

# ① THE OPTIMAL TRANSPORT PROBLEM

- $X, Y$  complete and separable metric spaces
- $c: X \times Y \rightarrow [0, +\infty]$  Borel
- $\mu \in \mathcal{P}(X), \nu \in \mathcal{P}(Y)$

$$(M) \quad \inf \left\{ \int_X c(x, T(x)) d\mu(x) \mid T_{\#} \mu = \nu \right\}$$

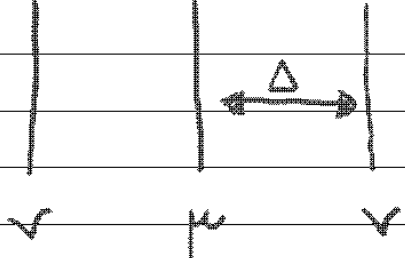
NOTATION:  $T_{\#}: \mathcal{P}(X) \rightarrow \mathcal{P}(Y)$ ,  $T_{\#} \mu(E) = \mu(T^{-1}(E))$

$$(S \circ T)_{\#} = S_{\#} \circ T_{\#}, \quad \int_Y \varphi dT_{\#} \mu = \int_X \varphi \circ T d\mu$$

Monge's problem can be ill-posed because:

- 1) No admissible  $T$  exists
- 2) The infimum is not attained
- 3) The constraint is not weakly req. closed

- $\mu = \delta_x$        $r = (\delta_y + \delta_z)/2, \quad y \neq z$

-   $\inf(M) = \Delta$ , not attained  
 $c(x, y) = |y - x|^p, \quad p > 0$

$T$  Lipschitz, 1-1.

$$T_{\#}(\rho_1 \mathcal{L}^m) = \rho_2 \mathcal{L}^m$$



$$\rho_2(T(x)) \text{ a.e. } |\nabla T(x)| = \rho_1(x) \text{ a.e.}$$

Kantorovich's weak formulation (1941):

$$(K) \quad \inf \left\{ \int_{X \times Y} c(x, y) d\gamma(x, y) \mid \gamma \in \Gamma(\mu, \nu) \right\}$$

$$\Gamma(\mu, \nu) = \left\{ \gamma \in \mathcal{P}(X \times Y) \mid (\pi_X)_\# \gamma = \mu, (\pi_Y)_\# \gamma = \nu \right\}$$

Equivalently:  $\gamma \in \Gamma(\mu, \nu) \iff \mu(A) = \gamma(A \times Y), \nu(B) = \gamma(X \times B)$

$\gamma(A \times B) =$  "The mass in  $A$  sent to  $B$ "

$\mu \times \nu \in \Gamma(\mu, \nu)$  !!! In addition,  $\Gamma(\mu, \nu)$  is seq.  
weakly closed in  $\mathcal{P}(X \times Y)$  (w.r.t. duality with  $C_b(X \times Y)$ )

**THEOREM 1. (EXISTENCE OF OPTIMAL PLANS)** If  $c$  is lower semicontinuous, then  $(K)$  has a solution. Moreover, if  $c < \infty$ , then

$$\inf(M) = \min(K)$$

Idea of the proof:  $\gamma \rightarrow \int_{X \times Y} c d\gamma$  is weakly l.s.c.

In addition, by ULAM theorem,  $\exists C_m, K_m$  compact such that

$$\mu(X \setminus \bigcup_n C_m) = 0 \quad \text{and} \quad \nu(Y \setminus \bigcup_n K_m) = 0.$$

It turns out that

$$\gamma(X \times Y \setminus C_m \times K_m) \leq \mu(X \setminus C_m) + \nu(Y \setminus K_m) \quad \forall \gamma \in \Gamma(\mu, \nu),$$

so that  $\Gamma(\mu, \nu)$  is a tight family in  $\mathcal{O}(X \times Y)$  and

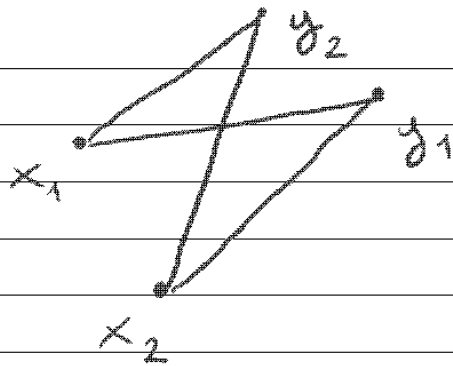
PROKHOROV theorem gives that  $\Gamma(\mu, \nu)$  is weakly relatively compact.

Can we recover an optimal  $T$  from an optimal  $\gamma$  ??

Def (C-MONOTONICITY)  $\Gamma \subset X \times Y$  is said to be c-monotone

if

$$\sum_{i=1}^N c(x_i, y_{\sigma(i)}) \geq \sum_{i=1}^N c(x_i, y_i) \quad \forall (x_i, y_i) \in \Gamma, \quad \sigma \in \bar{\Sigma}_N$$



**THEOREM 2.**  $c$  l.s.c.,  $\gamma$  optimal,  $\int c d\gamma < \infty$ .

Then  $\gamma$  is concentrated on a  $c$ -monotone and  $\sigma$ -compact set  $\Gamma \subset X \times Y$ . The converse holds if

$$(*) \quad c(x, y) \leq a(x) + b(y), \quad a \in L^1(\mu), \quad b \in L^1(\nu).$$

REMARK Condition (\*) is satisfied in many natural examples, but not all: for instance the OT problem in the WIENER space does not fit into this framework.

Theorem 2 leads in a natural way to the analysis of the properties of  $c$ -monotone sets, trying to understand how far are they from being graphs.



QUADRATIC COST FUNCTION:

$$X = Y = H, \quad c(x, y) = \frac{1}{2} \|x - y\|^2$$

$$\sum_{i=1}^N \|x_i - y_i\|^2 \leq \sum_{i=1}^N \|x_{\sigma(i)} - y_i\|^2 \quad \text{implies}$$

$$\sum_{i=1}^N \langle y_i, x_{\sigma(i)} - x_i \rangle \leq 0$$

SUBDIFFERENTIAL of a convex l.s.c. function  $F: H \rightarrow \mathbb{R} \cup \{+\infty\}$

$$\partial F(x) := \{ v \in H \mid F(y) \geq F(x) + \langle y - x, v \rangle \quad \forall y \in H \}.$$

Here  $x \in D(F) := \{ F < \infty \}$ .

**THEOREM 3 (ROCKAFELLAR)**  $\Gamma \subset H \times H$  is contained in the graph of  $\partial F$ , for some convex lsc  $F$  if, and only if,  $\Gamma$  is monotone

Proof of the "only if" implication: add the inequalities

$$\langle y_i, x_{\sigma(i)} - x_i \rangle \leq F(x_{\sigma(i)}) - F(x_i) \quad , \quad \text{if } y_i \in \partial F(x_i)$$

to obtain 
$$\sum_i \langle y_i, x_{\sigma(i)} - x_i \rangle \leq 0.$$

Combining theorem 2 and theorem 3, we know that the support of any optimal  $\gamma$  is contained in  $\{(x, y) \mid y \in \partial F(x)\}$  for some convex l.s.c.  $F: H \rightarrow \mathbb{R} \cup \{+\infty\}$ .

In  $H = \mathbb{R}^n$ , we know also that

$$\Sigma := \{x \in \mathbb{R}^n \mid \partial F(x) \text{ contains more than 1 point}\}$$

is Lebesgue negligible, and even  $(n-1)$ -dimensional.

**THEOREM 4.**  $X = Y = \mathbb{R}^m$ ,  $\mu(B) = 0 \quad \forall (m-1)$ -dimensional set  $B$ .

Then any optimal plan  $\gamma$  is induced by a transport  $T$ , that is the gradient of a convex function  $F$ . In addition  $T$  is unique.

Proof If  $T_1, T_2$  are optimal and  $\mu(\{T_1 \neq T_2\}) > 0$ , then

$$\gamma = \frac{1}{2}(\gamma_1 + \gamma_2), \quad \text{with} \quad \gamma_i = (\text{Id} \times T_i)_\# \mu, \quad i = 0, 1$$

is still optimal, and not induced by a transport. ■

EXTENSIONS : • MANIFOLDS (McCANN, FATHI-FIGALLI)

• CARNOT GROUPS (A-RIGOT, AGRACHEV-LEE,  
RIFFORD-FIGALLI)

• WIENER SPACES (FEYEL - USTUNEL)

$E$  Hilbert space,  $\gamma$  Gaussian measure in  $E$

$H = CM(\gamma) = \{ h \in E \mid (\tau_h)_\# \gamma \ll \gamma \}$ , a dense  
subspace of  $E$ , with a stronger Hilbert norm  $\|\cdot\|_H$

$$c(x, y) = \begin{cases} \frac{1}{2} \|x - y\|_H^2 & \text{if } x - y \in H \\ +\infty & \text{otherwise} \end{cases} \quad (\gamma(H) = 0 !!)$$

## DISTANCE, INTERPOLATION AND GEODESICS

$(X, d)$  metric space

$$\mathcal{P}_2(X) = \left\{ \mu \in \mathcal{P}(X) \mid \int d^2(x_0, x) d\mu(x) < \infty \text{ for some } x_0 \right\}$$

$$W_2^2(\mu, \nu) = \min(K), \quad \text{with } c(x, y) := \frac{1}{2} d^2(x, y)$$

$W_2$  is a distance: a "brutal" proof using transport maps

$$\text{is: } 2 W_2^2(\mu, \nu) = \int d^2(T(x), x) d\mu(x) \quad T_{\#} \mu = \nu$$

$$2 W_2^2(\nu, \sigma) = \int d^2(S(y), y) d\nu(y) \quad S_{\#} \nu = \sigma$$

$$\begin{aligned}
2W_2(\sigma, \mu) &\leq \|d(S \circ T, Id)\|_{L^2(\mu)} \\
&\leq \|d(S \circ T, S)\|_{L^2(\mu)} + \|d(S, Id)\|_{L^2(\mu)} \\
&= 2W_2(\mu, \nu) + 2W_2(\nu, \sigma)
\end{aligned}$$

In general, either use " $\inf(M) = \min(K)$ ", or introduce the concept of "composition of two plans"

$(\mathcal{P}_2(X), W_2)$  inherits many properties of  $X$ :

- complete if  $X$  is complete;
- compact if  $X$  is compact;

- a length space if  $X$  is a length space
- a PC (positively curved) space if  $X$  is a PC space.

**THEOREM 5.** (CONVERGENCE IN  $\mathcal{P}_2(X)$ )  $(X, d)$  complete.

Then  $\mu_n \rightarrow \mu$  in  $\mathcal{P}_2(X)$  iff

$$\mu_n \xrightarrow{C_b} \mu \quad \text{and} \quad \lim_{n \rightarrow \infty} \int d^2(x, x_0) d\mu_n(x) = \int d^2(x, x_0) d\mu(x).$$

let us prove that  $\mu_n \xrightarrow{W_2} \mu \Rightarrow \mu_n \xrightarrow{C_b} \mu$



By a monotone approximation argument, it suffices to consider a test function  $\varphi \in \text{lip}_b(X)$ . Then, with  $\gamma \in \Gamma_0(\mu, \mu_n)$ , we have:

$$\begin{aligned} \int \varphi d\mu_n &= \iint \varphi(y) d\gamma_n(x, y) \\ &= \iint \varphi(x) d\gamma_n(x, y) + o(1) = \int \varphi d\mu + o(1) \end{aligned}$$

because

$$\iint d(x, y) d\gamma_n \rightarrow 0.$$

## THEOREM 6 (GEODESICS IN $\mathcal{D}_2(H)$ )

If  $\gamma \in \Gamma_0(\mu, \nu)$ , then

$$\mu_t := ((1-t)x + ty)_\# \gamma \quad t \in [0, 1]$$

is a constant speed geodesic from  $\mu$  to  $\nu$ . Conversely, any constant speed geodesic has this representation for some (unique)  $\gamma$ .

Proof We have to show that  $W_2(\mu_s, \mu_t) = (t-s) W_2(\mu, \nu)$

By the triangle inequality, it suffices to show that:

$$W_2(\mu_s, \mu_t) \leq (t-s) W_2(\mu, \nu), \quad 0 \leq s \leq t \leq 1.$$

Indeed,  $\gamma_{s,t} := \left( \underbrace{(1-s)x + sy}_z, \underbrace{(1-t)x + ty}_w \right)_{\#} \gamma \in \Gamma(\mu_s, \mu_t)$

and

$$\int |z - w|^2 d\gamma_{s,t}(z, w) = (t-s)^2 \int |x - y|^2 d\gamma = (t-s)^2 W_2^2(\mu, \nu).$$

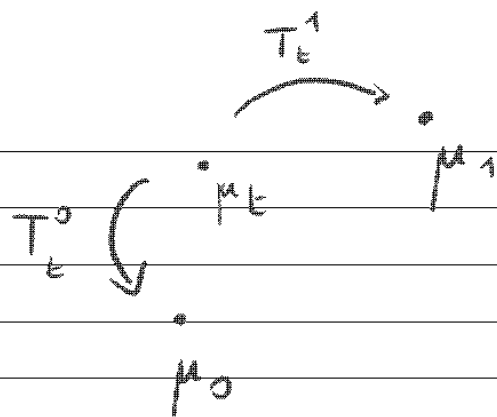


REMARK. The proof of the converse implication gives much more

information:  $\Gamma_0(\mu_t, \mu_0)$  and  $\Gamma_0(\mu_t, \mu_1)$ ,  $t \in (0, 1)$ , are

singularities, and induced by maps  $T_t^0, T_t^1$  with  $\text{lip}(T_t^0) \leq 1/(1-t)$

$$\text{lip}(T_t^1) \leq 1/t$$



In the case  $\mu_0 \ll \mathcal{L}^m$ ,  $H = \mathbb{R}^m$ , we have

$$\mu_t := ((1-t)\text{Id} + tT)_\# \mu_0 \quad (\text{McCANN})$$

$$T_t^0 = ((1-t)\text{Id} + tT)^{-1}$$

The estimate on the Lipschitz constant of  $T_t^0$  follows by the monotonicity of  $(1-t)\text{Id} + tT$ , larger than  $(1-t)\text{Id}$ :

$$\langle (1-t)x + tT(x) - (1-t)x' + tT(x'), x - x' \rangle \geq (1-t)|x - x'|^2$$

**THEOREM 7.** (BEHAVIOUR OF  $\mathcal{D}_2^{\alpha}(\mathbb{R}^m)$  under interpolation)

Let  $\mu, \nu \in \mathcal{D}_2(\mathbb{R}^m)$ ,  $\mu \ll \mathcal{L}^m$ . Then  $\mu_t \ll \mathcal{L}^m \quad \forall t \in [0, 1)$  and its density  $\rho_t$  is given by:

$$\rho_t = \frac{\rho}{\det \nabla T_t} \circ \left( T_t \Big|_{\Sigma^c} \right)^{-1}$$

Here:  $\rho$  is the density of  $\mu$ ;

$$T_t = (1-t) \text{Id} + tT;$$

$\Sigma$  is the singular (non differentiability) set of  $T$

As a consequence, we obtain the behaviour of internal energy functionals under geodesic interpolation:

$$\int U(p_t) dy = \int \det \nabla T_t \ U\left(\frac{p}{\det \nabla T_t}\right) dx, \quad t \in [0, 1)$$

"  $y = T_t(x)$  "

The proof that  $\mu_\varepsilon \ll L^n$  is simple: indeed,

$$\mathcal{L}^m(A) = 0 \Rightarrow \mathcal{L}^m(T_\ell^{-1}(A)) = 0 \Rightarrow \mu(T_\ell^{-1}(A)) = 0 \Rightarrow \mu_\ell(A) = 0.$$

$(T_\ell^{-1} \in \text{lip})$ 
 $(\mu \ll \mathcal{L}^m)$

# OPTIMAL TRANSPORTATION FROM THE DIFF. VIEWPOINT

HEURISTICS. Continuity equation

$$\frac{d}{dt} \mu_t + \nabla \cdot (v_t \mu_t) = 0$$

Formally  $\int \mu + \nabla \cdot (v \mu) = 0$ , so that

$$T_\mu \mathcal{D}_2(\mathbb{R}^m) \sim \{ -\nabla \cdot (v \mu) \mid v \in L^2(\mu, \mathbb{R}^m) \}$$

$$\text{Metric: } \langle -\nabla \cdot (v \mu), -\nabla \cdot (w \mu) \rangle_\mu := \int_{\mathbb{R}^m} \langle v, w \rangle d\mu$$

The induced "Riemannian" distance is:

$$\tilde{W}_2^2(\mu, \nu) := \inf \left\{ \int_0^1 \int |v_t|^2 d\mu_t : \mu_0 = \mu, \mu_1 = \nu, \frac{d}{dt} \mu_t + \nabla \cdot (v_t \mu_t) = 0 \right\}$$

THEOREM 8. (BENAMOU-BRENIER)  $W_2 = \tilde{W}_2$ .

$$\tilde{W}_2 \leq W_2$$

$$\text{let } \mu_t := (T_t)_\# \mu$$

LEMMA  $f \in L^2(\sigma, \mathbb{R}^m) \Rightarrow T_\#(f\sigma) \ll T_\#\sigma$  and its density

$$h \text{ satisfies } \|h\|_{L^2(T_\#\sigma)} \leq \|f\|_{L^2(\sigma)}$$



Instead,

$$\langle g, T_{\#}(f\sigma) \rangle = \langle g \circ T, f\sigma \rangle \leq \|f\|_{L^2(\sigma)} \|g \circ T\|_{L^2(\sigma)}$$

$$g \in C_c(\mathbb{R}^m; \mathbb{R}^m) \quad = \|f\|_{L^2(\sigma)} \|g\|_{L^2(T_{\#}\sigma)}$$

By Riesz theorem, this proves the lemma. ■

Now, a velocity field compatible with  $\mu_t$  is given by

$$V_t \mu_t = (T_t)_{\#} ((T - \text{Id}) \mu)$$

By the lemma  $v_t$  is well defined and

$$\|v_t\|_{L^2(\mu_t)} \leq \|T - \text{Id}\|_{L^2(\mu)} = W_2(\mu, \nu)$$

It follows that

$$\widetilde{W}_2^2(\mu, \nu) \leq \int_0^1 \|v_t\|_{L^2(\mu_t)}^2 dt \leq W_2^2(\mu, \nu).$$

Compatibility of  $v_t$ :

$$\begin{aligned} \frac{d}{dt} \int \varphi d\mu_t &= \frac{d}{dt} \int \varphi(T_t) d\mu = \int \langle \nabla \varphi(T_t), T - \text{Id} \rangle d\mu \\ &= \langle \nabla \varphi, (T_t)_\# ((T - \text{Id}) \mu) \rangle = \int \langle \nabla \varphi, v_t \rangle d\mu_t. \end{aligned}$$

$$\tilde{W}_2 \geq W_2$$

Let us assume that  $v_t$  is sufficiently regular to have a flow map  $X: [0,1] \times \mathbb{R}^n \rightarrow \mathbb{R}^n$  satisfying

$$1) \quad \frac{d}{dt} X(t, x) = v_t(X(t, x)), \quad X(0, x) = x$$

$$2) \quad X(t, \cdot) \# \mu_0 = \mu_t.$$

$$\text{By 1)} \quad |X(1, x) - x|^2 \leq \int_0^1 |v_t(X(t, x))|^2 dt$$

An integration w.r.t.  $x$  gives

$$W_2^2(\mu_0, \mu_1) \leq \int |X(1, x) - x|^2 d\mu(x) \leq \int_0^1 \int |v_t(X(t, x))|^2 d\mu(x) dt$$

$$= \int_0^1 \int |v_t|^2 d\mu_t dt.$$

In the general case a regularization argument is needed.

let  $\rho_\varepsilon(x) = (2\pi\varepsilon^2)^{-n/2} \exp(-|x|^2/2\varepsilon^2)$  be the heat kernel and set

$$\mu_t^\varepsilon := \mu_t * \rho_\varepsilon$$

$$V_t^\varepsilon \mu_t^\varepsilon := (V_t \mu_t) * \rho_\varepsilon$$

This regularization preserves the continuity equation. The key fact is that the convexity of the map

$$(z, t) \longrightarrow |z|^2/t \quad z \in \mathbb{R}^n, \quad t > 0$$

and JENSEN's inequality yield

$$\int |v_t^\varepsilon|^2 d\mu_t^\varepsilon \leq \int |v_t|^2 d\mu_t \quad \forall t \in [0, 1], \quad \varepsilon > 0.$$

Thus, together with the local Lipschitz condition

$$\|v_t^\varepsilon\|_{W^{1,\infty}(B_R; \mathbb{R}^n)} \in L^\infty(0, 1) \quad \forall \varepsilon, R > 0$$

can be used to show that the flow  $x^\varepsilon(t, x)$  is globally defined in  $[0, 1]$  for  $\mu$ -a.e.  $x$ , so that

$$W_2(\mu_0^\varepsilon, \mu_1^\varepsilon) \leq \int_0^1 \int |v_t^\varepsilon|^2 d\mu_t^\varepsilon dt \leq \int_0^1 \int |v_t|^2 d\mu_t.$$

letting  $\varepsilon \downarrow 0$ , the weak lower semicontinuity of

$$(\mu, \nu) \rightarrow W_2(\mu, \nu)$$

concludes the proof. ■

Now, we are going to make Otto's picture rigorous by relating a metric concept, the notion of ABSOLUTELY CONTINUOUS curve  $\mu_t$  with values in  $\mathcal{P}_2(H)$ , to a differential one.

## DEFINITION (ABSOLUTE CONTINUITY AND METRIC DERIVATIVE)

$x: [0,1] \rightarrow E$  is said to be absolutely continuous if

$$d(x(s), x(t)) \leq \int_s^t g(\tau) d\tau \quad \forall 0 \leq s \leq t \leq 1$$

for some  $g \in L^1(0,1)$ . The minimal  $g$  with this property is

$$|x'(t)| := \lim_{h \rightarrow 0} \frac{1}{|h|} d(x(t+h), x(t)).$$

**THEOREM 3.** Let  $\mu_t: [0, 1] \rightarrow \mathcal{P}_2(H)$  be absolutely continuous.

Then  $\exists v_t \in L^2(\mu_t; H)$  such that:

- 1) the continuity equation  $\frac{d}{dt} \mu_t + \nabla \cdot (v_t \mu_t) = 0$  holds;
- 2)  $\|v_t\|_{L^2(\mu_t)} \leq |\mu'_t|$  for a.e.  $t$ .

Conversely, if  $(v_t, \mu_t)$  satisfy (1) and  $\|v_t\|_{L^2(\mu_t)} \in L^1(0, 1)$ , then  $\mu_t$  is absolutely continuous, as a  $\mathcal{P}_2(H)$ -valued map, and

- 2')  $\|v_t\|_{L^2(\mu_t)} \geq |\mu'_t|$  for a.e.  $t$ .

By 2'), the velocity field  $v_t$  is unique and we call it tangent.



The "constructive" part of Theorem 3 is based on a duality argument, while the "converse" part relies on a smoothing scheme very much similar to the one used in the proof of the BB formula.

The duality argument provides also an alternative characterization of  $v_t$ :

$$1) \quad \frac{d}{dt} \mu_t + \nabla \cdot (v_t \mu_t) = 0, \quad \text{with } \|v_t\|_{L^2(\mu_t)} \in L^1(0,1)$$

$$2) \quad v_t \in \overline{\{ \nabla \varphi \mid \varphi \text{ smooth cylindrical} \}}^{L^2(\mu_t; H)}$$

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DEFINITION.  $T_\mu \mathcal{P}_2(H)$

In the case  $H = \mathbb{R}^n$  one can take  $\varphi \in C_c^\infty(\mathbb{R}^n)$

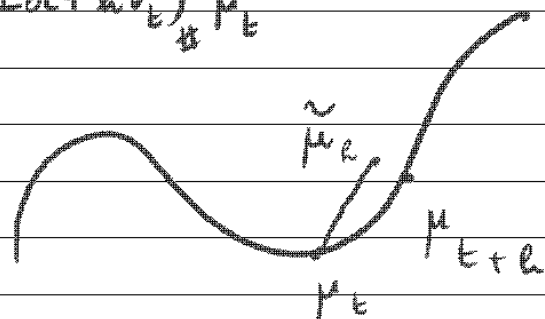
# THEOREM 10 (TANGENT FIELD AS LIMIT OF RESCALED OPTIMAL PLANS)

For a.e.  $t$ ,  $v_t$  satisfies:  $\forall \gamma_h \in \Gamma_0(\mu_t, \mu_{t+h})$ , we have

$$\tilde{\gamma}_h := \left(x, \frac{y-x}{h}\right)_{\#} \gamma_h \xrightarrow[\mathcal{G}_2(H \times H)]{h \rightarrow 0} (\text{Id} \times v_t)_{\#} \mu_t$$

Heuristically,  $(\text{Id} + h v_t)$  is "almost" optimal as  $h \rightarrow 0$ .

$$\tilde{\mu}_h = (\text{Id} + h v_t)_{\#} \mu_t \quad \frac{1}{h} \left( T_{\mu_t}^{\mu_{t+h}} - \text{Id} \right) \xrightarrow[L^2(\mu_t)]{} v_t \text{ as } h \rightarrow 0.$$



$$W_2(\mu_{t+h}, \tilde{\mu}_h) = o(h)$$