

Ricci curvature of Markov chains via convexity of the entropy

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Joint work with Matthias Erbar (Bonn)

Starting point: Displacement convexity of the entropy

Connection between:

- Boltzmann-Shannon entropy:

$$\text{Ent}(\mu) = \int_{\mathbf{R}^n} \rho(x) \log \rho(x) \, dx; \quad \frac{d\mu}{dx} = \rho$$

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- L^2 -Wasserstein metric:

$$W_2(\mu_0, \mu_1)^2 = \inf \left\{ \int_{\mathbf{R}^n \times \mathbf{R}^n} |x - y|^2 \, d\gamma(x, y) \right. \\ \left. : \gamma \text{ with marginals } \mu_0 \text{ and } \mu_1 \right\}$$

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Theorem (McCann '94)

The Boltzmann-Shannon entropy is **convex** along geodesics in $(\mathcal{P}(\mathbf{R}^n), W_2)$.

Starting Point: Ricci curvature and optimal transport

Theorem (Otto, Villani; Cordero-Erausquin, McCann, Schmuckenschläger; von Renesse, Sturm)

For a Riemannian manifold \mathcal{M} , TFAE:

- 1 Ric $\geq \kappa$ everywhere on \mathcal{M}
- 2 Displacement κ -convexity of the entropy, i.e.,

$$\text{Ent}(\mu_t) \leq (1-t)\text{Ent}(\mu_0) + t\text{Ent}(\mu_1) - \frac{\kappa}{2}t(1-t)W_2^2(\mu_0, \mu_1)$$

for all L^2 -Wasserstein geodesics $(\mu_t)_{t \in [0,1]}$ in $\mathcal{P}(\mathcal{M})$.

Synthetic Ricci curvature of metric measure spaces

Definition (Lott, Villani; Sturm)

A metric measure space (\mathcal{X}, d, m) satisfies $CD(\kappa, \infty)$ if any $\mu_0, \mu_1 \in \mathcal{P}_2(\mathcal{X})$ can be connected by a constant speed W_2 -geodesic $(\mu_t)_{t \in [0,1]}$ such that

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- Many geometric, analytic and probabilistic consequences
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But..... what about discrete spaces?

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- Suppose that $(\mu_{\alpha(t)})$ is a constant speed **geodesic**. Then:

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→ $(\alpha(t))$ is 2-Hölder, hence **constant**.

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LSV-definition does **not** apply to discrete spaces.

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How to make sense of gradient flows in metric spaces?

Let $\varphi : \mathbf{R}^n \rightarrow \mathbf{R}$ smooth and convex. For $u : \mathbf{R}_+ \rightarrow \mathbf{R}^n$ TFAE:

- 1 u solves the **gradient flow equation** $u'(t) = -\nabla\varphi(u(t))$.
- 2 u solves the **evolution variational inequality**

$$\frac{1}{2} \frac{d}{dt} |u(t) - y|^2 \leq \varphi(y) - \varphi(u(t)) \quad \forall y .$$

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The heat flow is the gradient flow of the entropy w.r.t W_2 , i.e.,

$$\partial_t \mu = \Delta \mu \quad \iff \quad \frac{1}{2} \frac{d}{dt} W_2(\mu_t, \nu)^2 \leq \text{Ent}(\nu) - \text{Ent}(\mu_t) \quad \forall \nu$$

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Many **extensions** have been proved:

- \mathbf{R}^n Jordan–Kinderlehrer–Otto
- Riemannian manifolds Villani, Erbar
- Hilbert spaces Ambrosio–Savaré–Zambotti
- Finsler spaces Ohta–Sturm
- Wiener space Fang–Shao–Sturm
- Heisenberg group Juillet
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Question

Is there a version of the JKO-Theorem for *discrete* spaces?

Setting

- \mathcal{X} : finite set
- $K : \mathcal{X} \times \mathcal{X} \rightarrow \mathbf{R}_+$ Markov kernel, i.e., $\forall x : \sum_y K(x, y) = 1$

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Relative Entropy

- For $\rho \in \mathcal{P}(\mathcal{X}) := \left\{ \rho : \mathcal{X} \rightarrow \mathbf{R}_+ \mid \sum_{x \in \mathcal{X}} \rho(x)\pi(x) = 1 \right\}$,

$$\text{Ent}(\rho) = \sum_{x \in \mathcal{X}} \rho(x) \log \frac{\rho(x)}{\pi(x)} .$$

Simplest non-trivial example: 2-point space

$$\mathcal{X} = \{-1, 1\}$$

$$K(-1, 1) = K(1, -1) = 1$$

$$\pi(-1) = \pi(1) = \frac{1}{2}$$



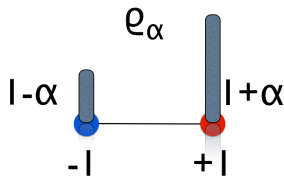
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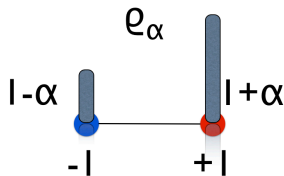
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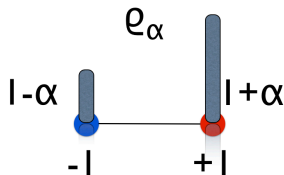
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Answer

NO! Reason:

$$W_2(\rho_\alpha, \rho_\beta) = \sqrt{2|\beta - \alpha|}.$$

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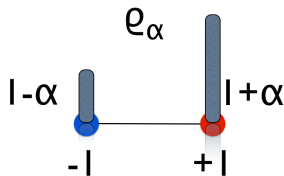
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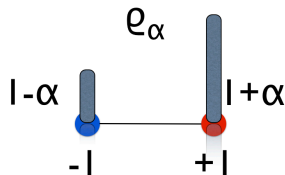
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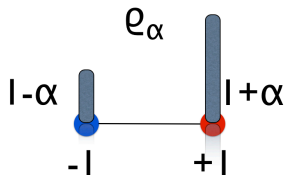
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Answer

YES!

Proposition [M. 2011]

The heat flow is the gradient flow of Ent w.r.t. the metric \mathcal{W} , where

$$\mathcal{W}(\rho_\alpha, \rho_\beta) := \frac{1}{\sqrt{2}} \int_\alpha^\beta \sqrt{\frac{\operatorname{arctanh} r}{r}} dr, \quad -1 \leq \alpha \leq \beta \leq 1.$$

How to define \mathcal{W} in the general discrete case?

In \mathbf{R}^n there is a **dynamical** characterisation of W_2 :

Benamou-Brenier formula in \mathbf{R}^n

$$W_2(\bar{\rho}_0, \bar{\rho}_1)^2 = \inf_{\rho, \Psi} \left\{ \int_0^1 \int_{\mathbf{R}^n} |\Psi_t(x)|^2 \rho_t(x) \, dx \, dt : \right. \\ \left. \partial_t \rho + \nabla \cdot (\rho \Psi) = 0, \quad \rho_0 = \bar{\rho}_0, \quad \rho_1 = \bar{\rho}_1 \right\}.$$

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- **Obstruction**: how to multiply probability densities and discrete gradients?

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$$W_2^2(\bar{\rho}_0, \bar{\rho}_1) = \inf_{\rho, \psi} \left\{ \int_0^1 \int_{\mathbf{R}^n} |\nabla \psi_t(x)|^2 \rho_t(x) dx dt \right\}$$

s.t. $\partial_t \rho + \nabla \cdot (\rho \nabla \psi) = 0 .$

Definition in the discrete case (M. 2011)

$$\mathcal{W}(\bar{\rho}_0, \bar{\rho}_1)^2 := \inf_{\rho, \psi} \left\{ \frac{1}{2} \int_0^1 \sum_{x, y \in \mathcal{X}} (\psi_t(x) - \psi_t(y))^2 \hat{\rho}_t(x, y) K(x, y) \pi(x) dt \right\}$$

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How should we define $\hat{\rho}$? \longrightarrow logarithmic mean

$$\hat{\rho}(x, y) := \int_0^1 \rho(x)^{1-\alpha} \rho(y)^\alpha d\alpha = \frac{\rho(x) - \rho(y)}{\log \rho(x) - \log \rho(y)}$$

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Remark

Related independent work by

- Chow, Huang, Li, and Zhou
- Mielke

Why the logarithmic mean?

Formal proof of the JKO-Theorem

① If (ρ_t, ψ_t) satisfy the cont. eq. $\partial_t \rho + \nabla \cdot (\rho \nabla \psi) = 0$, then

$$\frac{d}{dt} \text{Ent}(\rho_t) = -\langle \log \rho_t, \nabla \cdot (\rho_t \nabla \psi_t) \rangle = \langle \nabla \log \rho_t, \nabla \psi_t \rangle_{\rho_t} .$$

$$\longrightarrow \text{grad}_{W_2} \text{Ent}(\rho) = \nabla \log \rho$$

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provided $\psi = -\log \rho$.

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Logarithmic mean compensates for the lack of a discrete chain rule:

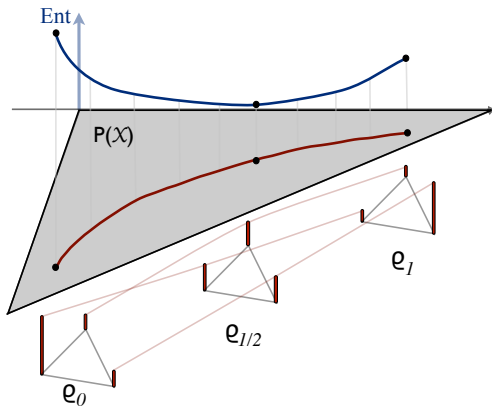
$$\rho(x) - \rho(y) = \hat{\rho}(x, y) (\log \rho(x) - \log \rho(y))$$

Ricci curvature of Markov chains

The discrete analogue of Lott–Sturm–Villani becomes:

Definition (Erbar, M. 2011)

We say that (\mathcal{X}, K, π) has Ricci curvature bounded from below by $\kappa \in \mathbf{R}$ if the entropy is κ -convex along geodesics in $(\mathcal{P}(\mathcal{X}), \mathcal{W})$.



Theorem (Erbar, M.)

Let (\mathcal{X}, K, π) be a reversible Markov chain. Let $\kappa > 0$.

- ① à la Bakry–Émery: $\text{Ric}(K) \geq \kappa \implies$ **modified log-Sobolev**, i.e.

$$\text{Ent}(\rho) \leq \frac{1}{2\kappa} \mathcal{E}(\rho, \log \rho) .$$

This implies $\text{Ent}(H_t \rho) \leq e^{-2\kappa t} \text{Ent}(\rho)$.

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Consequences: Sharp functional inequalities

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- ③ **mod. Talagrand** \implies [**spectral gap** and T_1]:

$$\|\varphi\|_{L^2(\mathcal{X}, \pi)}^2 \leq \frac{1}{\kappa} \mathcal{E}(\varphi, \varphi) \quad \text{and} \quad W_1(\rho, \mathbf{1})^2 \leq \frac{1}{\kappa} \text{Ent}(\rho) .$$

Ricci bounds: examples

- (Mielke 2012) For every finite reversible Markov chain:
 $\exists \kappa \in \mathbf{R}$ such that $\text{Ric}(K) \geq \kappa$.

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Theorem (Erbar, M. 2012)

Let $(\mathcal{X}_i, K_i, \pi_i)$ be reversible finite Markov chains and let (\mathcal{X}, K, π) be the product chain. Then:

$$\text{Ric}(\mathcal{X}_i, K_i, \pi_i) \geq \kappa_i \quad \implies \quad \text{Ric}(\mathcal{X}, K, \pi) \geq \frac{1}{n} \min_i \kappa_i$$

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- **Dimension-independent** bounds
- Sharp bounds for the **discrete hypercube** $\{-1, 1\}^n$

Gromov-Hausdorff convergence

- Let $\mathbf{T}_N^d = (\mathbf{Z}/N\mathbf{Z})^d$ be the discrete torus.
- Let \mathcal{W}_N be the renormalised transportation metric for simple random walk on \mathbf{T}_N^d .

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$(\mathcal{P}(\mathbf{T}_N^d), \mathcal{W}_N) \rightarrow (\mathcal{P}(\mathbf{T}^d), W_2)$ in the sense of Gromov-Hausdorff.

- **Compatibility** between W_2 and \mathcal{W} .
- Main ingredient for proving **convergence of gradient flows**.

Further developments

- Systems of reaction-diffusion equations (Mielke)
- Fractional heat equations (Erbar)
- Dissipative quantum mechanics (Carlen, M.; Mielke)

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Thank you!