



# Variational convergence of metric gradient flows

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1. Geodesically convex functions and EVI flows in metric spaces
2. Minimizing movements and uniform error estimates
3. Mosco convergence and stability results: the Hilbertian case
4. Stability in the metric setting

## Motivation: gradient flows beyond Hilbert spaces

**Classical setting:** in a Hilbert space  $H$ , the gradient flow of a convex functional  $\phi : H \rightarrow (-\infty, +\infty]$  is

$$\dot{u}(t) \in -\partial\phi(u(t)), \quad t > 0.$$

**Key insight** [Jordan-Kinderlehrer-Otto '98, Ambrosio-Gigli-Savaré '05]:

Many important PDEs can be interpreted as gradient flows **in metric spaces without linear structure:**

- the **heat equation** is the gradient flow of the **Boltzmann entropy** in the Wasserstein space of probability measures;
- the **porous medium**, **Fokker-Planck**, and other nonlinear diffusion equations fit the same framework.

**Main question of this talk:**

*What happens when the energy functionals vary?*

*Can we deduce **convergence of the flows** from **variational convergence of the energies**?*

## The role of the EVI formulation

**In a Hilbert space**, convergence of gradient flows is equivalent to **Mosco convergence** of the energies [Attouch '84], and the proof relies on:

- **resolvent operators**  $(I + \tau\partial\phi)^{-1}$  (contractions),
- the **Crandall-Liggett exponential formula**.

**In a general metric space** these tools are not available.

Our approach is based on:

1. the **Evolution Variational Inequality (EVI)** formulation of gradient flows,
2. **quantitative error estimates** for the minimizing movement (implicit Euler) scheme,
3. a suitable notion of **variational convergence** for the energies and the metric slopes.

**For simplicity, we will present the results in the case  $\lambda = 0$**  (geodesic convexity without a lower curvature bound).

$(X, d)$  is a **complete metric space**.

$\phi : X \rightarrow (-\infty, +\infty]$  is a lower semicontinuous **geodesically convex functional** with non-empty domain  $D(\phi) := \{x \in X : \phi(x) < \infty\}$ .

### Geodesically convex functions

$\phi$  is **geodesically convex** if for every  $x_0, x_1 \in D(\phi) := \{x \in X : \phi(x) < \infty\}$  there exists a (minimal, constant speed) geodesic  $\gamma : [0, 1] \rightarrow X$  such that

$$\phi(\gamma_\vartheta) \leq (1 - \vartheta)\phi(x_0) + \vartheta\phi(x_1) \quad \vartheta \in [0, 1].$$

**Convexity along geodesics** generalizes convexity in linear spaces: the graph of  $\phi \circ \gamma$  lies below the chord joining  $\phi(x_0)$  and  $\phi(x_1)$ .

The **metric slope**  $|\partial\phi|$  is defined for  $x \in D(\phi)$  by

$$|\partial\phi|(x) := \limsup_{y \rightarrow x} \frac{(\phi(x) - \phi(y))_+}{\mathbf{d}(x, y)}$$

$|\partial\phi|(x) := +\infty$  if  $x \in X \setminus D(\phi)$ .

The metric slope is a **lower semicontinuous** map and plays the role of  $\min\{|v| : v \in \partial\phi(x)\}$  in the Hilbert space case.

It measures the **steepness** of  $\phi$  at  $x$ : a large slope means that  $\phi$  decreases rapidly near  $x$ .

### Evolution variational inequalities [Ambrosio-Gigli-S. '05]

A locally Lipschitz curve  $\mathbf{u} : (0, \infty) \rightarrow D(\phi)$  is a solution to the  $\text{EVI}(X, \mathbf{d}, \phi)$  if

$$\frac{1}{2} \frac{d}{dt} \mathbf{d}^2(\mathbf{u}(t), v) \leq \phi(v) - \phi(\mathbf{u}(t)) \quad \text{for a.e. } t > 0, v \in D(\phi).$$

### Gradient flows

A *gradient flow* of  $\phi$  is a family of continuous maps  $\mathbf{S}_t : \overline{D(\phi)} \rightarrow \overline{D(\phi)}$ ,  $t \geq 0$ , such that for every  $u_0 \in \overline{D(\phi)}$

$$\mathbf{S}_{t+h}(u_0) = \mathbf{S}_h(\mathbf{S}_t(u_0)) \quad \text{for every } t, h \geq 0, \quad \lim_{t \downarrow 0} \mathbf{S}_t(u_0) = \mathbf{S}_0(u_0) = u_0,$$

the curve  $t \mapsto \mathbf{S}_t(u_0)$  is a solution of  $\text{EVI}(X, \mathbf{d}, \phi)$ .

$(X, |\cdot|)$  is a **Hilbert space**,  $d(x, y) = |x - y|$ ,  $x \mapsto \phi(x)$  is convex.

$u$  a solution to  $\text{EVI}(X, d, \phi)$  if and only if  $u$  is locally Lipschitz and

$$\dot{u}(t) \in -\partial\phi(u(t)) \quad t > 0,$$

where  $v \in \partial\phi(x) \Leftrightarrow \langle v, y - x \rangle \leq \phi(y) - \phi(x)$  for every  $y \in X$ . In this case  
 $|\partial\phi|(x) = \min \{|v| : v \in \partial\phi(x)\}$ .

The **EVI formulation** follows from the differentiability of the squared distance:

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

**Similarly on a complete Riemannian manifold**  $(X, g)$ : if  $\phi \in C^2(X)$  with  $\text{Hess } \phi \geq 0$ , then  $u$  solves  $\text{EVI}$  iff  $\dot{u}(t) = -\nabla\phi(u(t))$ ,  $t > 0$ .

## Examples: RCD metric measure spaces

Let  $(X, d, \mathbf{m})$  be a complete metric measure space,  $\mathbf{m} \in \mathcal{P}(X)$ . Let  $\mathcal{X} := \mathcal{P}_2(X)$  endowed with the Wasserstein distance  $W_d$ .

The gradient flow of the entropy in  $\mathcal{X}$  is the heat flow. Let  $\phi(\mu) := \text{Ent}(\mu|\mathbf{m})$ ,

$$\text{Ent}(\mu|\mathbf{m}) = \int \rho \log \rho \, d\mathbf{m} \quad \text{if } \mu = \rho\mathbf{m}; \quad +\infty \text{ if } \mu \not\ll \mathbf{m}.$$

$(X, d, \mathbf{m})$  is RCD(0,  $\infty$ ) iff  $\text{Ent}(\cdot|\mathbf{m})$  admits a gradient flow in  $(\mathcal{P}_2(X), W_d)$ .

In this case  $S_t\mu = \mathbf{u}(t, \cdot)\mathbf{m}$  where  $\mathbf{u}$  solves

$$\partial_t \mathbf{u} = \Delta \mathbf{u} \quad t > 0; \quad \lim_{t \downarrow 0} W_d(\mathbf{u}(t, \cdot)\mathbf{m}, \mu) = 0.$$

Let  $\mathbf{u}^1, \mathbf{u}^2 \in C^0([0, +\infty); X)$  be solutions of  $\text{EVI}(X, \mathbf{d}, \phi)$ .

- **Contraction and uniqueness:**

$$\mathbf{d}(\mathbf{u}^1(t), \mathbf{u}^2(t)) \leq \mathbf{d}(\mathbf{u}^1(0), \mathbf{u}^2(0)) \quad \text{for every } t > 0.$$

In particular, for every  $\mathbf{u}_0 \in \overline{D(\phi)}$  there is at most one solution  $\mathbf{u}$  of  $\text{EVI}(X, \mathbf{d}, \phi)$  satisfying the initial condition  $\lim_{t \downarrow 0} \mathbf{u}(t) = \mathbf{u}_0$ .

- **Regularizing effect:**  $\mathbf{u}$  is *locally Lipschitz* in  $(0, +\infty)$  and the map  $t \mapsto \phi(\mathbf{u}(t))$  is *nonincreasing*. Moreover, the **energy identity** holds:

$$\frac{d}{dt} \phi(\mathbf{u}(t+)) = -|\dot{\mathbf{u}}(t+)|^2 = -|\partial\phi|^2(\mathbf{u}(t)) \quad \text{for every } t > 0.$$

## Minimizing movements

Given  $\tau > 0$  and  $U_\tau^0 \approx u_0$ , consider  $\Phi(\tau, U, V) := \frac{1}{2\tau} d^2(U, V) + \phi(V)$ .

$(U_\tau^n)_n$  is a **discrete minimizing sequence** if  $U_\tau^n \in \text{Argmin } \Phi(\tau, U_\tau^{n-1}, \cdot)$ , i.e.

$$\frac{1}{2\tau} d^2(U_\tau^{n-1}, U_\tau^n) + \phi(U_\tau^n) \leq \frac{1}{2\tau} d^2(U_\tau^{n-1}, V) + \phi(V) \quad \text{for every } V \in X.$$

The **discrete minimizing movement** is the piecewise constant interpolant  $\mathbf{U}_\tau$ :

$$\mathbf{U}_\tau(t) := U_\tau^n \quad \text{if } t \in ((n-1)\tau, n\tau].$$

### Minimizing Movements

$u$  is a **Minimizing Movement** in  $MM(\Phi, u_0)$  if there exists  $(\mathbf{U}_\tau)_\tau$  such that

$$u(t) = \lim_{\tau \downarrow 0} \mathbf{U}_\tau(t) \quad \text{for every } t > 0.$$

## Ekeland relaxation of minimizing movements

Consider a relaxation parameter  $\eta \in (0, 1)$ .

Applying **Ekeland's variational principle** to  $V \mapsto \Phi(\tau, U_\tau^{n-1}, V)$ , starting from  $W := U_\tau^{n-1}$  we find  $U_\tau^n$  satisfying

$$\frac{1}{2\tau} d^2(U_\tau^{n-1}, U_\tau^n) + \phi(U_\tau^n) \leq \frac{1}{2\tau} d^2(U_\tau^{n-1}, V) + \phi(V) + \eta d(U_\tau^{n-1}, U_\tau^n) d(U_\tau^n, V)$$

for every  $V \in X$ , and

$$\frac{1}{2\tau} d^2(U_\tau^{n-1}, U_\tau^n) + \phi(U_\tau^n) \leq \phi(U_\tau^{n-1}).$$

For  $\eta = 0$  we recover the usual Minimizing Movement scheme.

We call  $U_{\tau, \eta}$  any **piecewise constant interpolant** of a discrete  $\eta$ -Ekeland-minimizing sequence.

## Resolvent and Crandall-Liggett estimates in the Hilbert case

In a Hilbert space,  $\mathbf{A} := \partial\phi$ . The Minimizing Movement scheme is the backward Euler method, solved by the **resolvent**  $\mathbf{J}_\tau := (\mathbf{I} + \tau\mathbf{A})^{-1}$ :

$$\frac{\mathbf{u}_\tau^n - \mathbf{u}_\tau^{n-1}}{\tau} + \mathbf{A}\mathbf{u}_\tau^n \ni 0, \quad \mathbf{u}_\tau^n = \mathbf{J}_\tau \mathbf{u}_\tau^{n-1}.$$

$\mathbf{J}_\tau$  is a contraction  $\Rightarrow$  **Crandall-Liggett exponential formula**:

$$\mathbf{u}(t) = \lim_{n \rightarrow \infty} \mathbf{J}_{t/n}^n \mathbf{u}_0, \quad |\mathbf{u}(t) - \mathbf{u}_\tau(t)| \leq 2|\mathbf{A}(\mathbf{u}_0)| \sqrt{T\tau} \quad t \in [0, T].$$

The Ekeland minimization provides a solution of

$$\frac{\mathbf{u}_\tau^n - \mathbf{u}_\tau^{n-1}}{\tau} + \mathbf{A}\mathbf{u}_\tau^n \ni \Xi_\tau^n, \quad |\Xi_\tau^n| \leq \eta |\mathbf{u}_\tau^n - \mathbf{u}_\tau^{n-1}|.$$

### Theorem (A priori error estimates)

Suppose that  $\phi$  admits a gradient flow and fix a time interval  $T > 0$ . There exists a constant  $C = C(T)$  such that every discrete  $\eta$ -Ekeland minimizing movement  $\mathbf{U}_{\tau, \eta}$  and every solution  $\mathbf{u}$  of the  $\text{EVI}(X, \mathbf{d}, \phi)$  satisfy

$$\mathbf{d}(\mathbf{u}(t), \mathbf{U}_{\tau, \eta}(t)) \leq \mathbf{d}(\mathbf{u}_0, \mathbf{U}_{\tau, \eta}^0) + C|\partial\phi|(\mathbf{U}_{\tau, \eta}^0) \sqrt{\tau}.$$

In particular, if  $\mathbf{U}_{\tau, \eta}^0 = \mathbf{u}_0$ ,

$$\mathbf{d}(\mathbf{u}(t), \mathbf{U}_{\tau, \eta}(t)) \leq C|\partial\phi|(\mathbf{u}_0) \sqrt{\tau} \quad t \in [0, T].$$

### Corollary (Minimizing Movements coincide with gradient flows)

If  $\phi$  admits a gradient flow then  $\text{MM}(\Phi, \mathbf{u}_0)$  coincides with the solution  $\mathbf{u}(t) = \mathbf{S}_t \mathbf{u}_0$  of the  $\text{EVI}(X, \mathbf{d}, \phi)$  starting from  $\mathbf{u}_0$ .

## The stability problem and Mosco convergence

**The stability problem:** given  $\phi^h$  geodesically convex with gradient flows  $S^h$ , and  $\phi^h \rightarrow \phi$  in a suitable sense, **does  $S_t^h u \rightarrow S_t u$ ?**

In a Hilbert space we can use the **weak and strong topology**.

$\phi^h$  **Mosco-converges to  $\phi$**  if

$$\begin{aligned} x^h \rightharpoonup x &\Rightarrow \liminf_{h \rightarrow \infty} \phi^h(x^h) \geq \phi(x) \\ \forall x \in X \quad \exists (x^h)_h : &\quad x^h \rightarrow x, \quad \phi^h(x^h) \rightarrow \phi(x). \end{aligned}$$

Equivalently:  $\phi^h$   $\Gamma$ -converges to  $\phi$  w.r.t. **both** the weak and the strong topology:

$$\phi = M\text{-}\lim_{h \rightarrow \infty} \phi^h \iff w\text{-}\Gamma\text{-}\liminf_{h \rightarrow \infty} \phi^h = s\text{-}\Gamma\text{-}\limsup_{h \rightarrow \infty} \phi^h = \phi.$$

Suppose that  $X$  is a Hilbert space,  $(\phi^h)_{h \in \mathbb{N}}$ ,  $\phi$  l.s.c. and convex.

### Theorem (Benilan, Crandall, Pazy, Attouch)

The following properties are equivalent:

- **Convergence of the flows:** whenever  $u_0^h \rightarrow u_0 \in D(\phi)$ ,  $u_0^h \in D(\phi^h)$ , we have

$$\lim_{h \rightarrow \infty} S_t^h u_0^h = S_t u_0 \quad \text{for every } t \geq 0.$$

- **Mosco-convergence:**  $\phi^h \xrightarrow{M} \phi$ .
- **Convergence of the resolvents:**  $(I + \tau \mathbf{A}^h)^{-1} u_0 \rightarrow (I + \tau \mathbf{A})^{-1} u_0 \quad \forall u_0 \in X, \tau > 0$ .
- **Moreau-Yosida regularizations:** for every  $u \in X, \tau > 0$ :

$$\inf_v \phi^h(v) + \frac{1}{2\tau} \mathbf{d}^2(v, u) \rightarrow \inf_v \phi(v) + \frac{1}{2\tau} \mathbf{d}^2(v, u)$$

- **G-convergence of  $\mathbf{A}^h$ :**  $\forall v \in \mathbf{A}(u) \exists u^h, v^h: v^h \in \mathbf{A}^h u^h, u^h \rightarrow u, v^h \rightarrow v$ .

- Mosco limits of convex functionals are convex: the **limit semigroup  $\mathcal{S}$**  exists thanks to **Crandall-Liggett**.  
In metric spaces, the existence of the limit flow is not guaranteed a priori.
- Bounded sequences  $(x^h)_h$  are always weakly compact in Hilbert spaces; without strong coercivity, one must deal with **limits of  $\phi^h(x^h)$**  along weakly convergent sequences, so that the weak  $\Gamma$ -lim inf is involved.  
In metric spaces, a “weak” topology is generally missing.

- [MOSCO '67-'69] obstacle problems; [SPAGNOLO '67, DE GIORGI-SPAGNOLO '73] heat flow,  $\Gamma$ -convergence.
- [S.-VISINTIN '97, AMBROSIO-S.-ZAMBOTTI '09, PELETIER-S.-VENERONI '10]: **energies and metrics depend on  $h$ .**
- [AMBROSIO-GIGLI-S. '05-14, GIGLI-MONDINO-S. '15]: **stability of RCD spaces.**
- [ORTNER '05, SERFATY '11]: slopes, non-convex setting. [VISINTIN '14-'17]: **Fitzpatrick approach.**
- [MIELKE, ROUBICEK, STEFANELLI; MIELKE-ROSSI-S.]: **rate-independent** problems.

## Stability via error estimates

In the Hilbert space setting, stability follows by Crandall-Liggett estimate. Suppose that  $u_0^h \rightarrow u_0$ ,  $\sup_h |\mathbf{A}^h(u_0^h)| < \infty$ ,  $t \in [0, T]$  and  $\tau = T/N$ .

- First of all we replace the family  $\mathbf{u}^h(t) = \mathbf{S}_t^h(u_0^h)$  with the discrete minimizing movement  $\mathbf{U}_\tau^h$  starting from  $u_0^h$ : we pay an error  $\mathbf{d}(\mathbf{u}^h(t), \mathbf{U}_\tau^h(t)) \leq \varepsilon_\tau = C\sqrt{\tau}$ ,  $t \in [0, T]$ , **independent of  $h$** .
- Similarly, we replace the limit solution  $\mathbf{u}(t) = \mathbf{S}_t(u_0)$  with the discrete minimizing movement  $\mathbf{U}_\tau$  starting from  $u_0$  and we pay an error  $\mathbf{d}(\mathbf{u}(t), \mathbf{U}_\tau(t)) \leq \varepsilon_\tau = C\sqrt{\tau}$ ,  $t \in [0, T]$ .
- By triangle inequality

$$\mathbf{d}(\mathbf{u}^h(t), \mathbf{u}(t)) \leq \mathbf{d}(\mathbf{U}_\tau^h(t), \mathbf{U}_\tau(t)) + 2C\sqrt{\tau}$$

- We fix  $\varepsilon > 0$  and choose  $\tau$  sufficiently small, so that  $2C\sqrt{\tau} \leq \varepsilon/2$ .  $\mathbf{U}_\tau^h$  and  $\mathbf{U}_\tau$  depend on a **fixed** number of time steps  $N := T/\tau$ , and thus of iterations of the resolvent operators.

**The convergence of the resolvent** implies that for  $h \geq \bar{h}$  sufficiently big

$$\mathbf{d}(\mathbf{U}_\tau^h(t), \mathbf{U}_\tau(t)) \leq \varepsilon/2.$$

We deduce

$$\mathbf{d}(\mathbf{u}^h(t), \mathbf{u}(t)) \leq \varepsilon \quad \text{if } h \geq \bar{h}.$$

- We do not know a priori that the **limit flow  $S$  exists**.
- **Resolvent are not well defined** and we should use the relaxed Ekeland version.
- Even if resolvents are well defined, **they are not contractions**.
- A **“weak” topology is missing** in general.
- We would like to **study stability without assuming strong coercivity of the functionals**.
- Minimizers (and, even worse, Ekeland minimizers) do not behave well under  $\Gamma$ -convergence without coercivity.

## Main stability result: equivalent conditions (I)

Let  $\phi^h, \phi$  be l.s.c. and geodesically convex.

### Theorem (Muratori-S.)

The following properties are equivalent:

1. **Convergence of the flows:** whenever  $u_0^h \rightarrow u_0 \in D(\phi)$ ,  $u_0^h \in D(\phi^h)$ ,

$$\lim_{h \rightarrow \infty} S^h(t)u_0^h = S(t)u_0 \quad \text{for every } t \geq 0.$$

2. **Recovery sequence for  $\phi$  and  $|\partial\phi|$ :** for every  $u \in D(|\partial\phi|)$  there exist  $u^h \in D(|\partial\phi^h|)$  such that

$$u^h \rightarrow u, \quad \phi^h(u^h) \rightarrow \phi(u), \quad \limsup_{h \rightarrow \infty} |\partial\phi^h|(u^h) \leq |\partial\phi|(u).$$

3.  **$\Gamma$ -convergence for  $\phi^h$  and  $|\partial\phi^h|$ :**  $\Gamma\text{-}\lim_{h \rightarrow \infty} \phi^h = \phi$ ,  $\Gamma\text{-}\lim_{h \rightarrow \infty} |\partial\phi^h| = |\partial\phi|$ .

Condition 2 is the **metric counterpart** of the Hilbertian recovery condition  $u^h \rightarrow u$ ,  $v^h \rightarrow v$ ,  $v^h \in \mathbf{A}^h(u^h)$ ,  $\phi^h(u^h) \rightarrow \phi(u)$ .

## Main stability result: equivalent conditions (II)

### Theorem (continued)

The previous conditions are also equivalent to:

- 4. Qualified  $\Gamma$ -convergence:**  $\Gamma\text{-lim sup}_{h \rightarrow \infty} \phi^h \leq \phi$  and for every  $u \in D(\phi)$ ,  $\varepsilon > 0$  and  $\bar{\tau} > 0$ , there exists  $\tau \in (0, \bar{\tau})$  such that

$$\liminf_{h \rightarrow \infty} \inf_{B_\tau(u)} \phi^h \geq \inf_{B_\tau(u)} \phi - \varepsilon \tau$$

- 5. Local Yosida regularization:**  $\Gamma\text{-lim sup}_{h \rightarrow \infty} \phi^h \leq \phi$  and for every  $u \in D(\phi)$ ,  $\varepsilon > 0$  and  $\bar{\tau} > 0$ , there exists  $\tau \in (0, \bar{\tau})$  such that

$$\liminf_{h \rightarrow \infty} \inf_v \phi^h(v) + \frac{1}{2\tau} \mathbf{d}^2(u, v) \geq \inf_v \phi(v) + \frac{1}{2\tau} \mathbf{d}^2(u, v) - \varepsilon \tau$$

Conditions 4 and 5 are **purely variational** and do not involve the metric slope: they provide a **quantitative reinforcement** of  $\Gamma$ -convergence that controls the local behavior of the infima.

Suppose that  $X$  is endowed with a weaker topology  $\sigma$  such that

- $x_n \xrightarrow{\sigma} x$  yields  $\liminf_{n \rightarrow \infty} \mathbf{d}(x_n, y) \geq \mathbf{d}(x, y)$  for every  $y \in X$
- every  $(x^h)_h$  bounded sequence with  $\phi^h(x^h) \leq C$  admits a  $\sigma$ -convergence subsequence.

We suppose that  $\phi^h$  Mosco converges to  $\phi$  in  $(X, \mathbf{d}, \sigma)$ :

$$\begin{aligned} x^h \xrightarrow{\sigma} x &\Rightarrow \liminf_{h \rightarrow \infty} \phi^h(x^h) \geq \phi(x) \\ \forall x \in X \quad \exists (x^h)_h : &\quad x^h \rightarrow x, \quad \phi^h(x^h) \rightarrow \phi(x). \end{aligned}$$

### Theorem

*If  $\phi^h$  admits a gradient flow  $S^h$  then also  $\phi$  admits a gradient flow and  $S_t^h u_0^h \rightarrow S_t u_0$  for every  $u_0^h \in \overline{D(\phi^h)}$  converging to  $u_0$ .*

## Two basic estimates ( $\eta = 0$ )

(Ekeland) minimizers satisfy a discrete energy inequality:

$$\tau |\partial\phi|^2(\mathbf{U}_\tau^n) \leq \frac{\mathbf{d}^2(\mathbf{U}_\tau^{n-1}, \mathbf{U}_\tau^n)}{\tau} \leq \phi(\mathbf{U}_\tau^{n-1}) - \phi(\mathbf{U}_\tau^n) \quad (\star)$$

**Key point:** if  $(\star)$  holds up to an error  $\varepsilon > 0$  at the level  $h$ , one still gets

$$\mathbf{d}(\mathbf{u}^h(t), \mathbf{U}_\tau^h(t)) \leq \mathbf{d}(\mathbf{u}_0, \mathbf{U}_\tau^{0,h}) + C(\sqrt{\tau} + \sqrt{\varepsilon})$$

provided  $\mathbf{u}_0 \in D(|\partial\phi|)$  and  $\sup_{h,\tau} |\partial\phi|(\mathbf{U}_\tau^{0,h}) < \infty$ .

- Generate a sequence of (Ekeland) minimizers for  $\phi$  satisfying

$$\tau |\partial\phi|^2(\mathbf{U}_\tau^n) \leq \frac{\mathbf{d}^2(\mathbf{U}_\tau^{n-1}, \mathbf{U}_\tau^n)}{\tau} \leq \phi(\mathbf{U}_\tau^{n-1}) - \phi(\mathbf{U}_\tau^n), \quad (\star)$$

- Use  $\Gamma$  convergence of  $\phi^h$  and  $|\partial\phi^h|$  to approximate  $\mathbf{U}_\tau^n$  by sequences  $\mathbf{U}_\tau^{n,h}$  satisfying  $(\star)$  up to a small error  $\varepsilon > 0$  with

$$\lim_{h \rightarrow \infty} \sup_{t \in [0, T]} \mathbf{d}(\mathbf{U}_\tau(t), \mathbf{U}_\tau^h(t)) = 0.$$

- Use the a priori error estimate to prove that

$$\mathbf{d}(\mathbf{u}^h(t), \mathbf{U}_\tau^h(t)) \leq \mathbf{d}(\mathbf{u}_0, \mathbf{U}_\tau^{0,h}) + C(\sqrt{\tau} + \sqrt{\varepsilon})$$

- Combining the two estimates, we deduce

$$\limsup_{h, k \rightarrow \infty} \mathbf{d}(\mathbf{u}^h(t), \mathbf{u}^k(t)) \leq 2C(\sqrt{\tau} + \sqrt{\varepsilon})$$

which shows that  $(\mathbf{u}^h(t))_h$  is a Cauchy sequence, since  $\tau > 0$  and  $\varepsilon > 0$  are arbitrary.

Let  $(X, \mathbf{d}, \mathbf{m})$  be a  $\text{RCD}(K, \infty)$  metric measure space and let  $\varphi : X \rightarrow (-\infty, +\infty]$  be a continuous and geodesically convex function.

### Theorem (Sturm '14)

*If  $(X, \mathbf{d})$  is locally compact then  $\varphi$  admits a gradient flow.*

### Corollary

*The local compactness assumption can be removed.*

The proof relies on the construction of the gradient flow for the functional

$$\phi(\mu) := \int_X \varphi \, d\mu \quad \text{in } (\mathcal{P}_2(X), W_{\mathbf{d}})$$

by considering the approximation  $\phi^h(\mu) := \phi(\mu) + \frac{1}{h} \text{Ent}(\mu|\mathbf{m})$ . At least when  $\mathbf{m} \in \mathcal{P}(X)$  it is easy to check that the assumptions of the main stability result are satisfied.

### Main contributions:

- **Quantitative error estimates** for the minimizing movement scheme in metric spaces, uniform w.r.t. approximating functionals.
- A complete characterization of **variational convergence of gradient flows**: five equivalent conditions, from convergence of flows to qualified  $\Gamma$ -convergence.
- No need for strong coercivity or linear structure: the results apply to **Wasserstein gradient flows** and **RCD spaces**.

### Further results (not discussed here for simplicity):

- The full  $\lambda$ -convex and approximately convex settings.
- Stability under Gromov-Hausdorff convergence of the underlying metric spaces.

**Open:** convergence **rates** for the flows in terms of the approximation parameters.

Thank you!