

Thresholds for Some Basic Properties

Eight Lectures on Random Graphs: Lecture 2

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Graphs and Properties

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- *Monotone* = adding edges cannot violate it.

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- $\mathcal{G}_{n,p}$ = random order- n graph with edge probability p .
- *Whp* = with high probability (approaching 1 as $n \rightarrow \infty$).
- *Markov's Inequality*:
for a random variable $X \geq 0$ and a real $a > 0$

$$\Pr [X \geq a] \leq \frac{\mathbb{E} [X]}{a}.$$

Monotone Properties

Theorem For any monotone \mathcal{A} and $p_1 \leq p_2$

$$\Pr [\mathcal{G}_{n,p_1} \in \mathcal{A}] \leq \Pr [\mathcal{G}_{n,p_2} \in \mathcal{A}].$$

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$$\Pr [\mathcal{G}_{n,p_1} \in \mathcal{A}] \leq \Pr [\mathcal{G}_{n,p_2} \in \mathcal{A}].$$

Proof Define $p_0 \in [0, 1]$ by

$$p_1 + (1 - p_1) p_0 = p_2.$$

Let $G_1 \in \mathcal{G}_{n,p_1}$ and $G_0 \in \mathcal{G}_{n,p_0}$. Then $G_1 \cup G_0 \sim \mathcal{G}_{n,p_2}$.

$$\Pr [G_1 \in \mathcal{A}] \leq \Pr [G_1 \cup G_0 \in \mathcal{A}]. \blacksquare$$

Thresholds

$p_0 = p_0(n)$ is a *threshold* for a monotone property \mathcal{A} if $\forall p(n)$

$$\Pr [\mathcal{G}_{n,p} \in \mathcal{A}] \rightarrow \begin{cases} 0, & \text{if } p/p_0 \rightarrow 0, \\ 1, & \text{if } p/p_0 \rightarrow \infty. \end{cases}$$

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Example $p_0 = \frac{1}{n}$ is a threshold for having a cycle. Indeed,

• if $p = o(1/n)$, then

$$\Pr [\exists \text{ cycle}] \leq \mathbb{E} [\# \text{cycles}] = \sum_{i \geq 3}^n \binom{n}{i} \frac{(i-1)!}{2} p^i \leq \sum_{i \geq 3} (np)^i \rightarrow 0.$$

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• if $p > \frac{2+\varepsilon}{n}$, then $\mathbb{E} [e(G)] = p \binom{n}{2} > (2 + \varepsilon) \frac{n-1}{2}$.

By Chernoff's bound, whp $e(G) \geq n$.

Monotone \Rightarrow \exists Threshold

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Monotone $\Rightarrow \exists$ Threshold

Theorem (Bollobás-Thomason'87) Every non-trivial monotone property \mathcal{A} has a threshold.

Proof Choose $p_0 = p(1/2)$, i.e.

$$\Pr [\mathcal{G}_{n,p_0} \in \mathcal{A}] = 1/2.$$

p_0 exists as $f(p) = \Pr [\mathcal{G}_{n,p} \in \mathcal{A}]$ is a polynomial with $f(0) = 0$ and $f(1) = 1$.

$p_0 = p(1/2)$ is a threshold

Given $\varepsilon > 0$, let $(1 - \varepsilon)^m < 1/2$. Let $p < p_0/m$. Let $G_1, \dots, G_m \in \mathcal{G}_{n,p}$ and

$$H = G_1 \cup \dots \cup G_m \sim \mathcal{G}_{n,1-(1-p)^m}.$$

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As $1 - (1 - p)^m \leq pm \leq p_0$,

$$\frac{1}{2} \leq \Pr [H \notin \mathcal{A}] \leq \Pr [\forall i G_i \notin \mathcal{A}] = (1 - \Pr [\mathcal{G}_{n,p} \in \mathcal{A}])^m.$$

$$\Rightarrow \Pr [\mathcal{G}_{n,p} \in \mathcal{A}] < \varepsilon.$$

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Other direction: take $\mathcal{G}_{n,p_0} \cup \dots \cup \mathcal{G}_{n,p_0}$. ■

Connectivity Property \mathcal{C}

Idea 1: Connectivity = \exists spanning tree

$$E[\# \text{ spanning trees}] = n^{n-2} \cdot p^{n-1}.$$

The “window” is

$$p = (1 + o(1)) \frac{1}{n}.$$

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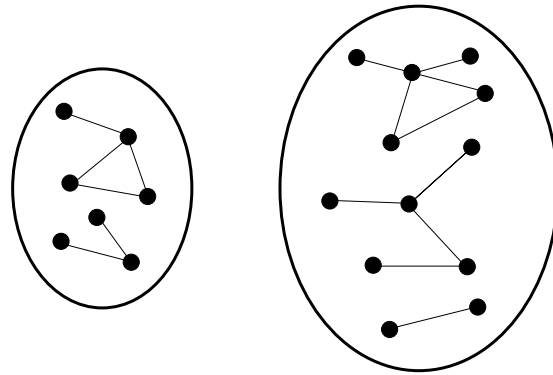
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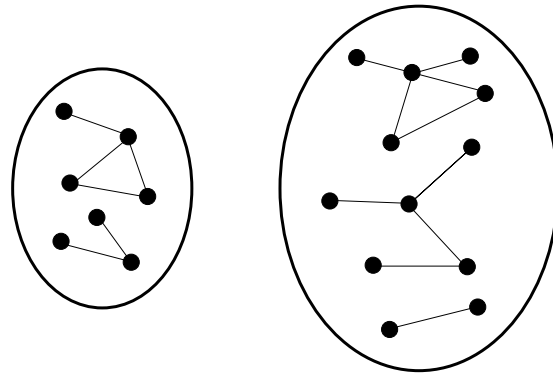
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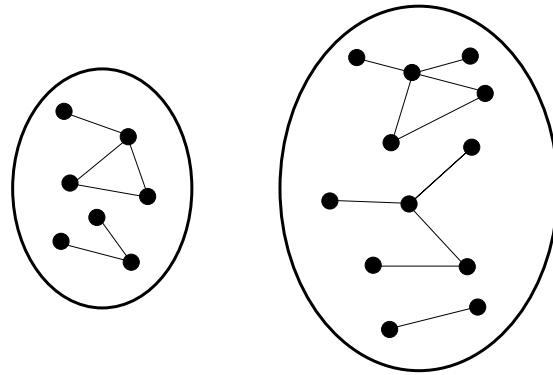
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Cuts or Isolated Components?

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Cuts or Isolated Components?

$C_k = \# k$ -components

Observation: $G \in \mathcal{C}$ iff $C_k = 0 \forall k \in [1, n/2]$.

Connectivity Threshold

Theorem (Erdős-Renyi'60) Let

$$p = \frac{\log n}{n} + \frac{c}{n}.$$

$$\text{Then } \Pr [\mathcal{G}_{n,p} \in \mathcal{C}] \rightarrow \begin{cases} e^{-e^{-c}}, & |c| = O(1), \\ 0, & c \rightarrow -\infty, \\ 1, & c \rightarrow +\infty. \end{cases}$$

In particular, $p_0(n) = \frac{\log n}{n}$ is a threshold for connectivity.

$$p = \frac{\log n}{n} + \frac{O(1)}{n}$$

Let $\Sigma := \sum_{k=2}^{\lfloor n/2 \rfloor}$.

$$\Pr \left[\sum C_k \geq 1 \right] \leq \mathbb{E} \left[\sum C_k \right]$$

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Let $\sum := \sum_{k=2}^{\lfloor n/2 \rfloor}$.

$$\begin{aligned} \Pr \left[\sum C_k \geq 1 \right] &\leq \mathbb{E} \left[\sum C_k \right] \\ &= \sum \mathbb{E} [C_k] \leq \sum \binom{n}{k} (1-p)^{k(n-k)} k^{k-2} p^{k-1} \end{aligned}$$

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$$\left[\binom{n}{k} \leq (en/k)^k \ \& \ (1-x) \leq e^{-x} \right]$$

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$$\left[\binom{n}{k} \leq (en/k)^k \ \& \ (1-x) \leq e^{-x} \right]$$

$$\leq n \sum \left(O(\log n) e^{-np+kp} \right)^k \rightarrow 0.$$

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$$\leq n \sum \left(O(\log n) e^{-np+kp} \right)^k \rightarrow 0.$$

Thus whp $C_2 = \dots = C_{\lfloor n/2 \rfloor} = 0$.

$$p = \frac{\log n}{n} + \frac{O(1)}{n} \quad \textbf{(cont.)}$$

Thus, whp $\mathcal{G}_{n,p} \in \mathcal{C}$ iff $C_1 = 0$ (i.e. no isolated vertices).

$$p = \frac{\log n}{n} + \frac{O(1)}{n} \quad \textbf{(cont.)}$$

Thus, whp $\mathcal{G}_{n,p} \in \mathcal{C}$ iff $C_1 = 0$ (i.e. no isolated vertices).

It is enough to prove $\Pr [C_1 = 0] \rightarrow e^{-e^{-c}}$ because

$$\begin{aligned} 0 &\leq \Pr [C_1 = 0] - \Pr [C \in \mathcal{C}] \\ &\leq \Pr [\exists i \in [2, n/2] C_i > 0] \rightarrow 0. \end{aligned}$$

Poisson Distribution with Mean μ

n independent trials, $\Pr[\text{success}] = \frac{\mu}{n}$, constant μ .

Poisson(μ) = # successes as $n \rightarrow \infty$.

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$$\Pr[i \text{ successes}] = \binom{n}{i} p^i (1-p)^{n-i} \rightarrow \frac{\mu^i e^{-\mu}}{i!}.$$

Isolated Vertices

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For fixed k

$$M_k[C_1] = (n)_k (1-p)^{k(n-1) - \binom{k}{2}} \rightarrow (e^{-c})^k = M_k[\text{Poisson}(e^{-c})].$$

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This is known to imply that $C_1 \rightarrow \text{Poisson}(e^{-c})$. In particular,

$$\Pr[C_1 = 0] \rightarrow e^{-e^{-c}}. \blacksquare$$

Sharp Threshold

Connectivity Threshold

p_0 is a *sharp threshold* for a monotone \mathcal{A} if $\forall \varepsilon > 0$ whp

$$\mathcal{G}_{n,(1-\varepsilon)p_0} \notin \mathcal{A} \text{ and } \mathcal{G}_{n,(1+\varepsilon)p_0} \in \mathcal{A}.$$

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Examples,

- connectivity: sharp,
- having a triangle: not sharp,
- having a cycle: ‘one-sided sharp’.

Friedgut's Theorem

Note: Sharp threshold $\Rightarrow \frac{\partial}{\partial p} \Pr [\mathcal{G}_{n,p} \in \mathcal{A}] \neq O\left(\frac{1}{p}\right)$.

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Theorem (Friedgut'99) If for a monotone \mathcal{A}

$$\frac{\partial}{\partial p} \Pr [\mathcal{G}_{n,p} \in \mathcal{A}] = O\left(\frac{1}{p}\right),$$

then $\forall \varepsilon > 0$ there is a finite family \mathcal{F} of graphs such that
 $\forall n, p$

$$\Pr [\mathcal{G}_{n,p} \in \mathcal{A} \triangle \{\text{an } \mathcal{F}\text{-subgraph}\}] \leq \varepsilon. \blacksquare$$

Applying Friedgut's Theorem

Difficult to apply: the type of $\mathcal{A} \cup \mathcal{B}$ depends on which one appears 'earlier'.

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Theorem (Achlioptas-Friedgut'99) For fixed $k \geq 3$ k -colorability has a sharp threshold.

Model $\mathcal{G}_{n,M}$

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Theorem For a monotone \mathcal{A} ,

$$\Pr [\mathcal{G}_{n,M} \in \mathcal{A}]$$

is a non-decreasing function of M . ■

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Theorem For any non-trivial monotone \mathcal{A} , there is a *threshold* M_0 , that is,

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Theorem Let $M = \frac{\log n + c}{n} \binom{n}{2}$. Then

$$\Pr [\mathcal{G}_{n,M} \in \mathcal{C}] \rightarrow \begin{cases} e^{-e^{-c}}, & |c| = O(1), \\ 0, & c \rightarrow -\infty, \\ 1, & c \rightarrow +\infty. \end{cases}$$

In particular, $M_0 = n \log n$ is a threshold for connectivity.

Connectivity of $\mathcal{G}_{n,M}$

Proof Enough to consider $|c| = O(1)$. Take small $\varepsilon > 0$. Let

$$p = \frac{\log n + c - \varepsilon}{n}.$$

Take $G \in \mathcal{G}_{n,p}$. Let $l = M - e(G)$. If $l \geq 0$, let

$$H = G + l \text{ random edges}; \quad H \sim \mathcal{G}_{n,M}.$$

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By Chernoff's bound, $\Pr [e(G) > M] \rightarrow 0$. Hence,

$$\Pr [\mathcal{G}_{n,M} \in \mathcal{C}] \geq \Pr [G \in \mathcal{C}] - \Pr [e(G) > M] \geq e^{-e^{-c+\varepsilon}} - o(1).$$

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Upper bound: remove edges from $\mathcal{G}_{n,p}$, $p = \frac{\log n + c + \varepsilon}{n}$. ■

Hitting Time Version

- *Random graph process:*

$G_0 = n$ isolated vertices;

$G_{M+1} = G_M +$ a random edge.

- *Hitting time* $\tau[\mathcal{A}] = \min\{M : G_M \in \mathcal{A}\}$.

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Theorem (Erdős-Renyi'60) Whp $\tau[\delta \geq 1] = \tau[\mathcal{C}]$.

Proof Let $\mathcal{B} = \{H : \delta \geq 1 \text{ \& } H \notin \mathcal{C}\}$.

Idea 1:

$$\Pr[\exists M : G_M \in \mathcal{B}] \leq \sum_M \Pr[G_{n,M} \in \mathcal{B}] \not\rightarrow 0.$$

Idea 2: Using $G_M \subset G_{M+1}$

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Fix large $c > 0$. Let $m_{\pm} = \lfloor \frac{\log n \pm c}{n} \rfloor$.

$$\begin{aligned} \Pr [\exists M : G_M \in \mathcal{B}] &\leq \Pr [\exists M \leq m_- \delta(G_M) \geq 1] \\ &+ \Pr [\exists M \in (m_-, m_+) G_M \in \mathcal{B}] \\ &+ \Pr [\exists M \geq m_+ G_M \notin \mathcal{C}] \\ &= p_- + p_0 + p_+ \end{aligned}$$

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Now,

$$\begin{aligned} p_+ &= \Pr [\mathcal{G}_{n, m_+} \notin \mathcal{C}] = 1 - e^{-e^{-c}} + o(1), \\ p_- &= \Pr [\delta(\mathcal{G}_{n, m_-}) \geq 1] = e^{-e^c} + o(1). \end{aligned}$$

The Old Trick

- **Recall:** $\mathcal{B} = \{H : \delta \geq 1 \ \& \ H \notin \mathcal{C}\}$.
- **Aim:** $\Pr [\exists M \in (m_-, m_+) G_M \in \mathcal{B}] \rightarrow 0$.

Let

$$p = \frac{\log n - c - \varepsilon}{n},$$
$$G \in \mathcal{G}_{n,p}.$$

Counting Components (Again)

Lemma Whp $C_1 \leq \log n$ and $C_2 = \dots = C_{\lfloor n/2 \rfloor} = 0$, i.e. G consists of at most $\log n$ isolated vertices and one component.

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Lemma Whp $C_1 \leq \log n$ and $C_2 = \dots = C_{\lfloor n/2 \rfloor} = 0$, i.e. G consists of at most $\log n$ isolated vertices and one component.

Proof

$$\mathbb{E}[C_1] = n(1-p)^{n-1} \leq e^{c+\varepsilon} + o(1) = O(1).$$

So $\Pr[C_1 > \log n] < \frac{\mathbb{E}[C_1]}{\log n} \rightarrow 0$.

We already proved that whp $C_2 = \dots = C_{\lfloor n/2 \rfloor} = 0$. ■

Process between m_- and m_+

$$\begin{aligned} \Pr [\exists M \in (m_-, m_+) : G_M \in \mathcal{B}] &\leq \Pr [e(G) > m_-] \\ &+ \Pr [C_1 > \log n] \\ &+ \Pr [\exists k \in [2, n/2] C_k = 0] \\ &+ m_+ \frac{\binom{\log n}{2}}{\binom{n}{2} - o(n^2)} \rightarrow 0. \end{aligned}$$

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Putting all together: whp $\tau[\mathcal{C}] = \tau[\delta \geq 1]$. ■

Some Spanning Subgraphs

$$\Pr [\mathcal{G}_{n,p} \in \mathcal{A}] \rightarrow \begin{cases} 0, & c \rightarrow -\infty, \\ 1, & c \rightarrow +\infty, \end{cases}$$

- **Erdős & Renyi'66:**

$$\mathcal{A} = \{\text{perfect matching}\}, n \text{ even}, p = \frac{\log n + c}{n}.$$

- **Korshunov'83, Komlós & Szemerédi'83:**

$$\mathcal{A} = \{\text{Hamiltonian}\}, p = \frac{\log n + \log \log n + c}{n}.$$

- **Riordan'00:**

$$\mathcal{A} = \{d\text{-dimensional cube}\}, n = 2^d, p = \frac{1}{4} + c \frac{\log d}{d}.$$

Perfect Matchings

$\mathcal{G}_{n,n,p}$: random subgraph of $K_{n,n}$, $\Pr[\text{edge}] = p$
(or random $n \times n$ 0/1-matrix).

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(or random $n \times n$ 0/1-matrix).

Theorem (Erdős & Renyi'64) Let $p = \frac{\log n + c}{n}$ and $G \in \mathcal{G}_{n,n,p}$.
Then

$$\Pr[G \text{ has a matching}] \rightarrow e^{-2e^{-c}}.$$

In particular, $p_0 = \frac{\log n}{n}$ is a sharp threshold.

Using Hall's Theorem

Proof No matching $\Leftrightarrow \exists S$ s.t.

- $|S| = |\Gamma(S)| + 1,$
- $|S| \leq \lceil n/2 \rceil,$
- $\forall x \in \Gamma(S) \quad |\Gamma(x) \cap S| \geq 2.$

Using Hall's Theorem

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- $|S| = |\Gamma(S)| + 1,$
- $|S| \leq \lceil n/2 \rceil,$
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$$\begin{aligned} \Pr [\exists \text{ such } S : |S| \geq 2] &\leq \mathbb{E} [\# \text{ such } S] \\ &\leq 2 \sum_{s=2}^{\lceil n/2 \rceil} \binom{n}{s} \binom{n}{s-1} \binom{s}{2}^{s-1} p^{2s-2} (1-p)^{s(n-s+1)} = o(ne^{-pn}). \end{aligned}$$

Using Hall's Theorem

Proof No matching $\Leftrightarrow \exists S$ s.t.

- $|S| = |\Gamma(S)| + 1,$
- $|S| \leq \lceil n/2 \rceil,$
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$\mathbb{E} [C_1] = 2n(1-p)^n \rightarrow 2e^{-c}$. As before $C_1 \rightarrow \text{Poisson}(2e^{-c})$. ■

Model $\mathcal{G}_{n,n,M}$

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Theorem Let $M = \frac{\log n + c}{n} n^2$ and $G \in \mathcal{G}_{n,n,p}$. Then

$$\Pr [G \text{ has a matching}] \rightarrow e^{-2e^{-c}}.$$

In particular, $M_0 = n \log n$ is a sharp threshold. ■

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$$\begin{aligned} \Pr[\exists |S| : |S| = 2] &\leq \Pr[G_{m_-} \text{ has such 2-element } S] \\ &+ \Pr[C_1(G_{m_-}) > \log n] \\ &+ \log n \Pr[\text{such 2-element } S \text{ is created}] \\ &\rightarrow 0. \blacksquare \end{aligned}$$

General Spanning Subgraphs

Theorem (Alon-Füredi'92) Let $v(H) = n$, $\Delta(H) \leq d$
 $D = d^2 + 1$ and $G \in \mathcal{G}_{n,p}$.

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Proof Let $F = H^2$; $\Delta(F) < D$.

Lemma (Hajnal-Szemerédi'70):

\exists F -stable sets $V_1 \cup \dots \cup V_D = V(F)$, each $|V_i| = \frac{n}{D} \pm 1$.

Take $V(G) = U_1 \cup \dots \cup U_D$ with $|U_i| = |V_i|$.

Partial H -Embeddings

Build $f_i : V_1 \cup \dots \cup V_i \rightarrow U_1 \cup \dots \cup U_i$ inductively.

Let $m = |V_{i+1}| = |U_{i+1}|$ and

$$F = \{ (u, v) : u \in U_{i+1}, v \in V_{i+1} \Gamma_H(v) \subset f_i(\Gamma_G(u)) \}.$$

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Observe $F \sim \mathcal{G}_{m,m,\geq p^d}$.

$$\begin{aligned} \Pr [\text{FAIL}] &= \Pr [\text{no matching}] \\ &= O(m e^{-pm}) = o(1/D). \end{aligned}$$

So, whp f_i exists $\forall i \in [D]$, i.e. $H \subset G$. ■

Random Edge-Weights

- **The model:** Random independent weights w_e , $e \in \binom{[n]}{2}$, each uniformly distributed in $(0, 1)$.
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Theorem (Frieze'85)

1. $E[w(T)] \rightarrow \zeta(3)$.
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Proof of 1. Let $G_p = ([n], \{e : w_e \leq p\}) \sim \mathcal{G}_{n,p}$.

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where $\kappa(G) = \#$ components of G .

Computing $E [w(T)]$

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$$= \sum_{k \geq 1} \frac{n^k k^{k-2}}{k!} \times \frac{(k-1)! (k(n-k))!}{(k(n-k+1))!} \rightarrow \sum_{k \geq 1} \frac{1}{k^3}.$$

Above the Connectivity Threshold

Lemma Let $p = \frac{3 \log n}{n}$. Then

$$\Pr [\mathcal{G}_{n,p} \notin \mathcal{C}] = o(1/n).$$

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Proof As before, we argue that

$$\sum_{k=1}^{\lfloor n/2 \rfloor} \mathbb{E} [C_k] = O(n e^{-pn}) = o(1/n). \blacksquare$$

Hence, $\mathbb{E} [w(T \setminus G_{3 \log n/n})] \leq n \times o(1/n) \rightarrow 0$.

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