

# VoteKit: A Python package for computational social choice research

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DOI: [10.xxxxxx/draft](https://doi.org/10.xxxxxx/draft)

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Submitted: 01 January 1970

Published: unpublished

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## Summary

The scholarly study of elections, known as *social choice theory*, centers on the provable properties of voting rules. Practical work in democracy reform focuses on designing or selecting systems of election to produce electoral outcomes that promote legitimacy and broad-based representation. For instance, the dominant electoral system in the United States is a one-person-one-vote/winner-take-all system, sometimes known as PSMD (plurality in single member districts); today, there is considerable reform momentum in favor of ranked choice voting because it is thought to mitigate the effects of vote-splitting and to strengthen prospects for minority representation, among other claimed properties.<sup>1</sup> Across the world, systems of election—and prospects for system change—vary substantially. From both a scholarly and a practical perspective, many questions arise about comparing the properties and tendencies of diverse systems of election in a rigorous manner.

VoteKit ([MGGG Redistricting Lab, 2023](#)) is a Python package designed to facilitate just that kind of analysis, bringing together multiple types of functionality. Users can:

1. Create synthetic *preference profiles* (collections of ballots) with a choice of generative models and behavioral parameters;
2. Read in real-world *cast vote records* (CVRs) as observed examples of preference profiles; clean and process ballots, including by deduplication and handling of undervotes and overvotes;
3. Run a variety of *voting rules* to ingest preference profiles and output winner sets and rankings; and
4. Produce a wide range of *summary statistics* and *data visualizations* to compare and analyze profiles and election outcomes.

## Statement of need

Social choice theory grew out of welfare economics in the mid-twentieth century and has been recognized as a deep and highly applicable area of economic theory, forming part of the basis for at least four Nobel Prize awards.<sup>2</sup> Since the 1990s, a new fusion of economics and computer science has emerged under the name of *computational social choice*, studying

<sup>1</sup>Recent ranked-choice voting reforms include the adoption of instant runoff voting (IRV) in Maine, Alaska, New York City, and single transferable vote (STV) in Portland, Oregon. Advocacy groups claiming various pro-democratic properties of ranked choice include [Campaign Legal Center](#), [FairVote](#), and many others.

<sup>2</sup>Nobel Laureates with significant work in social choice include Arrow, Sen, Maskin, and Myerson.

34 questions of complexity and design and further advancing the axiomatic study of elections.<sup>3</sup>  
35 But most of these innovations have been highly abstract, and there has been a significant gap  
36 in the literature—and in the landscape of software—between the theory and the practice of  
37 democracy.

38 On the software side, researchers have built a multitude of different packages for generating  
39 and analyzing elections.<sup>4</sup> Most packages, to our knowledge, handle just one part of the  
40 research arc; for instance, PrefSampling (Boehmer et al., 2024) generates profiles but does  
41 not conduct elections, while VoteLib (Jan Šimbera, 2020) *only* conducts elections. Others,  
42 like PrefLibTools (Nicholas Mattei and Simon Rey, 2022) and PrefVoting (Eric Pacuit and  
43 Wesley H. Holliday, 2022), provide support for generating profiles and conducting single-winner  
44 elections. Packages with multi-winner capability, like abcvoting (Lackner et al., 2023) or  
45 Apportionment (Martin Lackner, 2022), do not support ranked voting. To illustrate the gap  
46 this leaves, note that single transferable voting (STV), a voting system actually used for  
47 political election in six countries, is curiously absent. VoteKit is built to provide an end-to-end  
48 pipeline that supports ranked, scored, and approval profiles as well as single- and multi-winner  
49 elections, with an emphasis on practical applicability.

## 50 Area of need: Generative models

51 For one concrete example of a literature and software gap, consider the construction of  
52 *generative models*. This term is often associated with large language models as paradigms  
53 of artificial intelligence; here, what is being generated is realistic voting rather than realistic  
54 language. In this setting, a generative model of voting is a probability distribution on the  
55 set of all possible ballots that can be cast in a given election style; profiles can be sampled  
56 from a generative model to produce simulated or synthetic elections. Having sources of rich,  
57 varied, and realistic data is essential to an empirically grounded research program to probe  
58 the properties of voting rules. Good generative models are also essential to advise reformers  
59 deciding between options in a new locality, as they enable generation of synthetic profiles keyed  
60 to the scale, demographics, and election specs of that specific place. But most of the models  
61 in the literature, like the Impartial Culture model (all permutations of candidates are equally  
62 likely) or the Impartial Anonymous Culture model (sampling proportional to volume measure  
63 on the simplex of weighted averages of permutations) are mathematically tractable but highly  
64 unrealistic. This is bluntly described by Tideman and Plassman in a survey of generative  
65 methods: in their words, “None of the 11 models discussed so far are based on the belief that  
66 the associated distributions [...] might actually describe rankings in actual elections” (Tideman  
67 & Plassmann, 2010). They therefore recommend *spatial models* instead, which themselves are  
68 of dubious realism for the selection of political candidates.<sup>5</sup>

69 VoteKit implements many of the models described in those surveys, as well as newer mathe-  
70 matical models that give users the ability to generate profiles that are designed to comport  
71 with real-world ranking behavior and particularly to generate polarized elections. Two leading

<sup>3</sup>For example, a very active research direction in computational social choice theory has been the development of fairness axioms for approval elections, such as the definition called JR (justified representation) and its relatives, which have been extended to rankings. See (Aziz et al., 2017; Skowron et al., 2017) and their references.

<sup>4</sup>See for instance the extensive array of open-source tools on the Computational Social Choice (COMSOC) community page (Ulle Endriss and Simon Rey, n.d.) including the widely used collection of ranked data called PrefLib (Ulle Endriss and Simon Rey, n.d.). See also the materials provided by FairVote, including their DataVerse and GitHub (FairVote, n.d.). The ArXiv preprint (Boehmer et al., n.d.) provides an impressively comprehensive list of numerical experiments on elections. The PRAGMA Project (<https://perma.cc/2P6V-8ZER>) echoes our statement of need, noting that the current literature and software falls short in practical applicability and that the understanding of real and synthetic data is “very limited.”

<sup>5</sup>Spatial models assume voters rank by proximity in a metric space defined by issue positions or other attributes; the metric space may be latent, or unknown to voters, but it is presumed to universally govern the way voters rank candidates. See for instance (Burden, 1997), which introduces probabilistic voting keyed to proximity. Though spatial models have been argued to perform adequately to model roll call voting in Congress, their efficacy for selecting political representation is debatable. In a meta-analysis of 163 papers (Boehmer et al., n.d.), the authors report that Impartial Culture and Euclidean (spatial) models make up more than 75% of the election experiments found in 163 papers.

choices are based on classic statistical ranking mechanisms, called the Plackett–Luce (PL) and Bradley–Terry (BT) models; another model called the Cambridge Sampler (CS) draws from historical ranking data in Cambridge, MA city council elections. These models have flexible parameters, allowing users to vary voting bloc proportions, candidate strength within slates, and polarization between blocs. These parameters can be specified or randomly sampled.

## Area of need: Comparison and communication

In the realm of democracy reform, groups of stakeholders often ask researchers to provide modeling studies to decide on what shift to make in electoral systems, as the project list below makes clear. VoteKit implements voting rules that stakeholders often seek to compare, with parameters designed to be tailored by the user to the specific locality under study. Available voting rules include:

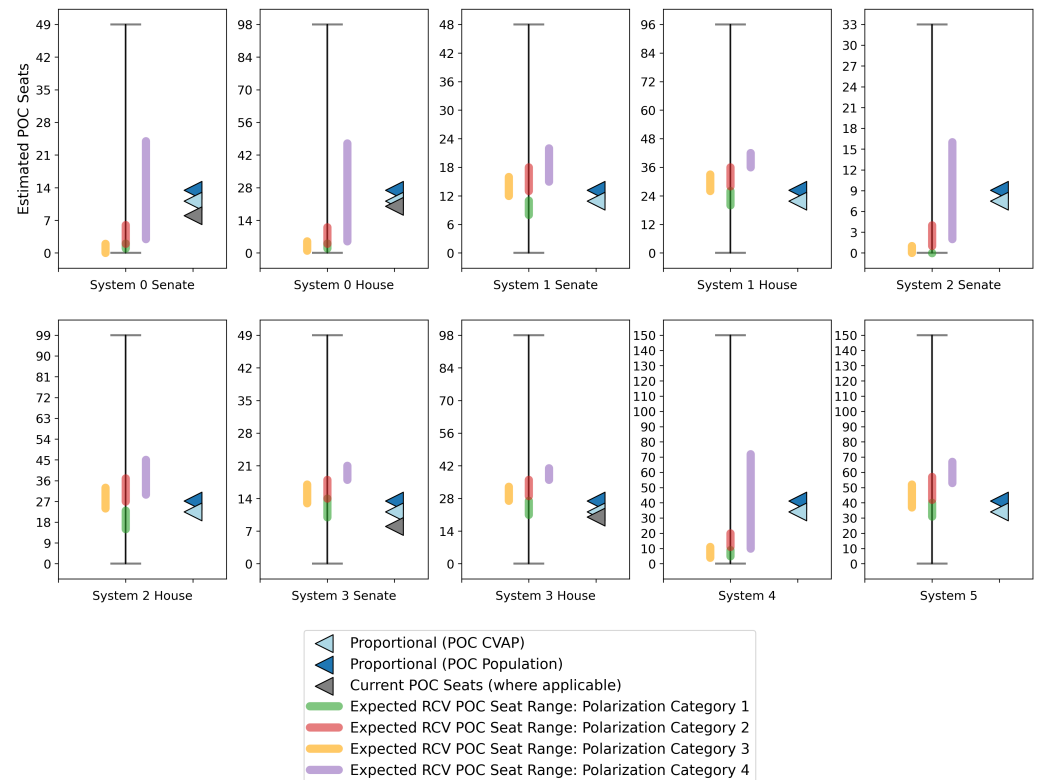
- **Ranking-based (ordinal).** Plurality/SNTV, STV and IRV, (generalized) Borda, Alaska<sup>6</sup>, Top-Two, Dominating sets/Smith method, Condo-Borda<sup>7</sup>, Sequential RCV.
- **Score-based (cardinal).** Range voting, Cumulative, Limited.
- **Approval-based (set).** Approval voting, Bloc plurality.

This list does not include every method that has attracted theoretical investigation; rather, it is oriented to methods used or considered for political representation, such as the final-four system in Alaska or the sequential RCV in Utah local elections. See generally (Amorós et al., 2016; Emerson, 2013; McCune et al., 2023; Reynolds et al., 2008; Tideman, 1995) for references. In addition, VoteKit is flexible enough to allow users to write custom voting rules.

Reform advocates also need to describe voting mechanisms and their likely outcomes effectively to members of their communities. The end-to-end pipeline provided by VoteKit allows advocates to toggle different system settings and compare expected outcomes. For example, Figure 1 is reprinted from a report on reform proposals for the chambers of the Washington state legislature, with Systems 0-3 as paired bicameral systems and Systems 4-5 as unicameral solutions. Using the codebase that formed the foundation of VoteKit, researchers compared the expected outcomes for minority representation under these six systems.

<sup>6</sup>Our model of the Alaska method is an SNTV/STV hybrid that uses single non-transferable vote to choose a set of finalists, then runs STV on the same preference profile to fill the seats. Alaska's elections run this with four finalists and one seat; the top-two system runs this with two finalists and one seat.

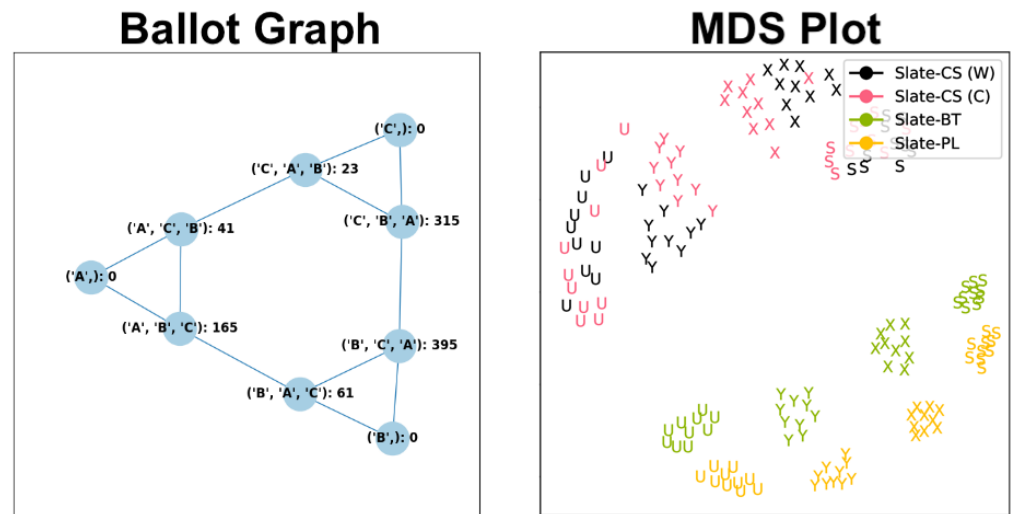
<sup>7</sup>This system orders candidates within dominating sets by Borda score. Note that this is distinct from Black's method (Black, 2012), which uses Borda score as a backup system in case the smallest dominating set is not a singleton.



**Figure 1:** A comparison of a variety of electoral systems and their effect on minority representation in a case study of reform proposals for the Washington state legislature (MGGG Redistricting Lab, 2021d).

## 99 Area of need: Resources for research

100 Previous research works such as (Elkind et al., 2017) have compared properties of earlier  
 101 generative models; VoteKit has functionality to fully replicate this work and facilitates robust  
 102 comparisons across a more comprehensive and up-to-date list of alternatives. It also offers  
 103 new analytical tools that will support research on elections. Some examples are shown in  
 104 Figure 2. At left is a *ballot graph*, where nodes are ballots weighted by their frequency in the  
 105 profile; a recent research paper shows that ballot graphs can be metrized to realize classical  
 106 statistical ranking distances, like Kendall tau and the Spearman footrule (Duchin & Tapp,  
 107 2024). VoteKit also implements a class of election distances, as surveyed in (Boehmer et al.,  
 108 2022). Choices for measuring the difference between two profiles on the same set of candidates  
 109 include  $L^p$  distance and Wasserstein (earth-mover) distance. At right is a multidimensional  
 110 scaling (MDS) plot of a different set of data, showing mutual  $L^1$  differences between generated  
 111 profiles across various selections of model (shown in colors) and candidate strength parameters  
 112 (shown with symbols), enabling comparisons in the style of (Szufa et al., 2020).



**Figure 2:** At left, the ballot graph for a 3-candidate election. There is one node per possible ballot, and the weights show the number of instances of that ballot in the profile. At right, a multidimensional scaling (MDS) plot for 160 synthetic profiles made with various generative models and candidate strength parameters for two slates of 3 candidates each. The MDS plot is a low-distortion planar embedding of those 160 profiles and their pairwise differences.

Finally, VoteKit interacts seamlessly with a wide range of actual vote data, such as thousands of political elections collected by FairVote and a cleaned repository of over 1000 Scottish STV local government elections (FairVote, n.d.; MGGG Redistricting Lab, n.d.). Previously, the use of real data in election research was often extremely limited; for instance, a recent survey reports that the single most popular “real-life” dataset has been a survey of 5000 respondents’ sushi preferences (Boehmer et al., n.d.).

## Projects

A significant number of white papers and scholarly articles have used VoteKit (and its predecessor codebase) in recent years. These include the following.

- A large number of case studies in ranked-choice modeling, such as studies for the city councils of Chicago, IL (MGGG Redistricting Lab, 2019b) and Lowell, MA (MGGG Redistricting Lab, 2019a); the state legislatures of Oregon and Washington (MGGG Redistricting Lab, 2021a, 2021d), and a range of county commissions and school boards across the Pacific Northwest (MGGG Redistricting Lab, 2021c, 2021b);
- A study modeling the impact of proposed legislation called the Fair Representation Act, which would convert U.S. Congressional elections to the single transferable vote system (MGGG Redistricting Lab, 2022);
- A detailed study isolating the impacts of varying hypotheses about voter behavior and candidate availability on the Massachusetts legislature (MGGG Redistricting Lab, 2024);
- A peer-reviewed article for an election law audience on the impact of STV elections on minority representation (Benadè et al., 2021);
- A peer-reviewed article for a CS/econ audience that probes whether STV delivers proportional representation (Benadè et al., 2024); and
- A peer-reviewed article for an CS/operations research audience on optimizing to “learn” blocs and slates in real-world elections (Duchin & Tapp, 2024).

## Acknowledgements

This work was initiated in a research cluster in Summer 2023, funded by the Democracy Fund and graciously hosted at the Faculty of Computing and Data Sciences at Boston University and the Tisch College of Civic Life at Tufts University. Major contributors to the initiation of the project include Brenda Macias, Emarie De La Nuez, Greg Kehne, Jordan Phan, Rory Erlich, James Turk, and David McCune. Earlier code contributions were made by Chanel Richardson, Anthony Pizzimenti, Gabe Schoenbach, Dylan Phelan, Thomas Weighill, Dara Gold, and Amy Becker; recent code contributions also come from Peter Rock, Kevin Quinn, and Divij Sinha. The authors also thank Deb Otis, Jeanne Clelland, and Michael Parsons for helpful feedback. FairVote's data repository in Dataverse ([https://dataverse.harvard.edu/dataverse/rcv\\_cvrs](https://dataverse.harvard.edu/dataverse/rcv_cvrs)) and RCV Cruncher code on GitHub ([https://github.com/fairvotereform/rcv\\_cruncher/](https://github.com/fairvotereform/rcv_cruncher/)) are excellent open-source efforts that were inspirational for the current project.

## References

- Amorós, P., Puy, M. S., & Martínez, R. (2016). Closed primaries versus top-two primaries. *Public Choice*, 167, 21–35. <https://doi.org/10.1007/s11127-016-0328-5>
- Aziz, H., Brill, M., Conitzer, V., Elkind, E., Freeman, R., & Walsh, T. (2017). Justified representation in approval-based committee voting. *Social Choice and Welfare*, 48(2), 461–485.
- Benadè, G., Buck, R., Duchin, M., Gold, D., & Weighill, T. (2021). *Ranked choice voting and proportional representation*.
- Benadè, G., Donnay, C., Duchin, M., & Weighill, T. (2024). Proportionality for ranked voting, in theory and practice. In *Preprint*. <https://mggg.org/PRVTP>.
- Black, D. (2012). *The theory of committees and elections*. Springer Dordrecht.
- Boehmer, N., Faliszewski, P., Janeczko, Ł., Kaczmarczyk, A., Lisowski, G., Pierczyński, G., Rey, S., Stolicki, D., Szufa, S., & Wąs, T. (n.d.). *Guide to numerical experiments on elections in computational social choice*.
- Boehmer, N., Faliszewski, P., Janeczko, Ł., Kaczmarczyk, A., Lisowski, G., Pierczyński, G., Rey, S., Stolicki, D., Szufa, S., & Wąs, T. (2024). *Guide to numerical experiments on elections in computational social choice*. <https://arxiv.org/abs/2402.11765>
- Boehmer, N., Faliszewski, P., Niedermeier, R., Szufa, S., & Wąs, T. (2022). Understanding distance measures among elections. In L. D. Raedt (Ed.), *Proceedings of the thirty-first international joint conference on artificial intelligence, IJCAI-22* (pp. 102–108). International Joint Conferences on Artificial Intelligence Organization. <https://doi.org/10.24963/ijcai.2022/15>
- Burden, B. C. (1997). Deterministic and probabilistic voting models. *American Journal of Political Science*, 41(4), 1150–1169. <http://www.jstor.org/stable/2960485>
- Duchin, & Tapp. (2024). Learning blocs and slates from observed elections. *Preprint*.
- Elkind, E., Faliszewski, P., Laslier, J.-F., Skowron, P., Slinko, A., & Talmon, N. (2017). What do multiwinner voting rules do? An experiment over the two-dimensional euclidean domain. *Proceedings of the AAAI Conference on Artificial Intelligence*, 31.
- Emerson, P. (2013). The original borda count and partial voting. *Social Choice and Welfare*, 40, 353–358. <https://doi.org/10.1007/s00355-011-0603-9>
- Eric Pacuit and Wesley H. Holliday. (2022). *Pref\_voting*.
- FairVote. (n.d.). *RCV cruncher*.



- 182 Jan Šimbera. (2020). *Votelib*.
- 183 Lackner, M., Regner, P., & Krenn, B. (2023). Abcvoting: A Python package for approval-  
184 based multi-winner voting rules. *Journal of Open Source Software*, 8(81), 4880. <https://doi.org/10.21105/joss.04880>
- 185
- 186 Martin Lackner. (2022). *Apportionment*.
- 187 McCune, D., Martin, E., Latina, G., & Simms, K. (2023). *A comparison of sequential*  
188 *ranked-choice voting and single transferable vote*. <https://arxiv.org/abs/2306.17341>
- 189 MGGG Redistricting Lab. (n.d.). *Scottish STV election repo*.
- 190 MGGG Redistricting Lab. (2019a). Findings on the city of lowell's election systems. In *White*  
191 *Paper*. <https://mggg.org/Lowell-Detailed-Report>.
- 192 MGGG Redistricting Lab. (2019b). Study of reform proposals for chicago city council. In  
193 *White Paper*. <https://mggg.org/publications/Chicago.pdf>.
- 194 MGGG Redistricting Lab. (2021a). Analysis of election systems for oregon state. In *White*  
195 *Paper*. <https://mggg.org/Oregon>.
- 196 MGGG Redistricting Lab. (2021b). Analysis of election systems for the chelan county,  
197 washington board of county commissioners. In *White Paper*. [https://mggg.org/Chelan\\_](https://mggg.org/Chelan_County)  
198 [County](https://mggg.org/Chelan_County).
- 199 MGGG Redistricting Lab. (2021c). Analysis of election systems for the tukwila, WA school  
200 district. In *White Paper*. <https://mggg.org/Tukwila>.
- 201 MGGG Redistricting Lab. (2021d). Analysis of election systems for washington state. In *White*  
202 *Paper*. <https://mggg.org/Washington>.
- 203 MGGG Redistricting Lab. (2022). Modeling the Fair Representation Act. In *White Paper*.  
204 <https://mggg.org/FRA-Report>.
- 205 MGGG Redistricting Lab. (2023). *VoteKit*.
- 206 MGGG Redistricting Lab. (2024). Comparing electoral systems for the massachusetts legislature.  
207 In *White Paper*. <https://mggg.org/MA-RCV>.
- 208 Nicholas Mattei and Simon Rey. (2022). *PrefLib-tools*.
- 209 Reynolds, A., Reilly, B., & Ellis, A. (2008). *Electoral system design: The new international*  
210 *IDEA handbook*. International Institute for Democracy; Electoral Assistance.
- 211 Skowron, P., Lackner, M., Brill, M., Peters, D., & Elkind, E. (2017). Proportional rankings.  
212 *International Joint Conference on Artificial Intelligence (IJCAI 2017)*.
- 213 Szufa, S., Faliszewski, P., Skowron, P., Slinko, A., & Talmon, N. (2020). Drawing a map of elec-  
214 tions in the space of statistical cultures. *Proceedings of the 19th International Conference*  
215 *on Autonomous Agents and MultiAgent Systems*, 1341–1349. ISBN: 9781450375184
- 216 Tideman, N. (1995). The single transferable vote. *Journal of Economic Perspectives*, 9(1),  
217 27–38. <https://doi.org/10.1257/jep.9.1.27>
- 218 Tideman, N., & Plassmann, F. (2010). *The structure of the election-generating universe*.  
219 <https://api.semanticscholar.org/CorpusID:119079594>
- 220 Ulle Endriss and Simon Rey. (n.d.). *COMSOC community site*.