

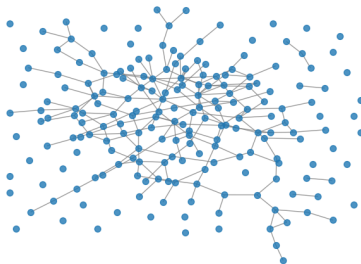
CS4423: Networks

## Week 8, Part 1: Introduction to Random Networks

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(05 March 2025)



# Class Test!

Answers (but not solutions)  
posted by 11 today.

Class Test 2pm tomorrow!

## Details:

- ▶ LENS reports: email Niall today!
- ▶ Locations: see announcement!
- ▶ Content:

- ▶ Similar to Problem Set 2
- ▶ Nothing from this week.
- ▶ No `networkx`
- ▶ Focus on skills, rather than theory.

- ▶ Bring a pen. And maybe a calculator (?).

- ▶ If you miss the test, for any reason, your grade will be based on the assignments (20%) and the final exam (80%).

4FMZ : CA116a

Everyone else:  
Larmor.

nothing on Betweenness  
Centrality.

# Outline

Today's notes are split between these slides, and a Jupyter Notebook.

- |   |                                 |   |                            |
|---|---------------------------------|---|----------------------------|
| 1 | Random Models of Networks       | 3 | Random samples             |
| 2 | Erdős-Rényi Random Graph Models | 4 | The two Erdős-Rényi Models |
|   | ■ Some examples                 |   | ■ Model A: $G_{ER}(n, m)$  |
|   |                                 |   | ■ Model B: $G_{ER}(n, p)$  |

Slides are at:

<https://www.niallmadden.ie/2425-CS4423>



# Random Models of Networks

## Random Models of Networks

One of the remaining “big” ideas for us to study in CS4423 is that of **Random Networks**. In a sense, we are not so interested in their randomness. It is more like we decide on the general structure of networks, but then choose a particular example by tossing a coin, or rolling dice.

What we are interested in:

- ▶ The **statistical properties** of very large networks, such as average degree, the number of 3-cycles, or the size of component.
- ▶ How well our random networks share these properties.

# Erdő-Rényi Random Graph Models

A **Random Graph**<sup>1</sup> is a *mathematical model* of a family of networks, where certain parameters (like the number of nodes and edges) have fixed values, but other aspects (like the actual edges) are randomly assigned.

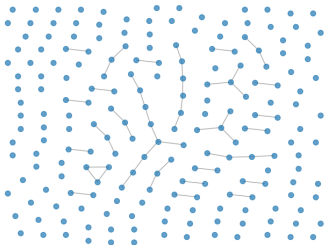
The simplest example of a random graph is in fact a network with fixed numbers  $n$  of nodes and  $m$  of edges, randomly placed between the vertices.

Although a random graph is not a specific object, many of its properties can be described precisely in the form of **expected values** or **probability distributions**.

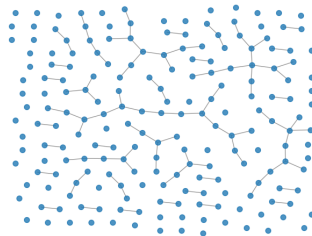
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<sup>1</sup>[https://en.wikipedia.org/wiki/Random\\_graph](https://en.wikipedia.org/wiki/Random_graph)

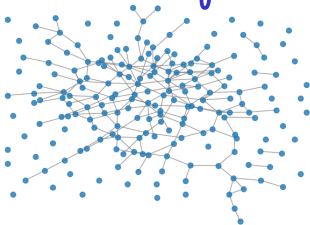
20 edges



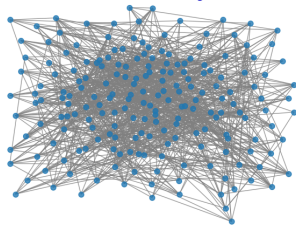
50 edges



100 edges



1000 edges



All have  $n = 200$  nodes.

## Random samples

$$N = \binom{n}{2}$$

Suppose our network  $G = (X, E)$  has  $|X| = n$  nodes. Then we know the most number of edges  $E$  can have is:

$$\binom{n}{2} = \frac{n!}{(n-2)!2!} = \frac{n(n-1)}{2}. \quad (\text{roughly } \frac{1}{2}n^2)$$

- Our goal is to randomly select edges on the vertex set  $X$ . That is, pick at random elements from the set  $\binom{X}{2}$  of pairs of nodes.
- So we need a procedure for selecting  $m$  from  $N$  objects randomly, in such a way that each of the  $\binom{N}{m}$  subsets of the  $N$  objects is an equally likely outcome.

- We first discuss sampling  $m$  values in the range  $\{0, 1, \dots, N-1\}$ .  
Recall  $\binom{X}{2}$  is the set of pairs drawn from  $X$ . So  $|\binom{X}{2}| = \binom{|X|}{2}$

# Random samples

1. Suppose we choose a natural number  $N$ , and real number  $p \in [0, 1]$
2. Then iterate over each element of the set  $\{0, 1, \dots, N - 1\}$ .
3. For each, we pick a random number  $x \in [0, 1]$ .
4. If  $x < p$ , we keep that number. Otherwise remove it from the set.

When we are done, how many elements do we expect in the set if  $p = m/N$  for some chosen  $m$ ?

And what is the likelihood of there being, say  $k$  elements in the set?



# Random samples

We are creating random samples. The size of each is a random number,  $k$ .

**Claim: Expected value:**  $E[k] = Np = m$ .

**Proof:** This is a **binomial distribution**<sup>2</sup>

- ▶ The probability of a specific subset of size  $k$  to be chosen is  $p^k(1-p)^{N-k}$ .
- ▶ There are  $\binom{N}{k}$  subsets of size  $k$ . So the probability  $P(k)$  of the sample to have size  $k$  is  $P(k) = \binom{N}{k} p^k (1-p)^{N-k}$ .

We use the following facts

- (i)  $j \binom{N}{j} p^j = \underbrace{Np}_{\text{circled in green}} \binom{N-1}{j-1} p^{j-1}$ ,
- (ii)  $(1-p)^{N-j} = (1-p)^{(N-1)-(j-1)}$ ,
- (iii)  $(p + (1-p))^r = 1$  for all  $r$ .

<sup>2</sup>[https://en.wikipedia.org/wiki/Binomial\\_distribution](https://en.wikipedia.org/wiki/Binomial_distribution)

# Random samples

Expected value:

$$\begin{aligned} E[k] &= \underbrace{\sum_{j=0}^N jP(j)}_{\text{weighted average of } j} = \sum_{j=0}^N j \underbrace{\binom{N}{j} p^j (1-p)^{N-j}}_{\text{Formula for } P(j)} \\ &= Np \underbrace{\left[ \sum_{l=0}^{N-1} \binom{N-1}{l} p^l (1-p)^{(N-1)-l} \right]}_{\text{From (i),(ii),(ii)}} = Np, \quad (1) \end{aligned}$$

*Handwritten notes: A green bracket under the sum in the second line is labeled "From (i),(ii),(ii)". A green arrow points from the expression in brackets to "1." on the right. Another green arrow points from the final result "Np" to the right.*

substituting  $l = k - 1$ ,

↑  
Exer: check this!

# Random samples

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Next week, we'll look at some computational examples, as well as an algorithm for choosing exactly  $m$  numbers from a set of  $N$ .

For now, we'll just assume it can be done...

Important:  
nodes are labelled

Uniformly selected edges

### ER Model $G_{ER}(n, m)$ : Uniform Random Graphs

Let  $n \geq 1$ , let  $N = \binom{n}{2}$  and let  $0 \leq m \leq N$ .

The model  $G_{ER}(n, m)$  consists of the ensemble of graphs  $G$  on the  $n$  nodes  $X = \{0, 1, \dots, n-1\}$ , and  $m$  randomly selected edges, chosen uniformly from the  $N = \binom{n}{2}$  possible edges.

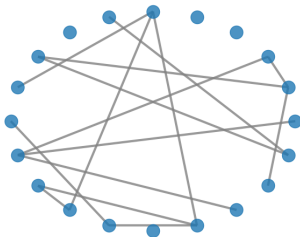
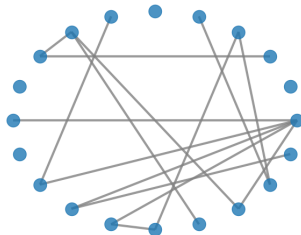
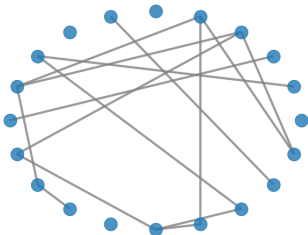
Equivalently, one can choose uniformly at random one network in the **set**  $\mathcal{G}(n, m)$  of *all* networks on a given set of  $n$  nodes with *exactly*  $m$  edges.

# The two Erdős-Rényi Models

Model A:  $G_{ER}(n, m)$

N

$n=20, m=20(?)$



Finished here  
Wed at 10,  
with an example  
of  $G(3,2)$   
on the board.

Randomly selected edges

**ER Model  $G_{ER}(n, p)$ : Random Edges**

Let  $n \geq 1$ , let  $N = \binom{n}{2}$  and let  $0 \leq p \leq 1$ .

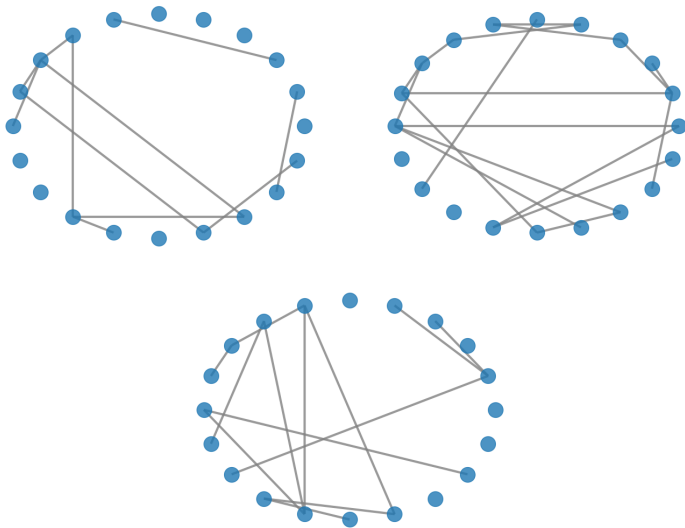
The model  $G_{ER}(n, p)$  consists of the ensemble of graphs  $G$  on the  $n$  nodes  $X = \{0, 1, \dots, n-1\}$ , with each of the possible  $N = \binom{n}{2}$  edges chosen with probability  $p$ .

The probability  $P(G)$  of a particular graph  $G = (X, E)$  with  $X = \{0, 1, \dots, n-1\}$  and  $m = |E|$  edges in the  $G_{ER}(n, p)$  model is

$$P(G) = p^m(1 - p)^{N-m}.$$

# The two Erdős-Rényi Models

Model B:  $G_{ER}(n, p)$



Of the two models,  $G_{ER}(n, p)$  is the more studied. They are many similarities, but do differ. For example:

1.  $G_{ER}(n, m)$  will have  $m$  edges with probability 1.
2. A graph in  $G_{ER}(n, p)$  will have  $m$  edges with probability  $\binom{N}{m} p^m (1-p)^{N-m}$ .

