# PACS Part 2, Lecture 4

## Connectivity Threshold in ER Graphs

- $p = \frac{\log n}{n}$  is a threshold function for connectivity in Erdős–Rényi graphs
- If  $p \leq \lambda \frac{\log n}{n}$  with  $\lambda < 1$ , then there is an isolated vertex a.a.s.
- If  $p \ge \lambda \frac{\log n}{n}$  with  $\lambda > 1$ , then the graph is connected a.a.s.

#### Size of Connected Components

- we will study the maximum size of a connected component in the disconnected case
- the connected component C(u) of vertex u can be constructed in the following manner
  - We keep a set L of live vertices, a set N of neutral vertices, and a set D of dead vertices.
  - Initially, at time t = 0, we have  $L(0) = \{u\}$ ,  $N(0) = V \setminus \{u\}$ , and D(0) = 0.
  - In every step  $t \ge 1$ , we choose a live vertex  $w \in L(t-1)$ , move it from L to D, and move all neutral neighbors of w from N to L.
  - The process stops at the earliest time T with  $L(T) = \emptyset$ .
  - We then have C(u) = D(T) and |D(T)| = T.
- Setting Z(t) = |N(t-1)| |N(t)|, we have the recurrence formulas

$$|L(t)| = |L(t-1)| - 1 + Z(t)$$
  
|N(t)| = |N(t-1)| - Z(t)  
|D(t)| = t

with |L(0)| = 1, |N(0)| = n - 1, and |D(0)| = 0.

• In particular, |N(t)| = n - t - |L(t)| and

$$|L(t)| = 1 + \sum_{s=1}^{t} (Z(s) - 1) = 1 - t + \sum_{s=1}^{t} Z(s)$$

• Thus:  $Z(t) \sim Bin(|N(t-1)|, p) = Bin(n-t+1-|L(t-1)|, p)$ 

## **Regime with Only Small Components**

- set p = c/n
- since we always have  $|N(t-1)| \leq n$ , we can upper bound the size T of the connected component by the length of a process Y(t) that satisfies the recurrence for |L(t)| and in which  $Z(t) \sim \operatorname{Bin}(n, p)$

- that is, Y(t) = Y(t-1) 1 + Z(t) and Y(0) = 1, with the process stopping when Y(T) = 0
- denote by  $T_{n,p}$  the length of the original graph process and by  $\tilde{T}_{n,p}$  the length of the process Y(t)
- if c < 1, applying Chernoff's bound, we have:

$$\begin{split} \mathbb{P}(T_{n,p} > t) &\leq \mathbb{P}(T_{n,p} > t) \\ &\leq \mathbb{P}(\operatorname{Bin}(nt,p) \geq t) \\ &= \mathbb{P}(\operatorname{Bin}(nt,p) \geq ct(1 + (1-c)/c)) \\ &\leq \exp\left(-\frac{ct}{3}\frac{(1-c)^2}{c^2}\right) \end{split}$$

- choosing  $t = a \log n$  with an appropriate constant a, this is  $\leq 1/n^2$
- the union bound then implies that all connected components have size  $\leq a \log n$  with high probability

## Birth of the Giant Component

- let now p = c/n with c > 1
- setting  $t^- = b \log n$  and  $t^+ = n^{2/3}$ , we define for a vertex v:
  - -v is small if  $|C(v)| \leq t^{-}$
  - -v is big if  $|L_v(t)| \ge \frac{c-1}{2}t$  for all  $t^- \le t \le t^+$  v is bad if it is neither big nor small
- If there are no bad vertices, then there is at most one big component (of super-logarithmic size).

Proof: For any pair (u, v) of big vertices, we have:

$$\mathbb{P}(C(u) \neq C(v)) \leq \mathbb{P}(\text{there are no edges between } L_u(t^+) \text{ and } L_v(t^+))$$
$$\leq (1-p)^{\left(\frac{c-1}{2}t^+\right)^2} \leq \exp\left(-\frac{c}{n}\frac{(c-1)^2}{4}n^{4/3}\right)$$
$$= \exp\left(-\frac{c(c-1)^2}{4}n^{1/3}\right) = O(1/n^3)$$

The union bound then shows that there is no such pair with high probability.

• If there are no bad vertices, then there is a giant component (of linear size).

Proof: Let X be the number of small vertices. We show that We have:

$$\mathbb{P}\left(\tilde{T}_{n,p} \le t^{-}\right) \le \mathbb{P}\left(T_{n,p} \le t^{-}\right) \le \mathbb{P}\left(\tilde{T}_{n-t^{-},p} \le t^{-}\right)$$

We will later show that, for  $n \ge \infty$ , the two outer terms converge to the same quantity  $p_e$ , which then shows  $\mathbb{E}X = (p_e + o(1))n$ . But first we will study the variance of  $N_s$  and apply Chebyshev's inequality.

Define the indicator variable  $X_u = 1$  iff u is a small vertex. Then  $X = \sum_{u \in V} X_u$ . We have

$$\operatorname{Var}(X) \le \mathbb{E}X^2 = \mathbb{E}X + \sum_{v} \mathbb{P}(X_v = 1) \sum_{u \ne v} \mathbb{P}(X_u = 1 \mid X_v = 1)$$

and

$$\sum_{u \neq v} \mathbb{P}(X_u = 1 \mid X_v = 1) = \sum_{\substack{u \neq v \\ u \in C(v)}} \mathbb{P}(X_u = 1 \mid X_v = 1) + \sum_{\substack{u \neq v \\ u \notin C(v)}} \mathbb{P}(X_u = 1 \mid X_v = 1)$$
$$\leq t^- + (p_e + o(1))n$$

which gives

$$\operatorname{Var}(X) \le \mathbb{E}X + (p_e + o(1))^2 n^2 = \mathbb{E}X + o\left((\mathbb{E}X)^2\right)$$

Applying Chebyshev's inequality with  $a = \delta \mathbb{E}X$  gives

$$\mathbb{P}\left(X/n \ge p_e\left(1+\delta\right)\right) \le \frac{1}{\delta}\left(\frac{1}{\mathbb{E}X} + o(1)\right) = o(1)$$

We can even let  $\delta \to 0$  very slowly.

- We left two tasks open: show that there are no bad vertices and the convergence to  $p_e$
- We first show that there are no bad vertices with high probability:

Let v be a bad vertex. Then there is some  $t^- < t \leq t^+$  with  $L_v(t) < \frac{c-1}{2}t.$  We have

$$\mathbb{P}\left(L_{v}(t) \leq \frac{c-1}{2}t\right) \leq \mathbb{P}\left(\operatorname{Bin}\left(t\left(n-t-\frac{c-1}{2}t\right), \frac{c}{n}\right) \leq \frac{c-1}{2}t\right) \\
\leq \mathbb{P}\left(\operatorname{Bin}\left(t\left(n-\frac{c+1}{2}t^{+}\right), \frac{c}{n}\right) \leq \frac{c-1}{2}t\right)$$

We have  $\mu = ct \left(1 - \frac{c+1}{2n}t^+\right)$ . Choosing  $\delta$  such that  $(1 - \delta)\mu = \frac{c+1}{2}t$ , we get  $\delta = \frac{c-1}{2c} + o(1)$  as  $n \to \infty$  and, by Chernoff's bound,

$$\mathbb{P}\left(L_u(t) \le \frac{c-1}{2}t\right) \le \exp\left(-\left(\frac{(c-1)^2}{8c} + O(n^{-1/3})\right)t\right)$$

and thus

$$\mathbb{P}(u \text{ is bad}) \le n^{2/3} \exp\left(-\left(\frac{(c-1)^2}{8c} + O(n^{-1/3})\right) b \log n\right)$$

which is O(1/n) for an appropriate choice of b.

## Galton–Watson Process

- To show convergence of  $\mathbb{P}(\tilde{T}_{n,p} \leq t^{-})$  and  $\mathbb{P}(\tilde{T}_{n-t^{-},p} \leq t^{-})$  to  $p_e$ , we introduce a parallel version of the process: the Galton–Watson branching process.
- There, we set  $Y_0 = 1$ , and  $Y_{n+1} = \sum_{i=1}^{Y_n} Z_i^{(n)}$  where the  $Z_i^{(n)}$  are i.i.d.
- Using the probability-generating function  $g_X(s) = \mathbb{E}s^X$  for integer-valued random variables X, we get  $g_{X_{n+1}}(s) = g_Z(g_{X_n}(s))$ .
- We have  $\mathbb{P}(X_n = 0) = g_Z^n(0)$  and thus  $p_e = g_Z(p_e)$  for the extinction probability  $p_e = \mathbb{P}(\exists n \colon X_n = 0)$ .
- For a binomial random variable Z, we have  $g_Z(s) = ((1-p) + ps)^n$ .
- Both probability-generating functions  $\left(1 + \frac{c(s-1)}{n}\right)^n$  and  $\left(1 + \frac{c(s-1)}{n}\right)^{n-t^-}$  converge to  $e^{c(1-s)}$ .
- Let  $p_e$  be the nontrivial  $(s \neq 1)$  solution of the equation  $s = e^{c(s-1)}$ . This solution is unique and asymptotically equal to the extinction probability.
- Using the approximation  $|\mathbb{P}(X \in A) \mathbb{P}(Y \in A)| \leq np^2$  for all sets A of nonnegative integers whenever  $X \sim \text{Bin}(n, p)$  and  $X \sim \text{Poi}(np)$ , we can show convergence of the finite-cutoff probabilities to  $p_e$ .