

Covariance-modulated optimal transport and gradient flows

Part II - Gradient flows

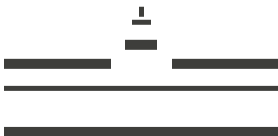
BIRS: Stochastic Mass Transport

André Schlichting

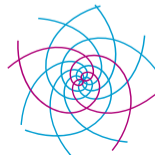
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Applied Mathematics Münster: Institute for Analysis and Numerics

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- Motivation
 - Bayesian inverse problems
 - sampling via Covariance-modulated SDE
 - mean-field limit
- Gradient flows induced by Covariance-modulated transport
- Geodesic convexity
- EVI formulations and functional inequality
- Entropy method

Inverse problems for parameter estimation

Parameter estimation: Given data $y \in \mathbb{R}^K$ and noise $\xi \in \mathbb{R}^d$

Find parameter x : $y = G(x) + \xi$ for given model $G : \mathbb{R}^d \rightarrow \mathbb{R}^K$.

Posterior density: For Gaussian noise $\xi \sim \mathcal{N}(0, \Gamma)$ and $x \sim \mathcal{N}(0, \Sigma)$

$$\pi(dx) \propto \exp(-f(x)) \quad \text{with} \quad f(x) = \frac{1}{2}|y - G(x)|_{\Gamma}^2 + \frac{1}{2}|x|_{\Sigma}^2$$

$G(x) = Ax$ \rightarrow
 f Gaussian

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Bayesian Inverse problem tasks

(1) Inversion: Find $x^* := \arg \max \pi(x)$ or **(2) Sampling** from π .

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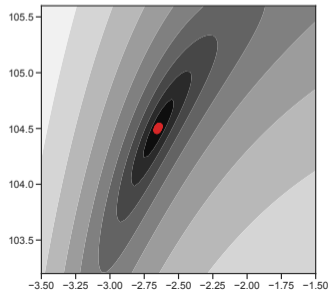
(1) **Ensemble Kalman Inversion (EKI):** Particle algorithm finding x^*

$$\dot{x}^{(j)} = -\frac{1}{J} \sum_{k=1}^J \left\langle G(x^{(k)}) - \bar{G}, G(x^{(j)}) - y \right\rangle_{\Gamma} x^{(k)} \quad \text{with} \quad \bar{G} := \frac{1}{J} \sum_{k=1}^J G(x^{(k)}).$$

Ensemble-Kalman-Sampling (EKS)

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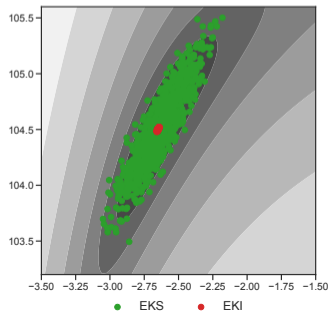
(1) **Inversion:** Find $x^* := \arg \max \pi(x)$ or (2) **Sampling** from π .

(2) **Ensemble Kalman Sampling (EKS):** SDE sampling $\pi \propto e^{-f}$ with J particles $\{X^{(j)}\}_{j=1}^J$

$$\dot{X}^{(j)} = -\frac{1}{J} \sum_{k=1}^J \left\langle G(X^{(k)}) - \bar{G}, G(X^{(j)}) - y \right\rangle_{\Gamma} X^{(k)} - \mathbf{C}(\rho^J) \Sigma^{-1} X^{(j)} + \sqrt{2 \mathbf{C}(\rho^J)} \dot{W}^{(j)},$$

with the **covariance** of the empirical measure $\rho^J = J^{-1} \sum_j \delta_{X^{(j)}}$.

$$\mathbf{C}(\rho^J) = \frac{1}{J} \sum_{k=1}^J \left(X^{(k)} - \bar{X} \right) \otimes \left(X^{(k)} - \bar{X} \right) \quad \text{and} \quad \bar{X} = \frac{1}{J} \sum_{k=1}^J X^{(k)}.$$



Ensemble-Kalman-Sampling (EKS)

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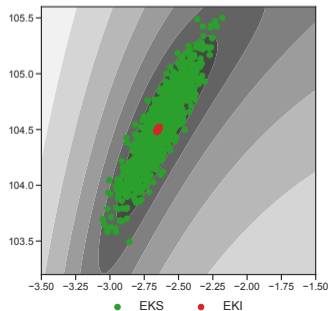
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Mean-field limit of EKS: $J \rightarrow \infty$ yields

$$\dot{X} = -\mathbf{C}(\text{law } X) \nabla f(X) + \sqrt{2 \mathbf{C}(\text{law } X)} \dot{W}.$$



Gradient flow for mean-field EKS

Mean-field EKS: $\dot{X}_t = -C(\rho_t)\nabla f(X_t) + \sqrt{2C(\rho_t)} \dot{W}_t$

Covariance: $C(\rho) = \int (x - M(\rho)) \otimes (x - M(\rho)) d\rho,$ **Mean:** $M(\rho) = \int x d\rho.$

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Covariance modulated Gradient Flows

$$\partial_t \rho = \nabla \cdot (\rho C(\rho) \nabla \mathcal{F}'(\rho)), \quad \text{with energy} \quad \mathcal{F}(\rho) = \int \log \rho d\rho + \int f d\rho.$$

$$\partial_t \rho = D^2 : (C(\rho_x) \rho_t) + \nabla \cdot (\rho_t \nabla f) = \nabla \cdot (\rho C(\rho) \nabla (\log \rho + f))$$

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⇒ **Cov-modulated metric tensor:** Formal metric for gradient flow description

$$\mathcal{W}_{\text{cov}}(\rho^0, \rho^1)^2 = \inf \left\{ \int_0^1 \int \langle \nabla \phi_t, C(\rho_t) \nabla \phi_t \rangle d\rho_t dt : \partial_t \rho_t + \nabla \cdot (\rho_t \underbrace{C(\rho_t) \nabla \phi_t}_{=V}) = 0, \rho_0 = \rho^0, \rho_1 = \rho^1 \right\}$$

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Question: Can the covariance-modulation improve convergence rates?

Convexity of energy (formal)

Splitting of Cov-OT:

$$\mathcal{W}_{\text{cov}}(\rho^0, \rho^1)^2 = \mathcal{W}_{0, \mathbb{1}}(\bar{\rho}^0, \bar{\rho}^1)^2 + D(\rho^0, \rho^1)^2.$$

shape *moments*

Cov-constrained OT:

$$\mathcal{W}_{0, \mathbb{1}}(\rho^0, \rho^1) = \inf_{(\rho, \psi)} \left\{ \int_0^1 \int |\nabla \psi|^2 d\rho_t dt : \begin{array}{l} M(\rho_t) = 0, \\ C(\rho_t) = \mathbb{1} \end{array} \right\}.$$

Constrained geodesics:

$$\partial_t \rho_t + \nabla \cdot (\rho_t \nabla \psi_t) = 0$$

$$\partial_t \psi_t + \frac{1}{2} |\psi_t|^2 + \underbrace{x \cdot \Theta x}_{\text{wavy}} = 0$$

$$\Theta = \int \nabla \psi \otimes \nabla \psi d\rho$$

$$|\dot{\rho}_0|_{\mathcal{W}_{0, \mathbb{1}}}^2 = \int |\nabla \psi|^2 d\rho = \text{tr } \Theta$$

Convexity of energy (formal)

Theorem (Improved constrained convexity)

Let $U : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}_{\geq 0}$ be an internal energy satisfying the *McCann*-conditions and set

$$\mathcal{U}(\rho) = \int U(\rho) \, dx$$

then for any **constrained** geodesic follows

$$\left. \frac{d^2}{dt^2} \mathcal{U}(\rho_t) \right|_{t=0} \geq |\dot{\rho}_0|_{\mathcal{W}_{0,1}} \int P(\rho) \geq 0,$$

with the *pressure* $P(r) = rU'(r) - U(r)$.

\Rightarrow Entropy $U(r) = r \log r$ ($P(r) = r$) is **1-convex**.

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 Quadratic potential energies $\mathcal{V}(\rho) = \frac{1}{2} \int |x|_B^2 \, d\rho$ are **constant** along **constrained** geodesics.

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Normalization and splitting



Setting: $f(x) = \frac{1}{2}|x - x_0|_B^2$

$$\partial_t \rho_t = \nabla \cdot (C(\rho) \nabla (\rho + \rho B^{-1}(x - x_0))).$$

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Square-root: Given $C_t = C(\rho_t)$ solve

$$\dot{A}_t = \frac{1}{2} \dot{C}_t A_t^{-T} \quad A_0 A_0^T = C_0$$



Non-symmetric!

Independent of choice of $\sqrt{C_0}$

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Normalization map: For $m \in \mathbb{R}^d$, $A \in \mathbb{R}^{d \times d}$

$$T_{m,A}(x) = A^{-1}(x - m).$$

Normalization and splitting

Evolution of moments and shape

Moment equations: $m_t = M(\rho_t)$, $C_t = C(\rho_t)$

$$\begin{cases} \dot{m}_t = -C_t B^{-1}(m_t - x_0) \\ \dot{C}_t = 2C_t(\mathbb{1} - B^{-1}C_t) \end{cases}$$

Normalized shape evolution:

$$\eta_t = (T_{A_t, m_t})\# \rho_t : \quad \partial_t \eta_t = \Delta \eta_t + \nabla \cdot (\eta_t x)$$

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Ornstein-Uhlenbeck evolutions:

$$C_t = ((1 - e^{-2t})B^{-1} + e^{-2t}C_0^{-1})^{-1}.$$

⇒ explicit sharp convergence rates possible!

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EVI and consequences

Relative entropy:

$$\mathcal{E}(\rho|\rho_\infty) = \int \log \frac{\rho}{\rho_\infty} d\rho$$

$$\rho(dx)_\infty \propto e^{-\frac{1}{2}|x-x_0|_B^2} dx$$

EVI and consequences

Theorem (Shape EVI)

For any $\nu \in \mathcal{P}_{0,1}$ holds

$$\frac{d}{dt} \mathcal{W}_{0,1}(\eta_t, \nu)^2 + \mathcal{W}_{0,1}(\eta_t, \nu)^2 \leq \mathcal{E}(\nu|\eta_\infty) - \mathcal{E}(\eta_t|\eta_\infty)$$

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Exponential stability of shape:

$$\mathcal{W}_{0,1}(\eta_t^1, \eta_t^2) \leq e^{-t} \mathcal{W}_{0,1}(\eta_0^1, \eta_0^2).$$

Independent of quadratic potential B !

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Corollary (HWI inequality)

For any $\nu \in \mathcal{P}_{0,1}(\mathbb{R}^d)$

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Moment part: Explicitly estimated via ODEs!

Recall: $D((m_0, C_0), (m_1, C_1)) \stackrel{?}{=} \inf \left\{ \int_0^1 \dot{m}_t \cdot C_t^{-1} \dot{m}_t + \frac{1}{4} \operatorname{tr}(\dot{C}_t C_t^{-1} \dot{C}_t C_t^{-1}) \right\}$

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Independent of quadratic potential B !



Intrinsic result! Comparison with W_2 will contain pre-factors [Carrillo-Vaes 2021]

Convergence to equilibrium



Splitting of entropy

$$\mathcal{E}(\underline{\rho} | \mathbf{N}_{x_0, B}) = \mathcal{E}(\underline{\eta} | \mathbf{N}_{0, \mathbb{1}}) + \mathcal{E}(\mathbf{N}_{M(\rho), C(\rho)} | \mathbf{N}_{x_0, B}).$$

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Evolution of shape+moments:

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$\rightarrow \partial_t \eta_t = \Delta \eta_t + \nabla \cdot (\eta_t x)$

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$$\dot{m}_t = -C_t B^{-1} (m_t - x_0)$$

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Comparison of norms:

$$|x|_B^2 \leq \left\| C_t^{\frac{1}{2}} B^{-1} C_t^{\frac{1}{2}} \right\|_2 |x|_{C_t}^2$$

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Convergence to equilibrium

Splitting of entropy

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$$\mathcal{E}(\mathbf{N}_{m_t, C_t} | \mathbf{N}_{x_0, B}) \leq \kappa(B, C_0) e^{-2t} \mathcal{E}(\mathbf{N}_{m_0, C_0} | \mathbf{N}_{x_0, B})$$

with

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- κ improves if $m_0 = x_0$
- Similar estimates for Fisher information
 \Rightarrow exponential smoothing of gradients

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Comparison to var-modulated gradient flows

Let $B \in \mathbb{R}_{\text{sym},+}^{d \times d}$ be fixed and consider energy

$$\mathcal{E}(\rho) = \int \log \rho \, d\rho + \frac{1}{2} \int \langle x, Bx \rangle \, d\rho.$$

Variance-modulated GF

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 Exponential rate depends on EVs of B !

 Only prefactors depend on EVs of B !

Summary and open questions

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- Covariance-weighted transport distance:
 - splitting of the distance in shape and moments
 - local existence of geodesics
 - improved convexity properties
- EKS is gradient flow of Covariance-modulated metric
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Thank you for your attention!