

## Problem Set 3

Due: November 30, 2023 (11:59 PM EST)

**Exercise 1** (Hardcore Model on  $K_{n,n}$ ). Consider the hardcore model with fugacity  $\lambda \geq 0$  on the complete bipartite graph  $K_{n,n}$ . Let  $\mu$  denote the Gibbs distribution, and  $Z(\lambda)$  denote the partition function.

- (a) Give a closed form expression for  $Z(\lambda)$ .
- (b) Prove that there is a constant  $C = C(\lambda) > 0$  such that for each  $0 \leq k \leq 2n - 2$ , there exists  $S \subseteq V$  with  $|S| = k$  and a pinning  $\tau : S \rightarrow \{\text{in}, \text{out}\}$  such that  $\lambda_{\max}(\Psi_{\mu^\tau}) \geq C \cdot (n - k)$ .  
(If it makes life convenient, you may assume  $k$  is even.)
- (c) Show that for every fixed  $\lambda \geq 0$  independent of  $n$ , there is a constant  $C = C(\lambda) > 0$  such that Glauber dynamics has spectral gap at most  $\exp(-Cn)$ . In particular, its worst-case mixing time is exponentially large.
- (d) Conclude that the “local-to-global theorem”, i.e. that  $O(1)$ -spectral independence for all pinnings implies inverse polynomial spectral gap, **cannot** admit an “average-case” analog of the following form:

“Let  $\mu$  be a probability distribution on  $\{\pm 1\}^n$ . Suppose for every  $\Omega(\log n) \leq k \leq n - 2$ , the conditional measure  $\mu^\tau$  is  $O(1)$ -spectrally independent with high probability (e.g.  $1 - \frac{1}{\text{poly}(n)}$ ) over a random pinning  $\tau \sim \mu_S$  on a uniformly random set  $S$  of  $k$  coordinates. Then Glauber dynamics for  $\mu$  has spectral gap at least  $1/\text{poly}(n)$ , and polynomial mixing time.”

**Exercise 2** (Continuous to Discrete). Let  $\mu$  be a probability measure on  $\{\pm 1\}^n$ .

- (a) Suppose  $\mu$  can be decomposed as

$$\mu(\sigma) = \int_{\mathbb{R}^n} (\mathcal{T}_x \nu)(\sigma) d\xi(x), \quad \forall \sigma \in \{\pm 1\}^n,$$

for some distribution  $\nu$  on  $\{\pm 1\}^n$  and some distribution  $\xi$  on  $\mathbb{R}^n$  satisfying the following properties:

- **Poincaré Inequality for Mixture:** There is a constant  $C > 0$  such that for every test function  $\psi : \mathbb{R}^n \rightarrow \mathbb{R}$ ,

$$\text{Var}_{x \sim \xi}[\psi(x)] \leq C \cdot \int_{\mathbb{R}^n} \|\nabla \psi(x)\|_2^2 d\xi(x). \quad (1)$$

- **Spectral Independence for Components:** There is a constant  $\eta > 0$  such that for every  $x \in \mathbb{R}^n$ , the exponential tilt  $\mathcal{T}_x \nu$  of  $\nu$  is  $\eta$ -spectrally independent.

Prove that for every  $x \in \mathbb{R}^n$ , the exponential tilt  $\mathcal{T}_x \mu$  of  $\mu$  is  $O(\alpha \eta^2)$ -spectrally independent. (Note that since this holds for all exponential tilts, we also get  $O(\alpha \eta^2)$ -entropic independence for all exponential tilts.)

- (b) Let  $A$  be a symmetric positive definite matrix. Show that

$$\exp\left(\frac{1}{2} \sigma^\top A \sigma\right) = (2\pi)^{-n/2} \int_{\mathbb{R}^n} \exp\left(-\frac{1}{2} \|x\|_2^2 + \langle A^{1/2} x, \sigma \rangle\right) dx$$

for every  $\sigma \in \mathbb{R}^n$ .<sup>1</sup>

---

<sup>1</sup>The “fancy name” for this is the *Hubbard–Stratonovich transform*. Up to scaling, the right-hand side is just a moment generating function.

- (c) Let  $\mu(\sigma) \propto \exp(\frac{1}{2}\sigma^\top A \sigma)$  be the Gibbs distribution over  $\{\pm 1\}^n$  of an Ising model with interaction matrix  $A \in \mathbb{R}^{n \times n}$  satisfying  $\frac{\epsilon}{2} \cdot \text{Id} \preceq A \preceq (1 - \frac{\epsilon}{2}) \cdot \text{Id}$  for some constant  $0 < \epsilon < 1$ . Note that we can always shift  $A$  by a diagonal matrix without affecting the distribution  $\mu$ , and so we can assume  $A$  is positive (semi)definite without loss of generality; the key here is ensuring a bound on the spectral diameter  $\lambda_{\max}(A) - \lambda_{\min}(A) \leq 1 - \epsilon$ . Prove that  $\mathcal{T}_x \mu$  is  $O(1/\epsilon)$ -spectrally independent for all  $x \in \mathbb{R}^n$ .

Since the emphasis here is on the connection between the continuous and discrete, you may use the following fact from the continuous world without proof:

**Theorem 0.1.** Let  $\xi(x) \propto \exp(-V(x))$  be an  $\alpha$ -strongly log-concave probability distribution on  $\mathbb{R}^n$ , i.e. for some constant  $\alpha > 0$ ,  $\nabla^2 V(x) \succeq \alpha \cdot \text{Id}$  for all  $x \in \mathbb{R}^n$ . Then  $\xi$  satisfies the Poincaré Inequality<sup>2</sup> Eq. (1) with constant  $C = \frac{1}{\alpha}$ .

- (d) **The Sherrington–Kirkpatrick Spin Glass:** Let  $G \sim \text{GOE}(n)$ , i.e.  $G_{ij} \sim \mathcal{N}(0, 1/n)$  i.i.d. for  $i < j$ ,  $G_{ji} = G_{ij}$  for  $i > j$ , and  $G_{ii} = 0$  for  $i \in [n]$ . Prove that if  $\beta < \frac{1-\epsilon}{4}$  for a constant  $0 < \epsilon < 1$ , then with high probability over the randomness of  $G$ , all exponential tilts of the Ising Gibbs measure with interaction matrix  $A = \beta G$  are  $O(1/\epsilon)$ -spectrally independent.<sup>3</sup> Conclude that w.h.p. over  $G$ , Glauber dynamics mixes in polynomial-time.<sup>4</sup> On the other hand, what is the largest value of  $\beta$  for which you can make path coupling contract?

For this problem, you may use the following standard fact from random matrix theory, which can be proved using the trace moment method.

**Theorem 0.2.** With high probability over  $G \sim \text{GOE}(n)$ , we have  $-2(1 + o(1)) \cdot \text{Id} \preceq G \preceq 2(1 + o(1)) \cdot \text{Id}$ .

<sup>2</sup>Under this strong log-concavity assumption, a log-Sobolev inequality also holds. This is a consequence of the beautiful Bakry–Émery theory.

<sup>3</sup>It is known that for this distribution, we have  $\|\Psi_\mu\|_{\ell_\infty \rightarrow \ell_\infty} \gtrsim \sqrt{n}$ , roughly for the same reason that  $\|G\|_{\ell_\infty \rightarrow \ell_\infty} \gtrsim \sqrt{n}$ .

<sup>4</sup>The naïve bound one would get for the mixing time is  $n^{1/\epsilon}$ , although using concentration bounds for the norms of random submatrices, one can bring this down to  $C^{1/\epsilon} n^c$  for (reasonable) universal constants  $C, c > 0$ . Using stochastic localization, one can actually establish the optimal  $O(n \log n)$ -mixing.