# Branching processes with cousin mergers and locality of hypercube's critical percolation 

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## Making Sense of formulas





## Contents

(1) Percolation

- Critical probability
- Exploration of components
(2) The structure of the hypercube
- Critical probability expansions
- The quest for a heuristic
(3) Branching processes with mergers
- Cousin mergers does not suffice
- A refined collision model


## Percolation

Given an underlying graph, keep each edge independently with prob. p


Critical Probability: Edge density where an giant component appears

- Infinite graphs: $p_{c}:=\inf \{p: P(|C(0)|=\infty)>0\}$


## Phase transition for Erdős-Rényi $G_{n, p}$

Critical Probability: Edge density where a giant component appears


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Critical Probability: Edge density where a giant component appears


If $p=\frac{1}{n}(1+\epsilon)$, whp largest component of size:

- Subcritical $\epsilon^{3} n \rightarrow-\infty: \quad L_{1}\left(G_{n, p}\right)=O(\log n)$
- Critical $\epsilon^{3} n \rightarrow a \in \mathbb{R}: \quad L_{1}\left(G_{n, p}\right)=\Theta\left(n^{2 / 3}\right)$
- Supercritical $\epsilon^{3} n \rightarrow \infty: \quad L_{1}\left(G_{n, p}\right)=\Theta(n)$

The critical window is of order $O\left(n^{-4 / 3}\right)$

## Percolation in finite graphs

First reference was Erdős-Rényi graphs $p_{c}=\frac{1}{n}$

- Finite graphs: $p_{c}:=$ ???
- How big can components be?


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A definition that works
For finite transitive graphs with $V$ vertices and degree $m$, fix $\lambda \in(0,1)$.
Let $p_{c}=p_{c}(\lambda)$ solve

$$
E_{p_{c}(\lambda)}[|C(0)|]=\lambda V^{1 / 3}
$$

Fact:

$$
p_{c}\left(\lambda_{1}\right)-p_{c}\left(\lambda_{2}\right)=O\left(m^{-1} V^{-1 / 3}\right)
$$

Why is $p_{c}=1 / n$ ?

## Detour to Galton-Watson trees

- Indiv. $v$ has random $\xi_{v}$ children independently from rest.

$$
Z_{n}=\# \text { indiv. at generation } n
$$

## Galton-Watson Survival

Average children $\mathbb{E}\left[\xi_{v}\right]=(1+\epsilon)$ determines

- $\epsilon \leq 0$ : Extinction w.p. 1
- $\epsilon>0$ : Survival with positive prob.

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Markovian process: Each generation only depends on previous one.

$$
Z_{n+1}=\sum_{i=1}^{Z_{n}} \xi_{v_{i}}
$$

Why is $p_{c}=1 / n ?$

## Exploring components approximation:

- When graph has high dimension, exploration on giant component goes on forever.
- On $G_{n, p}$, each vertex sees $\operatorname{Bin}(n-1, p)$ other vertices

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Heuristic for percolation: should use $p_{c}(n-1) \sim 1$ instead

If $\epsilon+O\left(n^{-1 / 3}\right)$ then there is $\epsilon^{\prime}=O\left(n^{-1 / 3}\right)$ with

$$
p=\frac{1}{n}(1+\epsilon)=\frac{1}{n-1}\left(1+\epsilon^{\prime}\right)
$$

## The case of the Hypercube $Q^{n}$

## Borgs et al. ['05, '06]; Hosftad, Slade ['05, '06], Hofstad, Nachmias ['12,'14]

There exists rational coefficients $a_{k}$ such that

$$
p_{c}=\sum_{k=1}^{K} a_{k} n^{-k}+O\left(n^{-K-1}\right)
$$

In particular,

$$
\begin{aligned}
p_{c} & =\frac{1}{n}+\frac{1}{n^{2}}+\frac{7}{2 n^{3}}+O\left(n^{-4}\right) \\
& =\frac{1}{n-1}+\frac{5}{2}(n-1)^{-3}+O\left(n^{-4}\right)
\end{aligned}
$$

- Based on lace expansion and triangle condition verification
- Window too small $O\left(n^{-1} 2^{-1 / 3}\right)$ to neglect any expansion term


## Exploration on lattice-like graphs

Goal: Count size of a vertex $v$ component

## Exploration tracks:

- Explored vertices $D_{k}$
- To-explore vertices $X_{k}$

$$
D_{0}=\{v\}, \quad X_{0}=\emptyset
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At each step $k$ :
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(2) Add not explored neighbors
of $w$ to $X_{k}$
(3) Move $w \in D_{k}$

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Branching process - collisions

Component size

## The hypercube's local structure

$\{0,1\}^{n}$ Representation

- Sequences with $n$ entries
- Crossing edge changes one entry: ( $0,0,1,0, \ldots, 0$ )
- Smallest cycle has length 4



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## Simple collisions

- Parent:
$(1,0,0,0, \ldots, 0)$
- Possible children:

$$
\begin{aligned}
& (1,0,1,0, \ldots, 0) \\
& (1,1,0,0, \ldots, 0)
\end{aligned}
$$

- Possible grandkid:
$(1,1,1,0, \ldots, 0)$


Two steps in exploration

## Project: Heuristic to recover $c=\frac{5}{2}$

The local structure of hypercube predicts coefficients of critical $p_{c}$

- 'Guess'

$$
p_{c}=(n-1)^{-1}+c(n-1)^{-3}
$$

and tune $c$ via the survival threshold of a branching process

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- Design: A modified Poisson branching process with suitable survival threshold.

| Exploration | Hypercube dim. $n$ | Branching |
| :---: | :---: | :---: |
| Average children | $1+\frac{5}{2}(n-1)^{-2}$ | $1+\epsilon$ |
| Cousin identification | $(n-1)^{-2}$ | $q$ |

$$
\begin{aligned}
& \operatorname{Bin}\left(n-1, p_{c}\right) \approx \operatorname{Poi}(1+\epsilon) \\
& (n-1) p_{c} \approx 1+\epsilon=1+c q
\end{aligned}
$$

## Branching process with cousin mergers

- Indiv. have $\xi_{v} \sim \operatorname{Poi}(1+\epsilon)$ children
- Independently with probability $q$, each pair of cousins becomes a single indiv.
- Multiple mergers allowed



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## Difficulties:

Non-Markovian Process: $Z_{0}, Z_{1}, \ldots, Z_{n}$ not enough to obtain $Z_{n+1}$
Non-monotonicity: No straightforward coupling gives monotonicity of survival

## Survival Gap

BP with cousin mergers (E., Penington, Skerman, '20+)
If $\xi_{v} \sim \operatorname{Poi}(1+\epsilon), \epsilon>0$ suff. small, then merger prob. $q$ determines

- $q \geq 2 \epsilon+K \epsilon^{2}$ : Extinction w.p. 1
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Partial Idea: Estimate average growth per generation

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\mathbb{E}\left[Z_{n+1}\right] \approx\left(1+\epsilon-\frac{q}{2}\right) \mathbb{E}\left[Z_{n}\right]
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& \approx(1+\epsilon) \mathbb{E}\left[Z_{n}\right]-\frac{q}{2} \mathbb{E}\left[Z_{n-1}\right]
\end{aligned}
$$

## Zoom-in on idea

Partial Idea: Estimate average growth per generation

$$
\mathbb{E}\left[Z_{n+1}\right] \approx(1+\epsilon) \mathbb{E}\left[Z_{n}\right]-\frac{q}{2}(1+\epsilon)^{4} \mathbb{E}\left[Z_{n-1}\right]+O\left(\epsilon^{2}\right)
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Recall. If $X \sim \operatorname{Poi}(\lambda)$, then $\mathbb{E}[X(X-1)]=\lambda^{2}$

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$\mathbb{E}[\#$ pairs of cousins per grandparent]
$=\mathbb{E}\left[\#\right.$ pairs of children $\left.\left\{v_{1}, v_{2}\right\}\right]$

- $\mathbb{E}\left[\xi_{v_{1}}\right] \mathbb{E}\left[\xi_{v_{2}}\right]$


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- $\mathbb{E}\left[\xi_{v_{1}}\right] \mathbb{E}\left[\xi_{v_{2}}\right]$
$=\frac{(1+\epsilon)^{2}}{2} \cdot(1+\epsilon)(1+\epsilon)$


## What went wrong?

Offspring distribution: Not all vertices can explore $n-1$ new edges

$$
\mathbb{E}\left[Z_{n+1}\right] \approx(1+\epsilon) \mathbb{E}\left[Z_{n}\right]-\frac{q}{2} \mathbb{E}\left[Z_{n-1}\right]-q \mathbb{E}\left[Z_{n-2}\right]-q \mathbb{E}\left[Z_{n-3}\right]
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$\mathbb{E}[\#$ pairs of aunt-niece per grandparent]

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$$

$$
\begin{aligned}
& \mathbb{E}[\# \text { pairs of aunt-niece per grandparent }] \\
= & \mathbb{E}\left[\# \text { pairs of children }\left(v_{1}, v_{2}\right)\right] \\
& \cdot \mathbb{E}\left[\# \text { grandchildren of } v_{1}\right]
\end{aligned}
$$

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- $\mathbb{E}\left[\#\right.$ grandchildren of $\left.v_{1}\right]$

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and
$\mathbb{E}[\#$ greatgrandchildren per indiv. $]=(1+\epsilon)^{4}$

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$\mathbb{E}[\#$ pairs of aunt-niece per grandparent]

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=\mathbb{E}\left[\# \text { pairs of children }\left(v_{1}, v_{2}\right)\right]
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- $\mathbb{E}\left[\#\right.$ grandchildren of $\left.v_{1}\right]$
$=(1+\epsilon)^{2}(1+\epsilon)^{2}=(1+\epsilon)^{4}$
and
$\mathbb{E}[\#$ greatgrandchildren per indiv. $]=(1+\epsilon)^{4}$
All these pairs give collisions with probability

$$
(n-1)^{-2} \sim q
$$

## Refining the cousin mergers model



## Process construction

From generation $n$ to $n+1$ :
(1) Reproduction: Indiv. at generation $n$ have children.
(2) Deletions: Keep authentic children w.p. $(1-q)^{k_{v}}$
(3) Collisions: Each pair of cousins flip biased coin,
(9) Identification: of pairs of cousins.

## Survival gets the mysterious coefficients!

## Refined BP with collisions (E., Penington, Skerman, ' $20^{+}$)

If $\xi_{v} \sim \operatorname{Poi}(1+\epsilon), \epsilon>0$ suff. small, then collision prob. $q$ determines

- $q \geq \frac{2}{5} \epsilon+K \epsilon^{2}$ : Extinction w.p. 1
- $q \leq \frac{2}{5} \epsilon-K \epsilon^{2}$ : Survival w. positive prob.

Partial Idea: There are collisions occurring 4 times as often


$$
\mathbb{E}\left[Z_{n+1}\right] \approx\left(\begin{array}{llll}
1+\epsilon & -\frac{1}{2} q & -\frac{2}{2} q & -\frac{2}{2} q
\end{array}\right)(1+O(\epsilon)) \mathbb{E}\left[Z_{n}\right] .
$$

## Summary

- We obtain a survival threshold for a variant of a branching process that mimics hypercube's exploration near criticality

- This sheds light on structures determining its critical probability


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