## Dynamics of Heterogeneous Voter Models

Sid Redner (physics.bu.edu/~redner)
Nonlinear Dynamics of Networks UMD April 5-9, 2010
T. Antal (BU $\rightarrow$ Harvard), N. Gibert (ENSTA), N. Masuda (Tokyo),
M. Mobilia (BU $\rightarrow$ Leeds), V. Sood (BU $\rightarrow$ NBI), D. Volovik (BU) NSF DMR0535503 \& DMR0906504

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The classic voter model
3 basic results
Voting on complex networks T.Antal, V. Sood new conservation law
two time-scale route to consensus short consensus time

Strategic voting (>2 states) long time-scale switching
Partisan voting selfishness vs. collectiveness ultraslow evolution
M. Mobilia, D.Volovik
N. Masuda, N. Gibert

# Classic Voter Model Clifford \& Sudury (1973) Holley \& Liggett (1975) 



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proportional rule

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I. Pick a random voter
2.Assume state of randomly-selected neighbor individual has no self-confidence \& adopts neighbor's state
3. Repeat I \& 2 until consensus necessarily occurs in a finite system

## Voter Model Evolution Dorric etal. (2001)

random initial condition:


$$
t=4 \quad t=16 \quad t=64 \quad t=256
$$

droplet initial condition:


## Lattice Voter Model: 3 Basic Properties

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I. Final State (Exit) Probability $\mathcal{E}\left(\rho_{0}\right)$

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## 3. Consensus Time

| dimension | consensus time |
| :---: | :---: |
| 1 | $\mathrm{~N}^{2}$ |
| 2 | $\mathrm{~N} \ln \mathrm{~N}$ |
| $>2$ | N |

## Voter Model on Complex Networks

C. Castellano, D.Vilon, A.Vespignani, EPL 63, I 53 (2003)
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illustrative example: complete bipartite graph


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$$
\begin{aligned}
& \begin{array}{c}
\text { pick site on } \\
\text { a sublattice } \\
\mathbb{y}
\end{array} \\
d N_{a} & =\frac{a}{a+b}\left[\frac{a-N_{a}}{a} \frac{N_{b}}{b}-\frac{N_{a}}{a} \frac{b-N_{b}}{b}\right] \\
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Subgraph densities: $\rho_{a}=N_{a} / a, \rho_{b}=N_{b} / b \quad d t=1 /(a+b)$

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\begin{aligned}
\rho_{a, b}(t) & =\frac{1}{2}\left[\rho_{a, b}(0)-\rho_{b, a}(0)\right] e^{-2 t}+\frac{1}{2}\left[\rho_{a}(0)+\rho_{b}(0)\right] \\
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& \rightarrow \frac{1}{2}\left[\rho_{a}(0)+\rho_{b}(0)\right] \quad \text { magnetization not conserved }
\end{aligned}
$$

## Voter Model on Complex Networks

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## Voter Model on Complex Networks


"flow" from high degree to low degree

## New Conservation Law

low degree


## New Conservation Law


to compensate different rates, consider: $\begin{aligned} & \text { degree-weighted } \\ & \text { Ist moment: }\end{aligned} \quad \omega=\frac{1}{\mu_{1}} \sum_{k} k n_{k} \rho_{k}$

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\begin{aligned}
& \mu_{1}=\text { av. degree } \\
& n_{k}=\text { frac. nodes of degree } k \\
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Exit Probability on Complex Graphs

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\mathcal{E}(\omega)=\omega
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Extreme case: star graph N nodes: degree I I node: degree N


## Exit Probability on Complex Graphs

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\mathcal{E}(\omega)=\omega
$$

Extreme case: star graph 0


Final state: all I with prob. I/2!

Route to Consensus on Complex Graphs

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complete bipartite graph


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## Consensus Time Evolution Equation

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warmup: complete graph
$T(\rho) \equiv$ av. consensus time starting with density $\rho$

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\begin{aligned}
T(\rho)= & \mathcal{R}(\rho)[T(\rho+d \rho)+d t] \\
& +\mathcal{L}(\rho)[T(\rho-d \rho)+d t] \\
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& \\
& L \\
0 & \rho \\
& 1-\boldsymbol{R}-\boldsymbol{L}
\end{aligned} \quad \begin{aligned}
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## Consensus Time on Complete Graph

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continuum limit:

$$
T^{\prime \prime}=-\frac{N}{\rho(1-\rho)}
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\end{aligned}
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continuum limit:

$$
T^{\prime \prime}=-\frac{N}{\rho(1-\rho)}
$$

solution:

$$
T(\rho)=-N[\rho \ln \rho+(1-\rho) \ln (1-\rho)]
$$

## Consensus Time on Heterogeneous Networks

$T\left(\left\{\rho_{k}\right\}\right) \equiv$ av. consensus time starting with density $\rho_{k}$ on nodes of degree $k$

$$
\begin{aligned}
T\left(\left\{\rho_{k}\right\}\right)= & \sum_{k} \mathcal{R}_{k}\left(\left\{\rho_{k}\right\}\right)\left[T\left(\left\{\rho_{k}^{+}\right\}\right)+d t\right] \\
& +\sum_{k} \mathcal{L}_{k}\left(\left\{\rho_{k}\right\}\right)\left[T\left(\left\{\rho_{k}^{-}\right\}\right)+d t\right] \\
& +\left[1-\sum_{k}\left[\mathcal{R}_{k}\left(\left\{\rho_{k}\right\}\right)+\mathcal{L}_{k}\left(\left\{\rho_{k}\right\}\right)\right]\right]\left[T\left(\left\{\rho_{k}\right\}\right)+d t\right] \\
\mathcal{R}_{k}\left(\left\{\rho_{k}\right\}\right) & =\operatorname{prob}\left(\rho_{k} \rightarrow \rho_{k}^{+}\right) \quad \mathcal{L}_{k}\left(\left\{\rho_{k}\right\}\right)=n_{k} \rho_{k}(1-\omega) \\
& =\frac{1}{N} \sum_{x}^{\prime} \frac{1}{k_{x}} \sum_{y} P(\downarrow,-\uparrow) \\
& =n_{k} \omega\left(1-\rho_{k}\right)
\end{aligned}
$$

## Consensus Time on Heterogeneous Networks

continuum limit:

$$
\sum_{k}\left[\left(\omega-\rho_{k}\right) \frac{\partial T}{\partial \rho_{k}}+\frac{\omega+\rho_{k}-2 \omega \rho_{k}}{2 N n_{k}} \frac{\partial^{2} T}{\partial \rho_{k}^{2}}\right]=-1
$$

## Molloy-Reed Scale-Free Network



## Consensus Time on Heterogeneous Networks

continuum limit:

$$
\sum_{k}\left[\left(\omega-\rho_{k}\right) \frac{\partial T}{\partial \rho_{k}}+\frac{\omega+\rho_{k}-2 \omega \rho_{k}}{2 N n_{k}} \frac{\partial^{2} T}{\partial \rho_{k}^{2}}\right]=-1
$$

now use $\quad \rho_{k} \rightarrow \omega \quad \forall k$
and

$$
\frac{\partial}{\partial \rho_{k}}=\frac{\partial \omega}{\partial \rho_{k}} \frac{\partial}{\partial \omega}=\frac{k n_{k}}{\mu_{1}} \frac{\partial}{\partial \omega}
$$

## Consensus Time on Heterogeneous Networks

continuum limit:

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\sum_{k}\left[\left(\omega-\rho_{k}\right) \frac{\partial T}{\partial \rho_{k}}+\frac{\omega+\rho_{k}-2 \omega \rho_{k}}{2 N n_{k}} \frac{\partial^{2} T}{\partial \rho_{k}^{2}}\right]=-1
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$$

to give

$$
\frac{\partial^{2} T}{\partial \omega^{2}}=-\frac{N \mu_{1}^{2} / \mu_{2}}{\omega(1-\omega)}
$$

## Consensus Time on Heterogeneous Networks

continuum limit:

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\sum_{k}\left[\left(\omega-\rho_{k}\right) \frac{\partial T}{\partial \rho_{k}}+\frac{\omega+\rho_{k}-2 \omega \rho_{k}}{2 N n_{k}} \frac{\partial^{2} T}{\partial \rho_{k}^{2}}\right]=-1
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$$

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$$
\frac{\partial^{2} T}{\partial \omega^{2}}=-\frac{N \mu_{1}^{2} / \mu_{2}}{\omega(1-\omega)} \quad \text { as } \quad \text { ase } \quad T^{\prime \prime}=-\frac{N}{\rho(1-\rho)}
$$

with effective size $N_{\text {eff }}=N \mu_{1}^{2} / \mu_{2}$

## Consensus Time for Power-Law Degree

 Distribution $n_{k} \sim k^{-\nu}$$$
T_{N} \propto N_{\mathrm{eff}}=N \frac{\mu_{1}^{2}}{\mu_{2}} \sim\left\{\begin{array}{ll}
N & \nu>3 \\
N / \ln N & \nu=3 \\
N^{2(\nu-2) /(\nu-1)} & 2<\nu<3 \\
(\ln N)^{2} & \nu=2 \\
\mathcal{O}(1) & \nu<2
\end{array}\right]
$$

## Strategic Voting



## Strategic Voter Model $\begin{aligned} & \text { D.Volovik, M. Mobilia, SR } \\ & \text { EPL } 85,48001(2009)\end{aligned}$

randomly-selected voter changes to any other state equiprobably (rate T)
majority-minority interaction: minority preferentially changes to majority (rate r)
rate equations ( $\mathrm{A}, \mathrm{B}$ majority; c minority):

$$
\begin{aligned}
\dot{A} & =T(B+c-2 A)+r A c \\
\dot{B} & =T(c+A-2 B)+r B c \\
\dot{c} & =T(A+B-2 c)-r(A+B) c
\end{aligned}
$$

## Phase Portrait



## Phase Portrait



## Slow Switching




## Partisan Voter Model N.Masad. . . Gbers. SR arXiv:I003.0768

## Partisan Voter Model N.Masuda, N. Gibert, SR arXiv:I003.0768



## Partisan Voter Model N.Masuda, N. Gibert, SR arXiv:I003.0768

个happy
democrat

density $D_{h}$ \begin{tabular}{l}
sad <br>
democrat <br>
density $D_{s}$

$\quad$

个ad <br>
republican <br>
density $R_{s}$

 

$\downarrow$ happy <br>
republican <br>
density $R_{h}$
\end{tabular}

partisan voting update:

## Partisan Voter Model N.Masuda, N, Gibert, SR

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partisan voting update:
I. Pick voter, pick neighbor (as in usual voter model);

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2a. If initial voter becomes happy by adopting neighboring state, change occurs with rate $I+\varepsilon$;

$$
\uparrow \downarrow \underset{1+\varepsilon}{\rightarrow} \downarrow \downarrow
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$$
\uparrow \downarrow \underset{\jmath+\varepsilon}{\rightarrow} \downarrow \downarrow
$$

2b. If initial voter becomes unhappy by adopting neighboring state, change occurs with rate $I-\varepsilon$.

$$
\downarrow \uparrow \rightarrow \uparrow \uparrow
$$

## Partisan Voter Model: Mean-Field Limit

rate equations:

$$
\begin{gathered}
\dot{D}_{h}=2 \epsilon D_{h} D_{s}+(1+\epsilon) D_{s} R_{s}-(1-\epsilon) D_{h} R_{h} \\
\dot{D}_{s}=-2 \epsilon D_{h} D_{s}+(1-\epsilon) D_{h} R_{h}-(1+\epsilon) D_{s} R_{s} \\
\text { and } R \leftrightarrow D
\end{gathered}
$$

Symmetric Case: $D=R=1 / 2$
$H \equiv D_{h}+R_{h}$
$=$ density of happy voters
$\Delta \equiv D_{h}-R_{h}=D_{h}-\left(\frac{1}{2}-R_{s}\right)=\rho-\frac{1}{2}$
$=$ density democratic voters $-\frac{1}{2}$


## Consensus Time on Finite Graphs



## Summary \& Outlook

## Voter model:

paradigmatic, soluble, (but hopelessly naive)
Voter model on complex networks:
new conservation law
meandering route to consensus
fast consensus for broad degree distributions

## Extensions:

strategic voting $\rightarrow$ minority suppressed partisan voting $\rightarrow$ selfishness forestalls consensus

Future:
"churn" rather than consensus
heterogeneity of real people
positive and negative social interactions $\rightarrow$ social balance

## Crass Commercialism

Aimed at graduate students, this book explores some of the core phenomena in non-equilibrium statistical physics. It focuses on the development and application of theoretical methods to help students develop their problem-solving skills.
The book begins with microscopic transport processes: diffusion, collision-driven phenomena, and exclusion. It then presents the kinetics of aggregation, fragmentation and adsorption, where the basic phenomenology and solution techniques are emphasized. The
following chapters cover kinetic spin systems, both from a discrete following hapers cove kives spin systris, and a continuum perspective; the role of disorder in non-
perspective; the kinetics of chemical reactions; and the properties of perspective; the $k$. The networks. The book contains 200 exercises to test students' complex networks. The book contains 200 exercises to test stude authors, containing supplementary material including solutions to some of the exercises, can be found at
www.cambridge.org/9780521851039.
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Cover illustration: Snapshot of a collision cascade in a perfectly elastic hardsphere gas in two dimensions due to a singli incident particle. Shown are the cloud of moving particles (Ted) and the stationary particles (blue) that have not
yet experienced any collisions. Figure courtes of Tibor Antal.

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