

Gentle Statistical Mechanics — Fourth HW problem set

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▷ **Exercise 1.** For spin configuration $\sigma : V \rightarrow \{\pm 1\}$ on a finite graph $G(V, E)$, consider some energy functional $H(\sigma)$. Use Lagrange multipliers to answer these:

- (a)• For any possible energy level $E \in \mathbb{R}$, among all probability measures μ on $\{\pm 1\}^{V(G)}$ with $\mathbf{E}_\mu[H(\sigma)] = E$, which are the measures maximizing the entropy $\text{Ent}(\mu) := -\sum_\omega \mu(\omega) \log \mu(\omega)$?
- (b)• Without fixing the energy level, what are measures maximizing $\lambda \mathbf{E}_\mu[H] + \text{Ent}(\mu)$ for $\lambda \in \mathbb{R}$ fixed?

As in class, the **Ising model** on a finite graph $G(V, E)$ is the random spin configuration $\sigma : V \rightarrow \{\pm 1\}$ defined as follows. Given an external magnetic field $h \in \mathbb{R}$, the Hamiltonian is

$$H_h(\sigma) := -h \sum_{x \in V(G)} \sigma(x) - \sum_{(x,y) \in E(G)} \sigma(x)\sigma(y),$$

and then the measure, at inverse temperature $\beta = 1/T \geq 0$, is

$$\mathbf{P}_{\beta,h}[\sigma] := \frac{\exp(-\beta H_h(\sigma))}{Z_{\beta,h}}, \quad \text{where } Z_{\beta,h} := \sum_{\sigma} \exp(-\beta H_h(\sigma)).$$

▷ **Exercise 2.** The partition function $Z_{\beta,h}$ contains a lot of information about the model:

- (a)• Show that the **expected total energy** is

$$\mathbf{E}_{\beta,h}[H] = -\frac{\partial}{\partial \beta} \ln Z_{\beta,h}, \quad \text{with variance } \text{Var}_{\beta,h}[H] = -\frac{\partial}{\partial \beta} \mathbf{E}_{\beta,h}[H].$$

- (b)• The **average free energy** or **pressure** is defined by $f(\beta, h) := (\beta|V|)^{-1} \ln Z_{\beta,h}$. Show that for the **average total magnetization** $M(\sigma) := |V|^{-1} \sum_{x \in V} \sigma(x)$, we have

$$m(\beta, h) := \mathbf{E}_{\beta,h}[M] = \frac{\partial}{\partial h} f(\beta, h).$$

- (c)• The **susceptibility** of the total magnetization to a change in the external magnetic field is

$$\chi(\beta, h) := \frac{1}{\beta} \frac{\partial}{\partial h} m(\beta, h) = \frac{1}{\beta} \frac{\partial^2}{\partial h^2} f(\beta, h).$$

Relate this quantity to $\text{Var}_{\beta,h}[M]$. Deduce that $f(\beta, h)$ is convex in h .

▷ **Exercise 3.♦♦** On a sequence of transitive finite graphs G_n , using the previous exercise and maybe accepting some extra assumptions, can you deduce from the Long-Range Order (LRO) condition

$$\text{Var}_{\beta,h=0}[M_n] \asymp 1$$

at some shared $\beta > 0$ and $h = 0$, the Spontaneous Symmetry Breaking (SSB) property

$$\lim_{h \rightarrow 0^+} \lim_{n \rightarrow \infty} \mathbf{E}_{\beta,h}[M_n] > 0 > \lim_{h \rightarrow 0^-} \lim_{n \rightarrow \infty} \mathbf{E}_{\beta,h}[M_n]?$$

▷ **Exercise 4.** Consider the Ising model on an interval, $\{\sigma_i : i = -n, \dots, n-1, n\}$, with no boundary condition, at any inverse temperature $\beta \in [0, \infty)$.

(a)• Show that $\{\sigma_i : i = -n, \dots, n-1, n\}$ is the trajectory of a stationary irreducible Markov chain on $\{-, +\}$.

(b)•• Show that $S_n := \sum_{i=-n}^n \sigma_i$ has $\text{Var}[S_n] \sim C_\beta n$, as $n \rightarrow \infty$, for some $C_\beta \in (0, \infty)$.

That is, in dimension 1, there is no long range order for any $\beta < \infty$.

The next exercise is basically repeated from a class several weeks ago, when we talked about the Harris-FKG inequality. Stochastic domination will be a key notion in the study of the Ising model.

▷ **Exercise 5.**• Let (Ω, \leq) be a partially ordered set; think of (\mathbb{R}, \leq) or $(\{0, 1\}^{\mathbb{Z}^d}, \subseteq)$. Let π a probability measure on $\Omega \times \Omega$ with the property that $\pi(\{(x, y) \in \Omega \times \Omega : x \leq y\}) = 1$. Let the first marginal of π be $\mu(A) := \pi(A \times \Omega)$ and the second marginal be $\nu(A) := \pi(\Omega \times A)$ for any event $A \subseteq \Omega$. So, π is a **monotone coupling** of μ and ν . Show that ν **stochastically dominates** μ : for any increasing (upward closed) event A , we have $\mu(A) \leq \nu(A)$. (In class we talked about stochastic domination and coupling for random variables, not probability measures, but if X is a random variable with values in Ω , then there is an associated measure $\mu(A) := \mathbf{P}[X \in A]$, and then everything is the same.)

By the way, Strassen's theorem says that the converse also holds: if ν stochastically dominates μ , then a monotone coupling does exist.

▷ **Exercise 6.** By constructing monotone couplings, show the following:

(a)• If $n \leq m$, then $\text{Binom}(m, p)$ stochastically dominates $\text{Binom}(n, p)$.

(b)• If $0 \leq p \leq q \leq 1$, then $\text{Binom}(n, q)$ stochastically dominates $\text{Binom}(n, p)$.

(c)• If S_n and \tilde{S}_n , $n \geq 0$ are nearest neighbour (± 1) random walks on \mathbb{Z} started at $S_0 = \tilde{S}_0 = 1$, with jump distributions

$$\mathbf{P}[S_{n+1} = S_n + 1 \mid S_n] = p \leq q = \mathbf{P}[\tilde{S}_{n+1} = \tilde{S}_n + 1 \mid \tilde{S}_n],$$

and $\tau := \inf\{n : S_n = 0\}$ and $\tilde{\tau} := \inf\{n : \tilde{S}_n = 0\}$, then $\tilde{\tau}$ stochastically dominates τ .