


① "Solutions"

Ex 1: By the same argument as in the lectures,

For  $P = \begin{pmatrix} e^\beta & e^{-\beta} \\ e^{-\beta} & e^\beta \end{pmatrix}$ , 

$Z_n^+$  is the "bottom right" coef of  $P^{2n+2}$ .

$$\left( Z_n^+ = \sum_{\substack{\sigma_{-n-1}=+1, \\ \sigma_{-n}, \dots, \sigma_n \in \{\pm 1\}, \\ \sigma_{n+1}=+1,}} P_{\sigma_{-n-1}, \sigma_{-n}} \dots P_{\sigma_n, \sigma_{n+1}} = P^{2n+2} [+, +] \right)$$

This can be written as

$$Z_n^+ = (0 \ 1) P^{2n+2} \begin{pmatrix} 0 \\ 1 \end{pmatrix}.$$

Similarly,  $\langle \sigma_0 \rangle^+ Z_n^+ = \sum_{\text{same}} \sigma_0 P_{\sigma_{-n-1}, \sigma_{-n}} \dots P_{\sigma_n, \sigma_{n+1}}$

$$= \sum_{\substack{\sigma_{-n-1}=+1, \\ \sigma_{-n}, \dots, \sigma_{-1} \in \{\pm 1\}, \\ \sigma_0=+1, \\ \sigma_1, \dots, \sigma_n \in \{\pm 1\}, \\ \sigma_{n+1}=+1}} P_{\sigma_{-n-1}, \sigma_{-n}} \dots P_{\sigma_{-1}, \sigma_0} P_{\sigma_0, \sigma_1} \dots P_{\sigma_n, \sigma_{n+1}}$$

$$- \sum_{\text{same with } \sigma_0 = -1} \dots$$

$$= P^{n+1} [+, +] P^{n+1} [+, +] - P^{n+1} [+, -] P^{n+1} [-, +].$$

$$= \left[ (0 \ 1) P^{n+1} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \right]^2 - \left[ (0 \ 1) P^{n+1} \begin{pmatrix} 1 \\ 0 \end{pmatrix} \right]^2 \quad \text{as } P \text{ is symmetric.}$$

②)  $S_0 \langle \sigma_0 \rangle^+ = \frac{[(01) P^{n+1} \begin{pmatrix} 0 \\ 1 \end{pmatrix}]^2 - [\cancel{P^{n+1}} (01) P^{n+1} \begin{pmatrix} 1 \\ 0 \end{pmatrix}]^2}{(01) P^{2n+2} \begin{pmatrix} 0 \\ 1 \end{pmatrix}}$

other solution: put a magnetic field  $h$  at  $0$ , then differentiate with resp. to  $h$ ...

Ex 2: We follow the hint.  
For  $\Lambda \subset \mathbb{Z}^d$ ,  $\beta > 0$ ,  $h > 0$ , let

$\forall \sigma \in \Omega = \{\pm 1\}^\Lambda$ ,  $Z_\beta = \frac{\exp\left(\beta \sum_{\substack{x \sim y \\ x, y \in \Lambda}} \sigma_x \sigma_y + \beta \sum_{\substack{x \sim y \\ x \in \Lambda, y \in \Lambda^c}} \sigma_x \overset{+1}{=} \sigma_y + h \sigma_0\right)}{Z_\beta}$

where  $Z_\beta = \dots$

We want to show  $\mu_{\Lambda, \beta, 0}^+ \leq_{st} \nu \leq_{st} \mu_{\Lambda, \beta, h}^+$ .

We use a criterion from the course.

Let  $\sigma \leq \sigma' \in \Omega$ , and  $x \in \Lambda$ .

$\frac{\mu_{\Lambda, \beta, 0}^+(\sigma^{+x})}{\mu_{\Lambda, \beta, 0}^+(\sigma^{-x})} = \exp\left(\beta \sum_{y \sim x} \sigma_y\right)$  +1 if  $y \in \Lambda$

// since  $\sigma \leq \sigma'$  and  $0 \leq h$ .

$\frac{\nu(\sigma'^{+x})}{\nu(\sigma'^{-x})} = \exp\left(2\beta \sum_{y \sim x} \sigma'_y + 2h \mathbb{1}_{x=0}\right)$

so  $\mu_{\Lambda, \beta, 0}^+ \leq_{st} \nu$ .

Similarly,  $\frac{\nu(\sigma^{+x})}{\nu(\sigma^{-x})} = \exp\left(2\beta \sum_{y \sim x} \sigma_y + 2h \mathbb{1}_{x=0}\right)$

//  $\frac{\mu_{\Lambda, \beta, h}^+(\sigma'^{+x})}{\mu_{\Lambda, \beta, h}^+(\sigma'^{-x})} = \exp\left(2\beta \sum_{y \sim x} \sigma'_y + 2h \mathbb{1}_{x=0}\right)$  so  $\nu \leq_{st} \mu_{\Lambda, \beta, h}^+$ .

③ Since  $\langle \sigma_0 \rangle$  is  $\nearrow$ ,  $\langle \sigma_0 \rangle \leq \langle \sigma_0 \rangle_{\Lambda, \beta, h}^+$

But  $\chi(\sigma_0) = \frac{\langle \sigma_0 e^{h\sigma_0} \rangle_{\Lambda, \beta, 0}^+}{\langle e^{h\sigma_0} \rangle_{\Lambda, \beta, 0}^+} = \frac{\langle \sigma_0 \sum_{n \geq 0} \frac{h^n \sigma_0^n}{n!} \rangle_{\Lambda, \beta, 0}^+}{\langle \sum_{n \geq 0} \frac{h^n \sigma_0^n}{n!} \rangle_{\Lambda, \beta, 0}^+}$

$$= \frac{\langle \sigma_0 \rangle_{\Lambda, \beta, 0}^+ \cosh h + \sinh h}{\langle \sigma_0 \rangle_{\Lambda, \beta, 0}^+ \sinh h + \cosh h}$$

$$\geq \frac{\sinh(h)}{\sinh(h) + \cosh(h)}$$

So  $\langle \sigma_0 \rangle_{\Lambda, \beta, h}^+ \geq \chi$ . As  $n \rightarrow \infty$ , we get that it is  $> 0$ .

Ex 3: 1)  $\langle \sigma_0 \rangle_{\beta, n}^+$  is  $\nearrow$  in  $\beta$ :

for  $\beta \leq \beta'$ ,

$$\langle \sigma_0 \rangle_{\beta', n}^+ = \frac{\langle \sigma_0 g \rangle_{\beta', n}^+}{\langle g \rangle_{\beta', n}^+}$$

where  $g = \exp(\beta' \sigma_0)$ .

$$\sum_{k \geq 0} \frac{1}{k!} (\beta' - \beta)^k (\sum_{x \sim y} \sigma_x \sigma_y)^k$$

$$= \sum_{S \subset V_n} \beta_S \nu_S, \beta_S \geq 0.$$

So  $\langle \sigma_0 g \rangle_{\beta', n}^+ = \sum \beta_S \langle \sigma_0 \sigma_S \rangle_{\beta, n}^+$

$$\geq \sum \beta_S \langle \sigma_0 \rangle_{\beta, n}^+ \langle \sigma_S \rangle_{\beta, n}^+$$

$$\geq \langle \sigma_0 \rangle_{\beta, n}^+ \langle g \rangle_{\beta, n}^+$$

This gives  $\langle \sigma_0 \rangle_{\beta, n}^+ \leq \langle \sigma_0 \rangle_{\beta', n}^+$ .

by Griffith

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④  $\langle \sigma_0 \rangle_{\beta, n}^+$  is  $\downarrow$  in  $n$ :

$$\langle \sigma_0 \rangle_{\beta, n}^+ = \frac{\langle \sigma_0 \mathbb{1}_{\forall x/d(0,x)=n, \sigma_x=+1} \rangle_{\beta, n+1}^+}{\langle \mathbb{1}_{\forall x/d(0,x)=n, \sigma_x=+1} \rangle_{\beta, n+1}^+}$$

by domain Markov prop.

Since  $\mathbb{1}_{\dots} = \prod_{x/d(0,x)=n} \frac{1+\sigma_x}{2} = \frac{1}{2^{P(P-1)^{n+1}}} \sum_{T \subset \{x/d(0,x)=n\}} \sigma_T$ ,

$$\langle \sigma_0 \rangle_{\beta, n}^+ = \frac{\sum_{T \dots} \langle \sigma_0 \sigma_T \rangle_{\beta, n+1}^+}{\sum_{T \dots} \langle \sigma_T \rangle_{\beta, n+1}^+} \geq \langle \sigma_0 \rangle_{\beta, n+1}^+.$$

2) By  $\downarrow$  in  $n$ ,  $\langle \sigma_0 \rangle_{\beta}^+$  is well defined.

By  $\uparrow$  in  $\beta$ ,  $\beta \mapsto \langle \sigma_0 \rangle_{\beta}^+$  is  $\uparrow$ .

$$\text{Let } \beta_c = \sup \{ \beta > 0 / \langle \sigma_0 \rangle_{\beta}^+ = 0 \} \\ = \inf \{ \beta > 0 / \langle \sigma_0 \rangle_{\beta}^+ > 0 \}.$$

3) By HTE,

$$\langle \sigma_0 \rangle_{\beta, n}^+ = \frac{\sum_{\eta/\partial\eta=\{0\}} \text{th}_{\beta}^{|\eta|}}{\sum_{\eta/\partial\eta=\emptyset} \text{th}_{\beta}^{|\eta|}}$$

where  $\eta \subset E_n$ ,  
 $\partial\eta := \{x \in V_n \setminus \partial V_n / \deg_{\eta}(x) \text{ odd}\}$

IF  $\partial\eta = \{0\}$ , then...  $\eta$  contains a path  $\gamma$  from  $0$  to  $\partial V_n$ .

For a given path  $\gamma$ ,

$$\frac{\sum_{\eta/\gamma \subset \eta, \partial\eta=\{0\}} \text{th}_{\beta}^{|\eta|}}{\sum_{\eta/\partial\eta=\emptyset} \text{th}_{\beta}^{|\eta|}} = \text{th}_{\beta}^{|\gamma|} \frac{\sum_{\eta \dots} \text{th}_{\beta}^{|\eta \setminus \gamma|}}{\sum_{\eta/\partial\eta=\emptyset} \text{th}_{\beta}^{|\eta|}}.$$

⑤)  $\leq t h \beta^{|\partial|}$  by the same argument as in the lectures

So  $\langle \sigma_0 \rangle_{\beta, n}^+ \leq \sum_{\gamma} t h \beta^{|\gamma|}$

Moreover,  $|\gamma| = n$  and there are (exactly!)  $p \cdot (p-1)^{n-1}$  paths  $0 \rightsquigarrow \partial V_n$ .

So  $\langle \sigma_0 \rangle_{\beta, n}^+ \leq p(p-1)^{n-1} (t h \beta)^n$

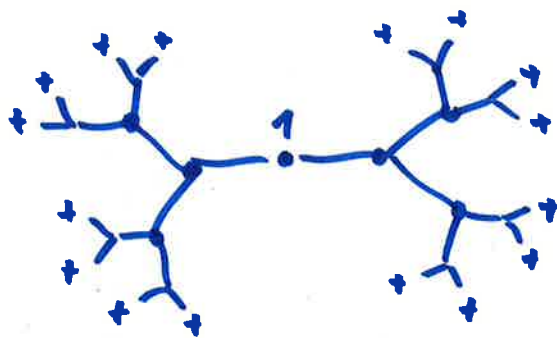
Therefore, if  $t h \beta < \frac{1}{p-1}$  ( $\Leftrightarrow \beta < \text{cograph}(\frac{1}{p-1})$ ),

$\langle \sigma_0 \rangle_{\beta, n}^+ \rightarrow 0$  and  $\langle \sigma_0 \rangle_{\beta}^+ = 0$ .

So  $\beta_c \geq \text{cograph}(\frac{1}{p-1})$ .

4) a)  $Z_n^+ = \sum_{\sigma_0 \in \{\pm 1\}} \sum_{\substack{\sigma: V_n \rightarrow \{\pm 1\} \\ \partial V_n \rightarrow +1}} \exp(\beta \sum_{\{x, y\} \in E_n} \sigma_x \sigma_y)$

Let  $T'_n$  be the tree whose root 1 has degree  $p-1$  and the other non-leaves vertices have degree  $p$ :

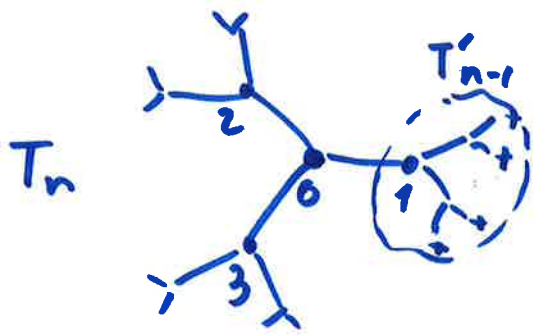


$T'_4$  (For  $p=3$ )

Let  $V'_n, \partial V'_n, E'_n$  vertices and edges of  $T'_n \dots$

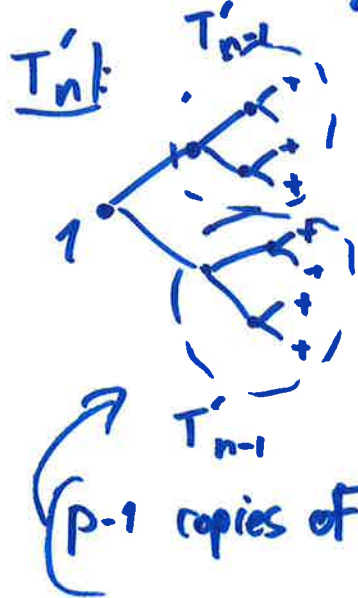
Let  $\chi_n(\sigma_0) = \sum_{\substack{\sigma \in \{\pm 1\}^{V'_n} \\ \sigma|_{\partial V'_n} = +1}} e^{\beta \sum_{\{x, y\} \in E'_n} \sigma_x \sigma_y} \exp(\beta \sum_{\{x, y\} \in E_n} \sigma_x \sigma_y)$

⑥) Then  $Z_n^+ = \sum_{\sigma_0 \in \{\pm 1\}} Y_{n-1}(\sigma_0)^P$ .



b) Let  $n \geq 1$ .

$$Y_n(\sigma_0) = \sum_{\substack{\sigma: V'_n \rightarrow \{\pm 1\} \\ \partial V'_n \rightarrow 1}} e^{\beta \sigma_0 \sigma_1} \exp\left(\beta \sum_{\{x,y\} \in E'_n} \sigma_x \sigma_y\right)$$



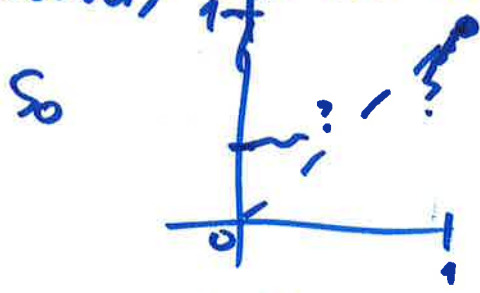
$$= \sum_{\sigma_1 \in \{\pm 1\}} e^{\beta \sigma_0 \sigma_1} (Y_{n-1}(\sigma_1))^{p-1}$$

$$\text{So } x_n = \frac{Y_n(-1)}{Y_n(+1)} = \frac{e^\beta Y_{n-1}(-1)^{p-1} + e^{-\beta} Y_{n-1}(+1)^{p-1}}{e^{-\beta} Y_{n-1}(-1)^{p-1} + e^\beta Y_{n-1}(+1)^{p-1}}$$

$$= \frac{e^\beta x_{n-1}^{p-1} + e^{-\beta}}{e^{-\beta} x_{n-1}^{p-1} + e^\beta} = \frac{1 + e^{2\beta} x_{n-1}^{p-1}}{e^{2\beta} + x_{n-1}^{p-1}} = F(x_{n-1})$$

$$c) \langle \sigma_0 \rangle_{\beta, n}^+ = \frac{\sum_{\sigma_0 \in \{\pm 1\}} \sigma_0 (Y_{n-1}(\sigma_0))^P}{\sum_{\sigma_0 \in \{\pm 1\}} (Y_{n-1}(\sigma_0))^P} = \frac{1 - x_{n-1}^P}{1 + x_{n-1}^P}$$

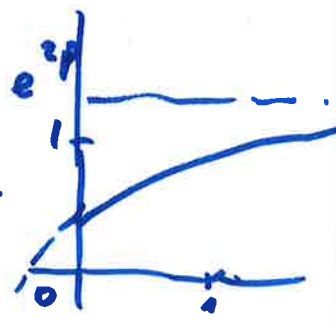
⑦ Moreover,  $F$  is continuous on  $[0, 1]$  and  $F(0) = e^{-2\beta}$ ,  
 $F(1) = 1$ ,  
 $\forall x \in [0, 1], F(x) \geq 0$  and  $F(x) \leq 1$ .



so  $(x_n)_n$  stays in  $[0, 1]$ .

~~Also,  $F$  is  $\uparrow$  on  $[0, 1]$~~

Also,  $F$  is  $\uparrow$  on  $[0, 1]$  (by composing  $\frac{1 + e^{2\beta}t}{e^{2\beta} + t}$  and  $x \mapsto x^{p-1}$ )!



so  $x_n$  is  $\uparrow$  so  $\lim_{n \rightarrow \infty} x_n$  exists (and is the smallest fixed point of  $F$  in  $[0, 1]$ )

We get  $\langle \sigma_0 \rangle_{\beta}^+ = \frac{1 - e^{\beta}}{1 + e^{\beta}}$ . This is 0 iff  $\beta = 1$ .

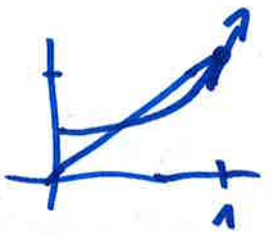
d) Quick way: let  $\beta \geq \text{arctanh}\left(\frac{1}{p-1}\right)$ .

We want to show that  $\beta \neq 1$  in that case.

This will show that  $\beta \geq \beta_c$ , so that  $\beta_c \leq \text{arctanh}\left(\frac{1}{p-1}\right)$ , and with 3) we get equality.

$$F'(x) = \dots = (p-1) \frac{e^{4\beta} x^{p-2} (e^{4\beta} - 1)}{(e^{2\beta} + x^{p-1})^2}$$

$$\text{so } F'(1) = (p-1) \frac{e^{4\beta} - 1}{(e^{2\beta} + 1)^2} = (p-1) \frac{e^{2\beta} - 1}{e^{2\beta} + 1} = (p-1) \tanh \beta > 1$$



So  $F$  has a fixed point  $< 1$  in  $[0, 1]$ .  
 so  $\beta \neq 1$ !

⑧) 5) The idea is to Taylor expand  $P(\beta) = F(P(\beta))$  around  $\beta_0$ , but I would need more time...

Ex 4: 1) For  $\sigma \neq \sigma'$ ,

if  $\exists i \in \{0, \dots, n-1\}$  / ~~if~~  $\forall j \neq i, \sigma_j = \sigma'_j$

and  $\sigma_i \neq \sigma'_i$ ,

~~so~~  $\sigma_i = +1, \sigma'_i = -1$   
then

$$P(\sigma, \sigma') = \frac{1}{Z} \exp\left(\frac{\beta}{n} \sum_{k=0}^{n-1} \sigma_k \sigma'_k\right) \frac{1}{n} \frac{\mu(\sigma')}{\mu(\sigma) + \mu(\sigma')}$$

$$= \frac{1}{n} \frac{\mu(\sigma) \mu(\sigma')}{\mu(\sigma) + \mu(\sigma')} = P(\sigma', \sigma).$$

otherwise,  $P(\sigma, \sigma') = 0 = P(\sigma', \sigma)$

(we didn't treat  $\sigma = \sigma'$ , for which symmetry is obvious)

2) P-ns on  $(\sigma, \sigma')$ ,  $F(\sigma, \sigma')$  is +2 if we chose a -1 spin in  $\sigma$  and make it  $\alpha + 1$ ,  
-2 if we chose a +1 spin in  $\sigma$  and make it  $\alpha - 1$ ,  
0 otherwise.

$$\text{so } f(\sigma) = E[F(\sigma, \sigma') | \sigma]$$

$$= \sum_{\substack{0 \leq i \leq n-1 \\ \sigma_i = +1}} \frac{1}{n} \cdot 2 \cdot \frac{\mu(\sigma^{-i})}{\mu(\sigma) + \mu(\sigma^{-i})} - \sum_{\substack{0 \leq i \leq n-1 \\ \sigma_i = -1}} \frac{1}{n} \cdot 2 \cdot \frac{\mu(\sigma^{+i})}{\mu(\sigma) + \mu(\sigma^{+i})}$$



(9) 
$$= \frac{2}{n} \sum_{i/\sigma_i=+1} \frac{\exp(-4 \sum_{j \neq i} \frac{\beta}{n} \sigma_j)}{1 + \exp(-4 \sum_{j \neq i} \beta \sigma_j)} - \frac{2}{n} \sum_{i/\sigma_i=-1} \frac{\exp(4 \sum_{j \neq i} \beta \sigma_j)}{1 + \exp(4 \sum_{j \neq i} \beta \sigma_j)}$$

changing  $\sigma_i$  affects the term  $(i,j)$  and the term  $(j,i)$

For  $\sigma_i=+1$ ,

$$\sigma_i - \tanh\left(\frac{2\beta}{n} \sum_{j \neq i} \sigma_j\right) = 1 - \tanh(\dots) = \frac{2e^{-2\beta \epsilon \dots}}{e^{2\beta \epsilon \dots} + e^{-2\beta \epsilon \dots}}$$

$$= \frac{\exp(-4 \sum_{j \neq i} \frac{\beta}{n} \sigma_j)}{1 + \exp(-4 \sum_{j \neq i} \beta \sigma_j)}$$

For  $\sigma_i=-1$ , similar.

So 
$$F(\sigma) = \frac{1}{n} \sum_{i=0}^{n-1} \left[ \sigma_i - \tanh\left(\frac{2\beta}{n} \sum_{j \neq i} \sigma_j\right) \right].$$

So P-a.s, as  $\sum \sigma_i$  and  $\sum \sigma'_i$  differ by at most 2,

$$|F(\sigma) - F(\sigma')| \leq \frac{1}{n} \left| \sum_{i=0}^{n-1} \sigma_i - \sum_{i=0}^{n-1} \sigma'_i \right| + \frac{1}{n} \sum_{i=0}^{n-1} \left| \tanh\left(\frac{2\beta}{n} \sum_{j \neq i} \sigma_j\right) - \tanh\left(\frac{2\beta}{n} \sum_{j \neq i} \sigma'_j\right) \right|$$

$$\leq \frac{2}{n} + \frac{1}{n} \cdot n \cdot \frac{4\beta}{n}$$

$$\leq \frac{2+4\beta}{n}$$

$$\leq \left| \frac{2\beta}{n} \sum_{j \neq i} \sigma_j - \frac{2\beta}{n} \sum_{j \neq i} \sigma'_j \right|$$

as  $|\tanh(x)| \leq 1$  so  $|\tanh(x) - \tanh(y)| \leq |x - y|$ .

$$\leq \frac{4\beta}{n}$$

3) 
$$E[F(\sigma)g(\sigma)] = E[E[F(\sigma, \sigma') | \sigma] g(\sigma)] = E[F(\sigma, \sigma') g(\sigma)]$$

$$= E[F(\sigma, \sigma') g(\sigma')] = -E[F(\sigma, \sigma') g(\sigma')] = -E[F(\sigma, \sigma') g(\sigma)]$$

$$E[F(\sigma)g(\sigma)] = \frac{1}{2} E[F(\sigma, \sigma') (g(\sigma) - g(\sigma'))]$$

(10) 4) We apply for  $g=f$ :

$$E[\beta^2(\sigma)] = \frac{1}{2} E[F(\sigma, \sigma') (f(\sigma) - f(\sigma'))]$$

$$\leq \frac{1}{2} E[|F(\sigma, \sigma')| |f(\sigma) - f(\sigma')|]$$

$$\leq \frac{2+4\beta}{n}$$

$|F(\sigma, \sigma')| \leq 2$   
p.o.s.  
and 2)

Moreover,  $E[\beta(\sigma)] = E[F(\sigma, \sigma')] = E[F(\sigma, \sigma)'] = -E[F(\sigma, \sigma)'] = 0$

So  $f$  is centered and has variance  $\leq \frac{2+4\beta}{n} \rightarrow 0$  as  $n \rightarrow \infty$ .

By Tchebitchev,

$$\mu_{\beta, n}(|f| > \epsilon) \leq \frac{1}{\epsilon^2} \frac{2+4\beta}{n} \rightarrow 0$$

$$5) |f(\sigma) - (M_n - \text{th}(2\beta M_n))|$$

$$= \frac{1}{n} \left| \sum_{i=0}^{n-1} \text{th}(2\beta M_n) - \text{th}\left(\frac{2\beta}{n} \sum_{j=i}^{n-1} \sigma_j\right) \right|$$

$$\leq \frac{1}{n} \sum_{i=0}^{n-1} \left| \text{th}\left(\frac{2\beta}{n} \sum_{j=0}^{n-1} \sigma_j\right) - \text{th}\left(\frac{2\beta}{n} \sum_{j=i}^{n-1} \sigma_j\right) \right|$$

$$\leq \left| \frac{2\beta}{n} \sum_{j=0}^{n-1} \sigma_j - \frac{2\beta}{n} \sum_{j=i}^{n-1} \sigma_j \right| \leq \frac{6\beta}{n}$$

$$\leq \frac{6\beta}{n}$$

(11) (9)  $\forall \epsilon > 0,$

$$\mu_{\beta, n}(|M_n - \text{th}(2\beta M_n)| > 2\epsilon)$$

$$\leq \mu_{\beta, n}(|f(r)| > 2\epsilon - \frac{4\beta}{n}) \text{ by } \underline{5)}$$

$$\leq \mu_{\beta, n}(|f(r)| > \epsilon) \text{ For } n \text{ large enough}$$

$$\xrightarrow[n \rightarrow \infty]{} 0 \text{ by } \underline{4}).$$

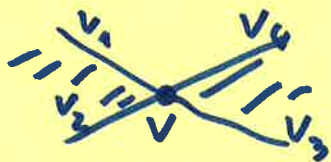
This implies that with high probability,

$M_n$  is close to one of the ~~Fixed~~ <sup>Zeros</sup> points of  $x \mapsto x - \text{th}(2\beta x) \dots$  But it doesn't tell us which one! For  $\beta \leq \frac{1}{2}$ , we do get that  $M_n$  is close to 0 (there is only one Fixed point).

For  $\beta > \frac{1}{2}$ , the Fixed points are 0 and  $\pm m^*(\beta)$ .

One idea would be to re-do this with  $h \neq 0$  (everything works), as then there is only one Fixed point with the same sign as  $h \dots$

Ex 5.1 For  $v \in V$ , let  $v_1, v_2, v_3, v_4$  be the 4 edges surrounding  $v$ :



in c-cw order starting from a white  $\rightarrow$  black transition.

We try to write  $w_{v_i} = w_i \dots$

$$w_v(w) = \hat{a}^{\circ}(w_1 w_4 + w_2 w_3) + \hat{b}^{\circ}(w_1 w_2 + w_3 w_4) + \hat{c}^{\circ}(w_1 w_2 w_3 w_4 + 1) + \hat{d}^{\circ}(w_1 w_3 + w_2 w_4)$$

(12) This gives  $\begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} = Z \begin{pmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & 1 & -1 \\ 1 & 1 & 1 & 1 \\ -1 & -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix}$   
 test every possibility of  $w_1, w_2, w_3, w_4$

$$\Leftrightarrow \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix} = \frac{1}{8} \begin{pmatrix} 1 & -1 & 1 & -1 \\ -1 & 1 & 1 & -1 \\ 1 & 1 & 1 & 1 \\ -1 & -1 & 1 & 1 \end{pmatrix} \begin{pmatrix} a \\ b \\ c \\ d \end{pmatrix}$$

so it works.

Then,  $Z(a, b, c, d) = \sum_{\omega \in \Omega_e} \prod_{v \in V} w_v(\omega)$

$$= \sum_{\omega \in \Omega} \prod_{v \in V} w_v(\omega)$$

$\Omega = \{\pm 1\}^E$   
 we take all  $\Omega = \{\pm 1\}^E$   
 but we set  $w_v(\omega) = 0$  when  $\omega$  is wrong at  $v$ .

$$= \sum_{\omega \in \Omega} \prod_{v \in V} \hat{a}(w_{v_1}, w_{v_2}, w_{v_3}, w_{v_4}) + \hat{b}(\dots) + \hat{c}(\dots) + \hat{d}(\dots)$$

We develop the product. This gives a sum on subsets of edges, where every edge can possibly be taken twice (once for each of its endpoints).  
 Switching with  $\sum_{\omega}$ , we see that each edge has to be taken 0 or 2 times (otherwise the sum is 0).

The remaining terms from the expansion can therefore be seen as subsets of edges (the ones taken twice, let's say) and since at every  $v \in V$  we made a choice in the product that is compatible with the local rule, we get

(13)

that the remaining terms are in fact elements of  $\underline{\underline{\Omega_e}}$ .

All in all,

$$Z(a,b,c,d) = 2^{|E|} Z(\hat{a}, \hat{b}, \hat{c}, \hat{d})$$

sum on  $\Omega$  for surviving terms.

By the "handshake lemma",  $4|V| = \sum_{v \in V} \deg v = 2|E|$ .

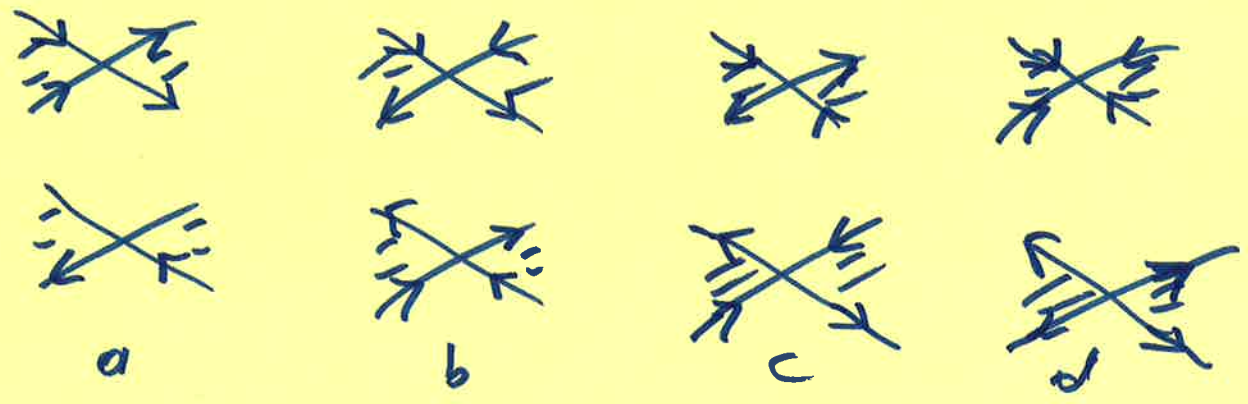
So  $2^{|E|} = 2^{2|V|}$ , and ~~also~~ so

$$\begin{aligned}
Z(a,b,c,d) &= \sum_{\omega \in \Omega_e} \prod_{v \in V} 4 \hat{w}_v(\omega) \\
&= Z(4\hat{a}, 4\hat{b}, 4\hat{c}, 4\hat{d}) \\
&= Z(a', b', c', d').
\end{aligned}$$

2) Yes. The two "d" config have to appear the same number of times.

To see that, orient every edge  $e \in E$  with black on its right when  $w_e = +1$ , and white on its right when  $w_e = -1$ .

The local possibilities become



16) Then the two "d" config are "sinks" and "sources", there has to be the same number of these.

So the transformation  $(a, b, c, d) \mapsto (a, b, c, -d)$  also leaves  $Z$  invariant.

And of course, all combinations of these two (you can check that they generate a group isomorphic to  $S_3$ ).

3) For instance, take an Ising model at temperature  $\beta$  on black faces, and  $\beta'$  on white faces.

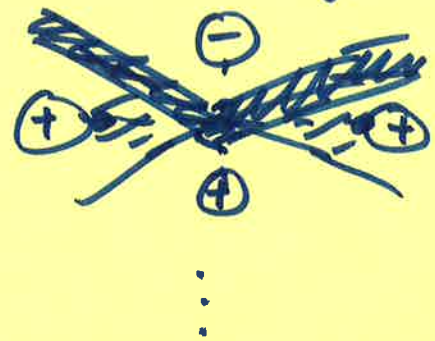
Then the  $+/-$  ~~are~~ interfaces are an eight-vertex model, with

$$a = \exp(\beta - \beta')$$

$$b = \exp(-\beta + \beta')$$

$$c = \exp(\beta + \beta')$$

$$d = \exp(-\beta - \beta').$$



You can even couple the two models, with 4-spins interactions...