

# Kinetic equations modelling complex systems in economics and sociology

Bertram Düring

Department of Mathematics



University of Sussex

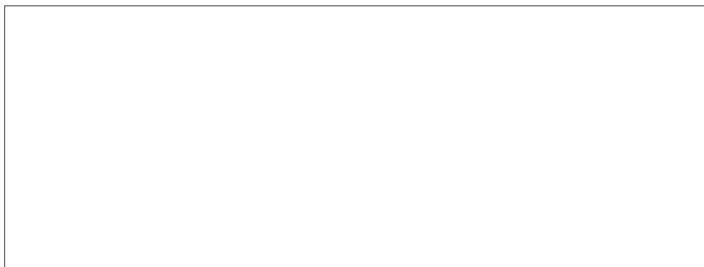
Joint work with

P. Markowich, D. Matthes, J.-F. Pietschmann,  
G. Toscani, M.-T. Wolfram

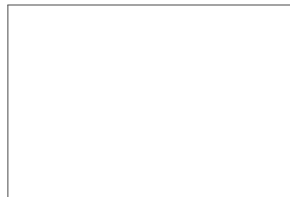
# Wealth distribution in an economy

**Wealth:** total value of a person's net assets (money, shares, property, buildings, . . . ) from which any debts are subtracted

**Exhibit:** Income distribution in the USA



- ▶ upper class ( $\approx 1\%$ ) well fitted by inverse power law (**Pareto tail**)
- ▶ important implications for **taxation, social peace, growth**



# Opinion formation

- ▶ How are political opinions formed?
- ▶ Sociology tells us that **interpersonal communication** and **opinion leaders** play a big role

- ▶ not limited to politics, but more general decision making: product choice, dissemination of new technologies,...

# Complex systems in economics and sociology

Examples:

- ▶ distribution of wealth in a society
- ▶ opinion formation (political opinions, product choice, ...)

Common features:

- ▶ large number of interacting agents (agent-based models?)
- ▶ model of full system not tractable
- ▶ quantities of interest are aggregates
- ▶ dynamics!
- ▶ emergent behaviour

↪ mathematical tools from kinetic theory/statistical mechanics

# From micro to macro

Conceptual approach:

- ▶ **continuous** independent **variables**:  
wealth/price/opinion and time
- ▶ describe dynamics of system by **microscopic interactions** among agents
- ▶ perform many interactions (analytically or numerically)
- ▶ observe emergent behaviour, patterns in **macroscopic distribution** of agents
- ▶ derive **partial differential equations** which (approximatively) govern the time-evolution of the density

Benefits:

- ▶ more (analytically and numerically) tractable model
- ▶ understanding role of parameters in the microscopic interactions for emergent behaviour
- ▶ PDE: nonlinear, anisotropic, nonlocal, degenerate

# Wealth distribution: quantity of interest

- ▶ We are interested in the **large-wealth behavior** of the distribution since it determines *a posteriori* if the model results fits real data
- ▶ let  $f(v)$  be probability density function of agents with wealth  $v$ , and consider the commulative density

$$F(w) = \int_w^{\infty} f(v) dv$$

- ▶ over 100 years ago, the Italian economist Vilfredo Pareto first quantified the large-wealth behavior to follow a **power-law distribution**
- ▶ i.e.  $F(w) \sim w^{-\mu}$  for  $w$  large

# Kinetic models of simple markets

**Idea:** model agents in simple market as colliding molecules in a Boltzmann gas:

Economic system	Particle dynamics
agents	molecules
agent's wealth	particle momentum
mean wealth	total momentum
transactions	collisions

However, complete analogy fails:

- ▶ no debts allowed → particle momenta non-negative
- ▶ risky investments → collision kernel with randomness
- ▶ individual agent's preferences → particles distinguishable

# Goals

- ▶ kinetic model of a simple economy to describe evolution of the *macroscopic* wealth distribution by means of *microscopic* interactions (trades) among individuals (agents)
- ▶ we will consider binary interactions of the form

$$v^* = p_1 v + q_1 w, \quad w^* = p_2 v + q_2 w$$

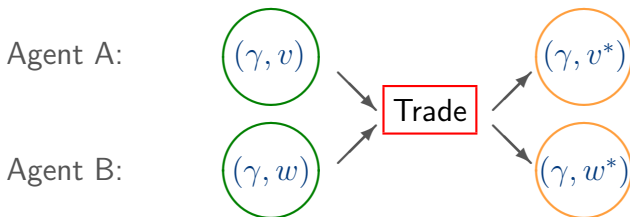
- ▶ wealth after trade non-negative (no debts)  $\rightarrow v^*, w^* \geq 0$
- ▶  $p_i, q_i$  can be **given constants or random quantities**
- ▶  $p_i, q_i$  should be nonnegative

**Benchmark:** steady states of good model should show Pareto tail



# Model of Cordier, Pareschi & Toscani (2005)

- $w/w^*$ ,  $v/v^*$  denote agents' wealths **before/after** trade
- single, fixed saving propensity  $\gamma \in (0, 1)$



- binary trade is characterized by

$$v^* = \gamma v + (1 - \gamma)w + \eta v$$

$$w^* = \gamma w + (1 - \gamma)v + \tilde{\eta} w$$

where risks of the market are described by  $\eta$  and  $\tilde{\eta}$ , random variables with same distribution and zero mean

# Boltzmann-type equation

## Paradigm ( $\sim 1985$ )

Wealth distribution function  $f(w, t)$  satisfies a *homogeneous Boltzmann equation* on  $\mathbb{R}_+$

$$\partial_t f = \mathcal{Q}(f, f), \quad f(w, 0) = f_0(w).$$

**Weak form:** For all smooth test functions  $\phi$  consider

$$\frac{d}{dt} \int_{\mathbb{R}_+} f(w, t) \phi(w) dw = \int_{\mathbb{R}_+} \mathcal{Q}(f, f) \phi(w) dw,$$

where

$$\begin{aligned} & \int_{\mathbb{R}_+} \mathcal{Q}(f, f) \phi(w) dw \\ &= \frac{1}{2} \left\langle \int_{\mathbb{R}_+} \int_{\mathbb{R}_+} [\phi(w^*) + \phi(v^*) - \phi(w) - \phi(v)] f(w) f(v) dw dv \right\rangle \end{aligned}$$

**Interpretation:** Pre-trade  $v, w$  change into post-trade  $v^*, w^*$

$\rightarrow$  to define a specific model, prescribe transitions  $(v, w) \rightsquigarrow (v^*, w^*)$

# Boltzmann equation: Fourier version

Existence of solutions to Boltzmann equation follows from results for the elastic Kac model

**Idea:** Boltzmann-like equation with constant kernel easily studied using *Fourier transform* (Bobylev, 1988)

Fourier version

$$\frac{\partial \hat{f}(t; \xi)}{\partial t} = \hat{Q}(\hat{f}, \hat{f})(t; \xi)$$

can be written as

$$\frac{\partial \hat{f}(t; \xi)}{\partial t} = \frac{1}{2} \langle \hat{f}(p_1 \xi) \hat{f}(q_1 \xi) + \hat{f}(p_2 \xi) \hat{f}(q_2 \xi) \rangle - \hat{f}(t; \xi)$$

# Unified approach

[B.D./Matthes/Toscani '08, Matthes/Toscani '08]

Use Fourier metric

$$d_s[f_1, f_2] = \sup_{\xi} [|\xi|^{-s} |\widehat{f}_1(\xi) - \widehat{f}_2(\xi)|], \quad s > 0$$

## Theorem

*Let  $f_1(t)$  and  $f_2(t)$  be two solutions of the Boltzmann equation, corresponding to initial values with same mean.*

*Let  $s \geq 1$  be such that  $d_s[f_{1,0}, f_{2,0}]$  is finite.*

*Then for all  $t \geq 0$*

$$d_s[f_1(t), f_2(t)] \leq \exp \left[ \left( \frac{1}{2} \left( \sum_{i=1}^2 \langle p_i^s + q_i^s \rangle \right) - 1 \right) t \right] d_s[f_{1,0}, f_{2,0}].$$

# Characteristic function

Introduce the characteristic function

$$\mathcal{S}(s) = \frac{1}{2} \left( \sum_{i=1}^2 \langle p_i^s + q_i^s \rangle \right) - 1$$

- ▶ obviously  $\mathcal{S}(0) = 1$
- ▶  $\mathcal{S}(1) = 0$  because of the conservation property
- ▶ sign of  $\mathcal{S}(s)$  is related to **number of moments of solution** which remain **uniformly bounded in time**
- ▶ Large-time behavior of solution **depends on sign of  $\mathcal{S}(s)$**

# Conclusions for CPT (2005) model

Mixing parameters are given by

$$\begin{aligned} p_1 &= \gamma + \eta, & q_1 &= 1 - \gamma \\ p_2 &= 1 - \gamma, & q_2 &= \gamma + \tilde{\eta} \end{aligned}$$

This implies **conservation in the mean**, i.e.  $\langle v^* + w^* \rangle = v + w$ .

Need to evaluate

$$\mathcal{S}(s) = (1 - \gamma)^s - 1 + \frac{1}{2} \langle (\gamma + \eta)^s + (\gamma + \tilde{\eta})^s \rangle$$

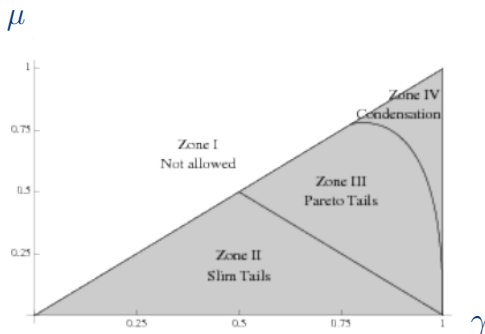
# Conclusions for CPT model: specific example

- ▶ Let  $\eta$  only assume values  $\pm\mu$ , with probability  $\frac{1}{2}$  each, where  $0 \leq \mu \leq \gamma$
- ▶ by varying  $\gamma$  and  $\mu$  one encounters variety of possible outcomes
- ▶ if  $\gamma + \mu \leq 1$ , trade is pointwise conservative  
→ all moments are finite (exponential tail)
- ▶ if  $\gamma + \mu > 1$ , evaluate

$$\mathcal{S}(s) = (1 - \gamma)^s - 1 + \frac{1}{2}(\gamma + \mu)^s + \frac{1}{2}(\gamma - \mu)^s$$

# Conclusions for CPT model: specific example

One obtains the following classification for  $f_\infty$



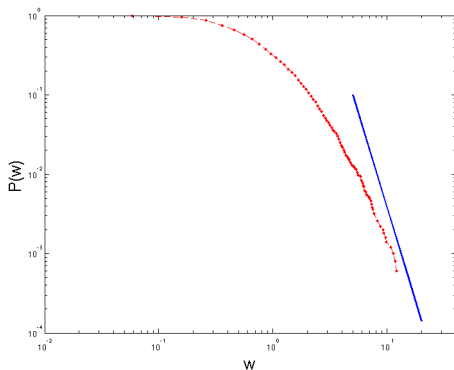
- ▶ Zone I: excluded by  $\mu \leq \gamma$
- ▶ Zone II: Socialism (exponential tails)
- ▶ Zone III: Capitalism (Pareto tails)
- ▶ Zone IV: Plutocracy (Dirac distribution)



# Numerical example for CPT model

Kinetic Monte Carlo simulation with  $N = 5000$  agents

- ▶ each agent starts with unit wealth
- ▶ uniform saving propensity  $\gamma \equiv 0.7$  and  $\eta = \pm 0.5$
- ▶ agents are randomly selected for trade events
- ▶ one time step =  $N$  collisions



- ▷ Pareto tail  $w^{-\bar{s}}$ ,
- ▷  $\bar{s} \approx 3.7$  ( $\mathcal{S}(\bar{s}) = 0$ )

# Kinetic models for opinion formation: literature

- ▶ Toscani (2006): opinion variable  $w \in \mathcal{I} = [-1, 1]$
- ▶ Boudin/Salvarani (2008): opinion variable  $w \in \mathcal{I}^p = [-1, 1]^p$
- ▶ Bertotti/Delitala (2009): opinion leadership, discrete model
- ▶ B.D./Markowich/Pietschmann/Wolfram (2009): opinion leadership, two species (PRSA)
- ▶ Motsch/Tadmor (2014), Heterophilious dynamics enhances consensus (SIAM Review)
- ▶ Pareschi/Toscani (2014): Book 'Interacting Multiagent Systems'
- ▶ ...

Connections to swarming/flocking models (Barbaro, Bertozzi, Cañizo, Carrillo, D'Orsogna, Slepčev,...) and 'sociophysics' literature (Galam *et al.*, Hegselmann/Krause,...)

# Opinion formation: Toscani (2006) model

- ▶ describe evolution of opinion distribution by **microscopic** interactions among individuals
- ▶ society develops a certain **macroscopic** opinion distribution
- ▶ opinion: continuous variable  $w \in \mathcal{I} = [-1, 1]$ , where  $\pm 1$  represent extreme opinions

Two individuals with pre-interaction opinion  $v$  and  $w$  meet  
 → post-interaction opinions  $v^*$  and  $w^*$  are given by

$$v^* = v - \gamma P(|v - w|)(v - w) + \eta_1 D(v),$$

$$w^* = w - \gamma P(|w - v|)(w - v) + \eta_2 D(w).$$

$\gamma \in (0, \frac{1}{2})$  : constant **compromise parameter**

$\eta_{1,2}$  : random variables with mean zero and variance  $\sigma^2$

modeling **self-thinking** through an exogenous, global access to information, e.g. through the press, television or internet.

# Homogeneous Boltzmann-like equation

Homogeneous society [Toscani '06]

↪ Boltzmann-like equation for the opinion distribution function  $f = f(w, t)$

$$\frac{\partial}{\partial t} f(w, t) = \frac{1}{\tau} \mathcal{Q}(f, f)(w)$$

$\tau$  : relaxation time

Collision operator in weak form:

$$\begin{aligned} & \int_{\mathcal{I}} \mathcal{Q}(f, f)(w) \phi(w) dw \\ &= \frac{1}{2} \left\langle \int_{\mathcal{I}^2} (\phi(w^*) + \phi(v^*) - \phi(w) - \phi(v)) f(v) f(w) dv dw \right\rangle \end{aligned}$$

for all smooth functions  $\phi(w)$

# Opinion leadership in sociology

- ▶ sociological concept trying to explain formation of opinions in a society
- ▶ going back to Lazarsfeld *et al.* (1944) studying US presidential elections 1940
- ▶ found out interpersonal communication to be much more influential than direct media effects
- ▶ **two-step flow of communication**: opinion leaders (active media users) select, interpret, modify, facilitate, and finally transmit information to less active parts of the population

# Opinion leader characteristics

Typical opinion leader characteristics are

- ▶ high confidence
- ▶ high self-esteem
- ▶ strong need to be unique (**public individuation**)
- ▶ socially active, highly connected (**scale-free network**)
- ▶ ability to withstand criticism

Although different, not easy to distinguish from followers

- ▶ still related to their followers
- ▶ opinion leadership is specific to a subject and can change over time

# Kinetic model with opinion leaders

[B.D./Markowich/Pietschmann/Wolfram, '09]

**Inhomogeneous** society consisting of two groups

- ▶ opinion leaders (highly self-confident, assertive, able to withstand criticism)
- ▶ followers

If two individuals from the same group meet ( $i = 1, 2$ )

$$v^* = v - \gamma_i P_i(|v - w|)(v - w) + \eta_{i1} D_i(v)$$

$$w^* = w - \gamma_i P_i(|w - v|)(w - v) + \eta_{i2} D_i(w)$$

Follower with opinion  $v$  meets opinion leader with opinion  $w$

$$v^* = v - \gamma_3 P_3(|v - w|)(v - w) + \eta_{11} D_1(v)$$

$$w^* = w$$

$\gamma_k \in (0, \frac{1}{2})$  : **compromise parameters**, which control the 'speed' of attraction of two different opinions

$\eta_{ij}$  : random variables with variance  $\sigma_{ij}^2$  and zero mean

# Boltzmann system

Distribution functions  $f_i = f_i(w, t)$  obey system of two Boltzmann-like equations, given by

$$\frac{\partial}{\partial t} f_i(w, t) = \sum_{j=1}^n \frac{1}{\tau_{ij}} \mathcal{Q}_{ij}(f_i, f_j)(w).$$

$\tau_{ij}$  : relaxation times

Collision operators in weak form:

$$\begin{aligned} & \int_{\mathcal{I}} \mathcal{Q}_{ij}(f_i, f_j)(w) \phi(w) dw \\ &= \frac{1}{2} \left\langle \int_{\mathcal{I}^2} (\phi(w^*) + \phi(v^*) - \phi(w) - \phi(v)) f_i(v) f_j(w) dv dw \right\rangle \end{aligned}$$

for all smooth functions  $\phi(w)$



# Fokker-Planck system

**Idea:** Derive macroscopic approximation

For  $\gamma \ll 1$ ,  $\tau = \gamma t$ , let  $g_i(w, \tau) = f_i(w, t)$ .

**Limit:**  $\gamma, \sigma_{ij} \rightarrow 0$  with  $\lambda_{ij} = \sigma_{ij}^2/\gamma$  fixed

$\rightarrow$  (scaled) densities converge to  $g_i(w, \tau)$ , which solve

$$\begin{aligned} \frac{\partial}{\partial \tau} g_1(w, \tau) = & \frac{\partial}{\partial w} \left( \left( \frac{1}{\tau_{11}} \mathcal{K}_1(w, \tau) + \frac{1}{2\tau_{12}} \mathcal{K}_3(w, \tau) \right) g_1(w, \tau) \right) \\ & + \left( \frac{\lambda_{11} M_1}{2\tau_{11}} + \frac{\lambda_{12} M_2}{4\tau_{12}} \right) \frac{\partial^2}{\partial w^2} (D_1^2(w) g_1(w, \tau)) \end{aligned}$$

$$\frac{\partial}{\partial \tau} g_2(w, \tau) = \frac{\partial}{\partial w} \left( \frac{1}{\tau_{22}} \mathcal{K}_2(w, \tau) g_2(w, \tau) \right) + \frac{\lambda_{22} M_2}{2\tau_{22}} \frac{\partial^2}{\partial w^2} (D_2^2(w) g_2(w, \tau))$$

with nonlocal drift operators

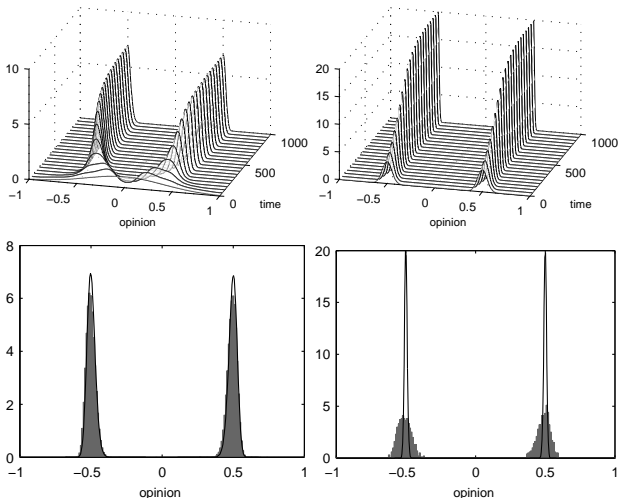
$$\mathcal{K}_i(w, \tau) = \int_{\mathcal{I}} P_i(|w - v|)(w - v) g_i(v, \tau) dv \quad \text{for } i = 1, 2$$

$$\mathcal{K}_3(w, \tau) = \int_{\mathcal{I}} P_3(|w - v|)(w - v) g_2(v, \tau) dv$$

# Numerical results

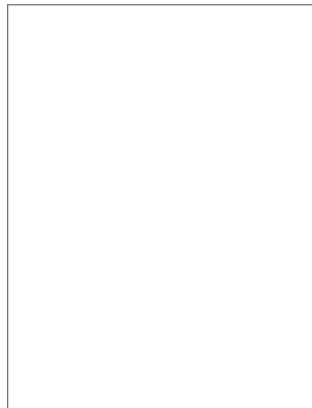
$$P_i(|v - w|) = \mathbf{1}_{\{|v-w| \leq r_i\}}, \quad r_i = 0.5$$

$$D_1(w) = D_2(w) = D(w) := (1 - w^2)^2$$



# Understanding Carinthia

- ▶ Carinthia is the southernmost state of Austria
- ▶ since 1999 elections the right-wing Freedom Party of Austria (FPÖ) became strongest party and continually improved, holding almost 45 % of the votes in 2008
- ▶ outcome strongly influenced by popularity of party leader Haider
- ▶ being considered populist, extreme-right or even antisemitic by many
- ▶ strongly acclaimed by his followers
- ▶ success less founded on political ideas than on authority of Haider himself



# Example: Carinthia state elections

Let's illustrate the behavior of our model under extreme conditions, like in Carinthia.

**Table:** Results of the state elections in Carinthia

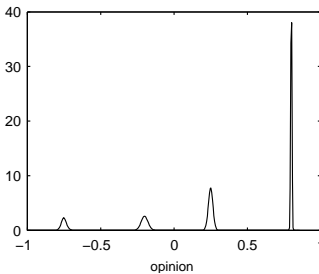
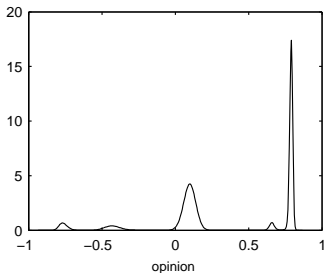
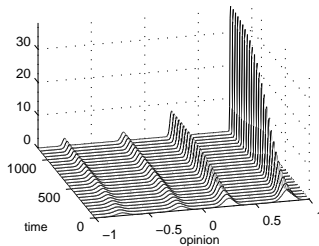
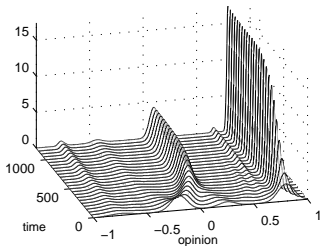
	Grüne	SPÖ	ÖVP	FPÖ	BZÖ
2004	6.7%	38.4 %	11.6 %	42.5 %	—
2009	5.2%	28.8 %	16.8 %	3.8 %	44.9 %

- ▶ initial distribution of normal people: weighted sum of Gaussians according to 2004 elections
- ▶ opinion leaders associated with the different parties with weights according to their influence
- ▶ other parameters

$$\alpha = 1.5, \quad \lambda = 3 \times 10^{-3}, \quad r_1 = r_2 = 0.2, \quad r_3 = 0.45,$$

$$\tau_{11} = \tau_{12} = 1, \quad \tau_{22} = 10, \quad \sigma_1 = 0.1, \quad \sigma_2 = 0.05.$$

# Numerical results



# Towards an inhomogeneous model

[B.D./Wolfram, '15]

More realistic models should depend on additional, independent variables, e.g.

- ▶ leadership 'strength'
- ▶ space
- ▶ ...

↪ **inhomogeneous** Boltzmann-type equation for  $f = f(x, w, t)$

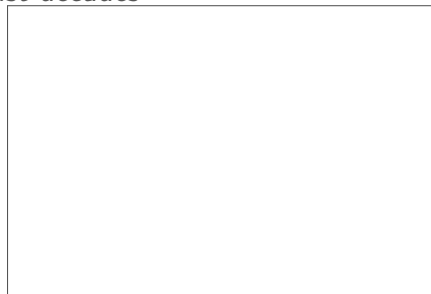
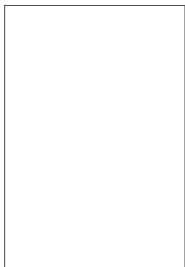
$$\frac{\partial}{\partial t} f + \operatorname{div}_x(\Phi(x, w)f) = \frac{1}{\tau} \mathcal{Q}(f, f)$$

- choice of the field  $\Phi = \Phi(x, w)$  which describes the opinion flux plays a crucial role
- in contrast to the physical situation where  $\Phi(w) = w$ , suitable choices need to be made
- example: inhomogeneous Boltzmann-type equation modelling political segregation in 'The Big Sort'

# Political segregation: The Big Sort

The Big Sort (Bishop/Cushion, 2008):

- clustering of individuals who share similar political opinions
- observed in USA over the last decades



- ▶ acclaimed in many newspapers and magazines
- ▶ former president Bill Clinton urged audiences to read the book
- ▶ but claims also met opposition from political sociologists

# Inhomogeneous model II: The Big Sort

**Landslide counties:** counties in which either candidate won or lost by 20 percentage points or more



→ number of landslide counties has doubled



# Inhomogeneous Boltzmann-type equation

opinion variable  $w \in [-1, 1]$

**additional variable:** position on map  $x \in \Omega \subset \mathbb{R}^2$

opinion distribution function  $f = f(x, w, t)$

supporters of the parties given by marginals

$$B(x, t) = \int_{-1}^0 f(x, w, t) dw, \quad R(x, t) = \int_0^1 f(x, w, t) dw$$

opinion dynamics ( $\sim$  Toscani, 2006):

$$v^* = v - \gamma K(|x - y|) P(|v - w|) (v - w) + \eta_1 D(v)$$

$$w^* = w - \gamma K(|y - x|) P(|v - w|) (w - v) + \eta_2 D(w)$$

We are lead to study

$$\frac{\partial}{\partial t} f + \operatorname{div}_x(\Phi(x, w)f) = \frac{1}{\tau} \mathcal{Q}(f, f).$$

# Fokker-Planck limit

Full inhomogeneous Boltzmann-type equation difficult to handle

**Idea:** Derive macroscopic approximation

For  $t' = \gamma t$ ,  $x' = \gamma x$  let  $g(x', w, t') = f(x, w, t)$ .

**Limit:**  $\gamma, \sigma \rightarrow 0$  with  $\lambda = \sigma^2/\gamma$  fixed

$\rightarrow$  (scaled) density converges to  $g(x, w, t)$ , which solves

$$\begin{aligned} & \frac{\partial}{\partial t} g(x, w, t) + \operatorname{div}_x(\Phi(x, w)g(x, w, t)) \\ &= \frac{\partial}{\partial w} \left( \frac{1}{\tau} \mathcal{K}(x, w, t)g(x, w, t) \right) + \frac{\lambda M(x)}{2\tau} \frac{\partial^2}{\partial w^2} (D^2(w)g(x, w, t)) \end{aligned}$$

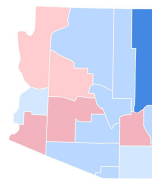
with nonlocal drift operator

$$\mathcal{K}(x, w, t) = \int_{\mathcal{I}} P(|w - v|)(w - v)g(x, v, t) dv$$

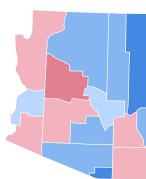
and mass  $M(x) = \int g(x, v, t) dv$

# Numerical example: The Big Sort in Arizona

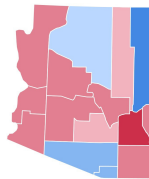
Results of the US presidential elections in the state of Arizona:



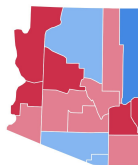
1992



1996



2000



2004

Colour intensities reflect election outcome in percent:

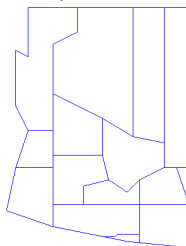
- ▶ dark blue (red) to Democrats (Republicans) 60–70%
- ▶ medium blue (red) to Democrats (Republicans) 50–60%
- ▶ light blue (red) to Democrats (Republicans) 40–50%

# Numerical results: The Big Sort in Arizona

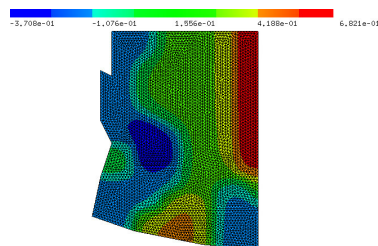
Initial distribution:  $f(x, w, 0)$  proportional to 1992 elections  
 Potential  $\Phi(x, w) = \text{sgn}(w)\nabla C(x)(1 - w^2)^\gamma$  is obtained from  
 1996 election results by solving

$$C(x) + \varepsilon\Delta C(x) = f_{1996}(x)$$

Solve the Fokker-Planck on  $\Omega \subset \mathbb{R}^2$  corresponding to the state  
 Arizona, divided into 15 electoral counties:



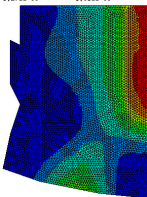
Domain  $\Omega$



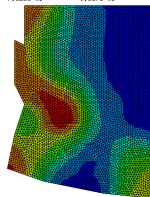
Potential  $C(x)$

# Numerical results: The Big Sort in Arizona

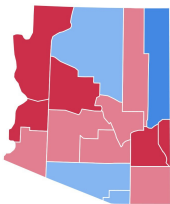
Simulation results:



Democrats



Republicans



→ patterns of the electoral results from 2004 are reproduced fairly well, except for the second county from the right in the Northeast (Navajo)

Election results 2004

# Summary and perspectives

## Summary:

- ▶ complex systems in economics and sociology  
↳ Boltzmann-type equations
- ▶ microscopic dynamics
- ▶ emergent behaviour
- ▶ examples:
  - ▶ wealth distribution
  - ▶ opinion dynamics under opinion leadership
  - ▶ political segregation: 'The Big Sort' in USA

## Perspectives:

- ▶ better understanding of input data (micro/macro)
- ▶ asymptotic behaviour of moments
- ▶ optimal control
- ▶ optimal strategies, mean-field games
- ▶ connection to graph-based approaches