

## Lecture 5.

Recap:

$\mathbb{P}_p$  - percolation measure  
on  $\mathbb{Z}^d$ ,  $d \geq 2$

$$\mathbb{P}_p(\infty \leftrightarrow \infty) = 0 \quad p_c \quad \mathbb{P}_p(0 \leftrightarrow \infty) > 0$$

We follow the proof of  
Demirici - Cepin and Tassion -  
for sharpness of the phase  
transition.

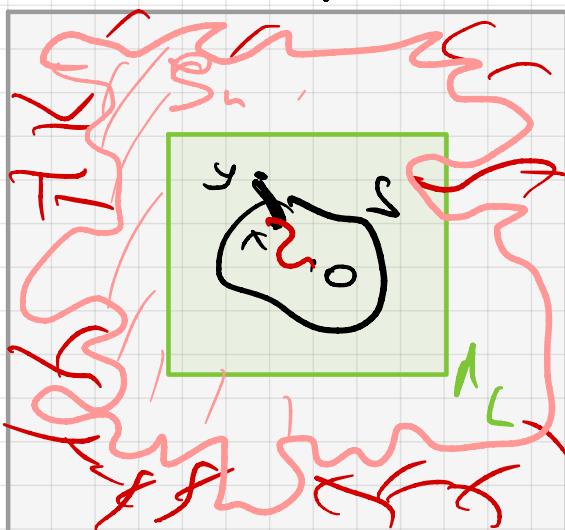
In exercises:

- $\Theta_n(p) := \mathbb{P}_p(0 \leftrightarrow \partial A_n)$

Take  $L < n$ ,  $S \subset A_{L-1}$ .

Then,

$$\Theta_n(p) \leq \Theta_{n-L}(p) \cdot p \sum_{x,y \in \Delta S} \mathbb{P}_p(0 \leftrightarrow x)$$



$$\Delta S := \{xy \in E : x \in S, y \notin S\}$$

$$\Theta_n(p) = \frac{1}{1-p} \left( E_p \sum_{x \in \Delta S_n} P_p(0 \rightarrow x) \right),$$

where  $S_n := \{x \in A_n : x \leftrightarrow 2A_n\}$

Key idea:

Define  $p_c$  differently!  
We will use a finite counter.

Def.

For  $S \subset \mathbb{Z}^d$  finite set of vertices  
and  $p \in (0, 1)$ , define

$$\ell_p(S) := p \sum_{x \in \Delta S} P_p(0 \rightarrow x)$$

Take

$$\tilde{p}_c := \sup \{p \in (0, 1) : \exists 0 \in S \subset \mathbb{Z}^d \text{- finite}, \\ \text{s.t. } \ell_p(S) < 1\}$$

Res. If  $0 \notin S$ , then  $\ell_p(S) = 0$ .

Proof (Show guess):

We prove it for  $\tilde{p}_c$ .

Part I: exp. decay below  $\tilde{P}_c$ .

Take any  $P < \tilde{P}_c$ .

By definition, there exists  $0 < S < 2^d$  finite, s.t.

$$\varphi_P(S) < 1.$$

Take  $L$ , s.t.  $S < \alpha_{L-1}$ .

Use our inequality for  $n=k \cdot L$ :

$$\Theta_{kL}(P) \leq \Theta_{(k-1)L}(P) \cdot \varphi_P(S)$$

iterate

$$\leq \Theta_{(k-2)L}(P) [\varphi_P(S)]^2$$

$$\dots \leq \Theta_L(P) [\varphi_P(S)]^{k-1}$$

$$\leq \varphi_P(S)^{k/2} = [\varphi_P(S)]^{n/2L}$$

Since  $\varphi_P(S) < 1$ , there exists  $C > 0$ , s.t.

$$\varphi_P(S) \leq e^{-C \cdot 2L}$$

Then,  $\Theta_n(P) \leq e^{-C \cdot n}$ .

Part II:

- Take  $P > \bar{P}_c$ .  $S_n := \{x \in d_n : x \notin \partial M\}$

We know:

$$\Theta_n'(P) = \frac{1}{P(1-P)} E_P [\varphi_P(S_n)]$$

That is

$$\varphi_P(S_n) \geq \prod_{o \in S_n} \begin{cases} \text{equals } 0 & \text{if } o \in S_n \\ 1 & \text{if } o \notin S_n \end{cases} \geq 1$$

Then:

$$\begin{aligned} \Theta_n'(P) &\geq \frac{1}{P(1-P)} \cdot E_P [\prod_{o \in S_n}] \\ &= \frac{1}{P(1-P)} \cdot P_P(o \in S_n) \\ &= \frac{1}{P(1-P)} \cdot (1 - \Theta_n(P)) \end{aligned}$$

This is a differential eq.  
or  $\Theta_n'(P)$ !

Then:

$$\frac{\Theta_n'(P)}{1 - \Theta_n(P)} \geq \frac{1}{P(1-P)}$$

$$\log \left( \frac{1}{1 - \Theta_n(P)} \right)' \geq \log \left( \frac{P}{1-P} \right)'$$

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Integrate it between  $\tilde{p}_c$  and  $p$ :

$$\frac{1}{1-\Theta_n(p)} - \underbrace{\frac{1}{1-\Theta_n(\tilde{p}_c)}}_{\frac{1}{1-\tilde{p}_c}} \geq \frac{p}{1-p} - \frac{\tilde{p}_c}{1-\tilde{p}_c}$$

$\frac{1}{1-\tilde{p}_c}$

$$\frac{\Theta_n(p)}{1-\Theta_n(p)} \geq \frac{p - \tilde{p}_c}{(1-p)(1-\tilde{p}_c)}$$

$$\Theta_n(p) \geq \frac{p - \tilde{p}_c}{(1-p)(1-\tilde{p}_c) + p - \tilde{p}_c} \geq C(p - \tilde{p}_c)$$

Tend  $n \rightarrow \infty$  and  
get the bound.

Put together Parts I & II:  
 $p_c = \tilde{p}_c$



## 8. Harris-FKG inequality

Edges in percolation are open/closed independently. Thus, for events  $A$  and  $B$  that are determined by disjoint sets of edges, we have

$$P_p(A \cap B) = P_p^2(A) - P_p(A)$$

What if  $A$  and  $B$  depend on the same edges?

For instance:

$$A := \boxed{\text{ } \text{ } \text{ }} = \left\{ \begin{array}{l} \text{left-right} \\ \text{crossing} \end{array} \right\}$$

$$B := \boxed{\text{ } \text{ } \text{ }} = \left\{ \begin{array}{l} \text{top-bottom} \\ \text{crossing} \end{array} \right\}$$

Can we estimate  $P_p(A \cap B)$ ?

Yes - if  $A$  and  $B$  are increasing!

## Theorem (Harris '60)

Let  $A, B$  be two decreasing events. Then,

$$P_p(A \cap B) \geq P_p(A) \cdot P_p(B).$$

More generally, if  $f, g$  are decreasing functions that are bounded, then they are positively correlated!

$$E_p(f \cdot g) \geq E_p(f) \cdot E_p(g).$$

Meaning:-

knowing that one increasing event occurs, increases the probability that any other increases event occurs.

Indeed, if  $P_p(B) > 0$ , then dividing by  $P_p(B)$ , we get:-

$$\frac{P(A|B)}{P_p} = \frac{P_p(A \cap B)}{P_p(B)} \geq P_p(A)$$

Since  $P_p(A|B)$  is the conditional probability of  $A$  given  $B$ .

Not surprising:

since  $B$  occurs, there are a lot of open edges.

These edges help  $A$  to occur

Recall:

This holds for a big family of models — it was shown for the first time by Fortuin-Kasteleyn-Ginibre (FKG inequality)

Proof (Kortis' inequality):

It's enough to prove if only for increasing bdd  $f, g$ :

we can take  $f := \mathbb{1}_A, g := \mathbb{1}_B$ .

$$\mathbb{E}_p(\mathbb{1}_A \cdot \mathbb{1}_B) \geq \mathbb{E}_p(\mathbb{1}_A) \cdot \mathbb{E}_p(\mathbb{1}_B)$$

$$P_p(A \cap B)$$

$$P_p(A)$$

$$P_p(B)$$

Sketch:

- reduce to the case where  $f, g$  depend on finite sets of edges;  $n$  edges
  - $n = 1$ : exact computation
  - $n - 1 \rightarrow n$ : induction step via conditional expectation
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Number all edges:

$$E = \{e_i\}_{i \geq 1}$$

For a configuration  $\omega \in \Omega$ ,  $\mathbb{F}^{\omega}$ :

$$\omega_i := \omega_{e_i}$$

Define a family of  $\sigma$ -algebras:

$$\mathcal{F}_n := \sigma(\{\omega_1, \dots, \omega_n\})$$

This is a filtration;

$$\mathcal{F}_1 \subseteq \mathcal{F}_2 \subseteq \mathcal{F}_3 \subseteq \dots$$

Step 1: Enough to prove for  $f, g$  that depend on finitely many edges.

Proof (Step 5):

Define  $f_n := \underbrace{E_p(f | F_n)}_{\text{this is a rand. var. which is } F_n\text{-measurable}}.$

this is a rand. var.  
which is  $F_n$ -measurable

In other words:

$$f_n(\omega_1, \dots, \omega_n)$$

$$= E_p(f | \omega_1 = \omega_1, \dots, \omega_n = \omega_n)$$

Clearly,  $f_n$  is a martingale.

$$\begin{aligned} E_p(f_{n+\delta} | F_n) &= E_p(E_p(f | F_{n+\delta}) | F_n) \\ &= E_p(f | F_n) = f_n \end{aligned}$$

since  $f$  is bounded, so is  $f_n$ .

Then, by the martingale convergence theorem,

$f_n$  converges a.s. to  $f$ .

exercise

Similarly

$$g_n := E_p(g | F_n) \xrightarrow{n \rightarrow \infty} g \text{ a.s.}$$

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Finally, we also have  
 $f_n g_n \xrightarrow{n \rightarrow \infty} f g$  a.s.

Assume first, for any  $n$ ,

$$\begin{aligned} I(E(f_n g_n)) &\geq I(E(f_n)) \cdot I(E(g_n)) \\ \downarrow & \quad \downarrow \quad \downarrow \\ I(E(fg)) &\geq I(E(f)) \cdot I(E(g)) \end{aligned}$$

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### Step 2:

Show that if  $f, g: [0, 1] \rightarrow \mathbb{R}$  are increasing, then

$$I(E_p(fg)) \geq I(E_p(f)) \cdot I(E_p(g)).$$

### Proof:

Shifting  $f, g$  by a constant does not change the inequality.

Then we can assume that

$$f(0) = g(0) = 0.$$

Since  $f, g$  are increasing, we have

$$f(s), g(s) \geq 0.$$

Now, let us do the computation

$$E_p(fg) = p \cdot f(s)g(r) + \underbrace{(1-p)f(0)g(r_0)}_{=0}$$

$$E_p(f) = p \cdot f(1) + \underbrace{(1-p)f(0)}_{=0}$$

$$E_p(g) = p \cdot g(1)$$

Then,

$$E_p(fg) - E_p(f) \cdot E_p(g)$$

$$= p \cdot f(1)g(s) - p \cdot f(s) \cdot p \cdot g(1)$$

$$= \underbrace{f(r)g(r)}_{\geq 0} \cdot \underbrace{(p-p^2)}_{=0} \geq 0$$

Step 2:

Assume that we know the inequality for functions that depend on  $n-1$  edges.

Show it for

$$f, g : \{0, 1\}^{\{e_1, \dots, e_n\}} \rightarrow \mathbb{R}.$$

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Proof:

We first fix

$$(\omega_1, \dots, \omega_{n-1}) = \{ \underbrace{\xi \in \{0, 1\}}_{\text{fixed}} \}_{i=1, \dots, n-1}$$

and then we average over them.

$$\mathbb{E}_P(fg) = \sum_{\substack{\xi \in \{0, 1\} \\ \zeta \in \{0, 1\}}} f(\xi, \zeta_n) g(\xi, \zeta_n) P_p(\xi, \zeta_n) P_p(\xi) \cdot P_p(\zeta_n)$$

$$= \sum_{\xi} \left( \sum_{\zeta \in \{0, 1\}} f(\xi, \zeta_n) g(\xi, \zeta_n) P_p(\zeta_n) \right) \cdot P_p(\xi)$$

$\underbrace{\mathbb{E}_P}_{\substack{\text{average} \\ \text{over } \zeta_n}} \left( \underbrace{f(\xi, \zeta_n)}_{\substack{\text{increasing} \\ \zeta \in \{0, 1\}}} \cdot \underbrace{g(\xi, \zeta_n)}_{\substack{\text{---} \\ \text{---}}} \right)$

$\xi$ -fixed

apply Hoeffding's  
(ineq. for  $n=1$ )

$$\mathbb{E}_P^{\zeta_n} (f(\xi, \zeta_n)) \cdot \mathbb{E}_P^{\zeta_n} (g(\xi, \zeta_n))$$

exercise

$$F_{n-1}(\xi)$$

$$g_{n-1}(\xi)$$

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Thus, we have:

$$\mathbb{E}_P(f \cdot g) \geq \sum_{\xi \in \{0,1\}^{k_1 \dots k_n}} f_{n-1}(\xi) \cdot g_{n-1}(\xi) \cdot P_P(\xi)$$

$$= \mathbb{E}_P(f_{n-1} \cdot g_{n-1})$$

Now consider  
the case  
for  $n-1$

$$\geq \mathbb{E}_P(f_{n-1}) \cdot \mathbb{E}_P(g_{n-1})$$

$$= \mathbb{E}_P(f) \cdot \mathbb{E}_P(g)$$



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