

SC9 Probability on Graphs and Lattices: Sheet #3

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Problem 1

Obtaining magnetisation from pressure.

Let $m_G(\omega) = \frac{1}{|V(G)|} \sum_{i \in V(G)} \sigma_i(\omega)$ be the magnetisation per site. Let $\Psi_{G;\beta,h} = \log(Z_{G;\beta,h})/|V(G)|$ be the pressure. Show that

$$\langle m_G \rangle_{G;\beta,h} = \frac{\partial \Psi_{G;\beta,h}}{\partial h}.$$

Proof.

$$\begin{aligned} \frac{\partial \Psi_{G;\beta,h}}{\partial h} &= \frac{1}{|V(G)|Z_{G;\beta,h}} \cdot \sum_{\omega \in \Omega_G} \frac{\partial e^{-\mathcal{H}_{G;\beta,h}(\omega)}}{\partial h} \\ &= \frac{1}{|V(G)|Z_{G;\beta,h}} \cdot \sum_{\omega \in \Omega_G} e^{-\mathcal{H}_{G;\beta,h}(\omega)} \cdot \frac{\partial(-\mathcal{H}_{G;\beta,h}(\omega))}{\partial h} \\ &= \frac{1}{|V(G)|Z_{G;\beta,h}} \cdot \sum_{\omega \in \Omega_G} \sum_{i \in V(G)} e^{-\mathcal{H}_{G;\beta,h}(w)} \cdot \sigma_i(\omega) \\ &= \frac{1}{|V(G)|} \cdot \sum_{i \in V(G)} \sigma_i(\omega) \cdot \sum_{\omega \in \Omega_G} \frac{e^{-\mathcal{H}_{G;\beta,h}(w)}}{Z_{G;\beta,h}} \\ &= m_G(\omega). \end{aligned}$$

□

Problem 2

Spatial Markov property.

Fix $(\beta, h) \in \mathbb{R}_{\geq 0} \times \mathbb{R}$, $\eta \in \{\pm 1\}^{\mathbb{Z}^d}$ and let $G' \subseteq G$ be finite subgraphs of \mathbb{Z}^d . Show that we have

$$\mu_{G; \beta, h}^\eta(\cdot \mid \sigma_i = \eta_i, \forall i \in V(G) \setminus V(G')) = \mu_{G'; \beta, h}^\eta(\cdot).$$

Proof.

$$\begin{aligned} \mathcal{H}_{G; \beta, h}^\eta(\omega \mid \sigma_i = \eta_i, \forall i \in V(G) \setminus V(G')) &= -\beta \sum_{(i,j) \in E(G) \cup \partial G} \sigma_i(\omega) \sigma_j(\omega) - h \sum_{i \in V(G)} \sigma_i(\omega) \\ &= \mathcal{H}_{G'; \beta, h}^\eta(\omega) - \beta \sum_{(i,j) \in E(G \setminus G') \cup \partial G} \sigma_i(\omega) \sigma_j(\omega) - h \sum_{i \in V(G \setminus G')} \sigma_i(\omega). \end{aligned}$$

Since all vertices in $V(G) \setminus V(G')$ are conditioned on η , we may write the last two sums as a constant $C = C(\eta)$. Thus,

$$\mathcal{H}_{G; \beta, h}^\eta(\omega \mid \sigma_i = \eta_i, \forall i \in V(G) \setminus V(G')) = \mathcal{H}_{G'; \beta, h}^\eta(\omega) + C.$$

Let

$$Z' = \sum_{\substack{\omega \in \Omega_G \\ \sigma_i = \eta_i, \forall i \in V(G) \setminus V(G')}} e^{-\mathcal{H}_{G'; \beta, h}^\eta(\omega) - C} = e^{-C} \cdot Z_{G'; \beta, h}.$$

It now follows that

$$\mu_{G; \beta, h}^\eta(\cdot \mid \sigma_i = \eta_i, \forall i \in V(G) \setminus V(G')) = \frac{e^{-\mathcal{H}_{G'; \beta, h}^\eta(\cdot) - C}}{e^{-C} \cdot Z_{G'; \beta, h}} = \frac{e^{-\mathcal{H}_{G'; \beta, h}^\eta(\cdot)}}{Z_{G'; \beta, h}} = \mu_{G'; \beta, h}^\eta(\cdot).$$

□

Problem 3

The measures $\mu_{G;\beta,h}^+$ and $\mu_{G;\beta,h}^-$ are extremal.

Let $f : \{\pm 1\}^{\mathbb{Z}^d} \rightarrow \mathbb{R}$ be an increasing function. Show that for all configurations $\eta \leq \omega$ in $\{\pm 1\}^{\mathbb{Z}^d}$ we have

$$\langle f \rangle_{G;\beta,h}^\eta \leq \langle f \rangle_{G;\beta,h}^\omega.$$

Proof. Let

$$R_{G;\beta,h}^{\omega,\eta}(\sigma) = \exp \left(\mathcal{H}_{G;\beta,h}^\eta(\sigma) - \mathcal{H}_{G;\beta,h}^\omega(\sigma) \right) = \exp \left(\beta \sum_{(i,j) \in \partial G} \sigma_i \cdot (\omega_j - \eta_j) \right).$$

Then

$$\begin{aligned} \langle f \rangle_{G;\beta,h}^\omega &= \sum_{\Omega_G^\omega} f(\sigma) \cdot \frac{e^{-\mathcal{H}_{G;\beta,h}^\omega(\sigma)}}{Z_{G;\beta,h}^\omega} \\ &= \sum_{\Omega_G^\omega} f(\sigma) \cdot \frac{e^{-\mathcal{H}_{G;\beta,h}^\eta(\sigma)} \cdot R_{G;\beta,h}^{\omega,\eta}(\sigma)}{Z_{G;\beta,h}^\omega} \\ &= \frac{Z_{G;\beta,h}^\eta}{Z_{G;\beta,h}^\omega} \sum_{\Omega_G^\eta} f(\sigma) \cdot R_{G;\beta,h}^{\omega,\eta}(\sigma) \cdot \frac{e^{-\mathcal{H}_{G;\beta,h}^\eta(\sigma)}}{Z_{G;\beta,h}^\eta} \\ &= \frac{Z_{G;\beta,h}^\eta}{Z_{G;\beta,h}^\omega} \langle f R_{G;\beta,h}^{\omega,\eta} \rangle_{G;\beta,h}^\eta. \end{aligned}$$

Note that

$$\sum_{\Sigma_G^\omega} \mu_{G;\beta,h}^\omega(\sigma) = \frac{Z_{G;\beta,h}^\eta}{Z_{G;\beta,h}^\omega} \sum_{\Sigma_G^\eta} \mu_{G;\beta,h}^\eta(\sigma) \cdot R_{G;\beta,h}^{\omega,\eta}(\sigma) = \frac{Z_{G;\beta,h}^\eta}{Z_{G;\beta,h}^\omega} \langle R_{G;\beta,h}^{\omega,\eta} \rangle_{G;\beta,h}^\eta = 1,$$

so

$$\frac{Z_{G;\beta,h}^\eta}{Z_{G;\beta,h}^\omega} = \frac{1}{\langle R_{G;\beta,h}^{\omega,\eta} \rangle_{G;\beta,h}^\eta}.$$

It now follows from the FKG inequality that

$$\langle f \rangle_{G;\beta,h}^\omega = \frac{\langle f R_{G;\beta,h}^{\omega,\eta} \rangle_{G;\beta,h}^\eta}{\langle R_{G;\beta,h}^{\omega,\eta} \rangle_{G;\beta,h}^\eta} \geq \frac{\langle f \rangle_{G;\beta,h}^\eta \cdot \langle R_{G;\beta,h}^{\omega,\eta} \rangle_{G;\beta,h}^\eta}{\langle R_{G;\beta,h}^{\omega,\eta} \rangle_{G;\beta,h}^\eta} = \langle f \rangle_{G;\beta,h}^\eta.$$

□

Problem 4

Spin at 0 and average magnetisation.

Define $m^+(\beta, h)$ and $m^-(\beta, h)$ as

$$\lim_{n \rightarrow \infty} \frac{\langle \sum_{i \in \Lambda(n)} \sigma_i \rangle_{\Lambda(n); \beta, h}^+}{|\Lambda(n)|}, \quad \lim_{n \rightarrow \infty} \frac{\langle \sum_{i \in \Lambda(n)} \sigma_i \rangle_{\Lambda(n); \beta, h}^-}{|\Lambda(n)|},$$

respectively. Show that $\langle \sigma_0 \rangle_{\beta, h}^\pm = m^\pm(\beta, h)$.

Hint: Show an upper and lower bound. Argue (and use) that for an increasing function f and subgraphs $G' \subseteq G \subseteq \mathbb{Z}^d$, $\langle f \rangle_{G', \beta, h}^+ \geq \langle f \rangle_{G, \beta, h}^+$, then establish bounds on the limsup and liminf using appropriate choices of G' , G and f .

Proof. We first note that

$$\langle \sigma_0 \rangle_{\beta, h}^+ = \langle \sigma_i \rangle_{\beta, h}^+.$$

for all $i \in \mathbb{Z}^d$. By Problem 3 and the strong Markov property, for increasing function f ,

$$\langle f \rangle_{G', \beta, h}^+ \geq \mathbb{E}_\alpha[\langle f \rangle_{G', \beta, h}^\alpha] = \langle f \rangle_{G, \beta, h}^+.$$

But then σ_i is increasing, so

$$\langle \sigma_0 \rangle_{\beta, h}^+ = \langle \sigma_i \rangle_{\beta, h}^+ \leq \langle \sigma_i \rangle_{\Lambda(n); \beta, h}^+,$$

for any $n \in \mathbb{N}$. Let

$$m_n^+(\beta, h) = \frac{\sum_{i \in \Lambda(n)} \langle \sigma_i \rangle_{\Lambda(n); \beta, h}^+}{|\Lambda(n)|}.$$

Then

$$\liminf_{n \rightarrow \infty} m_n^+(\beta, h) \geq \langle \sigma_0 \rangle_{\beta, h}^+.$$

On the other hand, fix an integer k . Let $\Lambda_i(k)$ denote the length k box centred at i . Assume n is sufficiently large and consider the sites $i \in \Lambda(n)$ such that $\Lambda_i(k) \subset \Lambda(n)$. Then

$$\langle \sigma_i \rangle_{\Lambda(n); \beta, h}^+ \leq \langle \sigma_i \rangle_{\Lambda_i(k); \beta, h}^+ = \langle \sigma_0 \rangle_{\Lambda(k); \beta, h}^+.$$

Thus

$$\limsup_{n \rightarrow \infty} m_n^+(\beta, h) \leq \langle \sigma_0 \rangle_{\Lambda(k); \beta, h}^+.$$

But then

$$\langle \sigma_0 \rangle_{\beta, h}^+ \leq \liminf_{n \rightarrow \infty} m_n^+(\beta, h) \leq \limsup_{n \rightarrow \infty} m_n^+(\beta, h) \leq \lim_{k \rightarrow \infty} \langle \sigma_0 \rangle_{\Lambda(k); \beta, h}^+ = \langle \sigma_0 \rangle_{\beta, h}^+.$$

The result for the negative case follows from symmetry. \square

Problem 5

Ising model on the complete graph.

The *Curie-Weiss* model is the Ising model on the complete graph K_N with N vertices. Despite the lack of geometry, there are still interesting phase transitions. Let $h = 0$ throughout.

- (i) Let $R(\omega) = \#\{i : \sigma_i(\omega) = 1\}$ be the number of + spins (which is N minus the number of - spins). Show that the probability that $R(\omega) = k$ is given by

$$\mu_{K_N; \beta, 0}(R(\omega) = k) = \frac{1}{\tilde{Z}_{N, \beta}} \binom{N}{k} \exp(-2\beta k(N - k))$$

for $0 \leq k \leq N$ (where $\tilde{Z}_{N, \beta}$ is a normalising constant).

Proof. Let $\omega \in \Omega_{K_N}^+$ such that $R(\omega) = k$. Let $C = e^{-\beta \binom{N}{2} - 2k(N - k)}$. Then

$$\mathcal{H}_{K_N; \beta, 0}(\omega) = -\beta \sum_{(i,j) \in E(K_N)} \sigma_i(\omega) \sigma_j(\omega) = -\beta \left(\binom{k}{2} + \binom{N-k}{2} - k(N-k) \right) = -\beta \left(\binom{N}{2} - 2k(N-k) \right).$$

But then

$$\begin{aligned} \mu_{K_N; \beta, 0}(R(\omega) = k) &= \frac{\binom{N}{k} \exp(-\beta \left(\binom{N}{2} - 2k(N-k) \right))}{\sum_{i=0}^N \binom{N}{i} \exp(-\beta \left(\binom{N}{2} - 2i(N-i) \right))} \\ &= \frac{C \binom{N}{k} \exp(-\beta(2k(N-k)))}{C \sum_{i=0}^N \binom{N}{i} \exp(-\beta(2i(N-i)))} \\ &= \frac{1}{\tilde{Z}_{N, \beta}} \binom{N}{k} \exp(-2\beta k(N-k)). \end{aligned}$$

□

- (ii) Consider taking the limit $N \rightarrow \infty$, with $\beta = \theta/N$ for some fixed $\theta > 0$. Investigate which k (approximately) maximises the probability in (i). Interpret your findings.

[Perhaps useful: from Stirling's formula, if $k = (\alpha + o(1))N$ as $N \rightarrow \infty$, then $\binom{N}{k} \approx \exp[N(f(\alpha) + o(1))]$, where $f(\alpha) = -\alpha \log \alpha - (1 - \alpha) \log(1 - \alpha)$.]

Proof.

$$\mu_{K_N; \theta/N, 0}(R(\omega) = k) = \frac{1}{\tilde{Z}_{N, \theta/N}} \binom{N}{k} \exp\left(\frac{-2\theta k(N-k)}{N}\right).$$

Write $k = (\alpha + o(1))N$ as $N \rightarrow \infty$. Let $f(\alpha) = -\alpha \log \alpha - (1 - \alpha) \log(1 - \alpha)$. By Stirling's formula,

$$\binom{N}{k} \approx \exp[N(f(\alpha) + o(1))].$$

Put $\phi(\alpha) = f(\alpha) - 2\theta(\alpha(1 - \alpha))$ and we have

$$\mu_{K_N; \theta/N, 0}(R(\omega) = k) = \frac{1}{\tilde{Z}_{N, \theta/N}} \exp(\phi(\alpha) + o(1)).$$

Note that the exponential function is increasing, so the maximum is achieved when $\phi(\alpha)$ is maximized.

□

Problem 6

FKG and BK in the context of the random-cluster model.

- One can show that the FKG inequality holds for the random-cluster model for all G when $q \geq 1$. Does it also hold for $q < 1$?

Proof. No. Consider graph G with two vertices and two parallel edges x, y . Let A_x and A_y be the event where edge x and y are open, respectively. Then

$$\mathbb{P}(\emptyset) = \frac{q^2(1-p)^2}{Z}, \quad \mathbb{P}(A_x) = \mathbb{P}(A_y) = \frac{qp}{Z}, \quad \mathbb{P}(A_x \cap A_y) = \frac{qp^2}{Z}, \quad Z = q^2(1-p)^2 + 2qp(1-p) + qp^2.$$

If the FKG inequality held, we would have

$$\frac{qp^2}{Z} \geq \left(\frac{qp}{Z}\right)^2 \implies Z \geq q.$$

But then

$$q \leq q^2(1-p)^2 + 2qp(1-p) + qp^2 \implies q \geq 1,$$

contradiction. \square

- Fix $q > 1$. Does the BK inequality hold for the random-cluster model on G with parameters p, q for all $G, p \in (0, 1)$?

Proof. No. Using the same events as in the previous proof, note that

$$\mathbb{P}(A_x \circ A_y) = \mathbb{P}(A_x \cap A_y) = \frac{qp^2}{Z}.$$

Since $\frac{q}{Z} < 1$ for $q > 1$,

$$\mathbb{P}(A_x \circ A_y) = \frac{qp^2}{Z} > \left(\frac{qp}{Z}\right)^2 = \mathbb{P}(A_x)\mathbb{P}(A_y),$$

contradicting the BK inequality. \square

Problem 7

Uniqueness of Gibbs measure for Ising model at high temperature.

- (i) Let G be a finite subgraph of \mathbb{Z}^d . Let $\beta > 0$ and $p \in (0, 1)$ be connected by $p = 1 - e^{-\beta}$. Fix $q \geq 2$.

Let $\phi_{G;p,q}^{\text{free}}$ and $\phi_{G;p,q}^{\text{wired}}$ be free and wired random-cluster measures on G respectively, and let $\mu_{G;\beta,q}$ and $\mu_{G;\beta,q}^b$ be Gibbs measures for the Potts model on G with free boundary conditions, and with all- b boundary conditions for some fixed colour b , respectively.

Show that for any $i, j \in V(G)$,

$$\begin{aligned}\mu_{G;\beta,q}(\sigma_i = \sigma_j) &= \frac{1}{q} + \frac{q-1}{q} \phi_{G;p,q}^{\text{free}}(i \leftrightarrow j) \\ \mu_{G;\beta,q}^b(\sigma_i = \sigma_b) &= \frac{1}{q} + \frac{q-1}{q} \phi_{G;p,q}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)).\end{aligned}$$

Proof. By Theorem 3.12, $\sigma_i = \sigma_j$ in the free random-cluster model with probability $1/q$ if i and j are not in the same cluster and probability 1 otherwise. Thus,

$$\mu_{G;\beta,q}(\sigma_i = \sigma_j) = \frac{1}{q}(1 - \phi_{G;p,q}^{\text{free}}(i \leftrightarrow j)) + \phi_{G;p,q}^{\text{free}}(i \leftrightarrow j) = \frac{1}{q} + \frac{q-1}{q} \phi_{G;p,q}^{\text{free}}(i \leftrightarrow j).$$

Similarly, $\sigma_i = \sigma_b$ in the wired random-cluster model with probability $1/q$ if i is disconnected from the boundary and probability 1 otherwise. Thus,

$$\mu_{G;\beta,q}^b(\sigma_i = \sigma_b) = \frac{1}{q}(1 - \phi_{G;p,q}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G))) + \phi_{G;p,q}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)) = \frac{1}{q} + \frac{q-1}{q} \phi_{G;p,q}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)).$$

□

- (ii) By applying the second identity in (i) to the $+$ boundary and $-$ boundary Ising models, show that at sufficiently high temperature (i.e. for β sufficiently small), there is a unique Gibbs measure for the Ising model in \mathbb{Z}^d .

[Hint: you may like to use the FKG inequality to compare connectivity probabilities for percolation with those for random-cluster with $q = 2$.]

Proof. Note that $-2^{k(\omega)}$ is an increasing function. By the FKG inequality,

$$-\phi_{G;p,2}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)) = \frac{\langle -2^{k(\omega)} \mathbf{1}_{i \leftrightarrow \mathbb{Z}^d \setminus V(G)} \rangle_{G;p,1}^{\text{wired}}}{\langle 2^{k(\omega)} \rangle_{G;p,1}^{\text{wired}}} \geq -\phi_{G;p,1}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)).$$

Thus,

$$\phi_{G;p,2}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)) \leq \phi_{G;p,1}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)).$$

This implies that vertex i is more likely to be connected to the boundary in the percolation model than in the random-cluster model with $q = 2$. Let (G_n) be an exhaustion of \mathbb{Z}^d by finite subgraphs. Then when $p < p_c$,

$$\lim_{n \rightarrow \infty} \phi_{G;p,2}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)) \leq \lim_{n \rightarrow \infty} \phi_{G;p,1}^{\text{wired}}(i \leftrightarrow \mathbb{Z}^d \setminus V(G)) = 0.$$

It now follows from (i) that

$$\lim_{n \rightarrow \infty} \mu_{G_n;\beta,q}^+(\sigma_i = +) = \lim_{n \rightarrow \infty} \mu_{G_n;\beta,q}^-(\sigma_i = -) = \frac{1}{q}.$$

□