

# Causal optimal transport and the Brownian transport map

---

Dan Mikulincer

MIT

Joint work with Yair Shenfeld

# Transport maps

Let  $X \sim \mu$  be a measure on  $\mathbb{R}^d$  and let  $G \sim \gamma$  stand for the standard Gaussian.

If  $\varphi$  is such that  $\varphi(G) \stackrel{\text{law}}{=} X$ , we call  $\varphi$  a *transport map*.

The existence and properties of such maps are useful for:

- Generative models and sampling algorithms.
- Understanding analytic properties of  $\mu$ .

# Transport maps

Let  $X \sim \mu$  be a measure on  $\mathbb{R}^d$  and let  $G \sim \gamma$  stand for the standard Gaussian.

If  $\varphi$  is such that  $\varphi(G) \stackrel{\text{law}}{=} X$ , we call  $\varphi$  a *transport map*.

The existence and properties of such maps are useful for:

- Generative models and sampling algorithms.
- Understanding analytic properties of  $\mu$ .

## Definition (Wasserstein distance between $\mu$ and $\gamma$ )

$$\mathcal{W}_2(\mu, \gamma) := \inf_{\pi} \left\{ \mathbb{E}_{\pi} [\|x - y\|^2] \right\}^{1/2}$$

where  $\pi$  ranges over all possible couplings of  $\mu$  and  $\gamma$ .

Brenier 87': There exists a transport map  $\psi^{\text{opt}} : \mathbb{R}^d \rightarrow \mathbb{R}^d$ :

$$\mathbb{E} [\|\psi^{\text{opt}}(G) - G\|^2] = \mathcal{W}_2^2(\mu, \gamma).$$

Caffarelli 00': If  $\mu$  is more log-concave than  $\gamma_d$ ,  $\psi^{\text{opt}}$  is 1-Lipschitz.

(strong log-concavity:  $-\nabla^2 \log \left( \frac{d\mu}{dx}(x) \right) \succeq \text{Id.}$ )

## Definition (Wasserstein distance between $\mu$ and $\gamma$ )

$$\mathcal{W}_2(\mu, \gamma) := \inf_{\pi} \left\{ \mathbb{E}_{\pi} [\|x - y\|^2] \right\}^{1/2}$$

where  $\pi$  ranges over all possible couplings of  $\mu$  and  $\gamma$ .

Brenier 87': There exists a transport map  $\psi^{\text{opt}} : \mathbb{R}^d \rightarrow \mathbb{R}^d$ :

$$\mathbb{E} [\|\psi^{\text{opt}}(G) - G\|^2] = \mathcal{W}_2^2(\mu, \gamma).$$

Caffarelli 00': If  $\mu$  is more log-concave than  $\gamma_d$ ,  $\psi^{\text{opt}}$  is 1-Lipschitz.

(strong log-concavity:  $-\nabla^2 \log \left( \frac{d\mu}{dx}(x) \right) \succeq \text{Id.}$ )

## Definition (Wasserstein distance between $\mu$ and $\gamma$ )

$$\mathcal{W}_2(\mu, \gamma) := \inf_{\pi} \left\{ \mathbb{E}_{\pi} [\|x - y\|^2] \right\}^{1/2}$$

where  $\pi$  ranges over all possible couplings of  $\mu$  and  $\gamma$ .

Brenier 87': There exists a transport map  $\psi^{\text{opt}} : \mathbb{R}^d \rightarrow \mathbb{R}^d$ :

$$\mathbb{E} [\|\psi^{\text{opt}}(G) - G\|^2] = \mathcal{W}_2^2(\mu, \gamma).$$

Caffarelli 00': If  $\mu$  is more log-concave than  $\gamma_d$ ,  $\psi^{\text{opt}}$  is 1-Lipschitz.

(strong log-concavity:  $-\nabla^2 \log \left( \frac{d\mu}{dx}(x) \right) \succeq \text{Id}.$ )

## Definition (Wasserstein distance between $\mu$ and $\gamma$ )

$$\mathcal{W}_2(\mu, \gamma) := \inf_{\pi} \left\{ \mathbb{E}_{\pi} [\|x - y\|^2] \right\}^{1/2}$$

where  $\pi$  ranges over all possible couplings of  $\mu$  and  $\gamma$ .

Brenier 87': There exists a transport map  $\psi^{\text{opt}} : \mathbb{R}^d \rightarrow \mathbb{R}^d$ :

$$\mathbb{E} [\|\psi^{\text{opt}}(G) - G\|^2] = \mathcal{W}_2^2(\mu, \gamma).$$

Caffarelli 00': If  $\mu$  is more log-concave than  $\gamma_d$ ,  $\psi^{\text{opt}}$  is 1-Lipschitz.

(strong log-concavity:  $-\nabla^2 \log \left( \frac{d\mu}{dx}(x) \right) \succeq \text{Id}.$ )

**Gaussian Poincaré inequality:** For any test function  $f$ ,

$$\text{Var}(f(G)) \leq \mathbb{E} [\|\nabla f(G)\|^2] .$$

In general,  $X \sim \mu$  satisfies a Poincaré inequality with constant  $C_p(\mu) > 0$ , if,

$$\text{Var}(f(X)) \leq C_p(\mu) \mathbb{E} [\|\nabla f(X)\|^2] .$$



**Gaussian Poincaré inequality:** For any test function  $f$ ,

$$\text{Var}(f(G)) \leq \mathbb{E} [\|\nabla f(G)\|^2] .$$

In general,  $X \sim \mu$  satisfies a Poincaré inequality with constant  $C_p(\mu) > 0$ , if,

$$\text{Var}(f(X)) \leq C_p(\mu) \mathbb{E} [\|\nabla f(X)\|^2] .$$

# An inequality of Brascamp and Lieb

## Theorem (Brascamp-Lieb 76')

*If  $\mu$  is more log-concave than  $\gamma_d$ , then  $C_p(\mu) \leq 1$ .*

*Proof (Cordero-Erausquin 02').*

$$\begin{aligned}\mathrm{Var}_\mu(f) &= \mathrm{Var}_{\gamma_d}(f \circ \psi^{\mathrm{opt}}) \leq \mathbb{E}_{\gamma_d} [\|\nabla(f \circ \psi^{\mathrm{opt}})\|^2] \\ &\leq \mathbb{E}_{\gamma_d} [\|\nabla\psi^{\mathrm{opt}}\|^2 \|\nabla f(\psi^{\mathrm{opt}})\|^2] = \mathbb{E}_\mu [\|\nabla f\|^2].\end{aligned}$$

□

# An inequality of Brascamp and Lieb

## Theorem (Brascamp-Lieb 76')

If  $\mu$  is more log-concave than  $\gamma_d$ , then  $C_p(\mu) \leq 1$ .

## Proof (Cordero-Erausquin 02').

$$\begin{aligned}\mathrm{Var}_\mu(f) &= \mathrm{Var}_{\gamma_d}(f \circ \psi^{\mathrm{opt}}) \leq \mathbb{E}_{\gamma_d} [\|\nabla (f \circ \psi^{\mathrm{opt}})\|^2] \\ &\leq \mathbb{E}_{\gamma_d} [\|\nabla \psi^{\mathrm{opt}}\|^2 \|\nabla f(\psi^{\mathrm{opt}})\|^2] = \mathbb{E}_\mu [\|\nabla f\|^2].\end{aligned}$$

□

# Bounded log-concave

If  $\mu$  is log-concave, but compactly supported on a ball of diameter  $R$ , then  $C_p(\mu) \lesssim R^2$ . Several proofs exist:

- Localization (Payne-Weinberger)
- Refined Brascamp-Lieb (Kolesnikov-Milman)
- Moment Maps (Klartag)

## Question

For such  $\mu$  is it necessarily true that there exists an  $R$ -Lipschitz  $\varphi$  with  $\varphi_*\gamma_d = \mu$ ?

# Bounded log-concave

If  $\mu$  is log-concave, but compactly supported on a ball of diameter  $R$ , then  $C_p(\mu) \lesssim R^2$ . Several proofs exist:

- Localization (Payne-Weinberger)
- Refined Brascamp-Lieb (Kolesnikov-Milman)
- Moment Maps (Klartag)

## Question

For such  $\mu$  is it necessarily true that there exists an  $R$ -Lipschitz  $\varphi$  with  $\varphi_*\gamma_d = \mu$ ?

# Motivation

A positive answer will not only recover known result but will also imply:

1. Dimension-free  $\Phi$ -Sobolev inequalities (generalizing both Poincaré and log-Sobolev).
2. Bounds for higher eigenvalues of the weighted Laplacian.
3. Isoperimetric inequalities.
4. Improved rates of convergence for the CLT.

# Motivation

A positive answer will not only recover known result but will also imply:

1. Dimension-free  $\Phi$ -Sobolev inequalities (generalizing both Poincaré and log-Sobolev).
2. Bounds for higher eigenvalues of the weighted Laplacian.
3. Isoperimetric inequalities.
4. Improved rates of convergence for the CLT.

# Motivation

A positive answer will not only recover known result but will also imply:

1. Dimension-free  $\Phi$ -Sobolev inequalities (generalizing both Poincaré and log-Sobolev).
2. Bounds for higher eigenvalues of the weighted Laplacian.
3. Isoperimetric inequalities.
4. Improved rates of convergence for the CLT.



# Motivation

A positive answer will not only recover known result but will also imply:

1. Dimension-free  $\Phi$ -Sobolev inequalities (generalizing both Poincaré and log-Sobolev).
2. Bounds for higher eigenvalues of the weighted Laplacian.
3. Isoperimetric inequalities.
4. Improved rates of convergence for the CLT.

# Gaussian mixtures

We call  $\mu = \gamma_d \star \nu$  a Gaussian mixture. It was recently proved by Bardet, Gozlan, Malrieu and Zitt that if  $\text{diam}(\text{supp}(\nu)) \leq R$ , then

$$C_p(\mu) \lesssim e^{R^2}.$$

Later, Chen, Chewi and Niles-Weed extended the result to the log-Sobolev inequality.

## Question

Suppose that  $\mu = \gamma_d \star \nu$  and  $\text{diam}(\text{supp}(\nu)) \leq R$ . Is there an  $e^{R^2}$ -Lipschitz  $\varphi$  with  $\varphi_* \gamma_d = \mu$ ?

# Gaussian mixtures

We call  $\mu = \gamma_d \star \nu$  a Gaussian mixture. It was recently proved by Bardet, Gozlan, Malrieu and Zitt that if  $\text{diam}(\text{supp}(\nu)) \leq R$ , then

$$C_p(\mu) \lesssim e^{R^2}.$$

Later, Chen, Chewi and Niles-Weed extended the result to the log-Sobolev inequality.

## Question

Suppose that  $\mu = \gamma_d \star \nu$  and  $\text{diam}(\text{supp}(\nu)) \leq R$ . Is there an  $e^{R^2}$ -Lipschitz  $\varphi$  with  $\varphi_* \gamma_d = \mu$ ?

Let  $\mu$  be log-concave and isotropic,

$$\int_{\mathbb{R}^d} x d\mu(x) = 0 \quad \int_{\mathbb{R}^d} x \otimes x d\mu(x) = \text{Id}.$$

A famous conjecture of Kannan-Lovász-Simonovits postulates,

$$C_p(\mu) \leq C.$$

Current best bound, due to Klartag and Lehec (building upon Chen):  $C_p(\mu) \leq \text{polylog}(d)$ .

It seems natural to ask whether we can find a Lipschitz map  $\varphi$  with  $\varphi_* \gamma_d = \mu$ ?

Let  $\mu$  be log-concave and isotropic,

$$\int_{\mathbb{R}^d} x d\mu(x) = 0 \quad \int_{\mathbb{R}^d} x \otimes x d\mu(x) = \text{Id}.$$

A famous conjecture of Kannan-Lovász-Simonovits postulates,

$$C_p(\mu) \leq C.$$

Current best bound, due to Klartag and Lehec (building upon Chen):  $C_p(\mu) \leq \text{polylog}(d)$ .

It seems natural to ask whether we can find a Lipschitz map  $\varphi$  with  $\varphi_* \gamma_d = \mu$ ?

Let  $\mu$  be log-concave and isotropic,

$$\int_{\mathbb{R}^d} x d\mu(x) = 0 \quad \int_{\mathbb{R}^d} x \otimes x d\mu(x) = \text{Id}.$$

A famous conjecture of Kannan-Lovász-Simonovits postulates,

$$C_p(\mu) \leq C.$$

Current best bound, due to Klartag and Lehec (building upon Chen):  $C_p(\mu) \leq \text{polylog}(d)$ .

It seems natural to ask whether we can find a Lipschitz map  $\varphi$  with  $\varphi_* \gamma_d = \mu$ ?

- In general, one cannot find a Lipschitz transport map from  $\gamma_d$  to  $\mu$ .
- The existence of such map implies sub-Gaussian tails of  $\mu$ , which is not true for all isotropic log-concave measures.
- However, E. Milman showed that for KLS, it is enough to have map which is 'Lipschitz on average'.

### Question

If  $\mu$  is log concave and isotropic, does there exists a map  $\varphi$  with  $\varphi_*\gamma = \mu$ , such that

$$\mathbb{E}_\gamma [\|D\varphi\|^2] \leq \text{polylog}(d)?$$

- In general, one cannot find a Lipschitz transport map from  $\gamma_d$  to  $\mu$ .
- The existence of such map implies sub-Gaussian tails of  $\mu$ , which is not true for all isotropic log-concave measures.
- However, E. Milman showed that for KLS, it is enough to have map which is 'Lipschitz on average'.

### Question

If  $\mu$  is log concave and isotropic, does there exists a map  $\varphi$  with  $\varphi_*\gamma = \mu$ , such that

$$\mathbb{E}_\gamma [\|D\varphi\|^2] \leq \text{polylog}(d)?$$



- In general, one cannot find a Lipschitz transport map from  $\gamma_d$  to  $\mu$ .
- The existence of such map implies sub-Gaussian tails of  $\mu$ , which is not true for all isotropic log-concave measures.
- However, E. Milman showed that for KLS, it is enough to have map which is 'Lipschitz on average'.

### Question

If  $\mu$  is log concave and isotropic, does there exists a map  $\varphi$  with  $\varphi_*\gamma = \mu$ , such that

$$\mathbb{E}_\gamma [\|D\varphi\|^2] \leq \text{polylog}(d)?$$

- In general, one cannot find a Lipschitz transport map from  $\gamma_d$  to  $\mu$ .
- The existence of such map implies sub-Gaussian tails of  $\mu$ , which is not true for all isotropic log-concave measures.
- However, E. Milman showed that for KLS, it is enough to have map which is 'Lipschitz on average'.

### Question

If  $\mu$  is log concave and isotropic, does there exists a map  $\varphi$  with  $\varphi_*\gamma = \mu$ , such that

$$\mathbb{E}_\gamma [\|D\varphi\|^2] \leq \text{polylog}(d)?$$

- In general, one cannot find a Lipschitz transport map from  $\gamma_d$  to  $\mu$ .
- The existence of such map implies sub-Gaussian tails of  $\mu$ , which is not true for all isotropic log-concave measures.
- However, E. Milman showed that for KLS, it is enough to have map which is 'Lipschitz on average'.

## Question

If  $\mu$  is log concave and isotropic, does there exists a map  $\varphi$  with  $\varphi_*\gamma = \mu$ , such that

$$\mathbb{E}_\gamma [\|D\varphi\|^2] \leq \text{polylog}(d)?$$

By slightly altering our perspective, we give a positive answer to the previous questions.

Let  $\Omega := C([0, 1], \mathbb{R}^d)$  stand for the Wiener space with the Wiener measure  $\gamma$ . We will let  $(B_t)_{t \in [0, 1]}$  denote a Brownian motion.

We consider Lipschitz mappings  $\Phi : \Omega \rightarrow \mathbb{R}^d$  with  $D\Phi$  bounded almost surely.

Derivatives are taken in the Malliavin sense.

By slightly altering our perspective, we give a positive answer to the previous questions.

Let  $\Omega := C([0, 1], \mathbb{R}^d)$  stand for the Wiener space with the Wiener measure  $\gamma$ . We will let  $(B_t)_{t \in [0, 1]}$  denote a Brownian motion.

We consider Lipschitz mappings  $\Phi : \Omega \rightarrow \mathbb{R}^d$  with  $D\Phi$  bounded almost surely.

Derivatives are taken in the Malliavin sense.

## Theorem (M.-Shenfeld)

Let  $\mu$  be a measure on  $\mathbb{R}^d$ . There exists map  $\Phi : \Omega \rightarrow \mathbb{R}^d$ , with  $\Phi_*\gamma = \mu$  and

1. If  $\mu$  is log-concave with  $\text{diam}(\text{supp}(\mu)) \leq R$ ,

$$\|D\Phi\| \leq R.$$

2. If  $\mu = \gamma_d \star \nu$  and  $\text{diam}(\text{supp}(\nu)) \leq R$ ,

$$\|D\Phi\| \leq e^{R^2}.$$

3. If  $\mu$  is log-concave and isotropic,

$$\mathbb{E}_\gamma [\|D\Phi\|^2] \leq d^{o(1)} \text{polylog}(d).$$

Recall the Cameron-Martin space

$$H := \{h \in \Omega \mid h_t = \int_0^t \dot{h}_s ds\}.$$

It is also characterized by the fact that  $B_t + g$  is absolutely continuous with respect to  $\gamma$ , iff  $g \in H$ .

Heuristically, for a random variable  $F$  we define the Malliavin derivative  $DF$ , as the Gateaux derivative in the  $H$  directions.

Recall the Cameron-Martin space

$$H := \{h \in \Omega \mid h_t = \int_0^t \dot{h}_s ds\}.$$

It is also characterized by the fact that  $B_t + g$  is absolutely continuous with respect to  $\gamma$ , iff  $g \in H$ .

Heuristically, for a random variable  $F$  we define the Malliavin derivative  $DF$ , as the Gateaux derivative in the  $H$  directions.



# Malliavin calculus 101

$H$  has a natural inner product,  $\langle h, h' \rangle_H := \int_0^1 \dot{h}_t \dot{h}'_t dt$ . Observe that  $DF : \Omega \rightarrow H$  and we denote by  $DF_t$ , by  $D_t F$ .

We say that a map  $F$  is  $R$ -Lipschitz (in the  $H$  directions), if  $\|DF\|_H \leq R$  almost surely. This definition is justified, since

$$\text{Var}_\gamma(F) \leq \mathbb{E}_\gamma [\|DF\|_H^2].$$

$H$  has a natural inner product,  $\langle h, h' \rangle_H := \int_0^1 \dot{h}_t \dot{h}'_t dt$ . Observe that  $DF : \Omega \rightarrow H$  and we denote by  $DF_t$ , by  $D_t F$ .

We say that a map  $F$  is  $R$ -Lipschitz (in the  $H$  directions), if  $\|DF\|_H \leq R$  almost surely. This definition is justified, since

$$\text{Var}_\gamma(F) \leq \mathbb{E}_\gamma [\|DF\|_H^2].$$

# First attempt

We can mimic Caffarelli's Euclidean optimal transport result. Two main issues to address:

- Need to define a Wasserstein metric on  $\Omega$ .
- Need to embed  $\mu$  in  $\Omega$ .

First, define a metric, which is compatible with  $H$ :

$$d_H(\omega, \omega') = \begin{cases} \|\omega - \omega'\|_H & \text{if } \omega - \omega' \in H \\ \infty & \text{otherwise} \end{cases} .$$

# First attempt

We can mimic Caffarelli's Euclidean optimal transport result. Two main issues to address:

- Need to define a Wasserstein metric on  $\Omega$ .
- Need to embed  $\mu$  in  $\Omega$ .

First, define a metric, which is compatible with  $H$ :

$$d_H(\omega, \omega') = \begin{cases} \|\omega - \omega'\|_H & \text{if } \omega - \omega' \in H \\ \infty & \text{otherwise} \end{cases} .$$

# First attempt

Define a measure  $\tilde{\mu}$  on  $\Omega$  by

$$\frac{d\tilde{\mu}}{d\gamma}(\omega) = \frac{d\mu}{d\gamma_d}(\omega_1),$$

and consider,

$$\min_{\Psi_*\gamma = \tilde{\mu}} \mathbb{E} \left[ d_H(\Psi(B.), B.)^2 \right].$$

Equivalently,

$$\min_{u_t} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right],$$

where  $B_1 + \int_0^1 u_t dt \sim \mu$ .

# First attempt

Define a measure  $\tilde{\mu}$  on  $\Omega$  by

$$\frac{d\tilde{\mu}}{d\gamma}(\omega) = \frac{d\mu}{d\gamma_d}(\omega_1),$$

and consider,

$$\min_{\Psi_*\gamma=\tilde{\mu}} \mathbb{E} \left[ d_H(\Psi(B.), B.)^2 \right].$$

Equivalently,

$$\min_{u_t} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right],$$

where  $B_1 + \int_0^1 u_t dt \sim \mu$ .

# First attempt

Define a measure  $\tilde{\mu}$  on  $\Omega$  by

$$\frac{d\tilde{\mu}}{d\gamma}(\omega) = \frac{d\mu}{d\gamma_d}(\omega_1),$$

and consider,

$$\min_{\Psi_*\gamma=\tilde{\mu}} \mathbb{E} \left[ d_H(\Psi(B.), B.)^2 \right].$$

Equivalently,

$$\min_{u_t} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right],$$

where  $B_1 + \int_0^1 u_t dt \sim \mu$ .

# First attempt

Define  $v_t^{\text{opt}} := \arg \min_{u_t} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right]$ .

Then,  $v_t^{\text{opt}}(\omega) = \psi^{\text{opt}}(\omega_1) - \omega_1$ , and  $\Phi^{\text{opt}}(\omega) = \omega + \int v_t dt$  satisfies,

- $\Phi_*^{\text{opt}} \gamma = \tilde{\mu}$ .
- $(\Phi_1^{\text{opt}})_* \gamma = \mu$ .

This is unsatisfactory.



# First attempt

Define  $v_t^{\text{opt}} := \arg \min_{u_t} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right]$ .

Then,  $v_t^{\text{opt}}(\omega) = \psi^{\text{opt}}(\omega_1) - \omega_1$ , and  $\Phi^{\text{opt}}(\omega) = \omega + \int v_t dt$  satisfies,

- $\Phi_*^{\text{opt}} \gamma = \tilde{\mu}$ .
- $(\Phi_1^{\text{opt}})_* \gamma = \mu$ .

This is unsatisfactory.

# First attempt

Define  $v_t^{\text{opt}} := \arg \min_{u_t} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right]$ .

Then,  $v_t^{\text{opt}}(\omega) = \psi^{\text{opt}}(\omega_1) - \omega_1$ , and  $\Phi^{\text{opt}}(\omega) = \omega + \int v_t dt$  satisfies,

- $\Phi_*^{\text{opt}} \gamma = \tilde{\mu}$ .
- $(\Phi_1^{\text{opt}})_* \gamma = \mu$ .

This is unsatisfactory.

## Second attempt

We consider an optimization problem adapted to the filtration of  $B_t$ .

Define  $v_t := \arg \min_{u_t \text{ adapted}} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right]$  and  $dX_t = dB_t + v_t dt$ .

Facts:

- $X_1 \sim \mu$  (this is the transport map).
- $\text{Ent}(\mu|\gamma) = \frac{1}{2} \int_0^1 \mathbb{E}[\|v_t\|^2] dt$ .
- $v_t$  is a martingale, with  $v_t(X_t) = \nabla \ln \left( P_{1-t} \left( \frac{d\mu}{d\gamma_d}(X_t) \right) \right)$ .

## Second attempt

We consider an optimization problem adapted to the filtration of  $B_t$ .

Define  $v_t := \arg \min_{u_t \text{ adapted}} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right]$  and  $dX_t = dB_t + v_t dt$ .

Facts:

- $X_1 \sim \mu$  (this is the transport map).
- $\text{Ent}(\mu|\gamma) = \frac{1}{2} \int_0^1 \mathbb{E}[\|v_t\|^2] dt$ .
- $v_t$  is a martingale, with  $v_t(X_t) = \nabla \ln \left( P_{1-t} \left( \frac{d\mu}{d\gamma_d}(X_t) \right) \right)$ .

## Second attempt

We consider an optimization problem adapted to the filtration of  $B_t$ .

Define  $v_t := \arg \min_{u_t \text{ adapted}} \mathbb{E} \left[ \int_0^1 \|u_t\|^2 dt \right]$  and  $dX_t = dB_t + v_t dt$ .

Facts:

- $X_1 \sim \mu$  (this is the transport map).
- $\text{Ent}(\mu || \gamma) = \frac{1}{2} \int_0^1 \mathbb{E}[\|v_t\|^2] dt$ .
- $v_t$  is a martingale, with  $v_t(X_t) = \nabla \ln \left( P_{1-t} \left( \frac{d\mu}{d\gamma_d}(X_t) \right) \right)$ .

# The Föllmer Drift - (Some) History

- Analogous problems were already considered by in the 30's, by Schrödinger.
- The process itself was first studied by Föllmer, in 85', who used it to derive a variational expression for entropy.
- It appeared implicitly in the works of Feyel and Üstünel, from 2004, in their study of infinite dimensional transportation problems.
- In the context of functional inequalities, the use of the Föllmer process was pioneered by Lehec in 2012.
- Lassalle identified the process as the solution to a causal transportation problem in 2013.

# The Brownian transport map

Recall that  $X_1 = B_1 + \int_0^1 \nabla \ln \left( P_{1-t} \frac{d\mu}{d\gamma_d}(X_t) \right) dt$ . It can be shown that

$$DX_t = I_d + \int_0^t \nabla^2 \ln \left( P_{1-s} \frac{d\mu}{d\gamma_d}(X_s) \right) DX_s ds.$$

We write  $\nabla v_t := \nabla^2 \ln \left( P_{1-t} \frac{d\mu}{d\gamma_d}(X_t) \right)$  and for  $h \in H$ , we calculate,

$$f_h(t) := \langle DX_t, h \rangle_H = \int_0^t \dot{h}_s ds + \int_0^t \nabla v_t \langle DX_s, h \rangle_H ds.$$

In particular,

$$\frac{d}{dt} f_h(t) = \dot{h}_t - \nabla v_t f_h(t).$$

# The Brownian transport map

Recall that  $X_1 = B_1 + \int_0^1 \nabla \ln \left( P_{1-t} \frac{d\mu}{d\gamma_d}(X_t) \right) dt$ . It can be shown that

$$DX_t = I_d + \int_0^t \nabla^2 \ln \left( P_{1-s} \frac{d\mu}{d\gamma_d}(X_s) \right) DX_s ds.$$

We write  $\nabla v_t := \nabla^2 \ln \left( P_{1-t} \frac{d\mu}{d\gamma_d}(X_t) \right)$  and for  $h \in H$ , we calculate,

$$f_h(t) := \langle DX_t, h \rangle_H = \int_0^t \dot{h}_s ds + \int_0^t \nabla v_t \langle DX_s, h \rangle_H ds.$$

In particular,

$$\frac{d}{dt} f_h(t) = \dot{h}_t - \nabla v_t f_h(t).$$



# The Brownian transport map

Recall that  $X_1 = B_1 + \int_0^1 \nabla \ln \left( P_{1-t} \frac{d\mu}{d\gamma_d}(X_t) \right) dt$ . It can be shown that

$$DX_t = I_d + \int_0^t \nabla^2 \ln \left( P_{1-s} \frac{d\mu}{d\gamma_d}(X_s) \right) DX_s ds.$$

We write  $\nabla v_t := \nabla^2 \ln \left( P_{1-t} \frac{d\mu}{d\gamma_d}(X_t) \right)$  and for  $h \in H$ , we calculate,

$$f_h(t) := \langle DX_t, h \rangle_H = \int_0^t \dot{h}_s ds + \int_0^t \nabla v_t \langle DX_s, h \rangle_H ds.$$

In particular,

$$\frac{d}{dt} f_h(t) = \dot{h}_t - \nabla v_t f_h(t).$$

# The Brownian transport map

Solving this differential equation, we get, for every  $h \in H$ ,

$$f_h(1) = \int_0^1 e^{\int_0^t \nabla v_s ds} \cdot \dot{h}(t) dt.$$

So,

$$D_t X_1 = e^{\int_0^t \nabla v_s ds},$$

and

$$\|DX_1\|_H^2 = \int_0^1 e^{2 \int_0^t \nabla v_s ds} dt.$$

# The Brownian transport map

Direct calculations show,

$$\nabla v_t := \nabla^2 \ln \left( P_{1-t} \frac{d\mu}{d\gamma}(X_t) \right) = \frac{\text{Cov}(\mu_t)}{(1-t)^2} - \frac{1}{1-t} \mathbb{I}_d,$$

where

$$\frac{d\mu_t}{dx} \propto \frac{d\mu}{d\gamma_d}(x) e^{\frac{-(x-X_t)^2}{2(1-t)}}.$$

If  $\text{diam}(\text{supp}(\mu)) \leq R$ , clearly,

$$\nabla v_t \leq \frac{R^2}{(1-t)^2} - \frac{1}{1-t}.$$

Moreover, by Brascamp-Lieb, if  $\mu$  is log-concave,

$$\nabla v_t \leq \frac{1}{t}.$$

# The Brownian transport map

Direct calculations show,

$$\nabla v_t := \nabla^2 \ln \left( P_{1-t} \frac{d\mu}{d\gamma}(X_t) \right) = \frac{\text{Cov}(\mu_t)}{(1-t)^2} - \frac{1}{1-t} \mathbb{I}_d,$$

where

$$\frac{d\mu_t}{dx} \propto \frac{d\mu}{d\gamma_d}(x) e^{-\frac{(x-X_t)^2}{2(1-t)}}.$$

If  $\text{diam}(\text{supp}(\mu)) \leq R$ , clearly,

$$\nabla v_t \leq \frac{R^2}{(1-t)^2} - \frac{1}{1-t}.$$

Moreover, by Brascamp-Lieb, if  $\mu$  is log-concave,

$$\nabla v_t \leq \frac{1}{t}.$$

# The Brownian transport map

Direct calculations show,

$$\nabla v_t := \nabla^2 \ln \left( P_{1-t} \frac{d\mu}{d\gamma}(X_t) \right) = \frac{\text{Cov}(\mu_t)}{(1-t)^2} - \frac{1}{1-t} \mathbb{I}_d,$$

where

$$\frac{d\mu_t}{dx} \propto \frac{d\mu}{d\gamma_d}(x) e^{\frac{-(x-X_t)^2}{2(1-t)}}.$$

If  $\text{diam}(\text{supp}(\mu)) \leq R$ , clearly,

$$\nabla v_t \leq \frac{R^2}{(1-t)^2} - \frac{1}{1-t}.$$

Moreover, by Brascamp-Lieb, if  $\mu$  is log-concave,

$$\nabla v_t \leq \frac{1}{t}.$$

We apply the two inequalities to  $\|DX_1\|_H^2 = \int_0^1 e^{2\int_t^1 \nabla v_s ds} dt$

## Theorem (M.- Shenfeld)

Consider  $X_1$  as a map from  $\Omega = C([0, 1], \mathbb{R}^d)$  to  $\mathbb{R}^d$ .

1. If  $\mu$  is log-concave with  $\text{diam}(\text{supp}(\mu)) \leq R$ ,

$$\|DX_1\| \leq R.$$

2. If  $\mu = \gamma_d \star \nu$  and  $\text{diam}(\text{supp}(\nu)) \leq R$ ,

$$\|DX_1\| \leq e^{R^2}.$$

- The second result follows by showing  $\nabla v_t \leq R^2$ .
- Can be extended to semi-log concave measures.

We apply the two inequalities to  $\|DX_1\|_H^2 = \int_0^1 e^{2\int_t^1 \nabla v_s ds} dt$

## Theorem (M.- Shenfeld)

Consider  $X_1$  as a map from  $\Omega = C([0, 1], \mathbb{R}^d)$  to  $\mathbb{R}^d$ .

1. If  $\mu$  is log-concave with  $\text{diam}(\text{supp}(\mu)) \leq R$ ,

$$\|DX_1\| \leq R.$$

2. If  $\mu = \gamma_d \star \nu$  and  $\text{diam}(\text{supp}(\nu)) \leq R$ ,

$$\|DX_1\| \leq e^{R^2}.$$

- The second result follows by showing  $\nabla v_t \leq R^2$ .
- Can be extended to semi-log concave measures.

# Further thoughts

We have demonstrated a Lipschitz map  $X_1 : \Omega \rightarrow \mathbb{R}^d$ .

It seems natural to ask whether  $X : \Omega \rightarrow \Omega$  is Lipschitz as well?

It turns out that there exist strongly log-concave measures, for which  $X$  is not Lipschitz, for any constant. This is contrast to the optimal transport map  $\Psi^{\text{opt}}$ , which is provably 1-Lipschitz.



# Further thoughts

We have demonstrated a Lipschitz map  $X_1 : \Omega \rightarrow \mathbb{R}^d$ .

It seems natural to ask whether  $X : \Omega \rightarrow \Omega$  is Lipschitz as well?

It turns out that there exist strongly log-concave measures, for which  $X$  is not Lipschitz, for any constant. This is contrast to the optimal transport map  $\Psi^{\text{opt}}$ , which is provably 1-Lipschitz.

## Further thoughts

We have demonstrated a Lipschitz map  $X_1 : \Omega \rightarrow \mathbb{R}^d$ .

It seems natural to ask whether  $X : \Omega \rightarrow \Omega$  is Lipschitz as well?

It turns out that there exist strongly log-concave measures, for which  $X$  is not Lipschitz, for any constant. This is contrast to the optimal transport map  $\Psi^{\text{opt}}$ , which is provably 1-Lipschitz.

# Future directions

- Can the results be extended to larger classes of measures?
- What about similar, but different, constructions on the Wiener space?
- Can similar results be proved for maps between finite dimensional spaces?
- In particular, can the results be recovered for the Brenier map?

*Thank You*

# The KLS connection

Instead of applying point-wise bounds, we could estimate

$$\mathbb{E} [\|DX_1\|_H^2] = \mathbb{E} \left[ \int_0^1 e^{2 \int_t^1 \nabla v_s(X_s) ds} \right].$$

For isotropic  $\mu$ , define  $\tau = \frac{1}{2} \wedge \inf\{t | \nabla v_t(X_t) \geq 2\}$ .

$$\int_0^1 \nabla v_t(X_t) \leq 2 + \int_{\tau}^1 \frac{1}{t} dt = 2 + \log(\tau).$$

So,

$$\mathbb{E} [\|DX_1\|_H^2] \leq e^4 \mathbb{E} \left[ \frac{1}{\tau^2} \right].$$

# The KLS connection

Instead of applying point-wise bounds, we could estimate

$$\mathbb{E} [\|DX_1\|_H^2] = \mathbb{E} \left[ \int_0^1 e^{2 \int_t^1 \nabla v_s(X_s) ds} \right].$$

For isotropic  $\mu$ , define  $\tau = \frac{1}{2} \wedge \inf\{t | \nabla v_t(X_t) \geq 2\}$ .

$$\int_0^1 \nabla v_t(X_t) \leq 2 + \int_{\tau}^1 \frac{1}{t} dt = 2 + \log(\tau).$$

So,

$$\mathbb{E} [\|DX_1\|_H^2] \leq e^4 \mathbb{E} \left[ \frac{1}{\tau^2} \right].$$

With the recent result of Yuansi Chen about the KLS constant, we prove:

## Theorem

*Let  $\mu$  be an isotropic log-concave vector in  $\mathbb{R}^d$ . Then,*

$$\mathbb{E} [\|DX_1\|_H^2] = d^{o(1)}.$$