

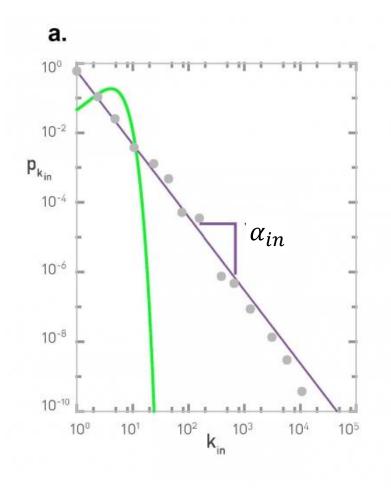
# Complex Network Systems

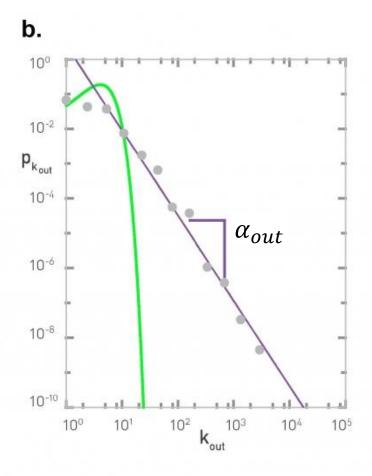
Scale-free networks

Ilche Georgievski

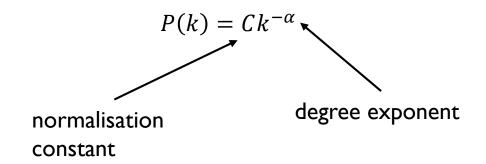
2019/2020 Winter

## Degree distribution of WWW



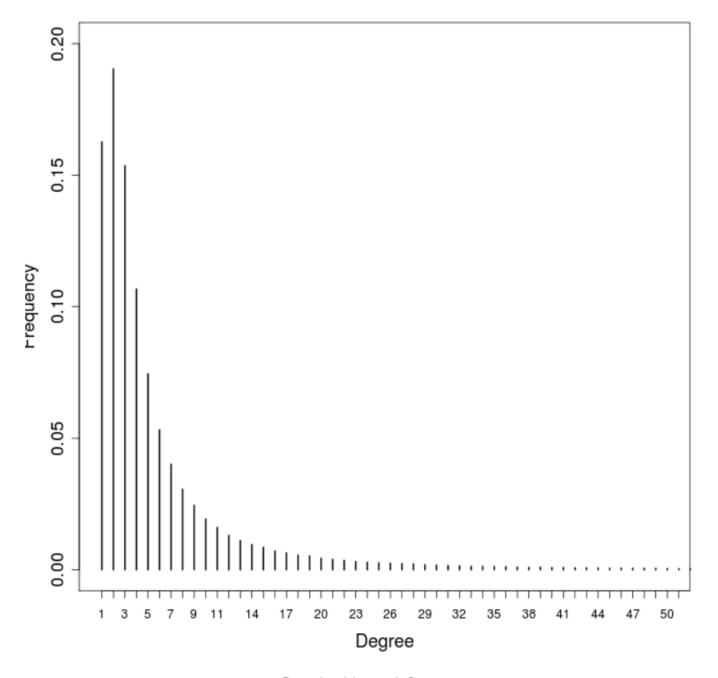


## Power-law degree distribution



## Power-law degree exponents

- Typical  $2 \le \alpha \le 3$ 
  - Web graph
    - $\alpha_{in} = 2.1$ ,  $\alpha_{out} = 2.4$  [Broder et al. 2000]
  - Autonomous systems
    - $\alpha = 2.4$  [Faloutsos 1999]
  - Actor collaborations
    - $\alpha = 2.3$  [Barabási-Albert 2000]
  - Citations to papers
    - $\alpha \approx 3$  [Redner 1998]
  - Online social networks
    - $\alpha \approx 2$  [Leskovec et al. 2007]



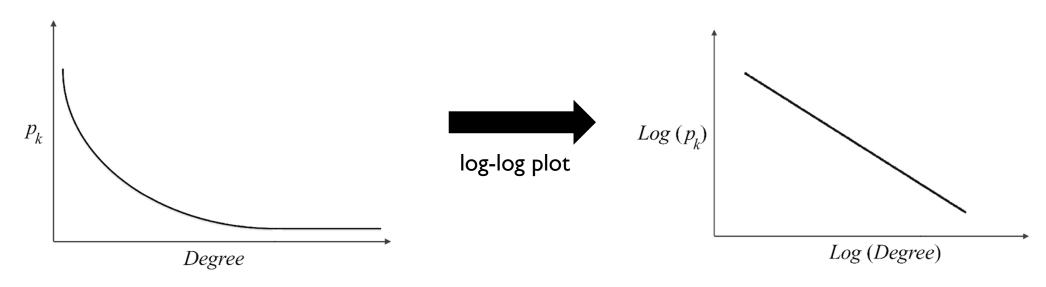
Tail contains 1098 highly collaborative scholars (0.16% of all authors)

Starting degree: III

$$\alpha = 4.4$$

$$p = 0.11$$

# Typical shape



(a) Power-Law Degree Distribution

(b) Log-Log Plot of Power-Law Degree Distribution

### Power-law distribution: Test

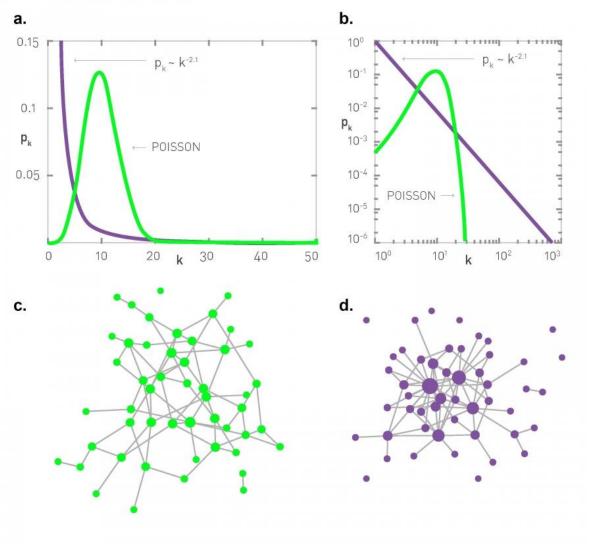
- Test whether a network follows a power-law distribution
  - I. Choose a popularity measure and compute it for the whole network
    - e.g., number of friends for all nodes
  - 2. Compute  $p_k$ 
    - Fraction of individuals having popularity k
  - 3. Plot a log-log graph, where the x-axis represents  $\log(k)$  and the y-axis represents  $\log(p_k)$
  - 4. Observe if there is a straight line. If yes, a power-law distribution exists

### Scale-free networks

A scale-free network is a network whose degree distribution follows a power law

### **HUBS**

### Poisson vs power-law distributions



## Poisson vs power-law distributions

Let us use the WWW to illustrate the properties of the high-k regime. The probability to have a node with k=100 is

- About  $P(100) \cong 10^{-30}$  in a Poisson distribution
- About if  $P(100) \cong 10^{-4}$  if P(k) follows a power law
- Consequently, if the WWW were to be a random network, according to the Poisson prediction, we would expect  $N_{k>100}\cong 10^{-18}$ , or none.
- For a power-law degree distribution, we expect about  $N_{k>100}=10^9$

### Random vs scale-free networks



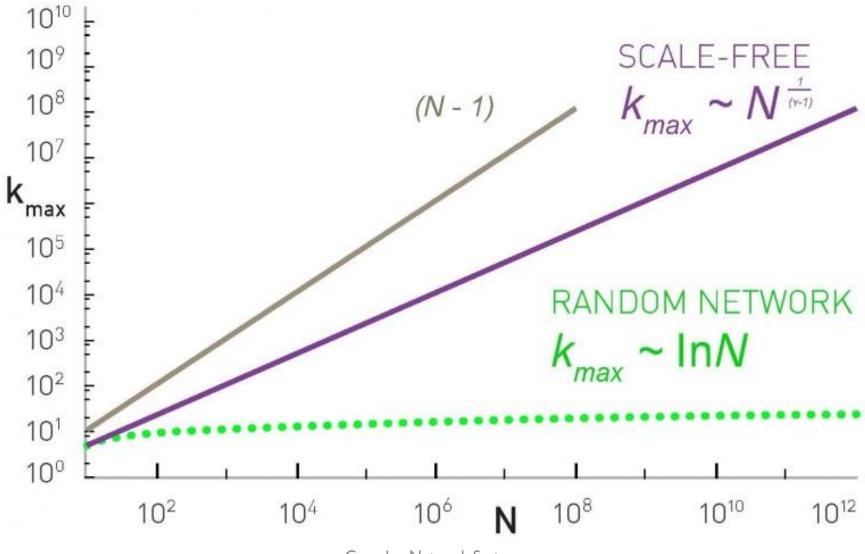
## Hubs are large in scale-free networks

Expected maximum degree,  $k_{max}$ 

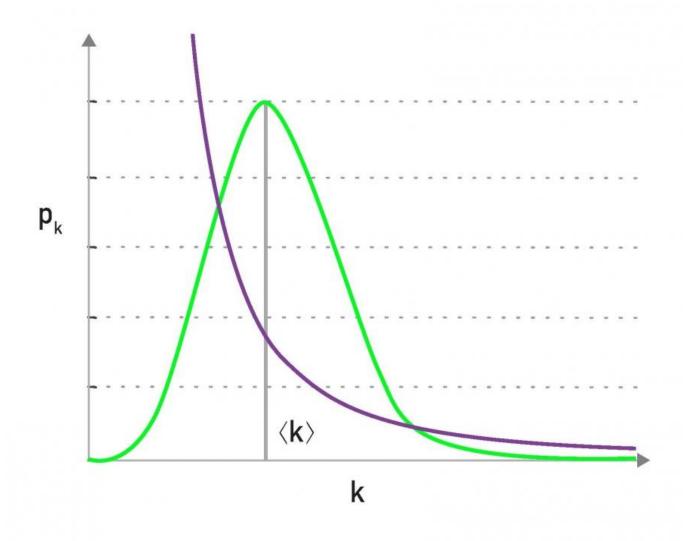
$$k_{max} = k_{min} N^{\frac{1}{\alpha - 1}}$$

- $k_{max}$  increases with the size of the network
  - — →the larger a system is, the larger its biggest hub
- For  $\alpha > 2$ ,  $k_{max}$  increases slower than N
  - → the largest hub will contain a decreasing fraction of links as *N* increases
- For  $\alpha = 2$ ,  $k_{max} \sim N$ .
  - $\rightarrow$  The size of the biggest hub is O(N)
- For  $\alpha < 2$ ,  $k_{max}$  increases faster than N: condensation phenomena
  - → the largest hub will grab an increasing fraction of links. Anomaly!

### Hubs are large in scale-free networks



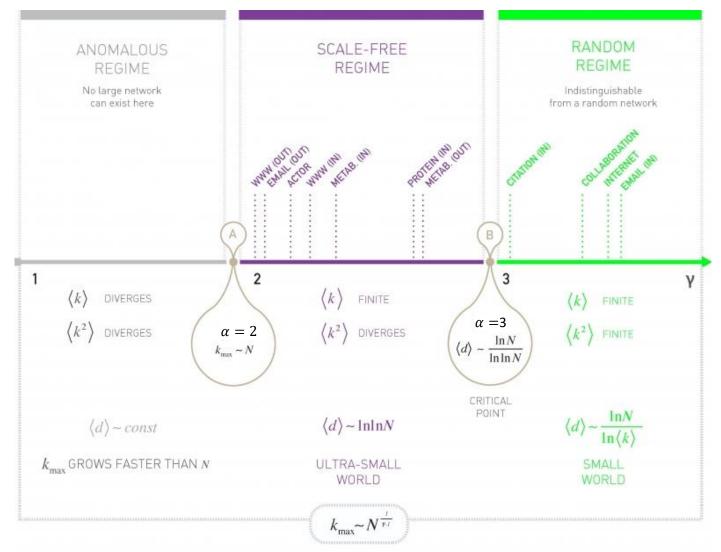
# Hubs are large in scale-free networks



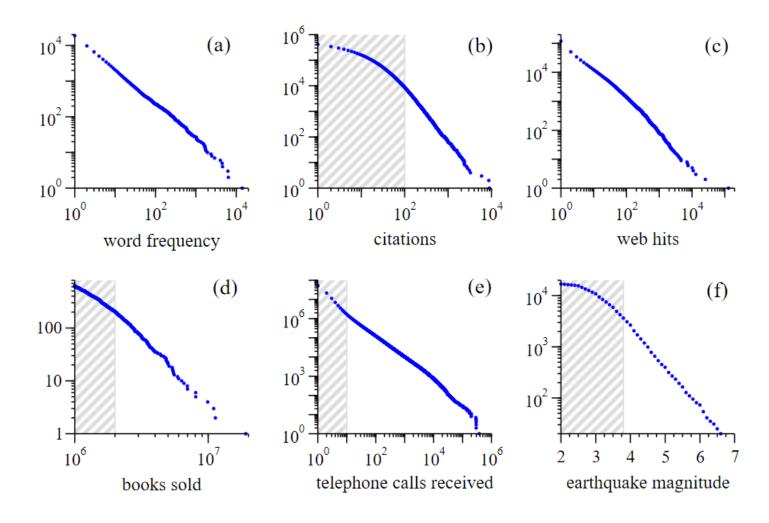
## Ultra-small property

- Shrinks the average path lengths. Therefore most scale-free networks of practical interest are not only "small" but are "ultra-small". This is a consequence of the hubs, that act as bridges between many small degree nodes.
- Changes the dependence of  $\bar{d}$  on the system size. The smaller is  $\alpha$ , the shorter are the distances between the nodes.
- Only for  $\alpha > 3$  we recover the  $\ln N$  dependence, the signature of the small-world property characterising random networks.

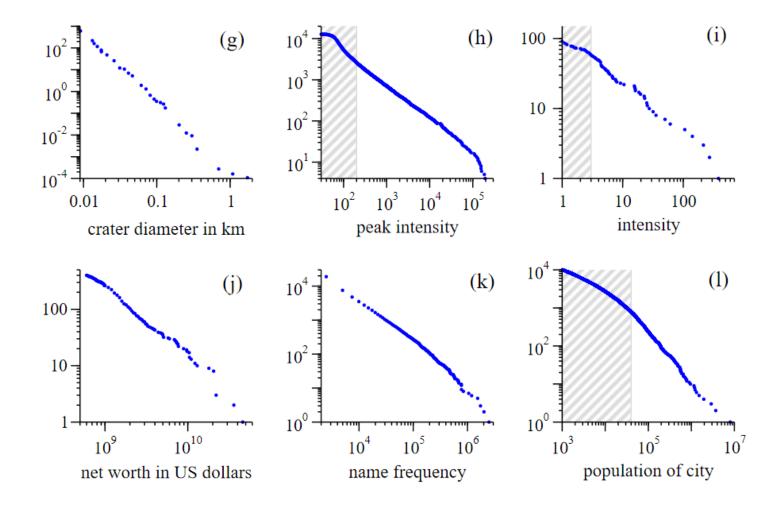
## Scale-free networks are degree dependent



### Examples: Power-law distribution



### More examples: Power-law distribution



### Sources

- Leskovec, J. Analysis of Networks, CS224W, Stanford University (2018), http://web.stanford.edu/class/cs224w/
- Mateos, G. Degrees, Power Laws and Popularity, University of Rochester, 2018.
- Zafarani, R., Abbasi, M.A. and Liu, H. Social Media Mining: An Introduction, Cambridge University Press, 2014.
- Barabási, A. Network Science, http://networksciencebook.com