

Modelling flocks and prices: jumping particles with an attractive interaction

Joint work with Miklós Zoltán Rácz and Bálint Tóth

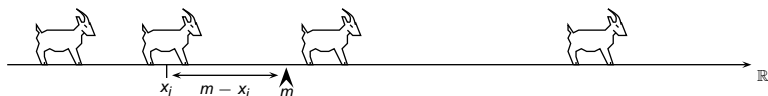
Márton Balázs

University of Bristol

Stochastic Analysis Seminar
Oxford, 10th March, 2014.

The model

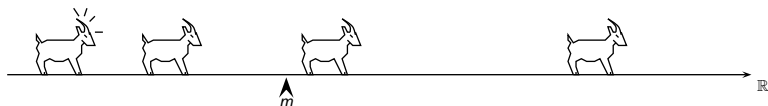
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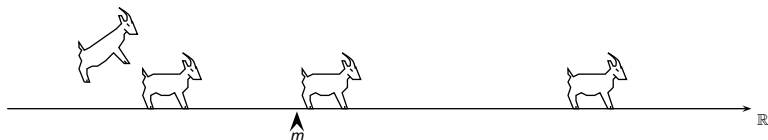
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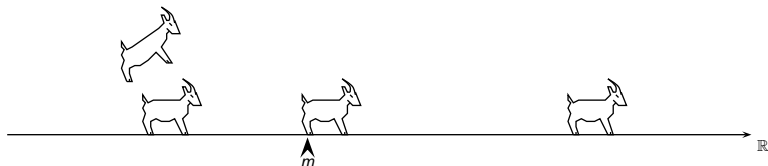
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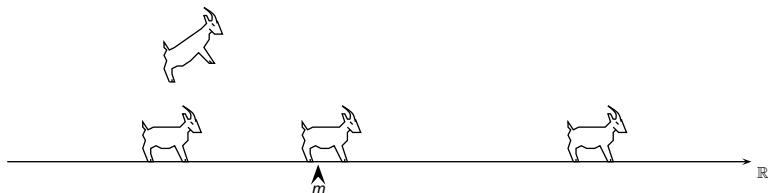
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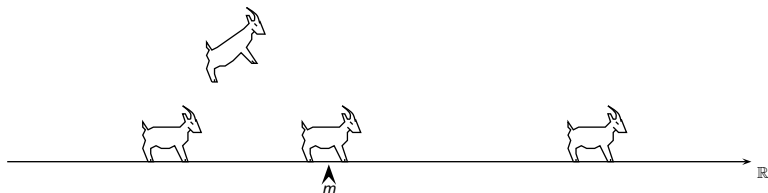
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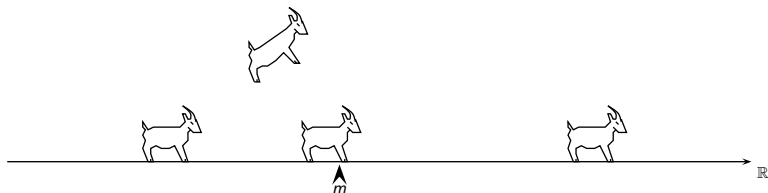
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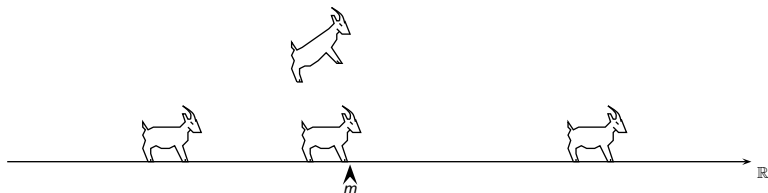
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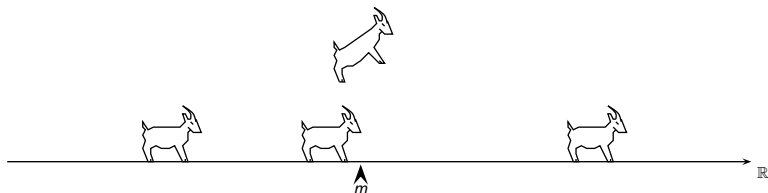
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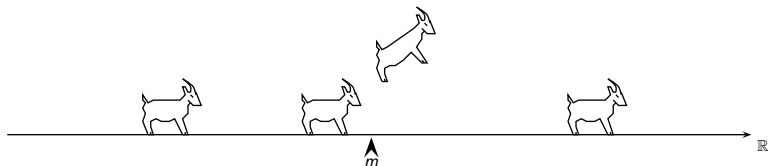
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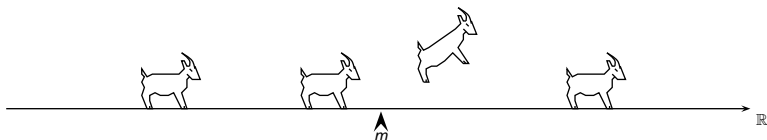
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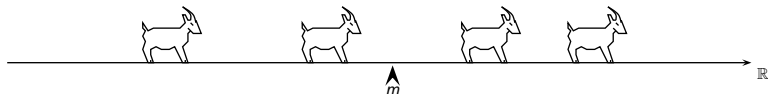
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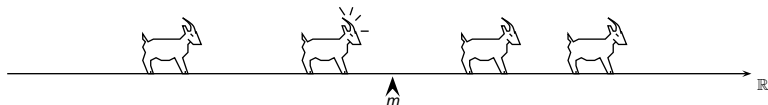
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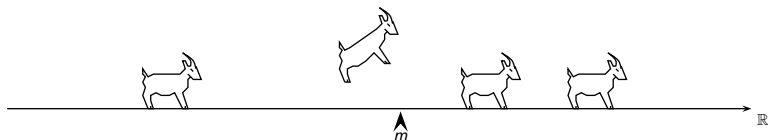
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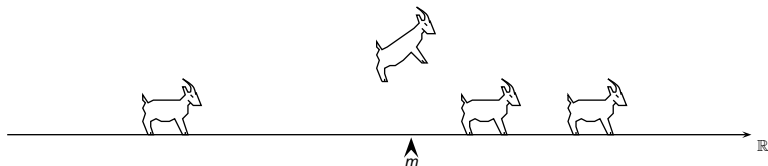
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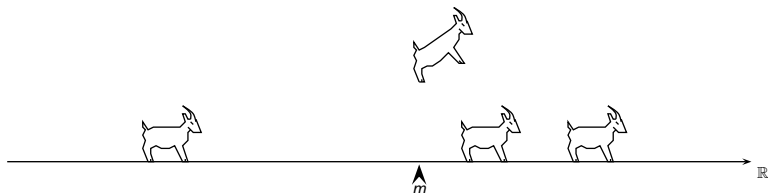
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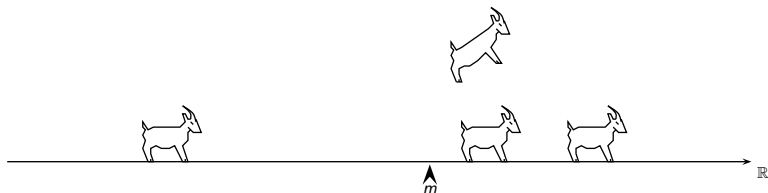
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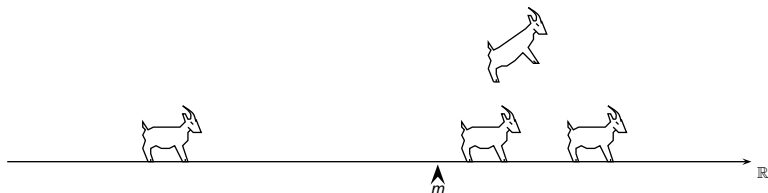
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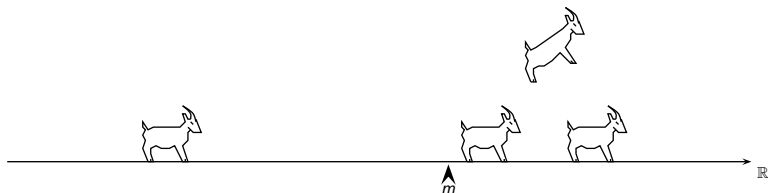
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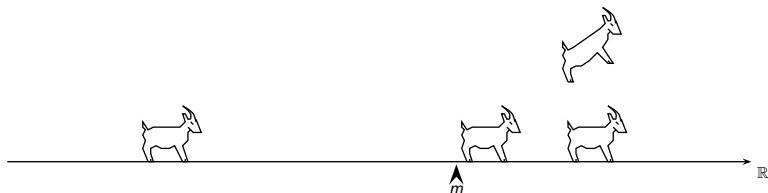
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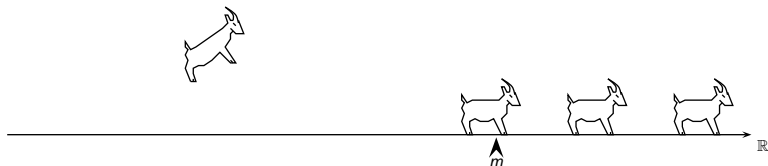
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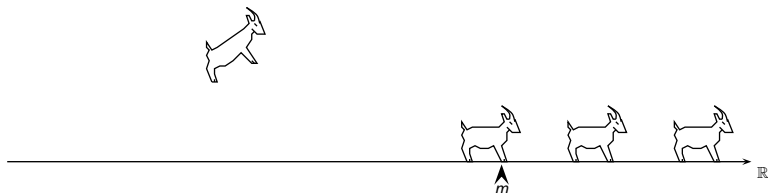
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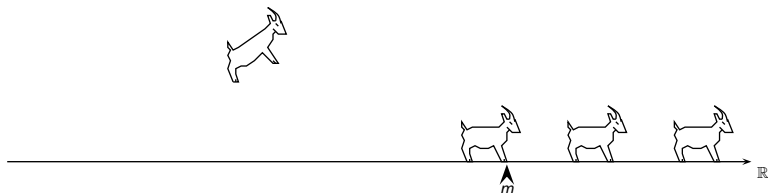
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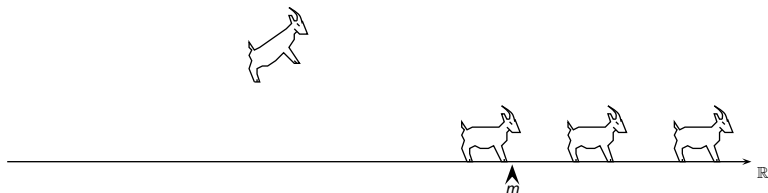
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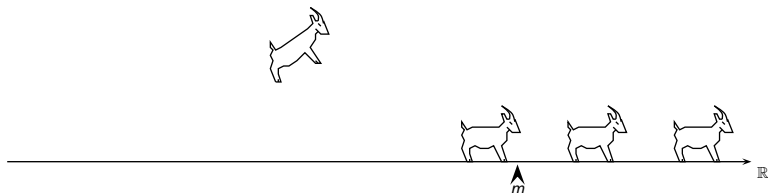
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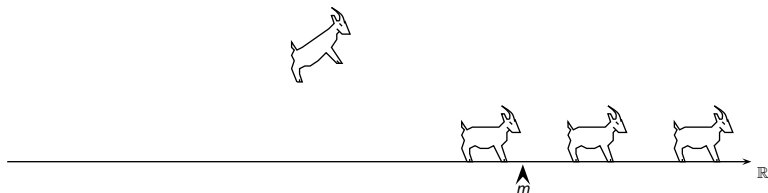
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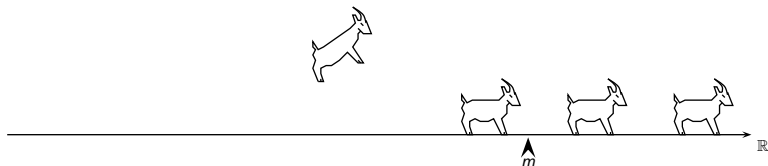
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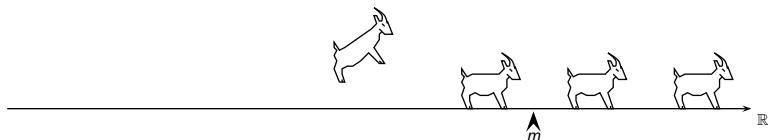
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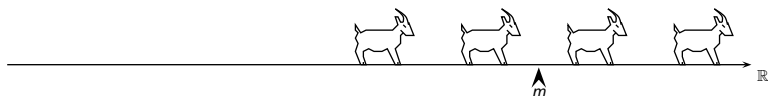
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Stationary distribution

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

The model

Can describe

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The model

Can describe

- ▶ motion of flocks, herds (as you have seen...),
- ▶ competing prices of goods (gyros / falafel / shawarma),
- ▶ prices of stocks, etc.

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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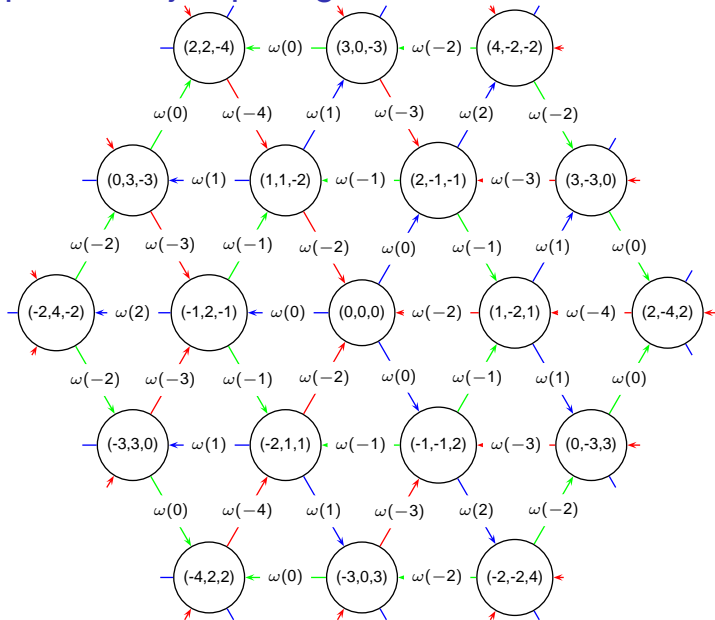
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$n = 3$ particles: already seems hopeless. The process is “very irreversible”.

$n = 3$ particles, jump lengths are deterministically 1



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

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

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

$$m(t) = ct \quad \text{and} \\ \varrho(x, t) = \varrho(x - ct).$$

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Plug this in to get

$$-c\varrho'(\mathbf{x}) = -w(\mathbf{x})\varrho(\mathbf{x}) + \int_{-\infty}^{\mathbf{x}} w(\mathbf{y})\varrho(\mathbf{y})\varphi(\mathbf{x} - \mathbf{y}) \, d\mathbf{y}, \\ 0 = \int_{-\infty}^{\infty} \mathbf{y}\varrho(\mathbf{y}) \, d\mathbf{y}.$$

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

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

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

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Hope to solve the recurrence relation on the \Im m line, then analytic continuation gives a hint on the form of $\widehat{\varrho}$, to be verified.

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

Taking the fluid limit

Recall the original mean field equation:

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or, for all f test functions:

$$\begin{aligned} \langle f, \mu(t) \rangle - \langle f, \mu(0) \rangle &= \int_0^t \langle \{ \mathbf{E}[f(\mathbf{x} + \mathbf{Z})] - f(\mathbf{x}) \} w(\mathbf{x} - m(s)), \mu(s) \rangle \, ds, \\ m(s) &= \langle \mathbf{x}, \mu(s) \rangle. \end{aligned}$$

Here \mathbf{E} refers to expectation of \mathbf{Z} w.r.t. the jump length distribution.

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Problem: bounded functions and “just measures” are not enough!

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

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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* _{j} can jump without *goat* _{j} , 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**.

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Let

$$\begin{aligned} A_{t,f}(\mu) &:= \langle f, \mu(t) \rangle - \langle f, \mu(0) \rangle \\ &\quad - \int_0^t \langle \{ \mathbf{E}[f(\mathbf{x} + \mathbf{Z})] - f(\mathbf{x}) \} w(\mathbf{x} - m(s)), \mu(s) \rangle ds \\ &= \langle f, \mu(t) \rangle - \langle f, \mu(0) \rangle - \int_0^t L \langle f, \mu(s) \rangle ds, \\ m(s) &= \langle \mathbf{x}, \mu(s) \rangle. \end{aligned}$$

Recall that the mean field equation was

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

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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} .

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

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