

## chap3 InFinite Volume Measures

The goal of this chapter is to construct probability measures for the Ising model on  $\mathbb{Z}^d$  ( $d \geq 1$ ), that is, probabilities on

$\Omega = \{\pm 1\}^{\mathbb{Z}^d}$ . There are two approaches:

- Construct a "limit in distribution" not really because the sets of definitions are different  
of our measures  $\mu_{\lambda_n, \beta}^+$ . As  $n \rightarrow \infty$ , do we get a proba  
 $\mu_\beta^+$ ? Does it depend on boundary conditions?

For instance, for  $\Theta$  b.c., do we have  $\mu^- = \mu^+$ ?

- Find an "axiomatic description" for what we want the measure to be, and show that they can be satisfied (the "DLR" conditions, or a "Gibbs measure")

Dobrushin, Lanford, Ruelle

In fact we will prove:

will be constructed from  $\mu_{\lambda_n, \beta}^{+/-}$ .

Theo  $\forall \beta > 0$ , there exists probabilities  $\mu^+, \mu^-$  on  $\Omega$  such that:

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Field
- $\mu^+, \mu^-$  are Gibbs measures (satisfy DLR)
  - $\mu^+, \mu^-$  are translation-invariant, ergodic, positively associated (FKG)
  - If  $\nu$  is a Gibbs measure on  $\Omega$ ,
- $$\mu^- \leq_{st} \nu \leq_{st} \mu^+$$

Moreover, the following are equivalent:

- (i)  $\mu^+ = \mu^-$
- (ii)  $\exists!$  Gibbs measure
- (iii)  $\langle \sigma_0 \rangle^+ = \langle \sigma_0 \rangle^-$ .

(All of that will be defined)

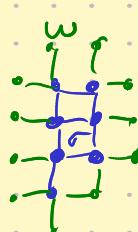
## I- Construction

### A) Generic boundary conditions

Recall that for  $\Lambda \subset \mathbb{Z}^d$ , we had  $\partial\Lambda = \bar{\Lambda} - \Lambda$ ,  $\bar{\Lambda} = \Lambda \cup \partial\Lambda$ ,  $E = \{(x,y) / x,y \in \Lambda \text{ or } x \in \Lambda, y \in \partial\Lambda\}$ ,  $\Omega = \{\pm 1\}^\Lambda$ .

Let also  $\tilde{E} = \{(x,y) / x,y \in \Lambda\}$ .

Let  $w \in \{-1, 0, +1\}^{d^2}$ ,  $\beta > 0$ ,  $h \in \mathbb{R}$ . We define



$$\forall \sigma \in \Omega, \mu_{\Lambda, \beta, h}^w(\sigma) = \frac{1}{Z_{\Lambda, \beta, h}^w} \exp \left( \beta \sum_{\{(x,y) \in \tilde{E}\}} \sigma_x \sigma_y + \beta \sum_{\{(x,y) \in E\}} \sigma_x w_y + h \sum_{x \in \Lambda} \sigma_x \right)$$

For  $w \equiv +1$ , we recover  $\mu_{\Lambda, \beta, h}^+$ .

For  $w \equiv -1$ , we denote  $\mu_{\Lambda, \beta, h}^- := \mu_{\Lambda, \beta, h}^w$ .

Finally, for  $w \equiv 0$ , we get "free boundary conditions" denoted  $\mu_{\Lambda, \beta, h}^f$ .

### B) Topology

Now let  $\Lambda_n = [-n, n]^d$ ,  $\Omega_n = \{\pm 1\}^{\Lambda_n}$ ,  
and  $\Omega = \{\pm 1\}^{\mathbb{Z}^d}$ .

We want to construct measures on  $\Omega$  (from those on  $\Omega_n$ ). This usually requires a bit of topology, and an existence argument like Kolmogorov's extension theorem, or Riesz theorem.

We endow  $\Omega$  with the product topology. By Tychonoff,  $\Omega$  is compact. Moreover, this topology is consistent with a metric such as  $d(\sigma, \sigma') = \frac{1}{n}$  where  $n = \sup \{k \geq 1 / \sigma = \sigma' \text{ on } \Lambda_k\}$ .

We also equip it with the product  $\sigma$ -algebra, which is generated by local events (or cylinders):

$$\{\sigma \in \Omega / \sigma|_{\Lambda_n} \in A\} \text{ where } A \in \Omega_n, n \geq 1.$$

A function  $F: \Omega \rightarrow \mathbb{R}$  is said to be a local function

If there exists  $n \geq 1$  s.t.  $f(\tau)$  only depends on  $\tau|_{\tau_n}$ .

Let  $\text{Loc}(\Sigma)$  be the space of local functions.

Prop: The closure of  $\text{Loc}(\Sigma)$  for the  $\|\cdot\|_\infty$  norm contains  $C_c(\Sigma)$  (the continuous functions on  $\Sigma$ ).  $f > 0 \Rightarrow \phi(f) > 0$

Cor: If  $\phi: \text{Loc}(\Sigma) \rightarrow \mathbb{R}$  is a continuous, positive linear form, then it can be extended into a continuous positive linear form on  $C_c(\Sigma)$ .

Riesz  $\rightsquigarrow$  Then, there exists a unique measure  $\mu$  on  $\Sigma$  st  $\forall f \in C_c(\Sigma), \phi(f) = \int f d\mu$ .

proof of Prop: Let  $f \in C_c(\Sigma)$ . As  $\Sigma$  is compact (Tychonoff),  $f$  is uniformly continuous.

Let  $\varepsilon > 0$ , take  $\delta$  s.t.  $d(\sigma, \sigma') \leq \delta \Rightarrow |f(\sigma) - f(\sigma')| \leq \varepsilon$ .

Let  $\sigma = \left\lceil \frac{1}{\delta} \right\rceil$ . Let  $g(\sigma) = f(\sigma_n)$  where  $\sigma_n = \begin{cases} \sigma & \text{on } \bar{\tau}_n \\ +1 & \text{on } \mathbb{C} \setminus \bar{\tau}_n \end{cases}$ .

Then  $\forall \sigma \in \Sigma, d(\sigma, \sigma_n) \leq \frac{1}{n}$  by def.  $\leq \delta$

so  $|f(\sigma) - g(\sigma)| \leq \varepsilon$ , so  $\|f - g\|_\infty \leq \varepsilon$ .  $\square$

Proof of Cor: The first part is clear.

For the second part, see Riesz Theo:

A continuous linear form on  $C_c(\Sigma)$  corresponds to a unique signed measure  $\mu$   $\phi(f) = \int f d\mu$ .

Here,  $C_c(\Sigma) = C(\Sigma)$  as  $\Sigma$  is compact, and  $\mu \geq 0$  by positivity of  $\phi$ .  $\square$

Remark: we want to prove that  $\mu$  is a proba, that is  $\phi(1) = 1$ .

9) Definition of  $\mu^+, \mu^-$  via limits

Remark: For  $f \in \text{Loc}(\Sigma)$ , its expectation  $\mu_{\text{min}, p, h}^{+/-}(f)$  is well-defined for  $n$  large enough.  $\leftarrow$  Check!

$\forall \beta > 0, h \in \mathbb{R}$ ,

Theo There exists probability measures  $\mu_{\rho_h}^+, \mu_{\rho_h}^-$  on  $\mathbb{R}$  characterized by

$$\forall f \in \text{Loc}(\mathbb{R}), \quad \mu_{\lambda_n, \beta, h}^+(f) \xrightarrow{n \rightarrow \infty} \mu_{\beta, h}^+(f)$$

$$\mu_{\lambda_n, \beta, h}^-(f) \xrightarrow{n \rightarrow \infty} \mu_{\beta, h}^-(f).$$

These measures don't depend on the choice of  $(\lambda_n)_{n \geq 0}$  (with  $(\lambda_n) \nearrow$  and  $\bigcup_n \lambda_n = \mathbb{Z}^d$ )

Proof: We prove this for  $h=0$ .

For  $h \neq 0$ , we don't have Griffith's ineq (which we used for monotonicity). Prove the Theo using FKG ineq (next chapter) mimicking the proof.

Let  $f \in \text{Loc}(\mathbb{R})$ . We claim that there exists  $\pi \in \mathbb{N}^\mathbb{N}$  s.t.  $f$  can be written as

$$f = \sum_{A \in \lambda_n} \hat{f}_A \circ \sigma_A \quad \text{recall: } \prod_{x \in A} \sigma_x.$$

Indeed, equip the space of functions on  $\{\pm 1\}^{\mathbb{Z}^d}$  with the scalar product  $(f, g) := \frac{1}{|\lambda_n|} \sum_{\sigma \in \{\pm 1\}^{\mathbb{Z}^d}} f(\sigma)g(\sigma)$ , then it

is easy to see that the  $(\sigma_A)_{A \in \lambda_n}$  are orthonormal, and there is  $2^{|\lambda_n|}$  of them, so it is a basis.

$$\text{So } \mu_{\lambda_n}^+(f) = \sum_{A \in \lambda_n} \hat{f}_A \langle \sigma_A \rangle_{\lambda_n}^+$$

We know that  $\langle \sigma_A \rangle_{\lambda_n}^+$  is  $\downarrow$  in  $n$  (of prev. chapter, monotonicity in domain)

so  $\lim_n \mu_{\lambda_n}^+(f)$  exists, we denote it by  $\phi(f)$ .

$$\text{Then } |\phi(f)| \leq \lim_n |\mu_{\lambda_n}^+(f)| \leq \|f\|_\infty$$

So  $\phi$  is a continuous linear form on  $\text{Loc}(\mathbb{R})$ , and clearly  $\geq 0$ . By section B), it is represented by a unique measure  $\mu^+$ . Moreover,  $\phi(1)=1$  so  $\mu^+$  is a proba.

The independance on  $\lambda_n$ , and adaptation to  $\mu^-$ , are left as exercises.

Discrete Fourier transform.

## II - Positive association (FKG)

### A) Stochastic domination

A Function  $f: \Omega \rightarrow \mathbb{R}$  is said to be **increasing** if either  $\{\pm 1\}^n$  or  $\{\pm 1\}^{\mathbb{Z}^d}$ , both work!

$$\sigma \leq \sigma' \Rightarrow f(\sigma) \leq f(\sigma')$$

$\forall \sigma, \sigma' \leq \sigma'$

we should say "nondecreasing" but it's too long.

Example:  $\tau \mapsto \tau_x$  is increasing.

- $\Delta \mapsto \Delta_A$  for  $|A| \geq 2$  is not!

- $\tau \mapsto \prod_{x \in A, \tau_x = +1} 1$  is increasing. Note that it is  $\prod_{x \in A} \frac{1+\tau_x}{2}$ .

Let  $\mu, \nu$  be two probabilities on  $\Omega$ . We say that

$\mu$  is **stochastically dominated by**  $\nu$ ,

denoted  $\mu \leq_{st} \nu$ , if  $\forall f$  increasing measurable bounded,

(" $\mu$  tends to have more  $\oplus$  spins than  $\nu$ .)

$$\int f d\mu \leq \int f d\nu.$$

Example: (Site Bernoulli percolation).

Let  $p \in [0, 1]$ , and  $\text{Rad}_p = p \delta_1 + (1-p) \delta_{-1}$ ,

then For  $p \leq p'$ ,  $\text{Rad}_p^{\otimes n} \leq_{st} \text{Rad}_{p'}^{\otimes n}$ .

Indeed, let  $(U_n)_{n \in \mathbb{N}}$  be iid  $\text{Unif}([0, 1])$ , then

$$X: \Lambda \rightarrow \{-1, +1\}$$

$$x \mapsto \begin{cases} 1 & \text{if } U_x \leq p \\ -1 & \text{if } U_x > p \end{cases}$$

$$Y: \Lambda \rightarrow \{-1, +1\}$$

$$x \mapsto \begin{cases} 1 & \text{if } U_x \leq p' \\ -1 & \text{if } U_x > p' \end{cases}$$

We have  $X \sim \mu = \text{Rad}_p^{\otimes n}$ ,  $Y \sim \nu = \text{Rad}_{p'}^{\otimes n}$ , and  $X \leq Y$  a.s.

So For  $f \nearrow$ ,  $f(X) \leq f(Y)$  a.s. so  $E[f(X)] \leq E[f(Y)]$

$$\int f d\mu \leq \int f d\nu$$

The method of the example (coupling) is in fact generic:

Theo: Let  $\mu, \nu$  be probabilities on  $\mathcal{S}$ . Here  $\mathcal{S} = \{\pm 1\}^n$  or  $\{\pm 1\}^{\mathbb{Z}^d}$   
but works for  $\mathcal{S}$  Polish  
ordered space.

The following are equivalent:

$$(i) \mu \leq_{st} \nu$$

(ii) There exists random variables  $X, Y$  defined on the same proba. space such that

$$X \sim \mu, \quad Y \sim \nu \quad \text{and} \quad X \leq Y \text{ a.s.}$$

(Not proved, see Werner's "Percolation et Modèle d'Ising" for  $\mathcal{S}$  finite, Lindvall '99 for generic case)

### B) Glauber dynamics

For the moment, we take  $\Lambda \subset \mathbb{Z}^d$ , and  $\mathcal{S} = \{\pm 1\}^\Lambda$ .

Also,  $\beta > 0, h \in \mathbb{R}, w \in \{\pm 1\}^{\partial \Lambda}$ .

Q: How does one simulate the measure  $\mu_{\Lambda, \beta, h}^w$ ?

(a priori we have  $2^{|\Lambda|}$  configs to consider, so it may be impossible to list them all and compute their proba!)

Also  $Z$  itself requires all configs to compute, so we can't even know the proba of a single config  $\sigma$ !)

Idea: Create a Markov chain on  $\mathcal{S}$  whose invariant distri is  $\mu_{\Lambda, \beta, h}^w$ . ↓ denoted  $\mu$  later on.

Starting from a config  $\sigma \in \mathcal{S}$ , select a  $x \in \Lambda$  uniformly at random. Let  $\sigma^{+x}$  be the config  $\sigma$  with a +1 at  $x$ ,

$$\sigma^{-x} \xrightarrow{\quad} \sigma \xrightarrow{\quad} -1 \xrightarrow{\quad}$$

easy to compute:  
 $Z$  and most of  
(the exponential)  
cancel out!

Then  $\frac{\mu(\sigma^{+x})}{\mu(\sigma^{+x}) + \mu(\sigma^{-x})}$  is the conditional proba of

seeing +1 at  $x$  conditionnally on  $\sigma|_{\Lambda \setminus \{x\}}$

So we set  $\sigma_x$  to be +1 with proba  $\frac{\mu(\sigma^{+x})}{\mu(\sigma^{+x}) + \mu(\sigma^{-x})}$

and to be -1

$$\xrightarrow{\quad}$$

$\frac{\mu(\sigma^{-x})}{\mu(\sigma^{+x}) + \mu(\sigma^{-x})}$

(regardless of the previous  $\sigma_x$ ).

In other words, we consider the Markov Chain on  $\mathcal{S}$  with transition matrix

$\forall \sigma \in \mathcal{S}$ ,

$$\forall x \in \mathbb{N}, \quad P(\sigma, \sigma^{\pm x}) = \frac{1}{|N|} \frac{\mu(\sigma^{\pm x})}{\mu(\sigma^+ x) + \mu(\sigma^- x)}$$

Prop: This MC (is irreducible, recurrent  $\Rightarrow$  aperiodic) has (unique) invariant distri  $\mu$ .

Proof: Clearly it is irred; since  $\mathcal{S}$  is finite it is rec.  $\Rightarrow$   $P(\sigma, \sigma) > 0$  so aper.

Moreover, it is reversible for  $\mu$ !

Let  $\sigma, \sigma' \in \mathcal{S}$ . Then either  $P(\sigma, \sigma') = 0$  (then  $P(\sigma', \sigma) = 0$ ) or  $\exists n \in \mathbb{N} / \sigma' = \sigma^{\pm x}$  and then

$$\begin{aligned} \mu(\sigma) P(\sigma, \sigma') &= \frac{1}{|N|} \frac{\mu(\sigma) \mu(\sigma')}{\mu(\sigma^+ x) + \mu(\sigma^- x)} \\ &= \frac{1}{|N|} \frac{\mu(\sigma') \mu(\sigma)}{\mu(\sigma'^+ x) + \mu(\sigma'^- x)} \\ &= \mu(\sigma') P(\sigma', \sigma). \end{aligned}$$



This may not look like much, but we deduce a "local criterion" for stochastic domination:

Prop let  $\mu, \nu$  be probabilities on  $\mathcal{S} = \{\pm 1\}^N$ , such that

for any  $\sigma \leq \sigma' \in \mathcal{S}$  and  $x \in \mathbb{N}$ ,

"Holley criterion"

$$\frac{\mu(\sigma^+ x)}{\mu(\sigma^- x)} \leq \frac{\nu(\sigma'^+ x)}{\nu(\sigma'^- x)}$$

$$\left( \Leftrightarrow \frac{\mu(\sigma^+ x)}{\mu(\sigma^+ x) + \mu(\sigma^- x)} \leq \frac{\nu(\sigma'^+ x)}{\nu(\sigma'^+ x) + \nu(\sigma'^- x)} \right)$$

by direct check.

Then  $\mu \leq_{st} \nu$ .

Prop Let  $(X_n)_{n \geq 0}$ ,  $(Y_n)_{n \geq 0}$  be two Markov chains on  $\mathcal{S}$ , with  $X_n$  having the previous transition matrix for  $\mu$ ,  $Y_n$

They are coupled s.t. at each step, we take the same

site  $x \in A$ , and whenever we set it to  $\uparrow$  in  $X_n$ , we also set it to  $\uparrow$  in  $Y_n$ . By the assumption, this is possible (cf direct check).

Then starting with  $X_0 \leq Y_0$ , clearly,

$$\text{a.s., } X_n \leq Y_n.$$

So if  $f$  is  $\uparrow$  and bounded,

$$\mathbb{E}[f(X_n)] \leq \mathbb{E}[f(Y_n)]$$

$$\int f d\mu \quad (\text{since } X_n \xrightarrow{d} \mu)$$

$$\int f d\nu \quad (Y_n \xrightarrow{d} \mu)$$



## 9 Finite volume FKG (Fortuin, Kasteleyn, Ginibre; 1971)

Theo For any  $1 \in \mathbb{Z}^d$ ,  $p > 0$ ,  $h \in \mathbb{R}$ ,  $w \in \{\pm 1\}^{\partial 1}$ , the measure  $\mu_{1,p,h}^w$  satisfies the **FKG inequality**:

"positive association"

$\forall f, g : \Omega \rightarrow \mathbb{R}$  increasing

$$\langle f \cdot g \rangle_{1,p,h}^w \geq \langle f \rangle_{1,p,h}^w \langle g \rangle_{1,p,h}^w.$$

For instance, take  $f = 1_A$ ,  $g = 1_B$ , where  $A, B \subset \Omega$  are

increasing events ( $\sigma \in A$  and  $\sigma \leq \sigma' \Rightarrow \sigma' \in A$ ). Then

$$\langle 1_A 1_B \rangle \geq \langle 1_A \rangle \langle 1_B \rangle : A, B \text{ are positively correlated.}$$

Proof: Let  $f, g$  be  $\uparrow$ . Adding a constant to  $g$  does not change the inequality, so we suppose  $g \geq 0$  ( $\forall \sigma \in \Omega, g(\sigma) \geq 0$ ).

$$\text{Let } \mu = \mu_{1,p,h}^w, \text{ and } \nu(g) = \frac{g(\sigma)}{\langle g \rangle_{1,p,h}^w} \mu(\sigma).$$

One checks that  $\nu$  is also a proba on  $\Omega$ .

We will show  $\mu \leq_{st} \nu$ .

$$\text{This gives the result because } \int f d\mu \leq \int f d\lambda = \frac{1}{\langle g \rangle} \sum_{x \in \Lambda} f(x) g(x) \mu(x)$$

$$= \frac{\langle fg \rangle}{\langle g \rangle}.$$

We use the property: it is enough to show that for  $\sigma \leq \sigma'$  and  $x \in \Lambda$ ,  $\frac{\mu(\sigma^{+x})}{\mu(\sigma^{-x})} \geq \frac{\nu(\sigma'^{+x})}{\nu(\sigma'^{-x})}$ .

$$\text{But } \frac{\nu(\sigma^{+x})}{\nu(\sigma'^{-x})} = \frac{\nu(\sigma^{+x}) \mu(\sigma'^{-x})}{\nu(\sigma'^{-x}) \mu(\sigma'^{-x})} \geq \frac{\mu(\sigma^{+x})}{\mu(\sigma'^{-x})}$$

So it is enough to show that  $\sigma \mapsto \frac{\mu(\sigma^{+x})}{\mu(\sigma^{-x})}$  is  $\mathcal{P}$  int ('then apply at  $\sigma \leq \sigma'$ )

$$\text{But it is } \sigma \mapsto \exp \left( \beta \sum_{y \sim x} ((+1)\sigma_y - (-1)\sigma_y) + 2h \right)$$

$$= \exp(+2\beta \sum_{y \sim x} \sigma_y + 2h), \text{ clearly } \mathcal{P} \text{ int.}$$



Remark: We see that if  $\mu$  is a measure on  $\mathbb{Z}^d$  s.t.

$$\forall \sigma \leq \sigma', \frac{\mu(\sigma^{+x})}{\mu(\sigma^{-x})} \leq \frac{\mu(\sigma'^{+x})}{\mu(\sigma'^{-x})}, \text{ then } \mu \text{ satisfies FKG.}$$

$$\text{In fact it is enough to have } \forall \sigma, \forall x, y, \frac{\mu(\sigma^{+x-y})}{\mu(\sigma^{-x-y})} \leq \frac{\mu(\sigma^{+x+y})}{\mu(\sigma^{-x+y})}$$

also "Holley"

Application: For  $n_A = \prod_{x \in A} \frac{1+\sigma_x}{2}$ , we get

$$\langle n_A, n_B \rangle_{1, \beta, h}^\omega \geq \langle n_A \rangle_{1, \beta, h}^\omega \langle n_B \rangle_{1, \beta, h}^\omega$$

Exercise Construct  $\mu_{\beta, h}^{+/-}$  as limits using this inequality,  
for  $h \in \mathbb{R}$ .

For + b.c. (or "free" b.c., ie no constraint on  $\sigma_0$ )

Remark: Griffiths:  $\langle \sigma_A, \sigma_B \rangle_{1, \beta}^+ > \langle \sigma_A \rangle_{1, \beta}^+ \langle \sigma_B \rangle_{1, \beta}^+$   
not  $\mathcal{P}$ !  
any b.c.

$h=0$  (works for  $h \neq 0$ )

FKG:  $\langle n_A, n_B \rangle_{1, \beta, h}^\omega \geq \langle n_A \rangle_{1, \beta, h}^\omega \langle n_B \rangle_{1, \beta, h}^\omega$   
 $\approx$  functions.  
 $h \in \mathbb{R}$

Prop: Let  $\Lambda \subset \mathbb{Z}^d$ ,  $\beta > 0$ . Let  $h \leq h'$  and  $\omega \leq \omega' \in \{\pm 1\}^{\partial \Lambda}$ .

Then  $\mu_{1, \beta, h}^\omega \leq \mu_{1, \beta, h'}^{\omega'}$ . Monotonicity (stochastic order).

Proof: Exercise (use Holley criterion). in b.c. and in  $h$ .

## D) Infinite volume FKG

Cor: The measures  $\mu_{\beta, h}^{+/-}$  on  $\Omega = \{\pm 1\}^{\mathbb{Z}^d}$  satisfy the FKG inequality:

$\forall f, g : \mathcal{L} \rightarrow \mathbb{R}$  increasing and in  $\text{Loc}(\mathcal{L})$  or  $\mathcal{C}(\mathcal{L})$ ,

$$\langle f g \rangle_{\beta, h}^{+/-} \geq \langle f \rangle_{\beta, h}^{+/-} \langle g \rangle_{\beta, h}^{+/-}.$$

Proof: We just put together finite FKG and limits ...  $\square$

## III - Gibbs measures

### A) Domain Markov property



Prop Let  $\Delta \subset \Lambda \subset \mathbb{Z}^d$ ,  $\beta > 0$  and  $h \in \mathbb{R}$ .

Let  $w \in \{\pm 1\}^{\partial \Lambda}$  and  $w' \in \{\pm 1\}^{\partial \Delta}$  be compatible:

And  $\omega|_{\partial \Delta} = w|_{\partial \Delta}$ .

Then  $\forall \eta \in \{\pm 1\}^\Delta$ ,

$$\mu_{\Lambda, \beta, h}^w(\sigma|_{\Delta} = \eta | \forall x \in \partial \Delta \cap \Lambda, \sigma_x = w'_x) = \mu_{\Delta, \beta, h}^{w'}(\eta).$$

Proof. Suppose for simplicity  $\partial \Delta \subset \Lambda$ .

The left-hand side is

$$\frac{\sum_{\sigma \in \Omega_\Lambda \text{ s.t. } \sigma|_{\Delta} = \eta, \sigma|_{\partial \Delta} = w'} \mu_\Lambda^w(\sigma)}{\sum_{\sigma \in \Omega_\Lambda \text{ s.t. } \sigma|_{\partial \Delta} = w'} \mu_\Lambda^w(\sigma)}.$$

By a direct computation, if  $\sigma \in \Omega_\Lambda$  s.t.  $\sigma|_{\partial \Delta} = w'$ ,

$$\mu_\Lambda^w(\sigma) \propto \mu_\Delta^{w'}(\sigma|_\Delta) \cdot \mu_{\Lambda \setminus \Delta}^{w, w'}(\sigma|_{\Lambda \setminus \Delta}) \cdot \exp\left(\beta \sum_{\substack{x, y \in \Lambda \\ x \sim y}} w'_x w'_y + h \sum_{x \in \Lambda} w'_x\right)$$

↑  
terms  $\Delta - \Delta$   
and  $\Delta - \partial \Delta$

↑  
terms  $\Lambda \setminus \Delta - \Lambda \setminus \Delta$   
and  $\Lambda \setminus \Delta - \partial \Delta$   
and  $\partial \Delta - \cdots$

↑  
terms  $\partial \Delta - \partial \Delta$ . indep of  $\sigma$

So we may write it as

$$\frac{\mu_{\Delta}^{w'}(\eta) \cdot \sum_{j \in \Delta \setminus \bar{\Delta}} \mu_{\Delta \setminus \bar{\Delta}}^{w,w'}(j)}{\left( \sum_{j \in \Delta} \mu_{\Delta}^{w'}(j) \right) \cdot \left( \sum_{j \in \Delta \setminus \bar{\Delta}} \mu_{\Delta \setminus \bar{\Delta}}^{w,w'}(j) \right)} = \mu_{\Delta}^{w'}(\eta)$$

□

Exercise: Using the domain Markov prop and the prop  $\mu_{\Lambda \setminus \bar{\Delta}}^w \leq \mu_{\Delta}^w$ , show that for  $\Delta \subset \Lambda \subset \mathbb{Z}^d$ ,

$$\mu_{\Lambda}^+ \leq \mu_{\Delta}^+ \quad (\text{for } \mathbb{P} \text{ Functions on } \{\pm 1\}^{\Delta})$$

and  $\mu_{\Lambda}^- \geq \mu_{\Delta}^-$ . Stochastic domination in domain

### B) DLR conditions

Let  $\beta > 0, h \in \mathbb{R}$ . A measure  $\mu$  on  $\mathcal{S} = \{\pm 1\}^{\mathbb{Z}^d}$  is said to be a **Gibbs measure** (for the Ising model at  $\beta, h$ ) if  $\Lambda \subset \mathbb{Z}^d$ ,  $w \in \{\pm 1\}^{\Lambda}$ , and any event  $E$  depending only on the sites in  $\mathbb{Z}^d \setminus \Lambda$ ,

"DLR conditions"

$$\mu(\cdot | \{\sigma|_{\Lambda} = w\} \cap E) = \mu_{\Lambda, \beta, h}^w(\cdot)$$

In other words, conditionnally on  $\sigma|_{\Lambda}$ ,  $\sigma|_{\Lambda}$  and  $\sigma|_{\Lambda^c}$  are indep and  $\sigma|_{\Lambda} \sim \mu_{\Lambda, \beta, h}^w$ .

Prop The measures  $\mu^+$  and  $\mu^-$  are Gibbs measures.

Proof: Let  $\Lambda \subset \mathbb{Z}^d$ , and let  $A, B$  be local events st  $A$  depends only on  $\Lambda$  and  $B$  depends only on  $\Lambda^c$ .

$$\mu^+(A | \{\sigma|_{\Lambda} = w\} \cap B) = \frac{\mu^+(A \text{ and } \sigma|_{\Lambda} = w \text{ and } B)}{\mu^+(\sigma|_{\Lambda} = w \text{ and } B)}$$

$$= \lim_{n \rightarrow \infty} \frac{\mu_{1_n}^+(A \text{ and } \sigma|_{\partial 1_n} = w \text{ and } B)}{\mu_{1_n}^+(\sigma|_{\partial 1_n} = w \text{ and } B)}$$

with  $n$  large enough s.t.  $B$   
depends only on  $1_n \cap 1$

By Spatial Markov Prop, this is

$$\mu_{1_n}^+(A | \sigma|_{\partial 1_n} = w \text{ and } B) = \mu_1^w(A).$$

As local events generate the  $\sigma$ -algebra, this is enough.  $\square$

### Extremality

It is natural to wonder if there are more Gibbs measures. There can be but we have found the extremal ones. If  $\mu^- \neq \mu^+$ , consider  $t\mu^- + (1-t)\mu^+$ . For  $t \in [0, 1]$ .

Prop If  $\mu$  is a Gibbs measure,

$$\mu^- \leq \mu \leq \mu^+. \quad (\text{For bounded } T \text{ Functions, local or continuous...})$$

Proof: Let  $f$  be a local, increasing function. For all  $n \geq 1$ ,  
and  $w \in \{-1, 1\}^{\partial 1_n}$ ,

$$\langle f \rangle_{1_n}^- \leq \langle f \rangle_{1_n}^w \leq \langle f \rangle_{1_n}^+. \quad (\text{by a proposition})$$

$$\text{so } \langle f \rangle_{1_n}^- \leq \sum_{w \in \{-1, 1\}^{\partial 1_n}} \underbrace{\mu(\sigma|_{\partial 1_n} = w)}_{\int f d\mu} \langle f \rangle_{1_n}^w \leq \langle f \rangle_{1_n}^+.$$

Then we let  $n \rightarrow \infty$ .

For continuous  $f$ , we approximate with local functions.  $\square$

Cor: The following are equivalent:

- L (i)  $\mu^+ = \mu^-$
- (ii)  $\exists!$  Gibbs measure

For  $h=0$ , we will see that this holds when  $\beta < \beta_c$ . (high temp)

## IV - Translation invariance

For  $x \in \mathbb{Z}^d$ , let  $\Theta_x : \mathbb{Z}^d \rightarrow \mathbb{Z}^d$  be the translation by  $x$ .  
 We extend it to events:  $\sigma \in \Theta_x(A)$  iff  $\sigma \circ \Theta_x^{-1} \in A$ .

Example: for  $A = \{\tau_0 = +1\}$ ,  $\Theta_x(A) = \{\tau_x = +1\}$ .

Theo The measures  $\mu^+, \mu^-$  are **translation invariant**.  
 L  $(\forall x \in \mathbb{Z}^d, \mu^\pm \circ \Theta_x = \mu^\pm)$

Proof: Let  $n \in \mathbb{N}^*$ . Then  $\mu_{1_n}^+ \circ \Theta = \mu_{x+1_n}^+$ .

But  $x+1_n \subset \Lambda_{n+1 \| x \|_\infty}$ , so

$\mu_{x+1_n}^+ \geq_{st} \mu_{\Lambda_{n+1 \| x \|_\infty}}^+$  (by an exercise!)

As  $n \rightarrow \infty$ , we get  $\mu^+ \circ \Theta \geq_{st} \mu^+$ .

Moreover, for  $n$  large enough,  $\Lambda_{n-1 \| x \|_\infty} \subset x+1_n$ , so  
 similarly  $\mu^+ \circ \Theta \leq_{st} \mu^+$ .

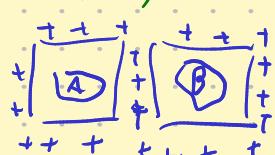
This implies  $\mu^+ \circ \Theta = \mu^+$ . The same works for  $\mu^-$ .  $\square$

Theo The measures  $\mu^+, \mu^-$  are **ergodic**:

L if  $A$  is an event that is invariant under all translations  
 of  $\mathbb{Z}^d$ , then  $\mu^\pm(A) \in \{0, 1\}$ .

Proof: We first show that if  $A, B$  are local, increasing events,  
 then they mix:  $\mu^+(A \cap \Theta_n(B)) \xrightarrow{n \rightarrow \infty} \mu^+(A) \mu^+(B)$  <sup>"asympt. indep."</sup>  
 Translation for  $x = (x_1, 0, \dots, 0)$  =

Indeed,  $\mu^+(A \cap \Theta_n(B)) \leq \mu^+(A \cap \Theta_n(B) \cap \{\tau_{\delta(\Lambda_{n/2})} = +\})$  and



FKG for  $\mu^+$ .

$\tau_{\delta(\Lambda_{n+1_{n/2}})} = +\}$

$$\begin{aligned}
 &= \mu_{\wedge_{n_2}}^+(A) \mu_{n+n_{n_2}}^+(\Theta_n(B)) \text{ by Gibbs} \\
 &= \mu_{\wedge_{n_2}}^+(A) \mu_{\wedge_{n_2}}^+(B) \\
 &\xrightarrow{n \rightarrow \infty} \mu^+(A) \mu^+(B)
 \end{aligned}$$

So  $\liminf_n \mu^+(A \cap \Theta_n(B)) \leq \mu^+(A) \mu^+(B)$ ,

but also  $\mu^+(A \cap \Theta_n(B)) \geq \mu^+(A) \mu^+(\Theta_n(B))$  by FKG

$$= \mu^+(A) \mu^+(B) \text{ by translation invariance}$$

So  $\mu^+(A \cap \Theta_n(B)) \xrightarrow{n \rightarrow \infty} \mu^+(A) \mu^+(B)$ . So all local,  $\nearrow$  events mix.

By monotone class lemmas, we deduce that any two events mix.

Now show that for  $A \nearrow \text{local}$ ,  $M_A = \{B \text{ s.t. } A, B \text{ mix}\}$  is a monotone class, and that  $\nearrow$  local events generate the whole  $\sigma$ -algebra,

then for any event  $B$  consider  $M_B = \dots$

Now for  $A$  translation invariant, we get

$$\begin{aligned}
 (\mu^+(A \cap \Theta_n(A))) &\xrightarrow{n \rightarrow \infty} \mu^+(A)^2 \\
 \mu^+(A \cap A) = \mu(A) &\quad \text{so } \mu^+(A) = \mu(A)^2 \quad \text{so } \mu^+(A) \in \{0, 1\}
 \end{aligned}$$

Same for  $\mu^-$ .

Exercise: Show that for  $f, g \in \text{Loc}(\omega)$ ,

$$\langle f \cdot g \rangle_{00\omega}^+ \xrightarrow{\text{short range correlations}} \langle f \rangle^+ \langle g \rangle^+.$$

$$\text{Deduce } \langle \nabla_0 \nabla_0 \rangle^+ \xrightarrow{n \rightarrow \infty} (\langle \nabla_0 \rangle^+)^2.$$

We conclude with the last missing part of the theorem announced at the beginning of this chapter:

Theo Let  $p > 0$ ,  $h \in \mathbb{R}$ . The following are equivalent:

|

- (i)  $\mu^+ = \mu^-$
- (ii)  $\langle \nabla_0 \rangle^+ = \langle \nabla_0 \rangle^-$

↑  
average of  $\nabla_0$  for the infinite volume measure  $\mu_{p,h}^+$ .

We already had a definition in the previous chapter

for  $h=0$ . Check that they are the same?

$$\langle \nabla_0 \rangle_{p,h}^+ = \lim_n \langle \nabla_0 \rangle_{p,h,n}^+$$

$$\langle \nabla_0 \rangle_{p,h}^- = \lim_n \langle \nabla_0 \rangle_{p,h,n}^-.$$

Proof: (i)  $\Rightarrow$  (iii) is clear.

(iii)  $\Rightarrow$  (i): Let  $A \in \mathbb{Z}^d$ .

$$\frac{\sum_{x \in A} n_x}{|A|} \quad \prod_{x \in A} \frac{1 + \sigma_x}{2}$$

The Function  $\sum_{x \in A} n_x - n_A$  is increasing,

$$\text{So } \left\langle \sum_{x \in A} n_x - n_A \right\rangle^- \leq \left\langle \sum_{x \in A} n_x - n_A \right\rangle^+.$$

$$\text{So } \langle n_A \rangle^+ - \langle n_A \rangle^- \leq \sum_{x \in A} \langle n_x \rangle^+ - \langle n_x \rangle^- = \frac{|A|}{2} (\langle \sigma_x \rangle^+ - \langle \sigma_x \rangle^-)$$

by translation invariance

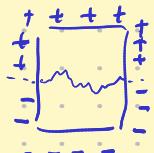
So (iii) implies  $\forall A \in \mathbb{Z}^d, \langle n_A \rangle^+ = \langle n_A \rangle^-$ .

This gives  $\forall A \in \mathbb{Z}^d, \langle \sigma_A \rangle^+ = \langle \sigma_A \rangle^-$ .

As the  $\langle \sigma_A \rangle$  are a basis of local functions, we get  $\mu^+ = \bar{\mu}$ .  
and  $h=0$  □

Remark: At low temp ( $\beta > \beta_c$ ), there exists Gibbs measures  $\neq \mu^\pm$ , and in particular there might exist some that are not translation-invariant (in particular this implies that they are not mixtures  $t\mu^- + (1-t)\mu^+$ ).

For instance, taking the limit of Dobrushin b.c.



← — —

- in dim 2, we get  $\frac{1}{2}\mu^- + \frac{1}{2}\mu^+$  (translation invariant)
- in dim  $> 3$ , we get another Gibbs measure that is not translation invariant.

See Friedli-Velenik, 3. 10.7.

## IV- Free energy and magnetization

An object related to the uniqueness of Gibbs measures is the free energy, or pressure. We will see that its derivative in  $h$  is the magnetization

### A) Existence of the Free energy

We will take  $(\Lambda_n)$  st

- $\Lambda_n, \Lambda_n \subset \mathbb{Z}^d$
- $\Lambda_n, \Lambda_n \subset \Lambda_{n+1}$
- $\cup \Lambda_n = \mathbb{Z}^d$
- $\frac{|\Lambda_n|}{|\Lambda_{n+1}|} \rightarrow 0$

$$(\Lambda_n) \nearrow \mathbb{Z}^d$$

Theo Let  $\beta > 0, h \in \mathbb{R}$ . Let  $(\Lambda_n) \nearrow \mathbb{Z}^d$  be such that  $\frac{|\partial \Lambda_n|}{|\Lambda_n|} \rightarrow 0$ .

Let  $(w_n)$  be a sequence of b.c. ( $\forall n, w_n \in \{-1, 0, +1\}^{\partial \Lambda_n}$ ). Then the limit

$$f(\beta, h) = \lim_{n \rightarrow \infty} \frac{1}{|\Lambda_n|} \log(Z_{\Lambda_n, \beta, h}^{w_n})$$

exists, and is independent of the b.c.  $(w_n)$ , in  $\mathbb{R}$  and of the domains  $(\Lambda_n)$ .

The function  $f$  is called the free energy.

Proof •  $(w_n)$  doesn't matter: Let  $(w'_n)$  be another seq. of b.c., then

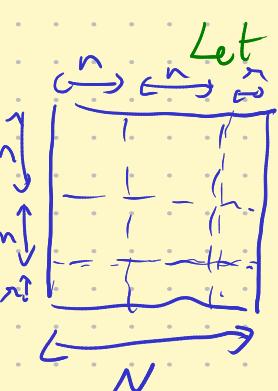
$$e^{-2\beta |\partial \Lambda_n|} \leq \frac{Z_{\Lambda_n, \beta, h}^{w_n}}{Z_{\Lambda_n, \beta, h}^{w'_n}} \leq e^{2\beta |\partial \Lambda_n|} \quad (\text{because true for } \mu(\sigma) \text{ for any } \sigma!)$$

$$\text{so } -2\beta \frac{|\partial \Lambda_n|}{|\Lambda_n|} \leq \frac{1}{|\Lambda_n|} (\log Z_{\Lambda_n}^{w_n} - \log Z_{\Lambda_n}^{w'_n}) \leq 2\beta \frac{|\partial \Lambda_n|}{|\Lambda_n|}.$$

If the limit exists, it is indeed indep of  $(w_n)$ .

• We take  $(w_n)$  all free b.c., and  $\Lambda_n = [(-n, n)]^d$

let  $1 \leq n \leq N$ . We write  $N = kn + r$ ,  $k \geq 0$ ,  $r \in [0, n]$ .



Let us compare  $Z_{\Lambda_N}^f$  with  $(Z_{\Lambda_n}^f)^{k^d}$ .

For  $r=0$ : We number the boxes of size  $n$  as

$$B_1, B_2, \dots, B_{k^d}.$$

$$\text{Let } D_m = \bigcup_{i=1}^m B_i.$$

Then

$$Z_{D_{m+1}}^f = \sum_{\sigma \in \Sigma_{D_{m+1}}} \exp \left( \beta \sum_{\substack{x, y \in E_{m+1} \\ \text{say } x \sim y}} \sigma_x \sigma_y + h \sum_{x \in D_{m+1}} \sigma_x \right)$$

$$= \sum_{\sigma \in \Sigma_{D_m}} \exp \left( \beta \sum_{\substack{x, y \in E_m \\ \text{say } x \sim y}} \sigma_x \sigma_y + h \sum_{x \in D_m} \sigma_x \right) \cdot Z_{\Lambda_n}^{w_\sigma}$$

by beginning of proof  $\rightarrow Z_{\Lambda_n}^f + \exp[O(|\partial \Lambda_n|)]$

1	4	7
2	5	8
3	6	9

$k^d$

$$\text{This gives } Z_{\Lambda_N}^F = (Z_{\Lambda_n}^F)^{k^d} \exp(\underbrace{O(k^d |\Lambda_n|)}_{= O(k^d n^{d-1})} ) = O(k N^{d-1})$$

So we have, more precisely,  $\exists c > 0$  s.t.

$$-c k N^{d-1} \leq \log Z_{\Lambda_N}^F - k^d \log Z_{\Lambda_n}^F \leq c k N^{d-1}$$

$$\text{So } \frac{1}{|\Lambda_N|} \log Z_{\Lambda_N}^F \leq \frac{k^d}{|\Lambda_N|} \log Z_{\Lambda_n}^F + \frac{c k N^{d-1}}{|\Lambda_N|}$$

check that  $n > 0$  still works

$$|\Lambda_N| = N^d \\ = k^d n^d \\ |\Lambda_n| = n^d$$

$$\text{So } \limsup_n \frac{1}{|\Lambda_N|} \log Z_{\Lambda_N}^F \leq \frac{1}{|\Lambda_n|} \log Z_{\Lambda_n}^F + \frac{c}{n}$$

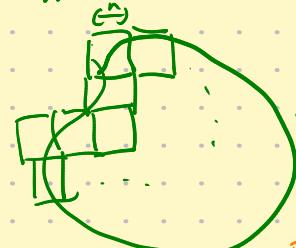
$$\leq \liminf_n \frac{1}{|\Lambda_n|} \log Z_{\Lambda_n}^F$$

So the lim exists.

• Independence on  $(\Lambda_n)$ : idea:

left as exercise:  
see Friedli-Velenik, 3.2.2.

try.



• For  $n$  large enough  
 $\frac{1}{n^d} \log(Z_{\Lambda'_n}^F)$   
close to  $f(\beta, h)$

• So (same as  $D_m$  before),

$[\Lambda'_N]_0 = [\Lambda'_N]$  also close to  $f$ .

• But  $\Lambda'_N \sqcup [\Lambda'_N]$

$\leq 2\Lambda'_N \cdot n^d$ .

... so close together.

Exercise: Find a sequence  $(\Lambda_n) \nearrow \mathbb{Z}^d$  and  $(w_n)$  s.t.

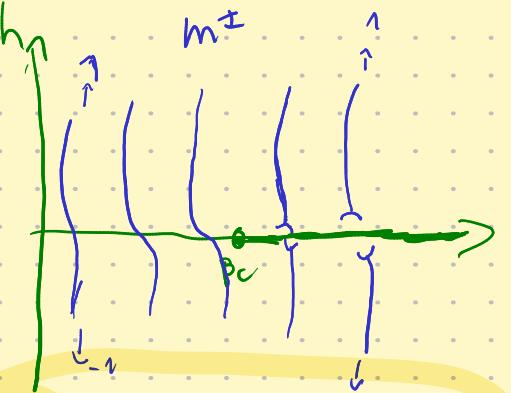
$$\text{st } \frac{1}{|\Lambda_n|} \log Z_{\Lambda_n}^{w_n} \not\rightarrow f(\beta, h)$$

## B) Analytic properties of the magnetization

We now work towards seeing the critical phase (where the phase transition occurs) as the point where the magnetization (and the free energy) is non-analytic.

Let  $m^+(\beta, h) := \langle \sigma_0 \rangle_{\beta, h}^+$  and  $m^-(\beta, h) := \langle \sigma_0 \rangle_{\beta, h}^-$ .

We would like to show that the profile of these functions is the same as in the Curie-Weiss model:



$m^\pm$  is analytic on  $(0, \infty) \times \mathbb{R}$   $\setminus \{\beta_c, +\infty\} \times \{0\}$ ,  
it is  $\nearrow$  in  $h$ .

$\text{tr}(\beta, h) / h \neq 0$  or  $\beta < \beta_c$ ,  
 $m^+(\beta, h) = m^-(\beta, h)$

For  $h=0$  it looks like:

so  $m^+$  is right-continuous in  $h$ .  $m^-$  left

So by the main Theo of section,  $\exists$  Gibbs measure there!

It is hard to get all these prop, in particular that  $m^\pm$  are continuous at  $(\beta_c, 0)$  (so they are both zero), but we will have the tools for it in the next chapter. Let us prove parts of this.

Remark:  $m^+(\beta, h) = -m^-(\beta, -h)$  by symmetry.

Prof:

- For any fixed  $\beta > 0$ ,  $h \mapsto m^+(\beta, h)$  is  $\nearrow$  and right-continuous  
 $m^-(\beta, h)$  is  $\nwarrow$  and left-continuous
- For any fixed  $h \geq 0$ ,  $\beta \mapsto m^+(\beta, h)$  is  $\nearrow$ .
- For any fixed  $h \leq 0$ ,  $\beta \mapsto m^-(\beta, h)$  is  $\searrow$ .

Proof: • By the remark, we just have to prove the props for  $m^+$ .

• Let  $\beta > 0$ . We know that there is monotonicity (stochastic domination) in  $h$ , so we have  $\forall h \leq h'$ ,  $m^+(\beta, h) \leq m^+(\beta, h')$ . (passing to the limit in  $\lambda$ )

(other proof:  $\frac{\partial}{\partial h} \langle \sigma_0 \rangle_{\beta, h}^+$   $= \frac{\partial}{\partial h} \left( \frac{\sum \sigma_0 \exp(\beta \sum_x \sigma_y + h \sum_x)}{\sum \exp(\beta \sum_x \sigma_y + h \sum_x)} \right)$   
For  $\lambda \ll \beta$ )

$$= \frac{\sum_n (\sigma_0 \sum_x - \sigma_0 \sum_x') \exp(\beta \sum_n \sigma_y + h \sum_n)}{\sum_n}$$

$$= \sum_n (\langle \sigma_0 \sum_x \rangle_{\beta, h}^+ - \langle \sigma_0 \sum_x' \rangle_{\beta, h}^+ \langle \sigma_0 \sum_x \rangle_{\beta, h}^+)$$

$\geq 0$  By FKG.

For right-continuity, we use the lemma:

Lemma: Let  $(a_{n,m})$  be a double sequence of real numbers which is bounded above and increasing.

$$n \leq n' \text{ and } m \leq m' \Rightarrow a_{n,m} \leq a_{n',m'}$$

(Exercise)

Then

$$\lim_n \lim_m a_{n,m} = \lim_m \lim_n a_{n,m} = \sup_{n,m} a_{n,m}.$$

so  $\mathbb{P}(h_m) \rightarrow h$ ,

$$\begin{aligned} \lim_m m(\beta, h_m) &= \lim_m \lim_n \langle \sigma_0 \rangle_{1_n, \beta, h_m}^+ \\ &= \lim_n \lim_m \langle \sigma_0 \rangle_{1_n, \beta, h_m}^+ \quad \text{by Lemma} \\ &= \lim_n \langle \sigma_0 \rangle_{1_n, \beta, h}^+ \quad \text{because for fixed } 1_n, \\ &\quad h \mapsto \langle \sigma_0 \rangle_{\beta, h}^+ \text{ is contin.} \\ &= m(\beta, h). \quad (\text{because differentiable!}) \end{aligned}$$

• (Careful, For  $\beta \leq \beta'$ , we do not necessarily have  $\mu_{1,\beta,h}^w \leq \mu_{1,\beta',h}^w$ )

Let  $h > 0$ .  $\frac{\partial}{\partial \beta} \langle \sigma_0 \rangle_{1,\beta,h}^+ = \dots$  as before

$$= \sum_{\{(x,y)\} \in E} \underbrace{\langle \sigma_0 \sigma_x \sigma_y \rangle_{1,\beta,h}^+ - \langle \sigma_0 \rangle^+ \langle \sigma_x \sigma_y \rangle_{1,\beta,h}^+}_{\geq 0 \text{ by Griffith with } h > 0}$$

■ (not proved before, but we will show it in next chapter)

Prop: If  $\beta > 0$ ,  $h \in \mathbb{R}$ ,

$$\lim_{h' \rightarrow h} m^+(\beta, h') = m^-(\beta, h).$$

As a result:

Theo: The Following are equivalent:

$$(i) \mu_{\beta,h}^+ = \mu_{\beta,h}^- \quad (\exists! \text{ Gibbs measure})$$

(iv)  $m^+(\beta, \cdot)$  is continuous at  $h$ .

Proof of Theo: We know that (i)  $\Leftrightarrow m^+(\beta, h) = m^-(\beta, h)$ , and since  $m^+$  is right-cont, with the Prop,  $m^+$  is continuous at  $h$  ( $\Leftrightarrow m^-(\beta, h) = m^+(\beta, h)$ )

Proof of Prop: •  $m^-(\beta, h) = \lim_{h' \nearrow h} m^-(\beta, h')$  by left-continuity

$\leq \lim_{h' \nearrow h} m^+(\beta, h')$  by stochastic dom.

in b.c.  $\sim$  clearly enough to have equality

• Let  $h' < h$ . We show in fact  $a := m^+(\beta, h') \leq m^-(\beta, h)$ .

Let next we write  $S = \sum_{x \in A_n} \Delta_x$ ,  $S' = \sum_{\substack{x, y \in A \\ y \in \partial A_n}} \Delta_{xy}$ .

Then  $\langle S \rangle_{A_n, \beta, h'}^+ = \sum_{x \in A_n} \langle \Delta_x \rangle_{A_n, \beta, h'}^+ \leq |A_n| a$ .

( $\forall x$ , in  $n$  this  $\rightarrow$  to  $m^+(\beta, h) = a$ )

Let  $\varepsilon > 0$ .

$$\begin{aligned} \mu_{A_n, \beta, h'}^+(S \leq (a-\varepsilon)|A_n|) &= \mu_{A_n, \beta, h'}^+(|A_n| - S \geq (1-a+\varepsilon)|A_n|) \\ &\stackrel{\text{Markov}}{\leq} \frac{|A_n| - \langle S \rangle_{A_n, \beta, h'}^+}{(1-a+\varepsilon)|A_n|} \leq \frac{1-a}{1-a+\varepsilon} \leq 1 - \frac{\varepsilon}{2}. \end{aligned}$$

Similarly,  $\langle S \rangle_{A_n, \beta, h}^- \geq |A_n| b$

$$\begin{aligned} \text{and } \mu_{A_n, \beta, h}^-(S \geq (b+\varepsilon)|A_n|) &= \mu_{A_n, \beta, h}^-(|A_n| - S \geq (1+b+\varepsilon)|A_n|) \\ &\leq \frac{1+b}{1+b+\varepsilon} \leq 1 - \frac{\varepsilon}{2}. \end{aligned}$$

$$So \frac{\varepsilon}{2} \leq \mu_{A_n, \beta, h}^+(S > (a-\varepsilon)|A_n|)$$

$$= \frac{\langle 1_{S > (a-\varepsilon)|A_n|} \exp((h'-h)S + 2\beta S') \rangle_{A_n, \beta, h}}{\langle \exp((h'-h)S + 2\beta S') \rangle_{A_n, \beta, h}}$$

$$\leq \frac{\exp((h'-h)(a-\varepsilon)|A_n| + 4\beta d|A_n|)}{\langle 1_{S < (b+\varepsilon)|A_n|} \exp((h'-h)S) \rangle_{A_n, \beta, h}}$$

$$\leq \frac{\exp((h'-h)(a-\varepsilon)|A_n| + 4\beta d|A_n|)}{\langle 1_{S < (b+\varepsilon)|A_n|} \exp((h'-h)S) \rangle_{A_n, \beta, h}}$$

$$\leq \frac{\exp((h'-h)(a-b-2\varepsilon)|\Lambda_n| + 4\beta d|\partial\Lambda_n|)}{m_{\Lambda_n, \beta, h}^-(S \subset (b+\varepsilon)|\Lambda_n|)}$$

$$\text{So } \left(\frac{\varepsilon}{2}\right)^2 \leq \exp[(h'-h)(a-b-2\varepsilon)|\Lambda_n|] \cdot \exp(4\beta d|\partial\Lambda_n|)$$

If we had  $a-b-2\varepsilon > 0$ , then as  $n \rightarrow \infty$ ,  
as  $h' - b \leq 0$  and  $(\Lambda_n) \supset (b|\Lambda_n|)$ , the right-hand-side  
would go to 0. Contradiction.

So  $a-b-2\varepsilon \leq 0$ , and  $a \leq b+2\varepsilon \dots$  so  $a \leq b$ , as claimed.  $\blacksquare$

This allows us to prove the announced profile for  $h \geq 0$ :

Theo  $\forall \beta > 0$ ,  $\forall h \geq 0$ , there is a unique Gibbs measure  $(\mu^- = \mu^+)$   
 $m^- = m^+$

Proof: We also need an inequality that will be proved in  
the next chapter, the GHS inequality:

Griffith-Hurst-Sherman

$$\boxed{\begin{aligned} \text{For } \Lambda \subset \mathbb{Z}^d, \beta > 0, h \geq 0, \forall x, y, z \in \Lambda, \\ \langle \sigma_x \sigma_y \sigma_z \rangle_{\Lambda, \beta, h}^+ - \langle \sigma_x \rangle_{\Lambda, h, \beta}^+ \langle \sigma_y \rangle_{\Lambda, h, \beta}^+ - \langle \sigma_y \rangle_{\Lambda, h, \beta}^+ \langle \sigma_z \rangle_{\Lambda, h, \beta}^+ - \langle \sigma_z \rangle_{\Lambda, h, \beta}^+ \langle \sigma_x \sigma_y \rangle_{\Lambda, \beta, h}^+ \\ + 2 \langle \sigma_x \rangle_{\Lambda, h, \beta}^+ \langle \sigma_y \rangle_{\Lambda, h, \beta}^+ \langle \sigma_z \rangle_{\Lambda, h, \beta}^+ \leq 0. \end{aligned}}$$

We assume it for now.

We prove that  $\forall \Lambda \subset \mathbb{Z}^d$ ,  $h \mapsto \langle \sigma_x \rangle_{\Lambda, h, \beta}^+$  is concave (on  $[0, \infty)$ )

This implies that the limit  $h \mapsto m(\beta, h)$  is concave (on  $[0, \infty)$ )  
so it is continuous (on  $(0, \infty)$ ).

We get the Theo for  $h > 0$ , but also for  $h < 0$  as  $m^+(\beta, h) = -m^-(\beta, -h)$

Let's go: For  $\Lambda \subset \mathbb{Z}^d$ ,  $\beta > 0$ ,  $h > 0$ , we differentiate twice:

$$\forall A \subset \mathbb{Z}^d, \frac{\partial}{\partial h} (\langle \sigma_A \rangle_{\Lambda, \beta, h}^+) = \dots = \sum_{x \in A} \langle \sigma_x \sigma_A \rangle_{\Lambda, \beta, h}^+ - \langle \sigma_x \rangle_{\Lambda, \beta, h}^+ \langle \sigma_A \rangle_{\Lambda, \beta, h}^+$$

$$\begin{aligned}
\text{and so, } \frac{\partial^2}{\partial h^2} (\langle \sigma_0 \rangle_{\Lambda, \beta, h}^+) &= \frac{\partial}{\partial h} \left( \sum_n \langle \sigma_n \sigma_0 \rangle - \langle \sigma_n \rangle \langle \sigma_0 \rangle \right) \\
&= \sum_n \frac{\partial}{\partial h} \langle \sigma_0 \sigma_n \rangle - \langle \sigma_0 \rangle \frac{\partial}{\partial h} \langle \sigma_n \rangle - \langle \sigma_n \rangle \frac{\partial}{\partial h} \langle \sigma_0 \rangle \\
&= \sum_{n,y} \langle \sigma_0 \sigma_n \delta_y \rangle - \langle \sigma_0 \sigma_x \rangle \langle \sigma_y \rangle - \langle \sigma_0 \rangle \langle \sigma_x \delta_y \rangle + \langle \sigma_0 \rangle \langle \sigma_x \rangle \\
&\quad - \langle \sigma_x \rangle \langle \sigma_0 \sigma_y \rangle + \langle \sigma_x \rangle \langle \sigma_0 \rangle \langle \sigma_y \rangle
\end{aligned}$$

$\leq 0$  By GHS!  $\blacksquare$ .

### G) Analytic prop. of the Free energy

Prop: The function  $f$  is convex on  $(0, \infty) \times \mathbb{R}$ .

Proof: Let  $(\beta, h), (\beta', h') \in (0, \infty) \times \mathbb{R}$  and  $t \in [0, 1]$ ,  
and  $\beta_t = t\beta + (1-t)\beta'$ ,  $h_t = th + (1-t)h'$ .

We want to show

$$f(\beta_t, h_t) \leq t f(\beta, h) + (1-t) f(\beta', h').$$

$$\begin{aligned}
\text{But } Z_{\Lambda_n, \beta_t, h_t}^f &= \sum_j \exp(\beta \sum_i \sigma_{ij} + h \sum_i \tau_{ij})^t \exp(\beta' \sum_i \sigma_{ij} + h' \sum_i \tau_{ij})^{1-t} \\
&\stackrel{\text{Holder}}{\leq} (Z_{\Lambda_n, \beta, h}^f)^t (Z_{\Lambda_n, \beta', h'}^f)^{1-t}. \text{ We take the log and voilà!} \quad \blacksquare
\end{aligned}$$

Cor: If  $\beta > 0$ , the function  $h \mapsto f(\beta, h)$  has left- and right derivatives at every point, and

$$\frac{\partial f}{\partial h}(\beta, h^-) = \langle \sigma_0 \rangle_{\beta, h}^-, \quad \frac{\partial f}{\partial h}(\beta, h^+) = \langle \sigma_0 \rangle_{\beta, h}^+.$$

• If  $\beta > 0$ ,  $\exists!$  Gibbs measure at  $(\beta, h)$  iff  $h \mapsto f(\beta, h)$  is differentiable at  $h$ .

We need a lemma for the proof:

Lemma (average magnetization):

$$\boxed{H_{\beta, h}, \langle \sigma_0 \rangle_{\beta, h}^+ = \lim_{n \rightarrow \infty} \frac{1}{|\Lambda_n|} \left\langle \sum_{x \in \Lambda_n} \sigma_x \right\rangle_{\Lambda_n, \beta, h}^+}$$

Proof of lemma:

- Let  $\eta > 0$  be s.t.  $(\langle \sigma_0 \rangle_{\Lambda_n}^+ - \langle \sigma_0 \rangle^+) \leq \varepsilon$ .

Then for  $n > 0$  large,



$$\begin{aligned} \sum_{x \in \Lambda_n} \langle \sigma_x \rangle_{\Lambda_n, \beta, h}^+ &\leq \sum_{x \in \Lambda_{n-r}} \langle \sigma_x \rangle_{\Lambda_n, \beta, h}^+ \quad \leftarrow (\text{monotonicity in domain}) \\ &\quad + \sum_{x \in \Lambda_n \setminus \Lambda_{n-r}} 1 \\ &\leq |\Lambda_{n-r}| (\langle \sigma_0 \rangle^+ + \varepsilon) + |\Lambda_n \setminus \Lambda_{n-r}|. \end{aligned}$$

$$\text{so } \limsup_n \frac{1}{|\Lambda_n|} \sum_{x \in \Lambda_n} \langle \sigma_x \rangle_{\Lambda_n, \beta, h}^+ \leq \langle \sigma_0 \rangle^+ + \varepsilon \text{ so } \leq \langle \sigma_0 \rangle^+.$$

- On the other hand,  $\langle \sigma_x \rangle_{\Lambda_n}^+ \geq \langle \sigma_0 \rangle_{\Lambda_{n+|h|}}^+ \geq \langle \sigma_0 \rangle_{\Lambda_{2n}}^+$ .  
So  $\frac{1}{|\Lambda_n|} \sum_{x \in \Lambda_n} \langle \sigma_x \rangle_{\Lambda_n}^+ \geq \langle \sigma_0 \rangle^+ - \varepsilon$  for  $n$  large enough.  
So  $\liminf_n \langle \sigma_x \rangle_{\Lambda_n}^+ \geq \langle \sigma_0 \rangle^+ - \varepsilon \Rightarrow \langle \sigma_0 \rangle^+ \boxed{\text{R}}$ .

Proof of Cor: By convexity we have left & right derivatives

$$\left( \frac{\partial f}{\partial h}(\beta, h^+) = \lim_{s \rightarrow 0^+} \frac{f(\beta, h+s) - f(\beta, h)}{s} \right) \boxed{h}$$

Moreover,  $\frac{\partial}{\partial h} \left( \log z_{\Lambda_n, \beta, h}^+ \right) = \langle \sum_{x \in \Lambda_n} \sigma_x \rangle_{\Lambda_n, \beta, h}^+$  (we see the average mogn!)

$$\begin{aligned} \text{so } \frac{f(\beta, h+s) - f(\beta, h)}{s} &= \lim_n \frac{1}{s|\Lambda_n|} \left( \log(z_{\Lambda_n, \beta, h}^+) - \log(z_{\Lambda_n, \beta, h+s}^+) \right) \\ &= \lim_n \frac{1}{s} \int_h^{h+s} \underbrace{\frac{1}{|\Lambda_n|} \langle \sum_{x \in \Lambda_n} \sigma_x \rangle_{\Lambda_n, \beta, g}^+}_{\geq \frac{1}{|\Lambda_n|} \langle \sum_{x \in \Lambda_n} \sigma_x \rangle_{\Lambda_n, \beta, h}^+ \text{ as } g \geq h} dg \end{aligned}$$

and  $\leq \frac{1}{|\Lambda_n|} \langle \sigma_x \rangle_{\Lambda_n, \beta, h+s}^+ \text{ as } g \leq h$

$$\text{So } \lim_{n \rightarrow \infty} \frac{1}{m_n} \langle \sum \phi_n \rangle_{n \geq 1, h}^+ \leq \frac{f(\beta, h+\delta) - f(\beta, h)}{\delta} \leq \lim_{\substack{n \rightarrow \infty \\ h \rightarrow 0}} \frac{1}{m_n} \langle \sum \phi_n \rangle_{n \geq 1, h}^+$$

We let  $\delta \rightarrow 0$  and use right-continuity of  $m^+$  to get

$$\frac{\partial}{\partial h} f(\beta, h^+) = m^+(\beta, h). \quad \text{Similar for left-deriv. } \square$$

Exercise Show that  $m^+(\beta, h) > 0$  for  $\beta > 0$  and  $h > 0$ .