

# On asymptotic behavior of the modified Arratia flow

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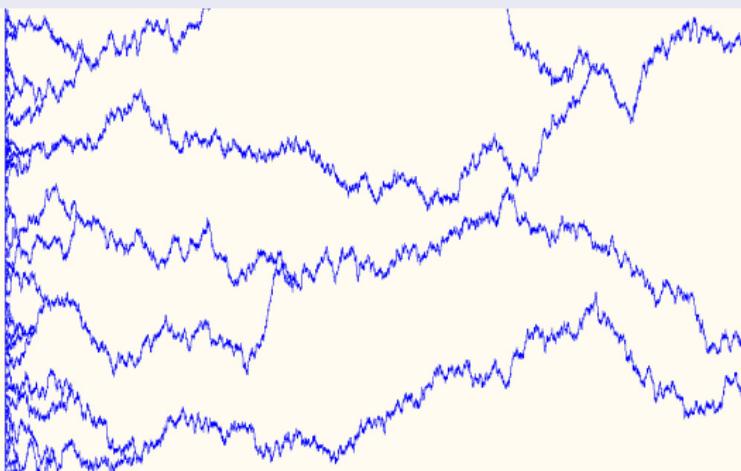
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# Arratia flow

The system of Brownian particles on the real line that

- ① start from all points of  $\mathbb{R}$ ;
- ② move independently up to the moment of meeting;
- ③ coalesce;
- ④ the diffusion is unchangeable and equals 1.

(R.A. Arratia '79)

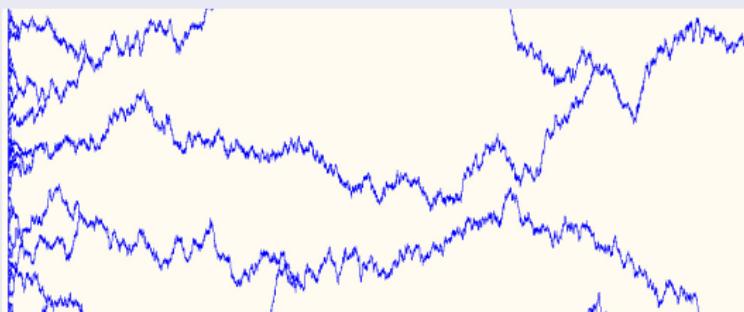


# Arratia flow

## The Arratia flow, the mathematical description

$\{x(u, t), t \in [0, T], u \in \mathbb{R}\}$  in  $D(\mathbb{R}, C[0, T])$  such that

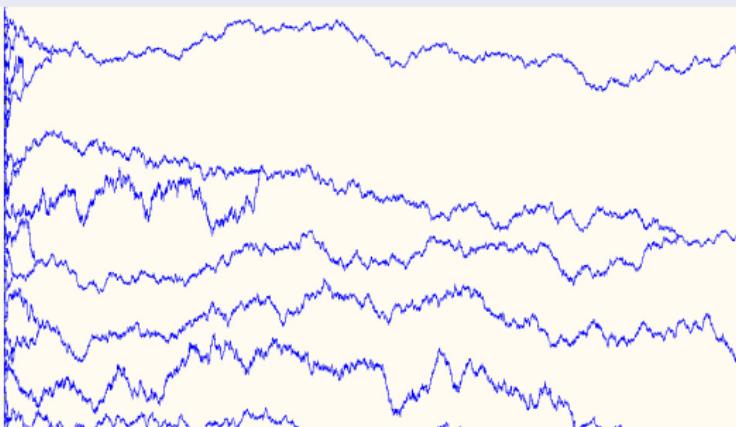
- ①  $x(u, \cdot)$  is a continuous square integrable martingale with respect to the joint filtration;
- ②  $x(u, 0) = u, \quad u \in \mathbb{R};$
- ③  $x(u, t) \leq x(v, t), \quad u < v;$
- ④  $\langle x(u, \cdot) \rangle_t = t;$
- ⑤  $\langle x(u, \cdot), x(v, \cdot) \rangle_t = 0, \quad t < \tau_{u,v} = \inf\{t : x(u, t) = x(v, t)\}.$



# Main object

We consider the system of diffusion particles on the real line which

- ① start from some set of points;
- ② move independently up to the moment of the meeting and then coalesce;
- ③ masses add after coalescing;
- ④ the diffusion is inversely proportional to the mass;



# Continuum system

## Theorem 1 (K. '14)

There exists a random element  $\{y(u, t), u \in [0, 1], t \in [0, T]\}$  in  $D([0, 1], C[0, T])$  such that

1°)  $y(u, \cdot)$  is a continuous square integrable martingale with respect to

$$\mathcal{F}_t = \sigma(y(u, s), u \in [0, 1], s \leq t), \quad t \in [0, T];$$

2°)  $y(u, 0) = u, u \in [0, 1];$

3°)  $y(u, t) \leq y(v, t), u < v;$

4°)  $\langle y(u, \cdot) \rangle_t = \int_0^t \frac{ds}{m(u, s)}, \quad m(u, t) = \text{Leb}\{v : y(v, t) = y(u, t)\};$

5°)  $\langle y(u, \cdot), y(v, \cdot) \rangle_t = 0, \quad t \leq \tau_{u,v} = \inf\{t : y(u, t) = y(v, t)\} \wedge T.$

# The reason of existence

## The main estimation

$$\mathbb{P}\{m(u, t) < r\} \leq \frac{2}{\sqrt{2\pi}} \int_0^{\frac{r^{3/2}}{\sqrt{t}}} e^{-\frac{x^2}{2}} dx \leq \frac{2r\sqrt{r}}{\sqrt{2\pi t}}.$$

## The mass growth

For  $\beta \in (0, \frac{3}{2})$

$$\mathbb{E} \frac{1}{m^\beta(u, t)} = \int_1^{+\infty} \mathbb{P} \left\{ m(u, t) < \frac{1}{r^{1/\beta}} \right\} dr \leq \frac{C}{\sqrt[3]{t^\beta}}.$$

(If particles start from interval  $[0, b]$ ,  $b \geq 1$ , then the constant  $C$  is independent of  $b$ .)

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# Connection with Wasserstein diffusion

## Wasserstein diffusion (M.-K. von Renesse, K.-T. Sturm '09)

$$d\mu_t = \beta \Delta \mu_t dt + \hat{\Gamma}(\mu_t) dt + \operatorname{div}(\sqrt{\mu} dW_t),$$

with

$$\langle f, \hat{\Gamma}(\nu) \rangle = \sum_{I \in \text{gaps}(\nu)} \left[ \frac{f''(I_+) + f''(I_-)}{2} - \frac{f'(I_+) - f'(I_-)}{|I|} \right].$$

## Evolution of particle mass in modified Arratia flow

Let  $\mu_t := \operatorname{Leb} \circ y(\cdot, t)^{-1}$  then

$$d\mu_t = \Gamma(\mu_t) dt + \operatorname{div}(\sqrt{\mu} dW_t),$$

where

$$\langle f, \Gamma(\nu) \rangle = \sum_{x \in \operatorname{supp}(\nu)} f''(x).$$

# Large deviations

## LDP (K., Max von Renesse '15)

The family of processes  $\{y(\varepsilon \cdot)\}_{\varepsilon > 0}$  satisfies a large deviations principle with the rate function

$$I(\varphi) = \begin{cases} \frac{1}{2} \int_0^T \|\dot{\varphi}(t)\|_{L_2(du)}^2 dt, & \varphi \in H, \\ +\infty, & \text{otherwise.} \end{cases}$$

$$\begin{aligned} H = \{&\varphi \in C([0, T], L_2^\uparrow([0, 1], du)) : \varphi(0) = \text{id} \text{ and} \\ &t \rightarrow \varphi(t) \in L_2([0, 1], du) \text{ is absolutely continuous}\}. \end{aligned}$$

Varadhan's formula for the modified Arratia flow

$$\mathbb{P}\{y(t) = g\} \sim e^{-\frac{\|\text{id} - g\|^2}{2t}}, \quad t \rightarrow 0.$$

$$\text{For } \mu_t := \text{Leb} \circ y(\cdot, t)^{-1}, \quad \mathbb{P}\{\mu_t = \nu\} \sim e^{-\frac{\mathcal{W}(\text{Leb}, \nu)^2}{2t}}, \quad t \rightarrow 0.$$

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# The aim of talk

## Asymptotic behavior of particles

To find  $\varphi(t)$  and  $\psi(t)$ ,  $t \geq 0$ , such that

$$\overline{\lim}_{t \rightarrow 0+} \frac{|y(u, t) - y(u, 0)|}{\varphi(t)} = 1 \quad \text{a.s.},$$

$$\overline{\lim}_{t \rightarrow 0+} \frac{m(u, t)}{\psi(t)} = 1 \quad \text{a.s.}$$

# Asymptotic behavior of Arratia flow $x$

## Law of the iterated logarithm

$$\overline{\lim}_{t \rightarrow 0^+} \frac{|x(0, t)|}{\sqrt{2t \ln \ln \frac{1}{t}}} = 1 \quad \text{a.s.}$$

## LIL for the cluster size

$$\overline{\lim}_{t \rightarrow 0^+} \frac{\nu(t)}{\sqrt{2t \ln \ln \frac{1}{t}}} \geq 1 \quad \text{a.s.}$$

$$\overline{\lim}_{t \rightarrow 0^+} \frac{\nu(t)}{2\sqrt{t \ln \ln \frac{1}{t}}} \leq 1 \quad \text{a.s.,}$$

where  $\nu(t) = \text{Leb}\{u : x(u, t) = x(0, t)\}$ .

(A.A. Dorogovtsev, A.V. Gnedin, M.B. Vovchanskii '12)

# Asymptotic behavior of the modified Arratia flow

## Asymptotic behavior (K. '16)

For all  $\varepsilon > 0$  and  $u \in [0, 1]$

$$\mathbb{P} \left\{ \lim_{t \rightarrow 0} \frac{m(u, t)}{\sqrt[3]{t} (\ln \frac{1}{t})^{1+\varepsilon}} = 0 \right\} = \mathbb{P} \left\{ \overline{\lim}_{t \rightarrow 0} \frac{m(u, t)}{\sqrt[3]{t} (\ln \frac{1}{t})^{-1-\varepsilon}} = +\infty \right\} = 1,$$

$$\mathbb{P} \left\{ \lim_{t \rightarrow 0} \frac{|y(u, t) - u|}{\sqrt[3]{t} (\ln \frac{1}{t})^{\frac{1}{2}+\varepsilon}} = 0 \right\} = \mathbb{P} \left\{ \overline{\lim}_{t \rightarrow 0} \frac{|y(u, t) - u|}{\sqrt[3]{t} (\ln \frac{1}{t})^{-\frac{1}{2}-\varepsilon}} = +\infty \right\} = 1.$$

$$\mathbb{E}m(u, t) \leq C \sqrt[3]{t}, \quad \mathbb{E} \frac{1}{m(u, t)} \leq \frac{C}{\sqrt[3]{t}}, \quad t \in (0, T].$$

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$$\mathbb{E}m(u, t) \leq C \sqrt[3]{t}.$$

Let  $\theta > 0$ ,  $\lambda < 1$  and  $t_n = \lambda^n$ ,  $n \in \mathbb{N}$ .

$$\begin{aligned} & \mathbb{P}\{m(u, t) > \theta \varphi(t), \text{ for some } t \in (t_{n+1}, t_n]\} \\ & \leq \mathbb{P}\{m(u, t_n) > \theta \varphi(t_{n+1})\} \leq \frac{1}{\theta \varphi(t_{n+1})} \mathbb{E}m(u, t_n) \\ & \leq \frac{C t_n^{\frac{1}{2\alpha+1}}}{\theta t_{n+1}^{\frac{1}{2\alpha+1}} \left( \ln \frac{1}{t_{n+1}} \right)^{1+\epsilon}} = \frac{C}{\theta \lambda^{\frac{1}{2\alpha+1}} \left( \ln \frac{1}{\lambda} \right)^{1+\epsilon} (n+1)^{1+\epsilon}}. \end{aligned}$$

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$$\mathbb{E}m(u, t) \leq C \sqrt[3]{t}, \quad \mathbb{E} \frac{1}{m(u, t)} \leq \frac{C}{\sqrt[3]{t}}, \quad t \in (0, T].$$

$$y(u, t) - u = w \left( \int_0^t \frac{ds}{m(u, s)} \right) \quad + \quad \text{LIL for Wiener process.}$$

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## Rescaling property

$$y_\rho(u, t) = \frac{1}{\rho} y(u\rho, t\rho^3), \quad u \in [0, 1/\rho], \quad \rho > 0.$$

$y_\rho$  satisfies the same properties as  $y$ , but is defined for  $u \in [0, 1/\rho]$ , i.e. it describes the system of particles starting from  $[0, 1/\rho]$ . Moreover

$$m_\rho(u, t) = \text{Leb}\{v : y_\rho(u, t) = y_\rho(v, t)\} = \frac{1}{\rho} m(u\rho, t\rho^3).$$

If  $\mathbb{E}m_\rho(0, T) \leq C$  for all  $\rho \in (0, 1]$ , then

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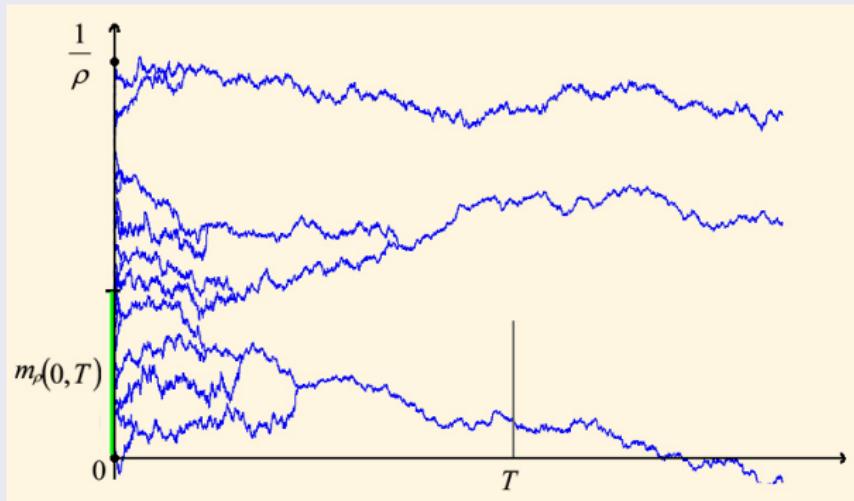
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Estimation  $\mathbb{E}m_\rho(0, T) \leq C$



$$\mathbb{E}m_\rho(0, T) = \mathbb{E} \int_0^{\frac{1}{\rho}} \mathbb{I}_{\{y_\rho(0, T) = y_\rho(u, T)\}} du = \int_0^{\frac{1}{\rho}} \mathbb{P}\{y_\rho(0, T) = y_\rho(u, T)\} du.$$

# Estimation $\mathbb{E}m_\rho(0, T) \leq C$

Set  $M_\rho(u, t) = y_\rho(u, t) - y_\rho(0, t)$ .

Using the Paley-Zygmund inequality, we can estimate

$$\begin{aligned}
 \mathbb{P}\{y_\rho(0, T) = y_\rho(u, T)\} &= 1 - \mathbb{P}\{M_\rho(u, T) > 0\} \\
 &\leq 1 - \frac{(\mathbb{E}M_\rho(u, T))^2}{\mathbb{E}M_\rho^2(u, T)} = 1 - \frac{u^2}{\mathbb{E}M_\rho^2(u, T)} \\
 &= 1 - \frac{u^2}{\text{Var } M_\rho(u, T) + u^2} = \frac{\text{Var } M_\rho(u, T)}{\text{Var } M_\rho(u, T) + u^2}.
 \end{aligned}$$

Estimation  $\mathbb{E}m_\rho(0, T) \leq C$

$$\begin{aligned}
 \text{Var } M_\rho(u, T) &= \mathbb{E} (M_\rho(u, T) - u)^2 \\
 &\leq \int_0^T \mathbb{E} \frac{1}{m_\rho(u, s)} ds + \int_0^T \mathbb{E} \frac{1}{m_\rho(0, s)} ds \leq C_1,
 \end{aligned}$$

where we used

$$\mathbb{E} \frac{1}{m_\rho(u, s)} < \frac{C}{\sqrt[3]{s}}, \quad s \in (0, T].$$

Thus

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## Description of the model

Let  $\mu$  be a probability measure on  $\mathbb{R}$ ,  $A = \text{supp } \mu$ .

$z(\xi, t)$  is the position of particles at time  $t$  which starts from  $\xi \in A$ .

$z(\xi, t)$ ,  $\xi \in A$ ,  $t \in [0, T]$ , should satisfy the same condition as  $y$ :

- 1)  $z(\xi, \cdot)$  is a continuous square integrable martingale;
- 2)  $z(\xi, 0) = \xi$ ,  $\xi \in A$ ;
- 3)  $z(\xi, t) \leq z(\eta, t)$ ,  $\xi < \eta$ ;
- 4)  $(z(\xi, \cdot))_t = \int_0^t \frac{ds}{m_\mu(\xi, s)}$ ;  $m_\mu(\xi, t) = \mu\{\eta : z(\eta, t) = z(\xi, t)\}$ ;
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## Description of the model

Let  $\mu$  be a probability measure on  $\mathbb{R}$ ,  $A = \text{supp } \mu$ .

$z(\xi, t)$  is the position of particles at time  $t$  which starts from  $\xi \in A$ .

$z(\xi, t)$ ,  $\xi \in A$ ,  $t \in [0, T]$ , should satisfy the same condition as  $y$ :

- 1)  $z(\xi, \cdot)$  is a continuous square integrable martingale;
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## Alternative description

Let  $\mu$  be the push forward of the Lebesgue measure on  $(0, 1)$

$$\mu = \text{Leb}|_{(0,1)} \circ g^{-1},$$

where  $g : (0, 1) \rightarrow \mathbb{R}$  is a non-decreasing right-continuous function.

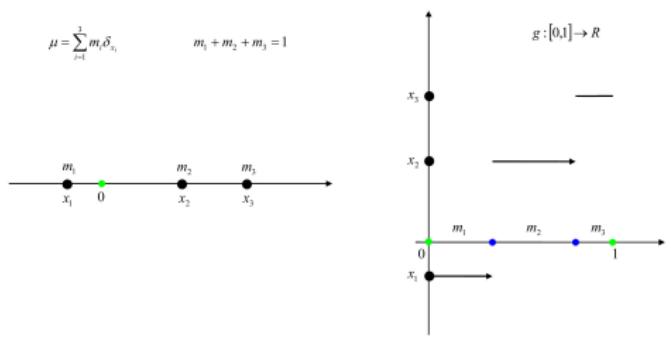
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## Example



## Alternative description

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### Theorem 3 (K. '16)

Let  $\|g\|_{L_{2+\varepsilon}} < \infty$  (it's equivalent to  $\int_{\mathbb{R}} |\xi|^{2+\varepsilon} \mu(d\xi) < \infty$ ). Then there exists  $y(u, t)$ ,  $u \in (0, 1)$ ,  $t \in [0, T]$ , in  $D([0, 1], C[0, T])$  satisfying

- 1°)  $y(u, \cdot)$  is a continuous square integrable martingale;
- 2°)  $y(u, 0) = g(u)$ ,  $u \in (0, 1)$ ;
- 3°)  $y(u, t) \leq y(v, t)$ ,  $u < v$ ;
- 4°)  $\langle y(u, \cdot) \rangle_t = \int_0^t \frac{ds}{m(u, s)}$ ,  $m(u, t) = \text{Leb}\{v : y(v, t) = y(u, t)\}$ ;
- 5°)  $\langle y(u, \cdot), y(v, \cdot) \rangle_t = 0$ ,  $t \leq \tau_{u,v} = \inf\{t : y(u, t) = y(v, t)\} \wedge T$ .

## Alternative description

Set  $A = \{g(u), u \in (0, 1)\}$  and

$$z(g(u), \cdot) = y(u, \cdot), \quad u \in (0, 1).$$

Since  $g(u) = g(v)$  implies  $y(u, \cdot) = y(v, \cdot)$ , the process  $z$  is well-defined.

If  $\xi = g(u)$ ,

$$z(\xi, 0) = z(g(u), 0) = y(u, 0) = g(u) = \xi$$

$$\begin{aligned} m_\mu(\xi, t) &= \mu\{\eta : z(\eta, t) = z(\xi, t)\} \\ &= \text{Leb}\{v : z(g(v), s) = y(u, s)\} \\ &= \text{Leb}\{v : y(v, s) = y(u, s)\} = m(u, t). \end{aligned}$$

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## Alternative description

### Asymptotic behavior (K. '16)

Let  $g'(u) \sim |u - u_0|^{\alpha-1}$ ,  $\alpha > \frac{1}{2}$ . Then for all  $\varepsilon > 0$

$$\mathbb{P} \left\{ \lim_{t \rightarrow 0} \frac{m(u_0, t)}{t^{\frac{1}{2\alpha+1}} (\ln \frac{1}{t})^{1+\epsilon}} = 0 \right\} = \mathbb{P} \left\{ \lim_{t \rightarrow 0} \frac{m(u_0, t)}{t^{\frac{1}{2\alpha+1}} (\ln \frac{1}{t})^{-1-\epsilon}} = +\infty \right\} = 1,$$

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### Remark

If  $\mu$  has the density  $h$  and  $h(\xi) \sim |\xi - \xi_0|^\beta$ , where  $\xi_0 = g(u_0)$  and  $-1 < \beta < 1$ , then  $\alpha = \frac{1}{1+\beta}$  and

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Thank you for your attention!