

## 1. Introduction

$$\mu(dx) \propto e^{-U(x)} dx \quad \text{on } S = \mathbb{R}^d \quad (\text{or Riemannian manifold})$$

Sampling from  $\mu$

## ① Reversible Markov Processes

$$dX_t = -\frac{1}{2} \nabla U(X_t) dt + dB_t$$

Brownian motion

Overdamped Langevin dynamics

→ Unadjusted Langevin algorithm (=Euler scheme), Metropolis-adjusted LA, Random Walk Metropolis, ...

$$Lg = \frac{1}{2} \Delta g - \frac{1}{2} \nabla U \cdot \nabla g$$

$$\mathcal{E}(f, g) = - (f, Lg)_{L^2(\mu)} = \frac{1}{2} \int \nabla f \cdot \nabla g \, d\mu$$

Relaxation time

$$t_{\text{rel}} = 1/\text{gap}(L)$$

Special gap

$$\text{gap}(L) = \inf_{f \neq 0} \frac{\mathcal{E}(f, f)}{\text{Var}_{\mu}(f)}$$

## EXAMPLE

1)  $S = (\mathbb{R}/\ell\mathbb{Z})^d$ ,  $U$  uniformly bounded  $\Rightarrow t_{\text{rel}} \sim \ell^2$

Slow diffusive motion

2)  $S = \mathbb{R}^d$ ,  $\nabla^2 U \geq m I_d$ ,  $m > 0$

$$\Rightarrow t_{\text{rel}} \leq \frac{2}{m}, \text{ sharp in Gaussian case}$$



Condition number dependence  $\Omega(\kappa)$  for discretizations

## ② Non-reversible Markov Processes

$$\hat{S} = \mathbb{R}^d \times \mathbb{R}^d = \{(x, v) : x, v \in \mathbb{R}^d\}$$

or tangent bundle of Riem. manifold

$$\hat{\mu} = \mu \otimes \kappa$$

$$\kappa = N(0, I_d)$$

$$d\hat{\mu} \propto e^{-H(x, v)} dx dv \quad H(x, v) = U(x) + \frac{1}{2} |v|^2$$

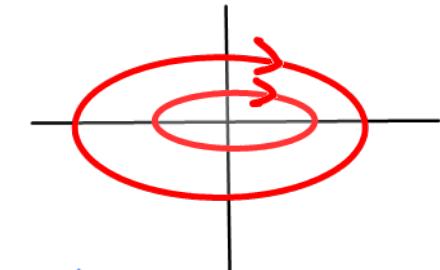
## Hamiltonian dynamics

$$\hat{L} = \mathbf{v} \cdot \nabla_{\mathbf{x}} - \nabla U(\mathbf{x}) \cdot \nabla_{\mathbf{v}}$$

$$dX_t = V_t dt$$

$$dV_t = -\nabla U(X_t) dt$$

$\leadsto$  MD



not ergodic  $\rightarrow$  add noise in v-variable

## Randomised Hamiltonian Monte Carlo

$\gamma > 0$  refreshment rate

$$\hat{L}_\gamma = \hat{L} + \gamma (\Pi_v - I)$$

$$(\Pi_v f)(x, v) = \int f(x, \omega) \kappa(d\omega)$$

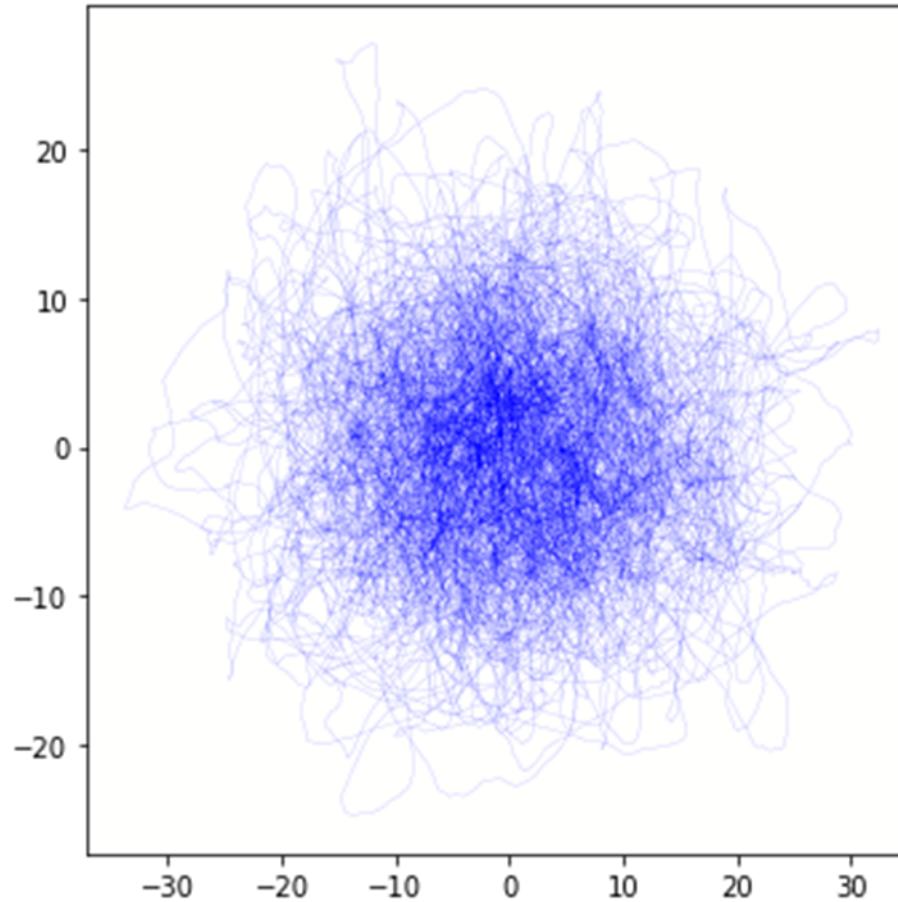
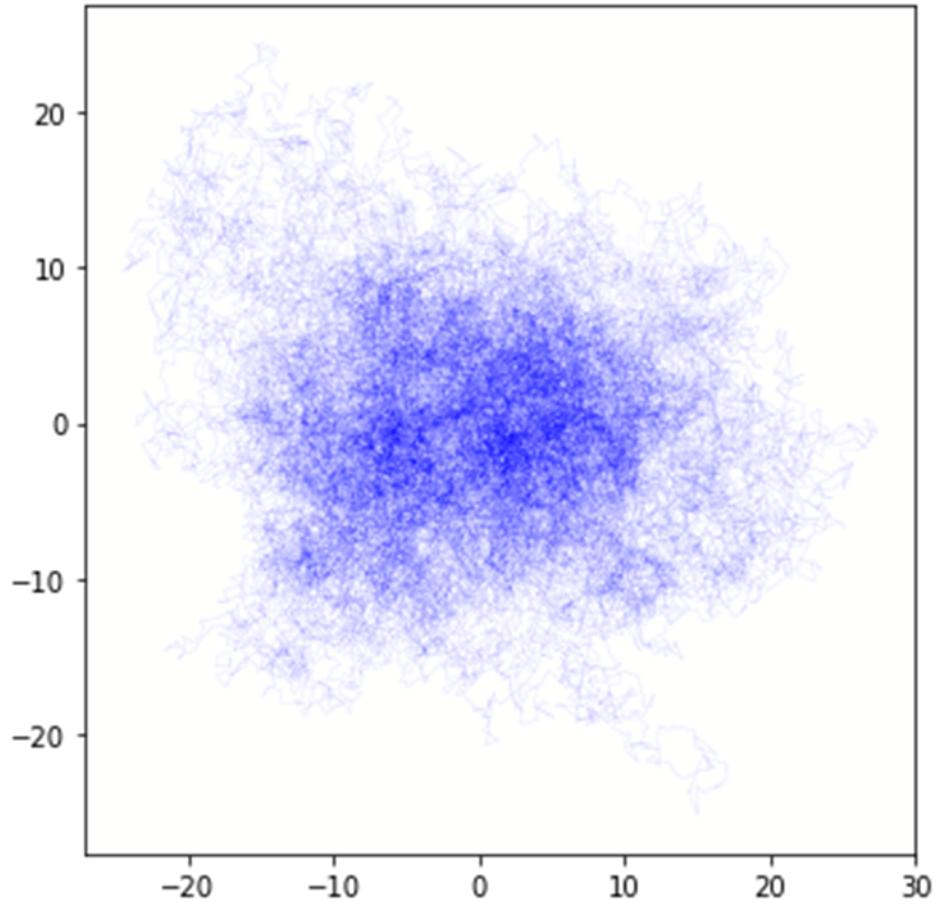
## Langevin dynamics (LD)

$$\hat{L}_\gamma = \hat{L} + \gamma L_{\text{ou}}$$

$$\hat{L}_{\text{ou}} = \Delta_{\mathbf{v}} - \mathbf{v} \cdot \nabla_{\mathbf{v}}$$

$$dX_t = V_t dt$$

$$dV_t = -\nabla U(X_t) dt - \gamma V_t dt + \sqrt{2\gamma} dB_t$$



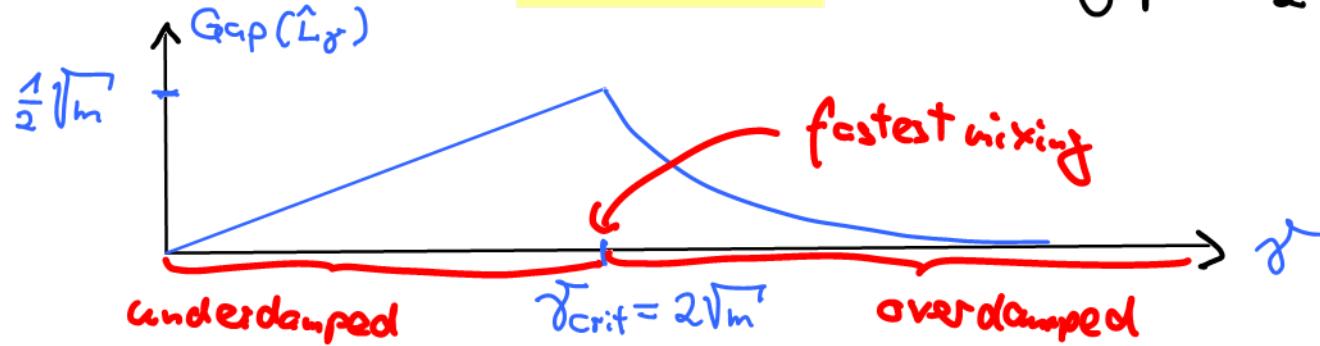
**Overdamped Langevin dynamics vs. critical Langevin dynamics in a quadratic potential.  
Each plot shows a single trajectory up to time  $t = 20000$ .**

- The sample path of critical LD changes its direction only slowly, and its empirical distribution gives a reasonable approximation of the Gaussian invariant measure.
- Conversely, due to random walk like behaviour, the empirical distribution of the sample path of OLD over the same time interval is still patchy and asymmetric.

EXAMPLE: Gaussian case

$$U(x) = m x^2/2$$

$$\text{Spectral gap} = \frac{\gamma}{2} - \sqrt{\left(\frac{\gamma^2}{4} - m\right)^+}$$



Diffusive to ballistic speed-up for  $\gamma \propto \sqrt{m}$

QUESTIONS

1) Relation between ① and ② ?



2) Consequence for relaxation times?

lower bound for relaxation times of lifts

3) Optimal lifts? Maximal speed-up?

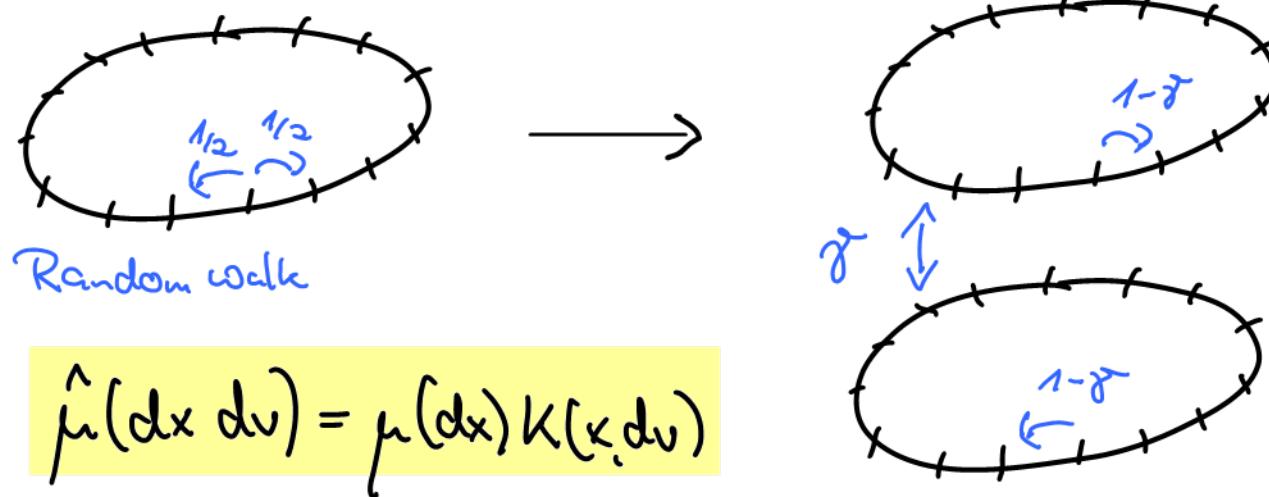
upper bound for relaxation times

## 2. Lifts: From Markov chains to diffusions

### a) Lifts of Markov chains

[Diaconis, Holmes, Neal 2000], [Chen, Lovasz, Pak 1999]

#### EXAMPLE



$$\hat{S} = S \times V$$

$$\hat{\mu}(dx dv) = \mu(dx) K(x, dv)$$

$\hat{P}$  transition kernel on  $\hat{S}$  with invariant measure  $\hat{\mu}$

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DEFINITION  $\hat{P}$  is a lift of  $P$  iff

$$(*) \int \hat{P}((x,v), A \times V) K(x, dv) = p(x, A) \quad \forall x \in S, A \subseteq V \text{ measurable}$$

**REMARK** This does not imply a corresponding relation between  $p^n$  and  $\hat{p}^n$ .

~ lift is infinitesimal property

Equivalent formulations of lift property

$$\pi(x, v) = x$$

$$(\ast\ast) \int \hat{p}(f \circ \pi)(x, v) k(x, dv) = (Qf)(x)$$

$\forall x \in S, f: S \rightarrow \mathbb{R}$  meas.+bounded

$$(\ast\ast\ast) \int \hat{p}(f \circ \pi) g \circ \pi d\hat{\mu} = \int p f g d\mu$$

$$\forall f, g \in L^2(\mu)$$

$$(\ast\ast\ast\ast) \int \hat{L}(f \circ \pi) g \circ \pi d\hat{\mu} = \underbrace{\int L f g d\mu}_{= -\mathcal{E}(f, g)}$$

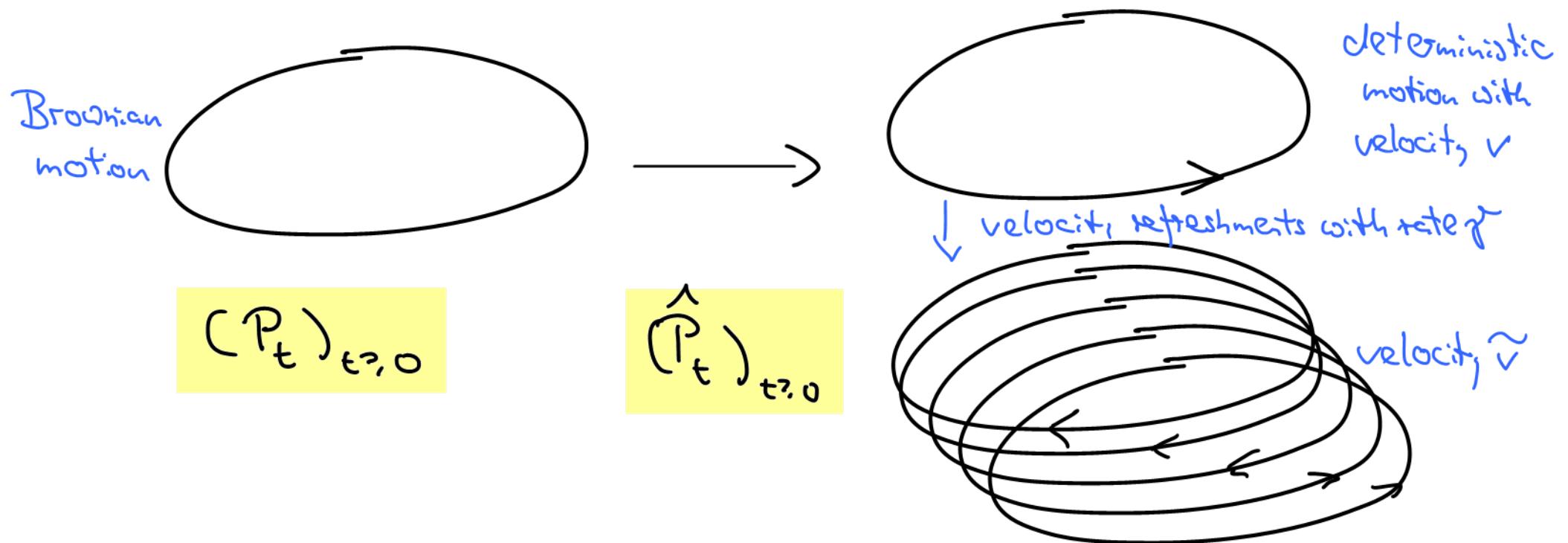
$$\forall f, g \in L^2(\mu)$$

$$\hat{L} = \hat{p} - I, L = p - I$$

"First order lift"

## b) From discrete to continuous time

EXAMPLE  $S = S^1, \mathcal{V} = \mathbb{R}, \mu = \text{Unif}(S), K(dv) = N(0,1)(dv)$



$$\int \hat{P}_t(f_{0\pi})(x,v) K(dv) = \mathbb{E}_{S_x \otimes K} [f(\hat{X}_t)] \quad \approx N(x, t^2) \quad \text{Second order lift}$$

$$= \mathbb{E}_{S_x} [f(X_{t^2})] + o(t^2) = (P_{t^2} f)(x) + o(t^2) \quad \text{as } t \downarrow 0$$

Equivalent formulation via generators:

$$(**) \quad \lim_{t \downarrow 0} \int \frac{\hat{P}_t(f_{0\pi}) - f_{0\pi}}{t^2} (x, v) K(dv) = (Lf)(x) \quad \forall f \in \text{Dom}(L)$$

$\underbrace{\qquad\qquad\qquad}_{= \frac{1}{t} \hat{L}(f_{0\pi}) + \frac{1}{2} \hat{L}^2(f_{0\pi}) + o(1)}$

$$(***) \quad \int \hat{L}(f_{0\pi})(x, v) K(dv) = 0 \quad \text{and} \quad \frac{1}{2} \int \hat{L}^2(f_{0\pi})(x, v) K(dv) = Lf(x)$$

c) Lifts of reversible diffusions

$(P_t)_{t \geq 0}$  Symmetric Markov semigroup on  $L^2(\mu)$ , generator  $L$

$(\hat{P}_t)_{t \geq 0}$  Markov semigroup on  $L^2(\hat{\mu})$ , generator  $\hat{L}$

DEFINITION [A.E., F. Lorler 2024]  $(\hat{P}_t)_{t \geq 0}$  is a 2<sup>nd</sup> order lift of  $(P_t)_{t \geq 0}$  iff

(i)  $f \in \text{Dom}(L) \Rightarrow f_{0\pi} \in \text{Dom}(\hat{L})$

(ii)  $\int \hat{L}(f_{0\pi}) g_{0\pi} d\hat{\mu} = 0 \quad \forall f, g \in \text{Dom}(L)$

(iii)  $\frac{1}{2} \int \hat{L}(f_{0\pi}) \hat{L}(g_{0\pi}) d\hat{\mu} = - \int f L g d\mu = E(f, g) \quad \forall f, g \in \text{Dom}(L)$

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(i)  $f \in \text{Dom}(L) \Rightarrow f \circ \pi \in \text{Dom}(\hat{L})$

(ii)  $\int \hat{L}(f \circ \pi) g \circ \pi \, d\hat{\mu} = 0 \quad \forall f, g \in \text{Dom}(L)$

(iii)  $\frac{1}{2} \int \hat{L}(f \circ \pi) \hat{L}(g \circ \pi) \, d\hat{\mu} = - \int f \wedge g \, d\mu = E(f, g) \quad \forall f, g \in \text{Dom}(L)$

EXAMPLES The following processes are all 2<sup>nd</sup> order lifts of OLD:

- 1) Hamiltonian dynamics
- 2) Langevin dynamics for any  $\gamma \geq 0$
- 3) Randomized HMC for any  $\gamma \geq 0$
- 4) Bouncy particle sampler

etc.

### 3. Bounds for relaxation times

$$t_{\text{rel}}(\hat{P}) := \inf \left\{ t \geq 0 : \|\hat{P}_t f\|_{L^2(\hat{\mu})} \leq \frac{1}{e} \|f\|_{L^2(\hat{\mu})} \quad \forall f \in L_0^2(\hat{\mu}) \right\}$$

non-asymptotic  $L^2$  relaxation time

**WARNING:** In general  $t_{\text{rel}}^{-1} \neq$  asymptotic decay rate,

$t_{\text{rel}} \neq$  asymptotic decorrelation time,  $t_{\text{rel}} \neq$  inverse spectral gap  $\triangleright$

**EXAMPLE**  $\|\hat{P}_t\|_{L_0^2(\hat{\mu}) \rightarrow L_0^2(\hat{\mu})} \leq C e^{-\lambda t} \Rightarrow t_{\text{rel}} \leq \frac{1}{\lambda} \log C$

#### a) General lower bound for relaxation times of lifts

**THEOREM [A.E.F. Lörler]** If  $(\hat{P}_t)$  is a 2<sup>nd</sup> order lift of  $(P_t)$  then

$$t_{\text{rel}}(\hat{P}) \geq \frac{1}{2\sqrt{2}} \sqrt{t_{\text{rel}}(P)}$$

**THEOREM [AET,Löder]** If  $(\hat{P}_t)$  is a 2<sup>nd</sup> order lift of  $(P_t)$  then

$$t_{\text{rel}}(\hat{P}) \geq \frac{1}{2\sqrt{2}} \sqrt{t_{\text{rel}}(P)}$$

**REMARK** Similar result for lifts of Markov chains by Chen, Lovasz, Pak (1999), but with "set time" instead of relaxation time.  
Proof via conductance.

**PROOF**

$$\lambda > \text{gap}(L) \Rightarrow \exists f \in L_0^2(\gamma) : E(f, f) \leq \lambda \|f\|_{L_0^2(\gamma)}^2$$

$$\Rightarrow \|\hat{L}(f \circ \pi)\|_{L_0^2(\hat{\gamma})}^2 \stackrel{\text{lift}}{=} 2E(f, f) \leq 2\lambda \|f\|_{L_0^2(\gamma)}^2 = 2\lambda \|f \circ \pi\|_{L_0^2(\hat{\gamma})}^2$$

$$\Rightarrow s(\hat{L}) := \inf_{\substack{g \in L_0^2(\hat{\gamma}) \\ g \neq 0}} \frac{\|\hat{L}g\|_{L_0^2(\hat{\gamma})}}{\|g\|_{L_0^2(\hat{\gamma})}} \leq \sqrt{2\text{gap}(L)}$$

Singular value gap, see also Chatterjee (2023)

One can show that  $t_{\text{rel}} \geq \frac{1}{2s(\hat{L})}$  (....) □

## b) Optimal lifts

### DÉFINITION

Let  $C \in [1, \infty)$ . A lift is called  $C$ -optimal iff

$$t_{\text{rel}}(\hat{P}) \leq \frac{C}{2\sqrt{2}} \sqrt{t_{\text{rel}}(P)} .$$

maximal acceleration up to factor  $C$

### EXAMPLE

Gaussian case: Critically damped LD is 5.46 optimal lift of OLD.

### THEOREM

Suppose that

(i)  $\mu$  satisfies Poincaré inequality with constant  $m \in (0, \infty)$ .

(ii)  $\nabla^2 u \geq -c_m$ ,  $c \in [0, \infty)$ .

(iii)  $\frac{1}{A} \leq \frac{\sigma}{\sqrt{m}} \leq A$ ,  $A \in [1, \infty)$ .

Then RHM is a  $C$ -optimal lift of OLD with  $C = 2\sqrt{2} \left( 482 \cdot \left( 6 + \frac{5c}{3} \right) \cdot A + 3 \right)$ .

Similar results hold for Langevin dynamics, on convex domains in  $\mathbb{R}^n$  (with reflection at boundary), and on Riemannian manifolds.

[AE, F. Lörler; work in progress]

**PROOF** Adaptation of Lu,Wang (2022) and Cao,Lu,Wang (2023).

Argument based on space-time Poincaré inequalities and divergence lemma, following approach in Albritton, Armstrong, Mourat, Novack (2023).

Simplifications by lift-framework.