State Legislative Aggregate Ideology Data
May 2018 Update[*]

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1   Introduction

The Shor-McCarty state legislative aggregate ideology data is being released as an update to the data underlying Shor and McCarty (2011). These are based on individual-level ideal point estimates (called "Shor-McCarty NPAT Scores") described fully in that article. Estimates are all in NPAT common ideological space to facilitate explicit comparisons across time and between states.

The new data now stretches through 2016, with and significant filling in of missing years. There are now 2,025 aggregate chamber-year observations, compared with 1,816 in the last release.

As we do with each new release, the individual level data underlying this release has been been extensively cleaned to minimize the random noise inherent in acquiring roll call votes from printed journals and online sources.

The newest version of the data and codebook can always be found on the American Legislatures web site (http://americanlegislatures.com). We are actively working on additional data, and will be releasing updated data sets over time.

2   Overview

Ideology and polarization estimates are measured at the chamber level. Not all state-years have measurements, due to the ragged availability of state legislative roll call data. The authors are actively working on additional data, and will be releasing updated data sets over time.

We generated bootstrapped error estimates for our scores at both individual and aggregate levels. At the individual level, we take advantage of the simulations that underlie the Bayesian IRT model to get a sense of the estimation uncertainty of legislator ideology in roll call space. We then simulate a large number of state-specific mapping coefficients for each state. Then, for each simulation of the legislator's ideal point, we randomly draw (with replacement) one set of mapping coefficients, predicting an NPAT score each time. The standard deviation of the simulations is our bootstrapped error estimate. At the aggregate level, we ge quantities of interest using the individual common space scores and their bootstrapped errors. For example, in each simulation, we calculate a chamber median. Iterating

over a large number of simulations, we take the standard deviation of the simulated medians as our measure of aggregate uncertainty.

3 Codebook

st
State abbreviation

alpha
State code, alphabetical order

fips
State FIPS code

icpsr
State ICPSR code

year
Chronological Year

*_chamber
Whole chamber median, by chamber

*_dem
Democratic party median, by chamber

*_rep
Republican party median, by chamber

*_majority
Majority party median, by chamber

*_diffs
Distance between party medians, by chamber. Our preferred measure of polarization.

*_distance
Average distance between any two members, by chamber. This is an alternative, party-free, measure of polarization.

*_dem_sd
Democratic party heterogeneity, by chamber

*_rep_sd
Republican party heterogeneity, by chamber

*_chamber_sd

Whole chamber heterogeneity, by chamber

`*_dem_error`
Democratic party median error estimate, by chamber

`*_rep_error`
Republican party median error estimate, by chamber

`*_chamber_error`
Whole chamber median error estimate, by chamber

`*_majority_error`
Majority party chamber median error estimate, by chamber

References

Shor, Boris and Nolan McCarty. 2011. "The Ideological Mapping of American Legislatures."
    American Political Science Review 105(3):530–551.