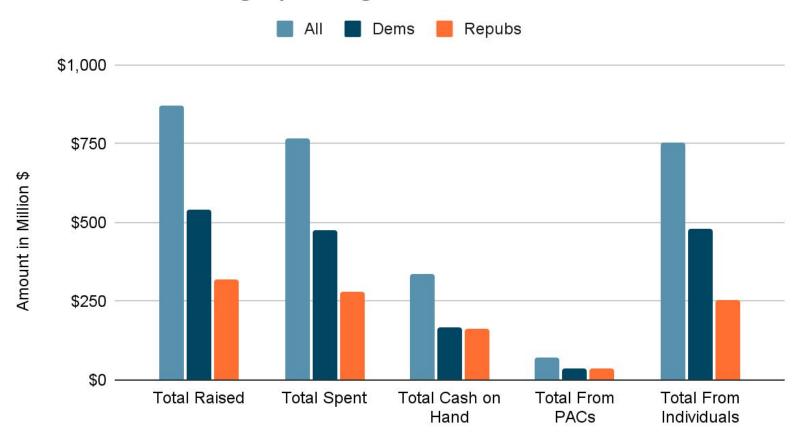
S117 US Senators Knowledge Graph

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Senator Fundraising/Spending 2022



Introduction

The purpose of our knowledge graph is to allow users to map financial contributors and vote history of senators in the 117th Senate with their publicly available statements via Twitter.

The graph uses publically available vote data, financial contribution data, and senator statements on twitter to allow a user to investigate how senators vote compared to public statements and their financial contributors. By using this graph it is possible to determine what entities a senator tweets about and how that corresponds to their voting record. Or it can be seen how much money individuals donated to campaigns for senators that voted specific ways on votes that mention topics.



Data Sources

Voteview.com





- csv files with Senate session 117 information
 - Senator biographical data ICPSR ID, State, Name
 - Senator party data Name, Party
 - Senate rollcall votes Roll Number, Date, Yay Count, Nay Count, Description
 - Senator vote record ICPSR, Roll Number, Cast
- Provides unique Senator ICPSR ID to identify unique individuals

OpenSecrets.org

- csv files with contribution amounts for the latest election that a senator was elected in by states
 - Name, Contributor, Individual Contributions, PAC Contributions, Total Contributions
- Contributors listed are the sum total of contributions from individuals that are associated with an organization
- Provides both corporate and political action committee contributions





Twitter

- csv listing from Kaggle.com of all current senator twitter handles to link individuals with tweets
 - Name, Handle
- json file of tweets from the list of twitter handles, about 10K tweets total
 - o handle:
 - created_at:
 - text:
 - public_metrics:
 - retweet count:
 - reply_count:
 - like_count
 - Quote_count
- Twitter API tutorial found at https://github.com/twitterdev/getting-started-with-the-twitter-api-v2-for-academic-research
- Limited in the tweet timeframe for free API, up to 3200 most recent tweets from user's timeline

Preparing Senators & Contributors

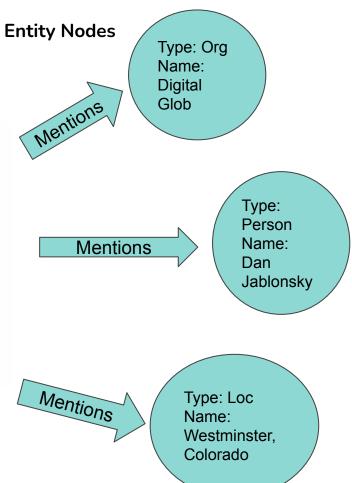
- 1) Extract a listing of all senators with a unique ICPSR ID from voteview biographical data
- 2) Generate a distinct set of nodes for senators with their ID, name, and party affiliation
- 3) Entity match firstname and lastname values from the master senators list with contribution csv for each senate race and confirm against their state
- 4) Find all unique contributors using names across all senate race files using entity matching
- 5) Create an edge linking each unique contributor to each unique senator with the edge titled DONATED_TO
 - a) Store the donation amount, number of individual from organization who donated, number of pacs from the donor, and race that was donated to as edge attributes

Preparing the Tweets with Entities

- 1) Match Senators to Twitter Accounts
 - a) Match on first name and last name, manually curate the results
- 2) Pull Tweets:
 - a) Use the twitter API to pull recent tweets on the senator's user timeline, max 100 tweets per senator twitterID, store as a JSON file
 - b) From JSON file, save the tweet itself and public metrics (likes, replies, retweets, quotes)
- 3) Tweet NLP:
 - a) Remove tweets with less than 5 words, remove '\n', and remove links with 'https://'
 - b) Conduct a NER on each tweet using Spacy, mapping each tweet to an entity type and entity name
- 4) Load the tweets in to Neo4j as individual nodes, with tweet text and public metrics as node attributes
- 5) Load the name entities as individual nodes, with the entity type and the entity name as node attributes
- 6) MATCH the tweet nodes to their corresponding entity node if the tweet mentions that named entity creating the edge MENTIONS



Tweet Node



Preparing Roll Calls with Entities

- 1) Process the S117 senate roll call information from voteview.com (S117_rollcalls.csv)
 - a) Save the bill number, date of bill, yea count, nay count, margin (yea nay), result, and bill description
- 2) Load in as individual node type to Neo4j with above attributes
- 3) Roll call NLP:
 - a) For the the bill descriptions, conduct a NER analysis to map descriptions to named entities
- 4) Load in the named entities from the roll call NER results as additional named entity nodes (with entity type and name as node attributes), merging with the ones from the twitter NER results
- 5) MATCH the roll call nodes to the corresponding named entity for which the bill description mentions, linking with the edge MENTIONS

Example Data from Roll Calls

		_desc	sult vo	vote_re	nay_count	yea_count	date	rollnumber
		tained	on Not S	Objecti	93	6	1/6/21	1
		tained	on Not S	Objecti	92	7	1/7/21	2
, of New Y	Haines, c	Danica	tion Av	Nomina	10	84	1/20/21	3
A bill to provide for an except			sed A	Bill Pas	27	69	1/21/21	4
Lloyd James Austin, of Georg			tion Llo	Nomina	2	93	1/22/21	5
Janet Louise Yellen, of Califor			tion Ja	Nomina	15	84	1/25/21	6
n, of New	Blinken,	ny John	tion Ar	Nomina	22	78	1/26/21	7
)	greed to	to Table	Motion	45	55	1/26/21	8
ide for rel	to provid	solution	ion A A	Resolu	17	83	1/26/21	9
Mayorkas	cholas M	andro Ni	Mot Al	Cloture	42	55	1/28/21	10
nery Butti	lontgome	r Paul M	tion Pe	Nomina	13	86	2/2/21	11
Mayorkas	cholas M	andro Ni	tion Al	Nomina	43	56	2/2/21	12
tion settir	resolutio	ncurrent	to Pı A	Motion	49	50	2/2/21	13
it-neutral	a deficit-	stablish	nent To	Amend	10	90	2/4/21	14
it-neutral	a deficit-	stablish	nent To	Amend	50	50	2/4/21	15
To establish a deficit-neutral			nent To	Amend	0	100	2/4/21	16
it-neutral	a deficit-	stablish	nent To	Amend	50	50	2/4/21	17
it-neutral	a deficit-	stablish	nent To	Amend	42	58	2/4/21	18
it-neutral	a deficit-	stablish	nent To	Amend	1	99	2/4/21	19

NER on vote_desc column

Entity Nodes

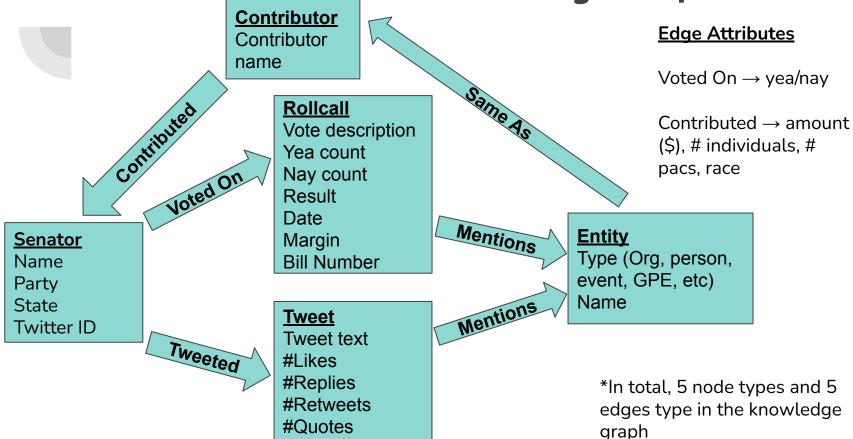
Linking Named Entities To Contributors

- 1) Process contributor names in contributors CSV file
 - a) Create data with removed punctuation, all lowercase, and removing org, inc, or llc
- 2) Merge roll call NER results file with twitter NER results file on the named entity
 - a) Create data with removed punctuation, all lowercase, and removing org, inc, or llc
- 3) Match the processed contributor names with the processed named entities, saving as a file
- 4) Load in the named entity contributor matching results as relationships to the Neo4j knowledge graph
 - a) Create the edge SAME_AS if a named entity is also a contributor

Linking Senators to their Votes and Tweets

- 1) Using S117 senator information from voteview.com (S117_members.csv), save the name, party, and state of each senator
- 2) Merge senator names with the Twitter account mapping
- 3) Load in the senators as their own nodes with name, party, state, and twitter account as node attributes
- 4) Senator Tweet Relationship:
 - a) MATCH each senator node to their tweet nodes creating an edge labeled TWEETED
- 5) Senator Votes on Roll calls:
 - a) Using the S117 voteview.com information (S117_votes.csv), MATCH each senator node to the roll call node they voted on creating an edge labeled VOTED
 - b) Store whether the senator voted yea or nay as an edge attribute

Knowledge Graph Schema



Next Steps

- Get an API key for followthemoney, which has more detailed contributor data such as industry. Or get contributor data manually from somewhere like dbpedia.
- Increase date range of tweets to provide a larger variety of public statements
- Curate NER results further
 - There are overlaps to what may be called an Organization or what may be called a Geopolitical entity, for example
- Add sentiment predictions on tweets to allow for searching entities that are positively mentioned vs negatively mentioned
- Conduct more rigorous matching of named entities to contributors
- Add news articles mentioning bills/senators into the knowledge graph
- Figure out a way to incorporate tweets that mention other users

Demo

We will now give a demo of our senator knowledge graph in Neo4j ...



Thank you!