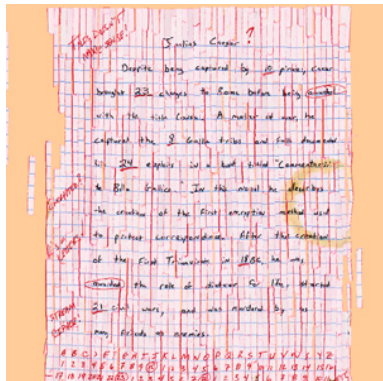


# Jigsaw percolation

Joint work with David Sivakoff

**Univerza na Primorskem**

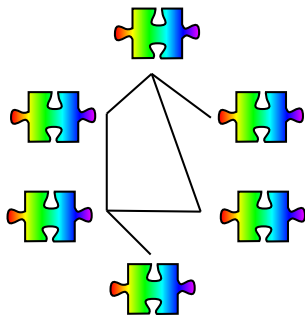
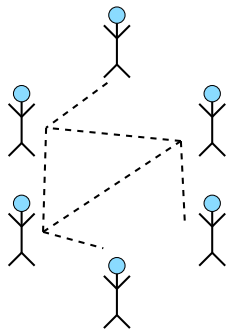
May 29, 2017



- 2011 DARPA shredder challenge: UCSD team used crowdsourcing to piece together shredded paper.
- Polymath Project: Timothy Gowers' experiment with "massively collaborative mathematics."
- How might people *cooperatively* combine their individual ideas to solve a problem?
- Coagulation model to form coalitions: discrete Smoluchowski dynamics with additional restrictions, say kin relations or spatial proximity.

# People and ideas (puzzle pieces)

A dynamic on two graphs with the same vertex set but different edges, introduced in a 2015 paper by Brummitt, Chatterjee, Dey, and Sivakoff, henceforth referred to as BCDS.



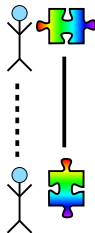
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- If two people know each other...



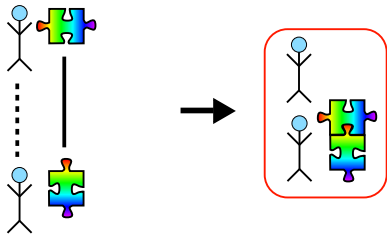
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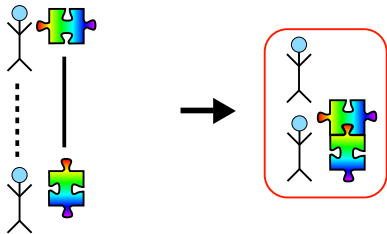
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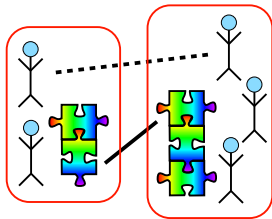


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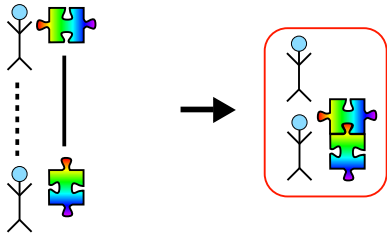


- Generally, if two groups with merged ideas **know each other** and **have compatible ideas**...

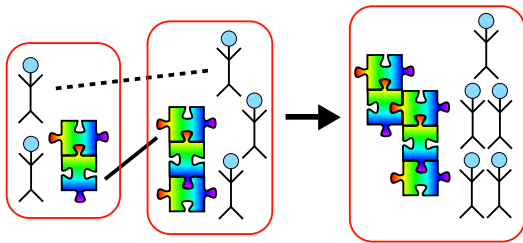


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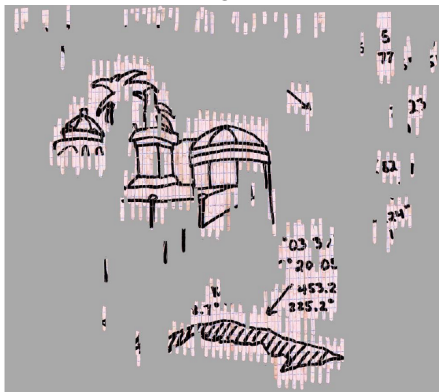
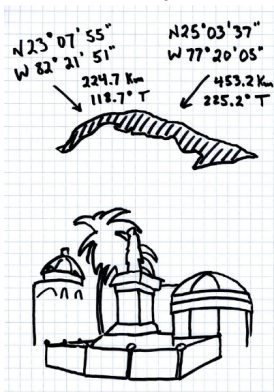
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# People and ideas (puzzle pieces)

This generates larger and larger partial solutions, as in this intermediate step from the DARPA challenge.

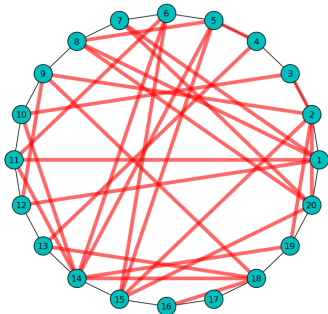


# Jigsaw percolation model

The vertex set  $V$  has  $N$  vertices.

- People Graph: Erdős-Rényi random graph  $(V, E_{\text{ppl}}) \sim G(N, p)$ .
- Puzzle Graphs: Connected deterministic graphs  $(V, E_{\text{puz}})$ .

Example:  $N = 20$ ,  
 $p = 0.15$ , and  $(V, E_{\text{puz}})$   
is the ring graph on 20  
vertices (i.e.,  $\mathbb{Z}_{20}$ ).



How connected must the people graph be to solve the puzzle?

# Basic jigsaw percolation

Start by the partition of  $V$  into singleton clusters. Iteratively create coarser partitions by merging existing clusters.

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Merge two clusters,  $W_1$  and  $W_2$ , if there is a puzzle edge between a pair  $v_1 \in W_1, v_2 \in W_2$ ; and a people edge between a pair  $v'_1 \in W_1, v'_2 \in W_2$ .

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The event `Solve` happens if eventually all vertices are in the same cluster.

# Adjacent-Edge (AE) jigsaw percolation

## Adjacent-Edge Merging Rule [BCDS]

Merge any two clusters,  $W_1$  and  $W_2$ , if there is a vertex  $v_1 \in W_1$  and vertices  $v_2, v'_2 \in W_2$  such that there is a puzzle edge between  $v_1$  and  $v_2$ , and a people edge between  $v_1$  and  $v'_2$ .

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**Open question:** Is Adjacent-Edge jigsaw percolation distinguishable from basic jigsaw percolation?

That is, are in some substantial way group connections more important than individual ones?

Our results (and their proofs) apply to both versions.

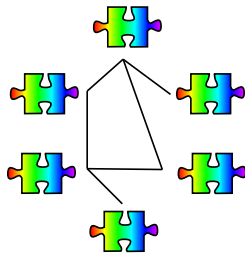
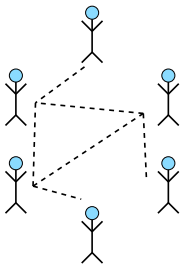
# AE jigsaw dynamics: a simple example

Each person has one unique piece of the puzzle.

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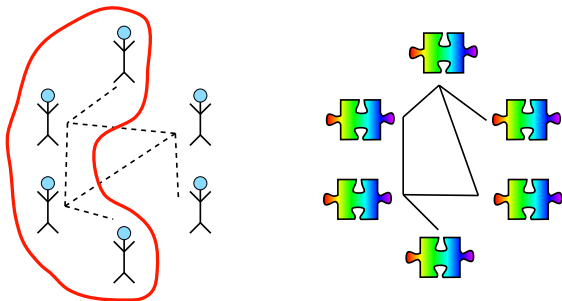
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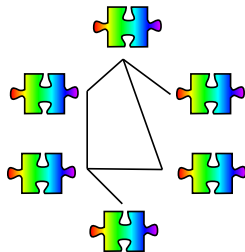
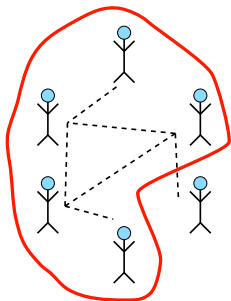


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Successively merge groups that know one another and have compatible puzzle pieces.

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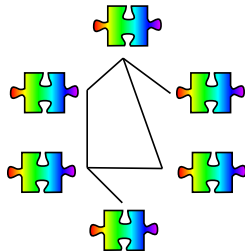
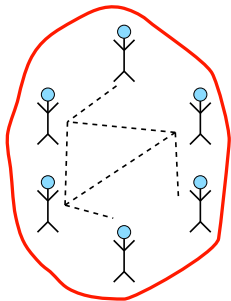
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Solved the puzzle!

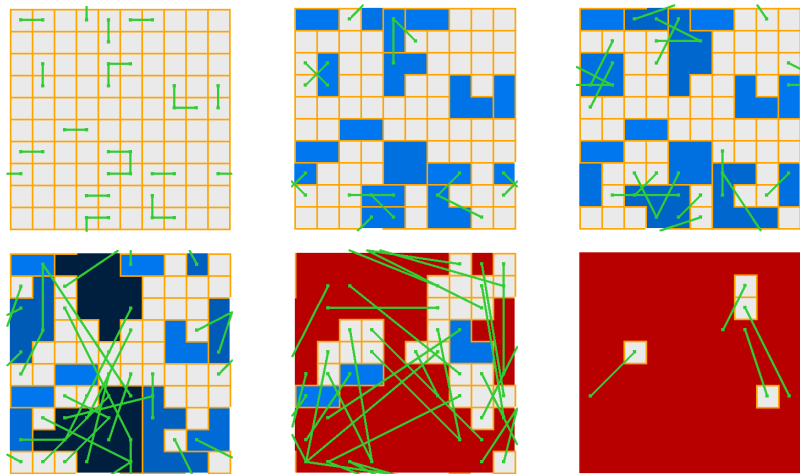


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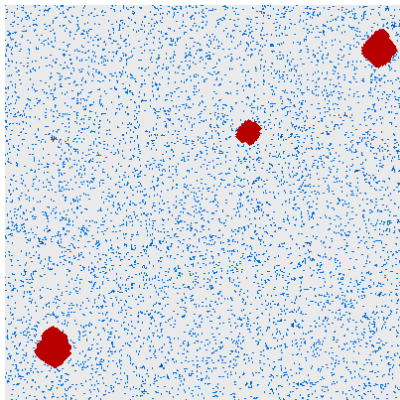
# AE jigsaw dynamics on a torus



Adjacent-Edge JP on  $10 \times 10$  torus (puzzle graph), with  $p = 0.11$  (people graph is  $G(10^2, 0.11)$ ), at times  $t = 0, \dots, 5$ .

# Jigsaw percolation on $\mathbb{Z}_n^2$ : nucleation

AE JP on  $\mathbb{Z}_n^2$  with  $n = 400$ ,  
 $\rho = 0.021$  at  $t = 31$ .



Apparently, solving the puzzle is caused by a local concentration of highly connected individuals that create a gradually growing partial solution by adding boundary pieces. This phenomenon is called **nucleation**.

# Results for general puzzle graphs

Setting: a sequence of connected puzzle graphs on  $N$  vertices and maximum degree  $D$ , with  $N \rightarrow \infty$ ; and Erdős-Rényi people graph with edge density  $p$  on the same vertex set.

## Theorem [BCDS]

If  $p = \lambda / \log N$  with  $\lambda > \pi^2/6$ , then

$$\mathbb{P}_p(\text{Solve}) \rightarrow 1.$$

## Theorem

If  $p = \mu / (D \log N)$  with  $\mu < 1/30$ , then

$$\mathbb{P}_p(\text{Solve}) \rightarrow 0.$$

Corollary: For puzzles of bounded degree, the transition between low and high probability of `Solve` occurs when  $p = \Theta(1/\log N)$ .

## Rule of thumb

On vertex-transitive graphs with  $N$  vertices of degree  $D$ , the transition between low and high  $\mathbb{P}(\text{Solve})$  typically occurs as  $\rho D \log N$  changes between a small constant and a large constant.



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This rule does not hold universally: if  $G_{\text{puz}}$  is the complete graph on  $N$  vertices,  $p \approx 1/(N \log N)$  is too small for connectivity of  $G_{\text{ppl}}$ !

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It does hold for many famous graphs.

## Theorem: Hypercube puzzle

Assume the puzzle graph is the hypercube  $\{0, 1\}^n$ . There exist constants  $c_1, c_2 > 0$  so that

$$\mathbb{P}(\text{Solve}) \rightarrow \begin{cases} 0 & \text{when } p \leq c_1/n^2 \\ 1 & \text{when } p \geq c_2/n^2 \end{cases}$$

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**Open question:** Sharp constant?

## Proof of the upper bound: local growth

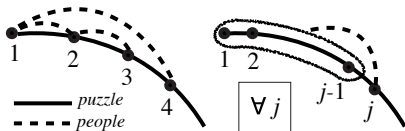
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$$\rho > (1 + \epsilon) \frac{\pi^2}{6 \log N} \implies \mathbb{P}_\rho(\text{Solve}) \rightarrow 1$$

Grow:  $j$  is people-connected to  $\{1, 2, \dots, j-1\}$ ,  
for all  $j \leq K$ .

Then, with  $g(x) = -\log(1 - e^{-x})$ ,

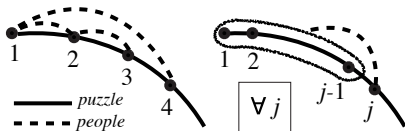


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$$\begin{aligned} \mathbb{P}_p(\text{Grow}) &\geq \prod_{j=1}^{\infty} \left(1 - (1 - p)^j\right) \geq \exp\left(-\sum_j g(pj)\right) \\ &\geq \exp\left(-p^{-1} \int_0^{\infty} g(x) dx\right) = \exp\left(-p^{-1} \cdot \frac{\pi^2}{6}\right) \\ &\geq N^{-1/(1+\epsilon)} \end{aligned}$$

# Proof of the upper bound: unstoppable clusters

Assume  $K \geq C(\log N)^2$  for a large enough  $C$ .

Assume you can create a successful instance of `Grow` on some set  $S$  of  $K$  connected vertices, and thus generate a cluster of size  $K$ . This cluster is **unstoppable**! Why?

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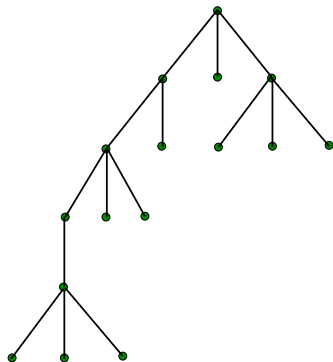
With high probability, all other vertices are people-connected to it:

$$\begin{aligned} & \mathbb{P}_p(\text{some vertex is not people-connected to } S) \\ & \leq N(1-p)^K \leq Ne^{-pK} \leq Ne^{-C \log N} = N^{1-C}. \end{aligned}$$



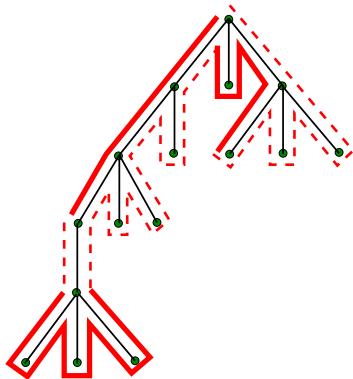
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Take a spanning tree for  $(V, E_{\text{puz}})$ , draw it on the plane,



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Take a spanning tree for  $(V, E_{\text{puz}})$ , draw it on the plane, then doubly-traverse the edges to find  $\Omega(N/K^2) \gg N^{1/(1+\epsilon)}$  edge-disjoint subtrees with  $K$  vertices. (Example:  $K = 4$ .)



## Lower bound: a necessary condition for `Solve`

If  $c$  is small enough, and  $D$  is the maximum degree of puzzle graph,

$$p < \frac{c}{D \log N} \implies \mathbb{P}_p(\text{Solve}) \rightarrow 0$$

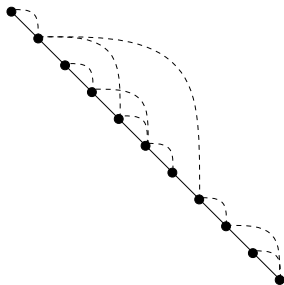
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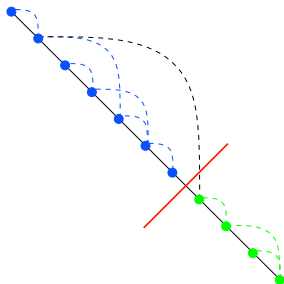
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- 4 Thus  $P(\text{Solve}) \leq N \log N \sup_{\alpha} (3D)^{\alpha \log N} (3c\alpha/D)^{\alpha \log N} \leq N^2 (18c)^{\log N}$ .

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*Sharp transition* occurs if, for every  $\epsilon > 0$ ,

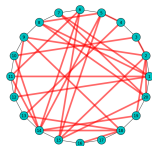
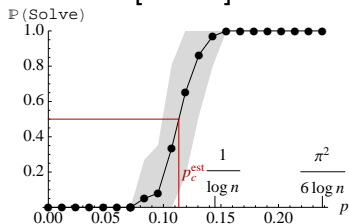
$$\mathbb{P}_{(1-\epsilon)p_c}(\text{Solve}) \rightarrow 0 \text{ and } \mathbb{P}_{(1+\epsilon)p_c}(\text{Solve}) \rightarrow 1.$$

Sharp transition is thought to be a nearly universal phenomenon, due to results by E. Friedgut, G. Kalai, and others. Their theorems do not cover Jigsaw percolation, as they depend on transitivity of random bits.

We can prove sharp transition only when we can establish precise asymptotics for  $p_c$ .

# Sharp transition for the ring puzzle

AE JP on  $\mathbb{Z}_n$  with  
 $n = 1000$ , averaged over  
200 trials [BCSD].



Theorem: ring puzzle.

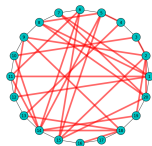
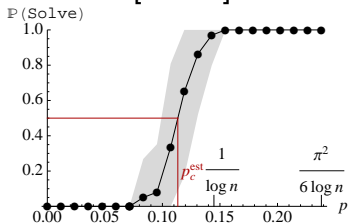
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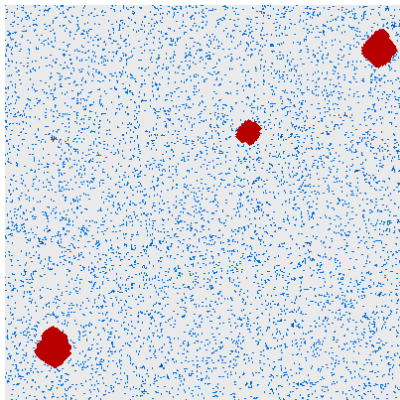
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**Open question:** Note that the sharp transition in the picture is significantly below  $\pi^2/(6 \log n)$ . Why?

# Jigsaw percolation on $\mathbb{Z}_n^2$

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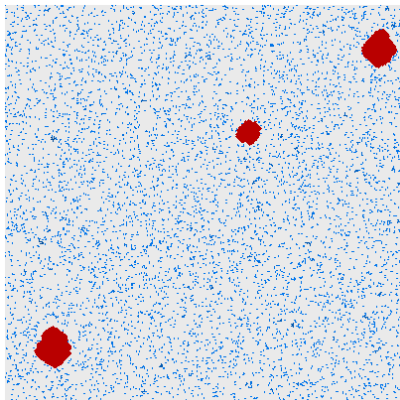
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**Open question:** Can these bounds be improved? Can sharp transition be proved? One obstacle: we know of no useful necessary condition other than connectivity!

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- 2 there is a vertex  $v_1 \in W_1$  with a puzzle neighbor and at least  $\sigma$  people neighbors in  $W_2$ . (Thus  $\sigma$  is the minimal number of “referees” needed to verify that a piece fits.)

## Merging Rules

Merge two clusters,  $W_1$  and  $W_2$ , if at least one of the following holds:

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## Theorem: scaling for large $\sigma$ .

Assume that the sequence of puzzle graphs has bounded degree. For  $N \geq N_0(\sigma)$ ,  $p_c$  is between two constants times  $\sigma^2 / \log N$ .

Requiring a lot of “referees” to verify a fit is very costly!

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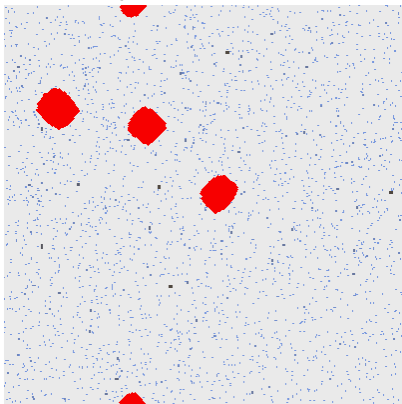
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- 3 there is a  $v_1 \in W_1$  with at least 2 puzzle neighbors in  $W_2$ .  
(On two-dimensional lattice graphs, corner pieces require no verification at all.)

# Free corner fit on $\mathbb{Z}_n^2$

JP ( $\sigma = 1$ , free corner fit) with  $n = 400$ ,  $p = 0.009$ , at  $t = 31$ .



Theorem: 2d-torus JP with free corner fit

Let  $\tau = 1$ ,  $\sigma \geq 1$ ,  $\theta = 2$ , and  $g(x) = -\log(1 - e^{-x})$ . Let

$$\begin{aligned}\lambda_c &= \int_0^\infty g\left(\frac{x^{2\sigma+1}}{\sigma!}\right) dx \\ &= \frac{(\sigma!)^{\frac{1}{2\sigma+1}} \Gamma\left(\frac{1}{2\sigma+1}\right) \zeta\left(\frac{2\sigma+2}{2\sigma+1}\right)}{(2\sigma+1)}.\end{aligned}$$

Then as  $n \rightarrow \infty$ ,

$$p_c(\log n)^{2+\frac{1}{\sigma}} \rightarrow \lambda_c^{2+\frac{1}{\sigma}},$$

with sharp transition.

# Origin of the power of $\log n$

Why is  $p_c \approx (\log n)^{2+1/\sigma}$ ?

- 1 Consider an  $L \times L$  square with  $L \approx p^{-\sigma/(2\sigma+1)} \ll p^{-1/2}$ . Then, the probability that a point on the boundary is  $G_{\text{ppp}}$ -connected to a point inside is on the order  $L(L^2 p)^\sigma = L^{2\sigma+1} p^\sigma \approx 1$ .

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- 3 This probability must exceed  $1/n^2$  for the puzzle to be solved, which gives the claimed power for  $p_c$ .



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