

Graph-Based Model Election Forecasting

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Polling Shortcomings

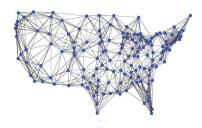


- ▶ Elections are fundamental to democratic society
 - \Rightarrow Draw widespread attention from the public \Rightarrow Forecasting
- Standard way to forecast election is through polling
- Polling is costly and resource intensive
 - \Rightarrow Classify population \Rightarrow Sample \Rightarrow Predict.
- ► Counties are the electoral units ⇒ Report accurate results
- ► Relationship between counties is non-linear ⇒ Model as a network

The Graph Model



- ► Election results are reported in terms of counties
- ► Counties as nodes on a graph ⇒ Election results as a graph signal
- Every pair of counties has a weighted connection
- Connection weight derived from historical voting patterns
- ► Determine the subset of counties to poll
 - ⇒ Exploit theory of sampling graph signals



Election Network



- ▶ Election network as a graph $\mathcal{G} = (\mathcal{N}, \mathcal{E}, \mathcal{W})$
 - $\Rightarrow \mathcal{N}$: counties; $(\mathcal{E}, \mathcal{W})$: historical voting patterns
- ▶ Election results x_i for each county \Rightarrow Signal $\mathbf{x} \in \mathbb{R}^n$
- ► Graph shift operator **S**: captures the strength of the connection
- ▶ **S** is a normal matrix: $SS^H = S^HS$ \Rightarrow Diagonalizable $S = V\Lambda V^H$
- ▶ Define a graph Fourier transform $\hat{\mathbf{x}} = \mathbf{V}^H \mathbf{x}$

$$\Rightarrow$$
 If $\hat{\mathbf{x}} = [\hat{\mathbf{x}}_k \ \mathbf{0}_{n-k}]^T \Rightarrow \mathbf{x}$ is k -bandlimited



$$\mathbf{S} = \begin{pmatrix} S_{11} & S_{12} & 0 & 0 & S_{15} & 0 \\ S_{21} & S_{22} & S_{23} & 0 & S_{25} & 0 \\ 0 & S_{23} & S_{33} & S_{34} & 0 & 0 \\ 0 & 0 & S_{43} & S_{44} & S_{45} & S_{46} \\ S_{51} & S_{52} & 0 & S_{54} & S_{55} & 0 \\ 0 & 0 & 0 & S_{64} & 0 & S_{66} \end{pmatrix}$$

Sampling of Graph Signals



- ightharpoonup x is k-bandlimited $\Rightarrow \hat{x}_k$ freq. components; \mathbf{V}_k eigenvectors
 - ⇒ Perform sampling and reconstruction
- ▶ Determine which $p \ge k$ nodes to sample \Rightarrow Sampling matrix **C**
- ▶ Find **C** such that (CV_k) is invertible
 - \Rightarrow Reconstruct $\tilde{\mathbf{x}} = \mathbf{V}_k(\mathbf{C}\mathbf{V}_k)^{-1}(\mathbf{C}\mathbf{x})$
- ▶ In the presence of noise $C^* = argmax{\sigma_{min}(CV_k)}$
 - \Rightarrow Select the nodes that let the least noise contaminate the signal
 - ⇒ Solvable by computationally tractable algorithms
- ▶ Determining nodes to sample ⇒ Selecting counties to poll

Results



- ▶ Built the network using Presidential election results from 1984-2008
- Added senatorial results for cycles between those same years
- Attempted to predict 2012 election results based on previous years
 - \Rightarrow Tested different ways of relating the data \Rightarrow Use covariance
- ▶ 23 counties to poll from states:
 - ⇒ Florida, California, Montana, Virginia, Massachusetts, etc
- ▶ Prediction relative error: 4.99% of aggregate votes over all counties
- Counties with worst prediction:
 - ⇒ Kenton (KY), Red River (TX), Windham (VT)
- ► Counties with best prediction:
 - \Rightarrow Morrill (NE), Washington (PA), Macon (TN)
- ▶ Prediction matched actual result: Democrat Win

Conclusions & Future Directions



- Used historic election results to identify voting patterns
- ▶ Built an election network with counties (electoral units) as nodes
- ▶ Defined election results as a signal defined on this graph
- Used sampling theory to determine which counties to poll
- ► Contrasted predicted results with latest 2012 elections
- Adjust the election model to incorporate verifiable census data
 - \Rightarrow Includes median income data, poverty index data, etc.
- ▶ Explore newly proposed methods to generate graph shift operator.
 - ⇒ Aggregate all available data into single shift operator.
- ▶ Apply the model to forecast the 2016 election.