

$$P(\Delta) \propto e^{-\frac{\Delta}{\alpha}} \quad (1)$$

- ▶ Δ – time difference between action for two nodes.
- ▶ α – inverse average time diff. between a node and its potential parents

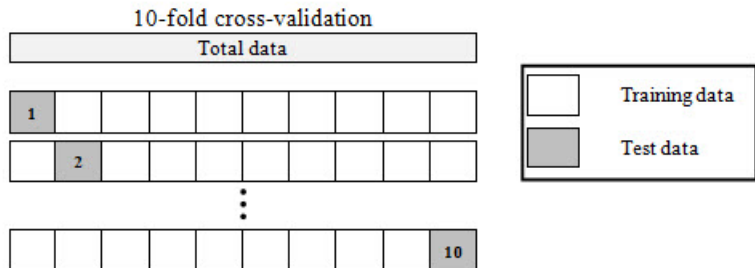
How Many Edges to Include in Network?



Model Selection

Cross Validation

- ▶ Jensen, D. D., & Cohen, P. R. (2000). **Multiple Comparisons in Induction Algorithms.** *Machine Learning*, 309338.



Strategy

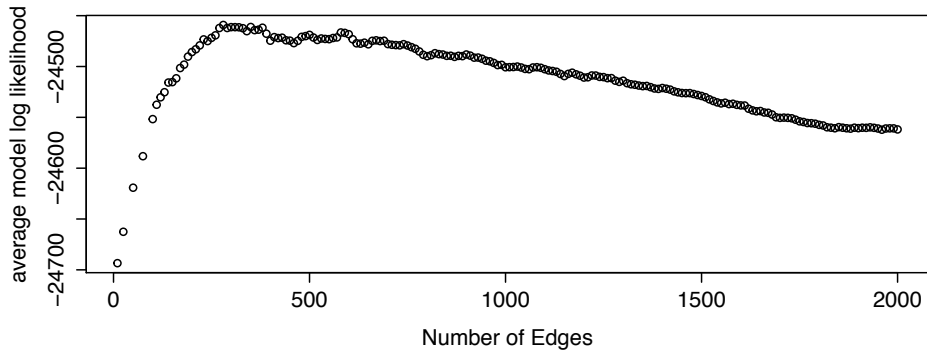
- ▶ Predict when Senators will cosponsor in held-out sample.
- ▶ Fit event history models for model selection.
- ▶ Optimization over # edges and hyper-parameter (10 80/20 splits)
- ▶ Grid Search!

Rare Events Logistic Regression

- ▶ King, G., & Zeng, L. (2001). **Logistic regression in rare events data.** *Political Analysis*, 9(2), 137-163.

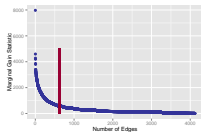


Use Model Log Likelihood For Selection.

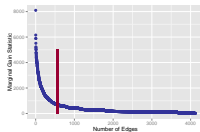


Cross Validation Results

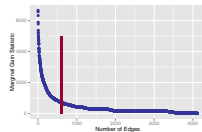
1981-1982



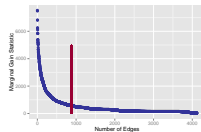
1983-1984



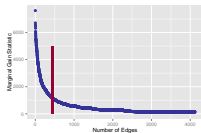
1985-1986



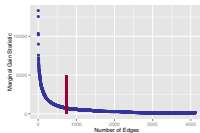
1987-1988



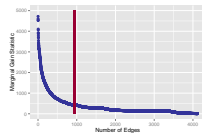
1989-1990



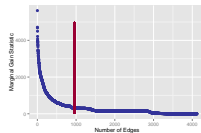
1991-1992



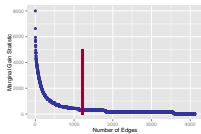
1993-1994



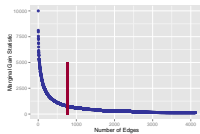
1995-1996



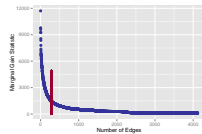
1997-1998



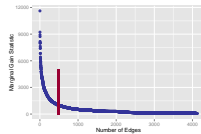
1999-2000



2001-2002

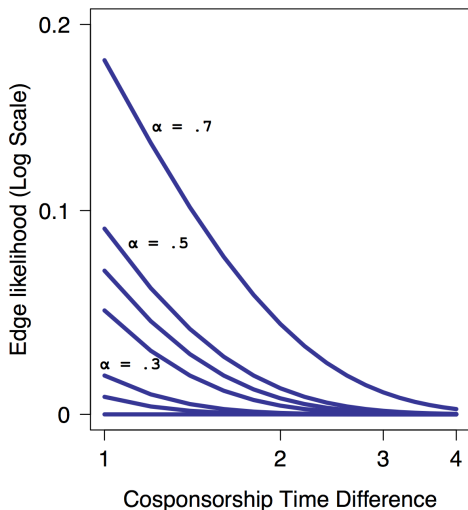


2003-2004



Large Variation in Optimal Network Size

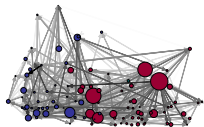
Session	Edges	Alpha
81-82	629	0.5
83-84	575	0.25
85-86	608	0.4
87-88	884	0.45
89-90	450	0.3
91-92	732	0.25
93-94	949	0.7
95-96	957	0.3
97-98	1207	0.1
99-00	779	0.4
01-02	298	0.25
03-04	493	0.5



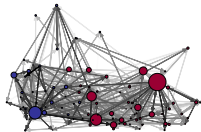
Influence Networks

Influence Networks: 1981-2004

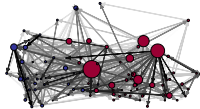
97th Senate



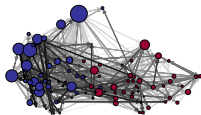
98th Senate



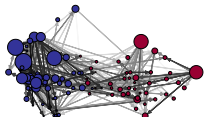
99th Senate



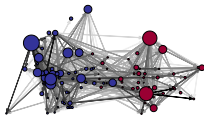
100th Senate



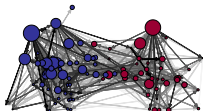
101st Senate



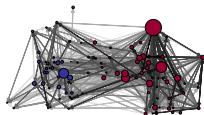
102nd Senate



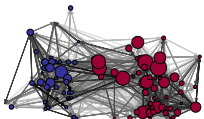
103rd Senate



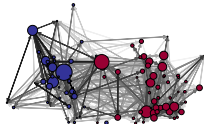
104th Senate



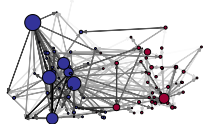
105th Senate



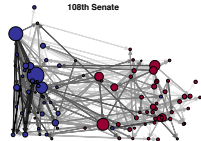
106th Senate



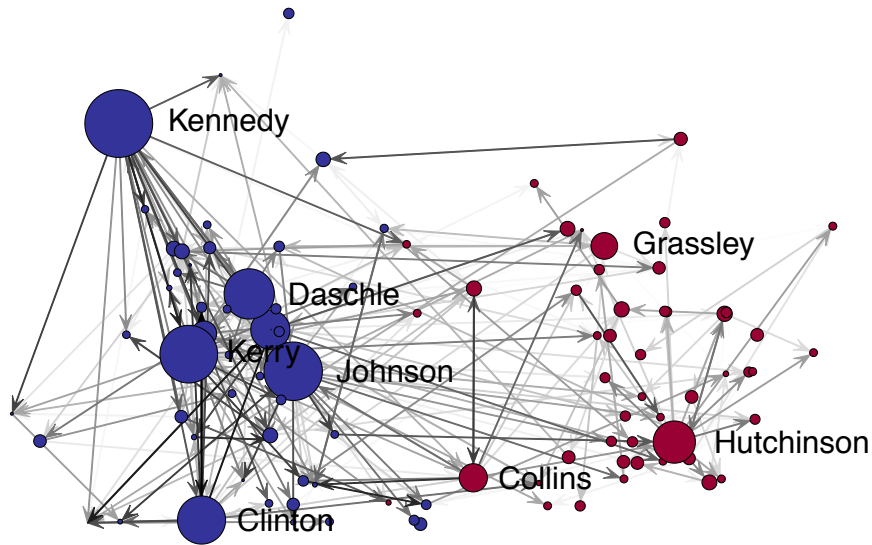
107th Senate



108th Senate

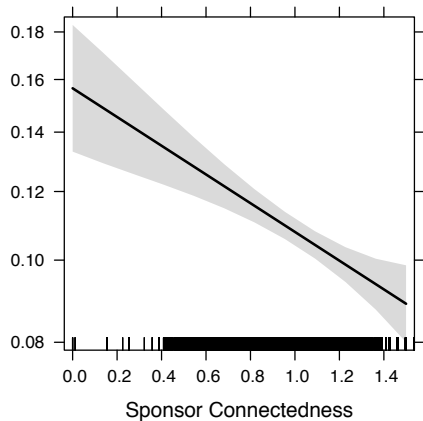
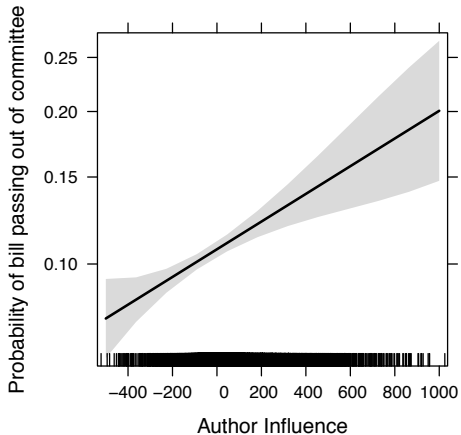


Example Influence Network 2001-2002

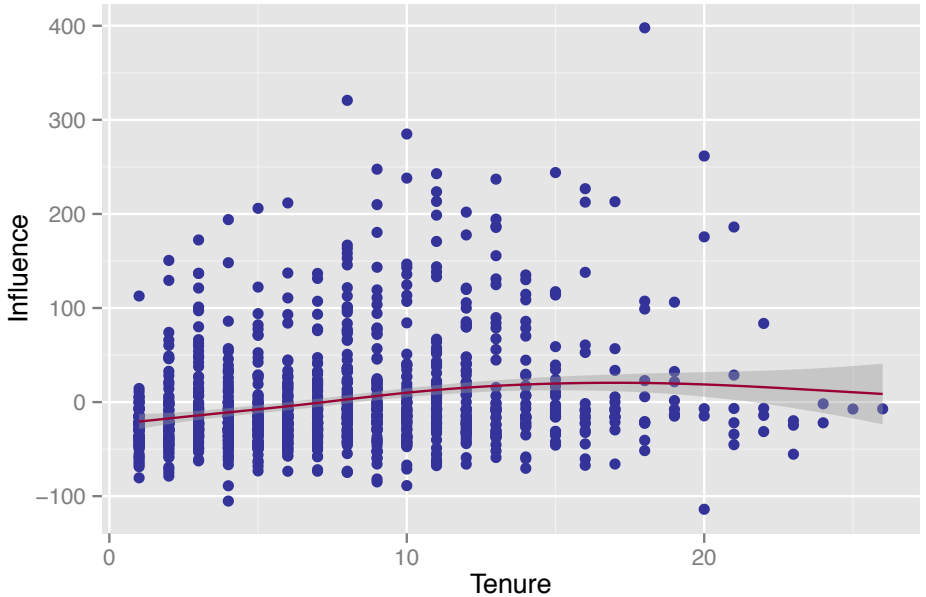


Applications

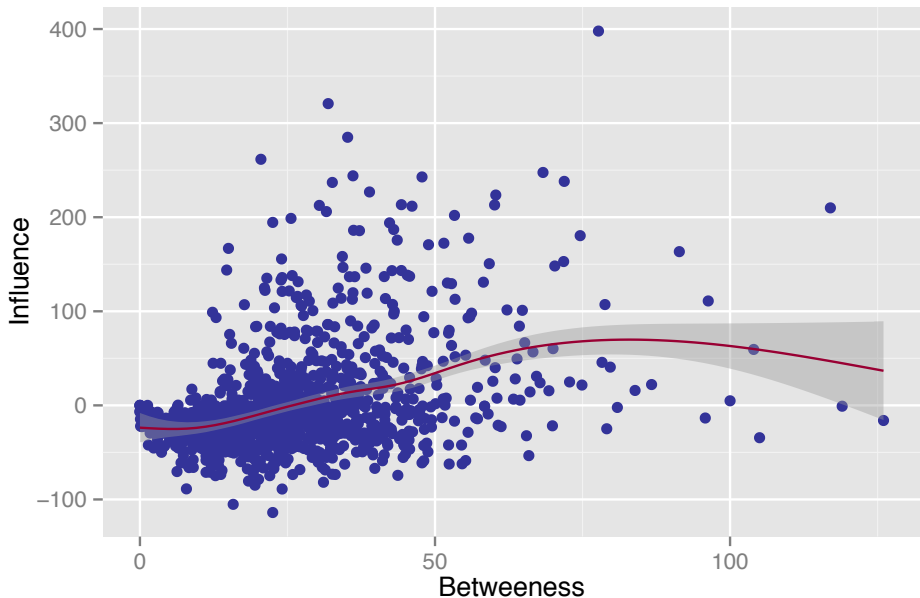
Predicting Bill Advance Using Author Influence



Tenure (Sessions of Congress) and Influence



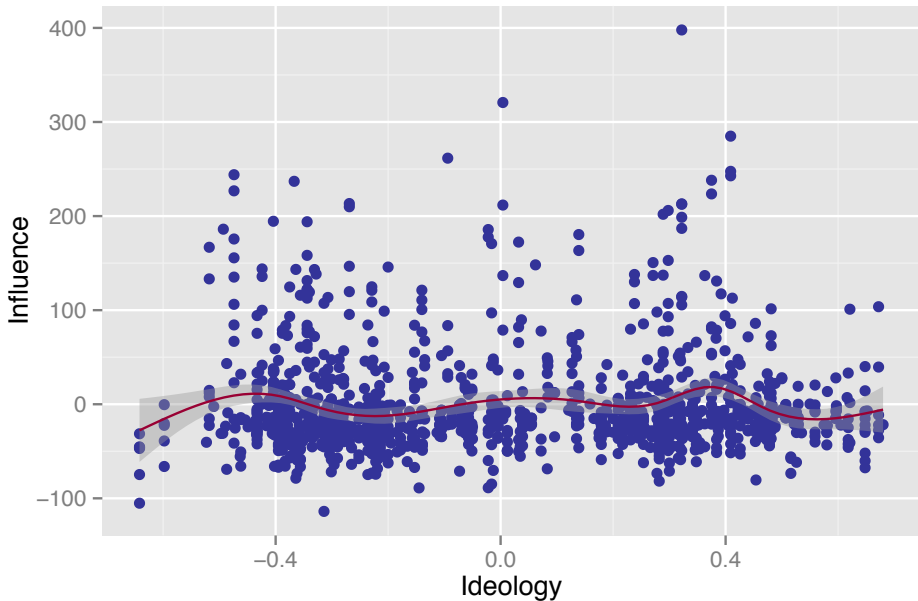
Co-Cosponsorship Betweenness and Influence



Research Agenda

- ▶ Provide influence score/network dataset
- ▶ Δ structure of inf. net. over time
- ▶ Temporal disaggregation
- ▶ Contact: mdenny@polsci.umass.edu
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Ideology and Influence



Density of Bill Author Influence 2001-2002

