



# Evolution of probability measures and Optimal Transport

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## Particle models

Evolution problems for discrete measures

Optimal transport for arbitrary measures

The metric definition of a gradient flow

Dissipative evolutions

Let us consider the evolution of a system  $\mathbf{X}(t) = (X_\omega(t))_{\omega \in \Omega} \in H^\Omega$  of  $N$  indistinguishable particles in the Euclidean space  $H = \mathbb{R}^d$ , parametrized by a finite set  $\Omega := \{\omega_1, \omega_2, \dots, \omega_N\}$ , governed by a first order system of differential equations

$$\dot{X}_\omega(t) = V_\omega(t) = F(\omega, \mathbf{X}(t)), \quad \omega \in \Omega$$

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where  $F : \Omega \times H^\Omega \rightarrow H^\Omega$  is invariant with respect to permutations  $\sigma \in \text{Sym}(\Omega)$  of the coordinates:

$$F(\sigma(\omega), \mathbf{X} \circ \sigma) = F(\omega, \mathbf{X}) \quad \text{for every permutation } \sigma \in \text{Sym}(\Omega), \sigma : \Omega \rightarrow \Omega.$$

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We may look for a robust description, suitable to pass to the limit as  $N \uparrow \infty$ .

- Finite dimensional Cauchy-Lipschitz theory <sup>1</sup>

$$F(\omega, \mathbf{X}) = A(X_\omega) + \frac{1}{N} \sum_{\eta \in \Omega} B(X_\omega - X_\eta), \quad A, B : H \rightarrow H \text{ (one-sided) Lipschitz.}$$

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- **Gradient flows:** <sup>2</sup> given a permutation invariant function

$$F(\omega, \mathbf{X}) := -\partial_{X_\omega} \Phi(\mathbf{X}), \quad \Phi : H^\Omega \rightarrow \mathbb{R} \text{ invariant by permutations}$$

$\Phi$  typically satisfies a suitable convexity condition but can be nonsmooth (subdifferential calculus): e.g. <sup>3</sup>

$$\Phi(\mathbf{X}) = R\left(\frac{1}{N} \sum_{\omega} T(X_\omega)\right) + \frac{1}{N^2} \sum_{\eta, \omega} W(X_\omega - X_\eta) + \frac{1}{N} \sum_{\omega} V(X_\omega)$$

$T : H \rightarrow \tilde{H}$  is a vector valued map,  $R : \tilde{H} \rightarrow \mathbb{R}$ ,  $W, V : H \rightarrow \mathbb{R}$ .

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- **Dissipative evolution**, contraction semigroups:  $H$  Hilbert or Banach space,  $F$  multivalued. E.g. the Lipschitz perturbation of a multivalued subgradient.

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## Measure description

$\Omega$  can be endowed with the normalized counting measure  $\mathbb{P} = \frac{1}{N} \sum_{\omega \in \Omega} \delta_{\omega}$  and the evolution of the system can be described by the discrete probability measures  $\mu_t \in \mathcal{P}^N(\mathbb{H})$

$$\mu_t := \frac{1}{N} \sum_{\omega \in \Omega} \delta_{X_{\omega}(t)} = \mathbf{X}(t)_{\#} \mathbb{P}, \quad \int \varphi d\mu_t = \frac{1}{N} \sum_{\omega \in \Omega} \varphi(X_{\omega}(t)) = \mathbb{E}[\varphi \circ \mathbf{X}(t)]$$

Recall that a map  $Z : \Omega \rightarrow \mathbb{H}$  acts on discrete measures as

$$Z_{\#} \delta_{\omega} = \delta_{Z(\omega)}, \quad Z_{\#} \left( \frac{1}{N} \sum_{\omega} \delta_{\omega} \right) = \frac{1}{N} \sum_{\omega} \delta_{Z(\omega)}.$$

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The evolution law involves the velocity vector measure

$$\mathbf{v}_t = \frac{1}{N} \sum_{\omega \in \Omega} \dot{X}_{\omega}(t) \delta_{X_{\omega}(t)} = \mathbf{v}_t \mu_t,$$

satisfying the **continuity equation**

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Notice that  $\dot{X}_{\omega}(t) = \dot{X}_{\eta}(t)$  for a.e.  $t$  such that  $X_{\omega}(t) = X_{\eta}(t)$ , so that  $\underline{v}_t$  is well defined a.e. on the support of  $\mu_t$ .

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**A PDE description:** suppose that there exists a smooth map  $\underline{F} : X \times \mathcal{P}^N(H) \rightarrow H$  such that

$$\underline{F}\left(y, \frac{1}{N} \sum_{\omega} \delta_{X_{\omega}}\right) = F(\omega, \mathbf{X}) \quad \text{whenever } X_{\omega} = y.$$

$$\begin{cases} \partial_t \mu_t + \nabla \cdot (\underline{v}_t \mu_t) = 0 & \text{in the distributional sense of } (0, \infty) \times X, \\ \underline{v}_t(x) = \underline{F}(x, \mu_t). \end{cases}$$

# Weak convergence

## Definition (Weak convergence)

A sequence of probability measures  $(\mu^N)_{N \in \mathbb{N}} \subset \mathcal{P}(H)$  converges to  $\mu \in \mathcal{P}(H)$  if

$$\lim_{N \rightarrow \infty} \int \varphi \, d\mu^N = \int \varphi \, d\mu \quad \text{for every } \varphi \in C_b(H) \text{ (or, equivalently, } C_c^\infty(H))$$



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If the map  $\underline{F}$  can be extended to  $X \times \mathcal{P}(H)$ , so that

$$(\mu^N)_{N \in \mathbb{N}} \subset \mathcal{P}^N(H), \quad \mu^N \xrightarrow{N \uparrow \infty} \mu \quad \Rightarrow \quad \underline{F}(\cdot, \mu^N) \rightarrow \underline{F}(\cdot, \mu) \quad \text{uniformly}$$

then one can hope to pass to the limit in the discrete PDE to obtain

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Questions:

1. How to provide a **quantitative description of weak convergence**.
2. How to provide a **quantitative description of continuity of velocity vector fields** and study the stability of the system with respect to the initial datum.
3. How to deal with **"discontinuous"** velocity vector fields.

## Evolution of discrete measures

$$\mu = \frac{1}{N} \sum_{\omega} \delta_{x_{\omega}}, \nu = \frac{1}{N} \sum_{\omega} \delta_{y_{\omega}}.$$

$$W_2^2(\mu, \nu) := \min_{\sigma \in \text{Sym}(\Omega)} \frac{1}{N} \sum_{\omega} |x_{\omega} - y_{\sigma(\omega)}|^2$$

$W_2$  is a distance between discrete measures.

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Permutations  $\leftrightarrow$  couplings:  $\gamma_{\sigma} = \frac{1}{N} \sum_{\omega} \delta_{(x_{\omega}, y_{\sigma(\omega)})} \in \Gamma^N(\mu, \nu) \subset \mathcal{P}(H \times H)$ ,

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Optimal couplings:

$$\Gamma_o(\mu, \nu) := \left\{ \gamma \in \Gamma^N(\mu, \nu) : W_2^2(\mu, \nu) = \int |x - y|^2 d\gamma \right\}$$

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## Intrinsic characterization of curves

Let  $t \mapsto \mu_t$ ,  $t \in [0, T]$ , be discrete measures (with  $N$  particles).

### Theorem

*The following properties are equivalent:*

1.  $(\mu_t)_{t \in [0, T]}$  satisfy the Lipschitz condition:

$$W_2(\mu_s, \mu_t) \leq L|t - s| \quad \text{for every } s, t \in [0, T].$$

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2. *There exists a family of vector measures  $\nu_t = \underline{\nu}_t \mu_t$ ,  $t \in (0, T)$ , such that*  
$$\int |\underline{\nu}_t|^2 d\mu_t \leq L^2 \text{ a.e. in } (0, T) \text{ and}$$

$$\partial_t \mu_t + \nabla \cdot \nu_t = 0 \quad \text{in the distributional sense of } (0, T) \times X.$$



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$$\partial_t \mu_t + \nabla \cdot \nu_t = 0 \quad \text{in the distributional sense of } (0, T) \times X.$$

3. There exist Lipschitz maps  $X_\omega : [0, T] \rightarrow X$  such that

$$\mu_t = \frac{1}{N} \sum_{\omega \in \Omega} \delta_{X_\omega(t)} \quad \text{and}$$

$$\frac{1}{N} \sum_{\omega} |\dot{X}_\omega(t)|^2 = |\dot{\mu}_t|_W^2 := \lim_{h \rightarrow 0} \left( \frac{W_2(\mu_{t+h}, \mu_t)}{|h|} \right)^2 \leq L^2 \quad \text{a.e. in } (0, T).$$

## The derivative of the Wasserstein distance

Let  $\mu_t = \frac{1}{N} \sum_{\omega} \delta_{X_{\omega}(t)}$  be a Lipschitz map of discrete measures and let  $\nu = \frac{1}{N} \sum_{\omega} \delta_{Y_{\omega}}$  be another discrete measure.

For every  $t \in [0, T]$  let us select an **optimal coupling**  $\gamma_t \in \Gamma_o(\mu_t, \nu)$  associated to an **optimal permutation**  $\sigma_t$ . For a.e.  $t \in (0, T)$  we have

$$\frac{d}{dt} \frac{1}{2} W_2^2(\mu_t, \nu) = \frac{1}{N} \sum_{\omega} \dot{X}_{\omega}(t) (X_{\omega}(t) - Y_{\sigma_t(\omega)}) = \int \underline{v}_t(x) \cdot (x - y) d\gamma_t(x, y).$$

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Similarly, if  $\nu_t = \frac{1}{N} \sum_{\omega} \delta_{Y_{\omega}(t)}$  is another Lipschitz curve with velocity vector  $\underline{w}_t$  we have

$$\begin{aligned} \frac{d}{dt} \frac{1}{2} W_2^2(\mu_t, \nu_t) &= \frac{1}{N} \sum_{\omega} (\dot{X}_{\omega}(t) - \dot{Y}_{\sigma_t(\omega)}(t)) (X_{\omega}(t) - Y_{\sigma_t(\omega)}(t)) \\ &= \boxed{\int (\underline{v}_t(x) - \underline{w}_t(y)) \cdot (x - y) d\gamma_t(x, y)}. \end{aligned}$$

## Dissipative fields

Suppose that  $(\mu_t)_{t \in [0, T]}$ ,  $(\nu_t)_{t \in [0, T]}$  are solutions of the evolution system

$$\partial_t \mu_t + \nabla \cdot (\underline{v}_t \mu_t) = 0, \quad \underline{v}_t(x) = \underline{F}(x, \mu_t)$$

$$\partial_t \nu_t + \nabla \cdot (\underline{w}_t \nu_t) = 0, \quad \underline{w}_t(x) = \underline{F}(x, \nu_t)$$

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This is equivalent to say that for every  $\mu, \nu$  every optimal permutation  $\sigma \in \text{Sym}(\Omega)$  and corresponding  $\gamma \in \Gamma_o(\mu, \nu)$

$$\begin{aligned} & \frac{1}{N} \sum_{\omega} (\underline{F}(X_{\omega}, \mu) - \underline{F}(Y_{\sigma(\omega)}, \nu))(X_{\omega} - Y_{\sigma(\omega)}) \\ &= \boxed{\int (\underline{F}(x, \mu) - \underline{F}(y, \nu)) \cdot (x - y) d\gamma(x, y) \leq 0} \end{aligned}$$

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More generally,  $\lambda$ -dissipativity

$$\int (\underline{F}(x, \mu) - \underline{F}(y, \nu)) \cdot (x - y) d\gamma(x, y) \leq \lambda \int |x - y|^2 d\gamma = \lambda W_2^2(\mu, \nu)$$

corresponds to the exponential behaviour

$$W_2^2(\mu_t, \nu_t) \leq e^{2\lambda t} W_2^2(\mu_0, \nu_0) \quad \text{for every } t \geq 0.$$

## Dissipative fields and geodesically convex functionals

**Displacement interpolation:** given  $\vartheta \in [0, 1]$ ,  $\sigma \in \Gamma_o(\mu, \nu)$

$$\mu_\vartheta := \frac{1}{N} \sum_{\omega} \delta_{X_\omega(\vartheta)}, \quad X_\omega(\vartheta) := (1 - \vartheta)X_\omega + \vartheta y_{\sigma(\omega)}.$$

$\vartheta \mapsto \mu_\vartheta$  is a **constant speed, minimal geodesic** connecting  $\mu$  to  $\nu$ :

$$\boxed{W_2(\mu_{\vartheta_1}, \mu_{\vartheta_2}) = |\vartheta_1 - \vartheta_2| W_2(\mu, \nu) \quad \text{for every } \vartheta_1, \vartheta_2 \in [0, 1].}$$

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A functional  $\mathcal{F}$  defined on discrete measures,

$$\mathcal{F}(\mu) = \Phi(\mathbf{X}) \quad \text{if } \mu = \frac{1}{N} \sum_{\omega} \delta_{X_{\omega}}$$

is **displacement convex** (or, equivalently, geodesically convex) if

$$\boxed{\mathcal{F}(\mu_{\vartheta}) \leq (1 - \vartheta)\mathcal{F}(\mu) + \vartheta\mathcal{F}(\nu).}$$

In this case the field  $\underline{F}$  satisfying  $\underline{F}(X_{\omega}, \frac{1}{N} \sum_{\omega} \delta_{X_{\omega}}) = -\partial_{X_{\omega}} \Phi(\frac{1}{N} \sum_{\omega} \delta_{X_{\omega}})$  is dissipative.



## Explicit Euler method

We can construct solutions to the evolution equation by means of the **Explicit Euler method**:

we fix a **step size**  $\tau > 0$ , an initial measure  $\bar{\mu} = \frac{1}{N} \sum_{\omega} \delta_{\bar{X}_{\omega}}$ , and we consider the measures  $\mu^{\tau,n} = \frac{1}{N} \sum_{\omega} \delta_{X_{\omega}^{\tau,n}}$  where the sequence of maps  $(X^{\tau,n})_{n \in \mathbb{N}}$  is defined by induction:

$$X_{\omega}^{\tau,0} = \bar{X}_{\omega}, \quad X_{\omega}^{\tau,n+1} = X_{\omega}^{\tau,n} + \tau F(X_{\omega}^{\tau,n}, X^{\tau,n})$$

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$$\mu^{\tau,n+1} = (E^{\tau,n})_{\#} \mu^{\tau,n}, \quad E^{\tau,n}(x) := x + \tau \underline{F}(x, \mu^{n,\tau})$$

If  $\mu_t^{\tau} := \mu^{\tau, \lfloor t/\tau \rfloor}$  is the **piecewise constant interpolation**,  $\mu_t^{\tau} = \mu^{\tau,n}$  if  $n\tau \leq t < (n+1)\tau$ , we get  $\lim_{\tau \downarrow 0} \mu_t^{\tau} = \mu_t$  solving

$$\partial_t \mu_t + \nabla \cdot (\underline{v}_t \mu_t) = 0, \quad \underline{v}_t(x) = \underline{F}(x, \mu_t), \quad \mu_0 = \bar{\mu}.$$

If  $\underline{F}$  arises as the gradient of a displacement convex functional  $\mathcal{F}$ , we can also use a **variational formulation of the implicit Euler method**.

In this case at each step one has to solve the equation

$$\left( E^{-\tau, n+1} \right)_{\#} \mu^{\tau, n+1} = \mu^{\tau, n}, \quad E^{-\tau, n+1}(x) := x - \tau \underline{F}(x, \mu^{n+1, \tau})$$

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<sup>4</sup>R. JORDAN, D. KINDERLEHRER, F. OTTO, The variational formulation of the Fokker-Planck equation. SIAM J. Math. Anal. (1998)  
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According to the **JKO -Minimizing Movement approach**

<sup>4</sup>, it is sufficient to select  $\mu^{\tau, n+1}$  among the minimizers of

$$\mu \mapsto \frac{1}{2\tau} W_2^2(\mu, \mu^{\tau, n}) + \mathcal{F}(\mu)$$

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The implicit Euler method is well adapted to deal with important example arising in diffusion equations <sup>5</sup> as the **Fokker-Planck equation**

$$\partial_t \mu_t + \nabla \cdot (\underline{v}_t \mu_t) = 0, \quad \underline{v}_t = -\nabla(\log u_t + V), \quad \mu_t = u_t \mathcal{L}^d \ll \mathcal{L}^d$$

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which can be interpreted as the gradient flow of the **Relative Entropy functional**

$$\mathcal{F}(\mu) := \int u(\log u + V) dx = \text{Ent}(\mu|m) \quad \text{if } \mu = u \mathcal{L}^d \ll \mathcal{L}^d, \quad m = e^{-V} \mathcal{L}^d.$$

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## Arbitrary probability measures

We can extend the notion of (Kantorovich-Rubinstein-)Wasserstein distance to arbitrary probability measures with finite quadratic moment:

$$\mu \in \mathcal{P}_2(\mathbb{H}), \quad \int |x|^2 d\mu(x) < +\infty$$

The role of couplings associated to permutations is now replaced by **arbitrary probability measures**  $\gamma \in \mathcal{P}(\mathbb{H} \times \mathbb{H})$  **with assigned marginals**  $\mu$  and  $\nu$ :

$$\Gamma(\mu, \nu) := \left\{ \gamma \in \mathcal{P}(\mathbb{H} \times \mathbb{H}) : \pi_{\#}^1(\gamma) = \mu, \pi_{\#}^2(\gamma) = \nu \right\}$$

where  $\pi^i(x_1, x_2) := x_i$ ,  $i = 1, 2$ .

$$W_2^2(\mu, \nu) := \min \left\{ \int |x_1 - x_2|^2 d\gamma(x_1, x_2) : \gamma \in \Gamma(\mu, \nu) \right\}$$

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$(\mathcal{P}_2(H), W_2)$  is a **complete and separable metric space** and the collection of discrete measures is a **dense** subset of  $\mathcal{P}_2(H)$  (it can be considered as the natural metric completion of set of discrete measures).

Couplings allow for splitting of mass: if  $\mu = \frac{1}{N} \sum_{\omega} \delta_{X_{\omega}}$ ,  $\nu = \frac{1}{N} \sum_{\omega} \delta_{Y_{\omega}}$  then an element of  $\Gamma(\mu, \nu)$  can be expressed as

$$\gamma = \sum_{\omega, \eta \in \Omega} m_{\omega, \eta} \delta_{(X_{\omega}, Y_{\eta})}, \quad \sum_{\eta} m_{\omega, \eta} = \sum_{\omega} m_{\omega, \eta} = 1/N.$$

and it is a remarkable consequence of Birkhoff theorem that for discrete measures

$$W_2^2(\mu, \nu) = \min_{\sigma \in \text{Sym}(\Omega)} \frac{1}{N} \sum_{\omega} |X_{\omega} - Y_{\sigma(\omega)}|^2$$

## Lipschitz curves in $\mathcal{P}_2(H)$

Let now  $\Omega := AC^2([0, T]; H)$ , the space of **absolutely continuous curves**  $\omega : [0, T] \rightarrow H$  such that  $\int_0^T |\dot{\omega}(t)|^2 dt < \infty$  endowed with the topology of uniform convergence. For every  $t \in [0, T]$  we consider the evaluation maps  $X(t) : \Omega \rightarrow H$ ,  $X(t, \omega) := \omega(t)$ .

### Theorem

*The following properties are equivalent:*

1.  $(\mu_t)_{t \in [0, T]}$  is **Lipschitz**:  $W_2(\mu_s, \mu_t) \leq L|t - s|$  for every  $s, t \in [0, T]$ .

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2. There exists a **Borel vector field**  $\underline{v} : (0, T) \times H \rightarrow H$  such that

$$\int |\underline{v}_t|^2 d\mu_t \leq L^2 \text{ a.e. in } (0, T) \text{ and } \partial_t \mu_t + \nabla \cdot (\underline{v}_t \mu_t) = 0 \text{ in } \mathcal{D}'((0, T) \times H).$$

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3. There exist a **Radon probability measure**  $\mathbb{P}$  on  $\Omega$  such that for every  $t \in [0, T]$   $\mu_t = X(t)_\# \mathbb{P}$  and

$$\int |\dot{X}(t, \omega)|^2 d\mathbb{P}(\omega) = |\dot{\mu}_t|_W^2 := \lim_{h \rightarrow 0} \left( \frac{W_2(\mu_{t+h}, \mu_t)}{|h|} \right)^2 \leq L^2 \text{ a.e. in } (0, T).$$

Moreover  $\dot{X}(t, \omega) = \underline{v}_t(X(t, \omega))$  for  $\mathbb{P}$ -a.e.  $\omega$ .

## Variational characterization of the Wasserstein velocity field

If  $(\mu_t)_{t \in [0, T]}$  is Lipschitz:  $W_2(\mu_s, \mu_t) \leq L|t - s|$  for every  $s, t \in [0, T]$  and  $\underline{v} : (0, T) \times H \rightarrow H$  is a Borel vector field such that

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then  $\underline{v}$  satisfies the **optimal condition**

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if and only if

$$\underline{v}_t \in \text{Tan}_{\mu_t}(\mathcal{P}_2(H)) = \overline{\left\{ \nabla \varphi : \varphi \in C_c^\infty(H) \right\}}^{L^2(\mu_t; H)} \text{ for a.e. } t \in (0, T).$$



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Moreover for almost every  $t \in (0, T)$  we have

$$W_2(\mu_{t+h}, (I + h\underline{v}_t)_\# \mu_t) = o(h) \text{ as } h \rightarrow 0.$$

“Splitting times” are negligible in  $(0, T)$ .

## The derivative of the Wasserstein distance

Let  $(\mu_t)_{t \in [0, T]}$  be a Lipschitz curve in  $\mathcal{P}_2(H)$  let  $\nu \in \mathcal{P}_2(H)$ . For every  $t \in [0, T]$  let us select an optimal coupling  $\gamma_t \in \Gamma_o(\mu_t, \nu)$ . For a.e.  $t \in (0, T)$  we have

$$\frac{d}{dt} \frac{1}{2} W_2^2(\mu_t, \nu) = \int \underline{v}_t(x) \cdot (x - y) d\gamma_t(x, y).$$

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Similarly, if  $(\nu_t)_{t \in [0, T]}$  is another Lipschitz curve with velocity vector field  $\underline{w}_t$  we have

$$\frac{d}{dt} \frac{1}{2} W_2^2(\mu_t, \nu_t) = \int (\underline{v}_t(x) - \underline{w}_t(y)) \cdot (x - y) d\gamma_t(x, y).$$

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## A metric definition of a gradient flow

If  $\mathbf{u}(t)$  is a solution of the gradient flow in  $\mathbb{R}^d$

$$\dot{\mathbf{u}}(t) = -\nabla\phi(\mathbf{u}(t)) \quad (\text{GF})$$

where  $\phi : \mathbb{R}^d \rightarrow \mathbb{R}$  is a smooth convex function, then for every test vector  $\mathbf{w} \in \mathbb{R}^d$  we have

$$\frac{d}{dt} \frac{1}{2} |\mathbf{u}(t) - \mathbf{w}|^2 = \langle \dot{\mathbf{u}}(t), \mathbf{u}(t) - \mathbf{w} \rangle = \langle -\nabla\phi(\mathbf{u}(t)), \mathbf{w} - \mathbf{u}(t) \rangle \leq \phi(\mathbf{w}) - \phi(\mathbf{u}(t))$$

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$$\langle \nabla\phi(\mathbf{u}), \mathbf{w} - \mathbf{u} \rangle \leq \phi(\mathbf{w}) - \phi(\mathbf{u}).$$

In fact,  $\mathbf{u}$  solves (GF) if and only if

$$\frac{d}{dt} \frac{1}{2} |\mathbf{u}(t) - \mathbf{w}|^2 \leq \phi(\mathbf{w}) - \phi(\mathbf{u}(t)) \quad \text{for every } \mathbf{w} \in \mathbb{R}^d \quad (\text{EVI})$$

## The Wasserstein case

Let  $\mathcal{F} : \mathcal{P}_2(\mathbb{H}) \rightarrow [0, +\infty]$  be a lower semicontinuous and displacement convex functional. We say that a locally Lipschitz curve  $(\mu_t)_{t>0}$  is an **EVI-solution of the gradient flow of  $\mathcal{F}$**  if for every  $\nu \in D(\mathcal{F}) \subset \mathcal{P}_2(\mathbb{H})$

$$\frac{d}{dt} W_2^2(\mu_t, \nu) \leq \mathcal{F}(\nu) - \mathcal{F}(\mu_t) \quad \text{a.e. in } (0, \infty). \quad (\text{EVI})$$

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### Theorem

For every initial datum  $\bar{\mu} \in D(\mathcal{F})$  there exists a unique EVI solution to (EVI) satisfying  $\lim_{t \downarrow 0} \mu_t = \bar{\mu}$ . Moreover,  $\mu_t$  is the uniform limit of the JKO-Minimizing Movement approximations, obtained by solving

$$\mu^{\tau, n+1} \in \operatorname{argmin}_{\mu} \left\{ \frac{1}{2\tau} W_2^2(\mu, \mu^{\tau, n}) + \mathcal{F}(\mu) \right\}, \quad \mu^{\tau, 0} := \bar{\mu}.$$



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The EVI formulation is equivalent to the continuity equation

$$\partial_t \mu_t + \nabla \cdot (\underline{v}_t \mu_t) = 0$$

where the Wasserstein velocity vector field  $\underline{v}$  satisfies for a.e.  $t > 0$

$$\int \langle \underline{v}_t(x), x - y \rangle d\gamma_t(x, y) \leq \mathcal{F}(\nu) - \mathcal{F}(\mu_t) \quad \text{for every } \gamma_t \in \Gamma_o(\mu_t, \nu).$$

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## Dissipative vector fields in $\mathcal{P}_2(H)$

In  $\mathcal{P}_2(H)$  a general dissipative vector field  $F$  can be represented by a map (possibly multivalued) from  $\mathcal{P}_2(H)$  to  $\mathcal{P}_2(H \times H)$ :

$$F(\mu) \in \mathcal{P}_2(H; \mu) := \left\{ \underline{F} \in \mathcal{P}_2(H \times H) : \pi_{\#}^1 \underline{F} = \mu \right\}.$$

By **disintegrating**  $\underline{F} = F(\mu)$  w.r.t.  $\mu$  we obtain a family of measures  $\underline{F}_x \in \mathcal{P}_2(H)$  which represent **probability laws on directions starting from  $x$** .

In the “regular case”  $\underline{F}_x$  is **concentrated on a single vector**  $\delta_{\underline{F}(x, \mu)}$  and therefore can be represented by a vector field  $\underline{F}(x, \mu)$  mapping  $H \times \mathcal{P}_2(H)$  into  $H$ .

## Dissipative vector fields in $\mathcal{P}_2(H)$

In  $\mathcal{P}_2(H)$  a general dissipative vector field  $F$  can be represented by a map (possibly multivalued) from  $\mathcal{P}_2(H)$  to  $\mathcal{P}_2(H \times H)$ :

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In the general case, we can allow for a **general probability measure**  $\underline{F}_x$  **depending on  $x$** . Denoting by  $(x, \nu)$  the variables in  $H \times H$  and by  $\exp^\tau(x, \nu) := x + \tau\nu$ ,  $F$  acts on  $\mu$  by the family of maps

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Notice that in the regular case we have one step of the explicit Euler method:

$$F^\tau(\mu) = E_{\#}^\tau \mu, \quad E^\tau(x) = x + \tau \underline{F}(x, \mu).$$

$$F^\tau(\mu) := \exp_{\sharp}^{\tau} \underline{E} = (x + \tau v)_{\sharp} \underline{E}, \quad \underline{E} = F(\mu).$$

$F$  is dissipative if

$$\boxed{\left. \frac{d}{d\tau} W_2(F^\tau(\mu), F^\tau(\nu)) \right|_{\tau=0+} \leq 0 \quad \text{for every } \mu, \nu \in \mathcal{P}_2(H).$$

In the regular case this condition reads as

$$\int \langle \underline{E}(x, \mu) - \underline{E}(y, \nu), x - y \rangle d\gamma(x, y) \leq 0 \quad \text{for some } \gamma \in \Gamma_o(\mu, \nu).$$

## The explicit Euler method

If  $F$  is a dissipative vector field and  $\tau > 0$  is a time step, we can construct the map  $F^\tau : \mathcal{P}_2(H) \rightarrow \mathcal{P}_2(H)$

$$F^\tau(\mu) := \exp_{\sharp}^{\tau} \underline{F} = (x + \tau v)_{\sharp} \underline{F}, \quad \underline{F} = F(\mu)$$

and therefore the sequence of explicit Euler approximations:

$$\mu^{\tau,0} := \bar{\mu} \text{ given, } \mu^{\tau,n+1} := F^\tau(\mu^{\tau,n}), \quad \mu_t^{\tau} := \mu^{\tau, \lfloor t/\tau \rfloor}$$

**Problems:** convergence of the method and characterization of the limit.

## A metric definition of dissipative evolutions

Let us now consider a solution  $u(t)$  of the differential equation in  $\mathbb{R}^d$

$$\dot{u}(t) = F(u(t)) \quad (\text{DE})$$

where  $F : \mathbb{R}^d \rightarrow \mathbb{R}^d$  is a dissipative vector field. For every vector  $w \in \mathbb{R}^d$  we have

$$\begin{aligned} \frac{d}{dt} \frac{1}{2} |u(t) - w|^2 &= \langle \dot{u}(t), u(t) - w \rangle = \langle \dot{u}(t) - Fw, u(t) - w \rangle + \langle Fw, u(t) - w \rangle \\ &\leq \langle Fw, u(t) - w \rangle. \end{aligned}$$

where we used the **dissipativity condition**

$$\langle Fu - Fw, u - w \rangle \leq 0 \quad \text{for every } u, w \in \mathbb{R}^d.$$

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<sup>6</sup>P. BÉNILAN, Solutions intégrales d'équations d'évolution dans un espace de Banach. C.R.A.S. (1972)



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Thus  $u$  solves (DE) if and only if <sup>6</sup>

$$\frac{d}{dt} \frac{1}{2} |u(t) - w|^2 \leq \langle Fw, u(t) - w \rangle \quad \text{for every } w \in \mathbb{R}^d \quad (\text{EVI})$$

How can we interpret  $\langle Fw, u(t) - w \rangle$  for probability vector fields?

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## A pseudo-scalar product

Notice that  $\langle Fw, u(t) - w \rangle$  can be expressed as

$$\langle Fw, w - u \rangle = \lim_{\tau \downarrow 0} \frac{1}{2\tau} \left( |w + \tau Fw - u|^2 - |w - u|^2 \right)$$

We can analogously define a **pseudo-scalar product**

$$\langle F(\nu), \mu \rangle_- := \lim_{\tau \downarrow 0} \frac{1}{2\tau} \left( W_2^2(F^\tau(\nu), \mu) - W_2^2(\nu, \mu) \right)$$

It is possible to prove that the limit always exists and can be expressed by a suitable coupling between  $F(\nu)$  and  $\gamma \in \Gamma_o(\nu, \mu)$ .

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Moreover,  $F$  is **dissipative if and only if**

$$\langle F(\mu), \nu \rangle_- + \langle F(\nu), \mu \rangle_- \leq 0.$$

# Convergence

Let  $F : \mathcal{P}_2(\mathbb{H}) \rightarrow \mathcal{P}_2(\mathbb{H} \times \mathbb{H})$  be a **dissipative probability vector field** satisfying the bound

$$\sup \left\{ \int |v|^2 dF(\mu) : \int |x|^2 d\mu \leq C \right\} < \infty \quad \text{for every } C > 0.$$

## Theorem (Cavagnari-Sodini-S.)

- For every  $\bar{\mu} \in \mathcal{P}_2(\mathbb{H})$  the discrete solutions  $\mu^\tau$  of **the explicit Euler method** **uniformly converge** to a unique limit  $\mu : [0, \infty) \rightarrow \mathcal{P}_2(\mathbb{H})$ .

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$$\frac{d}{dt} \frac{1}{2} W_2^2(\mu_t, \nu) \leq -\langle F(\nu), \mu \rangle_- \quad \text{for every } \nu \in \mathcal{P}_2(\mathbb{H}), \text{ a.e. in } (0, \infty).$$

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- Moreover, for every  $T > 0$  we have the optimal error estimate

$$W_2(\mu_t, \mu_t^\tau) \leq C\sqrt{\tau} \quad \text{for every } t \in [0, T].$$