t^{1/3}-scaling of current fluctuations: coupling results and problems

Joint with Júlia Komjáthy and Timo Seppäläinen

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Happy birthday Jóska!

Budapest, March 2008

The models ASEP Zero range Bricklayers

Current variance

Hydrodynamics Characteristics

Tool: the second class particle

Single Many second class particles

Results

Current fluctuations Microscopic convexity Coupling results Nonconvex, nonconcave



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The jump is suppressed if the destination site is occupied by another particle.

The Bernoulli(ϱ) distribution is time-stationary for any ($0 \le \varrho \le 1$). Any translation-invariant stationary distribution is a mixture of Bernoullis.




































































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Question: What is the time-order of $Var(h_{Vt}(t))$?

The *density* $\varrho := \mathbf{E}(\omega)$ and the *hydrodynamic flux* $H := \mathbf{E}[$ growth rate] both depend on a parameter of the stationary distribution.

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► The characteristics is a path X(T) where *e*(T, X(T)) is constant.

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So, $\dot{X}(T) = H'(\varrho) = : C$ is the *characteristic speed*.

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So, $\dot{X}(T) = H'(\varrho) = : C$ is the *characteristic speed*. If $H(\varrho)$ is convex or concave, then *the Rankine-Hugoniot speed* for densities ϱ and λ is

$$R = rac{H(\varrho) - H(\lambda)}{arrho - \lambda}.$$

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This would be the speed of a shock of densities ρ and λ .

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$${m C}={m H}'(arrho)>{m R}=rac{{m H}(arrho)-{m H}(\lambda)}{arrho-\lambda}$$

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Theorem (B. - Seppäläinen; also ideas from Bálint, Herbert, Michael Prähofer)

Started from (almost) equilibrium,

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The second class particle follows the characteristics, people have known this for a long time.

















































































































































Picture:

The position X(t) of \uparrow^0 follows the Rankine-Hugoniot speed *R*.

Convex flux (some cases of AZRP, ABLP):



Recall
$$C = H'(\varrho) > R = \frac{H(\varrho) - H(\lambda)}{\varrho - \lambda}$$

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Initial fluctuations are transported along the characteristics on this scale.

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$$0 < \liminf_{t \to \infty} \frac{\operatorname{Var}(h_{Ct}(t))}{t^{2/3}} \leq \limsup_{t \to \infty} \frac{\operatorname{Var}(h_{Ct}(t))}{t^{2/3}} < \infty.$$

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Both these arguments seem to be robust once $H(\varrho)$ is strictly convex or concave, and we have $Q(t) \ge X(t)$ (convex) or $Q(t) \le X(t)$ (concave).

There are limit distribution results for TASEP by Johansson 2000, Prähofer and Spohn 2001, Ferrari and Spohn 2006. Their methods are completely different, relying on combinatorial tricks and asymptotic analysis of certain determinants.



















































































































































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Model	<i>Η</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
			1	

Model	<i>H</i> (<i>ϱ</i>) is	$m_{\rm Q}(t)$	Hence	<i>t</i> ^{2/3} law
TASEP				

Model	<i>Η</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave			

Model	<i>H</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	\leq 0		

Model	<i>Η</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	\leq 0	$Q(t) \leq X(t)$	
	I			1

Model	<i>Η</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	\leq 0	$Q(t) \leq X(t)$	proved (BS.)

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TASEP	concave	\leq 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \leq X(t) + Err$	

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TASEP	concave	≤ 0	$Q(t) \leq X(t)$	proved (BS.)
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rate 1 TAZRP				

Model	<i>Η</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	≤ 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \leq X(t) + Err$	proved (<mark>BS</mark> .)
rate 1 TAZRP	concave			

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TASEP	concave	\leq 0	$Q(t) \leq X(t)$	proved (BS.)
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ASEP	concave	\leq Geom	$Q(t) \leq X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	≤ 0	$Q(t) \leq X(t)$	

Model	<i>Η</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	\leq 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \le X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	≤ 0	$Q(t) \leq X(t)$	ok (<mark>BK.</mark>)

Model	<i>Η</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	\leq 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \leq X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (<mark>BK.</mark>)
exp rate TABLP				

Model	<i>H</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	≤ 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \leq X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (<mark>BK</mark> .)
exp rate TABLP	convex			

Model	<i>H</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	≤ 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \le X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (BK.)
exp rate TABLP	convex	$\geq - \text{Geom}$		

Model	<i>Н</i> (<u></u>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	\leq 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \le X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (<mark>BK.</mark>)
exp rate TABLP	convex	\geq –Geom	$Q(t) \ge X(t) - Err$	

Model	<i>H</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	≤ 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \le X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (<mark>BK.</mark>)
exp rate TABLP	convex	\geq -Geom	$Q(t) \ge X(t) - Err$	seems ok (<mark>BS.</mark>)

Model	<i>H</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	\leq 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \le X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (<mark>BK.</mark>)
exp rate TABLP	convex	$\geq - \text{Geom}$	$Q(t) \ge X(t) - Err$	seems ok (<mark>BS.</mark>)
rate 1 AZRP				
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rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (<mark>BK.</mark>)
exp rate TABLP	convex	$\geq - \text{Geom}$	$Q(t) \ge X(t) - Err$	seems ok (<mark>BS</mark> .)
rate 1 AZRP	concave			
exp rate ABLP	convex			

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ASEP	concave	\leq Geom	$Q(t) \le X(t) + Err$	proved (<mark>BS</mark> .)
rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (<mark>BK.</mark>)
exp rate TABLP	convex	$\geq - \text{Geom}$	$Q(t) \ge X(t) - Err$	seems ok (<mark>BS.</mark>)
rate 1 AZRP	concave	\leq Geom		
exp rate ABLP	convex	$\geq - \text{Geom}$		

Model	<u> </u>	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	≤ 0	$Q(t) \leq X(t)$	proved (BS.)
ASEP	concave	\leq Geom	$Q(t) \leq X(t) + Err$	proved (BS.)
rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (BK .)
exp rate TABLP	convex	$\geq - \text{Geom}$	$Q(t) \ge X(t) - Err$	seems ok (<mark>BS</mark> .)
rate 1 AZRP	concave	\leq Geom	$Q(t) \le X(t) + Err$	
exp rate ABLP	convex	$\geq -$ Geom	$Q(t) \ge X(t) - Err$	

Model	<i>H</i> (<i>ϱ</i>) is	$m_Q(t)$	Hence	<i>t</i> ^{2/3} law
TASEP	concave	\leq 0	$Q(t) \leq X(t)$	proved (BS.)
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rate 1 TAZRP	concave	\leq 0	$Q(t) \leq X(t)$	ok (BK.)
exp rate TABLP	convex	$\geq - \text{Geom}$	$Q(t) \ge X(t) - Err$	seems ok (BS.)
rate 1 AZRP	concave	\leq Geom	$Q(t) \le X(t) + Err$	future (<mark>BK.</mark>)
exp rate ABLP	convex	$\geq - \text{Geom}$	$Q(t) \ge X(t) - Err$	future (<mark>BS</mark> .)

Linear models

There are asymmetric models with linear hydrodynamics:

- The random average process (RAP),
- The AZRP with linear rates = independent random walkers.

Linear models

There are asymmetric models with linear hydrodynamics:

The random average process (RAP),

The AZRP with linear rates = independent random walkers.

In their cases, we have

$$\lim_{t\to\infty}\frac{\operatorname{Var}(h_{Ct}(t))}{t^{1/2}}=\ldots,$$

even convergence of the finite-dimensional distributions of the $h_{Ct}(t)$ process to Gaussian limits is known (Seppäläinen 2005, Ferrari and Fontes 1998, B., Rassoul-Agha and Seppäläinen 2006).

And there are attractive asymmetric models with nonlinear, nonconvex and nonconcave hydrodynamics:

▶ 2-jump exclusion: $\stackrel{\bullet}{\longrightarrow}$ $\stackrel{\bullet}{\longrightarrow}$ $H(\varrho)$ is a cubic polynomial;

And there are attractive asymmetric models with nonlinear, nonconvex and nonconcave hydrodynamics:

▶ 2-jump exclusion: $\stackrel{\bullet}{\longleftarrow} \stackrel{\bullet}{\longrightarrow} H(\varrho)$ is a cubic polynomial;

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And there are attractive asymmetric models with nonlinear, nonconvex and nonconcave hydrodynamics:

- ▶ 2-jump exclusion: $\xrightarrow{\bullet} \xrightarrow{\bullet} \xrightarrow{\bullet} H(\varrho)$ is a cubic polynomial;
- A three-state process with variable rates (B. Tóth I. Tóth).





C < R $Q(t) \stackrel{?}{\leq} X(t)$





Inequality changes with the density... ?



Inequality changes with the density...? Any coupling *must be very very tricky*.

Thank you.