

# BST 267: Introduction to Social and Biological Networks

## Lab 4

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# Modeling Network Structure

- The goal of this lab is to learn how to generate ER random graphs and to learn about their phase transition
- **Deliverable:** Return these through Canvas: 1) Jupyter Notebook (.ipynb file) and 2) HTML version of the notebook (.html file)

# Question 1

- We will generate  $G(N, p)$  graphs without using the `nx.erdos_renyi_graph` function
- Write code that generates a graph from the  $G(N, p)$  ensemble (the classic Erdős-Rényi graph)
- The code will need two parameters: the number of nodes  $N$  and the connection probability  $p$
- Function `random.random` generates a random number in the  $[0, 1)$  interval
- We can use it to simulate the outcome of a Bernoulli variable:

```
1 import random
2 if random.random() < 0.5:
3     print("Heads")
4 else:
5     print("Tails")
```

- An alternative to `random.random` is `scipy.stats.bernoulli.rvs`:

```
1 from scipy.stats import bernoulli
2 bernoulli.rvs(p=0.2)
3 bernoulli.rvs(p=0.2, size=20)
```

- Use the following parameters:  $N = 100, p = 0.05$

- The easiest way to generate Erdős-Rényi graphs in NetworkX is using the `nx.erdos_renyi_graph` function:

```
1 import matplotlib.pyplot as plt
2 %matplotlib notebook
3 G = nx.erdos_renyi_graph(20, 0.15, seed=123)
4 nx.draw(G)
```

- Erdős-Rényi networks show a percolation (phase) transition when the link probability (link density)  $p \approx 1/N$  where  $N$  is the number of nodes in the network
- Since average degree in Erdős-Rényi networks is given by  $\langle k \rangle = p(N - 1) \approx pN$ , the transition happens when  $\langle k \rangle = 1$

## Question 2

- Study this transition by making a plot of the relative size of the largest connected component  $R_{LCC}(p)$  as a function of  $p$  around  $p \approx 1/N$
- You can use a small network to do this (e.g.,  $N = 50$ )
- Instead of investigating one realization for each value of  $p$ , calculate the average value of  $R_{LCC}(p)$  over 10 realizations and plot that against  $p$
- If your network is stored in  $G$ , you can extract the largest connected component of  $G$  using the following:

```
1 Gc = max(nx.connected_component_subgraphs(G), key=len)
```

- You can import the NumPy module using `import numpy as np`
- You can calculate the mean of a sequence using the `np.mean` function
- You can create a semi-logarithmic plot using the `plt.semilogx` function
- You can create a NumPy array (a type of sequence) of evenly spaced  $p$  values using `ps = np.logspace(-np.log10(N)-1, -np.log10(N)+1, 40)`