

Stochastic Localization with Non-Gaussian Tilts and Applications to Tensor Ising Models

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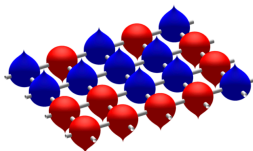
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Introduction

The Ising model

We consider the *Ising model* defined on the n -dim. hypercube $\mathcal{C}_n := \{-1, +1\}^n$.



Energy is given by the **Hamiltonian**

$$-\mathcal{H}(x) = \langle x, Jx \rangle + \langle h, x \rangle,$$

where J is the *interaction matrix* and h the *external field*.

The **Gibbs measure** is

$$\mu(x) = \frac{1}{Z_\beta} \exp(-\beta \mathcal{H}(x)).$$

Glauber Dynamics (GLD)

Markov chain $(X_t)_{t \geq 0}$ with $\text{Law}(X_t) \xrightarrow{t \rightarrow \infty} \mu$.

- Start from a configuration $x = (x_1, \dots, x_n)$.
- Repeat the following at each step
 - 1 pick a coordinate $i \in [n]$ uniformly at random,
 - 2 update $(X_t)_i$ according to $\mu|(X_t)_{j \neq i}$.

GLD has an associated reversible transition kernel P_{GLD} .

The **mixing time** is defined as

$$t_{\text{mix}}(\varepsilon) := \min \{t \geq 0 : \forall x \in \mathcal{C}_n \ \|P_{\text{GLD}}^t(x, \cdot) - \mu\|_{\text{TV}} \leq \varepsilon\}.$$

Spectral gap and Poincaré inequality

Consider the eigenvalues of P_{GLD} : $1 = \lambda_1 > \lambda_2 \geq \dots \geq \lambda_n > -1$. The *spectral gap* is defined as

$$\text{gap} := \lambda_1 - \lambda_2 = 1 - \lambda_2.$$

It holds:

$$t_{\text{mix}}(\varepsilon) \leq \left\lceil \frac{n}{\text{gap}} \left(\log \left(\frac{1}{\min_{x \in \mathcal{C}_n} \mu(x)} \right) + \log \left(\frac{1}{2\varepsilon} \right) \right) \right\rceil.$$

Fix $\varphi : \mathcal{C}_n \rightarrow \mathbb{R}$. For GLD, having gap is equivalent to satisfy a **Poincaré inequality**:

$$\text{Var}_{\mu}(\varphi) \leq C_P(\mu) \mathcal{E}_{\mu}(\varphi),$$

where

$$\mathcal{E}_{\mu}(\varphi) := \frac{1}{2} \sum_{x, y \in \mathcal{C}_n} (\varphi(x) - \varphi(y))^2 \mu(x) P_{\text{GLD}}(x, y),$$

and

$$\text{Var}_{\mu}(\varphi) = \int \varphi^2 d\mu - \left(\int \varphi d\mu \right)^2.$$

Stochastic Localization and two questions

Idea of Stochastic Localization (SL)

SL is a **measure-valued process** constructed by Eldan in 2013.

Start from a measure on \mathcal{C}_n (or \mathbb{R}^n) of the form $\frac{dm(x)}{dx} = \exp(-w(x))$.
Perturb it by a **Gaussian tilt**

$$\exp(-w(x)) \exp(\langle x, d \rangle - s \|x\|_2^2).$$

w must be "nice", for instance convex or quadratic.

SL definition

Fix μ on \mathcal{C}_n . Let $(B_t)_{t \geq 0}$ be a standard Brownian motion in \mathbb{R}^n .
SL is a process of the form $\mu_t = F_t \mu$, where F_t is the solution of

$$dF_t(x) = \langle x - a_t, C_t dB_t \rangle, \quad F_0(x) = 1,$$

- a_t is the *barycenter*:

$$a_t := \int x d\mu_t(x),$$

- C_t is the *driving matrix*: encodes constraints on the system.

Role of a_t

a_t allows μ_t to be a **probability measure**. In fact:

$$d \int \mu_t(x) dx = d \int F_t(x) \mu(x) dx = \left\langle \int x F_t(x) \mu(x) dx - a_t, C_t dB_t \right\rangle = 0.$$

SL is a Gaussian tilt

Apply Itô's formula:

$$d \log F_t = \langle x - a_t, C_t dB_t \rangle - \frac{1}{2} \|C_t(x - a_t)\|_2^2 dt.$$

We get:

$$\begin{aligned} \mu_t(x) &= \exp \left(\int_0^t \langle x - a_s, C_s dB_s \rangle - \frac{1}{2} \int_0^t \|C_s(x - a_s)\|_2^2 ds \right) \\ &\propto \exp(-\langle x, Q_t x \rangle + \langle L_t, x \rangle) \mu(x), \end{aligned}$$

where

- L_t is some adapted process in \mathbb{R}^n ,
- $Q_t = \frac{1}{2} \int_0^t C_s^2 ds$ is a matrix-valued process.

Can we localize μ with a non-Gaussian tilt?

Poincaré inequality via SL

Let τ be a suitable stopping time. Suppose we are able to prove

$$\mathrm{Var}_{\mu_\tau}(\varphi) \leq C_P(\mu_\tau) \mathcal{E}_{\mu_\tau}(\varphi),$$

Let $M_t := \int_{C_n} \varphi \, d\mu_t$. Then

$$\mathrm{Var}_\mu(\varphi) = \mathbb{E}[M_t] + \mathbb{E}[\mathrm{Var}_{\mu_t}(\varphi)].$$

Suppose $[M]_t = 0$ a.s. Then

$$\mathrm{Var}_\mu(\varphi) = \mathbb{E}[\mathrm{Var}_{\mu_\tau}(\varphi)] \leq C_P(\mu_\tau) \mathbb{E}[\mathcal{E}_{\mu_\tau}(\varphi)] \leq C_P(\mu_\tau) \mathbb{E}[\mathcal{E}_\mu(\varphi)].$$

Eldan-Koehler-Zeitouni work: role of C_t

Consider $\mu(x) \propto \exp(\langle x, Jx \rangle)$. C_t encodes two constraints

- ① $\mathbb{E}_{\mu_t}[\varphi]$ constant in time $\Rightarrow \text{Var}_{\mu_t}(\varphi)$ is a martingale.
- ② $J_t = J - \int_0^t C_s^2 ds$ decreasing, if $\text{rank}(J_t) \geq 2$.

Define $\tau = \min\{t : \text{rank}(J_t) = 1\}$. SL decomposes μ into a mixture of measures of the form

$$w_{u,v}(x) \propto \exp(\langle u, x \rangle^2 + \langle v, x \rangle).$$

It holds:

- ① $\|u\|^2 = \|J_t\|_{\text{op}} \leq \|J\|_{\text{op}}$
- ② $C_P(w_{u,v}) \leq (1 - 2\|u\|_2^2)^{-1}$

If $\|J\|_{\text{op}} < \frac{1}{2} \Rightarrow C_P(\mu) \leq (1 - 2\|J\|_{\text{op}})^{-1}$.

What happens if J is a fourth-order tensor?

Naïve answers

About fourth-order tensors

- $T \in (\mathbb{R}^n)^{\otimes 4}$.
- $T : \text{Hom}(\mathbb{R}^n, \mathbb{R}^n) \rightarrow \text{Hom}(\mathbb{R}^n, \mathbb{R}^n)$. " $n^2 \times n^2$ matrix"
- T symmetric: $T_{i_1, i_2, i_3, i_4} = T_{i_{\sigma(1)}, i_{\sigma(2)}, i_{\sigma(3)}, i_{\sigma(4)}}$ for any permutation σ .
- Injective norm: $\|T\|_{\text{inj}} = \sup_{x \in \mathbb{S}^{n-1}} |T(x, x, x, x)|$.

Naïve approach

Let $T \in (\mathbb{R}^n)^{\otimes 4}$. Consider $\nu(x) \propto \exp(T(x))$ on \mathcal{C}_n .

Define the following version of SL: $\nu_t = F_t \nu$, where F_t solves

$$dF_t(x) = \langle x^{\otimes 2} - a_t, M_t dW_t \rangle_{\text{HS}}, \quad F_0(x) = 1,$$

with

$$a_t = \int x^{\otimes 2} d\nu_t(x),$$

and where W_t is a Dyson Brownian motion, and M_t is a fourth-order tensor operating on matrices.

Problems

Apply Itô's formula:

$$\begin{aligned}\nu_t(x) &= \exp \left(\int_0^t \langle x^{\otimes 2} - a_s, M_s dW_s \rangle_{\text{HS}} - \frac{1}{2} \int_0^t \|M_s(x^{\otimes 2} - a_s)\|_{\text{HS}}^2 ds \right) \mu(x) \\ &\propto \exp \left(\left\langle x^{\otimes 2}, \left(T - \frac{1}{2} \int_0^t M_s^2 ds \right) x^{\otimes 2} \right\rangle_{\text{HS}} + \langle L_t, x^{\otimes 2} \rangle_{\text{HS}} \right),\end{aligned}$$

where

$$L_t = \int_0^t (M_s dW_s - M_s^2 a_s ds).$$

Problems:

- 1 $\langle L_t, x^{\otimes 2} \rangle_{\text{HS}}$ is not linear in x . Spectral gap can deteriorate.
- 2 We treat T as a *matrix* $\Rightarrow T - \frac{1}{2} \int_0^t M_s^2 ds$ collapses to a rank 1 matrix.

Tensor Stochastic Localization

Solution to problem 1

Problem: $L_t = \int_0^t (M_s dW_s - M_s^2 a_s ds)$ not linear in x .

Solution: eliminate it. How?

- 1 Remove barycenter. Choose v_t such that L_t is arbitrarily small.
- 2 Encode in M_t a new constraint to fix $\int 1 d\nu_t$

$\Rightarrow M_t$ encodes mass preservation and variance martingale.

$$dF_t(x) = \langle x^{\otimes 2} - v_t, C_t dW_t \rangle_{\text{HS}}, \quad F_0(x) = 1,$$

with $\mu_t = F_t \mu$ and where

- C_t is a 4-tensor operating on matrices such that $C_t dW_t$ is a matrix.
- v_t satisfies:

Proposition \exists adapted drift process $v_t^\delta \in \text{Mat}(\mathbb{R}^n, \mathbb{R}^n)$, st if X_t^δ satisfies

$$dX_t^\delta = C_t dW_t - C_t^2 v_t^\delta dt, \quad X_0^\delta = 0,$$

then $\sup_{t \geq 0} \|X_t^\delta\|_{\text{HS}} \leq \delta$ a.s.

Solution to problem 2

Define

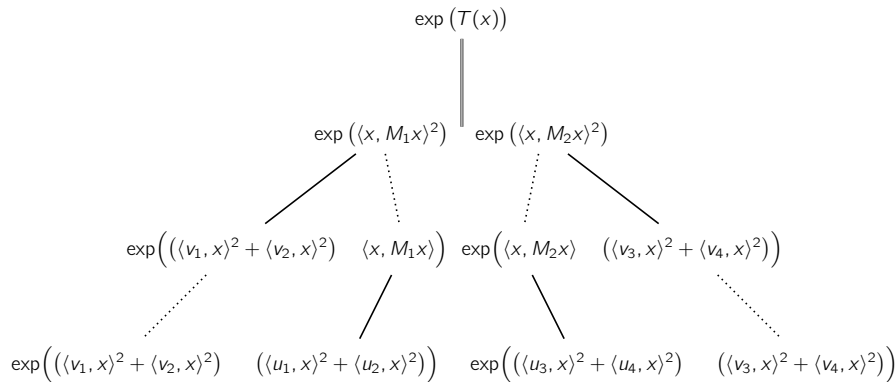
$$T_t := T - \frac{1}{2} \int_0^t C_s^2 ds \quad \text{and} \quad \tau = \inf\{t \geq 0 : \text{rank}(T_t) \leq 2\}.$$

We can decompose the measure

$$\begin{aligned} \mu(x) &= \mathbb{E}[\mu_\tau] \\ &\propto \mathbb{E}[\exp(\langle x^{\otimes 2} \otimes x^{\otimes 2}, M_1 \otimes M_1 + M_2 \otimes M_2 \rangle_{\text{HS}})] \\ &= \mathbb{E}[\exp(\langle x, M_1 x \rangle^2 + \langle x, M_2 x \rangle^2)] \end{aligned}$$

with $\|M_1\|_{\text{OP}}^2, \|M_2\|_{\text{OP}}^2 \leq \|T\|_{\text{inj}}$.

Decomposition of the measure $\mu(x) \propto \exp(T(x))$



Decomposition theorem

Let $\varphi : \mathcal{C}_n \rightarrow \mathbb{R}$ test function, let $\mu \propto \exp(T(x))$. Then, \exists **decomposition**

$$\mu = \int \mu_{\bar{u}, \bar{v}, \bar{w}, \ell} \eta(d\bar{u}, d\bar{v}, d\bar{w}, d\ell),$$

where $\bar{u} = \{u_1, u_2, u_3, u_4\}$, $\bar{v} = \{v_1, v_2, v_3, v_4\}$, and $\bar{w} = \{w_1, w_2\}$, with $u_i, v_i, w_j, \ell \in \mathbb{R}^n$ for $i \in [4]$ and $j = 1, 2$, and $\mu_{\bar{u}, \bar{v}, \bar{w}, \ell}$ are probability measures

$$\mu_{\bar{u}, \bar{v}, \bar{w}, \ell}(x) \propto \exp \left(\sum_{i,j=1}^2 \langle u_i, x \rangle^2 \langle v_j, x \rangle^2 + \sum_{i,j=3}^4 \langle u_i, x \rangle^2 \langle v_j, x \rangle^2 + \sum_{i=1}^2 \langle x, w_i \rangle^2 + \langle \ell, x \rangle \right).$$

Properties:

- For $i \in [4]$, $\|u_i\|_2^2, \|v_i\|_2^2 \leq 2\sqrt{\|T\|_{\text{inj}}}$, $\|w_i\|_2^2 \leq 4n\|T\|_{\text{inj}}$.
- **Variance decomposition**: $\text{Var}_{\mu}(\varphi) = \int \text{Var}_{\mu_{\bar{u}, \bar{v}, \bar{w}, \ell}}(\varphi) \eta(d\bar{u}, d\bar{v}, d\bar{w}, d\ell)$.

Spectral gap for tensor Ising model

Theorem Let μ be a measure on \mathcal{C}_n given by $\mu(x) \propto \exp(T(x))$. Assume $\|T\|_{\text{inj}} \leq \frac{1}{336n}$. Then

$$C_P(\mu) \leq \frac{1}{1 - 336n\|T\|_{\text{inj}}}.$$

Corollary T has independent $\mathcal{N}(0, n^{-3})$ off-diagonal entries and 0 everywhere else. Let μ be a measure on \mathcal{C}_n given by $\mu(x) \propto \exp(\beta T(x))$. If $\beta \lesssim \frac{1}{1205.568}$, then

$$C_P(\mu) \leq \frac{1}{1 - 1205.568\beta}.$$

Another approach

Trick: Normalized variance

Let ν be an unnormalized measure. Consider **normalized variance**

$$\overline{\text{Var}}_\nu(\varphi) = \mathbb{E}_\nu [\varphi^2] - \frac{\mathbb{E}_\nu [\varphi]^2}{\mathbb{E}_\nu [1]}.$$

Define $\tilde{\nu} = \frac{\nu}{\mathbb{E}_\nu[1]}$. Then

- $\overline{\text{Var}}_\nu(\varphi) = \mathbb{E}_\nu[1] \text{Var}_{\tilde{\nu}}(\varphi).$
- If $C_P(\tilde{\nu}) < \infty$. Then, $\forall \varphi : \mathcal{C}_n \rightarrow \mathbb{R}$

$$\overline{\text{Var}}_\nu(\varphi) \leq C_P(\tilde{\nu}) \mathbb{E}_\nu[1] \mathcal{E}_{\tilde{\nu}}(\varphi) = C_P(\tilde{\nu}) \mathcal{E}_\nu(\varphi).$$

Adjustments to TSL

We allow $\mathbb{E}_{\mu_t}[1]$ to vary.

$\Rightarrow C_t$ has **one** constraint: $\frac{\mathbb{E}_{\nu_t}[\varphi]}{\mathbb{E}_{\nu_t}[1]}$ is constant. This allows to preserve $\overline{\text{Var}}_{\nu_t}$.

$$\begin{array}{c}
 \exp(T(x)) \\
 | \\
 \exp(\langle x, Mx \rangle \langle x, Mx \rangle + \langle x, Rx \rangle) \\
 \begin{array}{ccc}
 \vdots & | & \vdots \\
 \exp(\langle x, Mx \rangle \langle v, x \rangle^2 + \langle x, Rx \rangle) \\
 | & \vdots & \vdots \\
 \exp(\langle u, x \rangle^2 \langle v, x \rangle^2 + \langle x, Rx \rangle) \\
 \vdots & \vdots & \diagdown \\
 \exp(\langle u, x \rangle^2 \langle v, x \rangle^2 + \langle x, w \rangle^2 + \langle \ell, x \rangle)
 \end{array}
 \end{array}$$

Decomposition theorem bis

Let $\varphi : \mathcal{C}_n \rightarrow \mathbb{R}$ and $\mu \propto \exp(T(x))$. Then, $\forall \delta > 0$, \exists decomposition

$$\mu = \int \mu_{u,v,w,\ell,\psi} \eta(du, dv, dw, d\ell, d\psi),$$

where with $u, v, w, \ell \in \mathbb{R}^n$ and $\psi : \mathcal{C}_n \rightarrow \mathbb{R}$, and $\mu_{u,v,w,\ell,\psi}$ is a non-negative measure of the form

$$\mu_{u,v,w,\ell,\psi}(x) \propto \exp(\langle u, x \rangle^2 \langle v, x \rangle^2 + \langle x, w \rangle^2 + \langle \ell, x \rangle + \psi(x)).$$

Properties:

- $\|u\|_2^2, \|v\|_2^2 \leq 2\sqrt{\|T\|_{\text{inj}}}, \|w\|_2^2 \leq 4n\|T\|_{\text{inj}}.$
- $\max_{x \in \mathcal{C}_n} |\psi(x)| \leq \delta.$
- Variance decomposition:
 $\text{Var}_\mu(\varphi) \leq \int \overline{\text{Var}}_{\mu_{u,v,w,\ell,\psi}}(\varphi) \eta(du, dv, dw, d\ell, d\psi) + \delta.$

Spectral gap for tensor Ising model bis

Theorem Let μ be a measure on \mathcal{C}_n given by $\mu(x) \propto \exp(T(x))$. Assume $\|T\|_{\text{inj}} \leq \frac{1}{56n}$. Then

$$C_P(\mu) \leq \frac{1}{1 - 56n\|T\|_{\text{inj}}}.$$

Corollary T has independent $\mathcal{N}(0, n^{-3})$ off-diagonal entries and 0 everywhere else. Let μ be a measure on \mathcal{C}_n given by $\mu(x) \propto \exp(\beta T(x))$. If $\beta \lesssim \frac{1}{200.928}$, then

$$C_P(\mu) \leq \frac{1}{1 - 200.928\beta}.$$

Thank you!