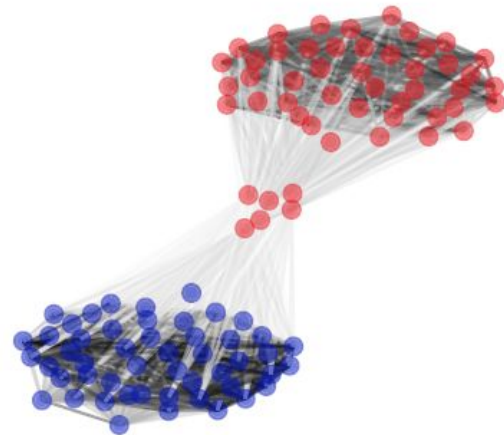


Centrality In Congressional Voting Networks

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Motivation for Project

- In the news, we can see many headlines of Congress and its members being at a standoff.
 - “... Congress is so polarized” (CNN)
 - “Polarization defines the midterm election” (Fox News)
 - “Measuring America’s Divide: ‘It’s Gotten Worse’” (NYT)
- There are many ways to quantify this, so we want to study one of them to see if this is really the case.
- One of the ways which would help not only quantify, but also visualize, polarization is **network science**, which is the study and modeling of interactions between objects.
 - A very common network we encounter is a social network, or even a group of friends
- The U.S. Senate can be expressed as a network.



Overview

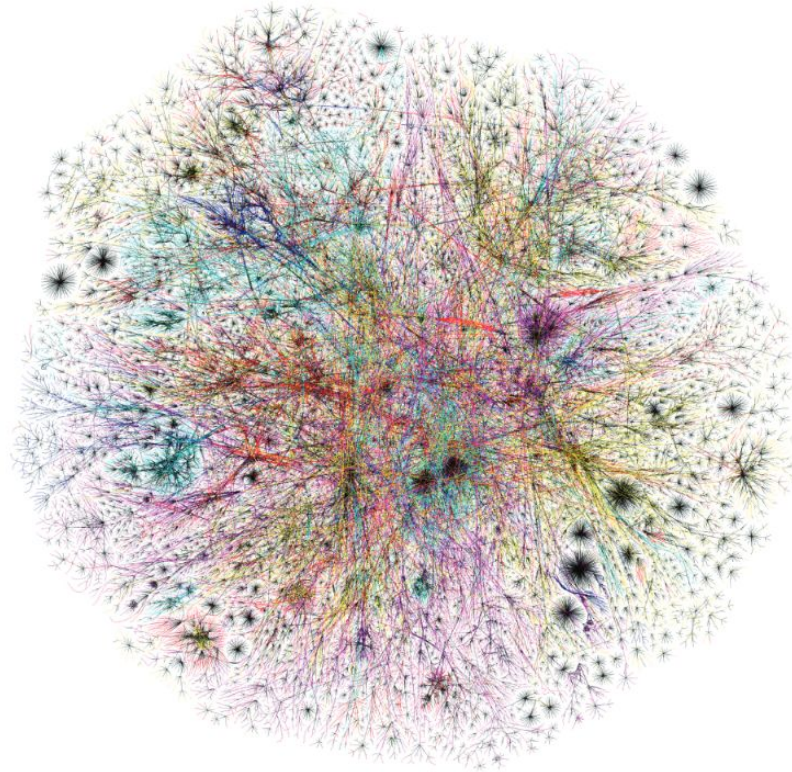
Part 1: Network Science Tools

Part 2: Constructing Congress as a Network

Part 3: Analysis of the Network

Part 4: Conclusions & Future Exploration

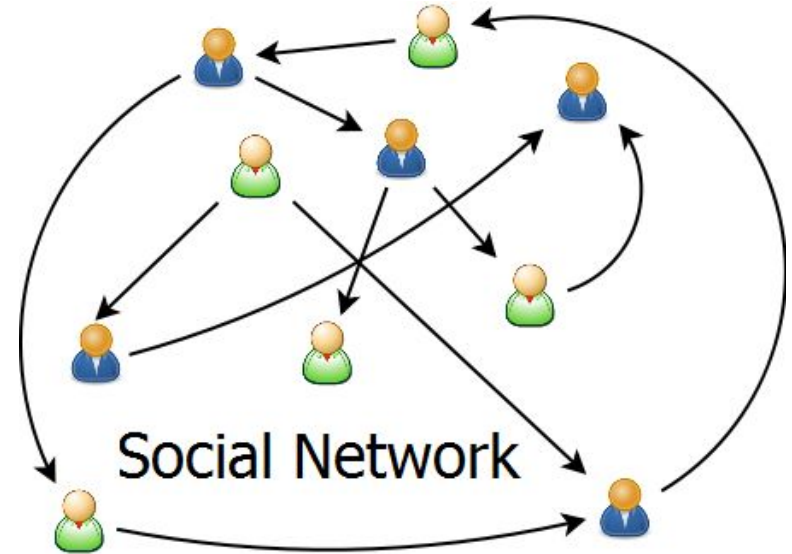
PART 1: Network Science Tools



Newman, Networks 2ed.

Motivations for network science

- Networks were developed as *sociograms* to study human interaction.
- **Network**: A mathematical object made up of **nodes** and **edges**, contained in some latent space.
- **Nodes** represent objects.
 - Examples include people, animals, and websites.
- **Edges** represent some relationship between the nodes.
 - Edges could represent friendship (people), predator/prey (animals), or links between pages (websites).



Josić, *Friendship Paradox*

Fundamental properties of a network

- The structural representation of a network in mathematics is an **adjacency matrix**.

$$A_{ij} = \begin{cases} 1 & (i, j) \in \mathcal{E} \\ 0 & \text{otherwise} \end{cases}$$

- The **degree** is a term used to describe the number of edges the are connected to the node.

$$d_i = \sum_{j=1}^N A_{ij}$$

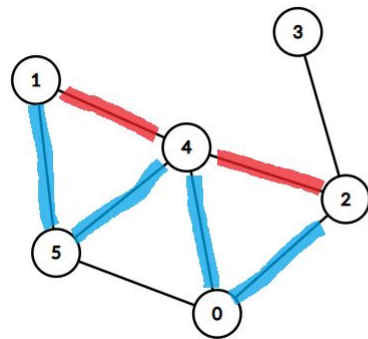
- The **number of edges** in a network is half the total degree of that network.

$$M = \frac{1}{2} \sum_{i=1}^N d_i$$

- The **density** (ρ) is a measure of how connected a graph is, defined as the ratio of the number of edges in the graph to the maximum number of edges possible.
 - High density indicates that the network is well connected.

Paths and connected components

- A **path** is a sequence of nodes where each consecutive pair of unique nodes is connected by an edge.
- A **shortest path** between two nodes is a path with the least number of nodes.
- A **connected component** is a set of nodes that are connected to at least one other node by an edge.



Path from 1 to 2

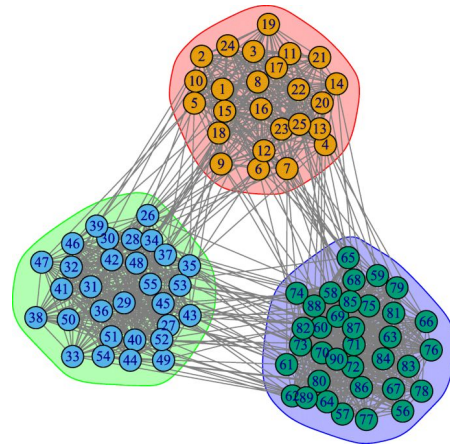
Shortest path from 1 to 2

Community structure in networks

- **Community structure** is a division of set of nodes in a network into smaller subgroups or communities.
 - These nodes are more densely related to each other.
 - Important because they show the functions of the network.
- **Modularity** shows us the extent to which a network can be divided into distinct, relatively independent modules or clusters.
 - Modules are defined as groups of nodes that have a higher density of connections among themselves.

$$Q = \frac{1}{4M} \text{Tr} \left(S^T \left[A - \frac{\mathbf{k}\mathbf{k}^T}{2M} \right] S \right)$$

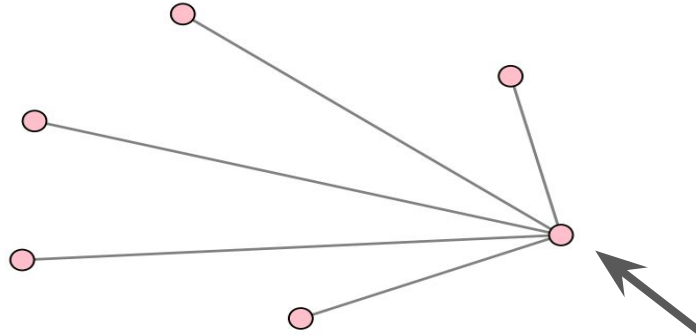
- We use **modularity maximization** to get a clear cut partition in the network.
 - Using statistical inference or iterative methods, we find the permutation of class identification that gives the largest modularity.



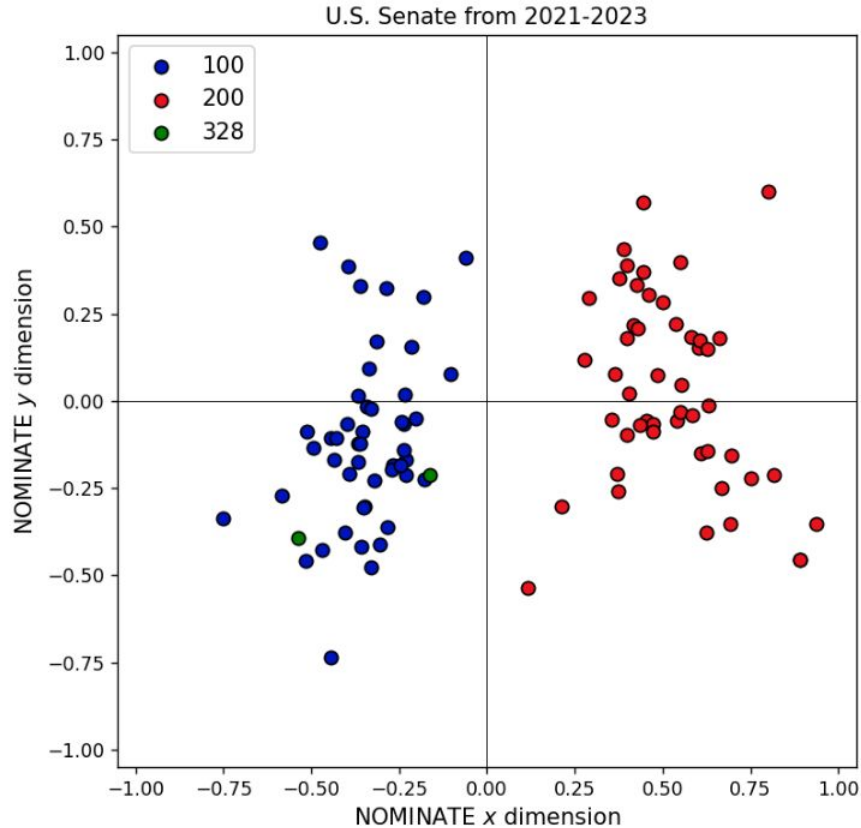
Lee, *A review of stochastic block models and extensions for graph clustering*

Centralities

- **Centrality** in a network determines the most important nodes, edges, or subgraphs.
- There are varying definitions of “important,” so there are many different measures of centrality:
 - **Degree**: a node’s centrality score is its degree.
 - **Eigenvector**: a node’s centrality score is proportional to the centrality of its neighbors.
 - This is the largest eigenvalue of the adjacency matrix and its corresponding eigenvector.
 - **Betweenness**: a node’s centrality score is the number of shortest paths from all nodes that include the node of interest, divided by all possible shortest paths.
 - This is very useful for determining connected components, community structure, etc.

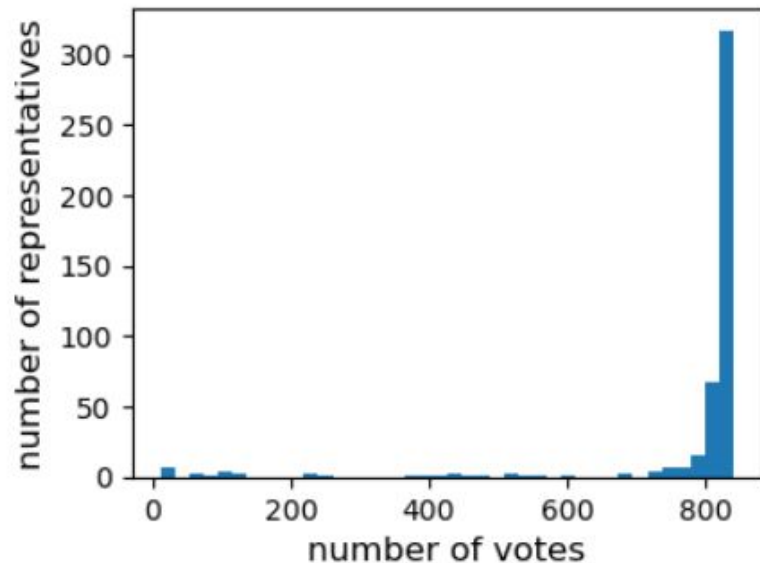


PART 2: Constructing Congress as a Network



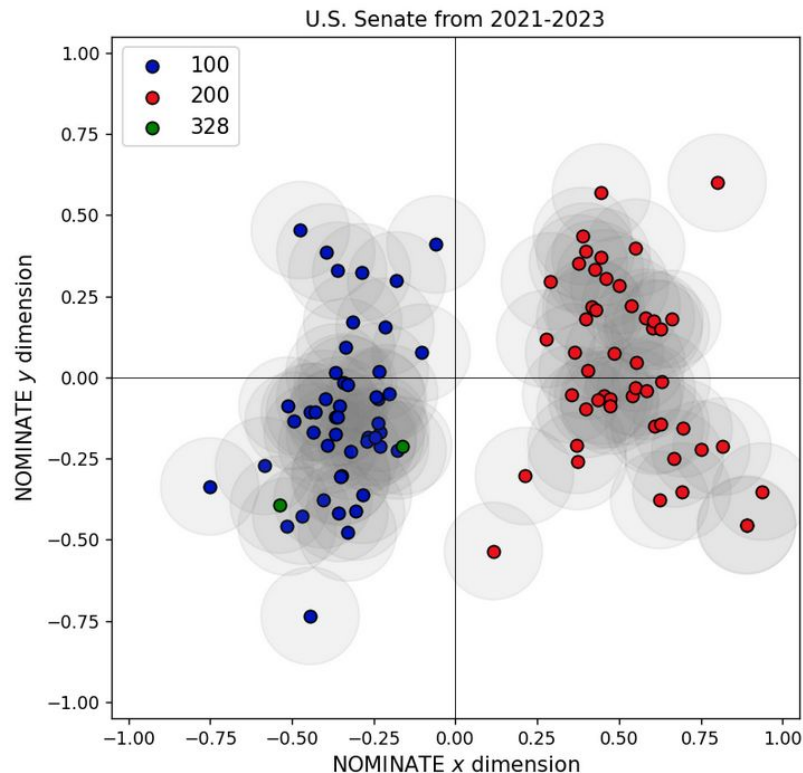
VoteView

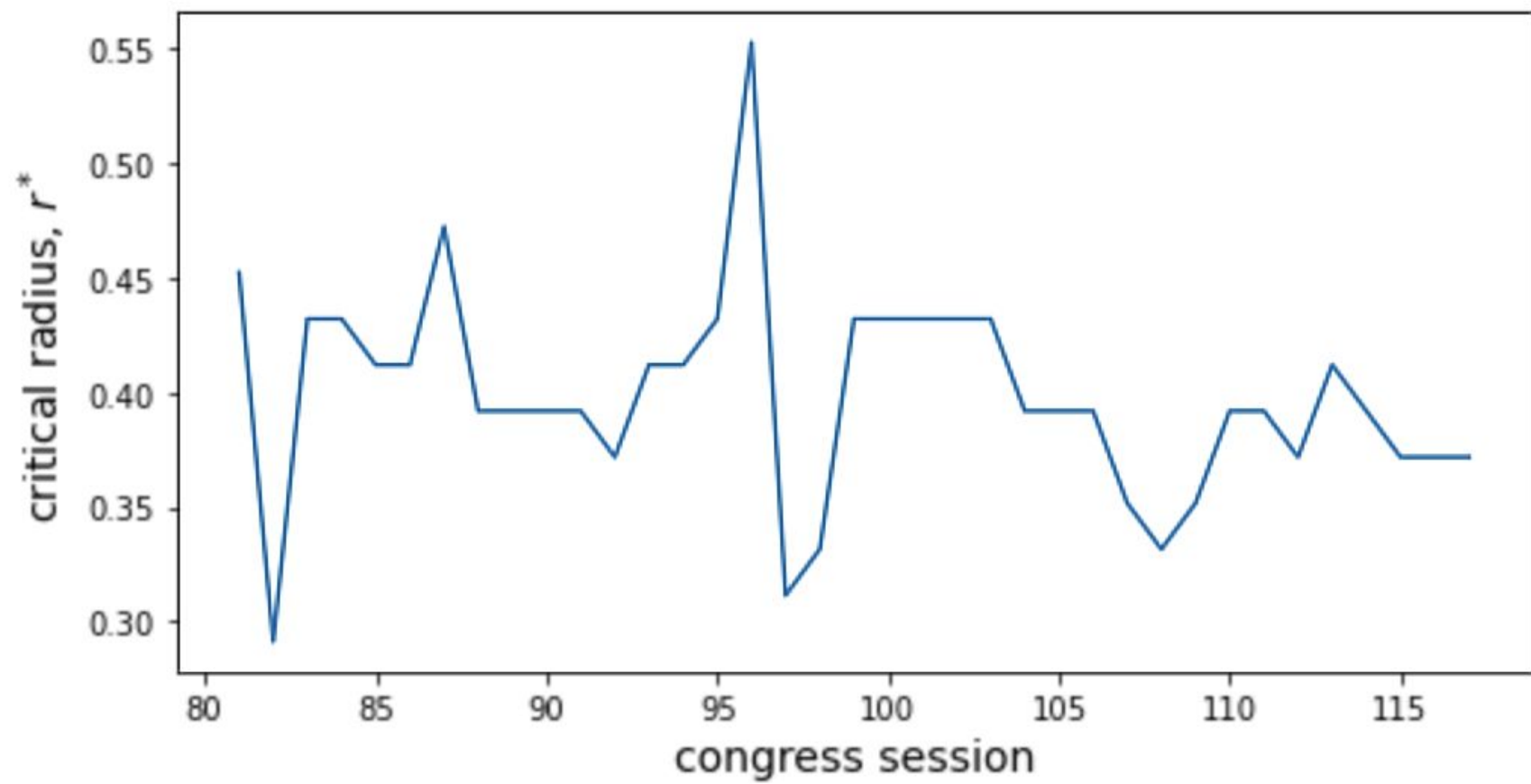
- VoteView is an online database that stores voting records and party ideologies for Congress.
 - We used the two NOMINATE dimensions as the criteria for our project.
 - NOMINATE represents the ideological similarity between congresspeople.
 - Both are represented as scores from a range of -1 to 1, which is an embedding.
- To make the data more reliable, we removed the Vice President and the senators that voted consistently less than the other senators.
 - We assumed that the data was normally distributed and removed statistically insignificant senators.



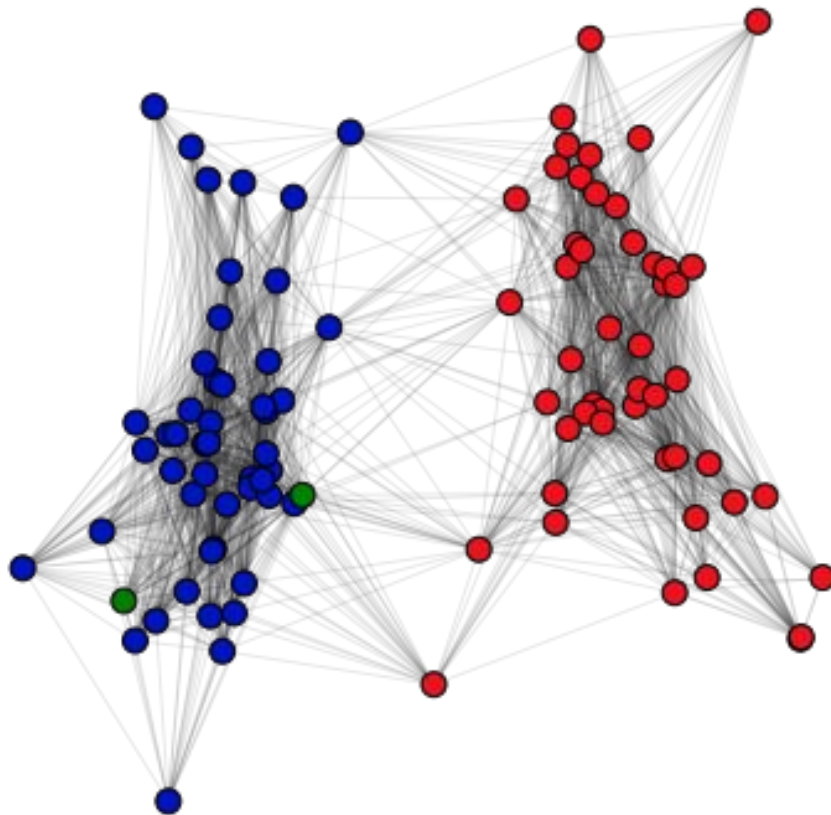
Forming the Network

- We considered Senates from the 81st onwards because this is the start of the most modern political landscape.
- Using Euclidean distance, we computed each senator's ε -ball.
 - Using modularity maximization under the constraint of there being only one connected component, we find ε .
 - We took the maximum ε of all the values for each session to be the ε to use for every session.
- To form the adjacency matrix, two nodes (NOMINATE points) are connected if they lie in each other's ε -ball.



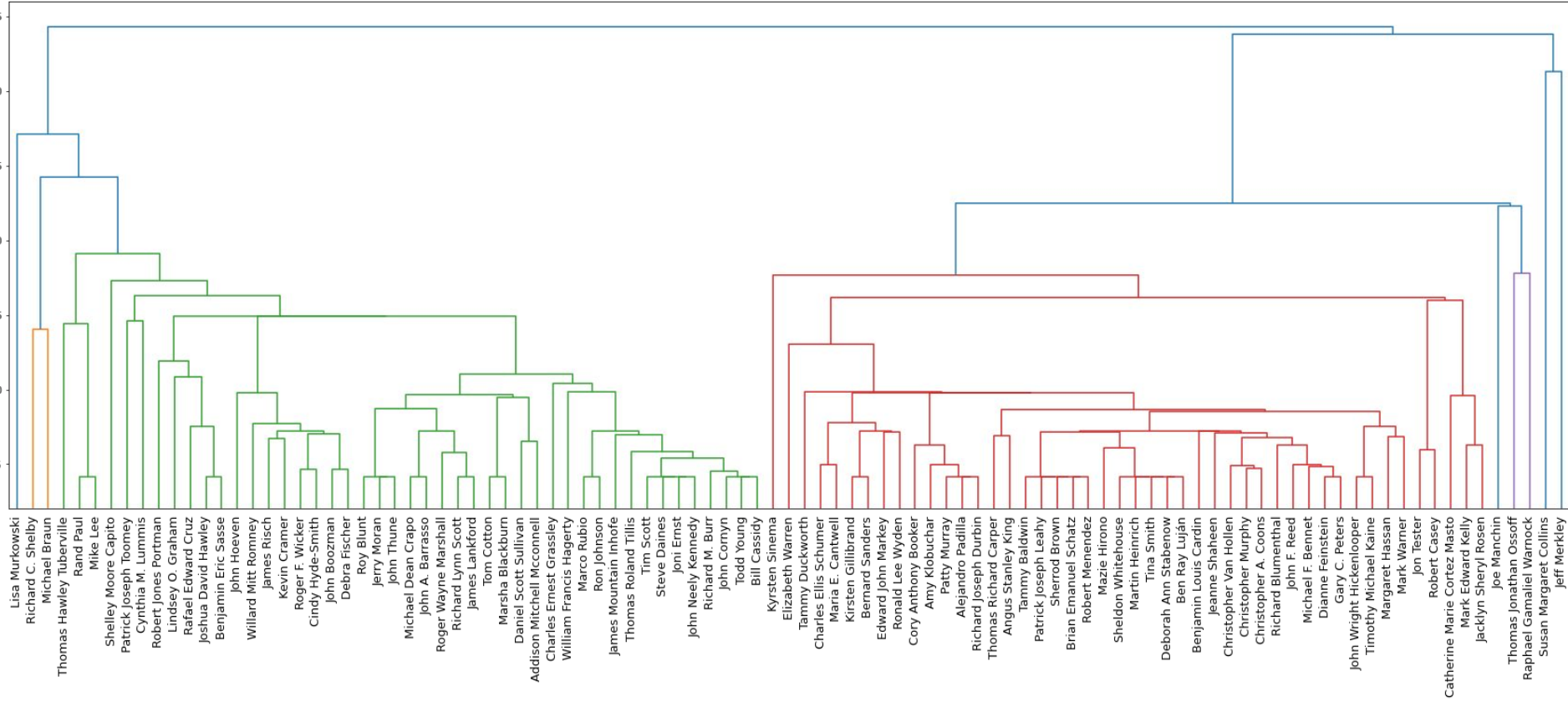


PART 3: Analysis of the Network



Dendrogram Algorithm

1. Begin by placing every senator at height 0.
 2. Compute inverse edge betweenness of the different senators.
 - a. Edge betweenness is like a similarity measure.
 3. Find the closest senators by inverse edge betweenness.
 - a. There can be multiple groups if (x, y, z) have distance d and (u, v, w) have distance d , but the distance between x and y is greater than d .
 4. Connect the closest senators at their inverse betweenness value.
 5. Combine the groups of senator into one pseudo-senator.
 6. Take the new distance between this pseudo-senator as the minimum distance between the senators in this group and the other senators.
 7. Repeat 2-6 until all senators are in one cluster.
- This is a “dendrogram;” at whatever height you cut the dendrogram forms the communities.



Node betweenness

- We transformed the data linearly for more interpretability.
 - Betweenness is already within $[0, 1]$, but we reversed order and shifted it so that 0 is most between and 1 is least between.

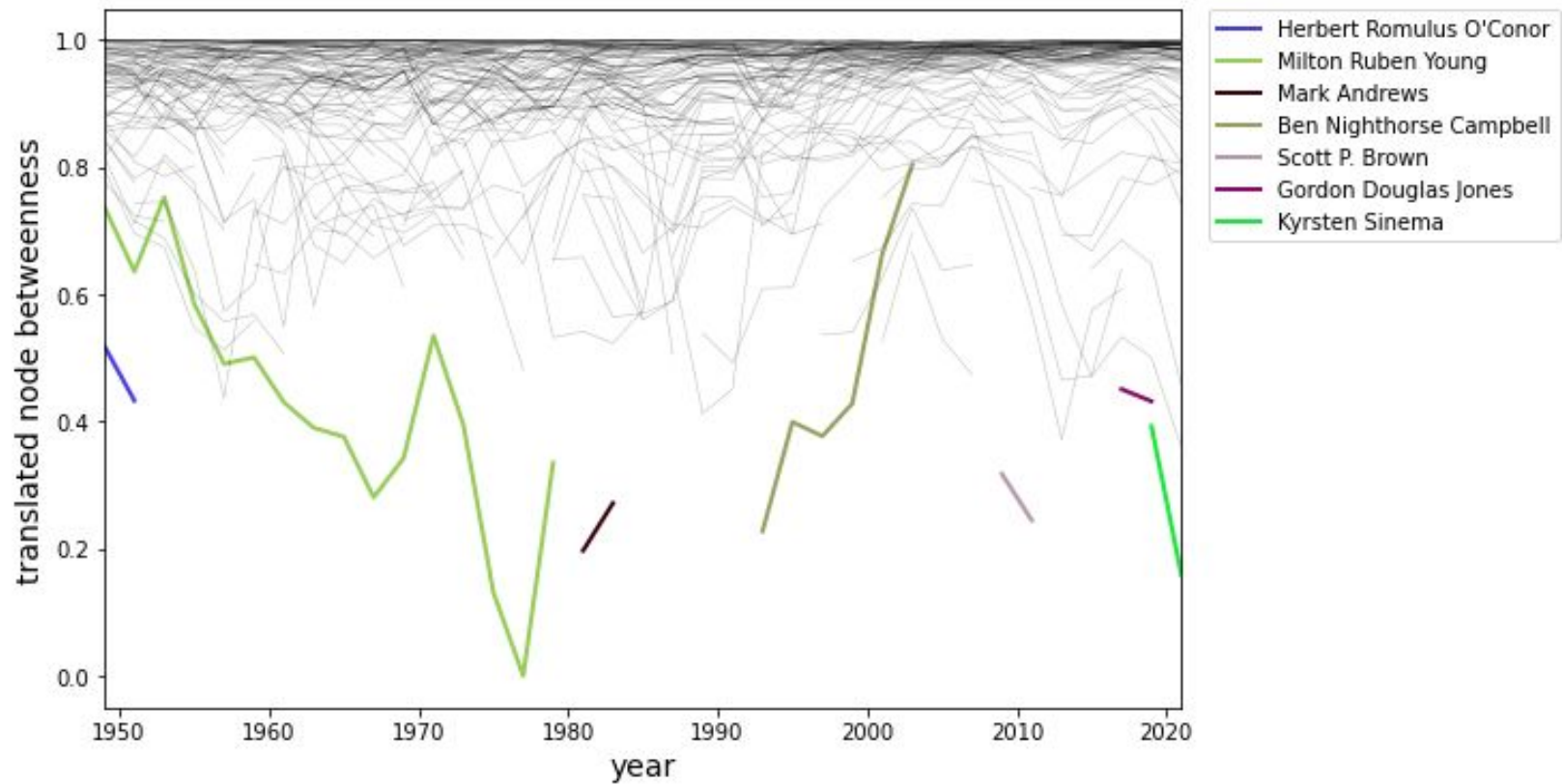
$$\vec{y} = 1 - \frac{\vec{x} - \min(\vec{x})}{\max(\vec{x}) - \min(\vec{x})}$$

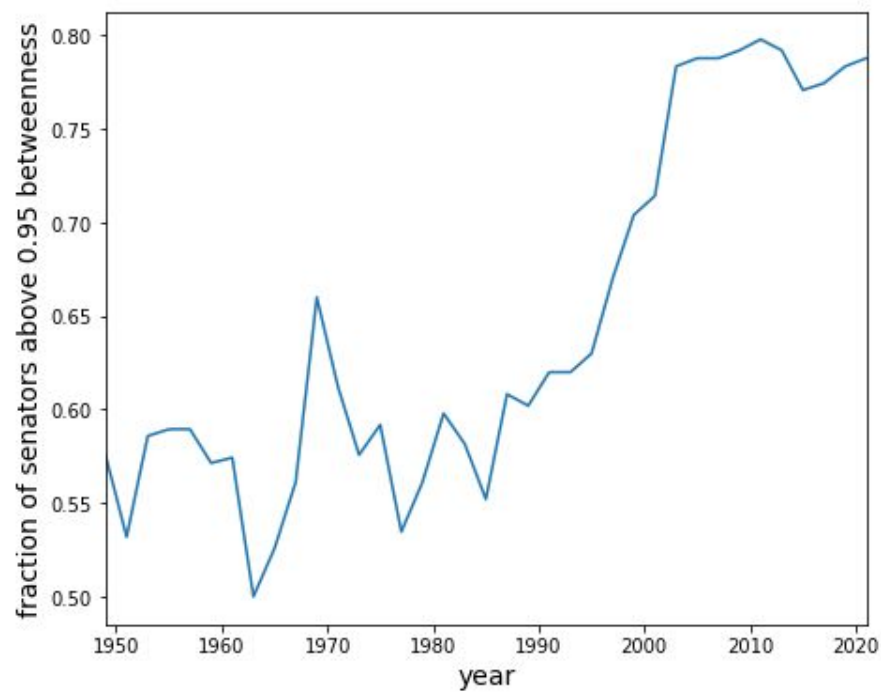
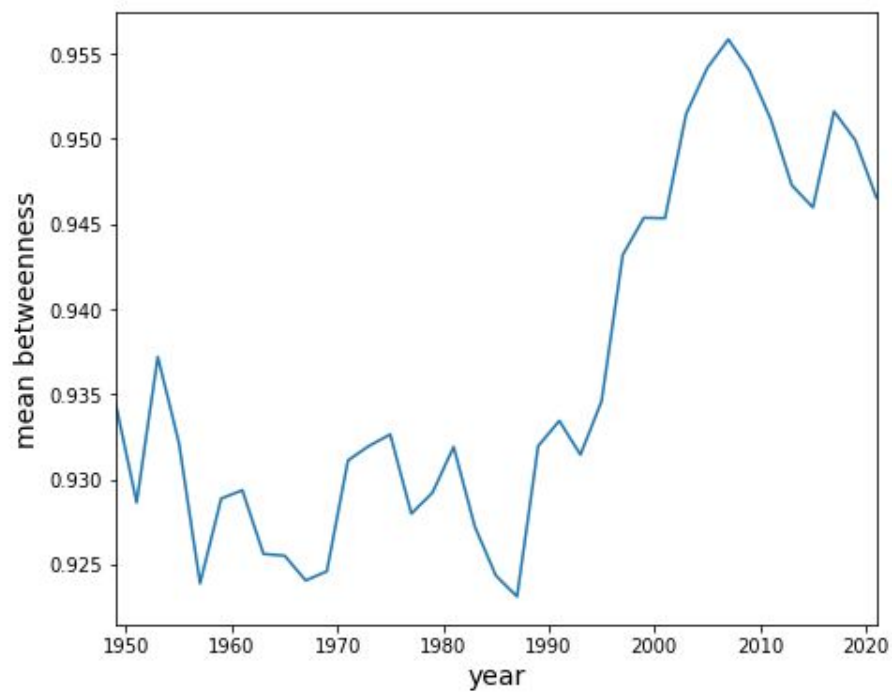
Three most between senators

- Kyrsten Sinema (D, AZ): 0.00
- Lisa Murkowski (R, AK): 0.24
- Shelley Moore Capito (R, WV): 0.35

Three least between senators

- Thomas Jonathan Ossoff (D, GA): 1.00
- Jeff Merkley (D, OR): 1.00
- Michael Braun (R, IN): 1.00





PART 4: Conclusion & Future Explorations

Conclusion

- Congress is a polarized body.
 - Network science is one such tool which can represent that.
- Networks are representations of graphs, which encode the relationships (edges) between nodes (objects).
 - Centralities can be used to measure the importance of nodes or edges.
- Some network centralities are typically associated with community structure and hence can be used as a proxy for polarization.
 - Node and edge betweenness are often used in parallel with community detection.
- We used the NOMINATE dimensions from the VoteView database to embed senators.
 - The NOMINATE dimensions convert high-dimensional voting patterns into 2D coordinates.
- We examined node betweenness for different sessions of the Senate.
 - We found that the average value is increasing over time.
 - We found that the fraction of people with high betweenness increases over time.

Future aspects

- We can repeat these analyses for the House of Representatives.
 - There are 400+ representatives, so the data is more rounded.
 - Polarization should be occurring in both chambers.
- Creating a different embedding to represent the network.
 - PCA, diffusion maps, eigenmaps for different choices of data (like voting patterns, cosponsorship, etc.)
- Using different types of centralities for a different analysis.
 - PageRank, eigenvector, degree, closeness, etc.
- Implementing machine learning on the data to predict important metrics.
 - Modularity, party classification, etc.

Questions?