

# Analysis of stochastic flows

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- $(X, \rho)$  is a complete separable metric space
- $\{\phi_{s,t}; 0 \leq s \leq t\}$  is a family of random mappings in  $X$
- For any  $0 \leq s_1 \leq s_2 \leq \dots \leq s_n < \infty$   $\phi_{s_1,s_2}, \dots, \phi_{s_{n-1},s_n}$  are independent.
- For any  $s, t, r \geq 0$   $\phi_{s,t}$  and  $\phi_{s+r,t+r}$  are equidistributed.
- For any  $r \leq s \leq t$  and  $u \in X$   $\phi_{r,s}\phi_{s,t}(u) = \phi_{r,t}(u)$ ,  $\phi_{r,r}$  is the identity map.
- The random process  $\{x(u, t) = \phi_{0,t}(u); t \geq 0\}$  describes the motion of a particle, which starts from the point  $u$

## Example

- $X = R^d$ , for every  $u$  the process  $x(u, t)$  is a solution to the Cauchy problem for SDE

$$dx(u, t) = a(x(u, t))dt + b(x(u, t))dw(t), x(u, 0) = u.$$

## Example

- The Harris flow of Brownian particles

$X = R$ ,  $\{x(u, t); u \in X, t \geq 0\}$  is a family of Brownian martingales with respect to a common filtration,  $x$  is order-preserving and

$$d \langle x(u_1, \bullet), x(u_2, \bullet) \rangle = \varphi(x(u_1, t) - x(u_2, t))dt,$$

where  $\varphi$  is a positive definite function.

- The Arratia flow

$$\varphi(x) = 1_{\{x=0\}}$$

## Properties

- The law of the one-point motions does not determine the flow
- Possibility of coalescence
- The map  $x(\bullet, t)$  can be discontinuous

# The relationships with SDE

$$dx_\varepsilon(u, t) = \int_{\mathbb{R}} \psi_\varepsilon(x_\varepsilon(u, t) - p) W(dp, dt)$$

$$\int_{\mathbb{R}} \psi_\varepsilon^2(u) du = 1, \text{ supp } \psi_\varepsilon \subset [-\varepsilon, \varepsilon]$$

## Theorem

*A. A. Dorogovtsev, 2005. The  $n$ -point motions of  $x_\varepsilon$  converge to the  $n$ -point motions of the Arratia flow when  $\varepsilon \rightarrow 0$ .*

## Theorem

*T. V. Malovichko, 2008. The same statement when*

$$\psi_\varepsilon^2 \rightarrow p_1 \delta_{-1} + p_2 \delta_1$$

- Does the flow of Brownian particles inherit the properties of solutions to SDE with a Gaussian noise? (Girsanov theorem, Large Deviations Principle, Clark representation, Krylov-Veretennikov expansion)
- Properties of the mappings  $\phi_{0,t} : X \rightarrow X$

$$dx_\varepsilon(u, t) = \sqrt{\varepsilon} \int \varphi(x_\varepsilon(u, t) - p) W(dp, dt)$$

$$x_\varepsilon(u, 0) = u$$

## Theorem

*Dorogovtsev, Ostapenko, 2009. Suppose that  $\varphi = \varphi_1 \star \varphi_2$ ,  $\varphi_i \in S(\mathbb{R})$ ,  $\mu = N(0, 1)$ .  $x_\varepsilon$  satisfies LDP in  $C([0; 1], L_2(\mathbb{R}, \mu))$  with the rate function*

$$I(z) = \inf \frac{1}{2} \int \int_0^{+\infty} h^2(p, t) dp dt,$$

$$dz(u, t) = \int \varphi(z(u, t) - p) h(p, t) dp dt, \quad z(u, 0) = u, \quad u \in \mathbb{R}$$

*$I(z) = +\infty$  if there is no such  $h$ .*

$$0 = u_0 < u_1 < \dots < u_n = 1,$$

$$\tau(u_0) = T, \tau(u_{k+1}) = \inf\{s : x(u_{k+1}, s) = x(u_k, s)\} \wedge T$$

## Theorem

*Dorogovtsev, 2006. The total time of free motion in the Arratia flow is finite.*

$$\sup \sum_{k=0}^n \tau(u_k) < +\infty \text{ a.s.}$$



## Corollary

*There exist the integrals*

$$\int_0^1 \int_0^{\tau(u)} a(x(u, s)) ds = \lim_{\max u_{k+1} - u_k \rightarrow 0} \sum_{k=0}^n \int_0^{\tau(u_k)} a(x(u_k, s)) ds$$

$$\begin{aligned} \int_0^1 \int_0^{\tau(u)} a(x(u, s)) dx(u, s) &= \\ &= L_2 - \lim_{\max u_{k+1} - u_k \rightarrow 0} \sum_{k=0}^n \int_0^{\tau(u_k)} a(x(u_k, s)) dx(u, s) \end{aligned}$$

## Theorem

*Dorogovtsev, 2006. The distribution of the Arratia flow with the drift  $a$  is absolutely continuous with respect to the distribution of the Arratia flow with the density*

$$\exp\left\{\int_0^1 \int_0^{\tau(u)} a(x(u,s)) dx(u,s) - \frac{1}{2} \int_0^1 \int_0^{\tau(u)} a^2(x(u,s)) ds\right\}$$

## Theorem

*Dorogovtsev, Ostapenko, 2009. The family  $\{x_\varepsilon(u, t) = x(u, \varepsilon t), \varepsilon > 0\}$  satisfies LDP with the rate function*

$$I(x) = \frac{1}{2} \int_0^1 \int_0^{\tau(u)} h'_t(u, t)^2 dt$$

## Theorem

*Dorogovtsev, 2008. The linear combinations of functions*

$$\exp\left\{\int_0^1 \int_0^{\tau(u)} a(u,s) dx(u,s) - \frac{1}{2} \int_0^1 \int_0^{\tau(u)} a^2(u,s) ds\right\}, a \in C([0, 1]^2)$$

*are dense in the space of all square-integrable functionals from the Arratia flow.*

## Definition

(A. V. Skorokhod) Strong random operator is a continuous in probability linear map from the space  $H$  to the space of all random elements in  $H$ .

## Example

- $H = L_2([0; 1])$ ,  $Af(t) = \int_0^t f(s)dw(s)$
- $H = L_2([0; 1])$ ,  $Af = f(\theta)$ ,  $\theta$  is uniformly distributed

*In general a strong random operator is not a randomly chosen bounded linear operator. It can be unbounded with probability one.*

A. V. Skorokhod. Random linear operators. Kiev: Nauk. dumka, 1978.

## Definition

A family of strong random operators  $\{G_{s,t}, 0 \leq s \leq t < \infty\}$  is referred to as a semigroup if the following conditions hold:

- 1 For any  $s, t, r \geq 0$ :  $G_{s,t}$  and  $G_{s+r,t+r}$  are equidistributed.
- 2 For any  $x \in H$ :  $E\|G_{0,t}x - x\|^2 \mapsto 0, n \mapsto \infty$ .
- 3 For any  $0 \leq s_1 \leq \dots \leq s_n < \infty$ :  $G_{s_1,s_2}, \dots, G_{s_{n-1},s_n}$  are independent.
- 4 For any  $r \leq s \leq t$ :  $G_{r,s}G_{s,t} = G_{r,t}$ ,  $G_{r,r} = I$ , where  $I$  is the identity operator.

# Semigroups related to stochastic flows

## Example

Stochastic semigroup related to the stochastic flow

$$G_{s,t}f(u) = f(\phi_{s,t}(u))$$

## Theorem

(Krylov-Veretennikov expansion) If

$$dx(u, t) = a(x(u, t))dt + b(x(u, t))dw(t), \quad x(u, 0) = u$$

then

$$\begin{aligned} f(x(u, t)) &= \\ &= \sum_{n=0}^{\infty} \int_{0 \leq s_1 \leq \dots \leq s_n \leq t} T_{t-s_n} B T_{s_n-s_{n-1}} B \dots T_{s_1} f(u) dw(s_1) \dots dw(s_n). \end{aligned}$$

Here  $\{T_t\}$  is a transition semigroup and  $B = b \frac{d}{du}$ .

## Theorem

*Dorogovtsev, 2010. If  $\{G_{s,t}\}$  is a multiplicative functional from the Wiener process  $W$ , then*

$$G_{0,t} = \sum_{n=0}^{\infty} \int_{0 \leq s_1 \leq \dots \leq s_n \leq t} Q_{t-s_n} B Q_{s_n-s_{n-1}} B \dots Q_{s_1} dW(s_1) \dots dW(s_n),$$

$$Q_t = EG_{0,t}, \quad B = \lim_{t \rightarrow 0+} \frac{1}{t} EG_{0,t} W(t).$$

**Remark.**  $G_{0,t}$  is a “solution” to the equation

$$dG_{0,t} = AG_{0,t}dt + BG_{0,t}dW(t)$$



# Semigroups of finite-dimensional projections

**Remark.** Suppose that  $\{G_t, 0 \leq t < \infty\}$  is a strongly continuous semigroup of bounded operators in a separable Banach space  $\mathcal{B}$ . Assume that  $\dim G_t(\mathcal{B}) < \infty$  for every  $t > 0$ . Then  $\dim \mathcal{B} < \infty$ .

## Example

Let  $H$  be a Hilbert space with an orthonormal basis  $\{e_k, k \geq 1\}$ . Consider the sequence  $\{n_k, k \geq 1\}$  of independent Poisson processes with intensities  $\{\lambda_k, k \geq 1\}$ . Suppose that

$$\forall \rho > 0: \sum_{k=1}^{\infty} \exp(-\rho \lambda_k) < +\infty,$$

$$v_{s,t}^k = \begin{cases} 0, & n_k(t) - n_k(s) > 0, \\ 1, & n_k(t) - n_k(s) = 0. \end{cases}$$

Then

$$G_{s,t}(u) = \sum_{k=1}^{\infty} (u, e_k) v_{s,t}^k e_k.$$

# Semigroups of finite-dimensional projections

## Theorem

*Let  $\{G_{s,t}, 0 \leq s \leq t < \infty\}$  be a semigroup of random finite-dimensional projections in a separable Hilbert space  $H$ . Then there exists an orthonormal basis  $\{e_k, k \geq 1\}$  in  $H$  and Poisson processes  $\{n_k, k \geq 1\}$  which have jointly independent increments, such that*

$$G_{s,t}(u) = \sum_{k=1}^{\infty} (u, e_k) v_{s,t}^k e_k,$$

*where for every  $k$*

$$v_{s,t}^k = \begin{cases} 0, & n_k(t) - n_k(s) > 0, \\ 1, & n_k(t) - n_k(s) = 0. \end{cases}$$

# Widths of compact sets

Let  $\{G_{s,t}, 0 \leq s \leq t < \infty\}$  be a random semigroup of finite-dimensional projections and let  $K$  be a compact subset of  $H$ . The behavior of the value

$$\zeta_K(t) = \max_{x \in K} \|x - G_{0,t}x\|$$

as  $t \rightarrow 0$  describes the geometry of the semigroup  $\{G_{s,t}, 0 \leq s \leq t < \infty\}$ .

## Example

Suppose that  $\lambda_n = n$ ,  $n \geq 1$ , Poissonian processes are independent and

$$K = \{x : (x, e_n)^2 \leq \frac{1}{n^2}, n \geq 1\}.$$

$$\varsigma_K(t)^2 = \sum_{n=1}^{\infty} \frac{\xi_n(t)}{n^2}$$

$$P - \lim_{t \rightarrow 0} \frac{\varsigma_K(t)}{\sqrt{t \ln t}} = 1$$

## Example

$$K = \{x : \sum_{n=1}^{\infty} n^2 (x, e_n)^2 \leq 1\}.$$

$$\varsigma_K(t)^2 = \max_{n: \xi_n(t)=0} \frac{1}{n^2}$$

$$\liminf_{t \rightarrow 0} \varsigma_K(t) \varphi(t) \geq 1, \quad \varphi(t) = \sqrt{\frac{2}{t} \ln t}.$$

$$d_n(K) = \inf_{\dim L=n} \max_{x \in K} \rho(x, K)$$

## Example

Define  $\alpha(t) = \dim G_{0,t}(H)$ .

$$\limsup_{t \rightarrow 0} \frac{t\alpha(t)}{2|\ln t|} \leq 1$$

$$\liminf_{t \rightarrow 0} \alpha(t)t|\ln t| \geq \frac{1}{2}.$$