

The Stein-log-Sobolev inequality and the exponential convergence of the Stein Variational Gradient Descent algorithm

From non-local to local



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The mobility equation

The following PDE denotes a general mobility equation

$$\partial_t \rho = \operatorname{div}(m(\rho) \mathcal{V}(\rho))$$

where

- $m : \mathbb{R} \rightarrow \mathbb{R}$ is some continuous mobility function,
- $\mathcal{V}(\rho)$ is some velocity that can have the form

$$\mathcal{V}(\rho_t) = \nabla U'(\rho_t) + \nabla V + \nabla(W * \rho_t).$$

Examples [Carrillo et al., 2024b]

- Lubrication model for thin films [Bertozzi, 1998]

$$m(\rho) = \rho^3, \quad \mathcal{V}(\rho) = -\sqrt{\text{Ca}} \nabla \rho + \nabla \Delta \rho.$$

- Cahn-Hilliard equation for phase separation in binary alloys [Cahn, 1998]

$$m(\rho) = 1 - \rho^2, \quad \mathcal{V}(\rho) = -\nabla \frac{\delta E}{\delta \rho}, \quad E(\rho) := \int_{\Omega} H(\rho(x)) + \frac{\epsilon^2}{2} |\nabla \rho(x)|^2 \, dx,$$

where $\rho : [0, \infty) \times \Omega \rightarrow [-1, 1]$.

- Chemotaxis with prevention of overcrowding [Burger et al., 2006]

$$m(\rho) = \rho(1 - \rho).$$

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Stein Variational Gradient Descent [Liu and Wang, 2016]

We want to approximate $\rho_\infty \propto e^{-V}$ for a potential $V : \mathbb{R}^d \rightarrow \mathbb{R}$.

SVGD is governed by a positive definite interaction kernel $K : \mathbb{R}^d \times \mathbb{R}^d \rightarrow \mathbb{R}$.

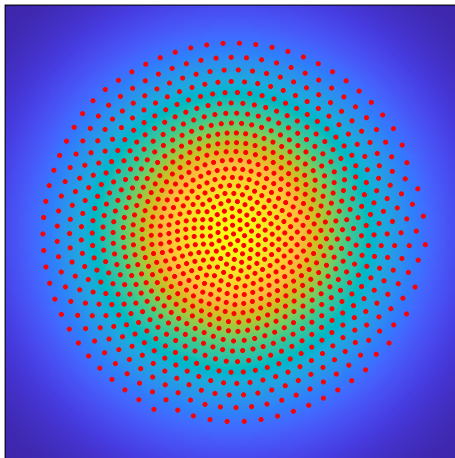
Given is some initial distribution of particles $X_0^1, \dots, X_0^N \in \mathbb{R}^d$.

SVGD has deterministic particle dynamics

$$X_{n+1}^i := X_n^i + \varepsilon_n \sum_{j=1}^N (\nabla_y K)(X_n^i, X_n^j) - K(X_n^i, X_n^j) \nabla V(X_n^j).$$

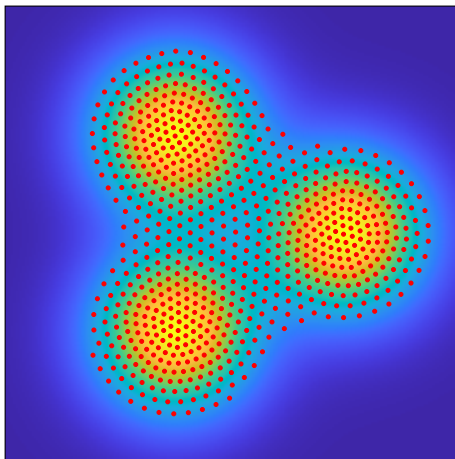
Computational Examples

Normal Gaussian



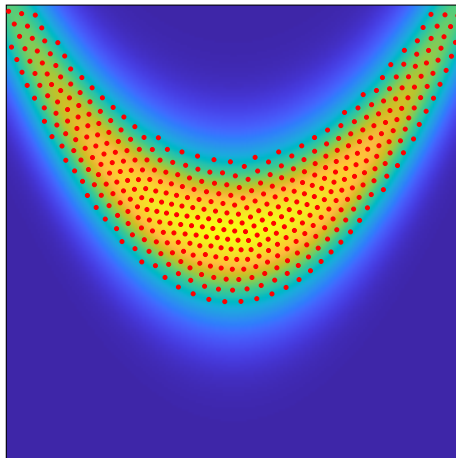
Computational Examples

Mixture of Gaussians



Computational Examples

Banana Distribution



Mean-field limit

Formal computations in [Liu, 2017] show that if $N \rightarrow \infty$, $\varepsilon \rightarrow 0$ and the kernel is translation invariant $K(x, y) := k(x - y)$, then the following evolution of distribution of particles holds true

$$\partial_t \rho = \operatorname{div}(\rho k * (\nabla \rho + \rho \nabla V)) = \operatorname{div}(\rho k * \nabla(\rho e^V) e^{-V}).$$

SVGD as a gradient flow

In fact we can interpret the previous equation as a gradient flow

$$\frac{d}{dt} \text{KL}(\rho_t \parallel \rho_\infty) = -\mathbb{D}^2(\rho_t \parallel \rho_\infty),$$

where we define the dissipation by

$$\mathbb{D}^2(\rho_t \parallel \rho_\infty) := \int_{\mathbb{R}^d} \nabla(\rho e^V) e^{-V} \cdot k * \nabla(\rho e^V) e^{-V} dx,$$

and the Kullback-Leibler divergence (or relative entropy)

$$\text{KL}(\rho_t \parallel \rho_\infty) := \int_{\mathbb{R}^d} \rho \ln(\rho) dx + \int_{\mathbb{R}^d} \rho V dx.$$

Stein-log-Sobolev inequality

We achieve an exponential rate of decay in the Kullback-leibler divergence if the **Stein-log-Sobolev inequality** holds for some fixed $\lambda > 0$

$$\lambda \text{KL}(\rho_t \parallel \rho_\infty) \leq \mathbb{D}^2(\rho_t \parallel \rho_\infty).$$

In this case we get

$$\text{KL}(\rho_t \parallel \rho_\infty) \leq e^{-\lambda t} \text{KL}(\rho_0 \parallel \rho_\infty).$$

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From non-local to local

Take $k \in L^1(\mathbb{R}^d)$ with $k \geq 0$, $\int_{\mathbb{R}^d} k \, dx = 1$ and k is positive definite. Define the approximation of the identity $k_\sigma := \sigma^{-d} k(x/\sigma)$ for $\sigma \in (0, 1)$. Let ρ^σ denote the solution of

$$\partial_t \rho^\sigma = \operatorname{div} (\rho^\sigma k_\sigma * (\nabla \rho^\sigma + \rho^\sigma \nabla V)).$$

We observe that $k_\sigma \rightharpoonup \delta_0$ as $\sigma \rightarrow 0$ in the sense of distributions. Then, if $\rho^\sigma \rightharpoonup \rho$ as $\sigma \rightarrow 0$, then one formally derives a weak solution to the following mobility equation

$$\partial_t \rho = \operatorname{div} (\rho (\nabla \rho + \rho \nabla V)) = \operatorname{div} (\rho^2 \nabla (\ln(\rho) + V)).$$

Exponential convergence rate for the continuity equation

Assume that the Stein-log-Sobolev inequality holds uniformly for all solutions $\{\rho^\sigma\}_\sigma$ with some fixed constant $\lambda > 0$. Then, we derive the uniform bound

$$\text{KL}(\rho_t^\sigma \parallel \rho_\infty) \leq e^{-\lambda t} \text{KL}(\rho_0 \parallel \rho_\infty).$$

By taking the limit $\sigma \rightarrow 0$ and using the lower semicontinuity of the Kullback-Leibler divergence, we see that the continuity equation would possess an exponential convergence rate

$$\text{KL}(\rho_t \parallel \rho_\infty) \leq e^{-\lambda t} \text{KL}(\rho_0 \parallel \rho_\infty).$$

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Main Theorem [Carrillo et al., 2024a]

We were inspired by [Duncan et al., 2023] to use the following ansatz form

$$K(x, y) = e^{V(x) - \frac{V_0(x)}{2}} k(x - y) e^{V(y) - \frac{V_0(y)}{2}}.$$

A sufficient criteria for the Stein-log-Sobolev inequality is

- $V \geq V_0$ for $V_0(x) = \alpha|x|^2 + \beta$,
- $k \in L^1(\mathbb{R}^d) + L^2(\mathbb{R}^d)$,
- There exists two constants $C_0, C_1 \geq 1$ such that

$$\frac{1}{C_0} \frac{1}{1 + |\xi|^2} \leq \hat{k}(\xi) \leq C_1 \frac{1}{1 + |\xi|^2}.$$

Existence of weak solution [Carrillo et al., 2024a]

If additionally $V \in C^1(\mathbb{R}^d) \cap H_{\text{loc}}^m(\mathbb{R}^d)$ for some $m > \frac{d}{2}$, then there exists a solution $\rho \in C^0([0, \infty), \mathcal{P}(\mathbb{R}^d))$ in the distributional sense to the non-local continuity equation

$$\partial_t \rho_t = \operatorname{div} \left(\rho_t e^{V - \frac{V_0}{2}} k * \left[(\nabla \rho_t + \rho_t \nabla V) e^{V - \frac{V_0}{2}} \right] \right). \quad (*)$$

Moreover, we have exponential decay in Kullback-Leibler divergence and the energy dissipation inequality

$$\operatorname{KL}(\rho_t \parallel \rho_\infty) \leq e^{-\lambda t} \operatorname{KL}(\rho_0 \parallel \rho_\infty),$$

$$\operatorname{KL}(\rho_t \parallel \rho_\infty) + \int_0^t \mathbb{D}^2(\rho_s \parallel \rho_\infty) \, ds \leq \operatorname{KL}(\rho_0 \parallel \rho_\infty).$$

Stein-log-Sobolev constant [Carrillo et al., 2024a]

The Stein-log-Sobolev constant has the form

$$\lambda = \lambda_0 (\alpha \wedge 1) e^\beta \frac{1}{C_0^2} \frac{1}{\|e^{-V}\|_{L^1}},$$

where $\lambda_0 > 0$ is some constant independent of dimension d , and we recall that α and β determine \mathbb{V}_0 , and C_0 is used for the lower bound of \hat{k} .

If we only vary k , then we see that the Stein-log-Sobolev constant only depends on C_0 .

Application of result

Let ρ^σ be the weak solution to $(*)$ with kernel $k_\sigma(x) := \sigma^{-d} k(x/\sigma)$.

We observe a uniform lower bound

$$\hat{k}_\sigma(x) = \hat{k}(\sigma x) \geq \frac{1}{C_0} \frac{1}{1 + |\sigma \xi|^2} \geq \frac{1}{C_0} \frac{1}{1 + |\xi|^2}.$$

Thus we see that the Stein-log-Sobolev inequality holds uniformly for all $\{k_\sigma\}_\sigma$.

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Passing to the limit

Let $\{\rho^\sigma\}_\sigma$ be the weak solutions to $(*)$.

If we assume $k = \omega * \omega$ for some measurable function $\omega : \mathbb{R}^d \rightarrow \mathbb{R}$ that satisfies certain constraints, and $\omega_\sigma(x) = \sigma^{-d}\omega(x/\sigma)$, then we need to pass to the limit in the following PDE

$$\partial_t \rho_t^\sigma = \operatorname{div} \left((\rho_t^\sigma e^{V - \frac{V_0}{2}}) \omega_\sigma * \left[(\nabla \rho_t^\sigma + \rho_t^\sigma \nabla V) e^{V - \frac{V_0}{2}} \right] * \omega_\sigma \right).$$

Passing to the limit

$$\partial_t \rho_t^\sigma = \operatorname{div} \left((\rho_t^\sigma e^{V - \frac{V_0}{2}}) \omega_\sigma * \left[(\nabla \rho_t^\sigma + \rho_t^\sigma \nabla V) e^{V - \frac{V_0}{2}} \right] * \omega_\sigma \right)$$

We restrict ourselves to the domain $[0, T] \times B_R$. A commutator argument shows that the above limit coincides with the one below

$$\partial_t \rho_t^\sigma = \operatorname{div} \left((\rho_t^\sigma e^{V - \frac{V_0}{2}} \chi) \omega_\sigma * \left[(\nabla \rho_t^\sigma + \rho_t^\sigma \nabla V) e^{V - \frac{V_0}{2}} \right] * \omega_\sigma \right)$$

where $\chi \in C_c^\infty(\mathbb{R}^d)$ satisfies $0 \leq \chi \leq 1$, $\chi = 1$ on B_R and $\chi = 0$ outside B_{2R} .

Passing to the limit

$$\partial_t \rho_t^\sigma = \operatorname{div} \left((\rho_t^\sigma e^{V - \frac{V_0}{2}} \chi) * \omega_\sigma \left[(\nabla \rho_t^\sigma + \rho_t^\sigma \nabla V) e^{V - \frac{V_0}{2}} \right] * \omega_\sigma \right)$$

The energy dissipation gives us weak compactness in $L^2([0, T] \times \mathbb{R}^d)$.

Passing to the limit

$$\partial_t \rho_t^\sigma = \operatorname{div} \left((\rho_t^\sigma e^{V - \frac{V_0}{2}} \chi) * \omega_\sigma \left[(\nabla \rho_t^\sigma + \rho_t^\sigma \nabla V) e^{V - \frac{V_0}{2}} \right] * \omega_\sigma \right)$$

Using the Aubin-Lions lemma we show compactness in $L^2([0, T] \times \mathbb{R}^d)$.

Passing to the limit

$$\partial_t \rho_t^\sigma = \operatorname{div} \left((\rho_t^\sigma e^{V - \frac{V_0}{2}} \chi) * \omega_\sigma \left[(\nabla \rho_t^\sigma + \rho_t^\sigma \nabla V) e^{V - \frac{V_0}{2}} \right] * \omega_\sigma \right)$$

We show weak compactness in $L^{2\frac{q+2}{q+1}}(0, T; H^{-m}(B_R))$ to derive uniform convergence in $C^0([0, T]; \mathcal{P}(\mathbb{R}^d))$ for $q := d \vee 3$ and some fixed $m > \frac{d}{2}$.

Combining everything allows us to pass to the limit.

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We've shown the existence of a solution $\rho \in C^0([0, \infty); \mathcal{P}(\mathbb{R}^d))$ in the sense of distributions to the following mobility equation

$$\partial_t \rho = \operatorname{div}(\rho^2 \nabla(\ln(\rho) + V)).$$

Moreover, we have shown that this equation possess an exponential decay in the Kullback-Leibler divergence (or relative entropy), for some $\lambda > 0$ depending only on V ,

$$\operatorname{KL}(\rho_t \parallel \rho_\infty) \leq e^{-\lambda t} \operatorname{KL}(\rho_0 \parallel \rho_\infty).$$

Thank you for your attention

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