

Introduction

Hamiltonian $H : \mathbb{R}^n \rightarrow \mathbb{R}$, Morse, growth at ∞
 Glauber dynamic $dX_t = -\nabla H(X_t) dt + \sqrt{2\varepsilon} dB_t$ at temperature $\varepsilon \ll 1$
 Gibbs measure $\mu(dx) = \frac{1}{Z_\mu} e^{-\frac{H}{\varepsilon}} dx$ with $Z_\mu = \int e^{-\frac{H}{\varepsilon}} dx$.
 Generator If law $X_0 = f_0\mu$, then law $X_t = f_t\mu$, where f_t solves
 Fokker-Planck equation $\partial_t f_t = Lf_t := \varepsilon \Delta f_t - \nabla H \cdot \nabla f_t$.

Definition PI and LSI

The measure μ satisfies the Poincaré inequality $PI(\varrho)$ with constant $\varrho > 0$ if

$$\forall f : \mathbb{R}^n \rightarrow \mathbb{R} \quad \text{var}_\mu(f) := \int \left(f - \int f d\mu \right)^2 d\mu \leq \frac{1}{\varrho} \int |\nabla f|^2 d\mu. \quad PI(\varrho)$$

and the logarithmic Sobolev inequality $LSI(\alpha)$ with constant $\alpha > 0$ if

$$\forall f : \mathbb{R}^n \rightarrow \mathbb{R} \quad \text{Ent}_\mu(f^2) := \int f^2 \log \frac{f^2}{\int f^2 d\mu} d\mu \leq \frac{2}{\alpha} \int |\nabla f|^2 d\mu. \quad LSI(\alpha)$$

Basic properties: $LSI(\alpha)$ implies $PI(\varrho)$ for μ with $\varrho \geq \alpha$.

$PI(\varrho)$ and $LSI(\alpha)$ imply exponential convergence to invariant measure μ :

$$PI(\varrho) \Rightarrow \text{var}_\mu(f_t) \leq \text{var}_\mu(f_0) e^{-2\varrho \varepsilon t} \quad \text{and} \quad LSI(\alpha) \Rightarrow \text{Ent}_\mu(f_t) \leq \text{Ent}_\mu(f_0) e^{-2\alpha \varepsilon t}.$$

Partitions and split of variance

Define for every local minima m_i of H its domain of attraction by

$$\Omega_i := \left\{ y \in \mathbb{R}^n : \lim_{t \rightarrow \infty} y_t = x_i, \dot{y}_t = -\nabla H(y_t), y_0 = y \right\}.$$

$\{\Omega_0, \Omega_1\}$ is partition of $\mathbb{R}^n = \bar{\Omega}_0 \cup \bar{\Omega}_1$ and $\Omega_0 \cap \Omega_1 = \emptyset$.

Introduce the restricted measures μ_0 and μ_1

$$\mu_i(dx) := \mu(dx)|_{\Omega_i} = \frac{1}{Z_i Z_\mu} e^{-\frac{H}{\varepsilon}} \Big|_{\Omega_i} dx, \quad Z_i := \mu(\Omega_i) \quad i = 0, 1.$$

\Rightarrow mixture representation: $\mu = Z_0 \mu_0 + Z_1 \mu_1$.

Motivated from [CM10] (discrete setting) and extended for LSI:

Lemma Splitting of the variance and entropy

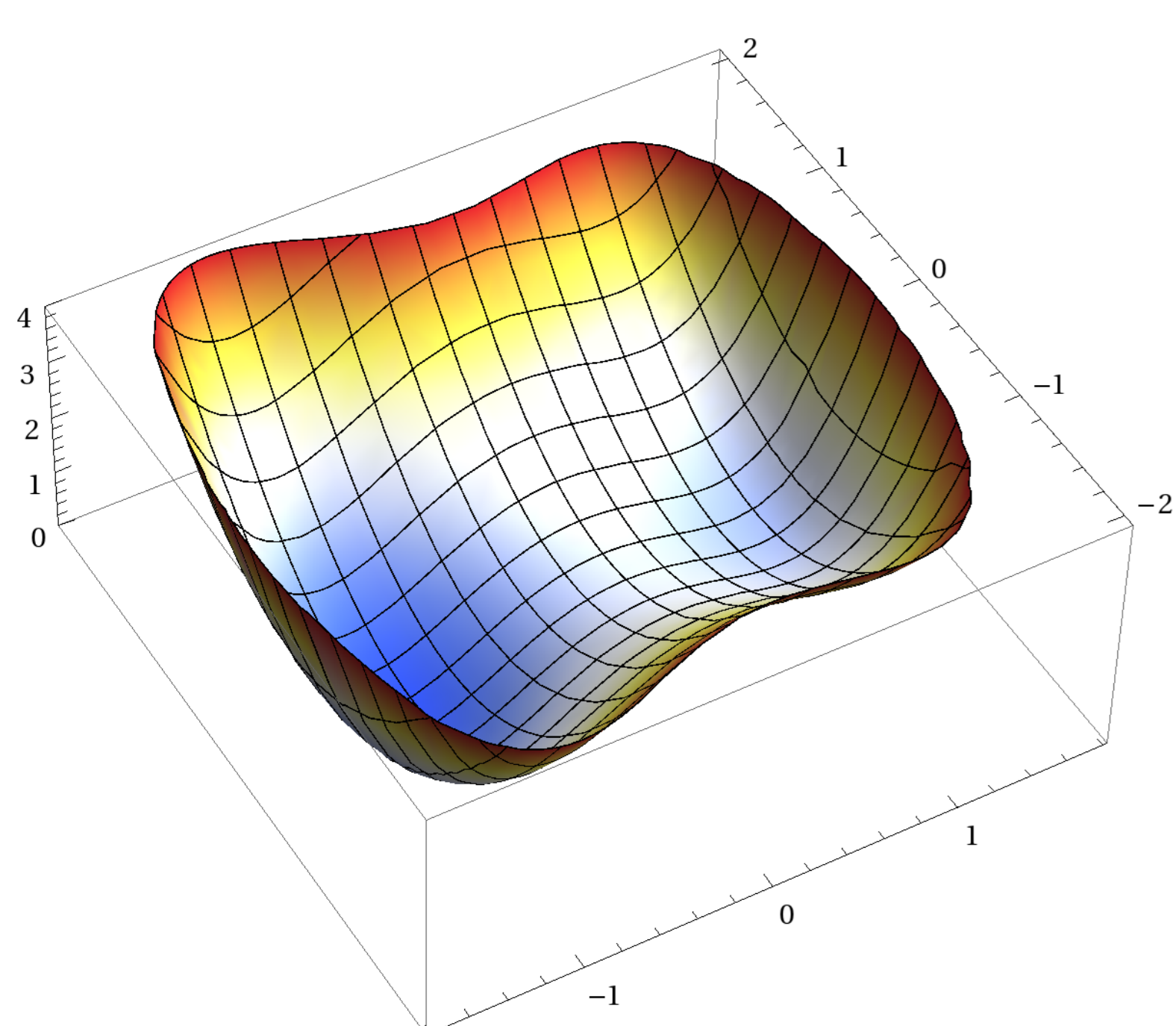
$$\text{var}_\mu(f) = \underbrace{Z_0 \text{var}_{\mu_0}(f) + Z_1 \text{var}_{\mu_1}(f)}_{\text{local variances}} + \underbrace{Z_0 Z_1 (\mathbb{E}_{\mu_0}(f) - \mathbb{E}_{\mu_1}(f))^2}_{\text{mean difference}}$$

$$\text{Ent}_\mu(f) \leq Z_0 \text{Ent}_{\mu_0}(f) + Z_1 \text{Ent}_{\mu_1}(f)$$

$$+ \frac{Z_0 Z_1}{\Lambda(Z_0, Z_1)} \left(\text{var}_{\mu_0}(f) + \text{var}_{\mu_1}(f) + (\mathbb{E}_{\mu_0}(f) - \mathbb{E}_{\mu_1}(f))^2 \right),$$

where $\Lambda(Z_0, Z_1) := \frac{Z_0 - Z_1}{\log Z_0 - \log Z_1}$ is the logarithmic mean.

Example of an energy landscape



A generic double well

Main results

Theorem 1 Local PI and LSI

The measures μ_0 and μ_1 satisfy $PI(\varrho_{\text{loc}})$ and $LSI(\alpha_{\text{loc}})$ with

$$\varrho_{\text{loc}}^{-1} = O(\varepsilon) \quad \text{and} \quad \alpha_{\text{loc}}^{-1} = O(1).$$

Proof: Use [CGW10] and explicitly construct a Lyapunov function.

Theorem 2 Mean difference estimate

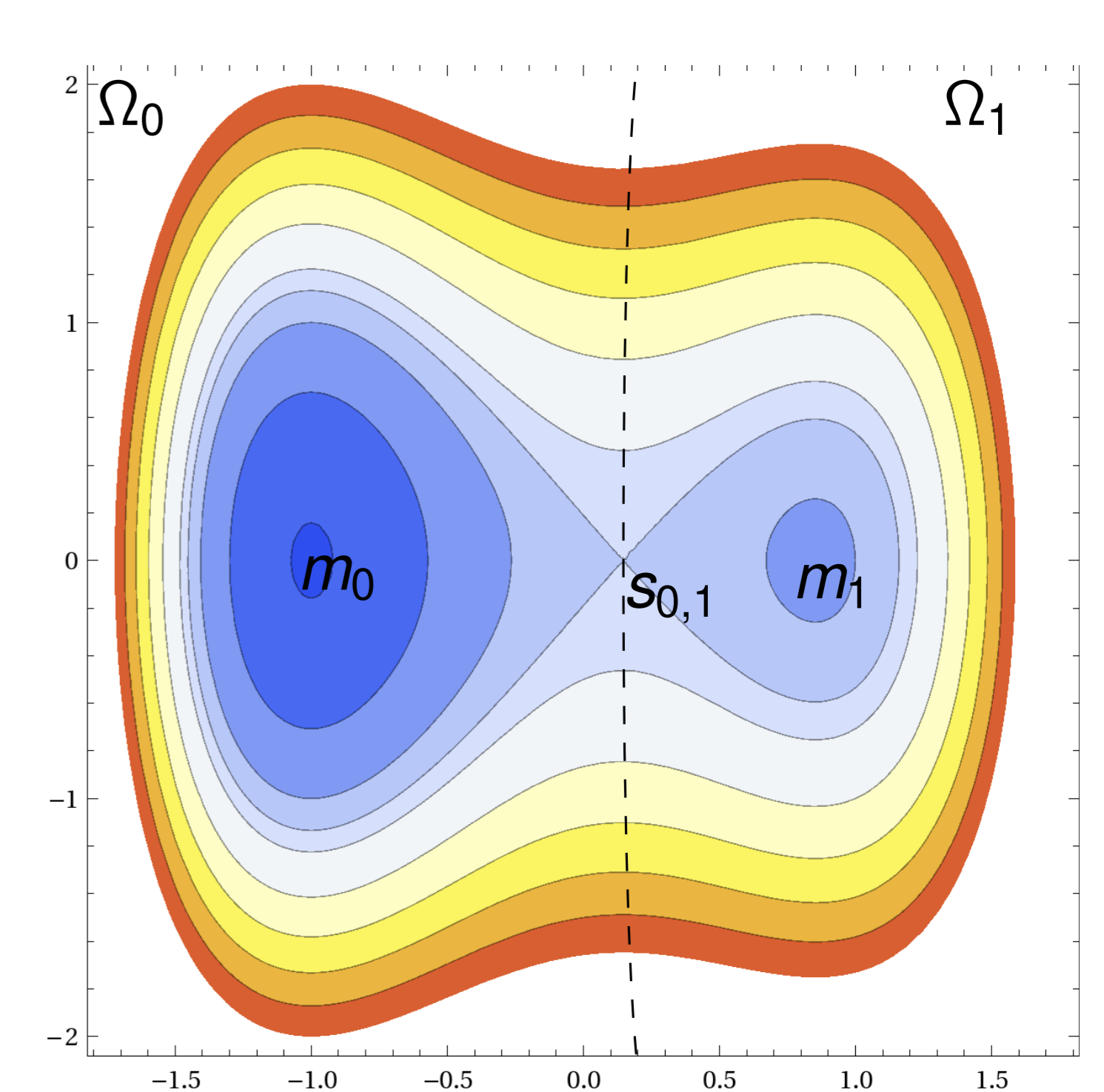
Notation: \lesssim is \leq up to factor $1 + o(1)$ as $\varepsilon \rightarrow 0$

$$(\mathbb{E}_{\mu_0}(f) - \mathbb{E}_{\mu_1}(f))^2 \lesssim \frac{Z_\mu}{(2\pi\varepsilon)^{\frac{n}{2}}} \frac{2\pi\varepsilon \sqrt{|\det \nabla^2 H(s_{0,1})|}}{|\lambda^-(\nabla^2 H(s_{0,1}))|} e^{-\varepsilon^{-1} H(s_{0,1})} \int |\nabla f|^2 d\mu,$$

where $\lambda^-(\nabla^2 H(s_{0,1}))$ is the negative eigenvalue of $\nabla^2 H(s_{0,1})$.

Proof: Coupling and transport argument

Example of the according contour



Ad Theorem 2: Coupling measures

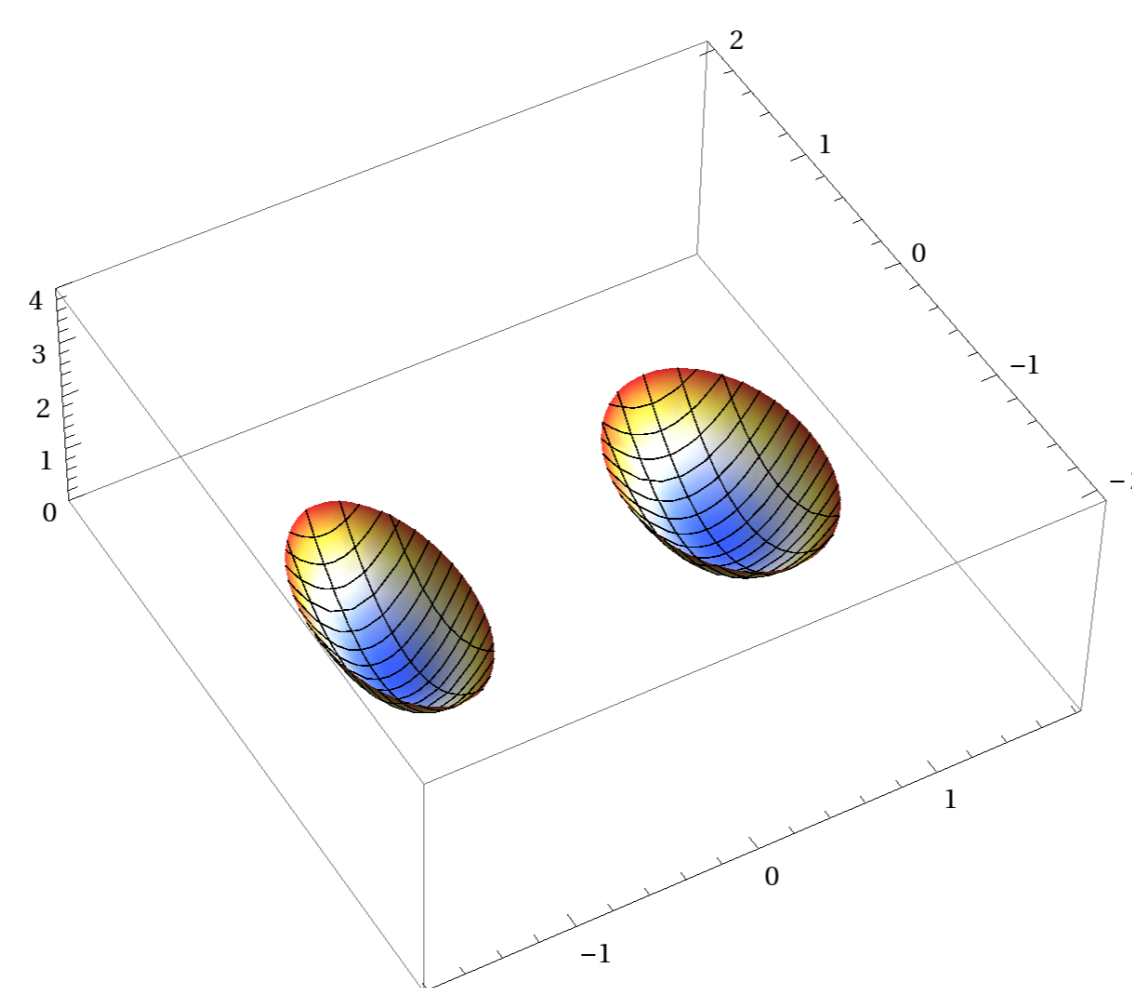
Truncated Gaussian measures ν_0 and ν_1 :

$$\nu_i \sim \mathcal{N}(m_i, \Sigma_i) \Big|_{B_{\sqrt{\varepsilon}}} \quad \text{where} \quad \Sigma_i^{-1} = \nabla^2 H(x_i).$$

Use ν_0 and ν_1 as coupling between μ_0 and μ_1 :

$$(\mathbb{E}_{\mu_0}(f) - \mathbb{E}_{\mu_1}(f))^2 \lesssim \underbrace{(\mathbb{E}_{\nu_0}(f) - \mathbb{E}_{\nu_1}(f))^2}_{\text{weighted transport cost}} + \sum_{i \in \{0,1\}} \underbrace{(\mathbb{E}_{\mu_i}(f) - \mathbb{E}_{\nu_i}(f))^2}_{\text{approximation bound}}$$

Approximation bound follows from Theorem 1.



Ad Theorem 2: Weighted transport cost

Definition weighted transport cost

For $\nu_0, \nu_1 \ll \mu$ the weighted transport cost $\mathcal{T}_\mu(\nu_0, \nu_1)$ is defined by

$$\mathcal{T}_\mu(\nu_0, \nu_1) = \inf_{\{\Phi_s\}_{s \in [0,1]}} \int \left| \int_0^1 \dot{\Phi}_s \circ \Phi_s^{-1} \frac{d\nu_s}{d\mu} ds \right|^2 d\mu,$$

where $\{\Phi_s\}_{s \in [0,1]}$ is a transport interpolation between ν_0 and ν_1 and $\nu_s = (\Phi_s)_\# \nu_0$ the pushforward of ν_0 under Φ_s .

Mean difference can be estimated by $\mathcal{T}_\mu(\nu_0, \nu_1)$:

$$\begin{aligned} (\mathbb{E}_{\nu_0}(f) - \mathbb{E}_{\nu_1}(f))^2 &= \left(\int \int_0^1 \frac{df(\Phi_s(x))}{ds} ds d\nu_0 \right)^2 = \left(\int \int_0^1 \langle \nabla f \circ \Phi_s, \dot{\Phi}_s \rangle d\nu_0 ds \right)^2 \\ &= \left(\int \left\langle \nabla f, \int_0^1 \dot{\Phi}_s \circ \Phi_s^{-1} \frac{d\nu_s}{d\mu} ds \right\rangle d\mu \right)^2 \leq \mathcal{T}_\mu(\nu_0, \nu_1) \int |\nabla f|^2 d\mu. \end{aligned}$$

Ansatz: explicit affine transport interpolation $\{\Phi_s\}_{s \in [0,1]}$ proofs Theorem 2.

Relation to $H^{-1}(\mu)$ norm: Note $\mathbb{E}_{\nu_0}(f) - \mathbb{E}_{\nu_1}(f) = (f, \nu_0 - \nu_1)_{L^2(\mu)}$

$$\text{Interpolation estimate} \quad (f, \nu_0 - \nu_1)_{L^2(\mu)}^2 \leq \|\nu_0 - \nu_1\|_{H^{-1}(\mu)}^2 \|f\|_{H^1(\mu)}^2.$$

Indeed, it holds $\mathcal{T}_\mu(\nu_0, \nu_1) = \|\nu_0 - \nu_1\|_{H^{-1}(\mu)}^2$.

\Rightarrow Dynamic representation of $H^{-1}(\mu)$ norm allows sharp estimates.

This representation generalizes to Wasserstein type distances [DNG09].

Discussion of main results

Theorem 1 + 2 reproves the lower bound of [BGK05] for PI and extends to LSI

Corollary Eyring-Kramers formula for PI and LSI

The measure μ satisfies $PI(\varrho)$ and $LSI(\alpha)$ with

$$\varrho \approx \frac{(2\pi\varepsilon)^{\frac{n}{2}}}{Z_0 Z_1 Z_\mu} \frac{|\lambda^-(\nabla^2 H(s_{0,1}))|}{2\pi\varepsilon \sqrt{|\det \nabla^2 H(s_{0,1})|}} e^{\frac{H(s_{0,1}) - H(x_1)}{\varepsilon}} \quad \text{and} \quad \alpha \gtrsim \Lambda(Z_0, Z_1) \varrho.$$

In special cases, an asymptotic evaluation of the factor $\frac{(2\pi\varepsilon)^{\frac{n}{2}}}{Z_0 Z_1 Z_\mu}$ leads to:

$$H(m_0) < H(m_1) : \quad \frac{\varrho}{\alpha} \lesssim O(\varepsilon^{-1})$$

$$H(m_0) = H(m_1) : \quad \frac{\varrho}{\alpha} \lesssim \frac{\frac{\kappa_0 + \kappa_1}{2}}{\Lambda(\kappa_0, \kappa_1)} = \frac{\text{arithmetic mean}}{\text{logarithmic mean}}, \quad \kappa_i := \sqrt{|\det \nabla^2 H(m_i)|}.$$

Especially, for $\kappa_0 = \kappa_1$ holds $\varrho \approx \alpha \Rightarrow$ LSI likes symmetry in the system.

Question: Relation to cut-off phenomenon of Markov chains?

Outlook and open questions

- discrete setting (Curie-Weiss model with random field)
- infinite dimension (stochastic PDEs like 1d stochastic Allen-Cahn equation)
- non-reversible processes (∇H replaced by $b \in C^2(\mathbb{R}^n, \mathbb{R}^n)$)
- other functional inequalities (interpolation between PI and LSI)

References

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- [CGW10] P. Cattiaux, A. Guillin, and L. Wu. A note on Talagrand's transportation inequality and LSI. *Probab. Theory Related Fields*, 148(1-2):285–304, 2010.
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