

Finite particle limit in the Ensemble Kalman Sampler

Louis Carillo

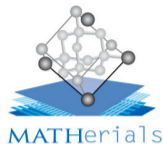
PhD under the supervision of [Tony Lelièvre](#), [Urbain Vaes](#) & [Gabriel Stoltz](#)

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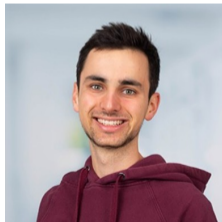


In collaboration with:

Urbain Vaes

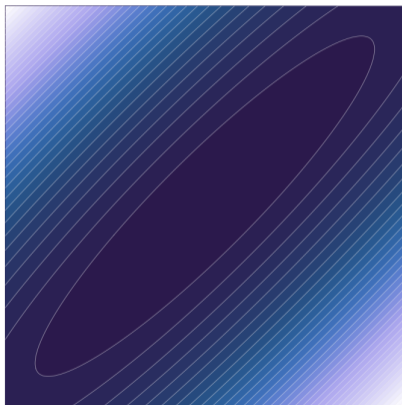


Nicolai Gerber



Julien Reygner





Sampling is **ubiquitous** in science nowadays
Today challenge: **anisotropy** in the target distribution

Take a Boltzmann distribution

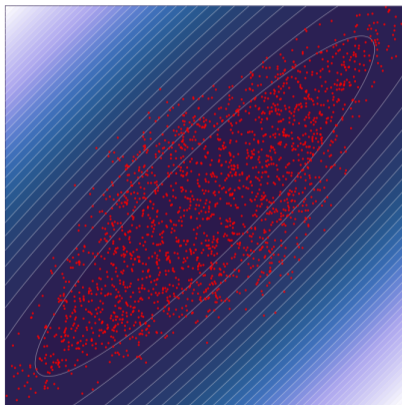
$$\mu_B(x) = \frac{1}{Z} e^{-V(x)}$$

with Z the normalization constant
and V an **anisotropic** potential

Exemple $V(x) = x^T A x$ with

$$\lambda_{\max}(A) \gg \lambda_{\min}(A)$$

Sampling **anisotropy**: Overdamped Langevin dynamics



Take N particles following

$$dX_t = -\nabla V(X_t) dt + \sqrt{2} dW_t$$

Studied for decades [1]

The invariant measure is $\mu_B^{\otimes N}$

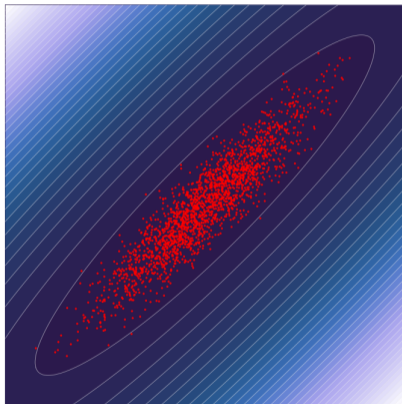
Problem:

convergence rate is constrained by $\lambda_{\min}(A)$

time step is constrained by $\lambda_{\max}(A)$

[1] Lelièvre, Stoltz *Acta Numerica* (2026)

Sampling **anisotropy**: Preconditioned Langevin dynamics



Take N particles following

$$dX_t = -C_\mu \nabla V(X_t) dt + \sqrt{2C_\mu} dW_t$$

C_μ is a **preconditioning matrix**
 \Rightarrow mitigate anisotropy[2]

Why?

$Y_t = C_\mu^{-\frac{1}{2}} X_t$ follows a **well-conditioned**
Overdamped Langevin dynamics

[2] Goodman, Weare *Commun. Appl. Math. Comput. Sci.* (2010)

Ensemble Kalman Sampler (EKS/ALDI)

Consider the following particle system

$$dX_t^n = -\mathcal{C}(\rho_{x_t}) \nabla V(X_t^n) dt + \frac{d+1}{N} (X_t^n - \mathcal{M}_{x_t}) + \sqrt{2\mathcal{C}(\rho_{x_t})} dW_t^n \quad (\text{EKS})$$

with $\mathcal{C}(\rho_{x_t})$ the **empirical covariance** of the finite particle ensemble
and **correction term** for **finite number** of particles

Idea introduced in [3]

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Idea introduced in [3]

- **Affine invariant** method, always well-conditioned
- **Gradient-free** approximation can be used
- **Square root approximation** of covariance for high-dimension

[3] Garbuno-Inigo, Nüsken, Reich *SIAM J. Appl. Dyn. Syst.* (2020)

Finite particle and mean-field limit: Invariant measure

Theorem (Garbuno-Inigo, Nüsken, Reich 2020)

For μ_B with gaussian tails, (EKS) is **ergodic** with a unique invariant measure $\mu_B^{\otimes N}$.

The particles are **interacting** though $C(\rho_{\mathcal{X}_t})$ yet they are as if they were **independent at equilibrium** as the invariant measure is a **product measure**

Open question:

What is the **convergence rate**?

Propagation of chaos has been established in [4] but only on **finite time intervals**

[4] Ding, Li *SIAM J. Math. Anal.* (2021)

Finite particle and mean-field limit: Main results

Theorem (Uniform-in-time propagation of chaos) [L.C., Gerber, Reygner, Vaes]

For $V(x) = \frac{1}{2}x^\top x$, consider $(X_t^n)_{n=1}^N$ following the (EKS) with initial condition $X_0 \stackrel{i.i.d.}{\sim} \rho_0$. Then, there exists $C > 0$ such that for any $p \geq 2$,

$$\sup_{t \geq 0, n \in \{1, \dots, N\}} W_p(\text{Law}(X_t^n), \text{Law}(\bar{X}_t)) \leq \frac{C}{\sqrt{N}}$$

with \bar{X}_t following the mean-field limit of (EKS) with initial condition $\bar{X}_0 \sim \rho_0$.

Theorem (Convergence to equilibrium) [L.C., Gerber, Reygner, Vaes 2026+]

In dimension 1, under the same assumptions, there exists $C > 0$ such that

$$W_2(\rho_{X_t}, \mu_B^{\otimes N}) \leq C e^{-(1+\varepsilon(N))t}$$

with $\varepsilon(N) \rightarrow 0$ as $N \rightarrow +\infty$.

Finite particle and mean-field limit: Main results

Theorem (Convergence to equilibrium) [L.C., Gerber, Reygner, Vaes 2026+]

In dimension 1, under the same assumptions, there exists $C > 0$ such that

$$W_2 \left(\rho_{\mathcal{X}_t}, \mu_{\mathbb{B}}^{\otimes N} \right) \leq C e^{-(1+\varepsilon(N))t}$$

with $\varepsilon(N) \rightarrow 0$ as $N \rightarrow +\infty$.

We will use a coupling between two system, $(X_t^n)_{n=1}^N$ and $(Y_t^n)_{n=1}^N$ following (EKS) with initial condition $\mathcal{X}_0 \stackrel{i.i.d.}{\sim} \rho_0$ and $\mathcal{Y}_0 \stackrel{i.i.d.}{\sim} \mu_{\mathbb{B}}$ respectively

Then

$$W_2 \left(\rho_{\mathcal{X}_t}, \mu_{\mathbb{B}}^{\otimes N} \right) \leq \mathbb{E} \left[\frac{1}{N} \sum_{n=1}^N |X_t^n - Y_t^n|^2 \right]^{\frac{1}{2}}$$

Idea of proof: difficulty with covariance moments

Consider the following **coupling** of two systems of particles following (EKS)

$$dX_t^n = -\mathcal{C}(\rho_{x_t})X_t^n dt + \frac{d+1}{N}(X_t^n - \mathcal{M}_{x_t}) dt + \sqrt{2\mathcal{C}(\rho_{x_t})} dW_t^n$$

$$dY_t^n = -\mathcal{C}(\rho_{y_t})Y_t^n dt + \frac{d+1}{N}(Y_t^n - \mathcal{M}_{y_t}) dt + \sqrt{2\mathcal{C}(\rho_{y_t})} dW_t^n$$

Key idea: If the covariances matrices $\mathcal{C}(\rho_{x_t})$ and $\mathcal{C}(\rho_{y_t})$ are **close**, then the two systems would be **close too**

Idea of proof: difficulty with covariance moments

Consider the following **coupling** of two systems of particles following (EKS)

$$\begin{aligned}dX_t^n &= -\mathcal{C}(\rho_{x_t})X_t^n dt + \frac{d+1}{N}(X_t^n - \mathcal{M}_{x_t}) dt + \sqrt{2\mathcal{C}(\rho_{x_t})} dW_t^n \\dY_t^n &= -\mathcal{C}(\rho_{y_t})Y_t^n dt + \frac{d+1}{N}(Y_t^n - \mathcal{M}_{y_t}) dt + \sqrt{2\mathcal{C}(\rho_{y_t})} dW_t^n\end{aligned}$$

Key idea: If the covariances matrices $\mathcal{C}(\rho_{x_t})$ and $\mathcal{C}(\rho_{y_t})$ are **close**, then the two systems would be **close too**

$$\begin{cases} \frac{d}{dt} \mathcal{E}_t^p \leq -2\mathcal{E}_t^p + \frac{C_p}{\sqrt{N}} (\mathcal{E}_t^{p+1} + \mathcal{F}_t^{p+1}) \\ \frac{d}{dt} \mathcal{F}_t^p \leq -2\mathcal{F}_t^p + \frac{C_p}{N} (\mathcal{E}_t^{p+1} + \mathcal{F}_t^{p+1}) \end{cases} \quad \text{with} \quad \begin{cases} \mathcal{E}_t^p = \mathbb{E} \left[\frac{1}{N} \sum_{n=1}^N |X_t^n - Y_t^n|^{2p} \right]^{\frac{1}{2p}} \\ \mathcal{F}_t^p = \mathbb{E} \left[\left\| C_{\rho_{x_t}}^{-1} - C_{\rho_{y_t}}^{-1} \right\|_F^{2p} \right]^{\frac{1}{2p}} \end{cases}$$

This leads to a **hierarchy of moments** that is difficult to close

Idea of proof: difficulty with covariance moments

The hierarchy of moments leads to results with a remainders:

$$\mathbb{E} \left[\frac{1}{N} \sum_{n=1}^N |X_t^n - Y_t^n|^2 \right] \leq C e^{-(2+\varepsilon(N))t} + \frac{C_{\mathbf{p}}}{N^{\mathbf{p}}}$$

for some \mathbf{p} arbitrarily large but fixed.

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for some \mathbf{p} arbitrarily large but fixed.

Better idea: before doing sticky coupling, [recenter the system](#):

$$U_t^n = C(\rho_{X_t})^{-\frac{1}{2}} (X_t^n - \mathcal{M}_{X_t})$$

It is easy to do in $d = 1$.

In larger dimensions, we need the Gradient and Hessian of $\varphi(A) = \sqrt{A}$ for A a positive definite symmetric matrix