



Dissipative evolution of Probability Measures

Giuseppe Savaré

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Department of Decision Sciences, Bocconi University, Milan, Italy

In collaboration with *G. Cavagnari, G. Sodini*

Probability vector fields and evolution

Dissipative operators and contraction semigroups in Hilbert spaces

Convergence of the Explicit Euler method and contraction semigroups

Problem: generation of contraction semigroups in
the space of Borel probability measures $\mathcal{P}_2(E)$
($E = \mathbb{R}^d$ Euclidean space or $E = \mathbf{H}$ Hilbert)
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$\Gamma(\mu, \nu) :=$ couplings between $\mu \in \mathcal{P}(E)$, $\nu \in \mathcal{P}(F)$, measures $\gamma \in \mathcal{P}(E \times F)$
whose marginals are μ and ν , e.g. $\gamma = (X, Y)_{\#}\mathbb{P}$, $X_{\#}\mathbb{P} = \mu$, $Y_{\#}\mathbb{P} = \nu$.

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$\Gamma_o(\mu, \nu)$: optimal couplings for the L^2 -Wasserstein distance. $\gamma_o \in \Gamma_o(\mu, \nu)$ iff

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

Probability vector fields in $\mathcal{P}_2(E)$

Tangent space: $\mathbf{TE} = \{(x, v) : x, v \in E\} \approx E \times E$, $\mathbf{x}(x, v) = x$, $\mathbf{v}(x, v) = v$.

In $\mathcal{P}_2(E)$ a **probability vector field** \mathfrak{F} can be represented by a map (possibly multivalued) from $D(\mathfrak{F}) \subset \mathcal{P}_2(E)$ to $\mathcal{P}_2(\mathbf{TE})$ such that

$$\text{for every } \underline{F} \in \mathfrak{F}(\mu) : \quad \mathbf{x}_{\#}\underline{F} = \mu.$$

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

In the “regular case” 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 $E \times \mathcal{P}_2(E)$ into E .

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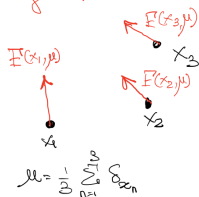
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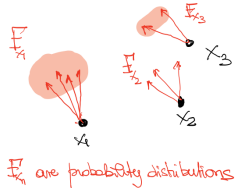
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In the general case, we can allow for a **general probability measure** F_x **depending on x** .

Regular case



General case



We want to study the evolution of probability measures driven by a PVF \mathfrak{F} , formally

$$\dot{\mu}_t = \mathfrak{F}(\mu_t) \quad t > 0.$$

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Example: finite dimensional Cauchy-Lipschitz theory ¹

\mathfrak{F} does not split particles and it is concentrated on the vector field $\underline{F}(x, \mu)$,

$$\mathfrak{F}(\mu) = (\text{Id} \times \underline{F}(\cdot, \mu))_{\#}\mu.$$

Examples are

$$\underline{F}(x, \mu) = A(x) + \int B(x - y) d\mu(y), \quad A, B : E \rightarrow E \text{ dissipative.}$$

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The curve $(\mu_t)_{t>0}$ solves the continuity equation

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

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- **Gradient flows** ² generated by a λ -geodesically convex functional $\mathcal{F} : \mathcal{P}_2(E) \rightarrow (-\infty, +\infty]$. \mathcal{F} can be nonsmooth (subdifferential calculus): e.g. ³

$$\mathcal{F}(\mu) = R\left(\int T(x) d\mu(x)\right) + \iint W(x-y) d\mu(x) d\mu(y) + \int V d\mu$$

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

$-\tilde{\mathcal{F}}$ is the multivalued **Wasserstein subdifferential of \mathcal{F}** .

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- **Dissipative evolution**, contraction semigroups: E Hilbert space, F multivalued. E.g. the Lipschitz perturbation of a multivalued subgradient. This case has been studied by Piccoli ⁴ in finite dimension with a different approach.

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Dissipative operators in Hilbert space

In a Hilbert space \mathbf{H} a (multivalued) map $B : D(B) \subset \mathbf{H} \rightrightarrows \mathbf{H}$ is **dissipative** if

$$\langle v - w, x - y \rangle \leq 0 \quad \text{for every } v \in Bx, w \in By$$

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This property has a **natural metric interpretation**: if we consider the curves

$$x(\tau) := x + \tau v, \quad y(\tau) := y + \tau w, \quad v \in Bx, w \in By$$

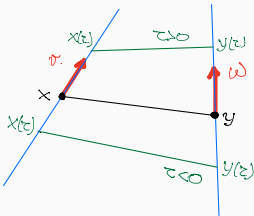
and their squared distance $D(\tau) := \frac{1}{2}|x(\tau) - y(\tau)|^2$

then

$$\langle v - w, x - y \rangle = D'(0) = \frac{1}{2} \frac{d}{d\tau} |x(\tau) - y(\tau)|^2 \Big|_{\tau=0} \leq 0$$

so that

$$|x(\tau) - y(\tau)|^2 \leq |x - y|^2 + \tau^2 |v - w|^2 = |x - y|^2 + o(\tau) \quad \text{as } \tau \downarrow 0$$



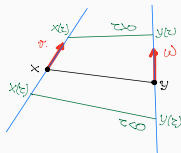
The resolvent

$D(\tau) = \frac{1}{2}|x(\tau) - y(\tau)|^2$ is **convex**, $D'(0) \leq 0$ yields

$$|x-y|^2 \leq |x(s)-y(s)|^2 \quad \text{for every } s < 0$$

If $x' - \tau Bx' = x$ and $y' - \tau By' = y$

then $|x' - y'|^2 \leq |x - y|^2$



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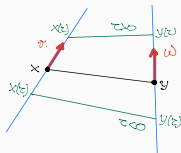
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$$J_\tau := (\text{Id} - \tau B)^{-1}, \quad x' = J_\tau(x) \Leftrightarrow x' - \tau Bx' = x \quad \text{is a contraction}$$

This property can be used to define dissipative operators in Banach spaces.



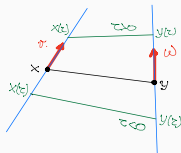
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B is **m-dissipative** (or maximal dissipative) if J_τ is defined in all the space \mathbf{H} : for every $x \in \mathbf{H}$ the equation

$$y - \tau B y \ni x \quad \text{has a (unique) solution } y = J_\tau x.$$

A particular case is the **subdifferential** $B = -\partial\Phi$ of a **convex l.s.c. function** $\Phi : \mathbf{H} \rightarrow (-\infty, +\infty]$: $x_\tau = J_\tau(x)$ if and only if

$$x_\tau \text{ minimizes } \quad y \mapsto \frac{1}{2\tau}|y - x|^2 + \Phi(y).$$

The Explicit and Implicit Euler methods

If B is everywhere defined, for every $x_\tau^0 \in \mathbb{H}$ one can solve the **Explicit Euler method**

$$\frac{x_\tau^n - x_\tau^{n-1}}{\tau} \in Bx_\tau^{n-1} \quad \Leftrightarrow \quad x_\tau^n = x_\tau^{n-1} + \tau Bx_\tau^{n-1} = (\text{Id} + \tau B)^n x_\tau^0, \quad n = 1, \dots$$

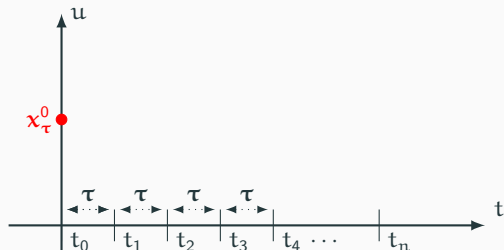
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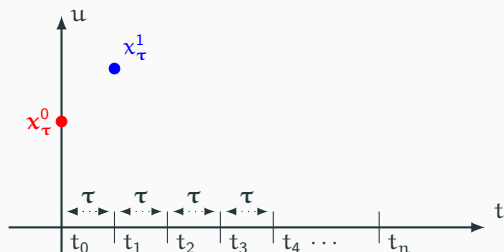
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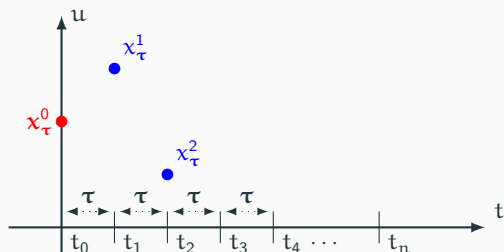
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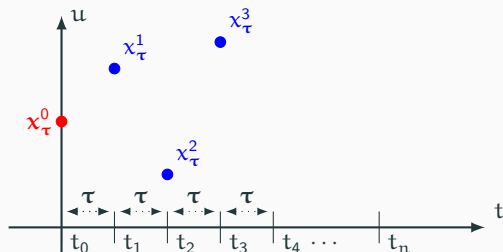
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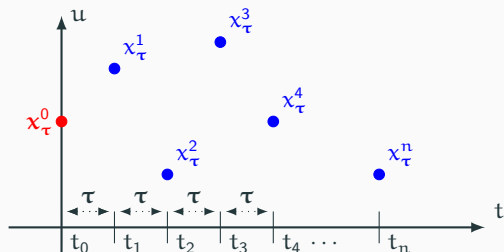
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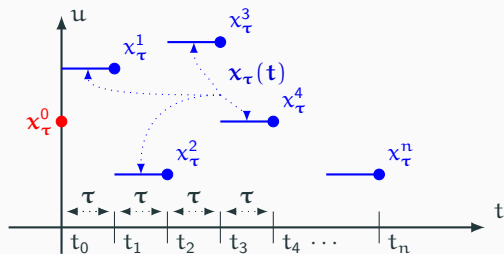
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\bar{x}_τ is the piecewise constant interpolant of the values $(x_\tau^n)_{n \in \mathbb{N}}$.

Convergence and characterization of the limit solution

Theorem (Crandall-Liggett '71)

If B is m -accretive, for every $x_0 \in \overline{D(B)}$ the discrete solutions \bar{x}_τ of the implicit Euler scheme converge uniformly to a limit curve $x \in C([0, \infty); H)$.

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If $x_0 \in D(B)$ then x is Lipschitz and solves

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$$\frac{1}{2} \frac{d}{dt} |x(t) - y|^2 \leq -\langle B(y), y - x(t) \rangle \quad \text{in } \mathcal{D}'(0, \infty), \quad \text{for every } y \in D(B).$$

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Formally

$$\begin{aligned} \frac{1}{2} \frac{d}{dt} |x(t) - y|^2 &= \langle Bx(t), x(t) - y \rangle = \langle Bx(t) - By, x(t) - y \rangle + \langle By, x(t) - y \rangle \\ &\leq \langle By, x(t) - y \rangle. \end{aligned}$$

PVFs and displacement extrapolation

In $\mathcal{P}_2(\mathbb{E})$ the role of the curve $\mathbf{x}(\tau) := \mathbf{x} + \tau\mathbf{B}(\mathbf{x})$ is played by

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

If $(\mathbf{X}, \mathbf{V})_{\sharp} \mathbb{P} = \underline{\mathbf{F}}$ we have

$$\underline{\mathbf{F}}(\tau) = (\mathbf{X} + \tau\mathbf{V})_{\sharp} \mathbb{P}$$

Semiconcavity of the Wasserstein distance

If $\underline{\mathbf{F}} \in \mathfrak{F}(\mu)$ and $\underline{\mathbf{G}} \in \mathfrak{F}(\nu)$, the map

$$D(\tau; \mu, \nu) := \frac{1}{2} W_2^2(\underline{\mathbf{F}}(\tau), \underline{\mathbf{G}}(\tau))$$

is not convex nor λ -convex for any $\lambda < 0$. In fact it is semiconcave, i.e.

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In particular

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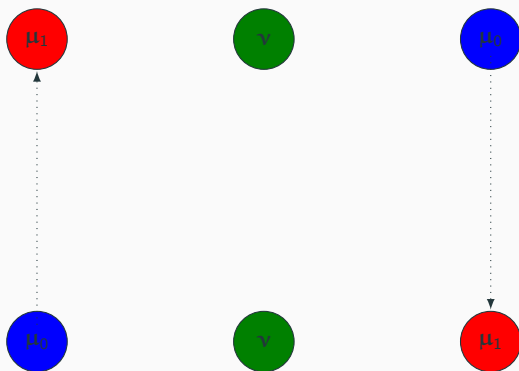
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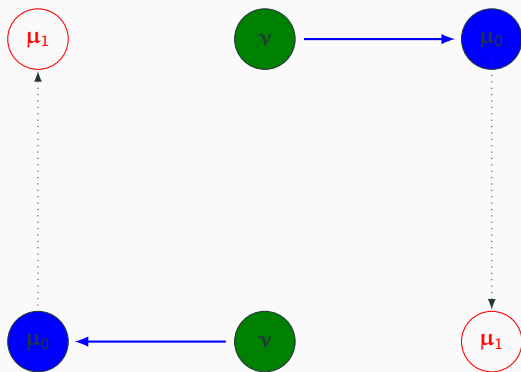
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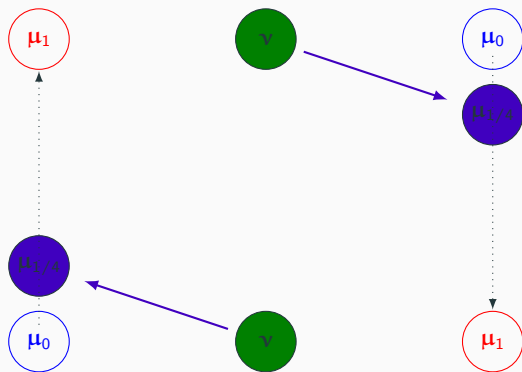
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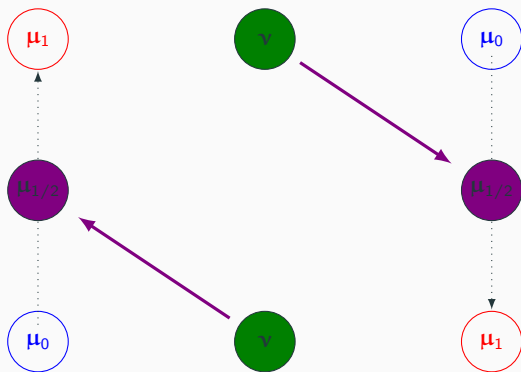
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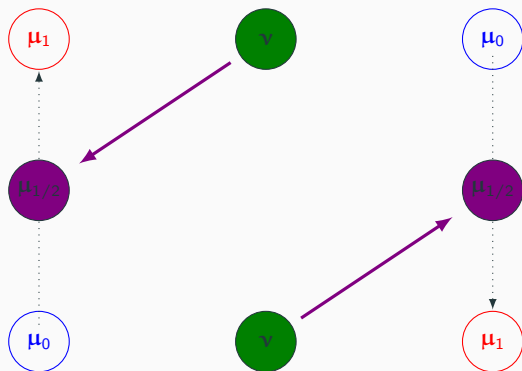
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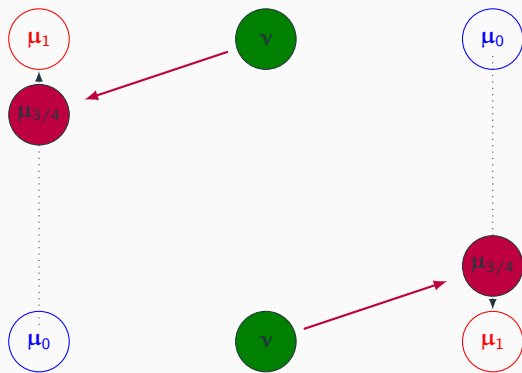
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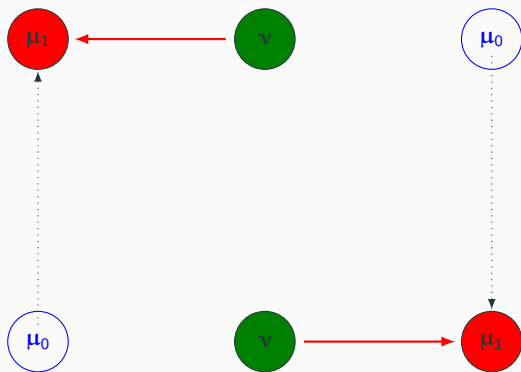
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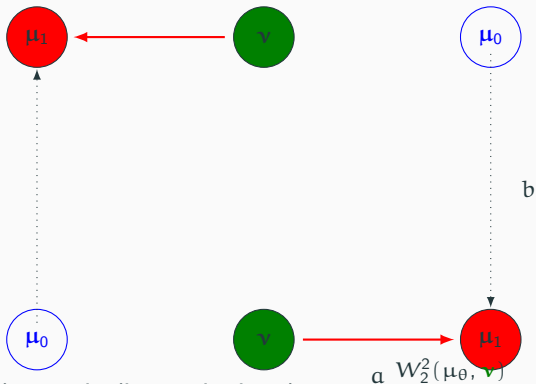
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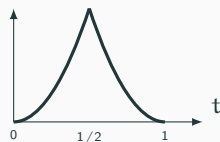
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The Wasserstein distance is given by

$$W_2^2(\nu, \mu_\theta) = \min \left(a^2 + b^2\theta^2, a^2 + b^2(1 - \theta)^2 \right)$$

It is not λ -convex, for any λ .



Metric dissipativity

We first compute the right derivative

$$\frac{d}{d\tau} \frac{1}{2} W_2^2(\underline{F}(\tau), \nu) \Big|_{\tau=0+}, \quad \underline{F}(\tau) = (\mathbf{x} + \tau \mathbf{v})_{\#} \underline{E}, \quad \underline{E} \in \mathfrak{F}(\mu)$$

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We introduce the set $\Gamma_o(\underline{F}, \nu)$ of couplings given by triple of random variables X, V, Y such that

$$(X, V)_{\#} \mathbb{P} = \underline{F}, \quad Y_{\#} \mathbb{P} = \nu \text{ and} \\ (X, Y)_{\#} \mathbb{P} \in \Gamma_o(\mu, \nu) \text{ is an } \text{optimal coupling} \text{ between } \mu \text{ and } \nu.$$

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\mathfrak{F} is (metrically) dissipative if [Cavagnari-Sodini-S.]

$$[\mathfrak{F}(\mu), v]_r + [\mathfrak{F}(v), \mu]_r \leq 0 \quad \text{for every } \mu, v \in \mathcal{P}_2(E).$$

If $\mathcal{F} : \mathcal{P}_2(E) \rightarrow (-\infty, +\infty]$ is a geodesically convex functional then its (opposite) Wasserstein subdifferential $\mathfrak{F} = -\partial_W \mathcal{F}$ is defined by

$$\underline{\mathbf{F}} \in \mathfrak{F}(\mu) \quad \Leftrightarrow \quad [\underline{\mathbf{F}}, \nu]_r \leq \mathcal{F}(\nu) - \mathcal{F}(\mu) \quad \text{for every } \nu \in D(\mathcal{F})$$

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The **Relative Entropy functional**

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

$\mathcal{F} \equiv +\infty$ on the discrete measures. \mathcal{F} is geodesically convex (i.e. convex along displacement interpolations, [McCann '97]) but not convex along arbitrary interpolation of measures: optimal interpolations avoid collisions!

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**.

⁵R. 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**⁵, at each step it is sufficient to select μ_τ^n among the minimizers of

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

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Let $\mathcal{F} : \mathcal{P}_2(E) \rightarrow (-\infty, +\infty]$ 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 $\nu \in D(\mathcal{F}) \subset \mathcal{P}_2(E)$

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Theorem (Ambrosio-Gigli-S.)

For every initial datum $\mu_0 \in \overline{D(\mathcal{F})}$ there exists a unique EVI solution to (EVI) satisfying $\lim_{t \downarrow 0} \mu_t = \mu_0$.

Moreover, μ is the uniform limit of piecewise constant interpolant μ_τ of the JKO-Minimizing Movement approximations, obtained by solving

$$\mu_\tau^n \in \operatorname{argmin}_{\mu} \left\{ \frac{1}{2\tau} W_2^2(\mu, \mu_\tau^{n-1}) + \mathcal{F}(\mu) \right\}, \quad \mu_\tau^0 := \mu_0$$

Uniform error estimate if $\mu_0 \in D(\mathcal{F})$:

$$W_2(\mu(t), \mu_\tau(t)) \leq C\sqrt{\tau}.$$

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- Metric **dissipativity does not imply contraction** of the resolvent.

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- Technical point: perturbations along \mathfrak{F} **can split particles**.
- \mathfrak{F} behaves well only along (optimal) displacement interpolations.

Probability vector fields and evolution

Dissipative operators and contraction semigroups in Hilbert spaces

Convergence of the Explicit Euler method and contraction semigroups

The explicit Euler method for dissipative evolutions

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 μ_0 . If \mathfrak{F} is a dissipative vector field and $\tau > 0$ is a time step, we consider the curves $\underline{F}(\tau)$ in $\mathcal{P}_2(E)$, $\underline{F} \in \mathfrak{F}(\mu)$

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

and therefore the sequence of explicit Euler approximations:

$$\mu_{\tau}^0 := \mu_0 \text{ given, } \mu_{\tau}^{n+1} := \underline{F}^n(\tau), \quad \underline{F}^n \in \mathfrak{F}(\mu_{\tau}^n), \quad \mu_{\tau}(t) := \mu_{\tau}^{\lfloor t/\tau \rfloor}$$

μ_{τ} is the **piecewise constant interpolation**, $\mu_{\tau}(t) = \mu_{\tau}^n$ if $n\tau \leq t < (n+1)\tau$.

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

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Easy “Lipschitz” estimate:

$$\frac{W_2(\mu_{\tau}^n, \mu_{\tau}^{n-1})}{\tau} \leq \left(\int |\nu|^2 d\underline{F}^n(x, \nu) \right)^{1/2}$$

$\mathcal{M}(\mu_0, \tau, L, T) :=$ set of discrete solutions μ_{τ} of the Explicit Euler method starting from μ_0 , defined up to the final time T , such that

$$\int |\nu|^2 d\underline{F}^n(x, \nu) \leq L^2 \quad \text{for every } n \leq \lfloor T/\tau \rfloor.$$

Theorem (Cavagnari-Sodini-S.)

Suppose that \mathfrak{F} is a dissipative MPVF.

- If $k \mapsto \tau(k) \downarrow 0$ is a vanishing sequence of step sizes and $\mu_k \in \mathcal{M}(\mu_0, \tau(k), L, T)$ for some $L \geq 0$, then the sequence of discrete solutions μ_k of the explicit Euler method uniformly converge to a unique limit $\mu : [0, T] \rightarrow \mathcal{P}_2(E)$.

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- We have the optimal error estimate

$$W_2(\mu(t), \mu_{\tau}(t)) \leq CL\sqrt{T\tau} \quad \text{for every } t \in [0, T].$$

Generation of a flow of contractions

Suppose that \mathfrak{F} is a dissipative MPVF such that $D(\mathfrak{F})$ contains all the measures with bounded support of $\mathcal{P}_b(E)$.

Suppose moreover that

- for every $\mu_0 \in \mathcal{P}_b(E)$ there exist $\rho, L > 0$ such that

$$W_2(\mu, \mu_0) < \rho \quad \Rightarrow \quad \exists \underline{F} \in \mathfrak{F}(\mu) : \text{supp}(v_{\#}\underline{F}) \subset B_L(0).$$

(local solvability of the Explicit Euler method)

- every $\underline{F} \in \mathfrak{F}$ is concentrated on the set

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Theorem

Then \mathfrak{F} generates a semigroup of contractions: for every $\mu_0 \in \mathcal{P}_2(E)$ there exists a unique continuous curve $\mu = S[\mu_0] \in C([0, \infty); \mathcal{P}_2(E))$ such that

$$\frac{d}{dt} \frac{1}{2} W_2^2(\mu(t), \nu) \leq -[F(\nu), \mu]_r \quad \text{for every } \nu \in D(\mathfrak{F}) \quad \mu(0) = \mu_0,$$

$$W_2(S_t[\mu_0], S_t[\nu_0]) \leq W_2(\mu_0, \nu_0) \quad \text{for every } \mu_0, \nu_0 \in \mathcal{P}_2(E), t > 0.$$

Barycentric property

Under the same conditions, let us also suppose that the sections $\mathfrak{F}(\mu)$ of \mathfrak{F} are convex and the graph of \mathfrak{F} is closed under strong-weak convergence: if a sequence $\underline{F}_n \in \mathfrak{F}(\mu_n)$ satisfies

$$\mu_n \rightarrow \mu \text{ in } \mathcal{P}_2(E), \quad \underline{F}_n \rightarrow \underline{F} \text{ in } \mathcal{P}(E \times E), \quad \sup_n \int |v|^2 d\underline{F}_n(x, v) < \infty$$

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Theorem

Every EVI solution $\mu : (0, \infty) \rightarrow D(\mathfrak{F})$ satisfies the barycentric property: for \mathcal{L}^1 -a.e. $t > 0$ there exists $\underline{F}_t \in \mathfrak{F}(\mu_t)$ such that

$$\frac{d}{dt} \int \zeta(x) d\mu_t(x) = \int \langle D\zeta(x), v \rangle d\underline{F}_t(x, v) \quad (\star)$$

for every smooth bounded Lipschitz cylindrical function $\zeta : E \rightarrow \mathbb{R}$.

Conversely, if $\mu : (0, \infty) \rightarrow D(\mathfrak{F})$ is absolutely continuous, it satisfies (\star) , and for a.e. $t > 0$ $\mu_t \in \mathcal{P}_2^+(E)$ or $\mathfrak{F}(\mu_t)$ contains a unique element concentrated on a map, μ is also an EVI solution.

(\star) is equivalent to $\partial_t \mu_t + \nabla \cdot (\mu_t v_t) = 0, \quad v_t = \text{proj}_{\text{Tan}(\mu_t)}(\underline{F}_t)$

- Everything can be easily extended to λ -dissipative probability vector fields.
- Evolutions do not split particles in dimension ≥ 2 ?
- Impose only local boundedness on \mathfrak{F}
- Implicit Euler scheme
- Stability and G-convergence
- ...