

# Distances in Scale-Free Networks

Introduction to Network Science

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Topic 10

# Contents

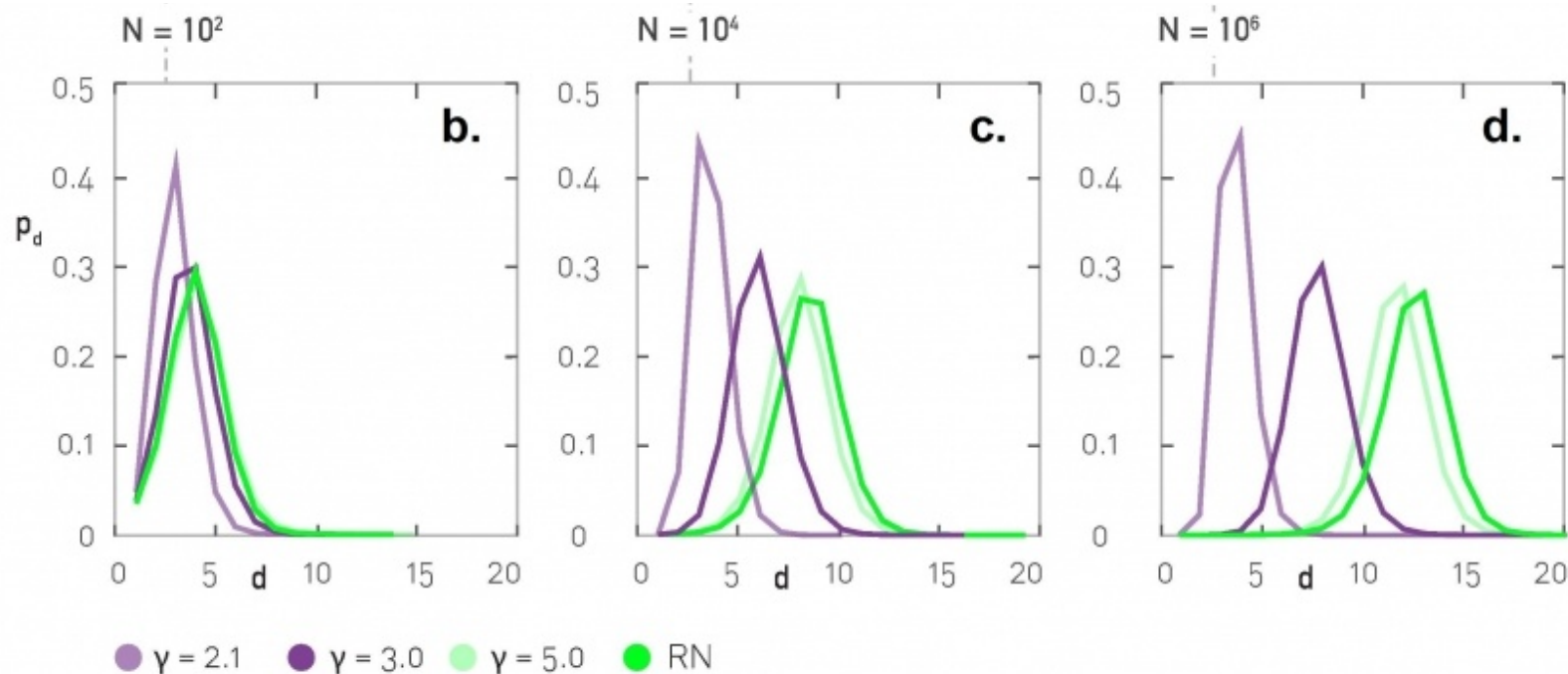
- Distance distribution of scale-free networks

# Sources

- Albert-László Barabási: Network Science. Cambridge University Press, 2016.
  - Follows almost section-by-section chapter 04
- URLs cited in the footer of specific slides

# Distance distributions: simulation results

Scale-free networks of increasing size,  $\langle k \rangle = 3$



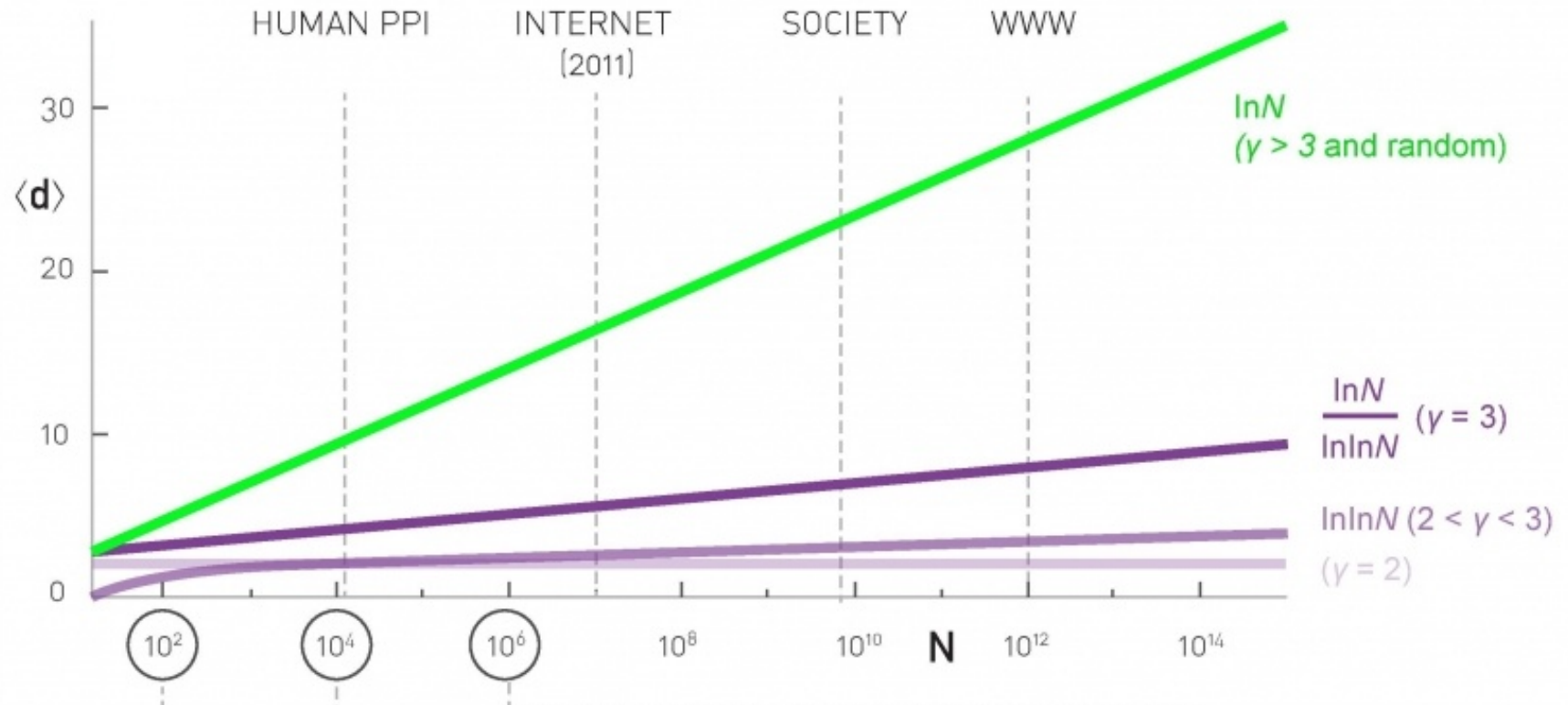
# Average distance

- Depends on  $\gamma$  and  $N$

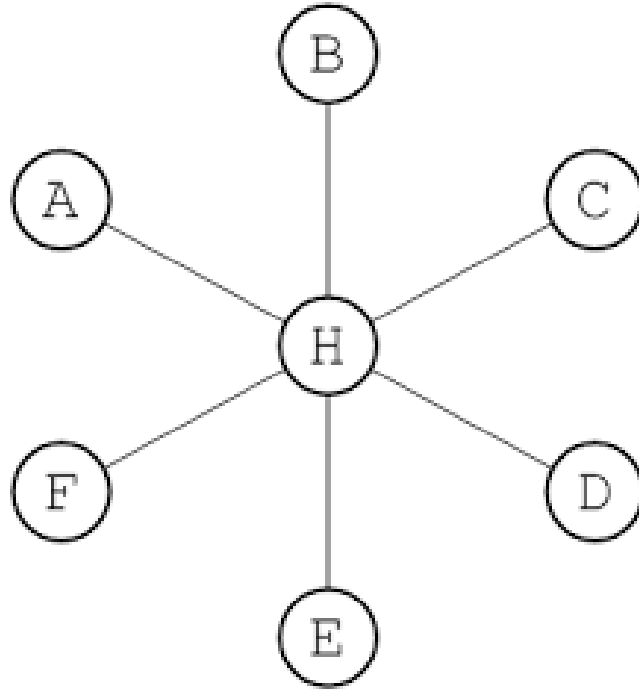
$$\langle d \rangle = \begin{cases} \text{const.} & \text{if } \gamma = 2 \\ \log \log N & \text{if } 2 < \gamma < 3 \\ \log N / \log \log N & \text{if } \gamma = 3 \\ \log N & \text{if } \gamma > 3 \end{cases}$$

← Same as in  
ER graphs

# Average distance and N



# Anomalous regime $\gamma = 2$



# Ultra-small world $2 < \gamma < 3$

- Average distance follows  $\log(\log(N))$
- Example (humans):

$$N \approx 7 \times 10^9$$

$$\log N \approx 22.66$$

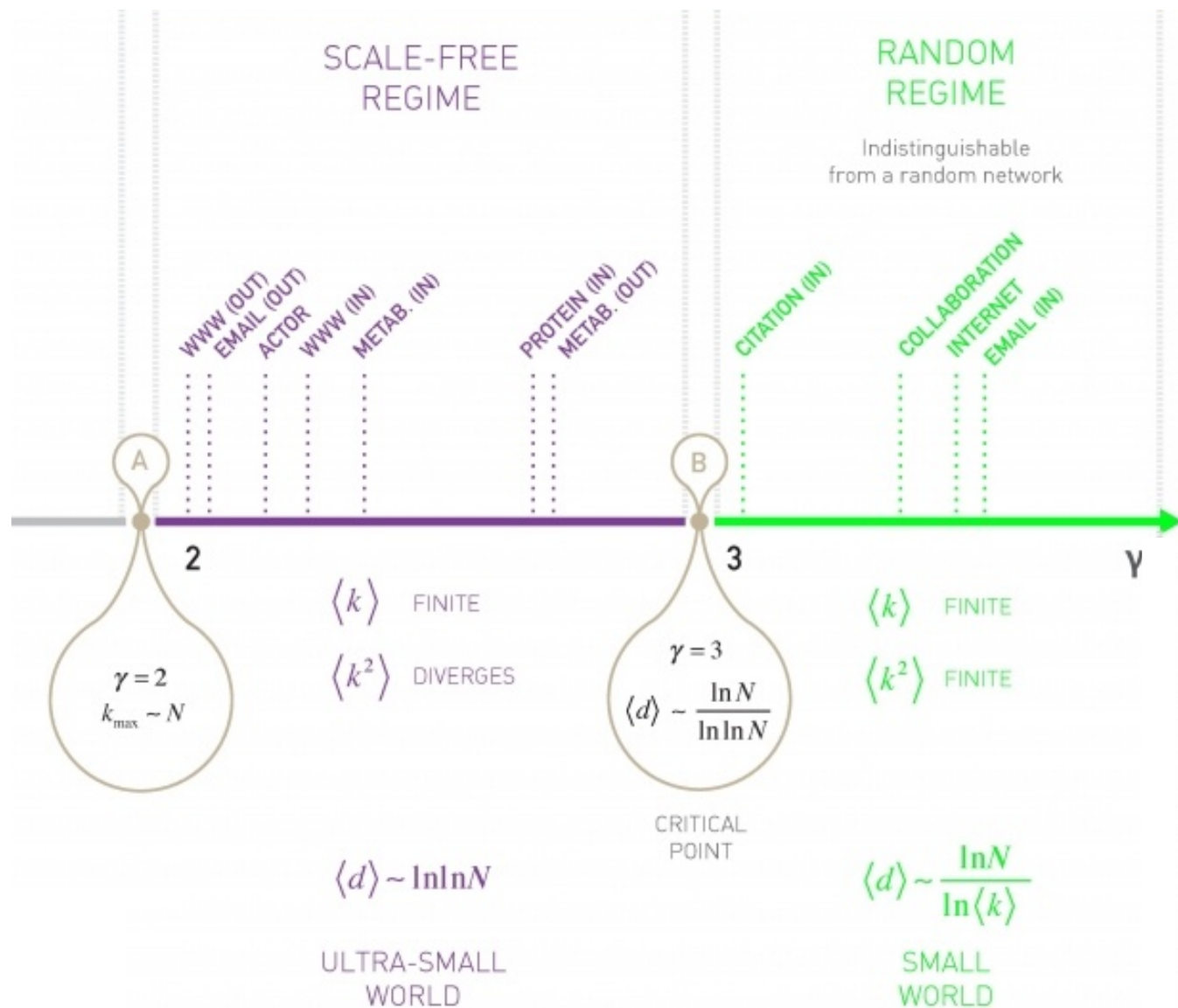
$$\log \log N \approx 3.12$$



# Small world $\gamma > 3$

- Average distance follows  $\log(N)$
- Similar to ER graphs where it followed  $\log(N)/\log(\langle k \rangle)$

The degree distribution exponent plays an important role



# When $\gamma > 3$

- In this case it is hard to distinguish this case from an ER graph
- In most real complex networks (but not all)

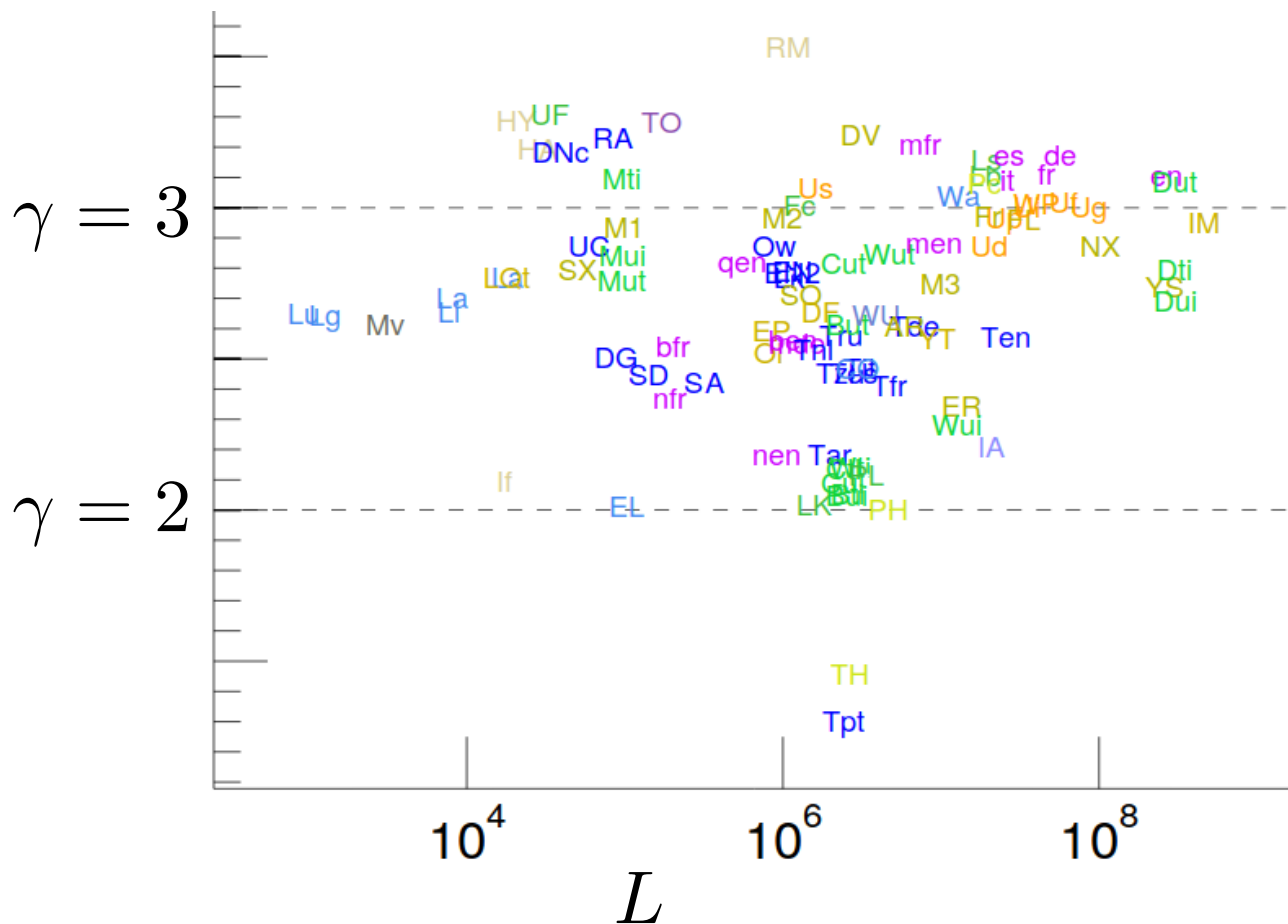
$$2 < \gamma < 3$$

# When $\gamma > 3$

- Remember  $k_{\max} = k_{\min} N^{\frac{1}{\gamma-1}}$        $N = \left( \frac{k_{\max}}{k_{\min}} \right)^{\gamma-1}$
- Observing the scale-free properties requires that  $k_{\max} \gg k_{\min}$ , e.g.  $k_{\max} = 10 k_{\min}$
- Then if  $\gamma = 5$ ,  $N > 10^8$
- Hence we won't find many such networks

# Examples

<http://konect.uni-koblenz.de/statistics/prefatt>



<b>EL</b>	Wikipedia elections
<b>LK</b>	Linux kernel mailing list threads
<b>Bul</b>	BibSonomy u-i
<b>Bti</b>	BibSonomy t-i
<b>Cul</b>	CiteULike u-i
<b>If</b>	Infectious
<b>PL</b>	Prosper loans
<b>Cti</b>	CiteULike t-i
<b>Wti</b>	Twitter t-i
<b>nen</b>	Wikinews (en)
<b>Tar</b>	Wikipedia talk, Arabic
<b>Wul</b>	Twitter u-i
<b>ER</b>	Epinions
<b>nfr</b>	Wikinews (fr)
<b>Tfr</b>	Wikipedia talk, French
<b>SD</b>	Slashdot
<b>Tzh</b>	Wikipedia talk, Chinese
<b>Tes</b>	Wikipedia talk, Spanish

Etc.

# The friendship paradox

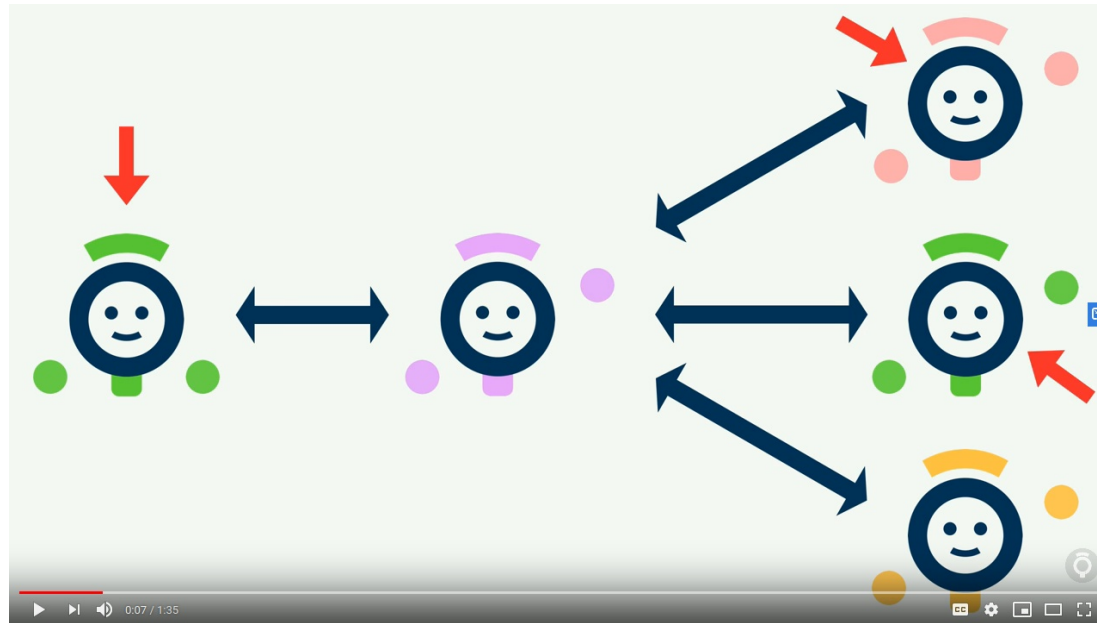
# The friendship paradox

- Take a random person  $x$ , what is the expected degree of this person?  $\langle k \rangle$
- Take a random person  $x$ , now pick one of  $x$ 's neighbors, let's say  $y$

What is the expected degree of  $y$ ?

**It is not  $\langle k \rangle$**

# Sampling bias and the friendship paradox (1'35'')



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



# Imagine you're at a random airport on earth

- Is it more likely to be ...  
a large airport or a small airport?
- If you take a random flight out of it ...  
will it go to a large airport or a small airport?

# An example of friendship paradox

- Pick a random airport on Earth
  - Most likely it will be a small airport
- However, no matter how small it is, it **will** have flights to big airports
- On average those airports will have much larger degree



Time	Flight	Airline	Destination	Gate	Exp.	Remarks
11:00	KA 376	DRAGONAIR	Hong Kong	4		Chk-in closed
12:25	DG 7792	tigerair	Singapore	1		On Time
12:25	QR 931	QATAR	Doha, Qatar	5		On Time
17:40	EK 339	Emirates	Dubai	5		On Time
00:50	OZ 708	ASIANA AIRLINES	Seoul Incheon	5		On Time
07:05	5J 150	ANA	Hong Kong	1		On Time
07:20	DG 7924	tigerair	Hong Kong	1		On Time
08:00	DG 7792	tigerair	Singapore	1		On Time
12:10	5J 537	ANA	Singapore	1		On Time
12:25	QR 931	QATAR	Doha, Qatar	5		On Time

# Exercise [B. 2016, Ex. 4.10.2]

## "Friendship Paradox"

- Remember  $p_k$  is the probability that a node has  $k$  "friends"
- If we randomly select a link, the probability that a node at any end of the link has  $k$  friends is  $q_k = C k p_k$  where  $C$  is a normalization factor
  - (a) Find  $C$  (the sum of  $q_k$  must be 1)

Answer in Nearpod Collaborate  
<https://nearpod.com/student/>  
Access to be provided during class

# Exercise [B. 2016, Ex. 4.10.2]

## "Friendship Paradox"

- If we randomly select a link, the probability that a node at any end of the link has  $k$  friends is  $q_k = C k p_k$  where  $C$  is a normalization factor
- $q_k$  is also the prob. that a randomly chosen node has a neighbor of degree  $k$

(b) Find its expectation  $E[q_k]$  which we will call  $\langle k_F \rangle$

Remember  $E[X] = \sum_{X_{\min}}^{X_{\max}} x \cdot P(X = x)$

# Exercise [B. 2016, Ex. 4.10.2]

## "Friendship Paradox"

(c) Compute the expected number of friends of a neighbor of a randomly chosen node in the case below

(d) compare with the expected number of friends of a randomly chosen node

$$N = 10000$$

$$\gamma = 2.3$$

$$k_{\min} = 1$$

$$k_{\max} = 1000$$

$$\langle k^n \rangle = C \frac{k_{\max}^{n-\gamma+1} - k_{\min}^{n-\gamma+1}}{n - \gamma + 1}$$

$$C = (\gamma - 1) k_{\min}^{\gamma-1}$$

# Code

```
def degree_moment(kmin, kmax, moment, gamma):  
    C = (gamma-1.0)*(kmin**(gamma-1.0))  
    numerator = (kmax**(moment-gamma+1.0) - kmin**(moment-gamma+1.0))  
    denominator = (moment-gamma+1.0)  
    return C * numerator / denominator
```

```
kavg = degree_moment(kmin=1, kmax=1000, moment=1, gamma=2.3)  
print(kavg)
```

3.787798988222529

```
ksqavg = degree_moment(kmin=1, kmax=1000, moment=2, gamma=2.3)  
print(ksqavg)
```

231.94329076177414

```
print(ksqavg / kavg)
```

61.23431879119234

# Summary

# Things to remember

- Regimes of distance and connectivity
- The friendship paradox



# Practice on your own

- Remember the regimes of a graph given  $\langle k \rangle$   
(It's useful to know this by heart)
- Estimate degree distributions and distance distributions for some graphs
- Draw a small graph, and sample from that graph until you're convinced  $\langle k_F \rangle > \langle k \rangle$