

Structure Preserving Discretizations of Wasserstein Gradient Flows

Part I of IV: The object to discretize

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Where do we need to go today?

Question

What is that alleged interesting structure of a Wasserstein gradient flow that a discretization should preserve?

What was a Wasserstein gradient flow again?

Let us rewrite Wasserstein
as Riemann.

An ODE analogy for gradient flows (1/3)

A function $e : \mathbb{R}^n \rightarrow \mathbb{R}$ defines an **euclidean gradient flow** via

$$\dot{x} = -\nabla e(x).$$

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Solutions $(x_t)_{t \geq 0}$ diminish e “as fast as possible”:

$$-\frac{d}{dt}e(x_t) = -\dot{x}_t \cdot \nabla e(x_t) = \|\dot{x}_t\|^2 = \|\nabla e(x_t)\|^2,$$

i.e., $(x_t)_{t \geq 0}$ invests all of its velocity into dissipating e .

An ODE analogy for gradient flows (2/3)

On a manifold $\mathcal{M}$ the tangent space $T_x\mathcal{M}$ at $x \in \mathcal{M}$ is formed by velocities $v \cong \dot{x}_0$ of curves $(x_s)_{s \in (-\sigma, \sigma)}$ with $x_0 = x$.

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A Riemannian tensor assigns to each $x \in \mathcal{M}$ a scalar product $\langle \cdot, \cdot \rangle_x$. It defines a norm on $T_x\mathcal{M}$ and a distance on $\mathcal{M}$:

$$\|v\|_x^2 = \langle v, v \rangle_x \quad \text{and} \quad \mathbf{d}(x_0, x_1)^2 = \inf_{(x_s)_{s \in [0,1]}} \int_0^1 \|v_s\|_{x_s}^2 \, ds.$$

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The differential $D_x e$ of a function $e : \mathcal{M} \rightarrow \mathbb{R}$ assigns to each $v \in T_x\mathcal{M}$ the directional derivative,

$$D_x e[v] = \left. \frac{d}{ds} \right|_{s=0} e(x_s) \quad \text{with} \quad v \cong \dot{x}_0.$$

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The gradient $\nabla_x e$ is the Riesz representative of $D_x e$ w.r.t. $\langle \cdot, \cdot \rangle_x$,

$$\langle \nabla_x e, v \rangle_x = D_x e[v] \quad \text{for all } v \in T_x\mathcal{M}.$$

An ODE analogy for gradient flows (3/3)

Accordingly, $e : \mathcal{M} \rightarrow \mathbb{R}$ defines a **Riemannian gradient flow** via

$$\dot{x} \cong -\nabla_x e.$$

By definition, solutions $(x_t)_{t \geq 0}$ satisfy

$$-\frac{d}{dt}e(x_t) = -D_{x_t}e[v_t] = \langle -\nabla_{x_t}e, v_t \rangle_{x_t} = \|\nabla_{x_t}e\|_{x_t}^2 = \|v_t\|_{x_t}^2,$$

i.e., all velocity is invested into dissipating e .

On the “manifold” of **probability densities**,

$$\mathcal{P}_2(\mathbb{R}^d) = \left\{ \rho \in L^1(\mathbb{R}^d) \left| \rho \geq 0, \int \rho \, dx = 1, \int |x|^2 \rho \, dx < \infty \right. \right\},$$

the **L^2 -Wasserstein metric** is a distance

$$\mathbb{W}(\rho_0, \rho_1)^2 = \int \left\{ \int |T(x) - x|^2 \rho_0(x) \, dx \left| T\# \rho_0 = \rho_1 \right. \right\}.$$

Can one define a Riemannian metric on $\mathcal{M} = \mathcal{P}_2(\mathbb{R}^d)$ such that $\mathbf{d} = \mathbb{W}$ (...and make the according gradient flow explicit)?

Otto's tangent space

The **tangent space** $T_\rho \mathcal{P}_2(\mathbb{R}^d)$ is identified with functions $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$ (modulo constants) such that

$$\|\varphi\|_\rho^2 := \int |\nabla \varphi|^2 \rho \, dx < \infty.$$

The tangent vector $\varphi \in T_\rho \mathcal{P}_2(\mathbb{R}^d)$ relates to curves $(\rho_s)_{s \in (-\sigma, \sigma)}$ s.t.

$$\dot{\rho}_s = -\operatorname{div}(\rho_s \nabla \varphi_s) \quad \text{with } \rho_0 = \rho, \varphi_0 = \varphi.$$

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Theorem ([Benamou&Brenier'00])

$$\mathbb{W}(\rho_0, \rho_1)^2 = \inf_{\rho_s} \int_0^1 \|\varphi_s\|_{\rho_s}^2 \, ds.$$

Otto calculus (3/3)

The norm $\|\cdot\|_\rho$ induces the **metric tensor**

$$\langle \varphi, \psi \rangle_\rho = \int \nabla \varphi \cdot \nabla \psi \rho \, dx.$$

The **gradient** ψ of $\mathbf{E}$ at ρ is thus characterized by

$$\int \nabla \varphi \cdot \nabla \psi \rho \, dx = \left. \frac{d}{ds} \right|_{s=0} \mathbf{E}(\rho_s) \quad \text{with} \quad \dot{\rho}_s = -\operatorname{div}(\rho \nabla \varphi).$$

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The **differential** of a functional $\mathbf{E} : \mathcal{P}_2(\mathbb{R}^d) \rightarrow \mathbb{R}$ at ρ is implicitly defined via $\delta \mathbf{E} / \delta \rho : \mathbb{R}^d \rightarrow \mathbb{R}$ such that

$$\int \zeta \frac{\delta \mathbf{E}}{\delta \rho} \, dx = \left. \frac{d}{ds} \right|_{s=0} \mathbf{E}(\rho + s\zeta) \quad \text{for all suitable } \zeta : \mathbb{R}^d \rightarrow \mathbb{R}.$$

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Gradient flow equation

$$\partial_t \rho = \operatorname{div} \left(\rho \nabla \left(\frac{\delta \mathbf{E}}{\delta \rho} \right) \right).$$

Does it get even more abstract?

There are interesting examples.

Standard example 1: Continuity equation

Potential energy

For an **external potential** $V : \mathbb{R}^d \rightarrow \mathbb{R}$ consider

$$E(\rho) = \int V(x)\rho(x) \, dx.$$

Then $\delta E / \delta \rho = V$, and accordingly,
the $\mathbb{W}$ -gradient flow is the **continuity equation**,

$$\partial_t \rho = \operatorname{div} (\rho \nabla V).$$

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Let $(x_t)_{t \geq 0}$ satisfy

$$\dot{x}_t = -\nabla V(x_t).$$

Then a distributional solution to the continuity equation is

$$\rho_t = \delta_{x_t}.$$

Standard example 2: Nonlocal aggregation equation

Interaction energy

For an even **interaction potential** $W : \mathbb{R}^d \rightarrow \mathbb{R}$ consider

$$E(\rho) = \frac{1}{2} \int \rho W * \rho \, dx = \frac{1}{2} \iint \rho(x) W(x - x') \rho(x') \, dx \, dx'.$$

Then $\delta E / \delta \rho = W * \rho$, and accordingly,
the $\mathbb{W}$ -gradient flow is the **nonlocal aggregation equation**,

$$\partial_t \rho = \operatorname{div} (\rho \nabla W * \rho).$$

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the $\mathbb{W}$ -gradient flow is the **nonlocal aggregation equation**,

$$\partial_t \rho = \operatorname{div} (\rho \nabla W * \rho).$$

Let N particle positions $x_1, \dots, x_N$ satisfy

$$\dot{x}_k = - \sum_{j \neq k} \nabla W(x_k - x_j).$$

Then a distributional solution to the nonlocal aggregation equation is

$$\rho_t = \frac{1}{N} \sum_k \delta_{x_k(t)}.$$

Standard example 3: Linear diffusion

Boltzmann entropy

Consider the Boltzmann entropy $E := H$, i.e.,

$$H(\rho) = \int \rho \log \rho \, dx.$$

Then $\delta E / \delta \rho = \log \rho$, and accordingly,
the $\mathbb{W}$ -gradient flow is the linear diffusion equation,

$$\partial_t \rho = \operatorname{div} (\rho \nabla \log \rho) = \Delta \rho.$$

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The solution for initial datum $\bar{\rho}$ is given by

$$\rho_t = H_t * \bar{\rho} \quad \text{with} \quad H_t(z) = (4\pi t)^{-d/2} \exp\left(-\frac{|z|^2}{4t}\right).$$

Standard example 4: Nonlinear Diffusion

Internal energy

For a convex **internal energy density** $h : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}$, consider

$$\mathbf{E}(\rho) = \int h(\rho) \, dx.$$

Then $\delta \mathbf{E} / \delta \rho = h'(\rho)$, and accordingly,
the $\mathbb{W}$ -gradient flow is the **nonlinear diffusion equation**,

$$\partial_t \rho = \Delta P(\rho) \quad \text{with} \quad P(r) = rh'(r) - h(r).$$

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Of particular interest: **power-type** energy resp. **homogeneous** diffusion,

$$\mathbf{E}(\rho) = \int \frac{\rho^m}{m-1} \, dx \quad \rightsquigarrow \quad \partial_t \rho = \Delta(\rho^m).$$

Standard example 4: Nonlinear Diffusion

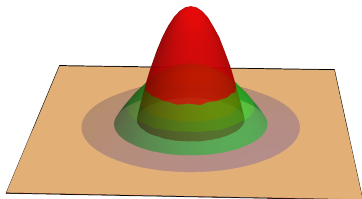
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Fourth order example 1: lubrication

Dirichlet energy

Consider the **Dirichlet energy**

$$E(\rho) = \frac{1}{2} \int |\nabla \rho|^2 dx.$$

Then $\delta E / \delta \rho = -\Delta \rho$, and accordingly,
the $\mathbb{W}$ -gradient flow is the **thin film equation** (a.k.a. Hele-Shaw flow),

$$\partial_t \rho = -\operatorname{div} (\rho \nabla \Delta \rho).$$



Fourth order example 2: quantum diffusion

Fisher information

Consider the [Fisher information](#)

$$F(\rho) = 2 \int |\nabla \sqrt{\rho}|^2 dx \quad .$$

Then $\delta E / \delta \rho = -\Delta \sqrt{\rho} / \sqrt{\rho}$, and accordingly,
the $\mathbb{W}$ -gradient flow is the [DLSS equation](#) (a.k.a. QDD4),

$$\partial_t \rho = -2 \operatorname{div} \left(\rho \nabla \left[\frac{\Delta \sqrt{\rho}}{\sqrt{\rho}} \right] \right) \quad .$$

Fourth order example 2: quantum diffusion

Fisher information

Consider the [Fisher information](#)

$$\mathbf{F}(\rho) = 2 \int |\nabla \sqrt{\rho}|^2 dx = \frac{1}{2} \int |\nabla \log \rho|^2 \rho dx.$$

Then $\delta \mathbf{E} / \delta \rho = -\Delta \sqrt{\rho} / \sqrt{\rho}$, and accordingly,
the $\mathbb{W}$ -gradient flow is the [DLSS equation](#) (a.k.a. QDD4),

$$\partial_t \rho = -2 \operatorname{div} \left(\rho \nabla \left[\frac{\Delta \sqrt{\rho}}{\sqrt{\rho}} \right] \right) = -\nabla^2 : (\rho \nabla^2 \log \rho).$$

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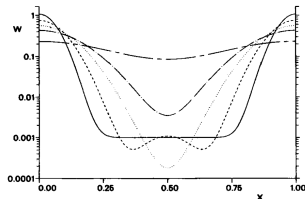
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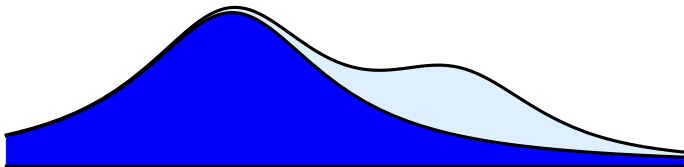
Exotic example 1: Muskat problem

With $a, b > 0$ and $\boldsymbol{\rho} = (\rho_1, \rho_2) \in \mathcal{P}_2(\mathbb{R})^2$, consider

$$\mathbf{E}(\boldsymbol{\rho}) = a \int \rho_1^2 \, dx + b \int (\rho_1 + \rho_2)^2 \, dx.$$

The gradient flow of $\mathbf{E}$ in the $\mathbb{W} \times \mathbb{W}$ -metric is the [Muskat problem](#),

$$\partial_t \boldsymbol{\rho} = \partial_x \left(\begin{pmatrix} (a+b)\rho_1 & b\rho_1 \\ \rho_2 & \rho_2 \end{pmatrix} \partial_x \boldsymbol{\rho} \right).$$



Exotic example 2: parabolic–parabolic Keller–Segel

Consider

$$\mathbf{E}(\rho, c) = \int \left(\rho \log \rho - \rho c + \frac{1}{2} |\nabla c|^2 + \frac{1}{2} c^2 \right) dx.$$

The gradient flow of $\mathbf{E}$ in the $\mathbb{W} \times L^2$ -metric is the [Keller–Segel system](#),

$$\partial_t \rho = \Delta \rho - \operatorname{div}(\rho \nabla c)$$

$$\partial_t c = \Delta c - c + \rho.$$



Got it . . . but why should I care?

There are benefits from
being a Wasserstein gradient flow.

Benefit 1: Variational form

Morally, the gradient flow is such that, at any $t > 0$,

$$\partial_t \rho_t = \operatorname{div}(\rho_t \nabla \varphi_t) \quad \text{where} \quad \varphi_t \in \operatorname{argmin}_{\varphi} \left(\frac{1}{2} \|\varphi\|_{\rho_t}^2 + D_{\rho_t} \mathbf{E}[\varphi] \right).$$

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Among many things, this implies **Hölder continuity**:

Hölder estimate

For $t_2 > t_1 > 0$,

$$\frac{\mathbb{W}(\rho_{t_1}, \rho_{t_2})^2}{t_2 - t_1} \leq \int_{t_1}^{t_2} \|\varphi_t\|_{\rho_t}^2 dt \leq \mathbf{E}(\rho_{t_1}) - \mathbf{E}(\rho_{t_2}).$$

Benefit 2: Dissipation

Morally, the gradient flow balances kinetic energy and **dissipation**,

$$\|\varphi_t\|_{\rho_t}^2 = |\partial \mathbf{E}(\rho_t)|^2 := -\frac{d}{dt} \mathbf{E}(\rho_t) = \int \left| \nabla \frac{\delta \mathbf{E}}{\delta \rho} \right|^2 \rho_t \, dx.$$

This provides **a priori estimates**:

Dissipation estimate

For $t_2 > t_1 > 0$,

$$\int_{t_1}^{t_2} |\partial \mathbf{E}(\rho_t)|^2 \, dt \leq \mathbf{E}(\rho_{t_2}) - \mathbf{E}(\rho_{t_1}).$$

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Example: The Boltzmann entropy $\mathbf{H}(\rho) = \int \rho \log \rho \, dx$ has dissipation

$$|\partial \mathbf{H}(\rho)|^2 = \int |\nabla \log \rho|^2 \rho \, dx = 4 \int |\nabla \sqrt{\rho}|^2 \, dx = 2\mathbf{F}(\rho),$$

i.e., twice the **Fisher information**. Thus, for $\partial_t \rho = \Delta \rho$,

$$2 \int_{t_1}^{t_2} \mathbf{F}(\rho_t) \, dt \leq \mathbf{H}(\rho_{t_1}) - \mathbf{H}(\rho_{t_2}).$$

Benefit 3: Positivity

Consider the thin film equation:

$$\partial_t \rho = -\operatorname{div}(\rho \nabla \Delta \rho)$$

From the PDE alone, it is **not** clear that solutions

- remain **non-negative** and
- **preserve mass**.

But the fact that it is formally the $\mathbb{W}$ -gradient flow of the Dirichlet energy

$$E(\rho) = \frac{1}{2} \int |\nabla \rho|^2 \, dx$$

is a strong indication — and actually provides an angle of attack.

And that's it???

There is one more significant thing:
Convexity!

A (fake) ODE analogy for convexity (1/2)

Consider the euclidean gradient flow $\dot{x} = -\nabla e(x)$ for a λ -uniformly convex energy $e : \mathbb{R}^n \rightarrow \mathbb{R}$, i.e.,

$$\nabla^2 e \geq \lambda \mathbf{1}.$$

A (fake) ODE analogy for convexity (1/2)

Consider the euclidean gradient flow $\dot{x} = -\nabla e(x)$ for a λ -uniformly convex energy $e : \mathbb{R}^n \rightarrow \mathbb{R}$, i.e.,

$$\nabla^2 e \geq \lambda \mathbf{1}.$$

Lemma

If $\lambda > 0$, then:

- e has a unique minimizer $x_\infty \in \mathbb{R}^n$,
- speed, dissipation, energy and distance to x_∞ decay exponentially,

$$\|\dot{x}_t\|^2 \leq e^{-2\lambda t} \|\dot{x}_0\|^2,$$

$$\|\nabla e(x_t)\|^2 \leq e^{-2\lambda t} \|\nabla e(x_0)\|^2,$$

$$e(x_t) - e(x_\infty) \leq e^{-2\lambda t} (e(x_0) - e(x_\infty)),$$

$$\|x_t - x_\infty\|^2 \leq e^{-2\lambda t} \|x_0 - x_\infty\|^2;$$

- the flow is λ -uniformly contractive,

$$\|\hat{x}_t - \check{x}_t\|^2 \leq e^{-2\lambda t} \|\hat{x}_0 - \check{x}_0\|^2.$$

A (fake) ODE analogy for convexity (2/2)

For $e \in C^2(\mathbb{R}^n)$, λ -uniform convexity is equivalent to

$$e(x_s) \leq (1-s)e(x_0) + se(x_1) - \frac{\lambda}{2}s(1-s)\|x_1 - x_0\|^2$$

for any choice of $x_0, x_1 \in \mathbb{R}^n$ and $s \in [0, 1]$,
where $(x_s)_{s \in [0,1]}$ is the straight line from x_0 to x_1 .

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where $(x_s)_{s \in [0,1]}$ is the straight line from x_0 to x_1 .

What about simply replacing

- points $x \in \mathbb{R}^n$ by densities $\rho \in \mathcal{P}_2(\mathbb{R}^d)$,
- euclidean norm $\|\cdot\|$ by Wasserstein metric $\mathbb{W}$, and
- straight line $(x_s)_{s \in [0,1]}$ by $\mathbb{W}$ -geodesic $(\rho_s)_{s \in [0,1]}$?

Definition

E is λ -geodesically convex if for any $\mathbb{W}$ -geodesic $(\rho_s)_{s \in [0,1]}$:

$$E(\rho_s) \leq (1-s)E(\rho_0) + sE(\rho_1) - \frac{\lambda}{2}s(1-s)\mathbb{W}(\rho_0, \rho_1)^2.$$

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Examples:

- the Boltzmann entropy H is 0-geodesically convex;
- if $\nabla^2 V \geq \lambda$, then the potential energy is λ -geodesically convex;
- if $\nabla^2 W \geq \lambda$, then the interaction energy is (almost) λ -geodesically convex;
- if $m > (d-1)/d$, then the internal energy with $h(r) = r^m/(m-1)$ is 0-geodesically convex;
- the Dirichlet energy is **not** λ -geodesically convex for any λ .

Geodesic convexity (2/3)

Back to the gradient flow $\partial_t \rho = \operatorname{div} \left(\rho \nabla \frac{\delta \mathbf{E}}{\delta \rho} \right)$.

Theorem

If $\mathbf{E}$ is λ -geodesically convex with $\lambda > 0$, then

- $\mathbf{E}$ possesses precisely one minimizer ρ_∞ ;
- speed, dissipation, energy excess and distance to ρ_∞ decay exponentially,

$$\begin{aligned}\|\varphi_t\|_{\rho_t}^2 &\leq e^{-2\lambda t} \|\varphi_0\|_{\rho_0}, \\ |\partial \mathbf{E}(\rho_t)|^2 &\leq e^{-2\lambda t} |\partial \mathbf{E}(\rho_0)|^2, \\ \mathbf{E}(\rho_t) - \mathbf{E}(\rho_\infty) &\leq e^{-2\lambda t} (\mathbf{E}(\rho_0) - \mathbf{E}(\rho_\infty)), \\ \mathbb{W}(\rho_t, \rho_\infty)^2 &\leq e^{-2\lambda t} \mathbb{W}(\rho_0, \rho_\infty)^2.\end{aligned}$$

- the flow is λ -uniformly contractive,

$$\mathbb{W}(\hat{\rho}_t, \check{\rho}_t)^2 \leq e^{-2\lambda t} \mathbb{W}(\hat{\rho}_0, \check{\rho}_0)^2.$$

Applications:

- Concerning **decay to equilibrium**, you will see a truly bombastic application in a couple of slides.
- Concerning **contractivity**, one obtains e.g. temporal robustness like: Solutions to $\partial_t \rho = \Delta \rho^m$ with $m > \frac{d-1}{d}$ satisfy

$$\mathbb{W}(\rho_{t+\tau}, \rho_t) \leq \mathbb{W}(\rho_\tau, \rho_0).$$

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This is easily seen for $m = 1$, but not trivial for $m \neq 1$.

Proof: would require a more precise definition of $\mathbb{W}$ -gradient flow.

As a weak remedy, we obtain **contractivity** under a slightly stronger hypothesis.

Displacement convexity (1/2)

Definition

A *segment* $(\rho_s)_{s \in [0,1]}$ w.r.t. η is such that

$$\rho_s = \nabla((1-s)\varphi_0 + s\varphi_1) \# \eta,$$

and φ_0, φ_1 are optimal, i.e.,

$$\mathbb{W}(\rho_i, \eta)^2 = \int |\nabla \varphi_i(x) - x|^2 \eta(x) \, dx \quad \text{for } i = 0, 1.$$

Note: Segments are more general than geodesics.

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Definition

E is **λ -displacement convex** if for any **segment** $(\rho_s)_{s \in [0,1]}$:

$$E(\rho_s) \leq (1-s)E(\rho_0) + sE(\rho_1) - \frac{\lambda}{2}s(1-s)\mathbb{W}(\rho_0, \rho_1)^2.$$

Note: λ -displacement convexity implies λ -geodesically convexity.
All of the aforementioned λ -geodesically convex functionals happen to be λ -displacement convex as well.

Displacement convexity (2/2)

Definition

$(\rho_t)_{t \geq 0}$ is an $(\text{EVI})_\lambda$ -solution if, for all $\eta \in \mathcal{P}_2(\mathbb{R}^d)$,

$$\frac{1}{2} \frac{d}{dt} \mathbb{W}(\rho_t, \eta)^2 + \frac{\lambda}{2} \mathbb{W}(\rho_t, \eta)^2 \leq \mathbf{E}(\eta) - \mathbf{E}(\rho_t).$$

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Theorem

If E is λ -geodesically convex,
then its gradient flow has an $(\text{EVI})_\lambda$ -solution.

Proof: will be given tomorrow, under the stronger hypothesis of λ -displacement convexity.

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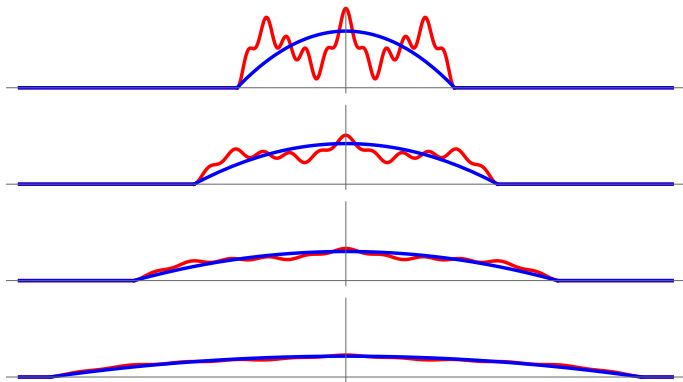
$(\text{EVI})_\lambda$ -solutions are λ -contractive.

... see computations on board ...

Can we go now?

Just one more thing!
(It's about atomic bombs!)

A “real” application (1/2)



A “real” application (2/2)

The porous medium equation

$$\partial_t u = \Delta u^m \quad (\text{PME}_m)$$

has the self-similar solution

$$u_t^*(x) = t^{-d\alpha} U_m(t^{-\alpha} x) \quad \text{with} \quad U_m(z) = \left(B_m - \frac{m-1}{2m} |z|^2 \right)_+^{\frac{1}{m-1}}.$$

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Theorem ([Otto'01])

$$\|u_t - u_t^*\|_{L^1} \leq C t^{-\alpha}.$$

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Proof: Use rescaling to rewrite (PME_m) as

$$\partial_t \rho = \Delta \rho^m + \operatorname{div}(x\rho),$$

which the $\mathbb{W}$ -gradient flow of the geodesically 1-convex functional

$$\mathbf{H}'_m(\rho) = \int \left(\frac{\rho^m}{m-1} + \frac{1}{2} |x|^2 \rho \right) dx.$$

And here is a peaceful spin-off (1/2)

An analogous result holds for the [DLSS equation](#)

$$\partial_t u = -2 \operatorname{div} \left(u \nabla \frac{\Delta \sqrt{u}}{\sqrt{u}} \right), \quad (\text{DLSS})$$

which has the [self-similar solution](#)

$$u_t^*(x) = t^{-d/4} U_1(t^{-1/4}x) \quad \text{with} \quad U_1(z) = (4\pi)^{-d/2} \exp \left(-\frac{1}{2} |z|^2 \right).$$

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Proof: Use [rescaling](#) to rewrite (DLSS) as

$$\partial_t \rho = -2 \operatorname{div} \left(\rho \nabla \frac{\Delta \sqrt{\rho}}{\sqrt{\rho}} \right) + \operatorname{div}(x\rho),$$

which the $\mathbb{W}$ -gradient flow of

$$\mathbf{F}'(\rho) = \mathbf{F}(\rho) + \frac{1}{2} \int |x|^2 \rho \, dx.$$

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And here is a peaceful spin-off (2/2)

Theory guarantees **1-displacement convexity** of the **relative entropy**,

$$\mathbf{H}'(\rho) = \int \rho \log \rho \, dx + \frac{1}{2} \int |x|^2 \rho \, dx.$$

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There is an intimate connection to the **relative Fisher information**:

$$\begin{aligned} |\partial \mathbf{H}'(\rho)|^2 &= \int |\nabla \log \rho + x|^2 \rho \, dx \\ &= \int (|\nabla \log \rho|^2 \rho + |x|^2 \rho + 2 \nabla \rho \cdot x) \, dx \\ &= 2\mathbf{F}(\rho) + \int |x|^2 \rho \, dx - 2d \int \rho \, dx \\ &= 2\mathbf{F}'(\rho) - 2d. \end{aligned}$$

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Theorem

$$\begin{aligned} \mathbf{H}'(\rho_t) - \mathbf{H}'(\rho_\infty) &\leq e^{-2t} (\mathbf{H}'(\rho_0) - \mathbf{H}'(\rho_\infty)), \\ \mathbf{F}'(\rho_t) - \mathbf{F}'(\rho_\infty) &\leq e^{-2t} (\mathbf{F}'(\rho_0) - \mathbf{F}'(\rho_\infty)). \end{aligned}$$

That's all for today

Thank you, and see you tomorrow!