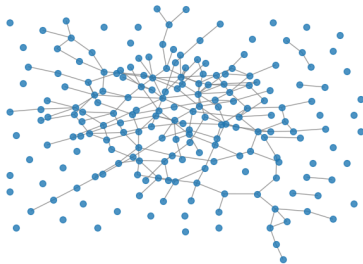


# Week 8, Part 1: Introduction to Random Networks

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# Class Test!

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%).

# 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 <ul style="list-style-type: none"><li>■ Some examples</li></ul> | 4 | The two Erdős-Rényi Models <ul style="list-style-type: none"><li>■ Model A: <math>G_{ER}(n, m)</math></li><li>■ Model B: <math>G_{ER}(n, p)</math></li></ul> |

Slides are at:

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



## 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.

# Erdős-Rényi Random Graph Models

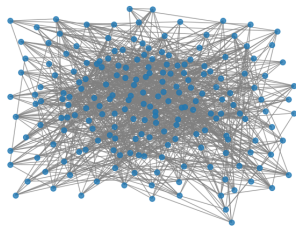
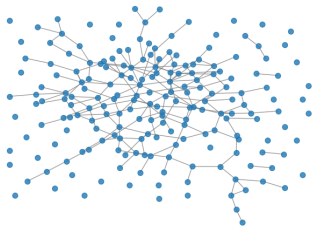
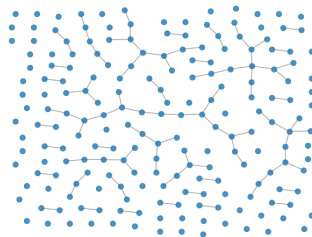
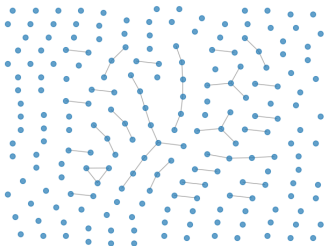
A **Random Graph**<sup>1</sup> is a *mathematical model* of a family of networks, where certain parameters (like the number of nodes and edges) have fixed values, but other aspects (like the actual edges) are randomly assigned.

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)



## Random samples

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

$$\binom{n}{2} = \frac{n!}{(n-2)!2!} = \frac{n(n-1)}{2}.$$

- ▶ 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  $\{0, 1, \dots, N-1\}$ .

## 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)  $j\binom{N}{j}p^j = Np\binom{N-1}{j-1}p^{j-1}$ ,
- (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:

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

substituting  $l = k - 1$ ,

## Random samples

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...

Uniformly selected edges

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

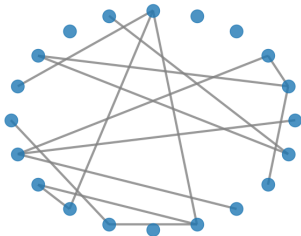
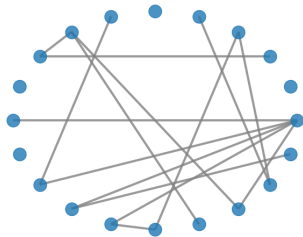
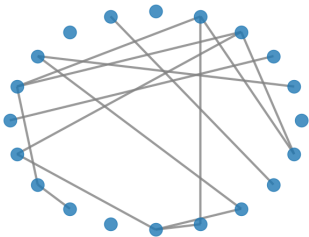
Let  $n \geq 1$ , let  $N = \binom{n}{2}$  and let  $0 \leq m \leq N$ .

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



Randomly selected edges

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

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

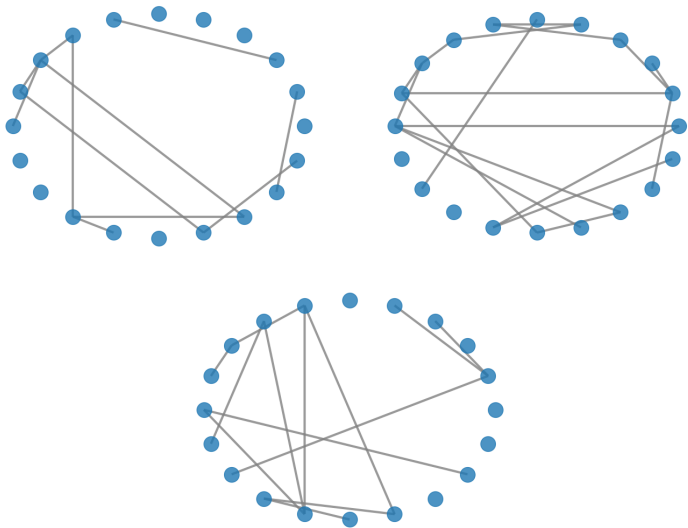
The model  $G_{ER}(n, p)$  consists of the ensemble of graphs  $G$  on the  $n$  nodes  $X = \{0, 1, \dots, n-1\}$ , with each of the possible  $N = \binom{n}{2}$  edges chosen with probability  $p$ .

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  $\binom{N}{m} p^m (1-p)^{N-m}$ .

