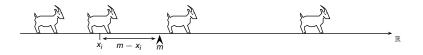
Modelling flocks and prices: jumping particles with an attractive interaction Joint work with Miklós Zoltán Rácz and Bálint Tóth

Márton Balázs¹

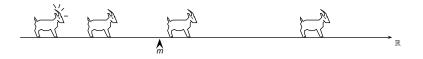
Budapest University of Technology and Economics MTA-BME Stochastics Research Group

Particle systems and PDE's U do Minho, December 5, 2012.

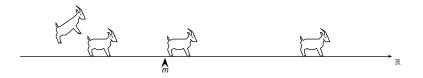
¹Bolyai Scholarship of the HAS; OTKA K100473;TAMOP422



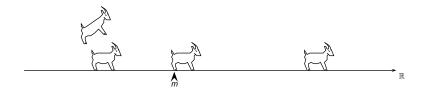
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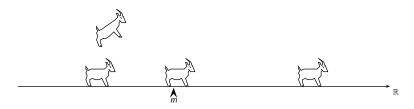
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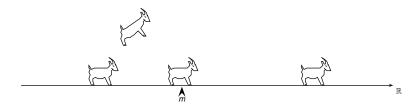
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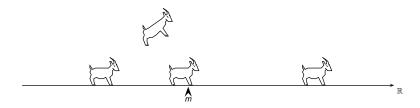
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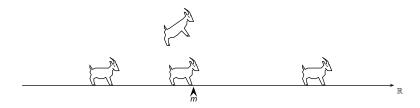
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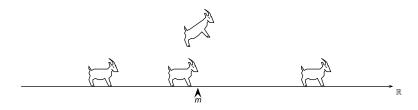
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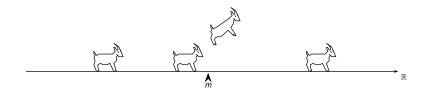
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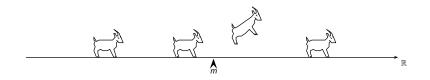
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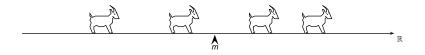
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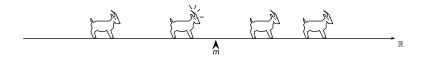
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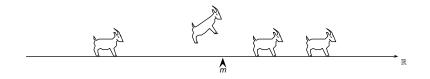
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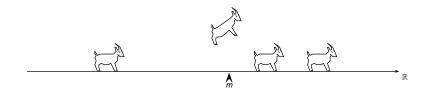
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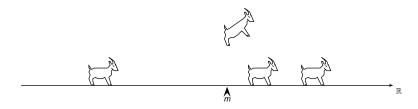
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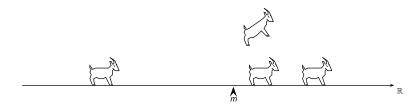
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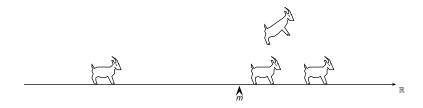
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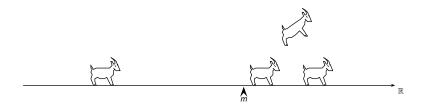
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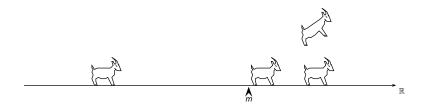
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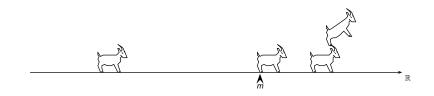
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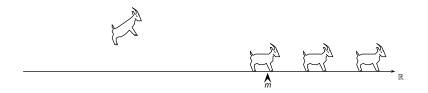
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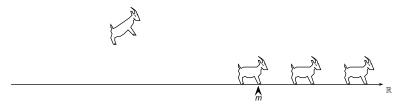
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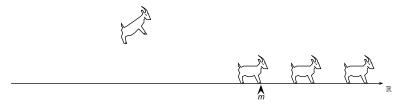
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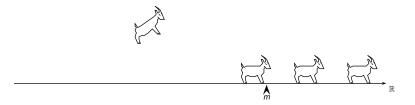
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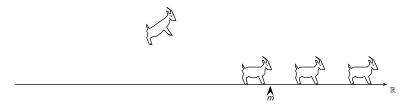
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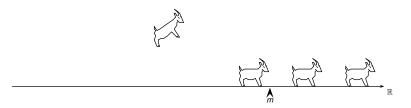
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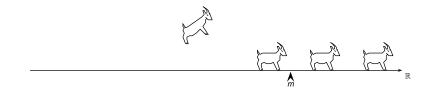
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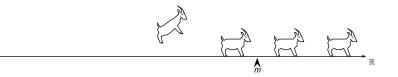
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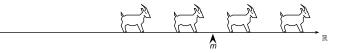
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Stationary distribution

Mean field equation

Exponential jumps Extreme value statistics Fourier methods

Fluid limit

Where do we live? Tightness The limit solves the mean field eq. Uniqueness

Questions

Can describe

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Found results of the types:

- rat race model (D. ben-Avraham, S.N. Majumdar, S. Redner 2007)
- interacting diffusions with linear drift (A. Greven et. al.),
- rank dependent drift of Brownian motions (S. Pal, J. Pitman 2008, S. Chatterjee, S. Pal 2009),
- relocation of random walking particles (A. Manita, V. Shcherbakov 2005),
- interacting jump processes (A. Greenberg, V.A. Malyshev, S.Yu. Popov 1995)
- multiplicative steps as well (I. Grigorescu, M. Kang 2010).

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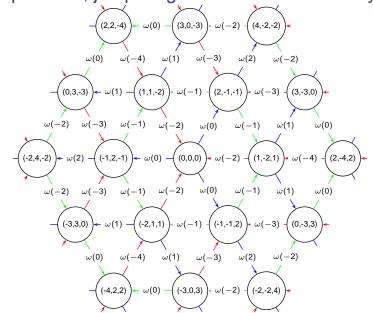
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n = 3 particles: already seems hopeless. The process is "very irreversible".

n = 3 particles, jump lengths are deterministically 1



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$$\begin{array}{l} \begin{array}{l} \begin{array}{l} \begin{array}{l} \text{jump rate at } x & \text{density at } x \end{array} \\ \hline \frac{\partial \varrho(x,t)}{\partial t} = & - w(x-m(t)) \cdot \varrho(x,t) \\ \\ \text{jump rate at } y & \text{density at } y & \text{prob to jump to } x \end{array} \\ + \int_{-\infty}^{x} w(y-m(t)) \cdot \varrho(y,t) \cdot \varphi(x-y) & \mathrm{d} y, \end{array}$$

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These equations conserve $1 = \int \rho(x, t) dx$ and give $\dot{m}(t) = \int w(x - m(t)) \cdot \rho(x, t) dx$.

We look for stationary solution of this equation as seen from the center of mass.

Idea: as $n \to \infty$, in a stationary distribution m(t) would stabilize. So assume

$$m(t) = ct$$
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Plug this in to get

$$\begin{aligned} -c\varrho'(x) &= -w(x)\varrho(x) + \int_{-\infty}^{x} w(y)\varrho(y)\varphi(x-y) \, \mathrm{d}y, \\ 0 &= \int_{-\infty}^{\infty} y\varrho(y) \, \mathrm{d}y. \end{aligned}$$

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Between t and t + dt, $dN(t) = e^{ct} dt$ many new Exp(1) particles try to break the record. So the probability that Y(t) jumps is

$$1 - (1 - e^{-Y(t)})^{e^{ct} dt} \simeq e^{ct - Y(t)} dt \qquad \text{(for large } Y(t)\text{)}.$$

And when it jumps, it jumps Exp(1). But we know that $Y(t) - ct + \log c$ converges to standard Gumbel.

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Method tested when φ(x) = e^{-x} (also seen before), hope to work with other φ's too.

Taking the fluid limit

Recall the original mean field equation:

$$\begin{aligned} \frac{\partial \varrho(\mathbf{x},t)}{\partial t} &= -w(\mathbf{x}-m(t)) \cdot \varrho(\mathbf{x},t) \\ &+ \int_{-\infty}^{\mathbf{x}} w(\mathbf{y}-m(t)) \cdot \varrho(\mathbf{y},t) \cdot \varphi(\mathbf{x}-\mathbf{y}) \, \mathrm{d}\mathbf{y}, \end{aligned}$$

or, for all f test functions:

$$egin{aligned} \langle f,\mu(t)
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angle \ &= \int_0^t ig\langle \left\{ \mathbf{E}[f(\mathbf{x}+\mathbf{Z})] - f(\mathbf{x})
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angle \, \, \mathrm{d}\mathbf{s}, \ m(\mathbf{s}) &= \langle \mathbf{x}, \, \mu(\mathbf{s})
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Here **E** refers to expectation of Z w.r.t. the jump length distribution.

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Define the *n*-particle empirical measure $\mu_n(t) = \frac{1}{n} \sum_{i=1}^n \delta_{x_i(t)}$. Goal:

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Problem: bounded functions and "just measures" are not enough!

Probability measures on \mathbb{R} with finite first moment: \mathcal{P}_1 .

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Wasserstein metric on \mathcal{P}_1 :

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Goal: convergence of the *n*-particle empirical measures $\mu_n(t)$ in the Skohorod space $D([0, \infty), \mathcal{P}_1)$.

Step 1: Tightness of ⟨f, µn(t)⟩ in D([0, ∞], ℝ); f bounded, continuous. (Grigorescu-Kang 2010)

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C-relative compactness

Method for these bounds: introduce *ghost goats*: they jump with rate $\sup_{x} w(x)$, they have the same jump length distribution as their planetary counterparts. Couple such that ghost goat_i can jump without goat_i, but not vice-versa. \rightsquigarrow increments of ghosts dominate increments of the planetary goats.

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For the compactness-type conditions, use again the ghost goats.

2. The limit solves the mean field eq.

Let

$$\begin{split} \mathbf{A}_{t,f}(\mu) &:= \langle f, \mu(t) \rangle - \langle f, \mu(0) \rangle \\ &- \int_0^t \left\langle \left\{ \mathbf{E}[f(\mathbf{x} + \mathbf{Z})] - f(\mathbf{x}) \right\} w(\mathbf{x} - m(\mathbf{s})), \, \mu(\mathbf{s}) \right\rangle \, \mathrm{d}\mathbf{s} \\ &= \langle f, \mu(t) \rangle - \langle f, \mu(0) \rangle - \int_0^t L \langle f, \mu(\mathbf{s}) \rangle \, \mathrm{d}\mathbf{s}, \\ \mathbf{m}(\mathbf{s}) &= \langle \mathbf{x}, \, \mu(\mathbf{s}) \rangle. \end{split}$$

Recall that the mean field equation was

$$A_{t,f}(\mu)=0.$$

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in probability.

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Step 2: If $\mu_n \Rightarrow \mu$ in $D([0, \infty], \mathcal{P}_1)$, then

$$A_{s,f}(\mu_n) \Rightarrow A_{s,f}(\mu)$$

in \mathbb{R} .

3. Uniqueness of solutions of the mean field eq.

Step 1: Look at the distance

$$d_{H}(\mu, \nu) := \sup_{f} |\langle f, \mu \rangle - \langle f, \nu \rangle|,$$

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 $\rightsquigarrow d_H(\mu(t), \nu(t)) \le d_H(\mu(0), \nu(0)) + c \int_0^t d_H(\mu(s), \nu(s)) ds$, apply Grönwall's inequality.

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