Non–extensive voter model for socio–economic phenomena

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Complex Physical and Social Systems Group members and a collaborator



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Physics of Risk About Topics Students Contribut

Wealth distribution in Talent vs Luck model

🞬 September 24, 2019 😫 Aleksejus Kononovicius 🛛 #Interactive models 🖉 #Agent-based models 🖉 #vealth 🖉 #ulent vs back

Here we continue our discussion on the Talent vs Luck model proposed by [1]. This time we show you the reason why this model is worthy of comparison to the **kinetic exchange models**. This model is capable of reproducing power-law wealth (capital in this model) distribution.

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The Math PhD Turning Basketball Into a Science

September 17, 2019 Aleksejus Kononovicius Invideo Instatistics Insports Indoorderg

In this video Bloomberg (renovmed source for news on various topics related to economics and business) introduces furturistic job sports data scientist. Actually this job is not a futuristic and there are some people who already do I. In this video Yuana Seric (PhD in Math and ex-basketball player) describes what she does everyday to help Philadelphia 76ers to gain dege against their opconests in the NBA.





COST P10 meeting in Vilnius

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Blogging since 2006: \sim 100 interactive models, \sim 170 texts w/o models.



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Non-extensive VM

http://rf.mokslasplius.lt

What this talk is about?

- Real life socio–economic systems are finite, yet their statistical properties are size (*N*) independent.
- Agent–based models are also finite, yet often the statistical properties of ABMs are *N*-dependent.
- Extensive models exhibit no fluctuations or only if *N* is finite.
- Non-extensive models exhibit fluctuations even if $N \to \infty$.

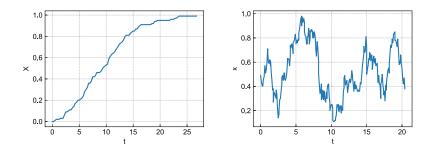


Agents we will (left) and will not (right) talk about.

Note: I use terms "extensive" and "non-extensive" rather loosely.



Extensive vs non-extensive



Typical time series of an extensive model (left) and a non-extensive model (right).

Figs.: http://rf.mokslasplius.lt/unidirectional-kirman-model/, http://rf.mokslasplius.lt/kirman-ants/.





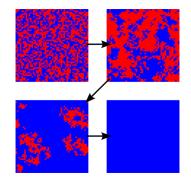
Non-extensive VM

Extensive and non-extensive voter models



2014-2020 Operational Programme for the European Union Funds Investments in Lithuania Originally defined similarly to a cellular automaton:

- Agents are the cells of a two dimensional grid.
- Each agent is in one of the two states: +1 or -1.
- During each time step:
 - agent (A) is selected,
 - its neighbor (B) is selected,
 - A copies the state of B.



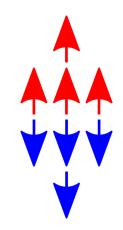
Consensus emerges in the classic voter model.

Collage from: http://rf.mokslasplius.lt/voter-model/



The Ising model is quite similar in formulation:

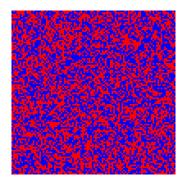
- Particles are placed on a grid.
- Each particle has a spin of +1 or -1.
- During each time step:
 - a particle is selected,
 - it changes the state to align with its neighbors
 - or flips randomly if there is no clear majority.



This state is stable for the Ising model, but unstable for the voter model.



- Social systems rarely reach a full consensus.
- To break the consensus lets include "thermal" noise. Namely, lets change our rule to:
 - During each time step:
 - agent (A) is selected,
 - with p_r A flips his state,
 - otherwise its neighbor (B) is selected
 - and A copies the state of B.



No consensus $(p_r = 0.05)$.



- During each time step at most one spin changes.
- Lets use mean-field approximation,
- to get probabilities of the two possible changes:

$$P(X_{+} \to X_{+} + 1) = \frac{X_{-}}{N} \left[p_{r} + (1 - p_{r}) \frac{X_{+}}{N} \right] = (1 - x) \left[p_{r} + (1 - p_{r}) x \right],$$

$$P(X_{+} \to X_{+} - 1) = \frac{X_{+}}{N} \left[p_{r} + (1 - p_{r}) \frac{X_{-}}{N} \right] = x \left[p_{r} + (1 - p_{r})(1 - x) \right],$$

here
$$x = \frac{X_+}{N}$$
.



Continuous limit of the noisy voter model

In continuous limit, any discrete one-step process of the following form:

$$P(X_+ \to X_+ \pm 1) = \lambda^{\pm} \Delta t,$$

is well approximated by the following stochastic differential equation:

$$\mathrm{d}x = \frac{\lambda^+ - \lambda^-}{N} \mathrm{d}t + \sqrt{\frac{\lambda^+ + \lambda^-}{N^2}} \mathrm{d}W.$$

For the noisy voter model (assuming $\Delta t = N^{-1}$), we get:

$$dx = p_r(1 - 2x)dt + \sqrt{\frac{1}{N} \left[2x(1 - x)(1 - p_r) + p_r \right]} dW.$$

Thus it is an **extensive model**.



Bass diffusion model

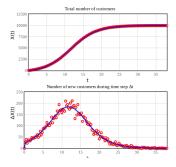
as a special case of the extensive voter model

- We consider consumers (+1) and potential consumers (-1) of a durable good.
- Forbid $+1 \rightarrow -1$:

$$P(X_+ \to X_+ - 1) = 0.$$

• Then in continuous limit:

$$\mathrm{d}x \approx (1-x) \left[p_r + (1-p_r) \mathbf{x} \right] \mathrm{d}t.$$



"Unidirectional" voter model vs Bass diffusion ODE.

Fig.: http://rf.mokslasplius.lt/unidirectional-kirman-model/

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- We need to make the imitation process N times more active.
- We can do it in framework we have been using until now, but the change is somewhat hard to conceptualize.
- So lets change the framework! Instead of probabilities lets consider event rates.
- We have only two possible events: generation (birth) and recombination (death) in respect to *X*₊.
- Rates must be positive, but otherwise are unconstrained, so:

$$\lambda^{+} = (N - X_{+}) \left[\varepsilon_{+} + X_{+} \right], \quad \lambda^{-} = X_{+} \left[\varepsilon_{-} + (N - X_{+}) \right].$$



Gillespie method (in general):

- Draw inter–event time τ_i from exponential distribution using total event rate λ .
- Update clock: $t_{i+1} = t_i + \tau_i$.
- With $p^{(k)} = \frac{\lambda^{(k)}}{\lambda}$ execute event k.
- Update total event rate $\lambda = \sum_k \lambda^{(k)}$.

In voter models we have just two events:

- with rate λ^+ we increment X_+ ,
- with rate λ^- we decrement X_+ .



Continuous limit for the non-extensive voter model

We know that:

$$\mathrm{d}x = \frac{\lambda^+ - \lambda^-}{N} \mathrm{d}t + \sqrt{\frac{\lambda^+ + \lambda^-}{N^2}} \mathrm{d}W.$$

For the non-extensive model we have:

$$\mathrm{d}x \approx \left[\varepsilon_+ \left(1 - x\right) - \varepsilon_- x\right] \mathrm{d}t + \sqrt{2x(1 - x)} \mathrm{d}W.$$

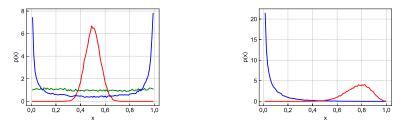
From this it can be easily shown that:

$$x \sim \mathcal{B}e(\varepsilon_+, \varepsilon_-)$$
.

The non-extensive voter model is also known as the Kirman model.

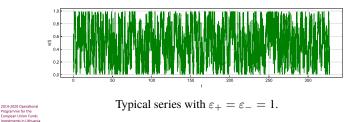


Lets see numerically



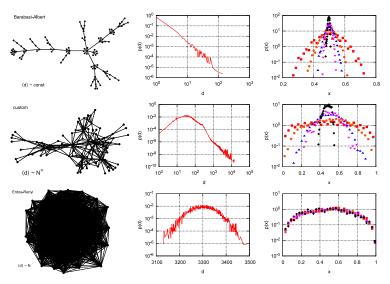
$$\varepsilon_{+} = \varepsilon_{-} = \{0.01, 1, 100\}$$





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Topology does matter!





Kononovicius & Ruseckas, EPJ B 87: 169 (2014).

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Non-extensive VM

Financial markets



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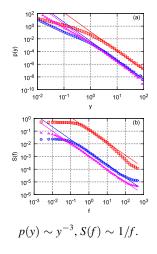
Fat-tails and long-range memory

- Let the two states represent chartist and fundamentalist traders.
- Let Walrass equilibrium hold.
- Then ignoring short-term fluctuations, the return is driven by:

$$y = \frac{x}{1-x}.$$

• Which is described by the following SDE:

$$\mathrm{d}y \approx \left(2 - \varepsilon_{cf}\right) y^{2+\alpha} \mathrm{d}t_s + \sqrt{2y^{3+\alpha}} \mathrm{d}W.$$



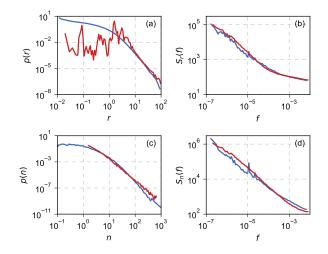
Kononovicius & Gontis, Physica A 391: 1309-1314 (2012); and many later works



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Latest paper: Combined order-book model



BTC stats (red) vs model stats (blue).



Kononovicius & Ruseckas, Physica A 525: 171-191 (2019); relies on: Kanazawa et al., PRL 120: 138301 (2018).

Non-extensive VM

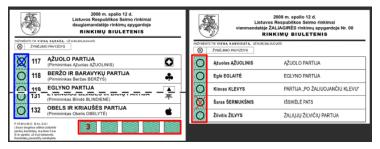
Lithuanian parliamentary elections

Based on: Kononovicius, Complexity 2017: 7354642 (2017).



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Example ballots



We are interested in political party performance (blue).

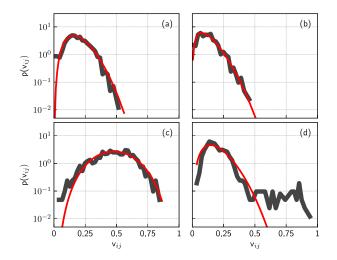
- Two-tier system.
- Held each 4 years.
- Polling station level data (~ 2000 data points).

Fig.: Central Electoral Commission



Data sources: CEC (full, in Lithuanian), GitHub (partial, in English)

PDFs over polling stations (1992)

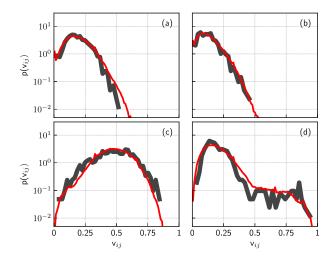


Three main parties ((a)-(c)) and other parties (d): empirical data (gray) vs Beta-fit (red).



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Multi-state non-extensive voter model (1992)



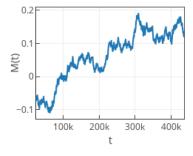
Three main parties ((a)-(c)) and other parties (d): empirical data (gray) vs the model

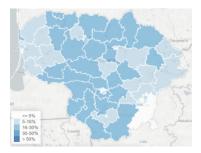


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What is wrong with this approach?

- Models are temporal, space is abstracted away.
- Data is spatial, and observe rarely in time.





Figs.: http://rf.mokslasplius.lt/q-voter-model/, http://rinkimurezultatai.lt

Independence assumption made in: (Kononovicius, 2017), (Sano *et al.*, 2017), (Braha & de Aguiar, 2017), (Fenner *et al.*, 2017) requestions for the Comparison of the co

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But what if spatial units are inter-dependent?

Complicated approach building on commuting patterns

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Fernandez-Garcia et al., PRL 112; 158701 (2014)

Compartmental voter model

The voter model, which is not a model for voters

Based on: Kononovicius, accepted to J. Stat. Mech., arXiv: 1906.01842 [physics.soc-ph]



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Socio-demographics impact voting behavior

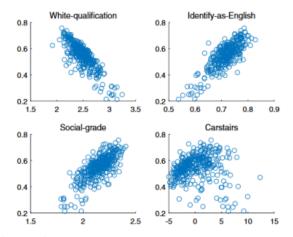


Fig. 2. (Color online) Scatter plots for the four chosen covariates; the y-values represent the proportion of Leave votes for a district and the x-values represent the values of the covariate for the district.



Fenner et al., Int J Mod Phys C 28: 1750132 (2017)

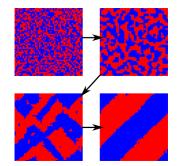
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During each time step:

- pick particle (A),
- pick its neighbor (B),
- A and B switch places based on how likely both states are according to the Boltzmann distribution.

In physical sense we are modeling transport phenomena and ignoring magnetization itself.



Kawasaki dynamics in Ising model with $T \ll T_c$.

Fig.: unpublished post on Physics of Risk

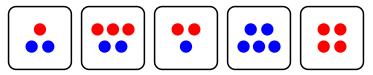
Kawasaki, Phys. Rev. 145 (1966) and Phys. Rev. 148 (1966)



Statics of the compartmental voter model

- Let there be *N* agents (people).
- Let each agent be one of T types (splits in socio-demographic category).
- Let those types be fixed.
- Let agents reside in *M* compartments (any real spatial unit).
- Let capacity of each district be C.

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





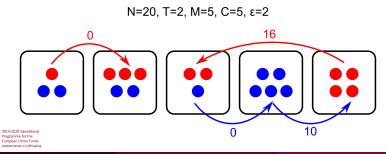
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Dynamics in the compartmental voter model

Let the migration rate between districts be $(i \rightarrow j \text{ for type } k)$:

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

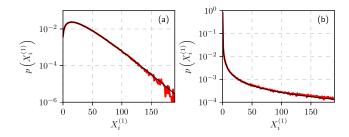
in the above $X_i^{(k)}$ is the number of type k agents in district *i*, $\varepsilon^{(k)}$ is individual transition rate for type k, N_j is total number of agents in district *j*.



- If capacity is infinite C = N, we know closed form expressions for the total entry/exit rates. Thus we can get closed form expression for the stationary distribution of $X_i^{(k)}$ for fixed *i* and *k*.
- If capacity is finite, we have to use detailed balance to get the stationary distribution. This works, but scales badly.
- The problem is that we are more interested in the compartmental distribution of $f_i^{(k)} = X_i^{(k)}/N_i$. This is likely impossible.



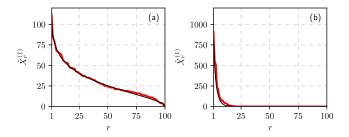
PDF of $X_i^{(k)}$ for the infinite capacity



Model (red) vs Beta-fit (black): N = 3000, T = 1, M = 100 and C = N (both), $\varepsilon^{(1)} = 2$ (a) and 0.03 (b).



CRSD of $X_i^{(k)}$ for the infinite capacity

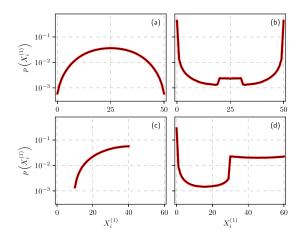


Model (red) vs Beta-fit (black): N = 3000, T = 1, M = 100 and C = N (both), $\varepsilon^{(1)} = 2$ (a) and 0.03 (b).

CRSD - abbr. compartmental rank-size distribution at fixed time.



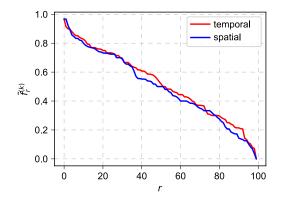
PDF of $X_i^{(k)}$ for the finite capacity



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)).



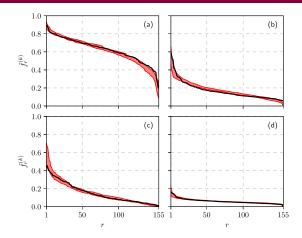
Spatio-temporal (RSD) symmetry



N = 2600, T = 2, M = 100, C = 30 and $\varepsilon = 2$



UK census 2011: Ethnic groups in London



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.

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Red areas show 95% confidence intervals for the model.

Summary



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Some key ideas

- Voter models are not only for the voters.
- Non-extensive voter model can quite nicely reflect social heterogeneity.
- You can remember, even if you have no memory. Non–linearity does the job for you.
- Typical opinion dynamics models correspond to Glauber dynamics, the data is more alike Kawasaki dynamics, but this difference doesn't seem to matter much.
- Collectively we are all curious spinsons.





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