

Optimizing the diffusion for sampling with overdamped Langevin dynamics Gabriel STOLTZ

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Outline

• Overdamped Langevin dynamics

- **General diffusion coefficients**
- Convergence rates

Characterization of the optimal diffusion

- Normalization of the diffusion
- Necessary conditions
- Approximation in the homogenized limit

Numerical results

- Numerical approximation of the optimal diffusion
- Numerical integration of overdamped Langevin dynamics
- **•** Efficiency gains from optimized diffusions

T. Leli`evre, G. A. Pavliotis, G. Robin, R. Santet and G. Stoltz, Optimizing the diffusion of overdamped Langevin dynamics, arXiv preprint 2404.12087

Overdamped Langevin dynamics

Computing average properties

Aim: Sample target measure $\mu(dq) = Z_\mu^{-1} \mathrm{e}^{-\beta V(q)}\,dq$ on $\mathcal{Q} = \mathbb{T}^d$ (assume $Z_u = 1$ in the remainder; configuration space = torus)

Main issue

Computation of high-dimensional integrals... Ergodic averages

$$
\mathbb{E}_{\mu}(\varphi) = \lim_{t \to +\infty} \widehat{\varphi}_t, \qquad \widehat{\varphi}_t = \frac{1}{t} \int_0^t \varphi(q_s) \, ds
$$

• One possible choice: overdamped Langevin dynamics $=$ Stochastic perturbation of gradient dynamics

$$
dq_t = -\nabla V(q_t) dt + \sqrt{\frac{2}{\beta}} dW_t
$$

• Other choices include Metropolis-like schemes

Properties of the standard overdamped Langevin dynamics

$$
\textbf{Generator} \ \mathcal{L} = -\nabla V(q) \cdot \nabla_q + \frac{1}{\beta} \Delta_q
$$

- elliptic generator hence irreducibility and ergodicity
- adjoints on $L^2(\mathcal{Q})$ versus $L^2(\mu)$

$$
\int_{\mathcal{Q}} (\mathcal{L}f) g = \int_{\mathcal{Q}} f\left(\mathcal{L}^{\dagger}g\right), \qquad \int_{\mathcal{Q}} (\mathcal{L}f) g d\mu = \int_{\mathcal{Q}} f\left(\mathcal{L}^{\star}g\right) d\mu
$$
\nflat adjoint $\mathcal{L}^{\dagger}\varphi = \text{div}_{q}\left((\nabla V)\varphi + \frac{1}{\beta}\nabla_{q}\varphi\right)$

-
- self-adjoint operator on $L^2(\mu),$ hence reversibility

$$
\mathcal{L}=-\frac{1}{\beta}\nabla_q^{\star}\nabla_q=-\frac{1}{\beta}\sum_{i=1}^d\partial_{q_i}^{\star}\partial_{q_i}=\mathcal{L}^{\star},\qquad \partial_{q_i}^{\star}=-\partial_{q_i}+\beta\partial_{q_i}V
$$

Invariance of canonical measure encoded as $\mathcal{L}^\dagger \mu$ or $\mathcal{L}^\star \mathbf{1}=0$

$$
\frac{d}{dt} \left[\mathbf{E}_{\mu} \left(\varphi(X_t) \right) \right] = \frac{d}{dt} \left(\int_{\mathcal{Q}} e^{t\mathcal{L}} \varphi \, d\mu \right) = \int_{\mathcal{Q}} \mathcal{L} \left(e^{t\mathcal{L}} \varphi \right) d\mu = 0
$$

Overdamped Langevin dynamics with multiplicative noise

Diffusion matrix $\mathcal{D}(q) \in \mathbb{R}^d$ (symmetric positive, not necessarily definite)

$$
dq_t = \left(-\mathcal{D}(q_t)\nabla V(q_t) + \frac{1}{\beta}\text{div}\mathcal{D}(q_t)\right)dt + \sqrt{\frac{2}{\beta}}\mathcal{D}^{1/2}(q_t) dW_t
$$

with div $\mathcal D$ the vector whose *i*-th component is the divergence of the *i*-th column of the matrix $\mathcal{D} = [\mathcal{D}_1, \ldots, \mathcal{D}_d]$

Two possible motivations:

- Compensate for anisotropic potential energy landscapes
- Reduce metastability

Generator still self-adjoint, invariant probability measure μ

$$
\mathcal{L}_{\mathcal{D}} = -\frac{1}{\beta} \nabla^{\star} \mathcal{D} \nabla = -\frac{1}{\beta} \sum_{i,j=1}^{d} \partial^{\star}_{q_j} \mathcal{D}_{i,j} \partial_{q_i}
$$

Behavior of overdamped Langevin dynamics for various D

Convergence of overdamped Langevin dynamics

Various measures of convergence, for instance

- asymptotic variance in central limit theorem
- \bullet convergence of the law at time t to the stationary distribution
- average exit time of a metastable well

Here: second option, in a $L^2(\mu)$ framework

Law at time t written as $\psi(t) = f(t)\mu$, so that $f(t) = e^{t\mathcal{L}_{\mathcal{D}}} f(0)$

$$
\mathbf{E}_{\psi(0)}\left(\varphi(X_t)\right) = \int_{\mathcal{Q}} \varphi f(t) \, d\mu = \int_{\mathcal{Q}} e^{t\mathcal{L}_{\mathcal{D}}}\varphi f(0) \, d\mu
$$

Typical convergence result: exponential convergence rate for ${\rm e}^{t\mathcal{L}_\mathcal{D}}$

$$
|| f(t) - \mathbf{1} ||_{L^2(\mu)} = ||e^{t\mathcal{L}_{\mathcal{D}}} (f(0) - \mathbf{1}) ||_{L^2(\mu)} \le e^{-\Lambda(\mathcal{D})t/\beta} || f(0) - \mathbf{1} ||_{L^2(\mu)}
$$

Implies bounds on the asymptotic variance

Obtaining an exponential rate of convergence

Spectral gap on
$$
H_0^1(\mu) = \left\{ u \in H^1(\mu) \middle| \int_{\mathbb{T}^d} u(q) \mu(q) dq = 0 \right\}
$$

$$
\Lambda(\mathcal{D}) = \inf_{u \in H_0^1(\mu) \setminus \{0\}} \frac{\int_{\mathbb{T}^d} \nabla u(q)^\top \mathcal{D}(q) \nabla u(q) \mu(q) dq}{\int_{\mathbb{T}^d} u(q)^2 \mu(q) dq}
$$

Desired inequality follows from a Gronwall estimate and $\frac{d}{dt}\left(\frac{1}{2}\right)$ 2 $\left\Vert \mathrm{e}^{t\mathcal{L}_{\mathcal{D}}}\varphi\right\Vert$ 2 $L^2(\mu)$ $\mathcal{L} = \left\langle e^{t\mathcal{L}_{\mathcal{D}}}\varphi, \mathcal{L}_{\mathcal{D}} e^{t\mathcal{L}_{\mathcal{D}}} \varphi \right\rangle_{L^2(\mu)} \leqslant -\frac{\Lambda(\mathcal{D})}{\beta}.$ β $\left\Vert \mathrm{e}^{t\mathcal{L}_{\mathcal{D}}}\varphi\right\Vert$ 2 $L^2(\mu)$

Criterion to choose D

Maximize the spectral gap $\Lambda(\mathcal{D})$

Possible choices:

- $\mathcal{D}=\left(\nabla^2 V\right)^{-1}$ for strongly convex potentials [Girolami/Calderhead 2011]
- \bullet $\mathcal{D} = e^{\beta V}$ [Roberts/Stramer 2002, Ghimenti/van Wijland/... 2023]

Characterization of the optimal diffusion

Need for normalization

Motivation: $\Lambda(\alpha \mathcal{D}) = \alpha \Lambda(\mathcal{D})$ and large $\mathcal D$ require smaller timesteps

 L^{∞} bounds trivial (saturate the constraint)

Chosen normalization: L^p_V $_{V}^{p}(\mathbb{T}^{d},\mathcal{M}_{a,b})$ (note $\mathcal{Q}=\mathbb{T}^{d})$, with associated norm

$$
\|\mathcal{D}\|_{L^p_V} = \left(\int_{\mathbb{T}^d} |\mathcal{D}(q)|_{\mathcal{F}}^p \,\mathrm{e}^{-\beta p V(q)}\,dq\right)^{1/p}
$$

and requirement $\mathrm{e}^{-\beta V}\mathcal{D} \in \mathcal{M}_{a,b}$ with (for $a,b\geqslant 0)$

$$
\mathcal{M}_{a,b} = \left\{ M \in \mathcal{S}_d^+ \, \middle| \, \forall \xi \in \mathbb{R}^d, \ a|\xi|^2 \leqslant \xi^\top M \xi \leqslant \frac{1}{b} |\xi|^2 \right\}
$$

Matrix norm compatible with order on symmetric positive matrices

Maximization performed on

$$
\mathfrak{D}_{p}^{a,b} = \left\{ \mathcal{D} \in L_{V}^{\infty}(\mathbb{T}^{d}, \mathcal{M}_{a,b}) \, \middle| \, \int_{\mathbb{T}^{d}} |\mathcal{D}(q)|_{\mathrm{F}}^{p} \, \mathrm{e}^{-\beta p V(q)} \, dq \leq 1 \right\}
$$

Existence of maximizer for $p \in [1,+\infty)$: For any $a \in [0,|\mathrm{Id}_d|_F^{-1}$ $\begin{bmatrix} -1 \\ F \end{bmatrix}$ and $b>0$ such that $ab\leqslant 1$, there exists $\mathcal{D}^\star_p\in\mathfrak{D}^{a,b}_p$ such that

$$
\Lambda(\mathcal{D}_p^\star) = \sup_{\mathcal{D} \in \mathfrak{D}_p^{a,b}} \Lambda(\mathcal{D})
$$

Moreover, for any open set $\Omega \subset \mathbb{T}^d$, there exists $q \in \Omega$ such that $\mathcal{D}_p^\star(q) \neq 0$.

Main arguments/properties:

- Λ is bounded (Poincaré inequality)
- Λ is concave (sup of linear functions in \mathcal{D})
- Λ is upper semicontinuous for the weak-* L^∞_V topology $(b>0)$
- the set $\mathfrak{D}^{a,b}_{p}$ is compact for the weak-* L^{∞}_{V} topology

Characterization of positive optimal diffusions $(1/2)$

Uniformly positive optimal diffusions \mathcal{D}_p^{\star} lead to degenerate eigenvalues

Precise statement:

- Frobenius norm $\left\vert \cdot\right\vert _{\mathrm{F}}$, Lebesgue exponent $p\in(1,+\infty)$ and $a=0$
- assume that there exists a maximizer \mathcal{D}_p^{\star} of Λ on $\mathfrak{D}_p^{a,b}$, with $\mathcal{D}^\star(q) \, \mathrm{e}^{-V(q)} \leqslant \frac{1}{h}$ $\frac{1}{b_+} \, \text{Id}_d$ for $b_+ > b$
- additionally $\mathcal{D}^\star_p\in C^0(\mathbb{T},\mathbb{R}_+)$ when $d=1$

If $\mathcal{D}_p^{\star} \geqslant c \mathrm{Id}_d$, then $\Lambda(\mathcal{D}_p^{\star})$ is a degenerate eigenvalue of $-\beta \mathcal{L}_{\mathcal{D}_p^{\star}}$.

Idea of proof: Proceed by contradiction and assume that the eigenvalue is simple. From the Euler–Lagrange equation (regular perturbation theory)

$$
\int_{\mathbb{T}^d} \delta \mathcal{D}(q) : \left(\nabla u_{\mathcal{D}^*_{p}} \otimes \nabla u_{\mathcal{D}^*_{p}}\right) \mu(q) \, dq = p\gamma \int_{\mathbb{T}^d} \left|\mathcal{D}^{\star}_{p}(q)\right|_{\mathcal{F}}^{p-2} \mathcal{D}^{\star}_{p}(q) : \delta \mathcal{D}(q) e^{-\beta p V(q)} \, dq,
$$

so that $\mathcal{D}_p^{\star} = \alpha_p \left| \mathcal{D}_p^{\star} \right|_{\text{F}}^{2-p}$ $\frac{2-p}{\rm F}\,{\rm e}^{\beta(p-1)V}\nabla u_{{\cal D}^\star_p} \otimes \nabla u_{{\cal D}^\star_p}$, contradicting ${\cal D}^\star_p(q) \geqslant c{\rm Id}_d$

Characterization of positive optimal diffusions (2/2)

Difficulty: Cannot directly rely on Euler-Lagrange equation

Strategy:
$$
\max_{\mathcal{D}\in\mathfrak{D}_p^{a,b}} f_{\alpha} = \text{regularize using softmax and pass to the limit}^1
$$
\n
$$
f_{\alpha}(\mathcal{D}) = \frac{\text{Tr}_{L^2(\mu)}(\mathcal{L}_{\mathcal{D}}e^{\alpha\mathcal{L}_{\mathcal{D}}})}{\text{Tr}_{L^2(\mu)}(e^{\alpha\mathcal{L}_{\mathcal{D}}}) - 1} = \frac{N_2\lambda_2 + \sum_{i\geqslant 3} N_i\lambda_i e^{\alpha(\lambda_i - \lambda_2)}}{N_2 + \sum_{i\geqslant 3} N_i e^{\alpha(\lambda_i - \lambda_2)}} \xrightarrow[\alpha \to +\infty]{} \lambda_2
$$

Can write Euler–Lagrange condition for f_α using spectral calculus

$$
\mathcal{D}_{p,\alpha}^{\star} = \gamma_{p,\alpha} \left| \mathcal{D}_{p,\alpha}^{\star} \right|_{\mathrm{F}}^{2-p} e^{\beta(p-1)V} \sum_{k \geq 2} \left[\frac{G_{\alpha} (1 + \alpha \lambda_{k,\alpha}) - \alpha H_{\alpha}}{G_{\alpha}^2} e^{\alpha \lambda_{k,\alpha}} \right] \nabla e_{k,\alpha} \otimes \nabla e_{k,\alpha}
$$
\nwith
$$
G_{\alpha} = \sum_{j \geq 2} N_j e^{\alpha \lambda_{j,\alpha}}, H_{\alpha} = \sum_{j \geq 2} N_j \lambda_j e^{\alpha \lambda_{j,\alpha}};
$$
 limit depends on $\lim_{\alpha \to +\infty} \alpha(\lambda_{j,\alpha} - \lambda_{2,\alpha})$
\nTypical example: $d = 1$, degeneracy of order 2 of first non zero eigenvalue

$$
\mathcal{D}_{p,\infty}^{\star}(q) = \tilde{\gamma}_{p,\infty} e^{\beta V(q)} \left(\left| e_{2,\infty}'(q) \right|^2 + \frac{e^{\eta} (1 + e^{\eta} + \eta)}{1 + e^{\eta} - \eta e^{\eta}} \left| e_{3,\infty}'(q) \right|^2 \right)^{1/(p-1)}
$$

Second term vanishes for a value $\eta^\star \approx 1.27$

¹Thank you Danny Perez for suggesting this!!

Approximation by homogenization theory

Diffusive time rescaling $\varepsilon q_{t/\varepsilon^2}\Rightarrow \overline{\mathcal{D}}^{1/2}B_t$ for effective diffusion $\overline{\mathcal{D}}$

Homogenized limit: for fixed D ,

- \bullet decrease the period: $\mathcal{D}_{\#,k}(q) = \mathcal{D}(kq)$ and $V_{\#,k}(q) = V(kq)$
- associated spectral gap

$$
\Lambda_{\#,k}(\mathcal{D}) = \min_{u \in H^1(\mathbb{T}^d) \setminus \{0\}} \left\{ \frac{\displaystyle \int_{\mathbb{T}^d} \nabla u^\top \mathcal{D}_{\#,k} \nabla u \, \mathrm{e}^{-\beta V_{\#,k}}}{\displaystyle \int_{\mathbb{T}^d} u^2 \, \mathrm{e}^{-\beta V_{\#,k}}} \,\,\right\} \int_{\mathbb{T}^d} u \, \mathrm{e}^{-\beta V_{\#,k}} = 0 \right\}
$$

converges to $\Lambda_{\mathrm{hom}}(\mathcal{D}),$ spectral gap of $-\mathcal{L}_{\overline{\mathcal{D}}}$ on $L^2(\mathbb{T}^d)$ with (1D case)

$$
\overline{\mathcal{D}} = \int_{\mathbb{T}^d} \mathcal{D}(q) \left(1 - w'_{\mathcal{D}}(q)^2\right) \mu(q) \, dq, \qquad \left[e^{-\beta V} \mathcal{D}(1 + w'_{\mathcal{D}})\right]' = 0
$$

Commutation optimization/homogenization: maximize $\Lambda_{\text{hom}}(\mathcal{D})$

$$
\mathcal{D}_{\text{hom}}^{\star}(q) = e^{\beta V(q)}
$$

Numerical results

Numerical discretization

Maximization of the spectral gap

- \bullet D piecewise constant, on uniform mesh
- **•** finite element approximation of test functions/eigenfunctions
- Sequential Least Squares Quadratic Programming algorithm for nonlinear eigenvalue problem with constraints

$$
A(D)U_D = \lambda(D)BU_D, \qquad U_D^{\top}BU_D = \text{Id}
$$

Discretization of the SDE

- **•** use Metropolis acceptance/rejection to ensure unbiased sampling
- $\frac{1}{2}$ and $\frac{1}{2}$ become deceptance, rejection to ensure ansies as Euler–Maruyama discretization
- lowered to $\mathrm{O}(\Delta t^{3/2})$ with dedicated (implicit) <code>HMC</code> algorithms 2

 2^2 Noble/De Bortoli/Durmus (2022), Lelièvre/Santet/Stoltz (2023) Gabriel Stoltz (ENPC/INRIA) Birmingham, Oct. 2024 17/25

Optimal diffusion / case $\eta \in (0, \eta^{\star})$

Potential $V(q) = \cos(2\pi q)$ $\eta \approx 0.51$

Spectral gaps: 30.47 (constant), 32.43 (homogenized), 36.75 (optimal)

Optimal diffusion / case $\eta = \eta^\star$

Potential $V(q) = \sin(4\pi q)(2 + \sin(2\pi q))$

Spectral gaps: 0.81 (constant), 10.6 (homogenized), 11.2 (optimal)

Influence of the lower bound

Spectral gap for various lower bounds a

Approximation by homogenized limit

Positive diffusion when periodizing

Fast convergence to the homogenized limit

Simulation of overdamped Langevin dynamics

Spectral gaps: 0.81 (constant), 10.6 (homogenized), 11.2 (optimal)

Metropolis rejection probabilities

Potential
$$
V(q) = \sin(4\pi q)(2 + \sin(2\pi q))
$$

Rejection probabilities for constant diffusion mostly where V maximal Rejection probabilities for optimized diffusion mostly where V minimal

Conclusion and perspectives

Normalization: numerical criterion (e.g. Metropolis rejection probability)

Scaling with dimension:³ diffusion depending only on some metastable degrees of freedom, e.g.

$$
D(q) = P_{\xi}^{\perp}(q) + a(\xi(q))P_{\xi}(q), \qquad P_{\xi} = \frac{\nabla \xi \otimes \nabla \xi}{\|\nabla \xi\|^2}
$$

where $\xi:\mathbb{R}^d\to\mathbb{R}^k$ (with $k\ll d)$ is a collective variable

Underdamped Langevin dynamics:

- o no variational framework
- \bullet optimization of constant diffusion⁴

⁴Chak/Kantas/Pavliotis/Lelièvre (2021)

 3 Lelièvre/Santet/Stoltz (2024)