

Compartmental voter model

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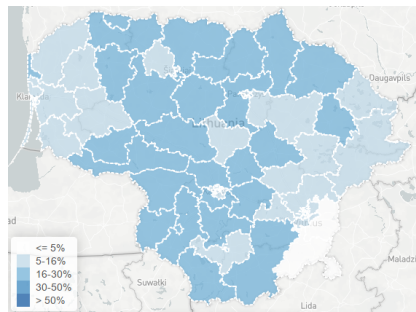
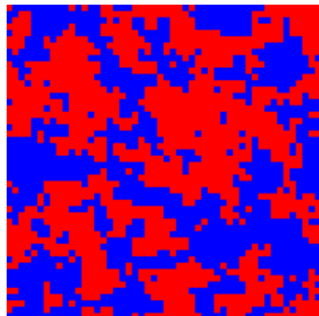
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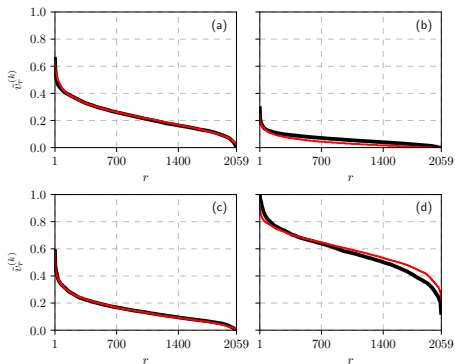
The core idea



- The models: single compartment, multiple points in time.
- The data: multiple compartments, single point in time.

Image sources: (left) <http://rf.mokslasplius.lt/voter-model/>, (right) <http://rinkimurezultatai.lt>

Simplest solution: independent compartments



Rank-size distribution of the vote shares for 4 main parties in the Lithuanian Seimas 1992 elections: model (red) vs data (black).

RSDs are used for observations over compartments, PDFs are used for observations over time.

Figure: Kononovicius, Complexity 2017: 7354642 (2017)

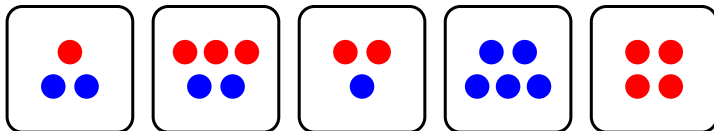
Other: (Sano *et al.*, 2017), (Braha & de Aguiar, 2017), (Fenner *et al.*, 2017)

Compartmental voter model

Static setup

- Consider N agents of T types.
- Let agent types be fixed.
- Let agents move between M compartments of capacity C .

$N=20, T=2, M=5, C=5$



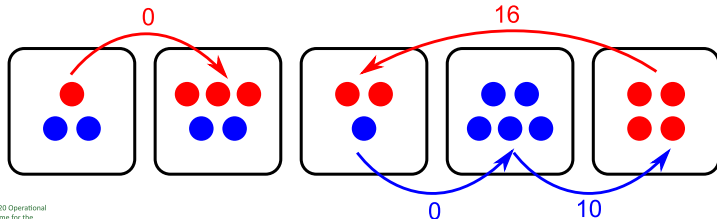
Movement dynamics

Consider movement rates (from i to j for type k):

$$\lambda_{(k)}^{i \rightarrow j} = \begin{cases} X_i^{(k)} \left(\varepsilon^{(k)} + X_j^{(k)} \right) & \text{if } i \neq j \text{ and } N_j < C, \\ 0 & \text{otherwise,} \end{cases}$$

here $X_i^{(k)}$ is the number of agents of type k in compartment i , while N_j is the total number of agents in compartment j .

$N=20, T=2, M=5, C=5, \varepsilon=2$



Infinite capacity ($C = N$)

Stationary distribution

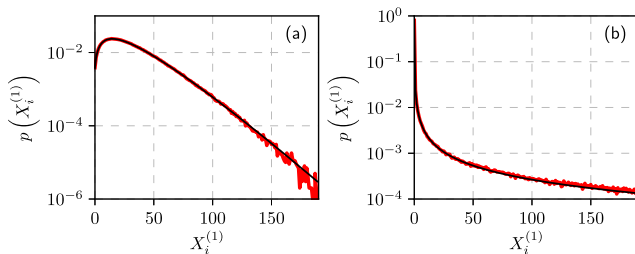
Given $C = N$ one can write closed form expression for the total entry and exit rates:

$$\lambda_{(k)}^{i+} = \sum_{j=1}^M \lambda_{(k)}^{j \rightarrow i} = \left[N^{(k)} - X_i^{(k)} \right] \left(\varepsilon^{(k)} + X_i^{(k)} \right),$$

$$\lambda_{(k)}^{i-} = \sum_{j=1}^M \lambda_{(k)}^{i \rightarrow j} = X_i^{(k)} \left([M - 1] \varepsilon^{(k)} + \left[N^{(k)} - X_i^{(k)} \right] \right).$$

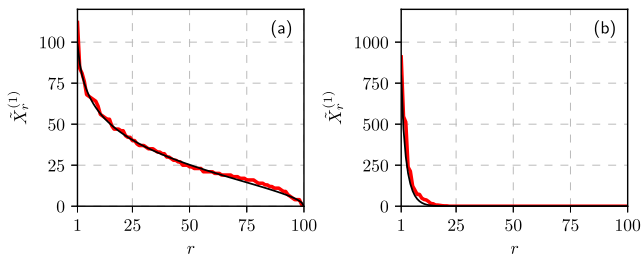
Which means that $X_i^{(k)}$ is distributed according to Beta-binomial distribution in this particular case.

Numerical verification of stationary distribution



Model (red curves): $N = 3000$, $T = 1$, $M = 100$ and $C = N$ (all cases), $\varepsilon^{(1)} = 2$ (a) and 0.03 (b). Beta-binomial fit (black curves): $N = 3000$, $\alpha = \varepsilon^{(1)}$ and $\beta = (M - 1)\varepsilon^{(1)}$ (all cases).

Numerical inquiry into rank–size distribution



Same simulation as in the previous slide: $N = 3000$, $T = 1$, $M = 100$ and $C = N$ (all cases), $\varepsilon^{(1)} = 2$ (a) and 0.03 (b).

Beta–binomial distribution provides a rather good fit as if compartments would be truly independent.

Finite capacity $C < N$

Obtaining stationary distribution

General idea

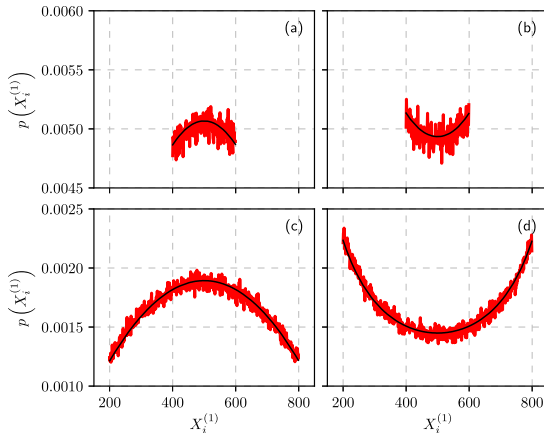
- In general we don't know close forms of the total entry and exit rates.
- Notice that the compartmental model is actually a multivariate finite-state Markov chain.
- In theory **we can reduce any finite-state Markov chain to one-dimensional Markov chain** by relabeling states.

Obtaining stationary distribution II

Technical implementation

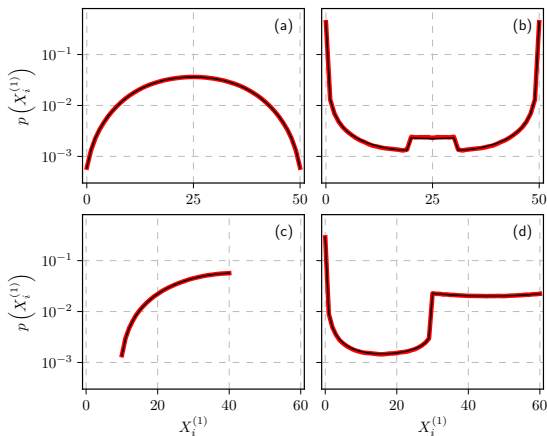
- Given $N = 10$, $T = 2$, $M = 2$, $C = 6$ and ε .
- The state vector is $\{X_1^{(1)}, X_1^{(2)}\}$.
- List all possible states: $\{0, 4\}, \dots, \{5, 1\}$.
- Relabel states: $\{0, 4\} \rightarrow 0, \{0, 5\} \rightarrow 1, \dots$
- Write down the transition matrix and find its eigensystem.
- Aggregate over eligible states to get the desired result:
$$P(X_1^{(1)} = 0) = P(\{0, 4\}) + P(\{0, 5\}) + P(\{0, 6\}).$$

Univariate MC: $T = 1, M = 2$



Model (red curves): $N = 1000$ (all cases), $C = 600$ ((a) and (b)) and 800 ((c) and (d)), $\varepsilon^{(1)} = 2$ ((a) and (c)) and 0.03 ((b) and (d)). Truncated Beta-binomial distribution (black curves): $N = 1000$, $\alpha = \varepsilon$, $\beta = (M - 1)\varepsilon$ (all cases).

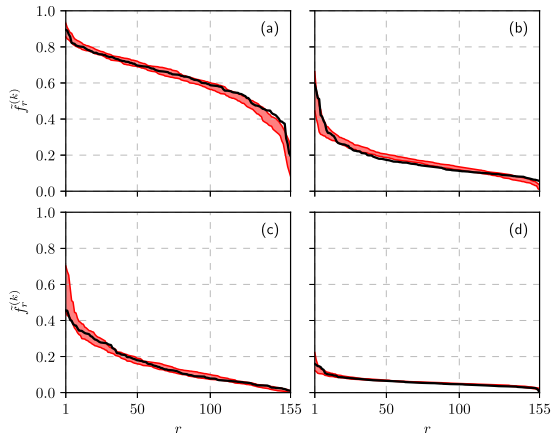
Simplest multivariate MCs



Model (red curves): $N = 100$, $M = 2$ and $T = 2$ ((a) and (b)), $N = 90$, $M = 3$ and $T = 1$ ((c) and (d)), $C = 40$ (c), 60 ((a) and (d)) and 80 (b), $\varepsilon = 2$ ((a) and (c)) and 0.03 ((b) and (d)). Fits (black curves) were obtained from the eigensystem.

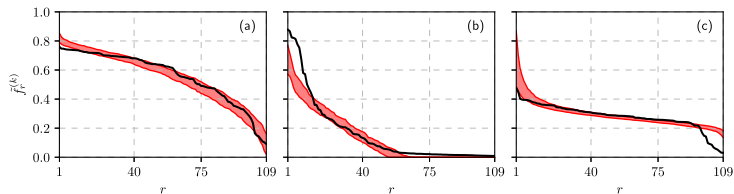
Empirical examples

Ethnic groups in London (UK census 2011)



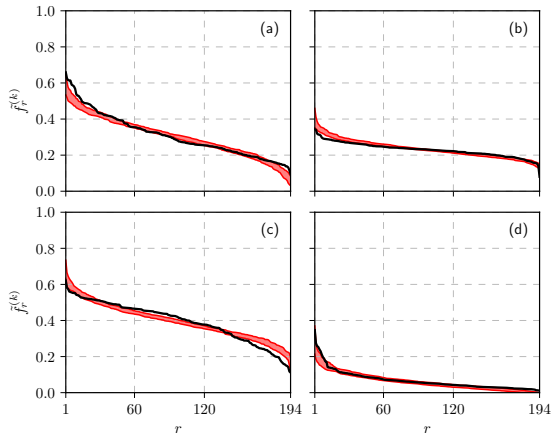
Considered groups (black curves): (a) White, (b) Asian, (c) Black, (d) other.
Model (red areas): $N^{(w)} = 48515$, $N^{(a)} = 12865$, $N^{(b)} = 11470$ and $N^{(o)} = 4495$ ($N = 77345$), $\varepsilon^{(w)} = 2.5$, $\varepsilon^{(a)} = 4$, $\varepsilon^{(b)} = 1.5$, $\varepsilon^{(o)} = 15$, $M = 155$,
 $C = 600$.

Religious groups in Leicester (UK census 2011)



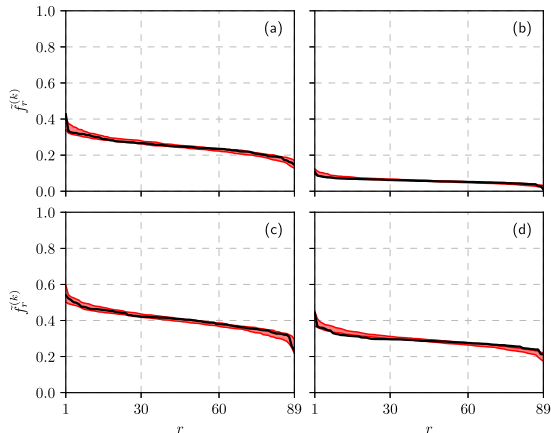
Considered groups (black curves): (a) Christians, (b) no religion, (c) other.
Model (red areas): $N^{(c)} = 30411$, $N^{(n)} = 8829$ and $N^{(o)} = 15151$ ($N = 54391$),
 $\varepsilon^{(c)} = 2.5$, $\varepsilon^{(n)} = 0.01$, $\varepsilon^{(o)} = 50$, $M = 109$, $C = 600$.

Working class in Sheffield (UK census 2011)



Considered groups (black curves): (a) higher, (b) intermediate and (c) lower occupations, (d) unemployed. Model (red areas): $N^{(1)} = 29876$, $N^{(2)} = 22310$, $N^{(3)} = 38218$ and $N^{(u)} = 6596$ ($N = 97000$), $\varepsilon^{(1)} = 3$, $\varepsilon^{(2)} = 50$, $\varepsilon^{(3)} = 12$, $\varepsilon^{(u)} = 2$, $M = 194$, $C = 600$.

Vote shares in Vilnius (LT Seimas election 1992)



Considered groups (black curves): (a) Sajudzio koalicija, (b) Lietuvos krikscioniu demokartu partija, (c) Lietuvos demokratine darbo partija, (d) other. Model (red areas): $N^{(s)} = 11125$, $N^{(l)} = 2581$, $N^{(d)} = 17978$ and $N^{(o)} = 12816$ ($N = 44500$), $\varepsilon^{(s)} = \varepsilon^{(l)} = \varepsilon^{(d)} = 25$, $\varepsilon^{(o)} = 75$, $M = 89$, $C = 600$.

Compartmental voter model

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Abstract

Numerous models in opinion dynamics focus on the temporal dynamics within a single spatial unit (e.g., country). While the opinions are often observed across multiple spatial units (e.g., polling stations) at a single point in time (e.g., elections). Aggregates of these observations, while quite useful in many applications, neglect the underlying spatial heterogeneity in opinions. To address this issue we build a simple compartmental agent-based model in which all agents have fixed opinions, but are able to change their compartments. We demonstrate that this model is able to generate compartmental rank-size distributions consistent with the empirical data.

1 Introduction

Most well-known models of opinion dynamics seem to imply that a steady state, either consensus or polarization, is inevitable [1-7]. However, local and spatial heterogeneity and ongoing exchange of opinions and cultural traits is a characterizing feature of social systems. Variety modifications of the well-known models were proposed to account for these features, such as inflexibility [8,9] or spontaneous flipping [10,11]. Some of the models were modified to account for the theories from the social sciences [12,13]. Effects of these modifications are still being actively reconsidered in context of network theory, non-linearity, complex contagion and applications towards financial markets [14-21]. Nevertheless even these modified models assume that opinion dynamics occur and

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Thank you!



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