

# Large deviations for Brownian intersection measures

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## Brownian paths do intersect

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- Look at their *path intersections*:

$$S_t = \bigcap_{i=1}^p W_i[0, t_i] \quad t = (t_1, \dots, t_p) \in (0, \infty)^p$$

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- Dvoretzky, Erdős, Kakutani and Taylor showed  $S_t$  is non-empty with positive probability iff

$$\begin{cases} d = 2, p \in \mathbb{N} \\ d = 3, p = 2 \\ d \geq 4, p = 1. \end{cases}$$

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- **Goal:** Make precise the above as  $t \uparrow \infty$  (in particular,  $d \geq 2$ ).

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- Le Gall (1986) looked at **Wiener sausages**:

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$$s_d(\epsilon) = \begin{cases} \pi^{-p} \log^p(\frac{1}{\epsilon}) & \text{if } d = 2 \\ (2\pi\epsilon)^{-2} & \text{if } d = 3 \text{ and } p = 2 \\ \frac{2}{\omega_d(d-2)} \epsilon^{2-d} & \text{if } d \geq 3 \text{ and } p = 1. \end{cases}$$

# Wiener Sausages

Intersection measure: scaling limit of Lebesgue measure on sausages

- Le Gall shows limit  $\epsilon \downarrow 0$  gives the **Brownian intersection measure**:

$$\lim_{\epsilon \rightarrow 0} \ell_{\epsilon,t}(A) = \ell_t(A) \text{ in } L^q \text{ for } q \in [1, \infty)$$

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- **Want to study:** Behavior of  $\frac{1}{t} \ell_t^{(i)}$ , as  $t \uparrow \infty$ .

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Path densities show up

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**Our Goal:** Similar statement for **intersection measure**  $\ell_t$ , for *large*  $t$ ?

# Large deviations: diverging time

occupation measure and intersection measure

- $\ell_t^{(1)}, \dots, \ell_t^{(p)}$  occupation measures of  $p$  paths running until time  $t$  in a bounded domain  $B$  until first exit times  $\tau_1, \dots, \tau_p$ .

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- Make sure no path exits  $B$  before time  $t$ :  
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Intersection densities as product of occupation densities

$$\mathbb{P}_t \left( \frac{\ell_t}{t^p} \approx \mu; \frac{\ell_t^{(1)}}{t} \approx \mu_1, \dots, \frac{\ell_t^{(p)}}{t} \approx \mu_p \right)$$

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else,

$$I = \infty \quad \text{identically}$$

# Large deviations: diverging time

Extension of classical theory

## Theorem (König/M (2011))

The family of tuples  $(\frac{\ell_t}{t^p}; \frac{\ell_t^{(1)}}{t}, \dots, \frac{\ell_t^{(p)}}{t})$  satisfies a LDP under  $\mathbb{P}_t$ , as  $t \uparrow \infty$ , with rate function

$$I(\mu; \mu_1, \dots, \mu_p) = \frac{1}{2} \sum_{i=1}^p \|\nabla \psi_i\|_2^2$$

if  $\mu$  and  $\mu_1, \dots, \mu_p$  have densities  $\psi^{2p}$  and  $\psi_1^2, \dots, \psi_p^2$  respectively,  $\psi_i \in H_0^1(B)$ ,  $\|\psi_i\|_2 = 1$  and  $\psi^{2p} = \prod_{i=1}^p \psi_i^2$ , else  $I = \infty$ .

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## Corollary

The family of measures  $(\frac{\ell_t}{t^p})$  satisfies a large deviation principle, under  $\mathbb{P}_t$ , as  $t \uparrow \infty$ , with rate function

$$J(\mu) = \inf \left\{ \frac{1}{2} \sum_{i=1}^p \|\nabla \psi_i\|_2^2 : \psi_i \in H_0^1(B), \|\psi_i\|_2 = 1, \prod_{i=1}^p \psi_i^2 = \frac{d\mu}{dx} \right\}$$

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$p = 1$ : We recover classical Donsker-Varadhan theory for one path.

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[König and Mörters (2001)]:

$$\lim_{a \rightarrow \infty} a^{-\frac{1}{p}} \log \mathbb{P}[\ell(U) > a] = -\Theta(U)$$

for

$$\Theta(U) = \inf \left\{ \frac{p}{2} \|\nabla \psi\|_2^2 : \psi \in H_0^1(B), \|1_U \psi\|_{2p}^2 = 1 \right\}.$$

# Minimisers and path behavior

## Euler-Lagrange equations

- Minimiser(s) to  $\Theta(U)$  exist(s).
- Every minimising  $\psi$  solves

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**Upshot:**  $\psi^{2p}$  should be the large- $a$  density of the **intersection measure** on  $U$ .

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## Law of large masses

- Let  $L = \frac{\ell}{\ell(U)}$  be the normalised probability on  $U$ .

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[König and Mörters (2005)]

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- **Large deviations:** What is the exponential decay rate?

# Intersection measure until exit times

Large deviations: diverging mass

## Theorem (König/M (2011))

The normalized probability measures  $L = \frac{\ell}{\ell(U)}$  satisfy a large deviation principle under  $\mathbb{P}(\cdot | \ell(U) > a)$ , as  $a \uparrow \infty$ , with rate function

$$\Lambda(\mu) = \inf \left\{ \frac{1}{2} \sum_{i=1}^p \|\nabla \psi_i\|_2^2 : \psi_i \in H_0^1(B), \prod_{i=1}^p \psi_i^2 = \frac{d\mu}{dx} \right\} - \Theta(U).$$

# Outlook

Questions we can chew on

- Study the variational formula for the rate function:

$$J(\mu) = \inf \left\{ \frac{1}{2} \sum_{i=1}^p \|\nabla \psi_i\|_2^2 : \psi_i \in H_0^1(B), \|\psi_i\|_2 = 1, \prod_{i=1}^p \psi_i^2 = \frac{d\mu}{dx} \right\}$$

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- Extend it to unbounded domains.: For  $p = 2, B = \mathbb{R}^3$