

# Preferential Attachment (BA Model)

Introduction to Network Science

Carlos Castillo

Topic 11

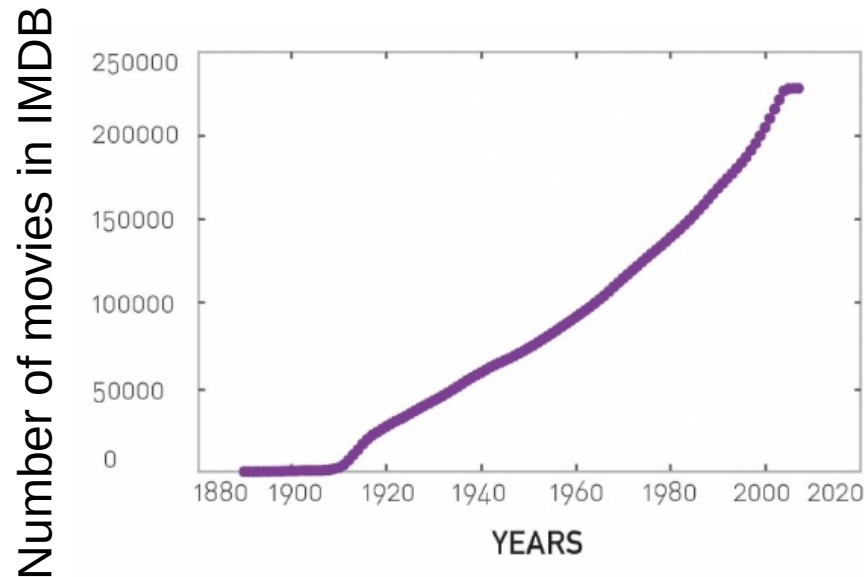
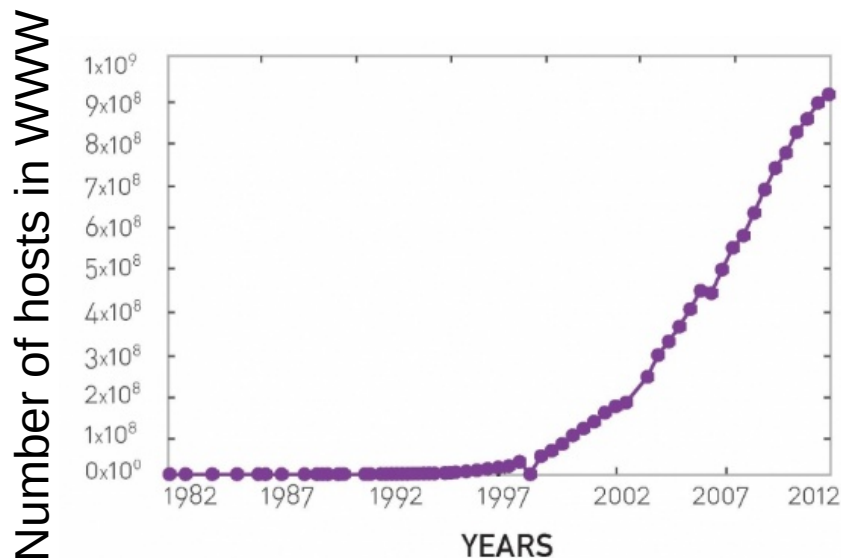
# Contents

- The uniform random attachment model
- The BA or preferential attachment model
- Degree distribution under the BA model
- Distance distribution under the BA model
- Clustering coefficient under the BA model

# Sources

- Albert-László Barabási (2016) Network Science
  - Preferential attachment follows chapter 05
- Ravi Srinivasan 2013 Complex Networks Ch 12
- Networks, Crowds, and Markets Ch 18
- Data-Driven Social Analytics course by Vicenç Gómez and Andreas Kaltenbrunner

# The number of nodes $N$ increases: we need models of network growth



# Preliminary: Uniform Random Attachment

# Growth in an ER network

- Two assumptions in ER networks:
  - There are  $N$  nodes that **pre-exist**
  - Nodes connect **at random**
- Let's challenge the first assumption

# Uniform Attachment

- Network starts with  $m$  fully-connected nodes
- Time starts at  $t_0=m$
- At every time step we add 1 node
- This node will have  $m$  outlinks

# Expected degree over time

- Probability of obtaining one link:  $m/t$ 
  - Decreases over time
- Expected degree of node born at  $m < i < t$

$$m + \frac{m}{i} + \frac{m}{i+1} + \frac{m}{i+2} + \dots + \frac{m}{t} \approx m \left( 1 + \log \left( \frac{t}{i} \right) \right)$$



# Tail of degree distribution

- How many nodes of degree larger than  $K$  are there at time  $t$ ? (Computation in “Advanced materials” at the end of these slides)

$$e^{-\frac{K-m}{m}}$$

- Decreases exponentially with  $K$ : it's vanishingly rare to find high-degree nodes

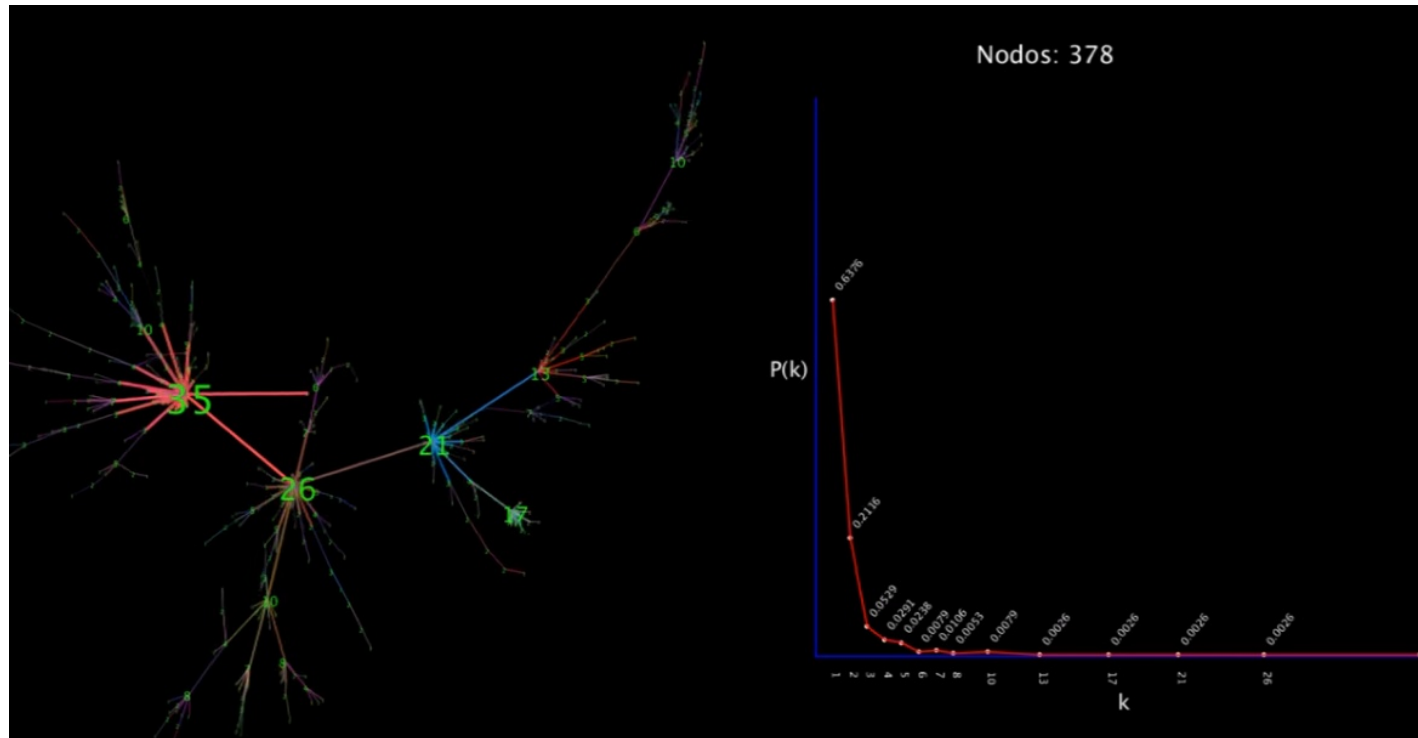
# Preferential Attachment

# Preferential attachment simulation



<https://www.youtube.com/watch?v=4GDqJVtPEGg>

# Degree distribution in simulation



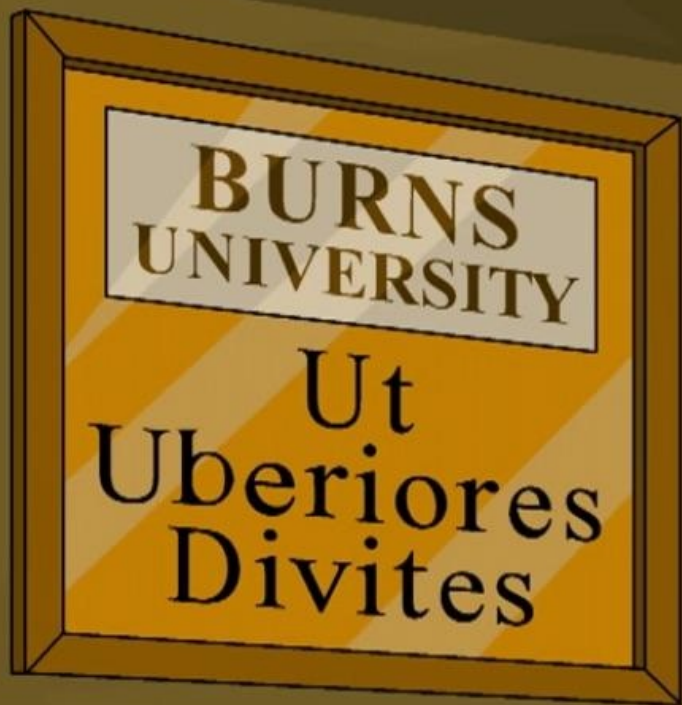
<https://www.youtube.com/watch?v=5RIQweqPT6A>

# We have seen what but not why

- Power-law degree distributions are prevalent
  - Why?
- Two assumptions in ER networks:
  - There are  $N$  nodes that **pre-exist**
  - Nodes connect **at random**
- Let's challenge both assumptions

# Growth

- Suppose there are two web pages on a topic, one with many inlinks the other with few, which one am I most likely to link to?
- Which scientific papers are read?
- Which book authors sell more?
- Which actors are more sought after?



Our motto: *Ut uberiores divites.*

# The Barabási-Albert (BA) model

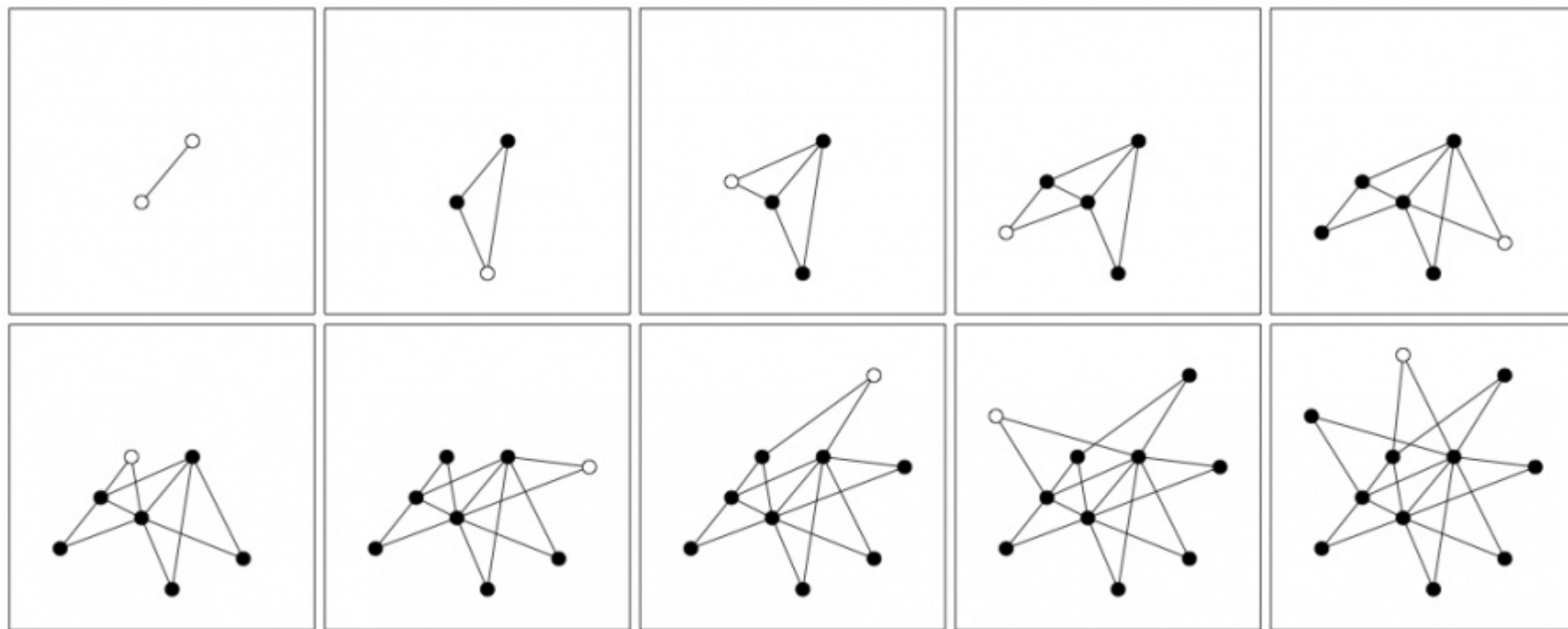
- Network starts with  $m_0$  nodes connected arbitrarily as long as their degree is  $\geq 1$
- At every time step we add 1 node
- This node will have  $m \leq m_0$  outlinks
- The probability of an existing node of degree  $k_i$  to gain one such link is

$$\Pi(k_i) = \frac{k_i}{\sum_{j=1}^{N-1} k_j}$$

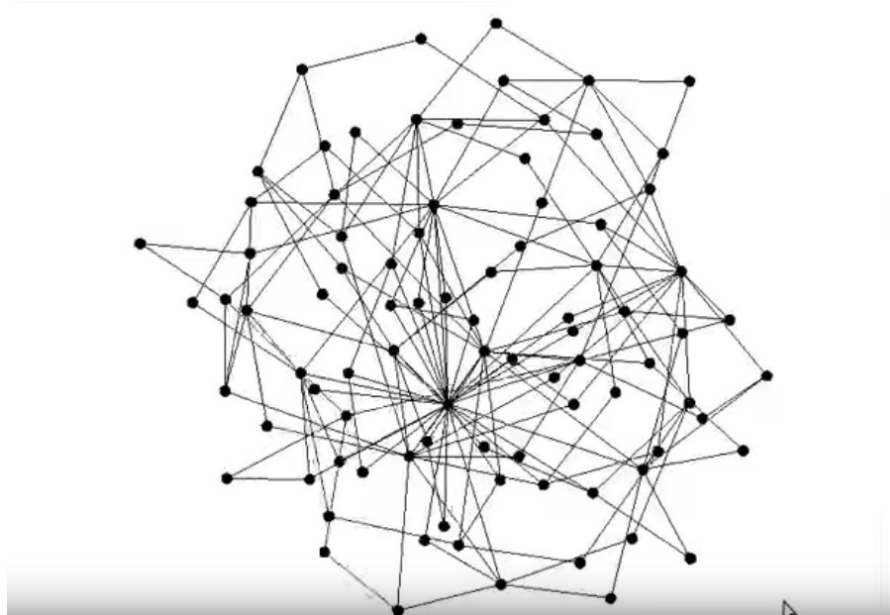
In an ER network,  $\Pi(k_i) = \frac{1}{N-1}$



# Example ( $m_0 = 2; m=2$ )



# Network growth with $m=2$



<https://www.youtube.com/watch?v=wocaGeNKn7Y>

# The Barabási-Albert (BA) model

- Network starts with  $m_0$  nodes connected arbitrarily as long as their degree is  $\geq 1$
- At every time step we add 1 node
- This node will have  $m$  outlinks ( $m \leq m_0$ )
- The probability of an existing node of degree  $k_i$  to gain one such link is 
$$\Pi(k_i) = \frac{k_i}{\sum_{j=1}^{N-1} k_j}$$

**Write the formula for  $N(t)$  and  $L(t)$ : at  $t=0$  the network has  $m_0$  nodes and  $L(0)$  links**

# Summary

# Things to remember

- Preferential attachment
- How to create a BA network step by step

# Practice on your own

- Describe step by step in pseudocode how to create a Barabási-Albert graph with  $N$  nodes having  $m_0$  starting nodes and  $m$  outlinks per node.
- For your pseudocode to be valid, if at any point there is a randomized step, you must indicate what is the probability of each possible outcome.

Advanced materials:  
Expected degree under  
uniform random attachment  
(not included in the exam)

# Expected degree in uniform random attachment using a differential equation

$$\frac{d}{dt}k_i(t) = \frac{m}{t}$$

Obtain  $k_i$

(1) Integrate between time  $i$  and time  $t$

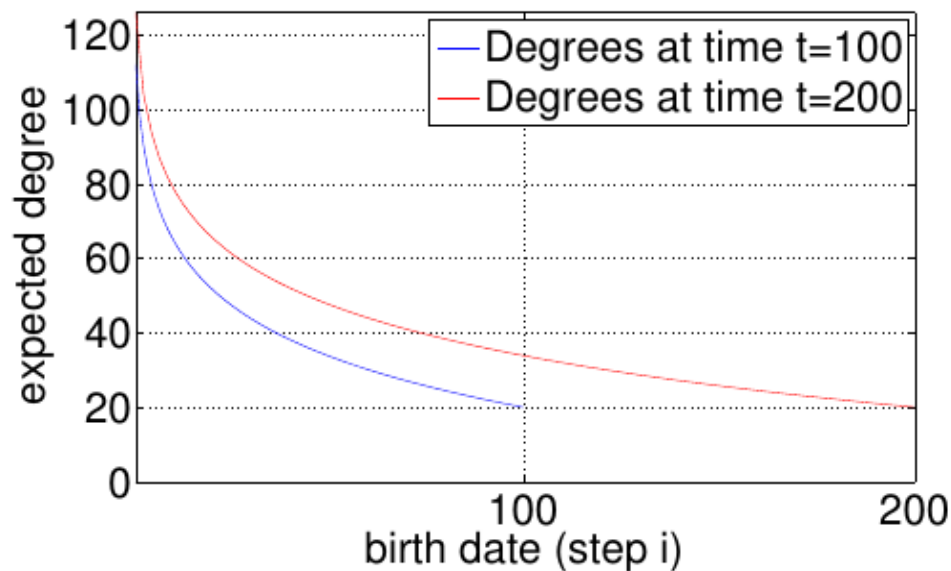
(2) Use initial condition  $k_i(i) = m$

$$\int \frac{1}{t} = \log t + C$$



# Degree distribution over time is not static

Degree of node born at time  $m < i < t = m \left( 1 + \log \left( \frac{t}{i} \right) \right)$



# Tail of degree distribution

How many nodes of degree larger than  $K$  are there at time  $t$ ?

The fraction is  $\frac{te^{-\frac{K-m}{m}}}{t} = e^{-\frac{K-m}{m}}$

**Decreases exponentially with  $K$ : it's vanishingly rare to find high-degree nodes**

$$m \left( 1 + \log \left( \frac{t}{i} \right) \right) > K$$

$$1 + \log \left( \frac{t}{i} \right) > \frac{K}{m}$$

$$\log \left( \frac{t}{i} \right) > \frac{K - m}{m}$$

$$\frac{t}{i} > e^{\frac{K-m}{m}}$$

$$i < te^{-\frac{K-m}{m}}$$