SAMUEL SHERMAN

You can find me at: github.com/scsherm linkedin.com/in/samuelcsherman

Congressional Bill Modeling

Data Collection

- Congress.gov
 - Bill text



- Sunlight Foundation API
 - **Bill data (JSON)**
 - Votes data (JSON)
- **MongoDB**





- Tools
 - Pandas, pymongo, sklearn, nltk, numpy, scipy, AWS ec2

Motivations

- Percent of yes votes for a given party
 - Polarization in government
- Whether a bill will reach a vote
 - Important features in bills
- Latent topics in bill text
 - Prevalence over time
 - Distinguishable characteristics between presidencies

Natural Language Processing

Stopwords

- Regular words: "and", "the", "be", "it", "there"
- Congressional: "amendment", "bill", "quorum", "act"

Maximum document frequency

TFIDF

- TF: Frequency of words across a single document
- IDF: Frequency of words across all documents

Non-Negative Matrix Factorization

Similarity within topics high

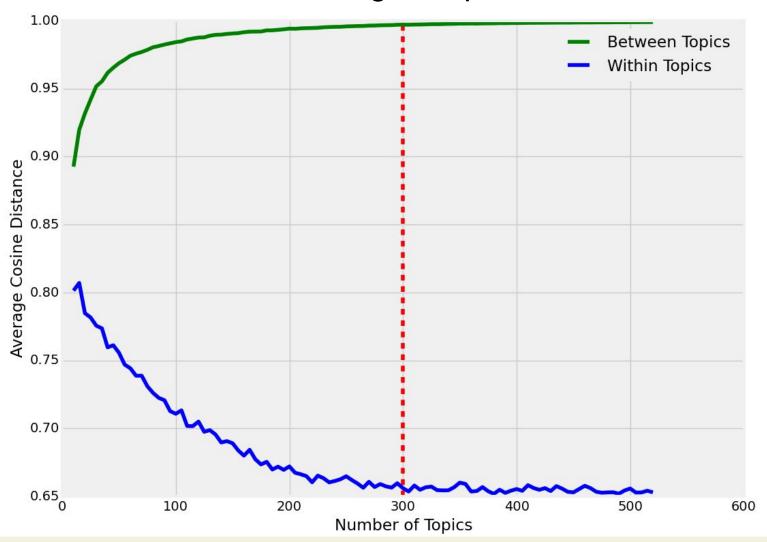


Similarity between topics low

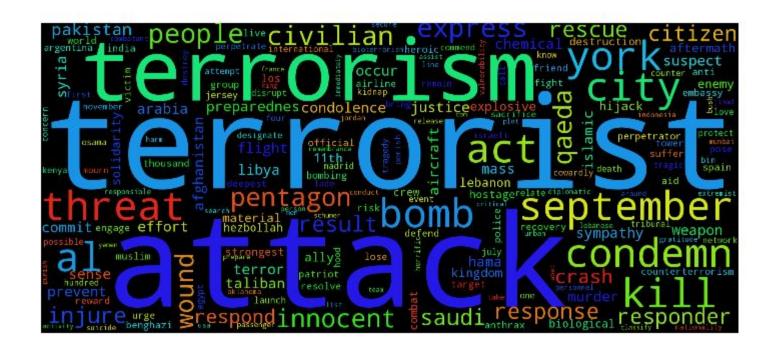




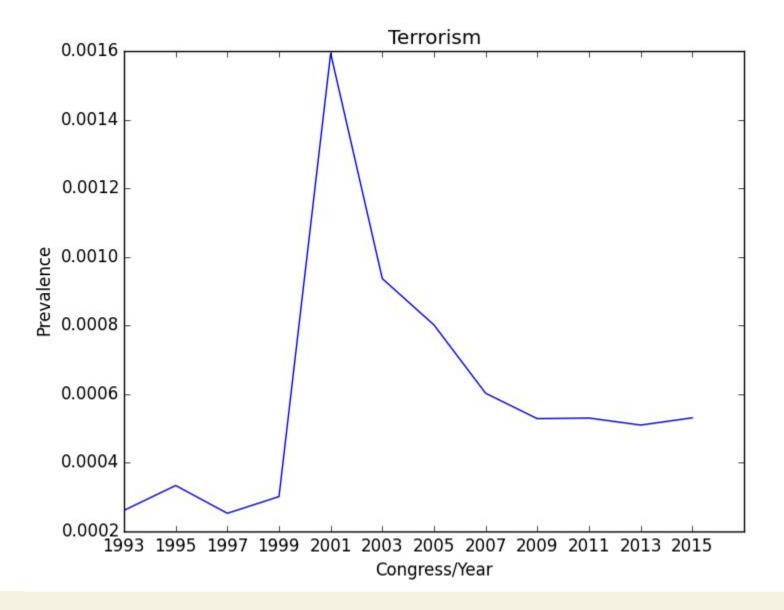
Choosing K Topics



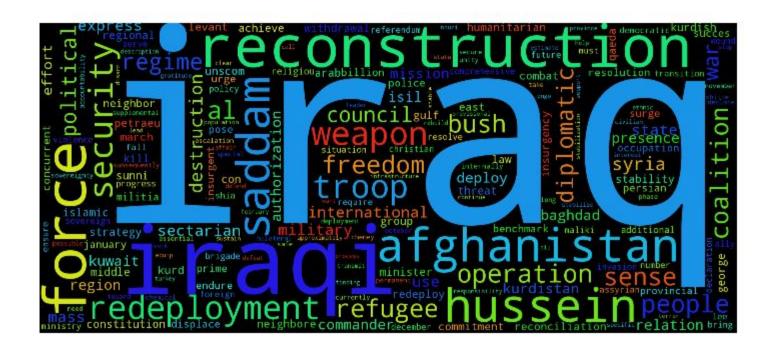
Measuring Topic Distance



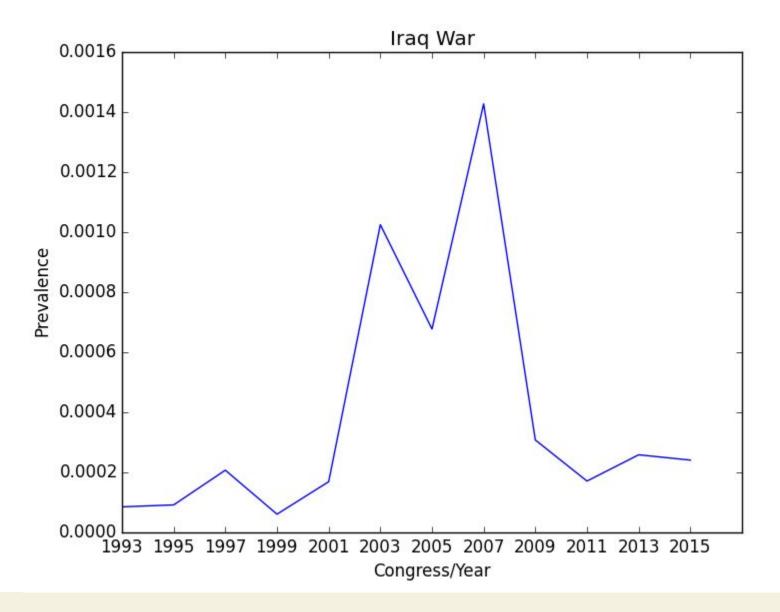
Terrorism



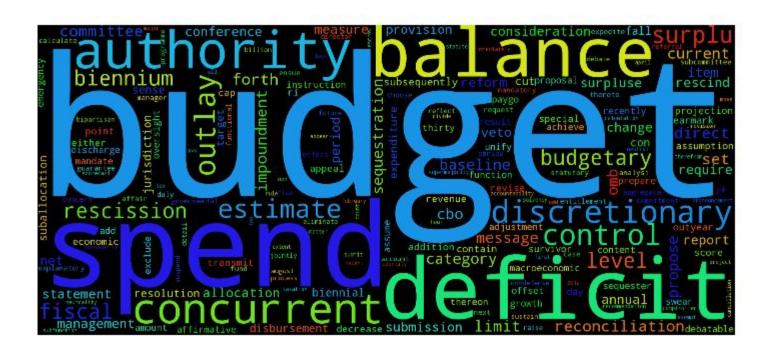
Terrorism in Bills Over Time



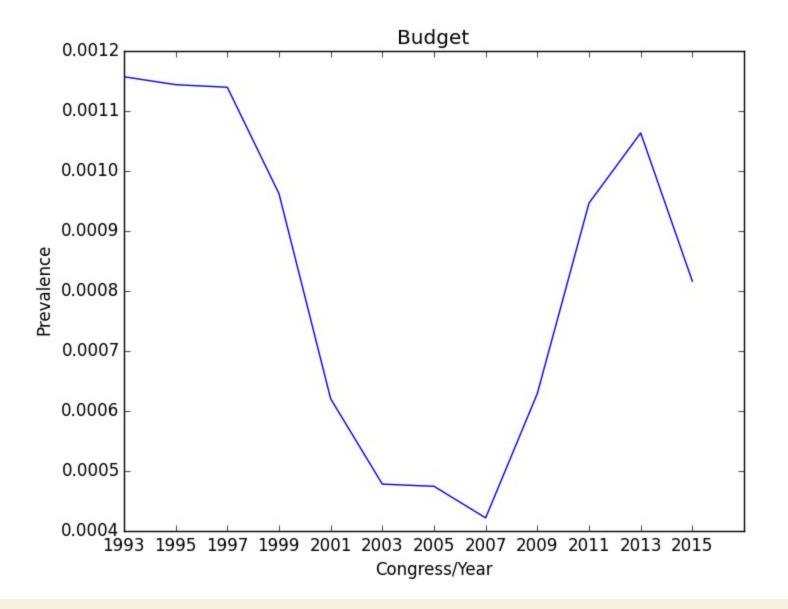
Iraq War



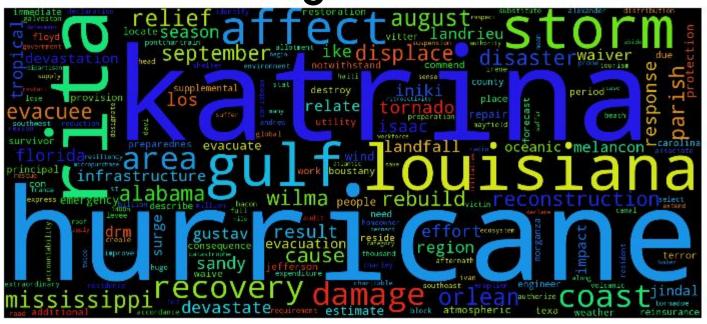
Iraq War in Bills Over Time



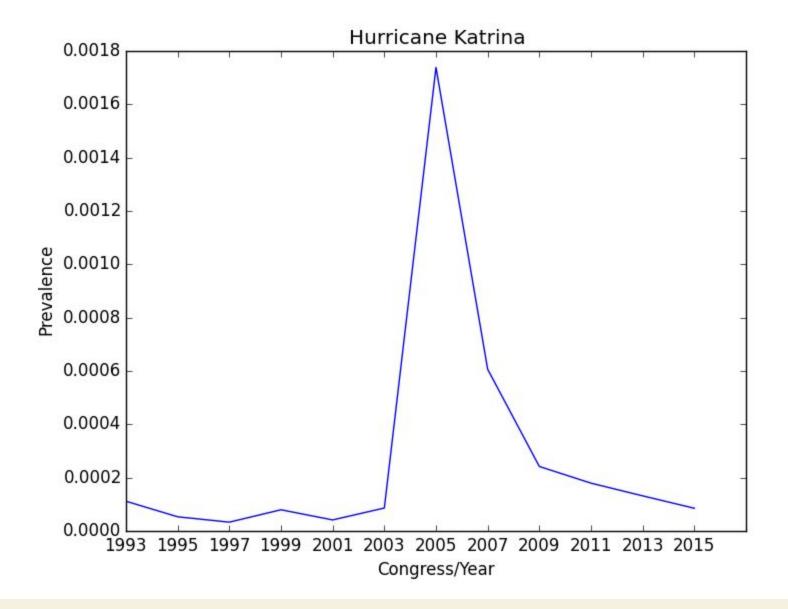
Budget



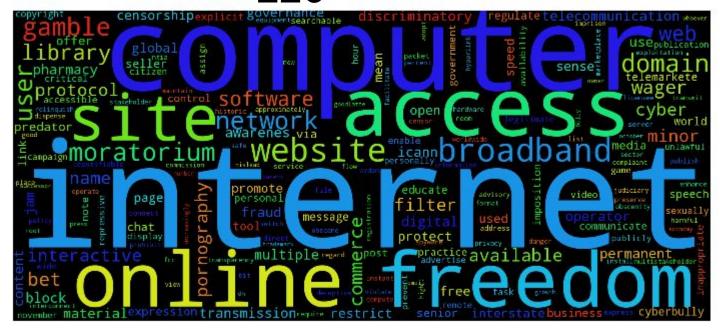
Budget in Bills Over Time



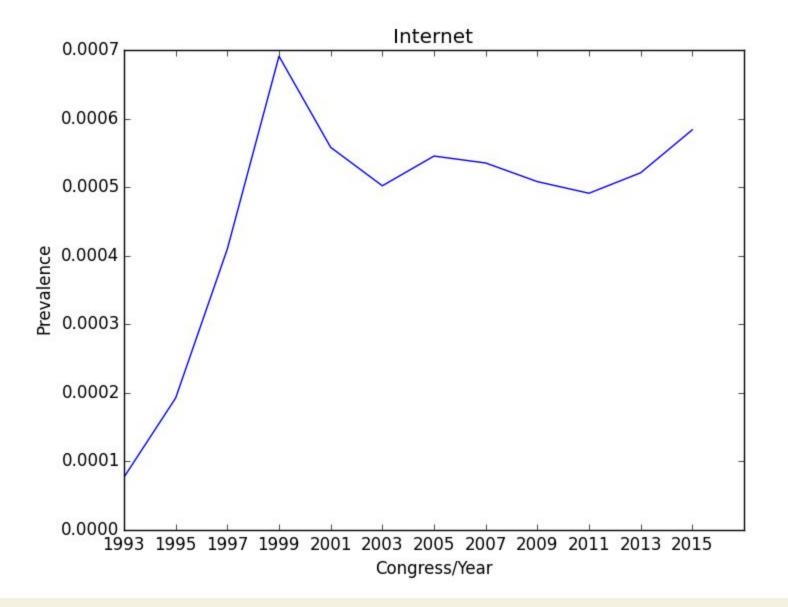
Hurricane Katrina



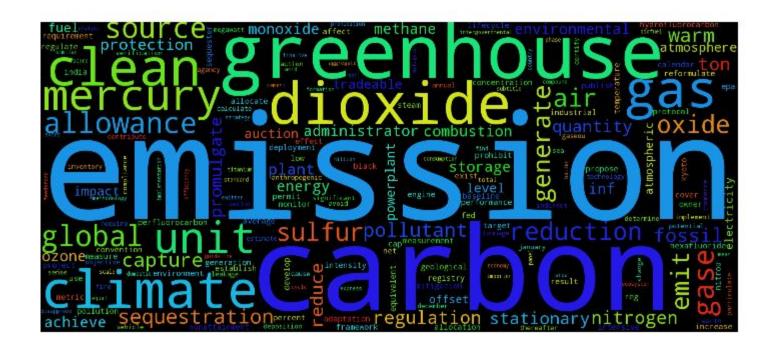
Hurricane Katrina in Bills Over Time



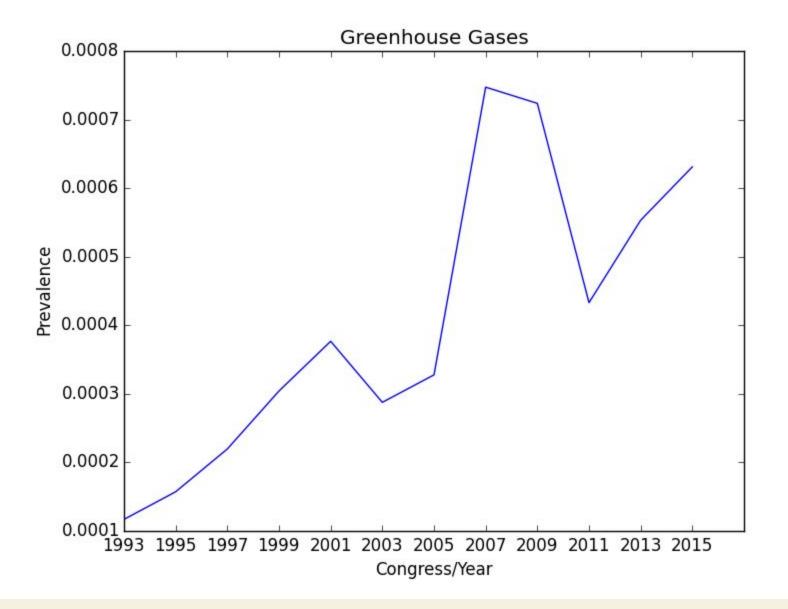
Internet



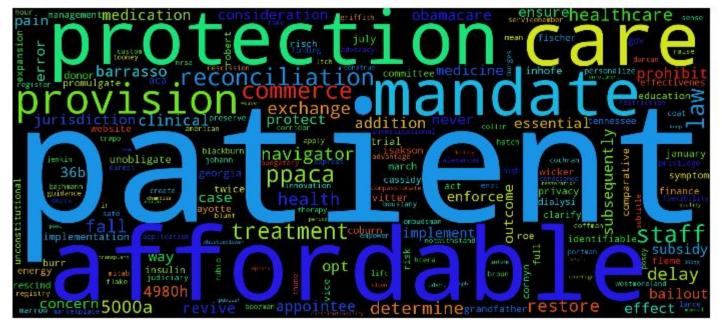
Internet in Bills Over Time



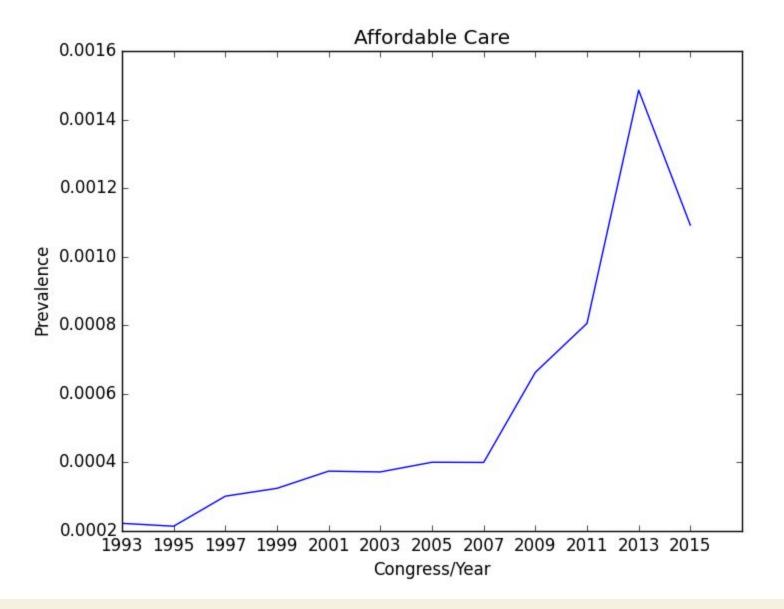
Greenhouse Gases



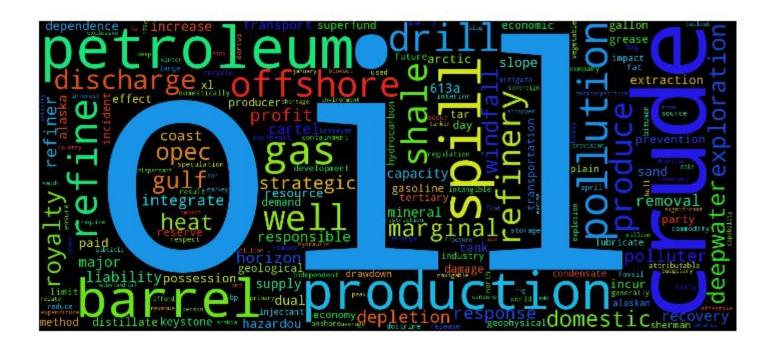
Greenhouse Gases in Bills Over Time

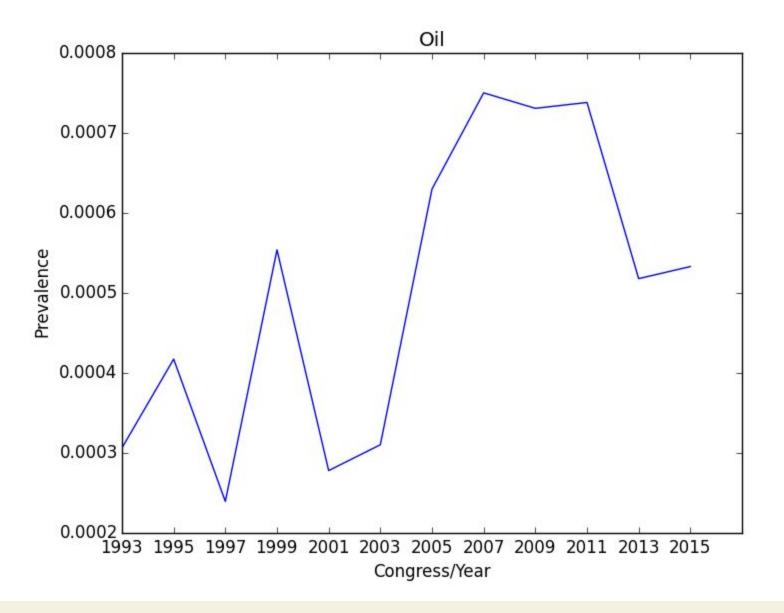


Affordable Care Act

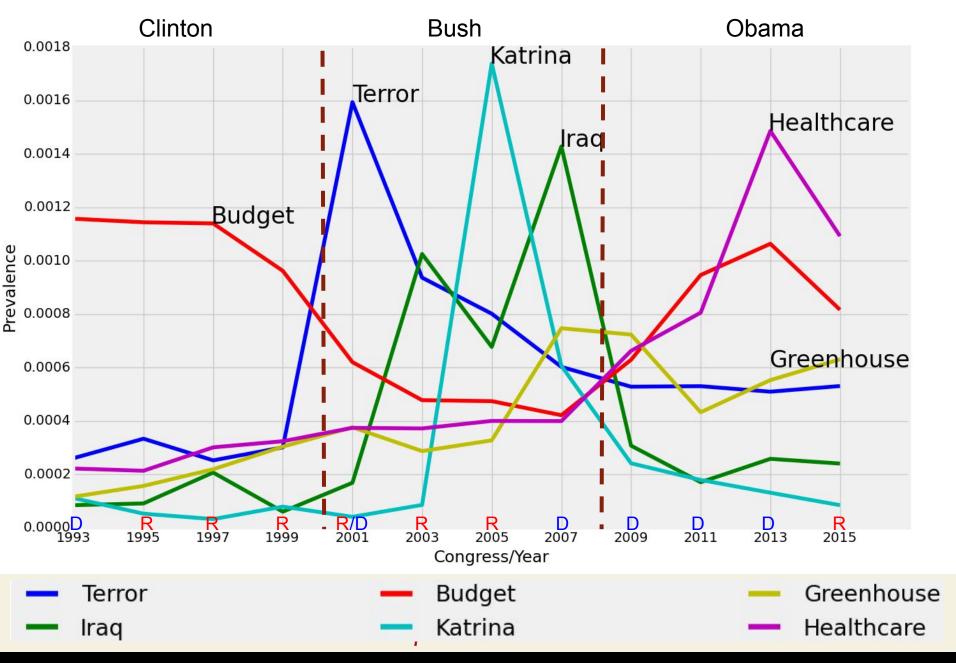


Affordable Care in Bills Over Time





Oil in Bills Over Time



Thanks!

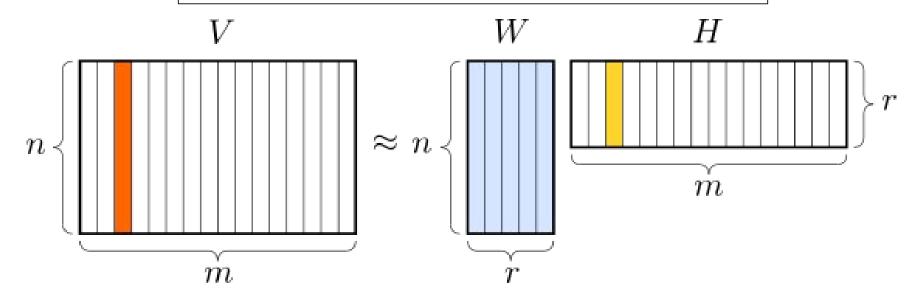
ANY QUESTIONS?

You can find me at:

linkedin.com/in/samuelcsherman github.com/scsherm scsherm@gmail.com

Appendix

Non-Negative Matrix Factorization

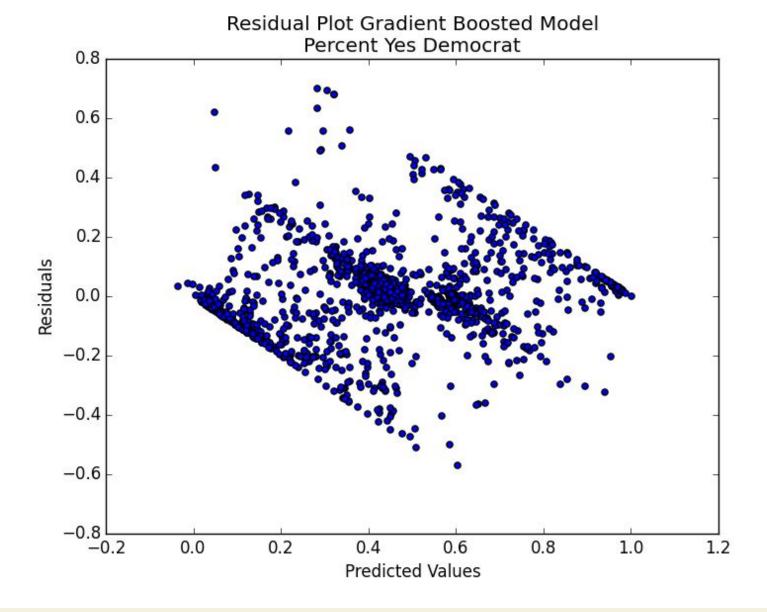


- Recommendation Systems
 - Null space
- Trend modeling/finance/stocks
 - Social Media
- Even Image Processing
 - Pattern/facial recognition

Regression

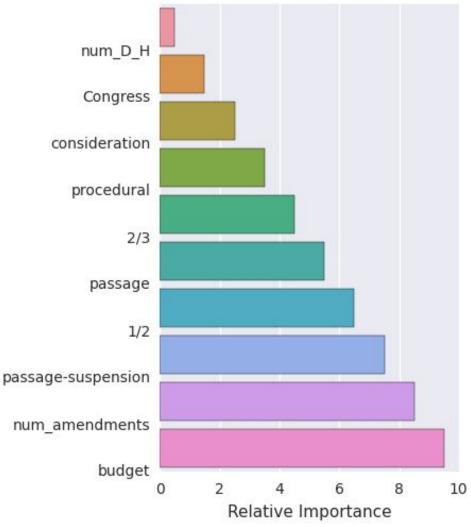
Percent Yes Democrat

	Mean Squared Error	Root Mean Squared Error	R ² score
Random Forest	0.0205	0.143	0.740
Bagging	0.0207	0.144	0.737
Linear	0.0417	0.204	0.417
Gradient Boosted	0.0203	0.143	0.743



Residuals

Gradient Boosted Regression Variable Importance Percent Yes Democrat



Feature Importances

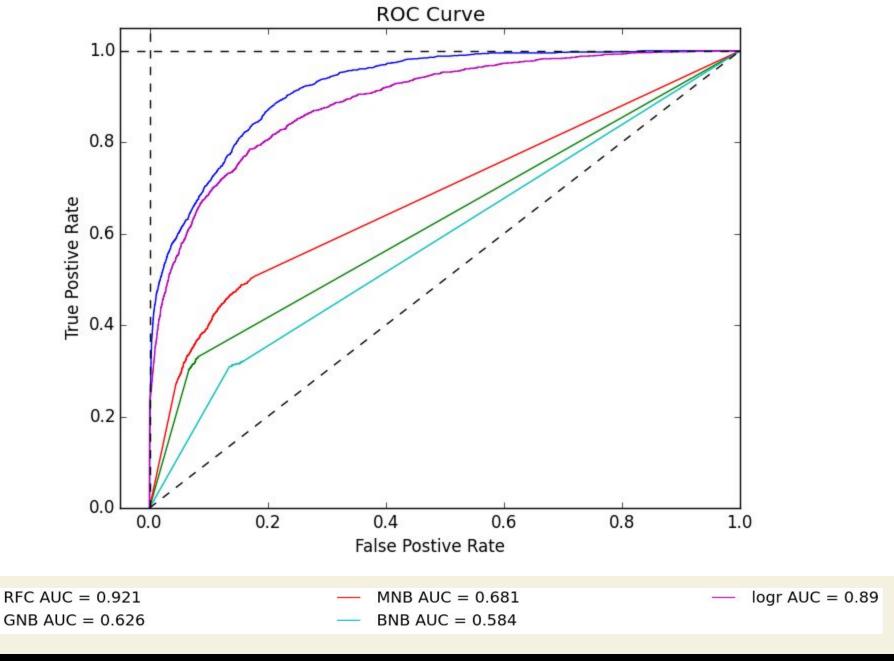


130,000 Bills

Since 1993

6% voted on

7,600



- Oversampling
 - SMOTE

Undersampling

Recall/Precision

Random Forest Classifier Variable Importance calendar resolve union num_amendments report whole commit adoption policy mission 0 2 6 8 10

Feature Importances

Relative Importance