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Iterated Schrödinger Bridge Approximation to Wasserstein Gradient Flows

Medha Agarwal, Zaid Harchaoui, Garrett Mulcahy, Soumik Pal

Background



$$x_i \leftarrow \sum_{j=1}^n K_{ij}^{\infty} W_V x_j$$
, where $K^{\infty} = \text{Sinkhorn}(C)$ and $C_{ij} = \left(W_Q x_i\right)^{\top} \left(W_K x_j\right)$
Query Matrix

Key Matrix

Under $(W_{O})^{\top}W_{K} = -W_{V} = I_{d}$, the infinite particles counterpart with bandwidth/temperature ε is

$$\rho \leftarrow (T_{\rho,\varepsilon})_{\#} \rho$$
 where $T_{\rho,\varepsilon}(x) = 2x - \int k_{\varepsilon}^{\infty}(x,y) \ y \ d\rho(y)$ and $k_{\varepsilon} = \text{Sinkhorn}(c/\varepsilon)$

Iteratively, with $\rho_0^{\varepsilon}=\rho_0$,

Sequence $\left(\rho_k^{\varepsilon}, k \in [T/\epsilon]\right)$ Piece-wise continuous curve $\left(\rho_t^{\varepsilon}, t \in [0,T]\right)$ $\rho_k^{\varepsilon} = \left(T_{\rho_{k-1}^{\varepsilon},\varepsilon} \right)_{\#} \rho_{k-1}^{\varepsilon}$ $\rho_t^{\varepsilon} = \rho_{|t/\varepsilon|}^{\varepsilon}, t \in [0,T]$

As $\varepsilon \to 0+$, what is the limit of the curve $(\rho_t^{\varepsilon}, t \in [0,T])$? Answer: Heat equation!

In fact, if $(\rho_t, t \in [0,T])$ is heat equation starting from ρ_0 ,

$$\lim_{\varepsilon \to 0} \sup_{k \in [T/\varepsilon]} \mathbb{W}_2^2(\rho_k^\varepsilon, \rho_{k\varepsilon}) = 0$$

Setting

Consider the entropy regularized optimal transport (EOT) problem between measures μ and ν

$$\min_{\pi \in \Pi(\mu,\nu)} \left\{ \|x - y\|^2 d\pi + \varepsilon \mathsf{KL}(\pi \mid \mid \mu \otimes \nu) \right\}$$

• argmin
$$\pi_{\mu,\nu,\varepsilon}$$
 is the Schrödinger bridge

Consider same marginal setting, i.e.
$$\mu = \nu = \rho$$
 with Schrödinger bridge $\pi_{\alpha,\varepsilon}$, then

$$\mathscr{B}_{\rho,\varepsilon}(x) = \mathbb{E}_{\pi_{\rho,\varepsilon}}[Y | X = x] = \int y \, k^{\infty}(x, y) \, d\rho(y) \, .$$

Therefore, for infinite particles, the self-attention update is

$$\rho_{k}^{\varepsilon} = \left(T_{\rho_{k-1}^{\varepsilon},\varepsilon}\right)_{\#} \rho_{k-1}^{\varepsilon} = \left(2I_{d} - \mathscr{B}_{\rho,\varepsilon}\right)_{\#} \rho_{k-1}^{\varepsilon}$$

Main contribution: Under suitable conditions, $2x - \mathscr{B}_{\rho,\varepsilon} \approx x - \frac{1}{2} \nabla \log \rho(x)$

Claim:
$$(\rho_k^{\varepsilon}, k \in \lfloor T/\varepsilon \rfloor)$$
 is a discrete approximate to the heat flow.

Wasserstein gradient flow is characterized by the continuity equation $\partial_t \rho_t + \nabla \cdot (v_t \rho_t) = 0$ where $v_t: \mathbb{R}^d \to \mathbb{R}^d$ is the velocity field. For heat flow, $v_t = -0.5 \nabla \log \rho_t$.

What about any gradient flow? Can we such approximations for them like self-attention for heat flow? Key Contribution: (1) Under suitable conditions, self-attention iterations converge to the heat flow as $\varepsilon \to 0$. (2) We write the iterated scheme for a general functional \mathcal{F} . (3) We prove the uniform convergence of the scheme to the gradient flow of KL divergence with respect to a log-concave density.

Iterative Scheme

Consider absolutely continuous curve $(\rho_t, t \in [0,T])$ satisfying $\partial_t \rho_t + \nabla \cdot (v_t, \rho_t) = 0$ where $v_t = v(\rho_t)$ is the velocity field. An example is gradient flow minimizing a functional $\mathcal{F}: \mathcal{P}(\mathbb{R}^d) \to \mathbb{R}$.

One-step Approximation

One discrete time approximation is analogous to explicit Euler scheme

$$S_{\varepsilon}^{1}(\rho) =$$

= $(id + \varepsilon v)_{\mu} \rho$ with $v = v(\rho)$. Explicit Euler (EE) Step Let $v = \nabla \psi$ for a smooth function ψ and there exists $\theta_{\epsilon} \in \mathbb{R} \setminus \{0\}$ s.t. $\Lambda(\epsilon) = [$ $\exp(2\theta_{e}\psi) < +\infty$.

Define the surrogate measure $\sigma_{\epsilon} := \Lambda(\epsilon)^{-1} \exp(2 \theta_{\epsilon} \psi)$ and the one-step update

$$SB^1_{\epsilon}(\rho) :=$$

Iterative Schemes

Explicit Euler Scheme

$$\left(\rho, S_{\varepsilon}^{1}(\rho), ..., S_{\varepsilon}^{N_{\varepsilon}}(\rho)\right), N_{\varepsilon} = [T/\varepsilon].$$

Remark: For heat equation (gradient flow of entropy functional), $\sigma_{e} = \rho$ and $\theta_{e} = -1$. For Fokker Planck equation (gradient flow of KL divergence with respect to ν), $\sigma_{\varepsilon} = (\rho/\nu)^{-\theta_{\varepsilon}}$ where the sign of θ_{ε} depends on the integrability of σ_{ε} .

Convergence

Theorem 1 (Tight Approximation of Same Marginal Schrödinger Bridge)

Let
$$\rho = e^{-g} \in \mathscr{P}(\mathbb{R}^d)$$
 with enough *regularity* such that there is strong solution to the Langevin SDE $dX_t = -\frac{1}{2}\nabla g(X_t) + dB_t$ with $X_0 \sim \rho$. Let $\ell_{\rho,\varepsilon} = \text{Law}(X_1, X_{\varepsilon})$, then $H(\ell_{\rho,\varepsilon} | \pi_{\rho,\varepsilon}) + H(\pi_{\rho,\varepsilon} | \ell_{\rho,\varepsilon}) \leq \varepsilon^2 C$

One-step Convergence

[heorem 2 (Single Step Convergence) If the surrogate measure σ_c satisfies a set of regularity conditions, then there exists a constant K > 0 such that $\mathbb{W}_2(S^1_{\varepsilon}(\rho), SB^1_{\varepsilon}(\rho)) < K\varepsilon C(\varepsilon)$. Under assumptions on surrogate measure σ_{ε} , $C(\varepsilon) = o(1)$.



Remark: (1) The one step convergence relies on the close approximation of the same-marginal Schrödinger bridge by the Langevin diffusion.



 $= ((1 - \theta_{\epsilon}^{-1})id + \theta_{\epsilon}^{-1}\mathscr{B}_{\sigma_{\alpha},\epsilon})_{\#}\rho$

Schrödinger Bridge (SB) Step

Schrödinger Bridge Scheme $\left(\rho, SB^{1}_{\varepsilon}(\rho), \dots, SB^{N_{\varepsilon}}_{\varepsilon}(\rho)\right), N_{\varepsilon} = [T/\varepsilon].$

For Gaussian marginals, $C(\varepsilon) = \mathcal{O}(\varepsilon^2)$

 $\lim_{\varepsilon} \varepsilon^{-1} \mathbb{W}_2(S^1_{\varepsilon}(\rho), SB^1_{\varepsilon}(\rho)) = 0$

Uniform Convergence

 $\mathbb{W}_{2}(\rho_{k\varepsilon}, SB_{\varepsilon}^{k}(\rho)) \leq \mathbb{W}_{2}(\rho_{k\varepsilon}, SB_{\varepsilon}^{1}(\rho_{(k-1)\varepsilon})) + \mathbb{W}_{2}(SB_{\varepsilon}^{1}(\rho_{(k-1)\varepsilon}), SB_{\varepsilon}^{k}(\rho))$



Experiments



