

# Lecture 2b: Phase Transition in Random Graphs

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Supplemental Material - 2024

# Analytical Results

## Distributions

Today we discuss about phase transition in random graphs. Recall on the *Erdős-Rényi class*  $\mathcal{G}_{n,p}$  of random graphs, the probability mass function on  $\mathcal{G}$ ,  $P : \mathcal{G} \rightarrow [0, 1]$ , is obtained by assuming that, as random variables, edges are independent from one another, and each edge occurs with probability  $p \in [0, 1]$ . Thus a graph  $G \in \mathcal{G}$  with  $m$  vertices will have probability  $P(G)$  given by

$$P(G) = p^m(1 - p)^{\binom{n}{2}}.$$

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## Distributions

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$$P(G) = p^m (1 - p)^{\binom{n}{2} - m}.$$

Recall the expected number of  $q$ -cliques  $X_q$  is

$$\mathbb{E}[X_q] = \binom{n}{q} p^{q(q-1)/2}$$

# Analytical Results

## Distributions

We shall also use  $\Gamma^{n,m}$  the set of all graphs on  $n$  vertices with  $m$  edges. The set  $\Gamma^{n,m}$  has cardinal

$$\binom{\binom{n}{2}}{m}.$$

In  $\Gamma^{n,m}$  each graph is equally probable.

# Analytical Results

## Cliques

The case of 3-cliques:  $\mathbb{E}[X_3] = \theta n^3 p^3$  ( $\theta \sim \frac{1}{6}$ ).

The case of 4-cliques:  $\mathbb{E}[X_4] = \theta n^4 p^6$  ( $\theta \sim \frac{1}{24}$ ).

The first problem we consider is the size of the largest clique of a random graph.

Note, finding the size of the largest clique (called *the clique number*) is a NP-hard problem.

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**Idea:** Analyze  $p$  so that  $\mathbb{E}[X_q] \approx 1$ .

- For  $p > \frac{1}{n}$  and large  $n$  we expect that graphs will have a 3-clique;
- For  $p > \frac{1}{n^{2/3}}$  and large  $n$ , we expect that graphs will have a 4-clique;

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Question: How sharp are these thresholds?

# Analytical Results

## 3-Cliques

### Theorem

Let  $p = p(n)$  be the edge probability in  $\mathcal{G}_{n,p}$ .

- 1 If  $p \gg \frac{1}{n}$  (i.e.  $\lim_{n \rightarrow \infty} np = \infty$ ) then  
 $\lim_{n \rightarrow \infty} \text{Prob}[G \in \mathcal{G}_{n,p} \text{ has a 3-clique}] \rightarrow 1.$
- 2 If  $p \ll \frac{1}{n}$  (i.e.  $\lim_{n \rightarrow \infty} np = 0$ ) then  
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# Analytical Results

## 4-Cliques

### Theorem

Let  $p = p(n)$  be the edge probability in  $\mathcal{G}_{n,p}$ .

- 1 If  $p \gg \frac{1}{n^{2/3}}$  (i.e.  $\lim_{n \rightarrow \infty} n^{2/3}p = \infty$ ) then  $\lim_{n \rightarrow \infty} \text{Prob}[G \in \mathcal{G}_{n,p} \text{ has a 4-clique}] \rightarrow 1$ .
- 2 If  $p \ll \frac{1}{n^{2/3}}$  (i.e.  $\lim_{n \rightarrow \infty} n^{2/3}p = 0$ ) then  $\lim_{n \rightarrow \infty} \text{Prob}[G \in \mathcal{G}_{n,p} \text{ has a 4-clique}] \rightarrow 0$ .

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Let  $m = m(n)$  be the number of edges in  $\Gamma^{n,m}$ .

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- 2 If  $m \ll n^{4/3}$  (i.e.  $\lim_{n \rightarrow \infty} \frac{m}{n^{4/3}} = 0$ ) then  $\lim_{n \rightarrow \infty} \text{Prob}[G \in \Gamma^{n,m} \text{ has a 4-clique}] \rightarrow 0$ .

# Analytical Results

## $q$ -Cliques

### Theorem

Let  $p = p(n)$  be the edge probability in  $\mathcal{G}_{n,p}$ . Let  $q \geq 3$  be an integer.

- 1 If  $p \gg \frac{1}{n^{2/(q-1)}}$  (i.e.  $\lim_{n \rightarrow \infty} n^{2/(q-1)} p = \infty$ ) then  $\lim_{n \rightarrow \infty} \text{Prob}[G \in \mathcal{G}_{n,p} \text{ has a } q\text{-clique}] \rightarrow 1$ .
- 2 If  $p \ll \frac{1}{n^{2/(q-1)}}$  (i.e.  $\lim_{n \rightarrow \infty} n^{2/(q-1)} p = 0$ ) then  $\lim_{n \rightarrow \infty} \text{Prob}[G \in \mathcal{G}_{n,p} \text{ has a } q\text{-clique}] \rightarrow 0$ .

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Let  $m = m(n)$  be the number of edges in  $\Gamma^{n,m}$ . Let  $q \geq 3$  be and integer.

- 1 If  $m \gg n^{2(q-2)/(q-1)}$  (i.e.  $\lim_{n \rightarrow \infty} \frac{m}{n^{2(q-2)/(q-1)}} = \infty$ ) then  
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# Analytical Results

## Markov and Chebyshev Inequalities

We want to control probabilities of the random event  $X_3(G) > 0$ . Two important tools:

- ① (Markov's Inequality) Assume  $X$  is a non-negative random variable. Then  $Prob[X \geq t] \leq \frac{\mathbb{E}[X]}{t}$ .
- ② (Chebyshev's Inequality) For any random variable  $X$ ,  $Prob[|X - E[X]| \geq t] \leq \frac{Var[X]}{t^2}$ .

where  $\mathbb{E}[X]$  is the mean of  $X$ , and  $Var[X] = \mathbb{E}[X^2] - |\mathbb{E}[X]|^2$  is the variance of  $X$ .

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where  $\mathbb{E}[X]$  is the mean of  $X$ , and  $\text{Var}[X] = \mathbb{E}[X^2] - |\mathbb{E}[X]|^2$  is the variance of  $X$ . *Quick Proof:*

$$\text{Prob}[X \geq t] = \int_t^\infty p_X(x) dx \leq \frac{1}{t} \int_t^\infty x p_X(x) dx \leq \frac{\mathbb{E}[X]}{t}.$$

$$\text{Prob}[|X - \mathbb{E}[X]| \geq t] = P[|X - \mathbb{E}[X]|^2 \geq t^2] \leq \frac{\mathbb{E}[|X - \mathbb{E}[X]|^2]}{t^2} = \frac{\text{Var}[X]}{t^2}.$$

# Analytical Results

## Proofs for the 3-clique case

For small probability: We shall use Markov's inequality to show

$Prob[X_3 > 0] \rightarrow 0$  when  $p \ll \frac{1}{n}$ :

$$Prob[X_3 > 0] = Prob[X_3 \geq 1] \leq \frac{E[X_3]}{1} = \frac{n(n-1)(n-2)}{6} p^3 = \theta n^3 p^3 \rightarrow 0.$$



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For large probability: Since  $\mathbb{E}[X_3] \rightarrow \infty$  it follows that  $Prob[X_3 > 0] > 0$ .

We need to show that  $Prob[X_3 = 0] \rightarrow 0$ . By Chebyshev's inequality:

$$Prob[X_3 = 0] \leq Prob[|X_3 - \mathbb{E}[X_3]| \geq \mathbb{E}[X_3]] \leq \frac{Var[X_3]}{|\mathbb{E}[X_3]|^2}$$

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Need the variance of  $X_3 = \sum_{(i,j,k) \in S_3} 1_{i,j,k}$ ,

$$X_3^2 = \sum_{(i,j,k) \in S_3} \sum_{(i',j',k') \in S_3} 1_{i,j,k} 1_{i',j',k'}.$$

# Analytical Results

## Proofs for the 3-clique case

$$\begin{aligned}
 X_3^2 = & \sum_{(i,j,k) \in \mathcal{S}_3(n)} 1_{i,j,k} + \sum_{(i,j,k) \in \mathcal{S}_3(n)} \sum_{l \in \mathcal{S}_1(n-3)} (1_{i,j,k} 1_{i,j,l} + 1_{i,j,k} 1_{j,k,l} + 1_{i,j,k} 1_{k,i,l}) + \\
 & + \sum_{(i,j,k) \in \mathcal{S}_3(n)} \sum_{u,v \in \mathcal{S}_2(n-3)} (1_{i,j,k} 1_{i,u,v} + 1_{i,j,k} 1_{j,u,v} + 1_{i,j,k} 1_{k,u,v}) + \\
 & + \sum_{(i,j,k) \in \mathcal{S}_3(n)} \sum_{(i',j',k') \in \mathcal{S}_3(n-3)} 1_{i,j,k} 1_{i',j',k'}
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 & + \sum_{(i,j,k) \in S_3(n)} \sum_{(i',j',k') \in S_3(n-3)} 1_{i,j,k} 1_{i',j',k'}
 \end{aligned}$$

$$\mathbb{E}[X_3^2] = |S_3| p^3 + 3|S_3|(n-3)p^5 + 3|S_3| \binom{n-3}{2} p^6 + |S_3| \binom{n-3}{3} p^6.$$

Thus

$$\text{Var}[X_3] = \mathbb{E}[X_3^2] - |\mathbb{E}[X_3]|^2 = \dots = \theta(n^3 p^3 + n^4 p^5 + n^5 p^6).$$

# Analytical Results

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and:

$$\text{Prob}[X_3 = 0] \leq \frac{\theta(n^3 p^3 + n^4 p^5 + n^5 p^6)}{\theta(n^6 p^6)} = \frac{1}{(np)^3} + \frac{1}{n} \rightarrow 0.$$

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Similar proofs for the other cases (4-cliques and  $q$ -cliques).

# Analytical Results

## Behavior at the threshold

In general we obtain a "coarse threshold". Recall a Poisson random variable  $X$  with parameter  $\lambda$  has p.m.f.  $Prob[X = k] = e^{-\lambda} \frac{\lambda^k}{k!}$ .

### Theorem

In  $\mathcal{G}_{n,p}$ ,

- 1 For  $p = \frac{c}{n}$ ,  $X_3$  is asymptotically Poisson with parameter  $\lambda = c^3/6$ .
- 2 For  $p = \frac{c}{n^{2/3}}$ ,  $X_4$  is asymptotically Poisson with parameter  $\lambda = c^6/24$ .

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### Theorem

In  $\Gamma^{n,m}$ ,

- 1 For  $m = cn$ ,  $X_3$  is asymptotically Poisson with parameter  $\lambda = 4c^3/3$ .
- 2 For  $m = cn^{4/3}$ ,  $X_4$  is asymptotically Poisson with parameter  $\lambda = 8c^6/3$ .



# Analytical Results

## Connected Components

$\mathcal{G}_{n,p}$  class of random graphs has a remarkable property in regards to the largest connected component. We shall express the result in the class  $\Gamma^{n,m}$ .

# Analytical Results

## Connected Components

### Theorem

- ① Let  $m = m(n)$  satisfies  $m \ll \frac{1}{2}n \log(n)$ . Then

$$\lim_{n \rightarrow \infty} \text{Prob}[G \in \Gamma^{n,m} \text{ is connected}] = 0$$

- ② Let  $m = m(n)$  satisfies  $m \gg \frac{1}{2}n \log(n)$ . Then

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- ③ Assume  $m = \frac{1}{2}n \log(n) + tn + o(n)$ , where  $o(n) \ll n$ . Then

$$\lim_{n \rightarrow \infty} \text{Prob}[G \in \Gamma^{n,m} \text{ is connected}] = e^{-e^{-2t}}$$

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In this case  $\frac{1}{2}n \log(n)$  is known as a *strong threshold*. 

# Numerical Results

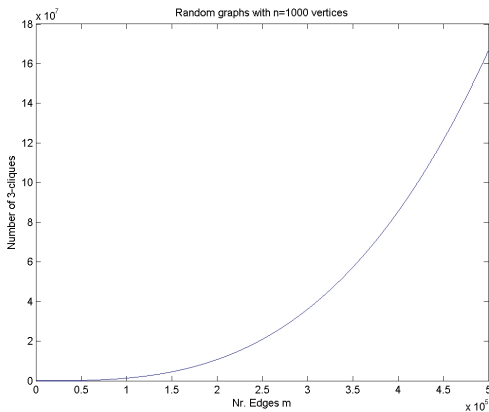
## 3-cliques & Connectivity results

Results for  $n = 1000$  vertices.

- 1 3-cliques. Recall  $\mathbb{E}[X_3] \sim m^3$
- 2 Connectivity. Recall the connectivity threshold is  $\frac{1}{2}n \log(n) = 3454$ .

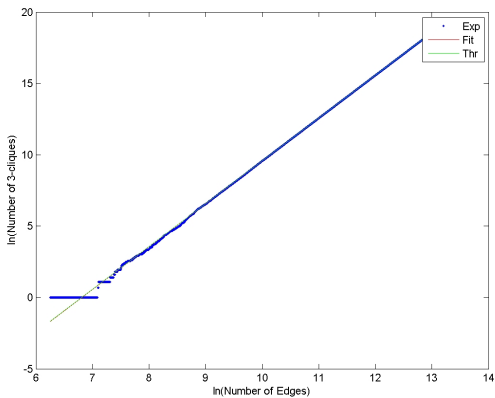
# Numerical Results

## 3-cliques



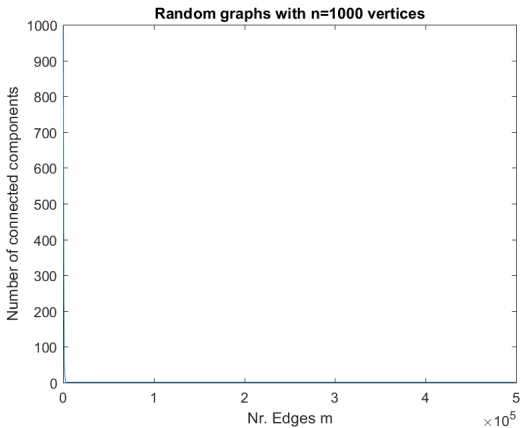
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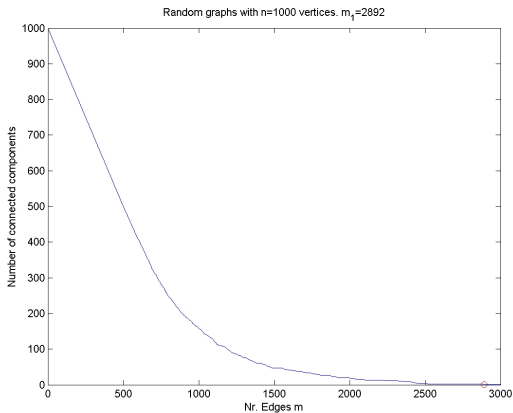
## Connectivity





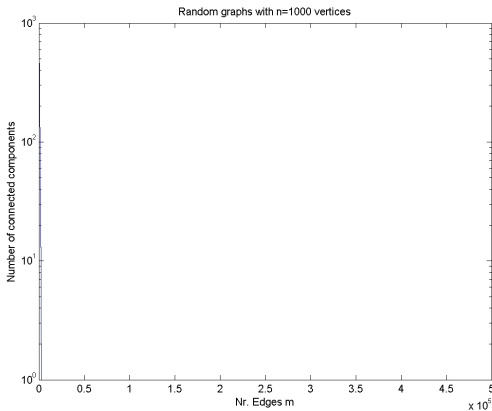
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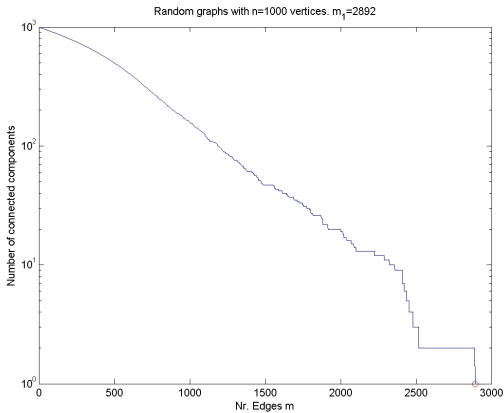
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# Numerical Results

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