Empirical Validation of Automated Redistricting Algorithms on the Virginia ${\rm House\ of\ Delegates\ District\ Map}$

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Literature Review

Get writing!

Method

Notes, this is still a rough draft.

Choice of Research Method

For this study, I chose to use the experimental design method because it will allow me to isolate the hypothetical impact of the redistricting algorithm from other possible confounding variables. This method also includes the use of a control group, which allows the researcher to establish causation.

Experimental Units

The experimental units for this study are the complete datasets for each election year in Viriginia. I have one dataset for each of these years: 2015, 2017, 2019. Every row in each dataset corresponds to a precinct, the smallest geographical unit by which votes are tabulated in Virginia. For each precinct, I have the following attributes: total population, population by race, total voting-age population (VAP)(population over the age of 18), VAP by race, total votes for the democratic House of Delegates (HOD) candidate, total votes for the Republican HOD candidate, and the total votes for any other HOD candidate. Additionally, each precinct has a polgon associated with it that represents its geographical shape.

Treatments

The treatments for this study are the three different redistricting algorithm that I'm comparing: Markov chain Monte Carlo (Fifield, Higgins, et al., 2020), Sequential Monte

Carlo (McCartan & Imai, 2020), and Random Seed Growth (Chen & Rodden, 2013). I'm using the implementations in the R programming language "redist" package (Fifield, Kenny, et al., 2020)

 ${\bf Results}$

Discussion

Conclusion

References

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 Using Markov Chain Monte Carlo. *Journal of Computational and Graphical*Statistics, 0(0), 1–14. https://doi.org/10.1080/10618600.2020.1739532
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