

Measuring Issue-Specific Preferences from Votes

Sooahn Shin

Department of Government, Harvard University

Motivation

Two challenges:

- ▶ How can we measure issue-specific ideal points?
 - ▷ Either use a subset of roll-call votes *or*;
 - ▷ Meta-data such as the text of the bills (Gerrish & Blei 2012)
- ▶ How can we interpret latent dimensions which jointly define a multidimensional policy space?
 - ▷ “Lost in translation”: the two-dimensions of *nominat*e have been translated into a number of different ways (Bensel 2016)

I propose a new model which ...

- ▶ arises from classic ideal point estimation and spatial voting theory
- ▶ estimates issue-specific hyperplane and ideal points

Setup

Data:

- ▶ roll-call votes $\{y_{ij}\}_{i=1,j=1}^{n,m}$ (source: *voteview*)
 - ▷ $y_{ij} = 1$ if legislator i votes Yea on roll-call j , and 0 o.w.
- ▶ issue label of bills $\{z_j\}_{j=1}^m$ (source: American Institutions Project)
 - ▷ $z_j \in \{1, \dots, k\}$ for each roll-call j

Issue-Specific Ideal Points Model

Data Generating Process:

$$\mathbf{u}_j \sim \mathbf{vMF}(\boldsymbol{\theta}_{z_j}, \rho) \quad (1)$$

$$\Pr(y_{ij} = 1) = \Phi(\mathbf{w}_j \mathbf{u}_j^\top \mathbf{x}_i - \alpha_j) \quad (2)$$

where $\mathbf{vMF}(\cdot)$ denotes von Mises-Fisher distribution and ...

- ▶ $\boldsymbol{\theta}_z$: *direction* vector of issue-specific hyperplane ($\|\boldsymbol{\theta}_z\| = 1$)
- ▶ \mathbf{u}_j : *direction* vector of the j th hyperplane ($\|\mathbf{u}_j\| = 1$)
 - ▷ orthogonal to *cutting line* in *nominat*e (Poole & Rosenthal 1997)
- ▶ w_j *magnitude* of the j th hyperplane ($w_j > 0$)
 - ▷ $w_j \mathbf{u}_j = \beta_j$, discrimination parameter (Clinton et al. 2004)
- ▶ \mathbf{x}_i : multidimensional ideal point
- ▶ α_j : difficulty parameter

Identification:

- ▶ Fix directionality of \mathbf{u}_j to avoid reflection invariance
 - ▷ Code the voting data so that $y_{ij} = 1$ indicates a conservative vote
 - ▷ Can implement this by anchoring a conservative legislator's votes

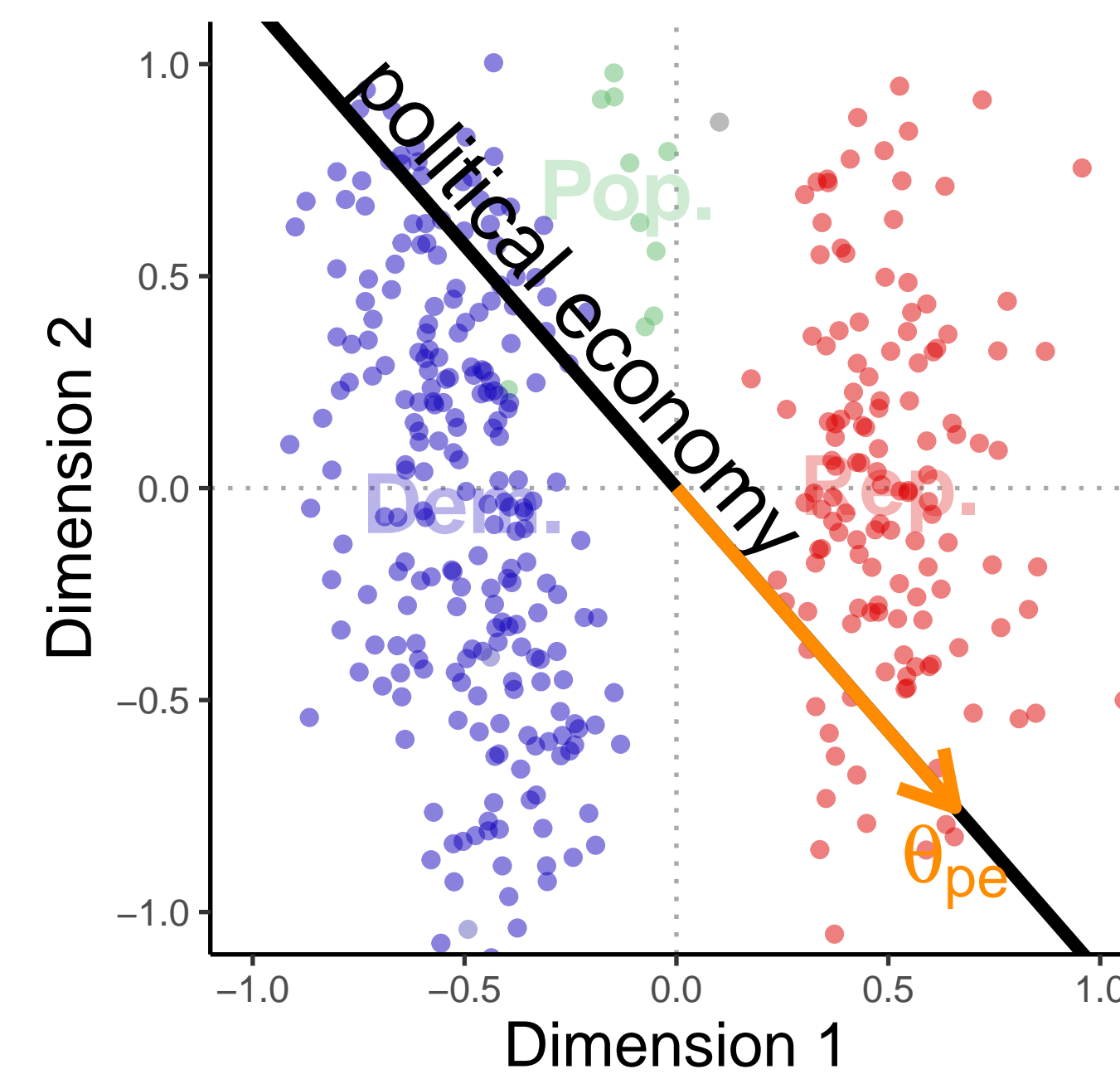
Issue-specific ideal points estimator:

$$\hat{\mathbf{x}}_i^*(z) := \frac{\hat{\boldsymbol{\theta}}_z^\top \hat{\mathbf{x}}_i}{\hat{\boldsymbol{\theta}}_z^\top \hat{\boldsymbol{\theta}}_z} \quad (3)$$

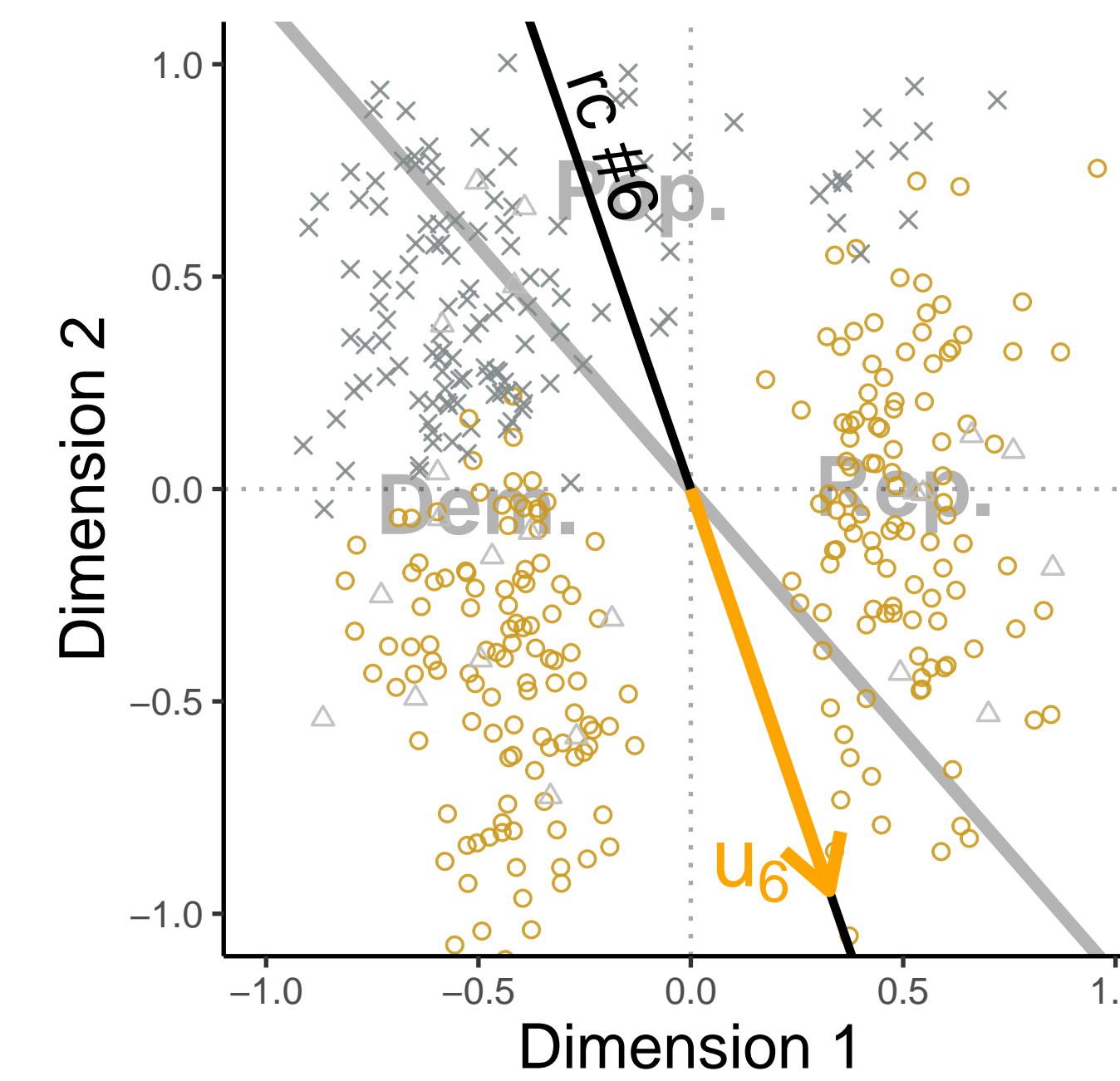
1. Estimate the issue-specific hyperplane $\boldsymbol{\theta}_z$
2. Project the ideal points onto the hyperplane

Application: U.S. Congress during the Gilded Age

Understanding the model based on spatial voting theory:



The 53rd H.R. (1893–1895)



Votes on Bland Amendment

$$\Pr(y_{ij} = 1) = p(U_i(\mathbf{o}_{yj}) > U_i(\mathbf{o}_{nj})) \quad (4)$$

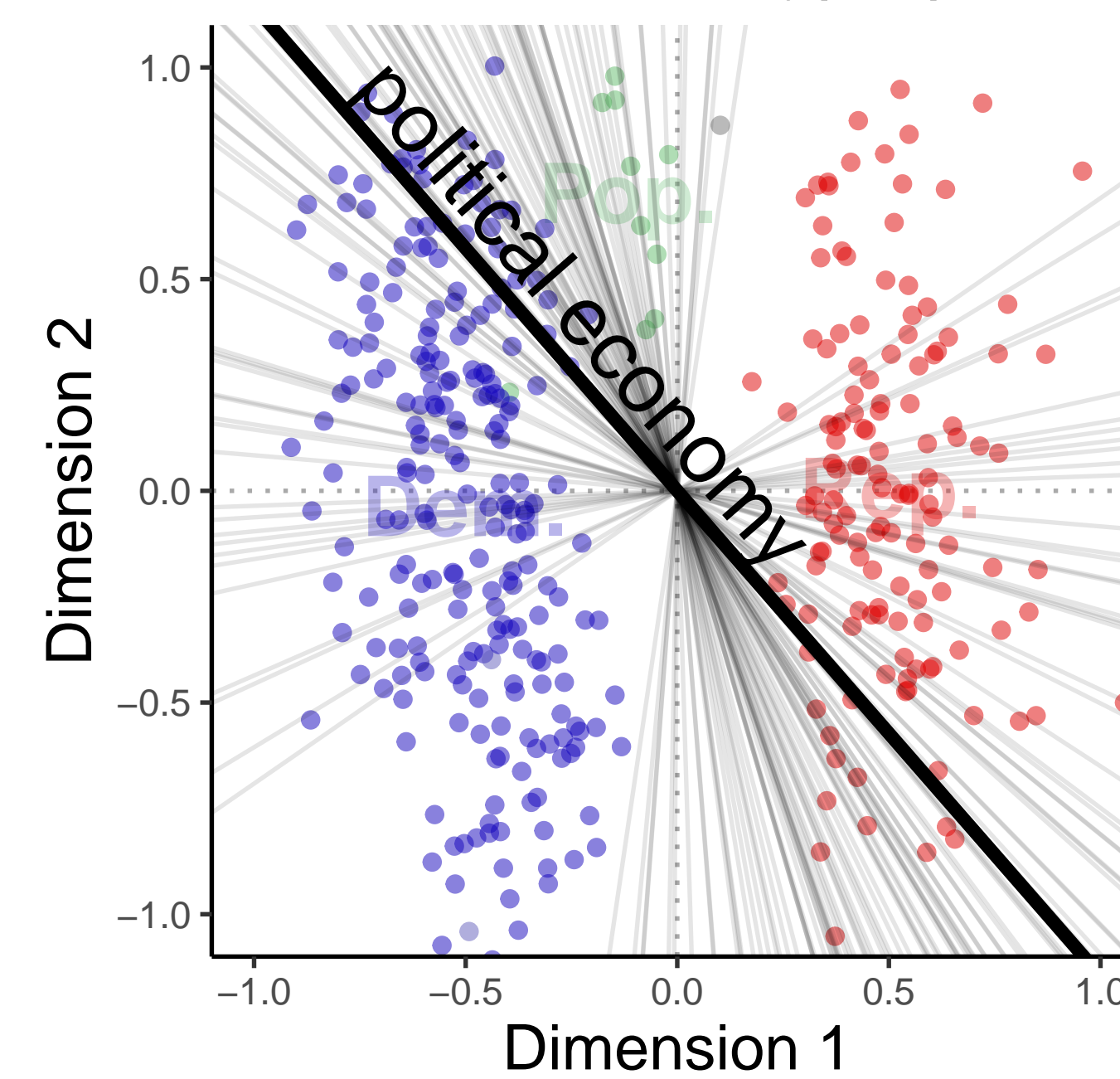
$$= p(-\|\mathbf{x}_i - \mathbf{o}_{yj}\|^2 + \epsilon_{yij} > -\|\mathbf{x}_i - \mathbf{o}_{nj}\|^2 + \epsilon_{nij}) \quad (5)$$

$$= p(2(\mathbf{o}_{yj} - \mathbf{o}_{nj})^\top \mathbf{x}_i - (\mathbf{o}_{yj}^\top \mathbf{o}_{yj} - \mathbf{o}_{nj}^\top \mathbf{o}_{nj}) > \epsilon_{nij} - \epsilon_{yij}) \quad (6)$$

$$= \Phi(\mathbf{w}_j \mathbf{u}_j^\top \mathbf{x}_i - \alpha_j) \quad (7)$$

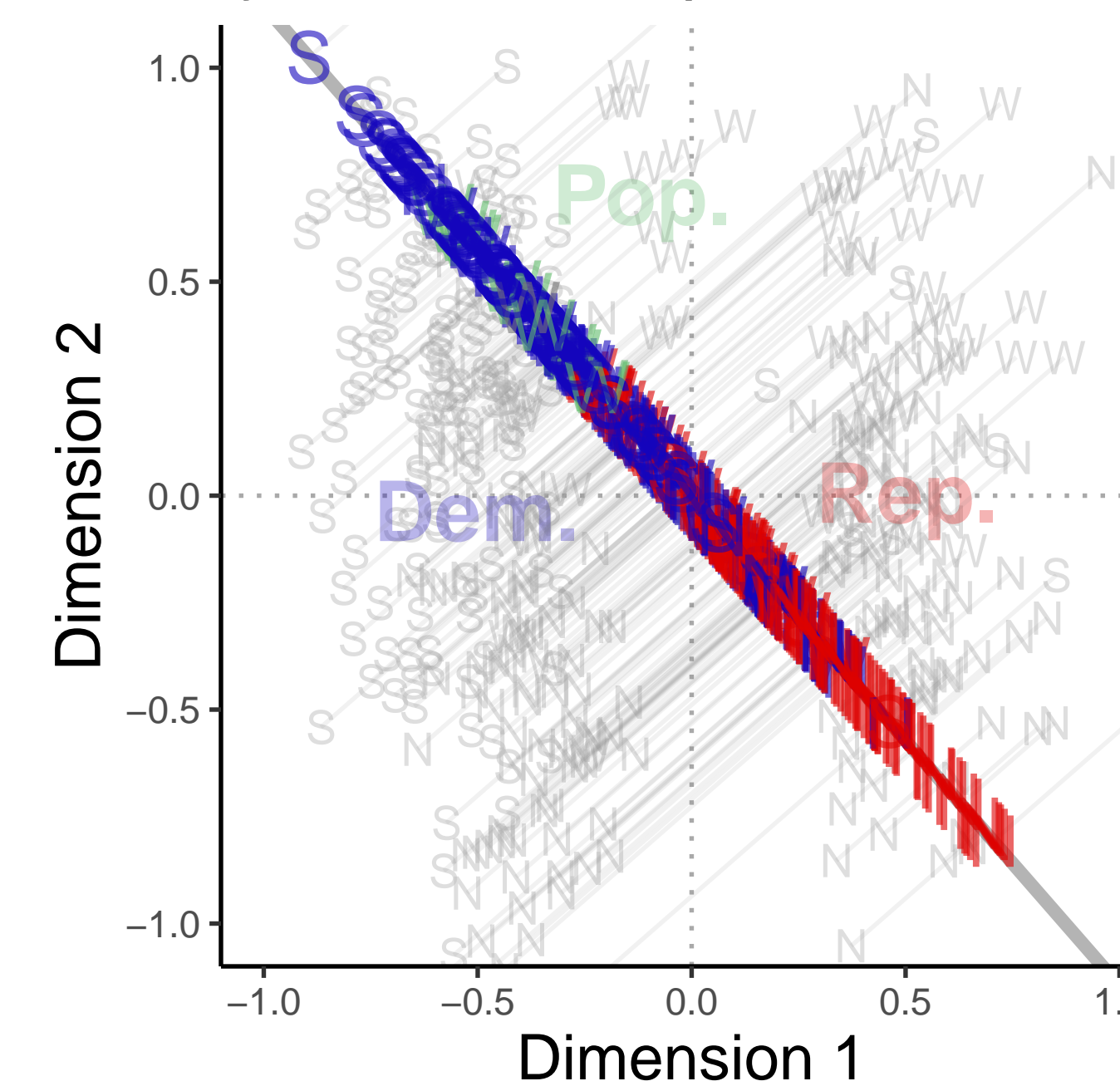
Two-step estimation:

1. Estimate the issue hyperplane



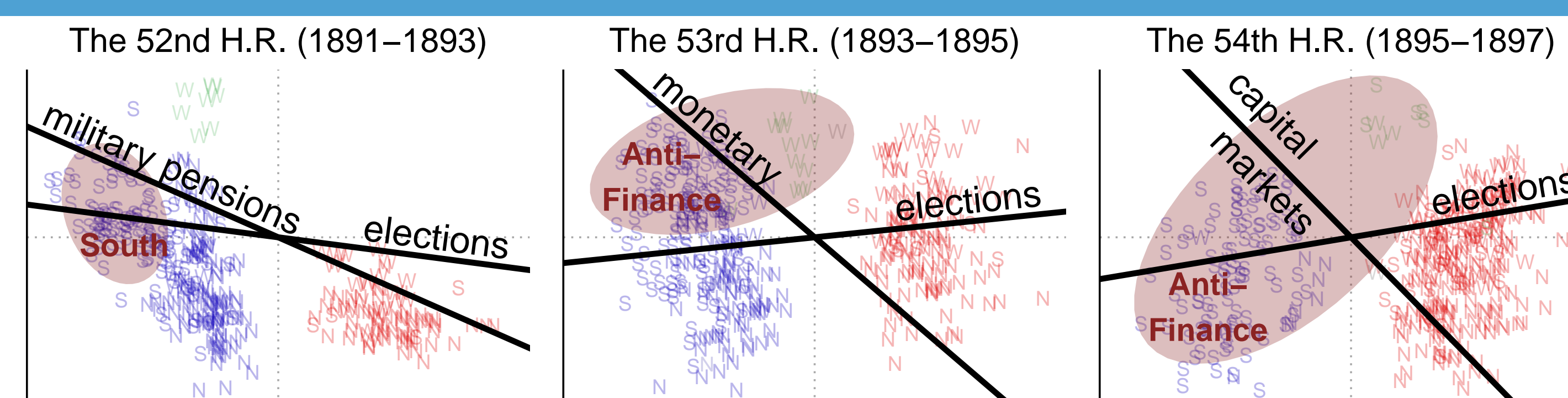
*Gray lines: \mathbf{u}_j 's such that $z_j = pe$

2. Project the ideal points



*Labels: State (South/ North/ West)

Main Findings: Partisan v. Sectional v. Sectoral Alignments



- ▶ “Southern Nation”: unity among Southern Dem. (Bateman et al. 2018)
- ▶ Division over gold: farmers/miners v. int'l fin./commerce (Frieden 2015)

Properties of Issue-Specific Ideal Points

Issue-specific voting behavior

$$y_{ij}^* := \mathbf{w}_j \boldsymbol{\theta}_{z_j}^\top \mathbf{x}_i^*(z_j) \boldsymbol{\theta}_{z_j} - \alpha_j \quad (8)$$

$$= \mathbf{w}_j \boldsymbol{\theta}_{z_j}^\top \mathbf{x}_i - \alpha_j \quad (9)$$

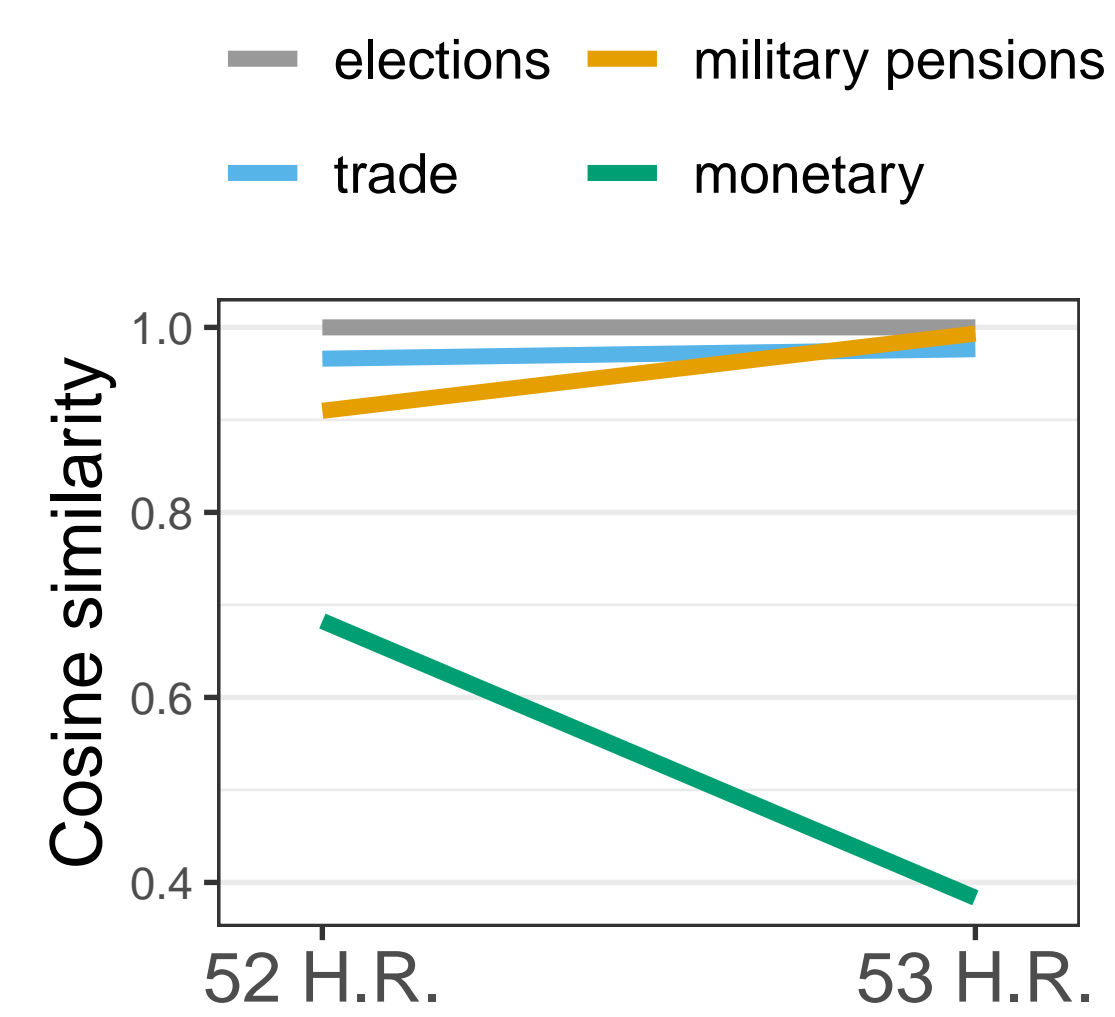
$$= \mathbb{E}_{\mathbf{u}_j}[\mathbf{1}\{y_{ij} = 1\}] | \boldsymbol{\theta}_{z_j}, \mathbf{w}_j, \mathbf{x}_i, \alpha_j \quad (10)$$

- ▶ A nice summary of issue-specific voting behavior w.r.t.
 - (1) stochastic noise of the utility and
 - (2) deviation of the j th roll call's hyperplane (\mathbf{u}_j) from the issue-specific hyperplane ($\boldsymbol{\theta}_z$).

Comparing issue hyperplanes

$$|\cos(\boldsymbol{\theta}_{\text{Issue A}}, \boldsymbol{\theta}_{\text{Issue B}})| \quad (11)$$

- ▶ A measure of “constraint”: the degree to which policy attitudes on issue A are aligned with those on issue B (Marble & Tyler 2021)
- ▶ Related measure: mean *item2vec* alignment (Sohn 2017)



Identified under rotations

$$\frac{\boldsymbol{\theta}_z^\top \mathbf{x}_i}{\boldsymbol{\theta}_z^\top \boldsymbol{\theta}_z} = \frac{(R\boldsymbol{\theta}_z)^\top R\mathbf{x}_i}{(R\boldsymbol{\theta}_z)^\top R\boldsymbol{\theta}_z} \text{ for } R^\top R = \mathbb{I} \quad (12)$$

- ▶ Rotational invariance of the likelihood does not threaten the identification.

Concluding Remarks: Issue and Dimension

The notion of *dimension* and its limitation:

- ▶ “The ‘spaces’ of interest are ultimately metaphors and both the dimensions spanning these spaces and agents’ positions on these dimensions are fundamentally unobservable (Benoit & Laver 2012, p.196)”
- ▶ “[F]rom a technical point of view, choosing the dimensionality of the policy space is essentially a model selection problem (Moser et al. 2021, p.146)”
- ▶ Multidimensional ideal points models are only identifiable with ad hoc conditions due to the rotational invariance

By incorporating *issue* labels in the model,

- ▶ The proposed method suggests a way to construct linear combinations of low dimensional latent traits, which explain the voting behavior for each substantive issue;
- ▶ And further provides a principled way to
 - (1) interpret the latent dimensions using issue hyperplanes and
 - (2) compare the alignments of different issues.