A COURSE ON THE ISING MODEL

PIET LAMMERS

Contents

Preface

These lecture notes are written progressively during the 2025 spring semester, as the course is taught at Sorbonne university in the M2 (second-year masters) programme. Its purpose is to give a broad introduction to the rigorous analysis of the Ising model. The main focus is on four techniques and their applications:

- The Peierls argument,
- The random-currents representation,
- The FKG inequality for the Ising spins,
- The FKG inequality for the random-cluster (FK) representation.

A basic understanding of analysis and probability theory is essential for following this course. Experience with other models in statistical mechanics (such as the Bernoulli percolation model) is a plus but by no means essential.

The appendices contain overviews of the main definitions, expansions, and inequalities in this text.

These notes are inspired by the lecture notes *Lectures on the Ising and Potts models on the hypercubic lattice* and the overview *100 Years of the (Critical) Ising Model on the Hypercubic Lattice*, both due to Hugo Duminil-Copin. The main text does not contain references at this stage; they will be added at a later time.

1. The Curie–Weiss model

At the end of the 19th century, Curie published his experimental results on *ferromagnetism*: the magnetic properties of metals. He made three striking observations.

- The magnetic strength of a metal varies with the temperature. Increasing the temperature decreases the magnetic strength.

- Each metal has a certain temperature, specific to that metal, at which the magnetic properties disappear entirely. We call this temperature the *Curie temperature*.
- Around the Curie temperature, the magnetic strength drops continuously to zero. In other words, the magnetic strength does not "jump" to zero.

The first observation singles out the temperature as the driving parameter of the system. This is good news for us, since the temperature may be regarded informally as the amount of "randomness" or "entropy" in the system, justifying a probabilistic analysis of the situation. The second observation implies that there is a *phase transition*: there is a special temperature (in this case the Curie temperature) at which the system's behaviour undergoes a qualitative change. The third observation entails an important property of this phase transition.

The first mathematical explanation for Curie's experimental results came from Weiss. He proposed the following mathematical axioms for studying the magnetic properties of metals.

- The metal consists of n atoms.
- Each atom acts like a small magnet in itself. It is in one of two states, denoted $\pm$.
- The total strength of the metal is obtained by summing the states of all atoms.
- Each atom interacts with all other atoms. The atoms prefer to *align*, that is, to be in the same state. The temperature regulates the strength of the interaction.

Physically, it makes sense that the temperature regulates the interaction strength. When atoms move slowly, they will stabilise, oriented in alignment with the magnetic field imposed by the other atoms. When atoms move fast, they will not bother with the states of the other atoms, and simply align themselves randomly. It is thus natural to think of the interaction strength as the *inverse temperature*.

Definition 1.1 (Curie-Weiss model). The Curie-Weiss model is the probability measure $\mathbb{P}_{n,\beta}^{\mathrm{CW}}$ on $\sigma \in \Omega := \{+, -\}^n$ defined via

$$\mathbb{P}(\sigma) := \mathbb{P}_{n,\beta}^{\mathrm{CW}}(\sigma) \propto e^{-H_{n,\beta}^{\mathrm{CW}}(\sigma)}; \qquad H(\sigma) := H_{n,\beta}^{\mathrm{CW}}(\sigma) := -\frac{\beta}{n}\sum_{i<j}\sigma_i\sigma_j,$$

where $n \in \mathbb{Z}_{\geq 1}$ and $\beta \in [0, \infty)$. The parameter β is called the *interaction strength* or *inverse temperature*. The function H is called the *Hamiltonian* and captures the *energy* in the system. The probability measure $\mathbb{P}$ is also called the *Boltzmann distribution*.

Let $n_+ = n_+(\sigma)$ denote the number of vertices with spin $+$ in a configuration $\sigma \in \Omega$. This is a random variable. Let us try to calculate the probability of the event $\{n_+ = k\}$, without worrying about the partition function (the normalising constant). One may easily check that the Hamiltonian satisfies

$$H(\sigma) = 2\frac{\beta}{n}n_+(n - n_+) + \mathrm{const}(n).$$

The distribution of n_+ can then be calculated as follows:

$$\mathbb{P}(\{n_+ = k\}) \propto \binom{n}{k}e^{-2\frac{\beta}{n}k(n-k)} \propto \frac{1}{k!(n-k)!}e^{-2\frac{\beta}{n}k(n-k)}. \tag{1}$$

Using Stirling's approximation for the factorials, we find that

$$\log\mathbb{P}(\{n_+ = k\}) \overset{\mathrm{Stirling}}{\approx} -nf_\beta(k/n) + \mathrm{const}(n);$$

$$f_\beta : [0, 1] \to \mathbb{R}, \ x \mapsto x\log x + (1 - x)\log(1 - x) + 2\beta x(1 - x).$$

If we fix β and send n to infinity, then we discover a large deviations principle for the random variable n_+/n with rate function f_β and speed n. In particular, the random variable n_+/n concentrates around the minimisers of the function f_β.

Exercise 1.2 (The rate function of the Curie–Weiss model). (1) Show that for small β, the function f_β has a single minimum at $x = 1/2$, which means that the random variable n_+/n concentrates around the value $1/2$.

(2) Show that for large β, the function f_β has two minima at $(1\pm m)/2$ for some $m > 0$, which means that the random variable n_+/n is concentrated around these minima. The value of m is called the *magnetisation*.

(3) Calculate the critical value for β. At this value, the second derivative of f_β vanishes at $x = 1/2$. What does this mean for the distribution of n_+/n? Estimate the order of magnitude of $\operatorname{Var} \frac{n_+}{n}$ as $n \to \infty$ for this value of β.

Remark 1.3 (Entropy versus energy in the Curie–Weiss model). Reconsider Equation (1). In this equation, the competition between the two factors is extremely transparent.

- First, there is a combinatorial term or *entropy*, which favours values k for the random variable n_+ such that the cardinality of the set $\{n_+ = k\}$ is large. This means that values $k \approx n/2$ are preferred.
- Second, there is the *energy* term, which favours values such that the energy is minimised. This favours configurations where as many spins as possible align.

The interaction parameter β allows us to put more emphasis on the entropy term or on the energy term. In the $n \to \infty$ limit, there is a precise value for β where the behaviour of the random system undergoes a qualitative change: a rudimentary example of a *phase transition*.

2. Ising's model and basic notions

While the competition between entropy and energy is transparent in the Curie–Weiss model, the model does not encode any kind of geometry. Indeed, all atoms interact equally with all other atoms. It would perhaps be more realistic to place the atoms on a Euclidean grid, and let the interactions strength between two atoms depend on their distance. In the simplest case, we could simply let each atom interact only with the atoms closest to it. This is called the *nearest-neighbour interaction*. We mainly focus on this setup in these lecture notes.

Wilhelm Lenz challenged his doctoral student Ernst Ising to solve this nearest-neighbour model for magnetism on the one-dimensional line graph $\mathbb{Z}$. Lenz was not entirely precise when posing this question, and it was Ising who first formulated a definition for the model under consideration. The model is therefore called the *Ising model* in his honour. We shall later derive Ising's result from a broader theorem (Theorem 4.4).

Definition 2.1 (Ising model). The Ising model on a finite graph $G = (V, E)$ with *inverse temperature* $\beta \in [0, \infty)$ is defined as follows. Let $\Omega := \{\pm 1\}^V$ denote the set of spin configurations on the vertices of the graph; a typical element of Ω is denoted by $\sigma = (\sigma_u)_{u \in V}$. Elements $\sigma \in \Omega$ are called *spin configurations*; elements σ_u are called *spins*. The *energy* or *Hamiltonian* of a spin configuration σ is given by

$$H_{G,\beta}^{\text{Ising}}(\sigma) := -\beta \sum_{uv \in E} \sigma_u \sigma_v.$$

We write $\mathbb{P}_{G,\beta}^{\text{Ising}}$ for the associated *Boltzmann distribution* or *Gibbs measure*:

$$\mathbb{P}_{G,\beta}^{\text{Ising}}(\sigma) := \frac{1}{Z_{G,\beta}^{\text{Ising}}} e^{-H_{G,\beta}^{\text{Ising}}(\sigma)},$$

where $Z_{G,\beta}^{\text{Ising}}$ is normalisation constant or *partition function* defined by

$$Z_{G,\beta}^{\text{Ising}} := \sum_{\sigma \in \Omega} e^{-H_{G,\beta}^{\text{Ising}}(\sigma)}.$$

We shall write $\langle\,\cdot\,\rangle_{G,\beta}^{\text{Ising}}$ for the expectation functional associated to this probability measure.

Remark. We shall often suppress subscripts and superscripts when they are clear from the context.

Remark. The mathematical community has widely adopted the terminology coming from the physics literature. We often prefer the symbol $\langle\,\cdot\,\rangle$ over $\mathbb{E}[\,\cdot\,]$ when taking expecations, expect when considering *conditional* expectations.

Exercise 2.2 (The edge graph). Consider the Ising model on the complete graph on the two vertices $V := \{x, y\}$ at inverse temperature $\beta \in [0, \infty)$.
- Calculate $\langle\sigma_x\rangle_\beta$.
- Calculate $\langle\sigma_x\sigma_y\rangle_\beta$.

Definition 2.3 (Correlation functions). Consider $\Omega := \{\pm\}^V$. Then for any finite subset $A \subset V$, we define $\sigma_A : \Omega \to \{\pm\}$, $\sigma \mapsto \prod_{x\in A} \sigma_x$. Its expectation $\langle\sigma_A\rangle$ in any probability measure $\langle\,\cdot\,\rangle$ on Ω is called a *correlation function*. If $|A| = n$ then $\langle\sigma_A\rangle$ is also called an *n-point correlation function*.

Remark (Flip-symmetry). The Ising model is *flip-symmetric* in the sense that the distribution of the spins is invariant under the transformation $\sigma \mapsto -\sigma$. This is because the Hamiltonian is invariant under this transformation.

Exercise 2.4 (Flip-symmetry). Consider the Ising model on a finite graph $G = (V, E)$.
- Prove that if $A \subset V$ contains an odd number of vertices, then $\langle\sigma_A\rangle = 0$.
- Prove that if $A \subset V$ contains an odd number of vertices and $x \in V$, then
$$\mathbb{E}[\sigma_A|\{\sigma_x = +\}] = \langle\sigma_A\sigma_x\rangle.$$
- Prove that if $A \subset V$ contains an even number of vertices and $x \in V$, then
$$\mathbb{E}[\sigma_A|\{\sigma_x = +\}] = \langle\sigma_A\rangle.$$

In practice, we are interested in the Ising model on finite portions of the square lattice $\mathbb{Z}^d$ endowed with nearest-neighbour connectivity. We now provide the definitions for this setup.

Definition 2.5 (Free boundary conditions). Let $G = (V, E)$ denote a locally finite graph and $\Lambda \subset V$ a finite set. Write Λ^{f} for the subgraph of G induced by Λ. Write $\langle\,\cdot\,\rangle_{\Lambda,\beta}^{\text{f}} := \langle\,\cdot\,\rangle_{\Lambda^{\text{f}},\beta}$ for the *free-boundary Ising model* in Λ at inverse temperature $\beta \in [0, \infty)$.

Definition 2.6 (Fixed boundary conditions). Let $G = (V, E)$ denote a locally finite graph and $\Lambda \subset V$ a finite set. Let $\partial\Lambda \subset V \setminus \Lambda$ denote the set of vertices adjacent to Λ. Write $\bar\Lambda$ for the graph defined by
$$V(\bar\Lambda) := \Lambda \cup \partial\Lambda; \qquad E(\bar\Lambda) := \{\{x, y\} \in E : \{x, y\} \cap \Lambda \neq \emptyset\}.$$
For any $\zeta \in \{\pm\}^{\partial\Lambda}$, we shall write $\langle\,\cdot\,\rangle_{\Lambda,\beta}^{\zeta}$ for the measure
$$\langle\,\cdot\,\rangle_{\Lambda,\beta}^{\zeta} := \mathbb{E}_{\bar\Lambda,\beta}[\,\cdot\,|\{\sigma|_{\partial\Lambda} = \zeta\}].$$
This is called the *fixed-boundary Ising model* with boundary conditions ζ. The boundary condition $\zeta \equiv \pm$ is of particular interest, and it is denoted $\langle\,\cdot\,\rangle_{\Lambda,\beta}^{\pm}$.

Exercise 2.7 (Markov property). Consider the Ising model on some finite graph $G = (V, E)$ at inverse temperature β. Fix some $\Lambda \subset V$ and let $(\Lambda_i)_i$ denote the partition of Λ into connected components. Let $\zeta \in \{\pm\}^{\Lambda^c}$, and consider the conditional probability measure $\mathbb{P}[\,\cdot\,|\{\sigma|_{\Lambda^c} = \zeta\}]$.
- Prove that $(\sigma|_{\Lambda_i})_i$ is a family of independent random variables in this measure.
- Prove that the law of $\sigma|_{\Lambda_i}$ is $\langle\,\cdot\,\rangle_{\Lambda_i}^{\zeta|_{\partial\Lambda_i}}$.

Hint. Decompose the Hamiltonian according to $H(\sigma) = C + \sum_i H_i(\sigma)$, where each H_i is measurable in terms of $\zeta|_{\partial \Lambda_i}$ and $\sigma|_{\Lambda_i}$.

Ising proved that in one dimension, the Ising model exhibits exponential decay of correlations at all temperatures. In other words, there is no phase transition. We now state his result, without a proof. While the proof is quite straightforward even with elementary methods, its proof becomes entirely trivial after the introduction of more recent methods.

Theorem 2.8 (Ising, 1924)**.** *Consider the finite domains* $\Lambda_n := \{-n, \dots, n\}$ *of the graph* $\mathbb{Z}$. *Then for any* $\beta \in [0, \infty)$, *there exists a constant* $c = c_\beta > 0$ *such that*

$$\langle \sigma_0 \rangle^+_{\Lambda_n, \beta} \leq \frac{1}{c} e^{-c_\beta \cdot n}.$$

Unfortunately, Ising wrongly conjectured that the same would be true in higher dimension. Disappointed with this prediction, he left academia.

3. Peierls' argument

Peierls disproved Ising's conjecture for the absence of phase transition in dimension $d \geq 2$.

Theorem 3.1 (Peierls, 1936)**.** *Consider the finite domains* $\Lambda_n := \{-n, \dots, n\}^2$ *of the square lattice graph* $\mathbb{Z}^2$. *Then for sufficiently large* $\beta \in [0, \infty)$, *we have*

$$\inf_n \langle \sigma_{(0,0)} \rangle^+_{\Lambda_n, \beta} > 0.$$

Proof. Our objective is to prove that $\mathbb{P}^+_{\Lambda_n, \beta}[\{\sigma_{(0,0)} = -\}] \leq \frac{1}{3}$ for all n. Fix n.

Consider the set $\Omega' \subset \Omega$ of spin configurations on $\bar{\Lambda}_n$ which assign $+$ to $\partial \Lambda_n$. The two-dimensional square lattice graph $G = \mathbb{Z}^2$ is planar, and therefore we may consider its planar dual G^*. For any spin configuration $\sigma \in \Omega'$, we let $\mathcal{I}(\sigma) \subset E(G^*)$ denote its *interface*, that is, the set of dual edges separating two spins with a *distinct* value. Notice that:

- The map $\sigma \mapsto \mathcal{I}(\sigma)$ is injective,
- If $\sigma_{(0,0)} = -$, then $\mathcal{I}(\sigma)$ contains at least one self-avoiding loop around $(0,0)$.

In particular, inclusion of events yields

$$\mathbb{P}^+_{\Lambda_n, \beta}[\{\sigma_{(0,0)} = -\}] \leq \mathbb{P}^+_{\Lambda_n, \beta}[\{\mathcal{I}(\sigma) \text{ contains a loop } \gamma \text{ around } (0,0)\}]. \tag{2}$$

We would now like to make a competition between entropy and energy appear, as for the Curie-Weiss model. The entropy comes from the choice of the loop γ; the energy comes into play when upper bounding the probability that a particular loop belongs to $\mathcal{I}(\sigma)$. For large β, energy wins over entropy, yielding the desired bound. Let us start with the energy bound.

Claim. For any fixed loop γ, we may bound $\mathbb{P}^+_{\Lambda_n, \beta}[\{\gamma \subset \mathcal{I}(\sigma)\}] \leq e^{-2\beta |\gamma|}$.

Proof of the claim. We would like to define a *loop erasure map* $\mathcal{E} : \{\gamma \subset \mathcal{I}(\sigma)\} \to \Omega'$, which has the property that it removes the loop γ from the interface, that is,

$$\mathcal{I}(\mathcal{E}(\sigma)) = \mathcal{E}(\sigma) \setminus \gamma.$$

It is easy to realise such a map: we simply define $\mathcal{E}$ such that it flips the sign of every vertex of Λ_n which is surrounded by γ. Since $\mathcal{I}$ is injective, the map $\mathcal{E}$ is also injective, and we have

$$\mathbb{P}^+_{\Lambda_n, \beta}[\{\gamma \subset \mathcal{I}(\sigma)\}] = \frac{\sum_{\sigma \in \text{Domain}(\mathcal{E})} e^{-H(\sigma)}}{\sum_{\sigma \in \Omega'} e^{-H(\sigma)}} \leq \frac{\sum_{\sigma \in \text{Domain}(\mathcal{E})} e^{-H(\sigma)}}{\sum_{\sigma \in \text{Image}(\mathcal{E})} e^{-H(\sigma)}} = e^{-2\beta |\gamma|}.$$

The last equality is easy, since for any $\sigma \in \{\gamma \subset \mathcal{I}(\sigma)\}$, we have

$$H(\mathcal{E}(\sigma)) = H(\sigma) - 2\beta|\gamma|,$$

since $\mathcal{E}$ removes precisely $|\gamma|$ disagreement edges from the interface. This proves the claim.

We use the energy bound to prove another interesting intermediate result.

Claim (Exponential decay of the loop length)**.** For any dual edge e, we have

$$\mathbb{P}^+_{\Lambda_n,\beta}[\{\mathcal{I}(\sigma) \text{ contains a loop of length at least } \ell \text{ through } e\}] \leq (3e^{-2\beta})^\ell \frac{1}{1 - 3e^{-2\beta}}$$

whenever $3e^{-2\beta} < 1$.

Proof of the claim. Let $\mathcal{L}_k$ denote the set of self-avoiding loops through e of length k. Notice that $|\mathcal{L}_k| \leq 3^k$. A union bound yields

$$\mathbb{P}^+_{\Lambda_n,\beta}[\{\mathcal{I}(\sigma) \text{ contains a loop of length at least } \ell \text{ through } e\}]$$

$$\leq \sum_{k=\ell}^\infty \sum_{\gamma \in \mathcal{L}_k} \mathbb{P}^+_{\Lambda_n,\beta}[\{\gamma \subset \mathcal{I}(\sigma)\}] \leq \sum_{k=\ell}^\infty |\mathcal{L}_k| \cdot e^{-2\beta k} \leq \sum_{k=\ell}^\infty 3^k \cdot e^{-2\beta k}$$

$$= (3e^{-2\beta})^\ell \frac{1}{1 - 3e^{-2\beta}}.$$

This is the desired bound.

Return to Equation (2). If $\mathcal{I}(\sigma)$ contains a loop around $(0,0)$, then this loop must hit $(k - \frac{1}{2}, 0)$ for some $k \in \mathbb{Z}_{\geq 1}$, and this loop must have at least k steps. Thus, another union bound yields

$$\mathbb{P}^+_{\Lambda_n,\beta}[\{\sigma_{(0,0)} = -\}] \leq \sum_{k=1}^\infty (3e^{-2\beta})^k \frac{1}{1 - 3e^{-2\beta}} = (3e^{-2\beta})\frac{1}{(1 - 3e^{-2\beta})^2}.$$

The right hand side is smaller than $\frac{1}{3}$ when β is sufficiently large, independently of n. $\qquad\square$

Remark. Peierls' is robust, in the sense that it can be adapted to many other models in statistical mechanics.

Exercise 3.2 (The Peierls argument in higher dimensions)**.** Now consider the square lattice graph $\mathbb{Z}^d$ in dimension $d \geq 3$. What is the structure of the interface in this case? Can we adapt Peierls' to prove magnetisation for sufficiently large β?

4. The high-temperature expansion

The previous section proved the Peierls argument. An essential ingredient was to view the Ising model in two dimensions through the *interfaces* of the spins. Such a transformation of the model may be viewed as a rudimentary version of an *expansion*. The interface perspective is sometimes called the *low-temperature expansion* because it works well in the low-temperature regime (when β is large). There are several useful expansions for the Ising model; each one of them is adapted to a different setting. In this section we discuss another expansion: the *high-temperature expansion*. As the name suggests, this expansion is well-adapted to situations where β is small, even though we can also use it to prove useful results in other regimes. Appendix **??** contains an overview of the expansions discussed in these notes, and may serve as a reference.

For a streamlined presentation, we will henceforth present all our expansions for the Ising model on *finite graphs without boundary conditions*. This obviously includes the free boundary conditions. It is straightforward to see that fixed boundary conditions also fit into this framework, see Definition 4.5 and Lemma 4.6 below.

Consider the Ising model on a finite graph G. We are typically interested in the correlation functions, defined via

$$\langle \sigma_A \rangle = \frac{\sum_{\sigma \in \Omega} \sigma_A e^{-H(\sigma)}}{\sum_{\sigma \in \Omega} e^{-H(\sigma)}} = \frac{\sum_{\sigma \in \Omega} \sigma_A \prod_{xy \in E} e^{\beta \sigma_x \sigma_y}}{\sum_{\sigma \in \Omega} \prod_{xy \in E} e^{\beta \sigma_x \sigma_y}}.$$

An *expansion* of the Ising model involves rewriting the sum

$$\sum_{\sigma \in \Omega} \sigma_A \prod_{xy \in E} e^{\beta \sigma_x \sigma_y}.$$

A typical expansion comes down to rewriting the exponential, for example:
- We may write $e^{\beta \sigma_x \sigma_y} = \cosh \beta + \sigma_x \sigma_y \sinh \beta$,
- We may write $e^{\beta \sigma_x \sigma_y} = \sum_{k=0}^{\infty} (\beta \sigma_x \sigma_y)^k / k!$,
- We may write $e^{\beta \sigma_x \sigma_y} = e^{-2\beta} + 2 \cdot \mathbb{1}(\sigma_x = \sigma_y) \sinh \beta$.

Every expansion comes with its own advantages and disadvantages. The high-temperature expansion derives from the first identity.

Definition 4.1 (High-temperature expansion)**.** Consider the Ising model on a finite graph $G = (V, E)$ at inverse temperature β. We consider *percolation configurations* $\omega \in \{0, 1\}^E$; each ω is also regarded a (random) set of edges. We write $\partial \omega \subset V$ for the set of vertices having *odd* degree in the graph (V, ω).

The *high-temperature expansion* is the measure $\mathbf{M}_{G,\beta}$ on $\omega \in \{0, 1\}^E$ defined by

$$\mathbf{M}[\omega] := \mathbf{M}_{G,\beta}[\omega] := (\cosh \beta)^{|E \setminus \omega|} (\sinh \beta)^{|\omega|}.$$

Theorem 4.2 (High-temperature expansion for correlation functions)**.** *Consider the Ising model on a finite graph $G = (V, E)$. Then for any $A \subset V$, we have*

$$Z \langle \sigma_A \rangle = 2^{|V|} \mathbf{M}[\{\partial \omega = A\}].$$

In particular, $Z = 2^{|V|} \mathbf{M}[\{\partial \omega = \emptyset\}].$

Proof. We claim that

$$Z \langle \sigma_A \rangle = \sum_{\sigma \in \Omega} \sigma_A \prod_{xy \in E} e^{\beta \sigma_x \sigma_y} \tag{3}$$

$$= \sum_{\sigma \in \Omega} \sigma_A \prod_{xy \in E} (\cosh \beta + \sigma_x \sigma_y \sinh \beta) \tag{4}$$

$$= \sum_{\sigma \in \Omega} \sigma_A \sum_{\omega \in \{0,1\}^E} \prod_{xy \in E} (\cosh \beta)^{1 - \omega_{xy}} (\sigma_x \sigma_y \sinh \beta)^{\omega_{xy}} \tag{5}$$

$$= \sum_{\omega \in \{0,1\}^E} (\cosh \beta)^{|E \setminus \omega|} (\sinh \beta)^{|\omega|} \sum_{\sigma \in \Omega} \sigma_A \sigma_{\partial \omega} \tag{6}$$

$$= \sum_{\omega \in \{0,1\}^E} (\cosh \beta)^{|E \setminus \omega|} (\sinh \beta)^{|\omega|} 2^{|V|} \mathbb{1}(A = \partial \omega) \tag{7}$$

$$= 2^{|V|} \mathbf{M}[\{\partial \omega = A\}].$$

Equations (3) and (4) come down to definitions and the identity for $e^{\beta \sigma_x \sigma_y}$. Swapping the sum and the product yields Equation (5). Equation (6) is a rearrangement of the terms, noting that $\prod_{xy} (\sigma_x \sigma_y)^{\omega_{xy}} = \sigma_{\partial \omega}$. Equation (7) is obtained by resolving the sum over σ. The final equation is the definition of $\mathbf{M}$. $\qquad \square$

We can use this theorem to state our first important *correlation inequality*.

Theorem 4.3 (First Griffiths inequality)**.** *Consider the Ising model on a finite graph $G = (V, E)$. Then for any $A \subset V$, we have $\langle \sigma_A \rangle \geq 0$.*

Proof. The previous theorem yields a nonnegative number for $Z\langle\sigma_A\rangle$. $\qquad\square$

Theorem 4.4 (Exponential decay at high temperature)**.** *Consider the Ising model with $+$ boundary conditions on the graph $\mathbb{Z}^d$ for $d \in \mathbb{Z}_{\geq 1}$. Then for any $\beta \in [0, \infty)$ such that $(2d-1)\tanh\beta < 1$, there exists a constant $c = c_{d,\beta} > 0$ such that*

$$\langle\sigma_x\rangle^+_{\Lambda,\beta} \leq \tfrac{1}{c}e^{-c\operatorname{Distance}(x,\Lambda^c)}$$

for any $x \in \mathbb{Z}^d$ and any domain $\Lambda \subset \mathbb{Z}^d$.

We would like to use the high-temperature expansion, but for this we must first write $\langle\,\cdot\,\rangle^+_\Lambda$ as an Ising model on a finite graph without boundary condition.

Definition 4.5 (Ghost vertex)**.** Let $G = (V, E)$ denote a locally finite graph, and $\Lambda \subset V$ a finite domain. We already defined the graphs Λ^{f} and $\bar{\Lambda}$. Now define the graph $\Lambda^{\mathfrak{g}}$ as follows: it is obtained from the graph $\bar{\Lambda}$ by replacing all vertices in $\partial\Lambda$ by a single distinguished vertex $\mathfrak{g}$, called the *ghost vertex*. Its vertex set is given by $V(\Lambda^{\mathfrak{g}}) := \Lambda \cup \{\mathfrak{g}\}$, and there is a natural bijection from $E(\bar{\Lambda})$ to $E(\Lambda^{\mathfrak{g}})$.

Notice that $\Lambda^{\mathfrak{g}}$ is a multigraph when some $x \in \Lambda$ is connected to multiple vertices in $\partial\Lambda$ in the graph $\bar{\Lambda}$, but this does not really affect our setup.

It is easy to see that the following lemma holds true.

Lemma 4.6. *Let $G = (V, E)$ denote a locally finite graph, and $\Lambda \subset V$ a finite domain. Then the distribution of $\sigma|_\Lambda$ is the same in the following two measures:*

$$\langle\,\cdot\,\rangle^+_\Lambda \qquad and \qquad \mathbb{E}_{\Lambda^{\mathfrak{g}}}[\,\cdot\,|\sigma_{\mathfrak{g}} = +].$$

Correlation functions can thus be expressed in terms of correlation functions on finite graphs via Exercise 2.4.

Proof overview of Theorem 4.4. We have

$$\langle\sigma_x\rangle^+_\Lambda = \langle\sigma_x\sigma_{\mathfrak{g}}\rangle_{\Lambda^{\mathfrak{g}}} = \frac{\mathbf{M}_{\Lambda^{\mathfrak{g}}}[\{\partial\omega = \{x, \mathfrak{g}\}\}]}{\mathbf{M}_{\Lambda^{\mathfrak{g}}}[\{\partial\omega = \emptyset\}]}.$$

If $\partial\omega = \{x, \mathfrak{g}\}$, then ω contains a self-avoiding walk γ from x to $\mathfrak{g}$. A union bound yields

$$\langle\sigma_x\rangle^+_\Lambda \leq \sum_\gamma \frac{\mathbf{M}_{\Lambda^{\mathfrak{g}}}[\{\partial\omega = \{x, \mathfrak{g}\}\} \cap \{\gamma \subset \omega\}]}{\mathbf{M}_{\Lambda^{\mathfrak{g}}}[\{\partial\omega = \emptyset\}]}.$$

The proof is now completed after performing the two steps of the Peierls argument:
- One bounds each term by $(\tanh\beta)^{|\gamma|}$,
- One bounds the number of walks γ of length n from x by $2d(2d-1)^{|\gamma|-1}$.

$\qquad\square$

Exercise 4.7. Fill in the details of the previous proof overview.

Let us summarise what we have proved so far.
- Theorem 4.4 implies that there is exponential decay of correlations when β is sufficiently small.
- In dimension $d = 1$, Theorem 4.4 also implies Ising's result (Theorem 2.8), since there was no requirement on β when $d = 1$.
- In dimension $d \geq 2$, we proved that there is *magnetisation* via the Peierls argument (Theorem 3.1). Thus, in dimension $d \geq 2$, there must be a phase transition, and we aim to investigate further.

The high-temperature expansion is typically used to find upper bounds on correlation functions. However, it is also possible to use it to find lower bounds. Let $G = (V, E)$ denote a finite graph. For any fixed set $Q \subset E$ of edges, we define the *XOR map*

$$\Xi_Q : \{0,1\}^E \to \{0,1\}^E, \ \omega \mapsto \omega\Delta Q,$$

where Δ denotes the symmetric difference of two sets. This map is an involution. Moreover, for any $A \subset V$, it restricts to a bijection

$$\Xi_Q : \{\partial\omega = A\} \to \{\partial\omega = A\Delta\partial Q\}. \tag{8}$$

The measure $\mathbf{M}$ is not invariant under the involution Ξ_Q, but it is easy to see how the map affects the measure. More precisely, for any $\eta \in \{0,1\}^E$, we have

$$\mathbf{M}[\{\omega = \Xi_Q(\eta)\}] = (\tanh\beta)^{|Q\setminus\eta| - |Q\cap\eta|} \cdot \mathbf{M}[\{\omega = \eta\}].$$

The prefactor is upper bounded by $(\tanh\beta)^{-|Q|}$. Thus, writing Ξ_Q^* for the pushforward map, we obtain

$$\Xi_Q^*\mathbf{M} \leq (\tanh\beta)^{-|Q|} \cdot \mathbf{M}.$$

For example, using the bijection in Equation (8), we obtain

$$\mathbf{M}[\{\partial\omega = A\}] \leq (\tanh\beta)^{-|Q|} \cdot \mathbf{M}[\{\partial\omega = A\Delta\partial Q\}]. \tag{9}$$

We have now proved the following result.

Lemma 4.8. *Consider the Ising model on a finite graph $G = (V, E)$. Then for any $A \subset V$ and any $Q \subset E$, we have*

$$\langle\sigma_A\sigma_{\partial Q}\rangle \geq (\tanh\beta)^{|Q|} \cdot \langle\sigma_A\rangle.$$

In particular,

$$\langle\sigma_x\sigma_y\rangle \geq (\tanh\beta)^{\mathrm{Distance}(x,y)}.$$

Proof. The first inequality is Equation (9). For the second inequality, simply set $A = \emptyset$ and let Q denote a shortest path from x to y. $\qquad\square$

This lemma complements Theorem 4.4 at high temperature, as the lemma asserts that the correlation functions cannot decay *faster* than exponentially at any temperature (except when $\beta = 0$).

5. THE RANDOM-CURRENTS EXPANSION

Previous sections explained the existence of a phase transition in dimension $d \geq 2$, by demonstrating that the qualitative behaviour of the model is different at low and high temperature. This section introduces a new representation of the Ising model which is adapted to studying the Ising model at and around the critical temperature.

Definition 5.1 (Currents). Let $G = (V, E)$ denote a graph. A *current* is a map $\mathbf{n} : E \to \mathbb{Z}_{\geq 0}$. We think of $(V, \mathbf{n})$ as a multigraph, where for each edge $uv \in E$ we have $\mathbf{n}_{uv}$ multi-edges between u and v. The set of *sources* $\partial\mathbf{n} \subset V$ of a current $\mathbf{n}$ is defined as the set of vertices $u \in V$ with an odd degree in the multigraph. We let $\hat{\mathbf{n}} := (\mathbf{n} \wedge 1) \in \{0,1\}^E$ denote the associated percolation, which simply contains the edges carrying at least one current.

If G is finite and $\beta \in [0, \infty)$, then the *weight* of a current is defined as

$$w(\mathbf{n}) := w_{G,\beta}(\mathbf{n}) := \prod_{xy \in E} \frac{\beta^{\mathbf{n}_{xy}}}{\mathbf{n}_{xy}!}.$$

The *random-currents measure* is the measure $\mathbb{M}_{G,\beta}$ on $(\mathbb{Z}_{\geq 0})^E$ defined by

$$\mathbb{M}[\mathbf{n}] := \mathbb{M}_{G,\beta}[\mathbf{n}] := w_{G,\beta}(\mathbf{n}).$$

Remark 5.2. Notice that $e^{-\beta|E|}\mathbb{M}_{G,\beta}$ is a probability measure in which $(\mathbf{n}_{xy})_{xy \in E}$ is a family of independent random variables with distribution Poisson(β).

Currents can be used to encode correlation functions of the Ising model.

Theorem 5.3 (Current representation of correlation functions)**.** *Consider the Ising model on a finite graph G at inverse temperature β. Let $A \subset V$ be a subset of vertices. Then*

$$Z\langle \sigma_A \rangle = 2^{|V|} \sum_{\mathbf{n}: \partial \mathbf{n} = A} w(\mathbf{n}) = 2^{|V|} \mathbb{M}[\{\partial \mathbf{n} = A\}].$$

In particular, the partition function is given by

$$Z = 2^{|V|} \sum_{\mathbf{n}: \partial \mathbf{n} = \emptyset} w(\mathbf{n}) = 2^{|V|} \mathbb{M}[\{\partial \mathbf{n} = \emptyset\}].$$

Proof. The numbers $\cosh \beta$ and $\sinh \beta$ are obtained from the expansion $\sum_{\mathbf{n}=0}^{\infty} \beta^{\mathbf{n}}/\mathbf{n}!$ of e^{β} by keeping only the even and odd terms respectively. The result then simply follows from the high-temperature representation (Theorem 4.2) by expanding $\cosh \beta$ and $\sinh \beta$. □

Remark 5.4 (Switching example)**.** The above setup opens the door to a powerfull technique called *switching*. Two elements are key to switching: Poisson randomness, and parity constraints (both of which are present in the currents representation of correlation functions). Let us quickly describe how switching may be applied to a simple example. Suppose that we record cars traversing a bridge on a road. Blue cars pass according to a Poisson point process with rate 1 (per second). Let X_B denote the number of blue cars that pass after recording β seconds. Can we easily prove, without a calculation, that

$$\mathbb{P}[\{X_B \text{ is even}\}] \geq \mathbb{P}[\{X_B \text{ is odd}\}]?$$

Suppose that there are also yellow cars, which arrive according to an independent Poisson process with the same rate. Let X_Y denote the number of yellow cars that passed. Suppose that, after waiting β seconds, $X_B + X_Y = N > 0$ cars passed. What is the *conditional* probability that X_B is even?

Well, we must have $\mathbb{P}[\{X_B \text{ is even}\}|\{X_B + X_Y = N\}] = 1/2$. Indeed, by the properties Poisson point processes, the n-th car (for $1 \leq n \leq N$) is blue or yellow with equal probability, independently of the other cars. Thus, we may condition on the colours of the first $N - 1$ cars, then flip a fair coin to decide the colour of the last car and thus the parity of X_B. Equivalently, we observe that *repainting* or *switching* the colour of the last car leaves the conditional distribution invariant.

But we cannot always do the switch. If $N = X_B + X_Y = 0$ then there is no car to repaint, and also $X_B = 0$. Thus, we conclude that

$$\mathbb{P}[\{X_B \text{ is even}\}] - \mathbb{P}[\{X_B \text{ is odd}\}] = \mathbb{P}[\{X_B + X_Y = 0\}] \geq 0.$$

Notice that we originally asked a question about blue cars, but introducing yellow cars allowed us to answer it. This is the essence of the switching lemma.

Lemma 5.5 (Explicit switching lemma)**.** *Let G denote a finite graph and $\beta \in [0, \infty)$. Consider the measurable pair $(\mathbf{n}, \mathbf{m}) \sim \mathbb{M}^2 = \mathbb{M}^2_{G,\beta}$. Fix $\mathbf{s} \in (\mathbb{Z}_{\geq 0})^E$ and $\eta \subset \hat{\mathbf{s}}$. Then for any $A \subset V$, we have*

$$\mathbb{M}^2[\{\mathbf{n} + \mathbf{m} = \mathbf{s}\} \cap \{\partial \mathbf{n} = A\}] = \mathbb{M}^2[\{\mathbf{n} + \mathbf{m} = \mathbf{s}\} \cap \{\partial \mathbf{n} = A \Delta \partial \eta\}].$$

Notice also that if $\mathbf{n} + \mathbf{m} = \mathbf{s}$, then $\partial \mathbf{s} = (\partial \mathbf{n}) \Delta (\partial \mathbf{m})$.

Slight variations of the proof will be used later.

Proof. Introduce the probability measure $\mathbb{P} :\propto \mathbb{M}^2[\{\mathbf{n} + \mathbf{m} = \mathbf{s}\} \cap (\,\cdot\,)]$. Our objective is to prove that

$$\mathbb{P}[\{\partial \mathbf{n} = A\}] = \mathbb{P}[\{\partial \mathbf{n} = A \Delta \partial \eta\}].$$

In the probability measure $\mathbb{P}$, the family $((\mathbf{n}_{xy}, \mathbf{m}_{xy}))_{xy \in E}$ is independent over the edges $xy \in E$, and each pair $(\mathbf{n}_{xy}, \mathbf{m}_{xy})$ follows the distrubution of two independent distributions Poisson(β) conditioned to sum to $\mathbf{s}_{xy}$. By analogy with the example of blue and yellow cars, we may interpret $\mathbb{P}$ as follows:

- $\mathcal{M} = \mathcal{M}_{\mathbf{s}}$ is the fixed multigraph $\{(xy, k) \in E \times \mathbb{Z}_{\geq 0} : k < \mathbf{s}_{xy}\}$,
- $\mathcal{A}$ is a uniformly random subset of $\mathcal{M}$,
- $\mathcal{B}$ is the complement of $\mathcal{A}$ in $\mathcal{M}$,
- $\mathbf{n}_{xy} = \mathbf{n}_{xy}(\mathcal{A})$ is the number of multi-edges in $\mathcal{A}$ between x and y,
- $\mathbf{m}_{xy} = \mathbf{m}_{xy}(\mathcal{B})$ is the number of multi-edges in $\mathcal{B}$ between x and y.

Define $\tilde{\eta} := \eta \times \{0\} \subset \mathcal{M}$. Since $\mathcal{A}$ is a uniformly random subset of $\mathcal{M}$ in the measure $\mathbb{P}$, the set $\mathcal{A} \Delta \tilde{\eta}$ is also uniformly random in $\mathcal{M}$. Therefore we have

$$\mathbb{P}[\{\partial \mathbf{n} = A\}] = \mathbb{P}[\{\partial \mathbf{n}(\mathcal{A}) = A\}] = \mathbb{P}[\{\partial \mathbf{n}(\mathcal{A} \Delta \tilde{\eta}) = A\}] = \mathbb{P}[\{\partial \mathbf{n} = A \Delta \partial \eta\}].$$

This is the desired equality. $\qquad\square$

Definition 5.6 (Percolation of currents). Let G denote a graph and $\mathbf{n}$ a current. Write

$$\{u \overset{\hat{\mathbf{n}}}{\longleftrightarrow} v\}$$

for the event there is an open path from u to v (u and v may represent vertices or sets of vertices). For fixed $S \subset V$, we shall also write $\mathcal{E}_S$ for the set

$$\{O \subset E : |C \cap S| \text{ is even for any connected component } C \subset V \text{ of } (V, O)\}.$$

Exercise 5.7. Let G denote a graph and $x, y \in V$ distinct vertices. Prove the following.

- If $S = \{x, y\}$, then $\{\hat{\mathbf{n}} \in \mathcal{E}_S\} = \{x \overset{\hat{\mathbf{n}}}{\longleftrightarrow} y\}$.
- If $\omega \in \mathcal{E}_S$, then we may find a subset $\eta \subset \omega$ with $\partial \eta = S$.
- If G is a finite graph and $\partial \mathbf{n} = S$, then $\hat{\mathbf{n}} \in \mathcal{E}_S$.
- For any $S \subset V$, the event $\{\hat{\mathbf{n}} \in \mathcal{E}_S\}$ is an increasing event of the current $\mathbf{n}$.

Lemma 5.8 (Switching lemma). *Let G denote a finite graph and $\beta \in [0, \infty)$. Consider the measurable pair $(\mathbf{n}, \mathbf{m}) \sim \mathbb{M}^2 = \mathbb{M}^2_{G,\beta}$. Then for any $A, B, S \subset V$ and for any bounded function $F : (\mathbb{Z}_{\geq 0})^E \to \mathbb{C}$, we have*

$$\mathbb{M}^2[F(\mathbf{n} + \mathbf{m})\mathbb{1}(\widehat{\mathbf{n} + \mathbf{m}} \in \mathcal{E}_S)\mathbb{1}(\partial \mathbf{n} = A)\mathbb{1}(\partial \mathbf{m} = B)]$$
$$= \mathbb{M}^2[F(\mathbf{n} + \mathbf{m})\mathbb{1}(\widehat{\mathbf{n} + \mathbf{m}} \in \mathcal{E}_S)\mathbb{1}(\partial \mathbf{n} = A\Delta S)\mathbb{1}(\partial \mathbf{m} = B\Delta S)].$$

Proof. By linearity of integration, it suffices to consider the case that $F(\mathbf{n} + \mathbf{m}) := \mathbb{1}(\mathbf{n} + \mathbf{m} = \mathbf{s})$ for some fixed current $\mathbf{s}$ with $\hat{\mathbf{s}} \in \mathcal{E}_S$. By the previous exercise, we may find some $\eta \subset \hat{\mathbf{s}}$ such that $\partial \eta = S$. The desired equality

$$\mathbb{M}^2[\mathbb{1}(\mathbf{n} + \mathbf{m} = \mathbf{s})\mathbb{1}(\partial \mathbf{n} = A)\mathbb{1}(\partial \mathbf{m} = B)]$$
$$= \mathbb{M}^2[\mathbb{1}(\mathbf{n} + \mathbf{m} = \mathbf{s})\mathbb{1}(\partial \mathbf{n} = A\Delta S)\mathbb{1}(\partial \mathbf{m} = B\Delta S)]$$

then follows by the explicit switching lemma. $\qquad\square$

In practice, we do not care so much about the function F, and simply set it to $F \equiv 1$. An important corollary of the switching lemma is the *second Griffiths inequality*.

Lemma 5.9 (Second Griffiths inequality). *Consider the Ising model on a finite graph G at inverse temperature β. Then for any $A, B \subset V$, we have $\langle \sigma_A \sigma_B \rangle - \langle \sigma_A \rangle \langle \sigma_B \rangle \geq 0$.*

The second Griffiths inequality is more subtle than the first, as it bounds a *difference* of correlation functions. This is typical for the switching lemma.

Proof. Claim that

$$\mathbb{M}^2[\{\partial \mathbf{n} = A\} \cap \{\partial \mathbf{m} = B\}]$$
$$= \mathbb{M}^2[\{\widehat{\mathbf{n} + \mathbf{m}} \in \mathcal{E}_B\} \cap \{\partial \mathbf{n} = A\} \cap \{\partial \mathbf{m} = B\}]$$
$$\overset{\text{switch}}{=} \mathbb{M}^2[\{\widehat{\mathbf{n} + \mathbf{m}} \in \mathcal{E}_B\} \cap \{\partial \mathbf{n} = A\Delta B\} \cap \{\partial \mathbf{m} = \emptyset\}]$$
$$\leq \mathbb{M}^2[\{\partial \mathbf{n} = A\Delta B\} \cap \{\partial \mathbf{m} = \emptyset\}].$$

The switch is just the switching lemma with $F \equiv 1$ and $S = B$. For the first equality, we simply observe that $\{\partial \mathbf{m} = B\} \subset \{\widehat{\mathbf{n} + \mathbf{m}} \in \mathcal{E}_B\}$ (see the exercise above). The inequality is inclusion of events.

By the random currents expansion of correlation functions (Theorem 5.3), the left- and rightmost expressions are given by

$$Z^2 \langle \sigma_A \rangle \langle \sigma_B \rangle \leq Z^2 \langle \sigma_A \sigma_B \rangle \langle \sigma_\emptyset \rangle.$$

Since $\langle \sigma_\emptyset \rangle = 1$, this is the desired inequality. $\qquad\square$

Exercise 5.10 (The two-point function as a metric)**.** Consider the Ising model on a finite graph G at inverse temperature $\beta > 0$. Prove that $V \times V \to [0, \infty]$, $(u, v) \mapsto -\log \langle \sigma_u \sigma_v \rangle_{G,\beta}$ defines a metric on V. Use directly the switching lemma, and not the second Griffiths inequality. What does the percolation event $\mathcal{E}_S$ look like in this case?

Definition 5.11 (Probability measures on currents)**.** Consider the Ising model on a finite graph G at inverse temperature β. For any $A \subset V$, define the probability measure $\mathbb{P}^A_{G,\beta}$ by

$$\mathbb{P}^A := \mathbb{P}^A_{G,\beta} := \frac{2^{|V|}}{Z_{G,\beta} \langle \sigma_A \rangle_{G,\beta}} \mathbb{M}_{G,\beta}[\mathbb{1}(\partial \mathbf{n} = A)(\,\cdot\,)].$$

For any $A_1, \ldots, A_n$, write $\mathbb{P}^{A_1,\ldots,A_n} := \mathbb{P}^{A_1} \times \cdots \times \mathbb{P}^{A_n}$.

Exercise 5.12 (Correlation functions in terms of sourceless currents)**.** Consider the Ising model on a finite graph G at inverse temperature β. Prove that for any $A \subset V$,

$$\langle \sigma_A \rangle^2 = \mathbb{P}^{\emptyset,\emptyset}[\{\widehat{\mathbf{n} + \mathbf{m}} \in \mathcal{E}_A\}].$$

Observe that we can now express all correlation functions in terms of a single fixed probability measure on sourceless random currents.

6. MONOTONICITY VIA THE SECOND GRIFFITHS INEQUALITY

Theorem 6.1 (Monotonicity in the temperature)**.** *Let G denote a finite graph and $A \subset V$ a finite set. Then the function $\beta \mapsto \langle \sigma_A \rangle_{G,\beta}$ is non-decreasing.*

Proof. We want to prove that

$$\frac{\partial}{\partial \beta} \langle \sigma_A \rangle_{G,\beta} = \frac{\partial}{\partial \beta} \left(\frac{\sum_\sigma \sigma_A \prod_{uv} e^{\beta \sigma_u \sigma_v}}{\sum_\sigma \prod_{uv} e^{\beta \sigma_u \sigma_v}} \right) \geq 0.$$

Since we are differentiating a fraction, it suffices to show that the numerator grows at a faster rate than the denominator, that is,

$$\frac{\frac{\partial}{\partial \beta} \sum_\sigma \sigma_A \prod_{uv} e^{\beta \sigma_u \sigma_v}}{Z \langle \sigma_A \rangle} \geq \frac{\frac{\partial}{\partial \beta} \sum_\sigma \prod_{uv} e^{\beta \sigma_u \sigma_v}}{Z}.$$

We perform the differential and then multiply each side by $\langle \sigma_A \rangle$, to see that this inequality is equivalent to

$$\sum_{xy} \frac{\sum_\sigma \sigma_x \sigma_y \sigma_A \prod_{uv} e^{\beta \sigma_u \sigma_v}}{Z} \geq \langle \sigma_A \rangle \sum_{xy} \frac{\sum_\sigma \sigma_x \sigma_y \prod_{uv} e^{\beta \sigma_u \sigma_v}}{Z}.$$

Each fraction may now be reinterpreted as a correlation function, so that the previous inequality is equivalent to

$$\sum_{xy} \langle \sigma_x \sigma_y \sigma_A \rangle \geq \langle \sigma_A \rangle \sum_{xy} \langle \sigma_x \sigma_y \rangle.$$

But this is just the second Griffiths inequality. $\qquad\square$

Exercise 6.2 (Regularity properties of the correlation functions in β)**.** Prove that the function $[0, \infty) \to \mathbb{R}$, $\beta \mapsto \langle \sigma_A \rangle_{G,\beta}$ in the above context is an analytic function.

Next, we want to prove monotonicity in domains. We first challenge the reader to prove the following exercise.

Exercise 6.3 (Conditioning on equality increases the correlation functions). Consider the Ising model on a finite graph G at inverse temperature β, and fix some subset $A \subset V$.

- Prove that for any two distinct vertices $u, v \in V$, we have

$$\mathbb{E}_{G,\beta}[\sigma_A | \{\sigma_u = \sigma_v\}] \geq \mathbb{E}_{G,\beta}[\sigma_A] = \langle \sigma_A \rangle_{G,\beta}.$$

- Prove for any $X \subset Y \subset V$, we have

$$\mathbb{E}_{G,\beta}[\sigma_A | \{\sigma \text{ is constant on } X\}] \leq \mathbb{E}_{G,\beta}[\sigma_A | \{\sigma \text{ is constant on } Y\}].$$

Lemma 6.4 (Monotonicity in domains). *Consider the Ising model on a locally finite graph $G = (V, E)$ at inverse temperature β. Consider two finite domains $\Lambda \subset \Lambda' \subset V$ and a subset $A \subset \Lambda$.*

- ***Free boundary.*** *We have $\langle \sigma_A \rangle^{\mathrm{f}}_{\Lambda,\beta} \leq \langle \sigma_A \rangle^{\mathrm{f}}_{\Lambda',\beta}$.*
- ***Wired boundary.*** *We have $\langle \sigma_A \rangle^{+}_{\Lambda,\beta} \geq \langle \sigma_A \rangle^{+}_{\Lambda',\beta}$.*

Proof for $\langle \cdot \rangle^{\mathrm{f}}$. We first prove the following claim: if G' and G'' are finite graphs on the same vertex set, and such that $E(G'') = E(G') \cup \{xy\}$, then

$$\langle \sigma_A \rangle_{G',\beta} \leq \langle \sigma_A \rangle_{G'',\beta}$$

for any $A \subset V(G')$. To prove the claim, we simply expand

$$\langle \sigma_A \rangle_{G'',\beta} = \frac{\langle e^{\beta \sigma_x \sigma_y} \sigma_A \rangle_{G'}}{\langle e^{\beta \sigma_x \sigma_y} \rangle_{G'}}.$$

Thus, we want to show that

$$\langle e^{\beta \sigma_x \sigma_y} \sigma_A \rangle_{G'} \geq \langle e^{\beta \sigma_x \sigma_y} \rangle_{G'} \langle \sigma_A \rangle_{G'}.$$

This follows from the second Griffiths inequality. We have now proved the claim.

Recall the definition of the finite graph Λ^{f}. Let $\tilde{\Lambda}^{\mathrm{f}} := ((\Lambda')^{\mathrm{f}}, E(\Lambda^{\mathrm{f}}))$; this is just the graph Λ^{f} supplemented with some isolated vertices $\Lambda' \setminus \Lambda$. The law of σ in $\langle \cdot \rangle_{\tilde{\Lambda}^{\mathrm{f}}}$ is just given by $\langle \cdot \rangle_{\Lambda^{\mathrm{f}}}$, with independent fair coin flips for the isolated vertices in $\Lambda' \setminus \Lambda$. Thus, it suffices to prove that

$$\langle \sigma_A \rangle^{\mathrm{f}}_{\Lambda} = \langle \sigma_A \rangle_{\tilde{\Lambda}^{\mathrm{f}}} \leq \langle \sigma_A \rangle_{(\Lambda')^{\mathrm{f}}} = \langle \sigma_A \rangle^{\mathrm{f}}_{\Lambda'}.$$

This follows from the claim. $\square$

Proof for $\langle \cdot \rangle^{+}$. Without loss of generality, $\Lambda' \setminus \Lambda = \{u\}$ for some vertex $u \in V$. We make all calculations in the graph $(\Lambda')^{\mathfrak{g}}$ with the ghost vertex: we get

$$\langle \sigma_A \rangle^{+}_{\Lambda'} = \mathbb{E}_{(\Lambda')^{\mathfrak{g}}}[\sigma_A | \{\sigma_{\mathfrak{g}} = +\}]; \qquad \langle \sigma_A \rangle^{+}_{\Lambda} = \mathbb{E}_{(\Lambda')^{\mathfrak{g}}}[\sigma_A | \{\sigma_{\mathfrak{g}} = +\} \cap \{\sigma_u = \sigma_{\mathfrak{g}}\}].$$

Assume first that $|A|$ is even for now. Then

$$\langle \sigma_A \rangle^{+}_{\Lambda} = \mathbb{E}_{(\Lambda')^{\mathfrak{g}}}[\sigma_A | \{\sigma_u = \sigma_{\mathfrak{g}}\}] \geq \mathbb{E}_{(\Lambda')^{\mathfrak{g}}}[\sigma_A] = \langle \sigma_A \rangle^{+}_{\Lambda'},$$

due to Exercise 6.3.

If $|A|$ is odd, we just need to replace the set A by $A' := A \cup \{\mathfrak{g}\}$. More precisely,

$$\langle \sigma_A \rangle^{+}_{\Lambda} = \mathbb{E}_{(\Lambda')^{\mathfrak{g}}}[\sigma_{A'} | \{\sigma_u = \sigma_{\mathfrak{g}}\}] \geq \mathbb{E}_{(\Lambda')^{\mathfrak{g}}}[\sigma_{A'}] = \langle \sigma_A \rangle^{+}_{\Lambda'},$$

where the inequality uses the same exercise.

Those are the desired inequalities. $\square$

Definition 6.5 (Infinite-volume limit). Let $G = (V, E)$ denote a locally finite graph. Write

$$\lim_{\Lambda \uparrow V} f(\Lambda) \qquad \text{for} \qquad \lim_{n \to \infty} f(\Lambda_n),$$

where $(\Lambda_n)_n$ is any increasing sequence of finite domains with $\cup_n \Lambda_n = V$. This notation makes sense only when the limit is independent of the precise choice of the sequence $(\Lambda_n)_n$, and is called the *thermodynamical limit* or *infinite-volume limit*.

Let $(\Omega, \mathcal{F})$ denote the measurable space $\Omega := \{\pm 1\}^V$ endowed with the product σ-algebra. For a domain Λ, we write $\mathcal{F}_\Lambda$ for the σ-algebra generated by spins in Λ. An observable $X : \Omega \to \mathbb{C}$ is called *local* if it is measurable with respect to $\mathcal{F}_\Lambda$ for some domain Λ.

Let $\mathcal{P}(\Omega, \mathcal{F})$ denote the set of all probability measures on this measurable space. We endow this set with the *local convergence topology*, which is defined as the topology making the map

$$\mathcal{P}(\Omega, \mathcal{F}) \to \mathbb{C}, \, \mu \mapsto \mu[X]$$

continuous for any local observable X.

Remark 6.6. This topology is sometimes known under different names in the literature (such as the *weak topology*). I like the name *local convergence topology* because it captures the essence quite literally: if the statistics of the measures within a fixed domain Λ converge, then we have local convergence.

Exercise 6.7. Prove that $\mathcal{P}(\Omega, \mathcal{F})$ is a compact space in this topology.

Theorem 6.8 (Existence of the thermodynamical limit). *Consider the Ising model on a locally finite graph G at inverse temperature β. Then there exists unique probability measures $\langle \cdot \rangle_{G,\beta}^{\mathrm{f}}, \langle \cdot \rangle_{G,\beta}^{+} \in \mathcal{P}(\Omega, \mathcal{F})$ such that*

$$\lim_{\Lambda \uparrow V} \langle X \rangle_{\Lambda, \beta}^{*} = \langle X \rangle_{G, \beta}^{*}$$

for $* \in \{\mathrm{f}, +\}$ and for any local observable $X : \Omega \to \mathbb{R}$. In other words,

$$\lim_{\Lambda \uparrow V} \langle \cdot \rangle_{\Lambda, \beta}^{*} =: \langle \cdot \rangle_{G, \beta}^{*}.$$

The measures $\langle \cdot \rangle_{G,\beta}^{*}$ are called the thermodynamical limits *or* infinite-volume limits.

Proof. Any local observable may be written as a finite linear conbination of observables of the form σ_A where A is a finite subset of V. The theorem then follows by compactness and Lemma **??**. $\square$

Exercise 6.9 (Continuity properties in β). Consider the Ising model on a locally finite graph $G = (V, E)$. Fix $A \subset V$ finite.

- The function $\beta \mapsto \langle \sigma_A \rangle_{G,\beta}^{*}$ is non-decreasing for $* \in \{\mathrm{f}, +\}$.
- The function $\beta \mapsto \langle \sigma_A \rangle_{G,\beta}^{\mathrm{f}}$ is left continuous.
- The function $\beta \mapsto \langle \sigma_A \rangle_{G,\beta}^{+}$ is right continuous.

Hint. Argue that $\beta \mapsto \langle \sigma_A \rangle_{G,\beta}^{\mathrm{f}}$ is a limit of a non-decreasing sequence of non-decreasing functions.

Definition 6.10 (Magnetisation and critical temperature). Let G be a vertex-transitive locally finite graph and u some distinguished reference vertex. The non-decreasing right-continuous function

$$m = m_G : [0, \infty) \to \mathbb{R}, \, \beta \mapsto \langle \sigma_u \rangle_{G, \beta}^{+}$$

is called the *magnetisation*.

The *critical (inverse) temperature* is defined via

$$\beta_c := \beta_c(G) := \inf\{\beta \in [0, \infty) : m(\beta) > 0\}.$$

We have already proved that $\beta_c \in (0, \infty)$ for $G = \mathbb{Z}^d$ in dimension $d \geq 2$, and that $\beta_c = \infty$ for $G = \mathbb{Z}$.

7. More correlation inequalities via switching

Our next goal is to prove the following theorem.

Theorem 7.1 (+ and − boundary conditions coincide when the magnetisation vanishes)**.** *Let G denote a connected locally finite graph, endowed with some reference vertex u. Then*

$$\langle \cdot \rangle_{G,\beta}^+ = \langle \cdot \rangle_{G,\beta}^- \qquad \Longleftrightarrow \qquad m_G(\beta) = 0.$$

The theorem can be proved using the following bound.

Exercise 7.2 (Pairing bound, difficult)**.** Consider the Ising model on a finite graph G at inverse temperature β. Let $A \subset V$ denote any finite subset, and fix $u \in A$. Use the switching lemma to prove that

$$\langle \sigma_A \rangle \leq \sum_{v \in A \setminus \{u\}} \langle \sigma_u \sigma_v \rangle \langle \sigma_{A \setminus \{u,v\}} \rangle.$$

Hint: argue that

$$\mathbb{1}(\mathbf{n} \in \mathcal{E}_A) \leq \sum_{v \in A \setminus \{u\}} \mathbb{1}(\mathbf{n} \in \mathcal{E}_{\{u,v\}}) \mathbb{1}(\mathbf{n} \in \mathcal{E}_A).$$

Conclude that

$$\langle \sigma_A \rangle \leq \sum_{\pi} \prod_{\{u,v\} \in \pi} \langle \sigma_u \sigma_v \rangle$$

where π ranges over the *pairings* of A, that is, the set of partitions of A in which each member has two elements.

Proof of Theorem 7.1. Notice that $\langle \cdot \rangle_{G,\beta}^+$ and $\langle \cdot \rangle_{G,\beta}^-$ are related by a global spin flip (the pushforward map corresponding to $\sigma \mapsto -\sigma$). Therefore all of the following are equivalent:

- $\langle \cdot \rangle_{G,\beta}^+ = \langle \cdot \rangle_{G,\beta}^-$,
- $\langle \cdot \rangle_{G,\beta}^+$ is invariant under the map $\sigma \mapsto -\sigma$,
- $\langle \sigma_A \rangle_{G,\beta}^+ = 0$ whenever $A \subset V$ has odd cardinal.

The implication "$\Longrightarrow$" is now obvious, and we focus on "$\Longleftarrow$". Suppose that $m(\beta) = 0$, that is, $\langle \sigma_u \rangle^+ = 0$ where u is the reference vertex. Fix $A \subset V$ with $|A|$ odd. It suffices to prove that $\langle \sigma_A \rangle^+ = 0$. We shall in fact give *two* proofs of this fact. In both proofs, we shall fix a sequence $(\Lambda_n)_n$ of increasing subsets of V with $\cup_n \Lambda_n = V$.

- *Proof 1, using the pairing bound.* For fixed n, we get

$$\langle \sigma_A \rangle_{\Lambda_n}^+ = \langle \sigma_{A \cup \{\Lambda_n^c\}} \rangle_{G_n'} \leq \sum_{v \in A} \langle \sigma_{\{v, \Lambda_n^c\}} \rangle_{G_n'} \langle \sigma_{A \setminus \{v\}} \rangle_{G_n'} = \sum_{v \in A} \langle \sigma_v \rangle_{\Lambda_n}^+ \langle \sigma_{A \setminus \{v\}} \rangle_{\Lambda_n}^+$$

$$\leq \sum_{v \in A} \langle \sigma_v \rangle_{\Lambda_n}^+ \to_{n \to \infty} 0.$$

  The first inequality is the pairing bound, the second the generic bound $\langle \sigma_{A \setminus \{v\}} \rangle_{\Lambda_n}^+ \in [0, 1]$, and the convergence follows from Exercise **??**.

- *Proof 2, using directly the high-temperature expansion.* We only consider $\beta > 0$, otherwise the spins are independent fair coin flips, and the result is automatic. Recall Exercise **??**. For fixed n, we get

$$(\tanh \beta)^{d_{\text{Transport}}(A, \{u\})} \cdot \langle \sigma_A \rangle_{\Lambda_n}^+ \leq \langle \sigma_u \rangle_{\Lambda_n}^+.$$

  On the left, the transport distance is calculated in the graph G_n' (the dependence on n is implicit). As $n \to \infty$, this transport distance stabilises at the finite transport distance in the infinite graph G. Since the right hand side tends to zero with n,

we know that the left hand side also tends to zero. Since the prefactor remains uniformly positive, we must have $\langle \sigma_A \rangle^+ = 0$. $\square$

Add: $\varphi_\beta(S)$ argument, Messager–Miracle-Solé

8. Mixing properties of the infinite-volume limit

In this section, we consider the Ising model on the infinite square lattice $\mathbb{Z}^d$.

Definition 8.1. Consider a measure $\langle \cdot \rangle \in \mathcal{P}(\Omega, \mathcal{F})$.

- For any $u \in \mathbb{Z}^d$, we define the *shift operator* $\tau_u : \Omega \to \Omega$ by

$$(\tau_u \sigma)_x = \sigma_{x-u}.$$

  An event A is *shift-invariant* if $\tau_u A := \{\tau_u \sigma : \sigma \in A\}$ for any $u \in \mathbb{Z}^d$.
- The measure is called *shift-invariant* if

$$\langle X \circ \tau_u \rangle = \langle X \rangle$$

  for any vertex $u \in \mathbb{Z}^d$ and for any bounded local observable X.
- The measure is called *mixing* if

$$\lim_{\|u\|_2 \to \infty} \left(\langle X(Y \circ \tau_u) \rangle - \langle X \rangle \langle Y \circ \tau_u \rangle \right) = 0$$

  for any bounded local observables X and Y.
- The measure is called *ergodic* if it is shift-invariant and

$$\langle \mathbb{1}_A \rangle \in \{0, 1\}$$

  for any shift-invariant event $A \in \mathcal{F}$.

Lemma 8.2 (Mixing implies ergodicity). *If a shift-invariant measure is mixing, then it is ergodic.*

Proof. Let $\langle \cdot \rangle$ denote a shift-invariant measure that is mixing, but not ergodic. We aim to derive a contradiction. Let A denote a shift-invariant event with $p := \langle \mathbb{1}_A \rangle \in (0, 1)$. We shall derive a contradiction by constructing two events which are both extremely correlated with A (using ergodicity), while also being almost independent (using mixing).

Fix $\varepsilon > 0$. By the martingale convergence theorem, there exists a finite domain $\Lambda \subset \mathbb{Z}^d$ and an $\mathcal{F}_\Lambda$-measurable event A_Λ such that $\langle \mathbb{1}_{A \Delta A_\Lambda} \rangle < \varepsilon$. Write $p' := \langle \mathbb{1}_{A_\Lambda} \rangle$; notice that $|p' - p| < \varepsilon$. Define $A_{\Lambda+u} := \tau_u A_\Lambda \in \mathcal{F}_{\Lambda+u}$. We claim that there exists some $u \in \mathbb{Z}^d$ such that:

$$\langle \mathbb{1}_{A_\Lambda \Delta A_{\Lambda+u}} \rangle < 2\varepsilon \qquad \text{and} \qquad \langle \mathbb{1}_{A_\Lambda \Delta A_{\Lambda+u}} \rangle \geq 2p'(1-p') - \varepsilon.$$

This yields the desired contradiction when ε is small enough.

The inequality on the left is easy to obtain: for any $u \in \mathbb{Z}^d$, the event $A_{\Lambda+u}$ also satisfies $\langle \mathbb{1}_{A \Delta A_{\Lambda+u}} \rangle = \langle \mathbb{1}_{A \Delta A_\Lambda} \rangle < \varepsilon$. By the triangular inequality, we have $\langle \mathbb{1}_{A_\Lambda \Delta A_{\Lambda+u}} \rangle < 2\varepsilon$.

For the inequality on the right, we use mixing: we get

$$\langle \mathbb{1}_{A_\Lambda \Delta A_{\Lambda+u}} \rangle = \langle \mathbb{1}_{A_\Lambda} + \mathbb{1}_{A_{\Lambda+u}} - 2\mathbb{1}_{A_\Lambda}\mathbb{1}_{A_{\Lambda+u}} \rangle \to_{\|u\|_2 \to \infty} p' + p' - 2p'p' = 2p'(1-p').$$

We then simply choose u such that $\|u\|_2$ is sufficiently large. $\square$

Theorem 8.3. *Consider the Ising model on $\mathbb{Z}^d$ at inverse temperature β.*

- **Wired boundary.** *The measure $\langle \cdot \rangle^+_{\mathbb{Z}^d, \beta}$ is mixing.*
- **Free boundary.** *The measure $\langle \cdot \rangle^f_{\mathbb{Z}^d, \beta}$ is mixing if $\lim_{\|y\|_2 \to \infty} \langle \sigma_x \sigma_y \rangle^f_{\mathbb{Z}^d, \beta} = 0$.*

Proof for $\langle \cdot \rangle^+$. It suffices to prove that for any finite sets $A, B \subset \mathbb{Z}^d$, we have

$$\lim_{\|u\|_2 \to \infty} \langle \sigma_A \sigma_{B+u} \rangle^+_{\mathbb{Z}^d, \beta} = \langle \sigma_A \rangle^+_{\mathbb{Z}^d, \beta} \langle \sigma_B \rangle^+_{\mathbb{Z}^d, \beta}.$$

By the second Griffiths inequality, we get

$$\langle \sigma_A \sigma_{B+u} \rangle^+_{\mathbb{Z}^d, \beta} \geq \langle \sigma_A \rangle^+_{\mathbb{Z}^d, \beta} \langle \sigma_B \rangle^+_{\mathbb{Z}^d, \beta}.$$

It suffices to prove the other inequality. Consider an extremely large finite domain $\Lambda \supset A \cup B$. Define $\Lambda_u := \Lambda \cup (\Lambda + u)$. If $\|u\|_2$ is sufficiently large, then Λ is not connected to Λ_u in the set Λ_u, and so we get

$$\limsup_{\|u\|_2 \to \infty} \langle \sigma_A \sigma_{B+u} \rangle^+_{\mathbb{Z}^d, \beta} \leq \limsup_{\|u\|_2 \to \infty} \langle \sigma_A \sigma_{B+u} \rangle^+_{\Lambda_u, \beta} = \langle \sigma_A \rangle^+_{\Lambda, \beta} \langle \sigma_B \rangle^+_{\Lambda, \beta}.$$

Sending $\Lambda \uparrow \mathbb{Z}^d$ yields the desired inequality. $\qquad\square$

Proof for $\langle \cdot \rangle^{\mathrm{f}}$. As for $+$ boundary conditions, it suffices to fix two finite sets $A, B \subset \mathbb{Z}^d$, and prove that

$$\limsup_{\|u\|_2 \to \infty} \langle \sigma_A \sigma_{B+u} \rangle^{\mathrm{f}}_{\mathbb{Z}^d, \beta} \leq \langle \sigma_A \rangle^{\mathrm{f}}_{\mathbb{Z}^d, \beta} \langle \sigma_B \rangle^{\mathrm{f}}_{\mathbb{Z}^d, \beta}.$$

Fix u, and consider an extremely large finite domain $\Lambda \supset A \cup (B + u)$. By the switching lemma, it is straightforward to deduce that

$$\langle \sigma_A \sigma_{B+u} \rangle^{\mathrm{f}}_{\Lambda, \beta} \leq \langle \sigma_A \rangle^{\mathrm{f}}_{\Lambda, \beta} \langle \sigma_{B+u} \rangle^{\mathrm{f}}_{\Lambda, \beta} + \sum_{x \in A, \, y \in B} \langle \sigma_x \sigma_{y+u} \rangle^{\mathrm{f}}_{\Lambda, \beta}.$$

By sending $\Lambda \uparrow \mathbb{Z}^d$, we get

$$\langle \sigma_A \sigma_{B+u} \rangle^{\mathrm{f}}_{\mathbb{Z}^d, \beta} \leq \langle \sigma_A \rangle^{\mathrm{f}}_{\mathbb{Z}^d, \beta} \langle \sigma_B \rangle^{\mathrm{f}}_{\mathbb{Z}^d, \beta} + \sum_{x \in A, \, y \in B} \langle \sigma_x \sigma_{y+u} \rangle^{\mathrm{f}}_{\mathbb{Z}^d, \beta}.$$

The error term vanishes by our additional assumption in the statement of the theorem. $\quad\square$

REFERENCES

CNRS and Sorbonne Université, LPSM
Email address: `piet.lammers@cnrs.fr`