

# Contracting Clusters of Critical Percolation

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Phenomena in High Dimensions, Samos 2007

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# Definition of $p_c$

- ▶ Let  $G$  be any infinite graph. Let  $0 \leq p \leq 1$ . Consider the random graph  $G_p$  that one gets by keeping every edge of  $G$  with probability  $p$ , independently for each edge.

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- ▶ Changing any finite set of edges cannot destroy or create an infinite cluster. Therefore  $\psi(p)$  is either 0 or 1.

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- ▶ Changing any finite set of edges cannot destroy or create an infinite cluster. Therefore  $\psi(p)$  is either 0 or 1.
- ▶ Therefore there exists some  $p_c$ , depending on  $G$ , such that  $\psi(p) = 0$  for  $p < p_c$  and  $\psi(p) = 1$  for  $p > p_c$ .

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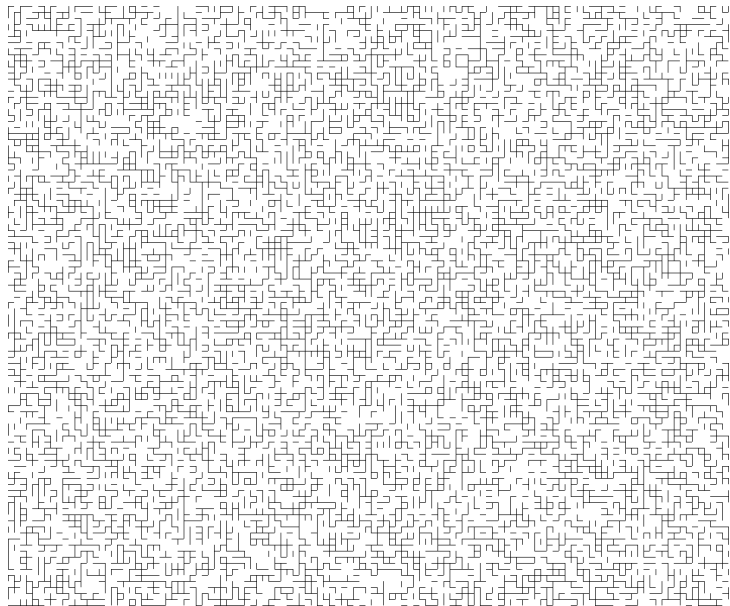
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# Percolation on $\mathbb{Z}^2$ , $p = 0.45$



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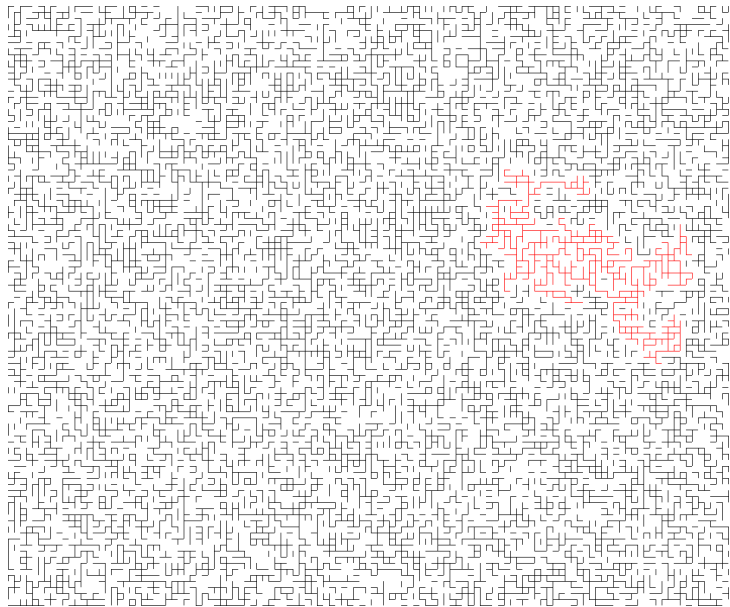
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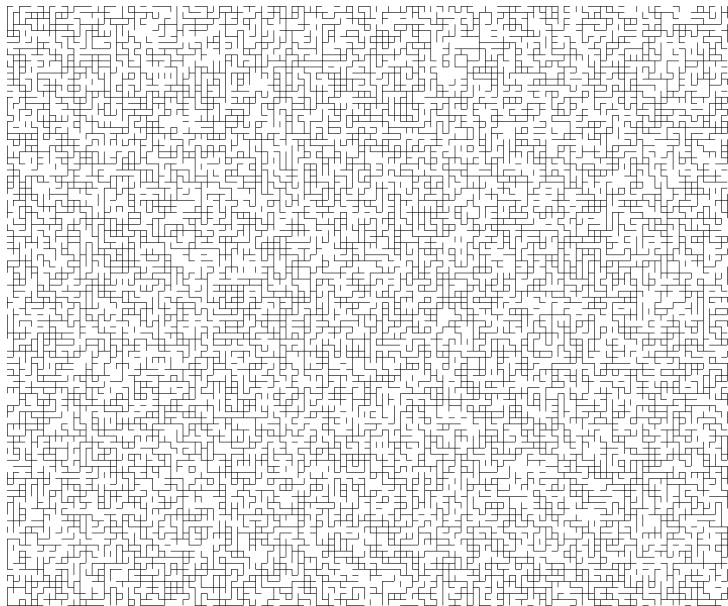
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# Percolation on $\mathbb{Z}^2$ , $p = 0.55$



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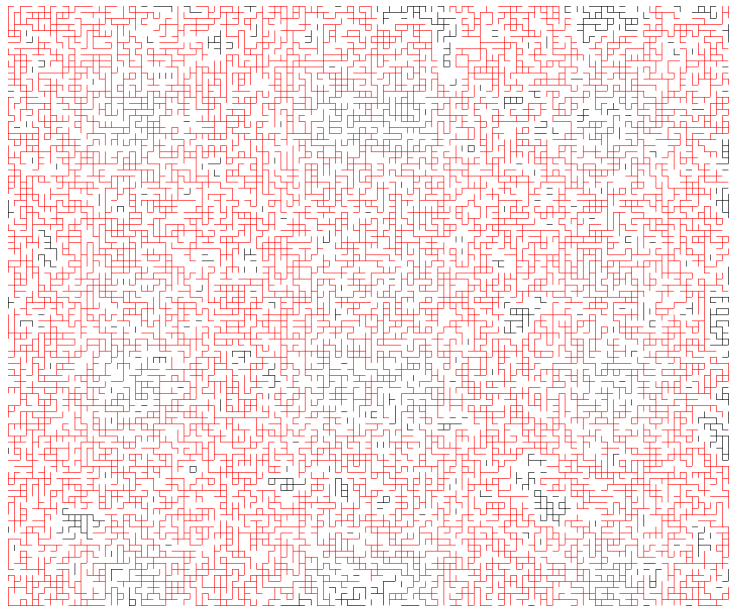
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# Percolation on $\mathbb{Z}^2$ , $p = 0.55$



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- ▶ for  $G = \mathbb{Z}$ ,  $p_c = 1$  and  $\psi(p_c) = 1$  (exercise).

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- ▶ for  $G = \mathbb{Z}$ ,  $p_c = 1$  and  $\psi(p_c) = 1$  (exercise).
- ▶ for  $G$  a  $d$ -regular tree,  $p_c = \frac{1}{d-1}$  and  $\psi(p_c) = 0$ . This is equivalent to a Galton-Watson branching process.

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- ▶ for  $G$  a  $d$ -regular tree,  $p_c = \frac{1}{d-1}$  and  $\psi(p_c) = 0$ . This is equivalent to a Galton-Watson branching process.
- ▶ The complete graph on  $n$  vertices exhibits similar behavior (even though it is finite) with “ $p_c = \frac{1}{n}$ ” and “ $\psi(p_c) = 0$ ”, Erdős & Rényi (1959).

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$$p \neq p_c$$

- ▶ In the subcritical case, component sizes decay exponentially in the volume, i.e. for every  $p < p_c$  there exist some  $\lambda > 0$  such that

$$\mathbb{P}(|\mathcal{C}| > n) \leq e^{-\lambda n}$$

where  $\mathcal{C}$  is the cluster containing the origin. Menshikov (1986), Aizenman & Barsky (1987).

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- ▶ In the supercritical case there exists one infinite cluster (Burton & Keane, 1989). The sizes of the finite clusters decay exponentially in the surface area, i.e. for every  $p > p_c$  there exists some  $\lambda$  such that

$$\mathbb{P}(n < |\mathcal{C}| < \infty) \leq e^{-\lambda n^{(d-1)/d}}$$

Grimmett & Marstrand (1990), Kesten & Zhang (1990).

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- ▶ In most senses, the supercritical cluster “looks like a stretched-out grid”.

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$$\rho = \rho_c$$

Some conjectures coming from the physics literature:

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$$\rho = \rho_c$$

Some conjectures coming from the physics literature:

(a). For  $d > 1$  there is no infinite cluster at the critical  $\rho$ .

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- (a). For  $d > 1$  there is no infinite cluster at the critical  $p$ .
- (b). The size of the critical cluster decays *polynomially\**, i.e.

$$\mathbb{P}(|\mathcal{C}| > n) \approx n^{-1/\delta}$$

for some  $\delta$ .

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- (c). Universality:  $\delta$  depends only on the dimension, and not on the specific grid (unlike, say,  $p_c$ ).

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- (d).  $\frac{91}{5} = \delta_2 > \delta_3 > \dots > \delta_6 = \delta_7 = \dots = 2$ . \*In  $d = 6$  there are logarithmic corrections. The conjecture for the value  $\frac{91}{5}$  is related to a conjecture that the distribution of large finite clusters is conformally invariant.

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What has been proved?

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- ▶  $d = 2$ : a, Kesten (1980), “b, d” Smirnov (2001); Lawler, Schramm & Werner (2001).

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- ▶  $d = 3, 4, 5, 6$ : not even a.

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# Random walk on a supercritical cluster

- ▶ Take some  $p > p_c$  and let  $\mathcal{C}$  be the infinite cluster, conditioned on  $0 \in \mathcal{C}$ . Examine random walk on  $\mathcal{C}$  starting from 0.

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- ▶ Properties that hold for almost any  $\mathcal{C}$  are called “quenched”. Properties that hold after averaging on the environment are called “annealed”.

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- ▶ So does the quenched, Sidoravicius & Sznitman (2004), Barlow (2004), Berger & Biskup (2006), Mathieu & Piatnitski (2006).

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- ▶ So does the quenched, Sidoravicius & Sznitman (2004), Barlow (2004), Berger & Biskup (2006), Mathieu & Piatnitski (2006).
- ▶ Results of the type “ $\mathcal{C}$  is like a grid”.

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# Random walk on the incipient infinite cluster

- ▶ Take critical percolation, and condition on the cluster of the origin  $\mathcal{C}$  to satisfy  $|\mathcal{C}| > n$ . Take  $n \rightarrow \infty$ . It turns out that the distributions of  $\mathcal{C}$  converge in the appropriate sense to a limit, Kesten (1986), van der Hofstadt & Járai (2004). This limit is called the incipient infinite cluster.

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- ▶ Random walk on the IIC is (like on all fractals), *subdiffusive*, that is

$$\mathbb{E}(R(n)) \leq Cn^{1/2-\epsilon}$$

Kesten (1986), Barlow, Járai, Kumagai & Slade (2007).

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- ▶ In  $d > 6$ , the exact exponent is  $\frac{1}{3}$ .

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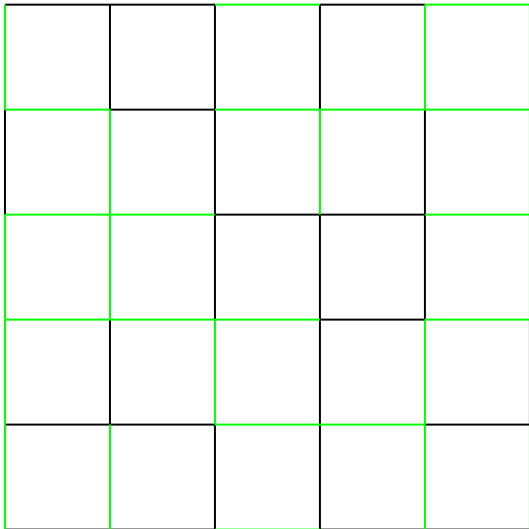
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No edges are removed, edges are only colored in two colors.



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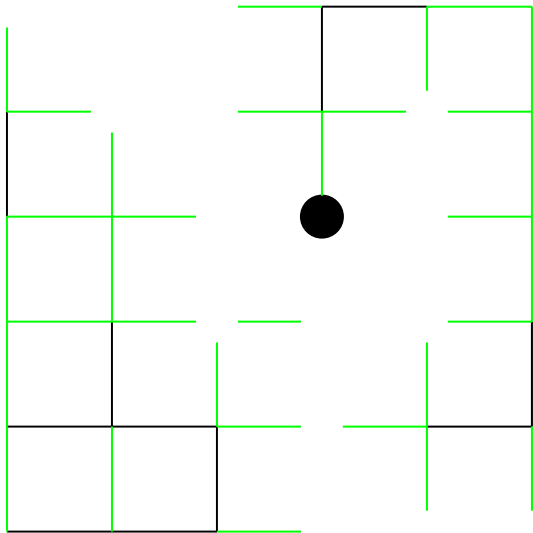
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Take a black cluster and replace it with a single vertex.



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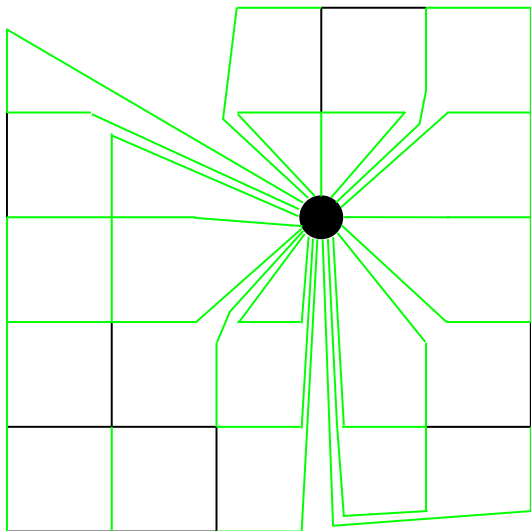
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Connect it to all edges which used to connect to the cluster.



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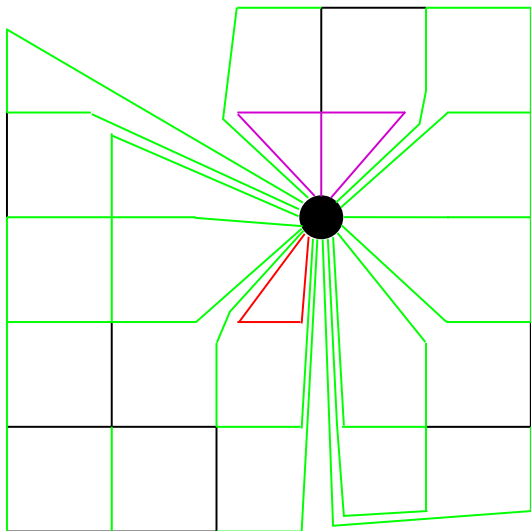
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Note that this can create loops and multiple edges.



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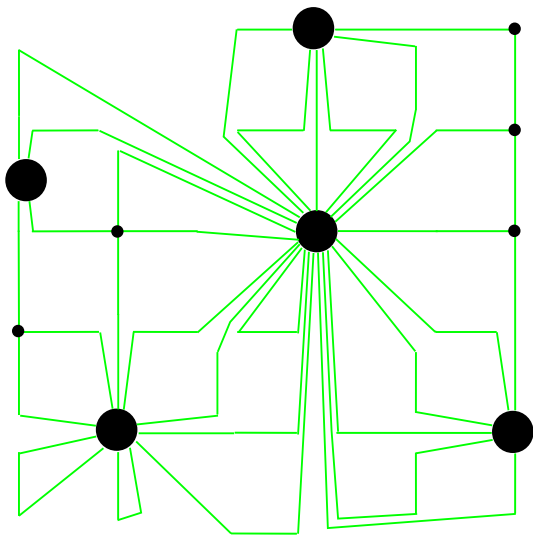
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Repeat for all clusters.



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# Which $p$ ?

- ▶ Formally, for every edge, independently and with probability  $p$ , identify its two end points. Call the resulting graph  $\text{CCP}_p$ .

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- ▶ If  $p < p_c$  the contracted clusters are small and do not affect the random walk on  $CCP_p$  significantly. This case would be amenable to the same techniques used to analyze random walk on the supercritical cluster.

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# Which $p$ ?

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- ▶ Hence we will focus on  $p = p_c$ , in which case we will call the graph CCCP.

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# Geometry

We have results for both  $d = 2$  and  $d > 6$ , but in this lecture we will focus on  $d > 6$ .

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- ▶ For any two  $x, y \in \mathbb{Z}^d$ , let  $d(x,y)$  denote the graph distance between  $x$  and  $y$ , i.e. the length of the shortest path in our graph. Then

$$d(x, y) \approx \log \log |x - y|.$$

For comparison, on a supercritical cluster,  $d(x, y) \approx |x - y|$ , while on the IIC  $d(x, y) \approx |x - y|^2$ .

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- ▶ The graph satisfies the same *isoperimetric inequality* as  $\mathbb{Z}^d$ , i.e. for any finite  $A \subset G$

$$|\partial A| \geq c|A|^{(d-1)/d}$$

where  $|\partial A|$  is the number of edges going out of  $A$ , and  $|A| = \sum_{v \in A} \deg v$  i.e. the total number of edges going out of vertices of  $A$ .  $\frac{d-1}{d}$  is sharp.

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## Random walks

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- The speed of the random walk on the graph, *measured in the Euclidean distance*, is  $\sqrt{n \log n}$  i.e.

$$\mathbb{E}(|R(n)|) \approx \sqrt{n \log n}.$$



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- ▶ The spectral gap of the Laplacian on a ball of radius  $R$  is between  $\frac{c}{R^2}$  and  $\frac{C \log R}{R^2}$ .

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- ▶ When  $d > 12$  this spectral gap is concentrated near its average. Uses the concentration property of Lipschitz functions.

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# Summary of results

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- ▶ The supercritical cluster behaves like the usual grid.
- ▶ The incipient infinite cluster behaves like a critical branching tree, embedded into  $\mathbb{Z}^d$  randomly (this is known as “integrated superbrownian excursion”).
- ▶ CCCP behaves in strange and unexpected ways. We don't have in mind any simple model that would reproduce all data above.

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We are now going to sketch some of the ideas that went into the proof of  $d(x, y) \approx \log \log |x - y|$ .

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- ▶ Any two increasing events are positively correlated, Fortuin, Kasteleyn & Ginibre (1971).

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# Percolation tools

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$$\mathbb{P}(x \leftrightarrow y, y \leftrightarrow z) \geq \mathbb{P}(x \leftrightarrow y)\mathbb{P}(y \leftrightarrow z)$$

where  $x \leftrightarrow y$  is the event that  $x$  and  $y$  belong to the same cluster.

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$$\mathbb{P}(\exists \text{two disjoint paths between } x \text{ and } y) \leq \mathbb{P}(x \leftrightarrow y)^2$$

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- ▶ For  $d > 6$ ,  $\mathbb{P}(x \leftrightarrow y) \approx |x - y|^{2-d}$ . Hara, van der Hofstad & Slade (2003).

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# First moment

- ▶ Let  $x, y \in \mathbb{Z}^d$ , and assume for simplicity that  $d = 8$ . What is the probability that one can jump from  $x$  to  $y$  by no more than 2 moves in CCCP (in general,  $\lfloor \frac{1}{2}d \rfloor - 2$  moves)?

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- ▶ This is equivalent to the existence of an edge  $(v, w)$  such that  $x \leftrightarrow v, w \leftrightarrow y$ .

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- ▶ We can estimate  $\mathbb{E}L$  using the FKG inequality:

$$\begin{aligned} \mathbb{E}L &= \sum_{(v,w)} \mathbb{P}(x \leftrightarrow v, w \leftrightarrow y) \\ &\geq \sum_{(v,w)} \mathbb{P}(x \leftrightarrow v) \cdot \mathbb{P}(w \leftrightarrow y) \end{aligned}$$

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 &\geq c|x - y|^8 \cdot (|x - y|^{-6})^2 = c|x - y|^{-4}
 \end{aligned}$$

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## Second moment

- ▶ Let now  $x_1, x_2, y_1$  and  $y_2 \in \mathbb{Z}^d$ , and assume they are all  $\approx r$  apart. Let  $L_i$  be, as before, the number of edges  $(v_i, w_i)$  such that  $x_i \leftrightarrow v_i$  and  $w_i \leftrightarrow y_i$ . We want to upper-bound  $\mathbb{E}L_1L_2 - \mathbb{E}L_1\mathbb{E}L_2$ .

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- ▶ By the BK inequality,

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- ▶ To estimate the last summand, assume for simplicity that it is the connections  $w_1 \leftrightarrow y_1$  and  $w_2 \leftrightarrow y_2$  that occur non-disjointly.

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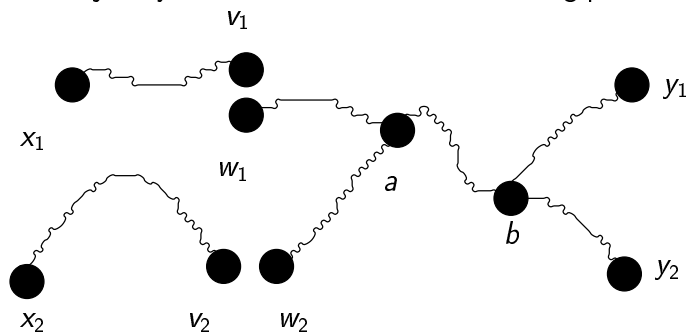
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## Second moment II

Examine the event that  $w_1 \leftrightarrow y_1$  and  $w_2 \leftrightarrow y_2$  occur non-disjointly. This must be as in the following picture:



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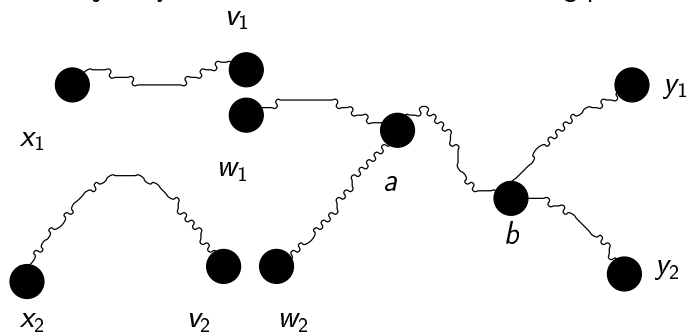
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Formally: there must exist  $a$  and  $b$  such that  $w_i \leftrightarrow a$ ,  $a \leftrightarrow b$ ,  $b \leftrightarrow y_i$  all disjointly.

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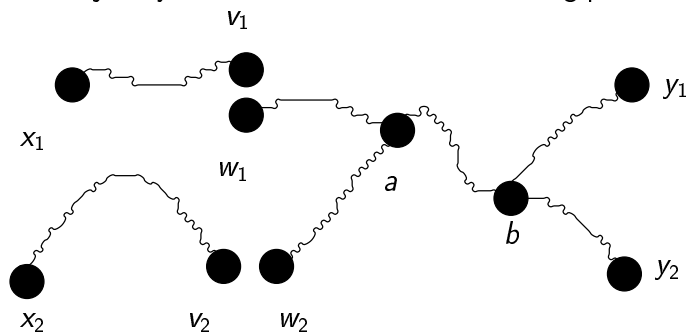
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Formally: there must exist  $a$  and  $b$  such that  $w_i \leftrightarrow a$ ,  $a \leftrightarrow b$ ,  $b \leftrightarrow y_i$  all disjointly. We use the BK inequality again and get that for each tuple  $(v_1, v_2, w_1, w_2, a, b)$  the probability is  $\leq (|x_1 - v_1| \cdot \dots \cdot |b - y_2|)^{-6}$ .

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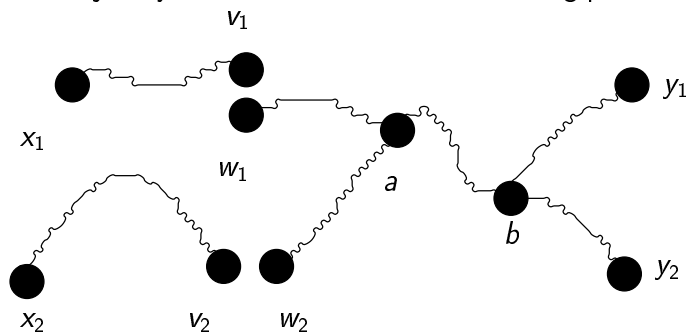
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- ▶ Recapitulating, we got that  $\mathbb{E}L \approx r^{-4}$  while  $\text{cov}(L_1, L_2) \approx r^{-10}$ .

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- ▶ Recapitulating, we got that  $\mathbb{E}L \approx r^{-4}$  while  $\text{cov}(L_1, L_2) \approx r^{-10}$ .
- ▶ Therefore if we have two large clusters at scale  $r$ , each has size  $\approx r^4$  and therefore the expected number of connections is  $\approx (r^4)^2 \cdot r^{-4} = r^4$  while the variance is only  $\approx (r^4)^4 \cdot r^{-10} = r^6$ . We get that they are connected with probability  $> 1 - cr^{-2}$ .

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- ▶ Similar diagrammatic bounds show the whole log log result. Roughly we show that at scale  $r$  there are clusters which go as far as  $r^2$ , so you can move between scales with a bounded number of jumps. We omit all further details.

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Thank you

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