Random Networks (ER Model)

Social Networks Analysis and Graph Algorithms

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Contents

- The ER model
- Degree distribution under the ER model

Sources

- A. L. Barabási (2016). Network Science Chapter 03
- Data-Driven Social Analytics course by Vicenç Gómez and Andreas Kaltenbrunner
- URLs cited in the footer of specific slides

Network Models

Network models

- Networks of many different types have similar properties:
 - Short paths
 - Many triangles
 - Skewed degree distributions
- Where do such properties come from?
- How do nodes connect to each other? How are triangles formed?
- We will study network models, i.e., sets of instructions to create networks

Why studying network models?

- Our models will be **stochastic**, i.e., randomized
- Running stochastic network models can let us check if they generate networks that look like real ones
- Almost invariably, the generated networks will be similar to actual networks in some ways, but different in other ways

The "Random Network" Erdös-Rényi (ER) Model

Sounds like "<u>ERDOSH</u> and <u>REGN</u>"



Paul Erdös (1913-1996)



Alfred Rényi (1921-1970)

Meeting people at a party

- You pick a random person
- Talk to that person for a while, if there are good vibes, you are connected
- Then pick another person
 - And repeat
- The result is what we call a random network



Formalization (Erdös-Rényi or ER) Sounds like "ERDOSH and REGN"

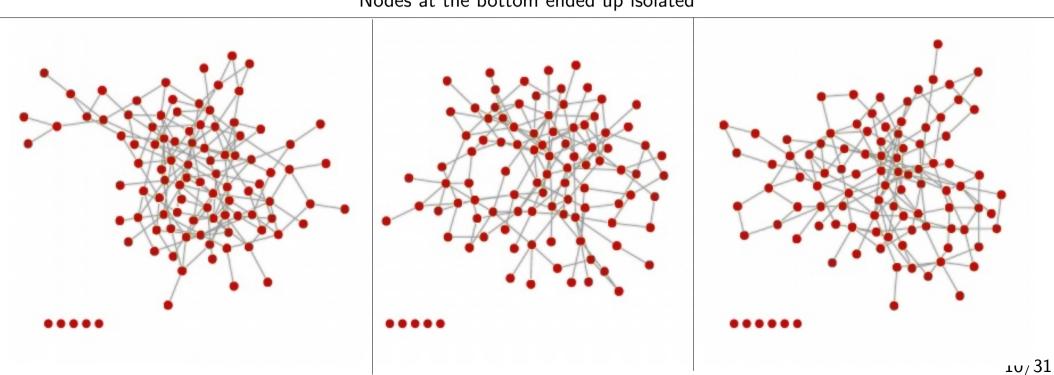
- For each pair of nodes in the graph
 - Perform a Bernoulli trial with probability p
 - "Toss a biased coin with probability p of landing heads"
 - If the trial succeeds, connect those nodes
 - "If the coin lands heads, connect those nodes"
- Repeat for all pairs $\frac{N(N-1)}{2}$

Example

(3 networks, same parameters)

$$N = 100, p = 0.03$$

Nodes at the bottom ended up isolated



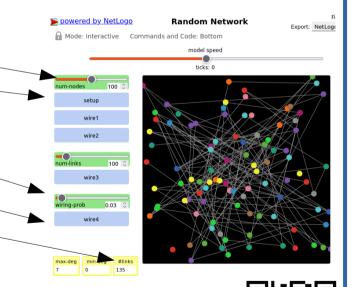
Exercise Guess formula for (L)

Go to netlogoweb.org/launch and select: "IABM Textbook / chapter 5 / Random Network"

• Execute the <u>"Random Network" program</u> in Netlogo Web

Select num-nodes N (e.g., 100)Click "setup"

- Select wiring-prob p (e.g., 0.03)
- Click "wire4"
- Write down "#links" L somewhere
- Repeat various times
- Guess a formula for $\langle L \rangle$ as a function of N and p





Pin board: https://upfbarcelona.padlet.org/chato/84a1nj59pkqpxvh3

Degree distribution

A key characteristic of a network: its degree distribution

- One of the most evident characteristics of a network is its degree distribution
 - Is this distribution very skewed? Or every node is close to some average?
 Is there a "typical" degree?
 - Does it look like the degree distribution predicted by a network formation model?
- We will spend a fair amount of time studying the degree distribution under various models

The binomial distribution

 The distribution of the probability of obtaining x successes in n independent trials, in which each trial has probability of succeeding p

$$p_x = \binom{n}{x} p^x (1-p)^{n-x}$$

$$\langle x \rangle = \sum_{x=0}^{n} x p_x = np$$

Degree distribution in ER model

- Simply a Binomial distribution
- Note that the maximum number of "successes" (links) of a node is N-1, hence:

$$p_k = {N-1 \choose k} p^k (1-p)^{N-1-k}$$
$$\langle k \rangle = p(N-1)$$

Expected number of links

Expected number of links

$$\langle L \rangle = p \cdot L_{\text{max}} = p \frac{N(N-1)}{2}$$

Average degree

$$\langle k \rangle = \frac{2 \langle L \rangle}{N} = p(N-1)$$

Exercise [B. 2016, Ex. 3.11.1]

Expected number of links and average degree

- Consider an ER graph with $N=3,000 p=10^{-3}$
 - 1) What is the expected number of links (L)?
 - 2) What is the average degree (k)?

$$\langle L \rangle = p \cdot L_{\text{max}} = p \frac{N(N-1)}{2}$$

$$\langle k \rangle = \frac{2\langle L \rangle}{N} = p(N-1)$$

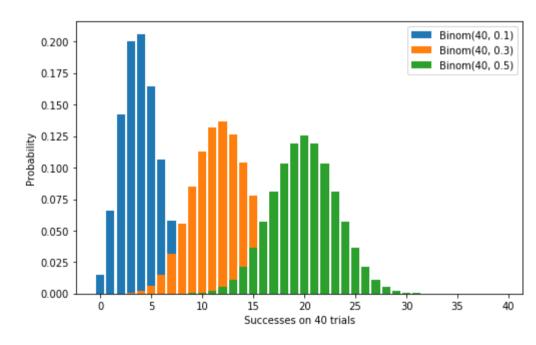
Degree distribution examples

• The peak is always at

$$\langle k \rangle = p(N-1)$$

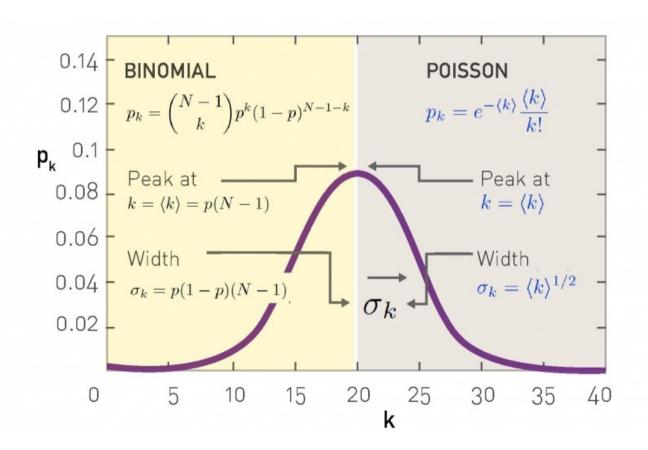
```
import numpy as np
from scipy.stats import binom
from matplotlib import pyplot as plt

x = np.arange(0, 40)
plt.figure(figsize=(8,5))
plt.bar(x, (binom(40, 0.1)).pmf(x), label='Binom(40, 0.1)')
plt.bar(x, (binom(40, 0.3)).pmf(x), label='Binom(40, 0.3)')
plt.bar(x, (binom(40, 0.5)).pmf(x), label='Binom(40, 0.5)')
plt.gca().legend()
plt.xlabel("Successes on 40 trials")
plt.ylabel("Probability")
plt.show()
```



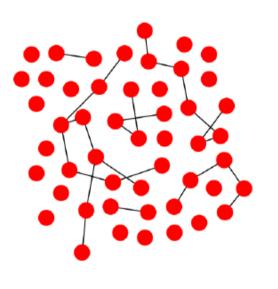


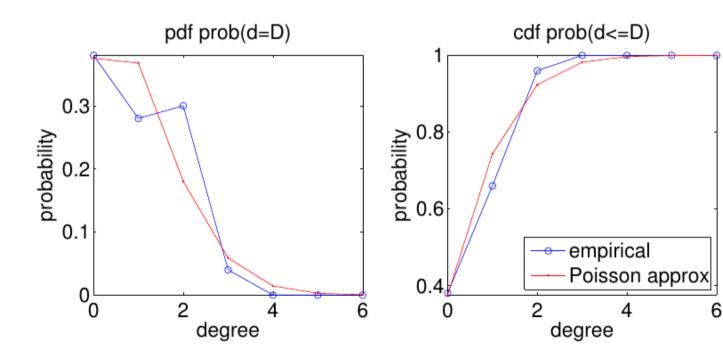
Approximation with a Poisson distribution for $\langle k \rangle \ll N$



More examples (1/6)

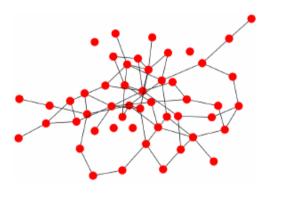
$$N = 50, p = 0.02, \langle k \rangle \approx 1$$

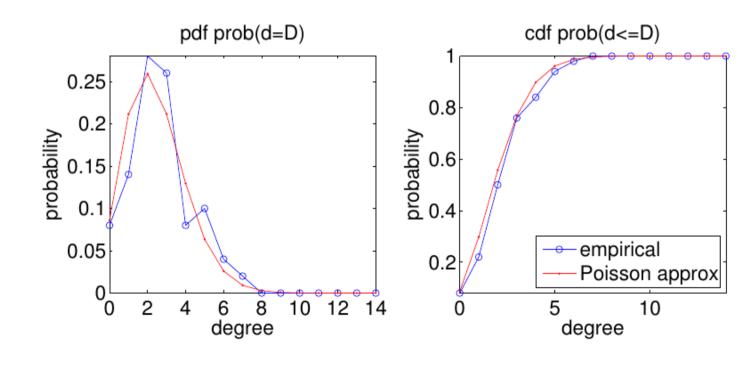




More examples (2/6)

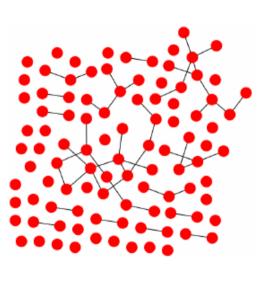
$$N = 50, p = 0.05, \langle k \rangle \approx 2.5$$

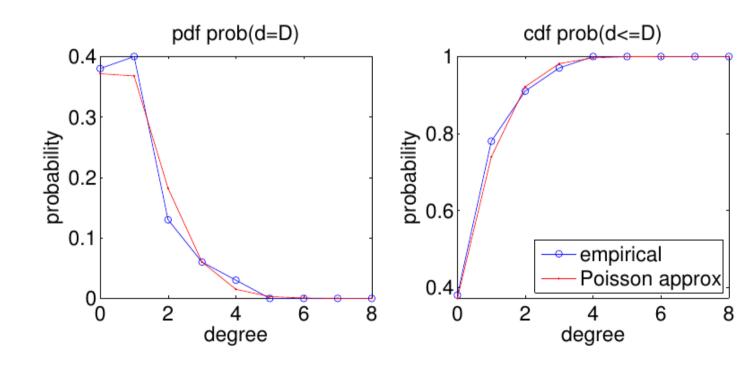




More examples (3/6)

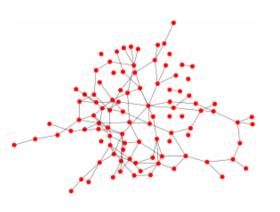
$$N = 100, p = 0.01, \langle k \rangle \approx 1$$

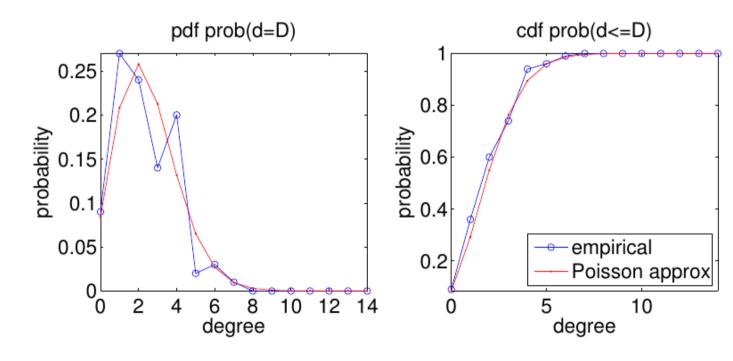




More examples (4/6)

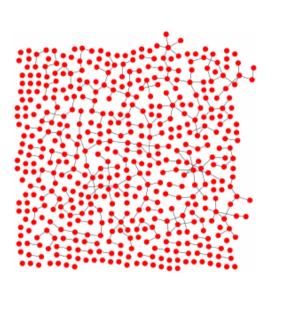
$$N = 100, p = 0.025, \langle k \rangle \approx 2.5$$

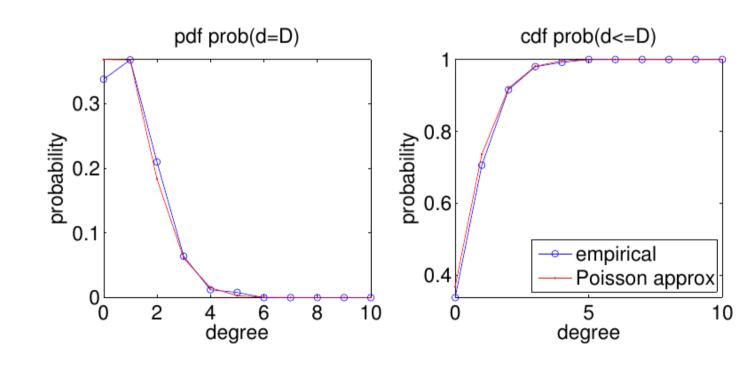




More examples (5/6)

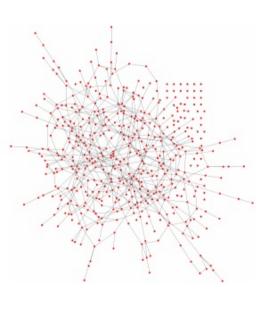
$$N = 500, p = 0.002, \langle k \rangle \approx 1$$

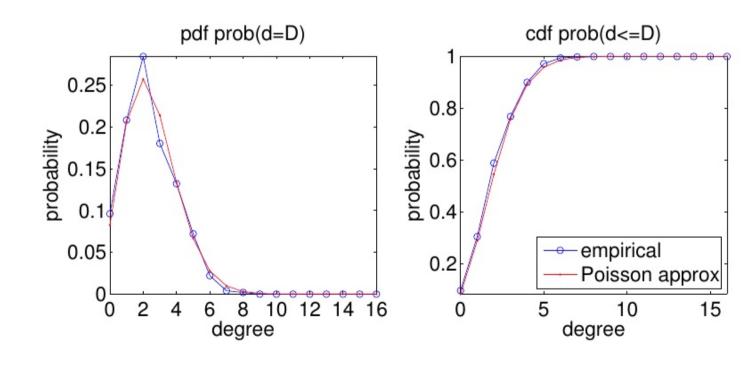




More examples (6/6)

$$N = 500, p = 0.005, \langle k \rangle \approx 2.5$$





"Back of the envelope" calculations

- Suppose $N = 7 \times 10^9$
- Suppose <k> = 1,000
 - A person knows the name of approx. 1,000 others
- Then on expectation $k_{max} = 1,185$
- <k $> \pm \sigma$ is the range from 968 to 1,032
- Is this realistic?

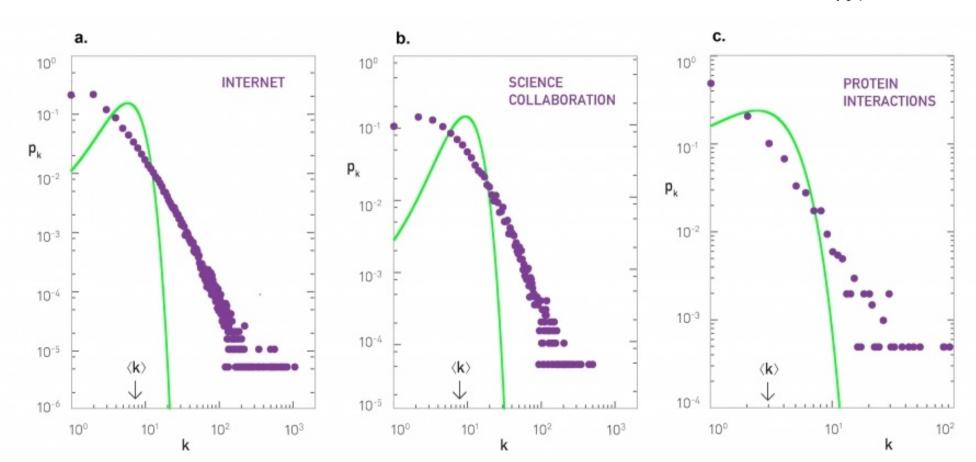
Survey: how many WhatsApp contacts do you have?



https://forms.gle/9xEYhzv2U5NrPQdH8

Real networks (green =

$$e^{-\langle k \rangle} \frac{\langle k \rangle^k}{k!}$$
)



Summary

Things to remember

- The ER model
- Degree distribution in the ER model

Practice on your own

• Indicate the expected number of edges of a network with N=256, p=0.25; then compare your solution with the one on this video:



https://www.youtube.com/watch?v=2DckiyysQy4