

Probabilistic Methods in Combinatorics Lecture 13

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Erdős-Rényi model

G(n,p): Erdős-Rényi random graphs

- n nodes



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- For each pair of vertices, create an edge independently with probability p.



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An example $G(3, \frac{1}{2})$:





The birth of random graph theory





Paul Erdős and A. Rényi, On the evolution of random graphs *Magyar Tud. Akad. Mat. Kut. Int. Kozl.* **5** (1960) 17-61.



The birth of random graph theory



ON THE EVOLUTION OF RANDOM GRAPHS

by

P. ERDÖS and A. RÉNYI

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1. Definition of a random graph

Let $E_{n,N}$ denote the set of all graphs having *n* given labelled vertices V_1, V_2, \cdots , V_n and *N* edges. The graphs considered are supposed to be not oriented, without parallel edges and without slings (such graphs are sometimes called linear graphs). Thus a graph belonging to the set $E_{n,N}$ is obtained by choosing *N* out of the possible $\binom{n}{2}$ edges between the points V_1, V_2, \cdots, V_n , and therefore the number of elements of $E_{n,N}$ is equal to $\binom{\binom{n}{2}}{N}$. A random graph $\Gamma_{n,N}$ can be defined as an element of $E_{n,N}$ chosen at random, so that each of the elements of $E_{n,N}$ have the same probability to be chosen, namely $1/\binom{\binom{n}{2}}{N}$. There is however an other slightly



 $\frac{\frac{c}{n}}{\frac{1}{n}}, \frac{\frac{1}{n}}{\frac{c'}{\frac{n}{n}}}, \frac{\frac{\log n}{n}}{\Omega\left(\frac{\log n}{n}\right)},$

()

the empty graph. disjoint union of trees. cycles of any size. the double jumps. one giant component, others are trees. G(n, p) is connected.

 $\begin{array}{ll} & \text{connected and almost regular.} \\ \Omega(n^{\epsilon-1}) & \text{finite diameter.} \\ \Theta(1) & \text{dense graphs, diameter is 2.} \\ 1 & \text{the complete graph.} \end{array}$



Range I p = o(1/n)

The random graph $G_{n,p}$ is the disjoint union of trees. In fact, trees on k vertices, for k = 3, 4, ... only appear when p is of the order $n^{-k/(k-1)}$.



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The random graph $G_{n,p}$ is the disjoint union of trees. In fact, trees on k vertices, for $k = 3, 4, \ldots$ only appear when p is of the order $n^{-k/(k-1)}$.

Furthermore, for $p = cn^{-k/(k-1)}$ and c > 0, let $\tau_k(G)$ denote the number of connected components of G formed by trees on k vertices and $\lambda = c^{k-1}k^{k-2}/k!$. Then,

$$\Pr(\tau_k(G_{n,p}) = j) \to \frac{\lambda^j e^{-\lambda}}{j!}$$

for $j = 0, 1, \ldots$ as $n \to \infty$.





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- In this range of p, $G_{n,p}$ contains cycles of any given size with probability tending to a positive limit.
- All connected components of $G_{n,p}$ are either trees or unicyclic components. Almost all (i.e., n o(n)) vertices are in components which are trees.
- The largest connected component of $G_{n,p}$ is a tree and has about $\frac{1}{\alpha}(\log n - \frac{5}{2}\log\log n)$ vertices, where $\alpha = c - 1 - \log c$.



Range III $p \sim 1/n + \mu/n$, the double jump

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- If $\mu = 0$, the largest component has size of order $n^{2/3}$.
- If $\mu > 0$, there is a unique giant component of size αn where $\mu = -\alpha^{-1} \log(1 \alpha) 1$.
- Bollobás showed that a component of size at least $n^{2/3}$ in $G_{n,p}$ is almost always unique if p exceeds $1/n + 4(\log n)^{1/2}n^{-4/3}$.





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- Except for one "giant" component, all the other components are relatively small, and most of them are trees.
- The total number of vertices in components which are trees is approximately n f(c)n + o(n).
- The largest connected component of $G_{n,p}$ has approximately f(c)n vertices, where

$$f(c) = 1 - \frac{1}{c} \sum_{k=1}^{\infty} \frac{k^{k-1}}{k!} (ce^{-c})^k.$$





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 If

$$p = \frac{\log n}{kn} + \frac{(k-1)\log\log n}{kn} + \frac{y}{n} + o(\frac{1}{n}),$$

then there are only trees of size at most k except for the giant component. The distribution of the number of trees of k vertices again has a Poisson distribution with mean value $\frac{e^{-ky}}{k \cdot k!}$.







Galton-Watson branching process: Let Z be a distribution over the non-negative integers. Starting with a single node, it gives Z children nodes. Each of children nodes have Z children independently. The process continues, each new offspring having an independent number A of children.



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- $Z_1, Z_2, \ldots, Z_t, \ldots$: a countable sequence of independent identically distributed variables, each have distribution Z.
- Y_t : the number of living children at time t.

$$Y_0 = 1$$

 $Y_t = Y_{t-1} + Z_t - 1.$



Let T be the total number of nodes in Galton-Watson process. There are two essentially different cases.

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- $Y_t > 0$ for all $t \ge 0$. In this case the Calton-Watson process goes on forever and $T = \infty$.
- $Y_t = 0$ for some $t \ge 0$. In this case, T is the least integer for which $Y_T = 0$. The Galton-Watson process stops with T nodes.



Poisson branching process

Let Z be the Poisson distribution with the expectation c. Write $T = T_c^{po}$. **Theorem:** If $c \le 1$, then T is finite with probability one. If c > 1, then T is infinite with probability y = y(c), where y is the unique positive real satisfying

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$$e^{-cy} = 1 - y.$$

Proof: Suppose c < 1. $\Pr(T > t) \le \Pr(\sum_{i=1}^{t} Z_i \ge t) < e^{-kt}$,

for some constant k. $\lim_{t\to\infty} \Pr(T > t) = 0.$









Suppose $c \ge 1$. Let $z = 1 - y = \Pr(T < \infty)$. Then

$$z = \sum_{i=0}^{\infty} \Pr(Z_1 = i) z^i = \sum_{i=0}^{\infty} e^{-c} \frac{c^i}{z^i} i! = e^{c(z-1)}.$$









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Hence $1 - y = e^{-cy}$. When c = 1, this equation has a unique solution y = 0. When c > 1, there are two solutions 1 and y(c). By Chernoff's equality, for any t

 $\Pr(\sum_{i=1}^{t} Z_i \le t) < e^{-\frac{(c-1)^2 t}{2c}}.$ There is a t_0 so that $\sum_{t \ge t_0} e^{-\frac{(c-1)^2 t}{2c}} < 1$. Thus, $y > \Pr(T = \infty \mid T \ge t_0) \Pr(T \ge t_0) > 0.$

Graph branching process



Graph branching process

Let C(v) denote the component of G(n, p), containing a vertex v. Explore C(v) using Breadth First Search (BFS). In this procedure all vertices will be live, dead, or neutral. The live vertices will be contained in a queue Q.



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Algorithm for computing C(v):

Push v into Q. Mark all vertices but v neutral. while(Q is not empty){ Pop Q and get w, mark w dead foreach(w' neutral){ if (ww' is an edge of G(n, p)){ mark w' live and push it into Q }



Analysis



In the graph branching process, let Y_t be the size of the queue at time t and N_t be the set of neutral vertices. Let N_t be the set of neutral vertices.

$$Z_t \sim B(N_{t-1}, p).$$

$$N_t \sim B(n-1, (1-p)^t).$$

If T = t it is necessary that $N_t = n - t$. We have

$$\Pr(|C(v)| = t) \le \Pr(B(n - 1, (1 - p)^t) = n - t).$$

Or, equivalently,



$$\Pr(|C(v)| = t) \le \Pr(B(n-1, 1 - (1-p)^t) = t - 1).$$

Comparison

Theorem: For any positive real c and any fixed integer k

$$\lim_{n \to \infty} \Pr(|C(v)| = k \text{ in } G(n, \frac{c}{n})) = \Pr(T_c^{po} = k).$$



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Proof: Let Γ be the set of k-tuples $\vec{z} = (z_1, z_2, \dots, z_k)$ of nonnegative integers such that the recursion $y_0 = 1$, $y_t = y_{t-1} + z_t - 1$ has $y_t > 0$ for t < k and $y_k = 0$.

$$\Pr(T^{gr} = k) = \sum \Pr(Z_i^{gr} = z_i, 1 \le i \le k)$$

$$\Pr(T^{po} = k) = \sum \Pr(Z_i^{po} = z_i, 1 \le i \le k).$$

Here both sums are over $\vec{z} \in \Gamma$.





Continue



Since $Z_{i-1} = n - O(1)$ and $B(Z_i, p)$ approaches the Poisson distribution, we have

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$$\Pr(T^{gr} = k) = \Pr(Z_i^{gr} = z_i, 1 \le i \le k)$$

$$= \prod_{i=1}^k \Pr(B(N_{i-1}^{gr}, p) = z_i)$$

$$\to \prod_{i=1}^k \Pr(B(Z_i^{po} = z_i))$$

$$= \Pr(T^{po} = k).$$



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Proof: We have $\Pr(T_c^{po} = k) = \lim_{n \to \infty} \Pr(|C(v)| = k)$ in G(n, p) with p = c/n.

$$\Pr(C(v) = k) \approx {\binom{n}{k-1}} k^{k-2} p^{k-1} (1-p)^{k(n-k)}$$
$$\rightarrow \frac{e^{-ck} (ck)^{k-1}}{k!}.$$





 $p = \frac{c}{n}$, $0 \le c \le 1$

$$\Pr(|C(v)| \ge u) \le (1 + o(1)) \Pr(T_c^{po} \ge u) \approx \sum_{k=u}^{\infty} e^{-ck} \frac{(ck)^{k-1}}{k!}.$$





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Setting $u = (c - 1 - \ln c)^{-1} \ln n + C \ln \ln n$, we have $\Pr(|C(v)| \ge u) \le o(\frac{1}{n \ln n})$.





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Thus, the size of largest component in G(n,p) is at most $(c-1-\ln c)^{-1}\ln n + O(\ln\ln n)$.

Most of them are trees. Then number of trees of size k is

$$(1+o(1))e^{-ck}\frac{(ck)^{k-1}}{k!}n.$$



Barely subcritical regimes



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The size of the largest component approaches $Kn^{2/3}\lambda^{-2}\ln n$.

