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

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joint work in progress with S. Penington, F. Skerman

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Making sense of formulas









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Contents

Percolation

- Critical probability
- Exploration of components

The structure of the hypercube

- Critical probability expansions
- The quest for a heuristic

Branching processes with mergers

- Cousin mergers does not suffice
- A refined collision model

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Percolation

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



Critical Probability: Edge density where an giant component appears

• Infinite graphs: $p_c := \inf\{p : P(|C(0)| = \infty) > 0\}$

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

Critical Probability: Edge density where a giant component appears





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

Critical Probability: Edge density where a giant component appears





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

- Subcritical $\epsilon^3 n \to -\infty$: $L_1(G_{n,p}) = O(\log n)$
- Critical $\epsilon^3 n \to a \in \mathbb{R}$: $L_1(G_{n,p}) = \Theta(n^{2/3})$
- Supercritical $\epsilon^3 n \to \infty$: $L_1(G_{n,p}) = \Theta(n)$

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

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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:

$$p_c(\lambda_1) - p_c(\lambda_2) = O(m^{-1}V^{-1/3})$$

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Detour to Galton-Watson trees

• Indiv. v has random ξ_v children independently from rest.

 $Z_n = \#$ indiv. at generation n

Galton-Watson Survival

Average children $\mathbb{E}[\xi_{\nu}] = (1 + \epsilon)$ 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.

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

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Exploring components approximation:

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

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Heuristic for percolation: should use $p_c(n-1) \sim 1$ instead

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

$$p=rac{1}{n}(1+\epsilon)=rac{1}{n-1}(1+\epsilon')$$

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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

$$p_c = \sum_{k=1}^{K} a_k n^{-k} + O(n^{-K-1})$$

In particular,

$$p_{c} = \frac{1}{n} + \frac{1}{n^{2}} + \frac{7}{2n^{3}} + O(n^{-4})$$
$$= \frac{1}{n-1} + \frac{5}{2}(n-1)^{-3} + O(n^{-4})$$

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

Goal: Count size of a vertex v component

Exploration tracks:

- Explored vertices D_k
- To-explore vertices X_k

$$D_0 = \{v\}, \qquad X_0 = \emptyset$$



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Branching process - collisions \approx Component size

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The hypercube's local structure

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

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



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Simple collisions

- Parent: (1,0,0,0,...,0)
- Possible children:

$$(1, 0, 1, 0, \dots, 0)$$

 $(1, 1, 0, 0, \dots, 0)$

• Possible grandkid: (1, 1, 1, 0, ..., 0)





Two steps in exploration

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

The local structure of hypercube predicts coefficients of critical pc

• '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. <i>n</i>	Branching
Average children	$1+\frac{5}{2}(n-1)^{-2}$	$1+\epsilon$
Cousin identification	$(n-1)^{-2}$	q

$$\operatorname{Bin}(n-1,p_c) \approx \operatorname{Poi}(1+\epsilon)$$

 $(n-1)p_c \approx 1+\epsilon = 1+cq$

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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: $Z_0, Z_1, ..., Z_n$ 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 $\xi_{v} \sim \text{Poi}(1 + \epsilon)$, $\epsilon > 0$ suff. small, then merger prob. q determines

- $q \ge 2\epsilon + K\epsilon^2$: Extinction w.p. 1
- $q \leq 2\epsilon K\epsilon^2$: Survival w. positive prob.

But ideal ratio was $q = \frac{2}{5}\epsilon$

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Partial Idea: Estimate average growth per generation

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

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Partial Idea: Estimate average growth per generation

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

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Partial Idea: Estimate average growth per generation

$$\mathbb{E}[Z_{n+1}] \approx (1+\epsilon)\mathbb{E}[Z_n] - \frac{q}{2}(1+\epsilon)^4\mathbb{E}[Z_{n-1}] + O(\epsilon^2)$$

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Recall. If $X \sim \text{Poi}(\lambda)$, then $\mathbb{E}[X(X-1)] = \lambda^2$

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 $\mathbb{E}[\# \text{ pairs of cousins per grandparent}]$

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$$\begin{split} & \mathbb{E}[\# \text{ pairs of cousins per grandparent}] \\ & = \mathbb{E}[\# \text{ pairs of children } \{v_1, v_2\}] \\ & \cdot \mathbb{E}[\xi_{v_1}] \mathbb{E}[\xi_{v_2}] \end{split}$$

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Partial Idea: Estimate average growth per generation

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 $\mathbb{E}[\# \text{ pairs of cousins per grandparent}] = \mathbb{E}[\# \text{ pairs of children } \{v_1, v_2\}] \\ \cdot \mathbb{E}[\xi_{v_1}]\mathbb{E}[\xi_{v_2}] \\ = \frac{(1+\epsilon)^2}{2} \cdot (1+\epsilon)(1+\epsilon)$

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

$$\mathbb{E}[Z_{n+1}] \approx (1+\epsilon)\mathbb{E}[Z_n] - \frac{q}{2}\mathbb{E}[Z_{n-1}] - q\mathbb{E}[Z_{n-2}] - q\mathbb{E}[Z_{n-3}]$$

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



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 $\mathbb{E}[\# \text{ pairs of aunt-niece per grandparent}] = \mathbb{E}[\# \text{ pairs of children } (v_1, v_2)] \\ \cdot \mathbb{E}[\# \text{ grandchildren of } v_1]$

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

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

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:

- Reproduction: Indiv. at generation n have children.
- **2** Deletions: Keep *authentic* children w.p. $(1-q)^{k_v}$
- Ollisions: Each pair of cousins flip biased coin,
- Identification: of pairs of cousins.

Survival gets the mysterious coefficients!

Refined BP with collisions (E., Penington, Skerman, '20⁺) If $\xi_{\nu} \sim \text{Poi}(1 + \epsilon)$, $\epsilon > 0$ 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



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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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