

# Minimizing movement scheme for time dependent gradient flows in probability space

Let  $M$  be a smooth closed manifold equipped with a smooth family of Riemannian metrics  $g_t$ .

We suppose that  $g_t$  evolves under the **Ricci flow**

$$\partial_t g = -2\text{Ric}(g) \quad t \in [0, T],$$

where  $\text{Ric}(g)$  denotes the Ricci tensor of  $g$ .

We consider the **forward heat equation** acting on functions

$$\partial_t u = \Delta u,$$

where  $\Delta = \Delta_{g(t)}$  denotes the Laplace-Beltrami operator.

By duality we obtain the **backward heat equation** on measures

$$\partial_t \nu = -\Delta \nu. \quad (1)$$

## Aim

We are interested in **reinterpreting the diffusion equation (1) as a dynamic gradient flow** for the **relative entropy** with respect to the **Kantorovich distance**.

## Method

At first we need to understand the meaning of gradient flows in a metric setting, where the metric is changing.

To establish existence of the trajectory, we apply the **time discrete approximation** of the gradient flow.

Let  $\mathcal{P}(M)$  be the space of **Borel probability measures** on  $M$ . We consider the following two objects on this space:

- the **Kantorovich distance** at time  $t$  between two measures  $\mu, \nu \in \mathcal{P}(M)$  is defined by

$$W_t(\mu, \nu) := \inf \left\{ \int_{M \times M} d_t^2(x, y) d\pi(x, y) \right\}^{1/2},$$

where the infimum is taken over all  $\pi \in \mathcal{P}(M \times M)$  such that  $\pi(A \times M) = \mu(A)$ ,  $\pi(M \times A) = \nu(A)$ ;

- the **relative entropy**  $H_t$  with respect to the volume measure  $m_t$ , which is defined by

$$H_t(\mu) := \int_M \rho \log \rho dm_t,$$

provided that  $\mu \in \mathcal{P}(M)$  is absolutely continuous with density  $\rho$ . Otherwise  $H_t(\mu) = \infty$ .

## Motivation (dynamic gradient flow)

Let  $V: [0, T] \times M \rightarrow \mathbb{R}$  be a smooth function,  $x: [0, T] \rightarrow M$  a smooth curve.  $G(a, b)$  denotes the Green function on  $[0, 1]$ .

$V$  is **dynamically convex**  $:\Leftrightarrow \text{Hess}_t V_t \geq \frac{1}{2} \partial_t g_t$

$$\Leftrightarrow V_t(\gamma_a) \leq (1-a)V_t(\gamma_0) + aV_t(\gamma_1) - \frac{1}{2} \int_0^1 G(a, b) \partial_t g_t(\dot{\gamma}_b, \dot{\gamma}_b) db$$

$$\forall d_t\text{-geodesics } (\gamma_a).$$

Then the following are equivalent

$$\dot{x}_t = -\nabla_t V_t(x_t)$$

$$\Leftrightarrow \partial_s \frac{1}{2} d_t^2(x_s, z)|_{s=t} \leq V_t(z) - V_t(x_t) - \frac{1}{2} \int_0^1 (1-b) \partial_t g_t(\dot{\gamma}_b, \dot{\gamma}_b) db \quad \forall d_t\text{-geodesics } \gamma \text{ from } x_t \text{ to } z.$$

For each  $x_0$  there exists **at most one** dynamic gradient flow for  $V$  starting in  $x_0$ .

## Definition

A functional  $V$  on  $\mathcal{P}(M)$  is **dynamically convex** if for every  $W_t$  geodesic  $(\gamma_a)$

$$V_t(\gamma_a) \leq (1-a)V_t(\gamma_0) + aV_t(\gamma_1) - \frac{1}{2} \int_0^1 G(a, b) \partial_t (|\dot{\gamma}_b|_t^2) db,$$

where  $|\dot{\gamma}_b|_t := \lim_{\epsilon \rightarrow 0} \frac{W_t(\gamma_{b+\epsilon}, \gamma_b)}{\epsilon}$  is the metric speed with respect to  $W_t$ .

A **dynamic gradient flow** for  $V$  is a curve  $(\mu_t)$  in  $\mathcal{P}(M)$  satisfying the **dynamic Evolution Variation Inequality (EVI)**, i.e.

$$\partial_s \frac{1}{2} W_t^2(\mu_s, \sigma)|_{s=t} \leq V_t(\sigma) - V_t(\mu_t) - \frac{1}{2} \int_0^1 (1-b) \partial_t (|\dot{\gamma}_b|_t^2) db \quad \forall \sigma,$$

where  $(\gamma_b)$  is a  $W_t$ -geodesic from  $\mu_t$  to  $\sigma$ .

## Backward dynamic convexity and backward EVI

The Ricci flow ensures the **backward dynamic convexity of the relative entropy**  $H$ , i.e.  $\tilde{H}$  is dynamically convex with respect to  $\tilde{W}$ , where  $\tilde{H}_t = H_{T-t}$ .

Since we are dealing with the backward heat equation, we have to deal with the EVI backwards in time, too, i.e.  $(\mu_t)$  satisfies the **backward EVI** for  $H$  with respect to  $W$  if  $(\tilde{\mu}_t)$  satisfies the EVI for  $\tilde{H}$  with respect to  $\tilde{W}$ , where  $\tilde{\mu}_t = \mu_{T-t}$ .

## Motivation (time discrete approximation)

Let  $F: \mathbb{R}^d \rightarrow \mathbb{R}$  be a convex differentiable functional. Consider the gradient flow  $(x_t)$  starting from  $\bar{x} \in \mathbb{R}^d$

$$\begin{cases} \dot{x}_t = -\nabla F(x_t), \\ x_0 = \bar{x}. \end{cases}$$



Fix a step size  $h > 0$ , set  $x_0 = \bar{x}$  and recursively determine  $x_n$  that minimizes

$$x \mapsto F(x) + \frac{|x - x_{n-1}|^2}{2h}. \quad (2)$$

The Euler-Lagrange equation of  $x_n$  is

$$\frac{x_n - x_{n-1}}{h} = -\nabla F(x_n),$$

which is a time discretization of  $\dot{x}_t = -\nabla F(x_t)$ .

Define the interpolation

$$x_t^h := x_n^h \text{ if } t \in ((n-1)h, nh].$$

Then as  $h \rightarrow 0$

$$x_t^h \rightarrow x_t \text{ and } x_t \text{ is a gradient flow starting in } \bar{x}.$$

Fix a time step  $h > 0$  and an initial value  $\bar{\mu}$ .

Recursively define the **sequence of minimizers**  $(\mu_n^h)$  according to the following discrete scheme

$$\mu_0^h := \bar{\mu}, \quad \mu_n^h := \arg \min_{\nu \in \mathcal{P}(M)} \left\{ H_{nh}(\nu) + \frac{1}{2h} W_{nh}^2(\mu_{n-1}^h, \nu) \right\}.$$

**Existence** of minimizers is provided by compactness and lower semicontinuity.

**Uniqueness** follows by usual strict convexity.

## Euler-Lagrange equation for the minimizers

The minimizers  $\mu_n^h$  satisfy the **discrete gradient flow equation**

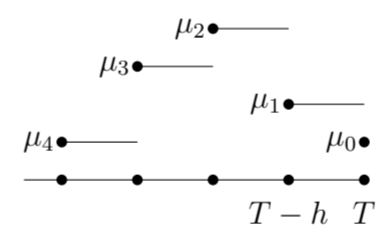
$$-\int_M \text{div}_{nh} \xi d\mu_n^h = \int_M \left\langle \frac{1}{h} \Psi_{\mu_n^h}^{\mu_{n-1}^h}, \xi \right\rangle_{nh} d\mu_n^h \quad \xi \in C_c^\infty(M, TM),$$

i.e. for  $\mu_n^h = \rho_n^h m_{nh}$  we have  $\nabla_{nh} \rho_n^h / \rho_n^h = \frac{1}{h} \Psi_{\mu_n^h}^{\mu_{n-1}^h}$ , where  $\Psi_{\mu_n^h}^{\mu_{n-1}^h}$  denotes the optimal transport vector field between  $\mu_n^h$  and  $\mu_{n-1}^h$ .

## Limit trajectory

Define piecewise constant interpolations  $\bar{\mu}_t^h$  backwards in time:

$$\bar{\mu}_T^h := \bar{\mu},$$

$$\bar{\mu}_t^h := \mu_n^h, \text{ for } t \in [T - nh, T - (n-1)h].$$


With the help of Arzela-Ascoli one obtains a **limit curve**  $(\mu_t)$  such that up to a subsequence

$$\bar{\mu}_t^{h_n} \rightarrow \mu_t \text{ weakly for all } t \text{ as } h_n \rightarrow 0.$$

## Questions

- Is the limit curve  $(\mu_t)$  the **solution to the backward heat equation (1)**?
- Does the limit curve  $(\mu_t)$  satisfy the **backward EVI** for  $H$  with respect to  $W$ ?

## Possible Generalizations

- $(M, g_t)$  is not compact but complete and evolves under a **super-Ricci flow**

$$\partial_t g \geq -2\text{Ric}(g).$$

- Consider instead of a manifold a family of metric measure space  $(X, d_t, m_t)_t$  evolving under a super Ricci-flow, which means that the relative entropy  $H$  is backward dynamically convex.

## References

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