

# Network Algorithms

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Some slides adapted from:

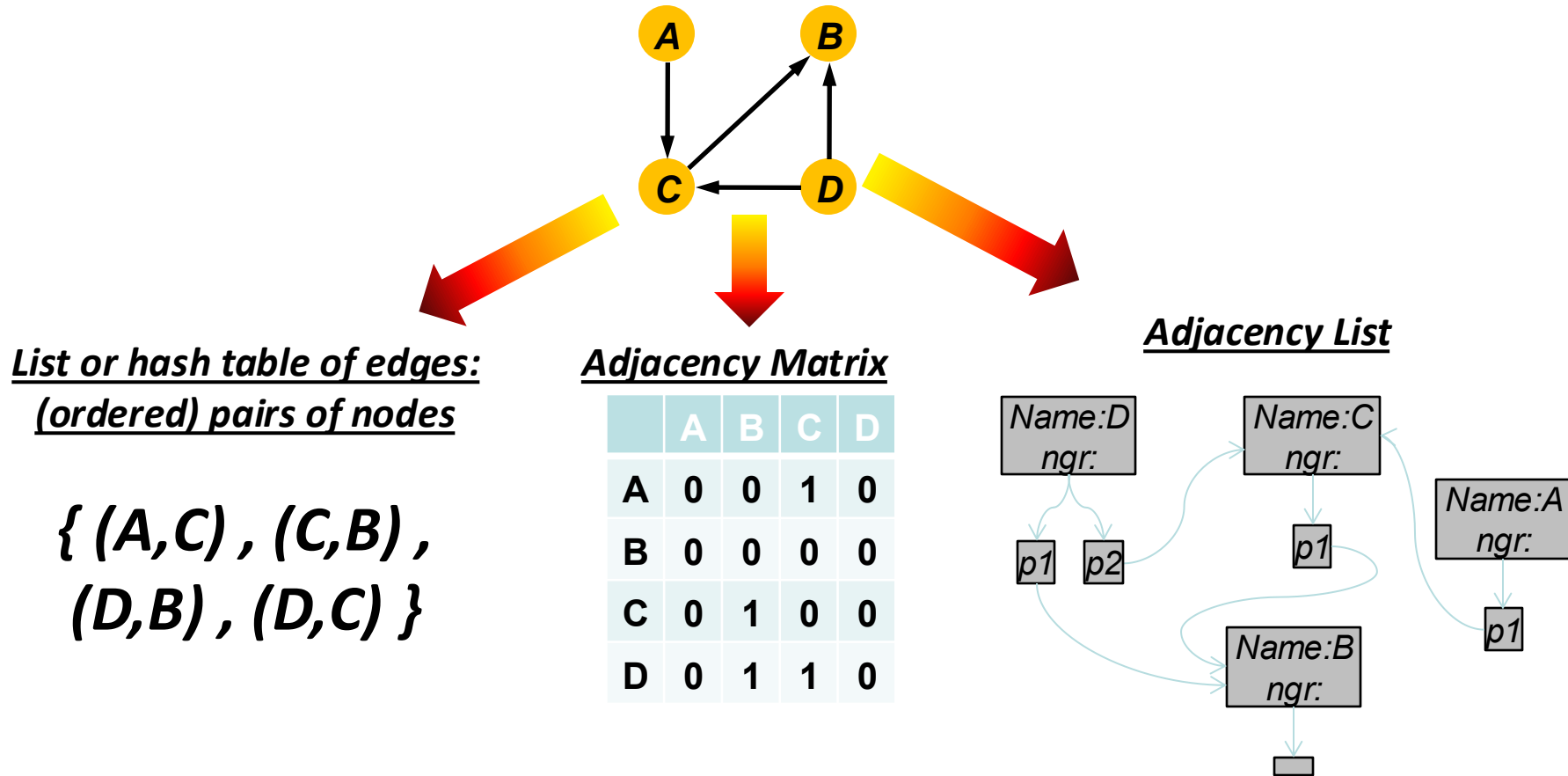
Networked Life (NETS) 112, Univ. of Penn., 2018, Prof. Michael Kearns

Determining the Diameter of Small World Networks, Frank W. Takes & Walter A. Kusters, Leiden University, The Netherlands

Structure and models of real-world graphs and networks, Jure Leskovec, Carnegie Mellon University

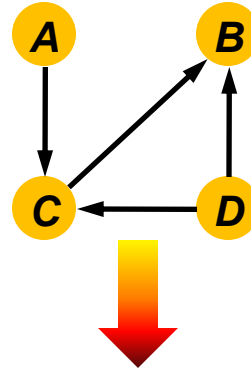
Complex (Biological) Networks, by Elhanan Borenstein, Roded Sharan, and Tomer Shlomi

# Traditional Computational Representations of Networks



- Which is the most useful representation?
- Should you use them in combination?

# A Unified Computational Representation of Networks



Adjacency Set

A: Adjacencies: {C}, Edges: {(A,C)}

B: Adjacencies: {}, Edges: {}

C: Adjacencies: {B}, Edges: {(C,B)}

D: Adjacencies: {B,C}, Edges: {(D,B), (D,C)}

- Time for `listAdjacent(v)` is  $O(\text{degree}(v))$
- Time for `areAdjacent(v,w)` is  $O(1)$  if sets have hash tables, like in Python

# Network Structures

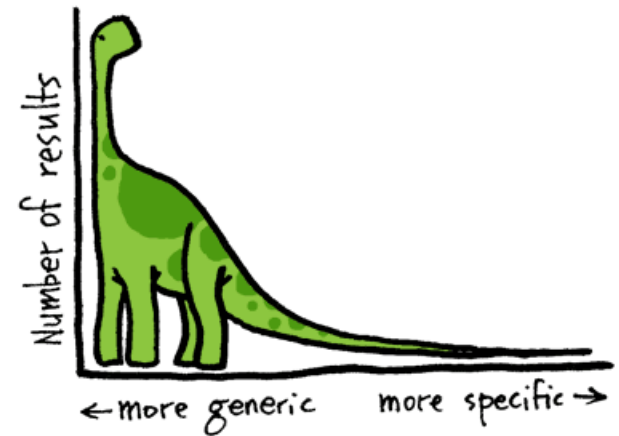
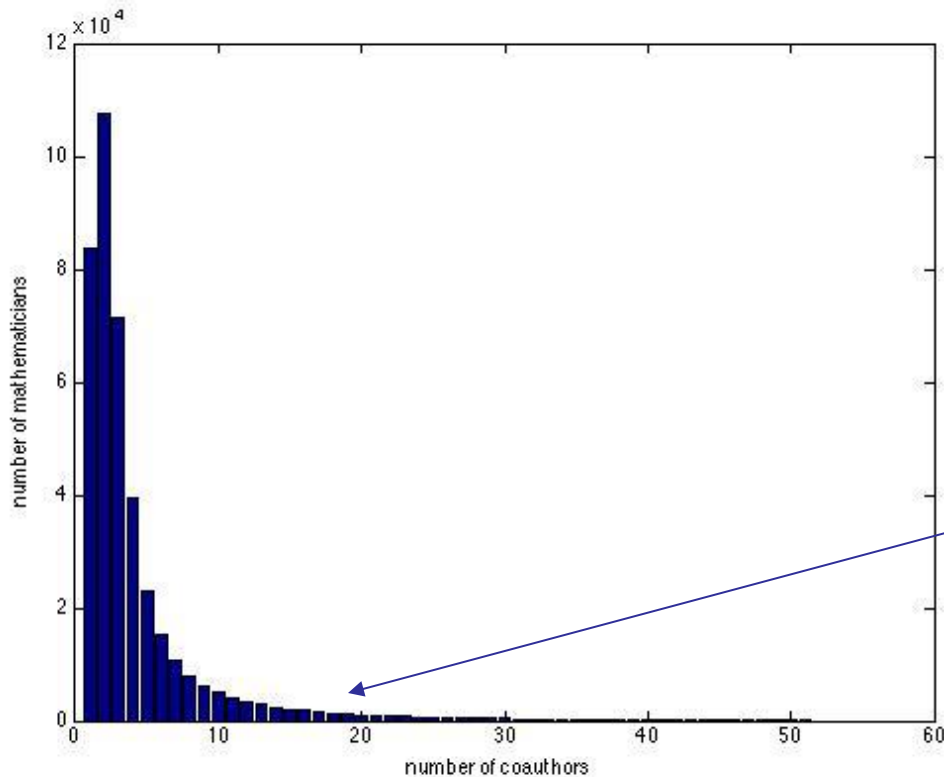
- Network structures characterize how networks “look”:
  - Large or small diameter?
  - Number of edges: sparse or dense?
  - Degree distributions: heavy/long tail with a power law?
  - Clustering coefficient: high or low?
- These are empirical phenomena
- How do you compute them?



Image from <https://matrix.berkeley.edu/research/social-networks-history>

# Degree Distribution

- x axis: number of neighbors (degree)
- y axis: number of vertices with that degree



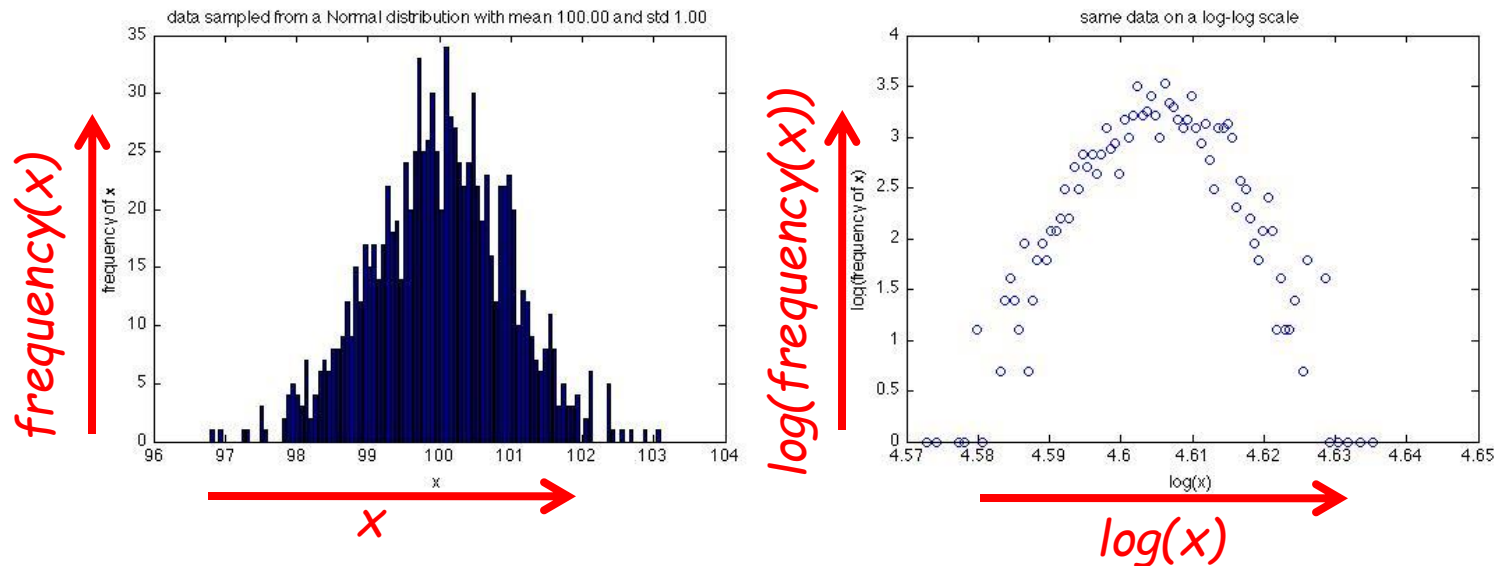
A long tail  
(also known as a "heavy tail")

# Degree Distribution Algorithm

1. Compute the degree,  $\deg(v)$ , of each vertex,  $v$ .
  - If  $G$  is represented as an adjacency list, count the number of elements in  $v$ 's list.
2. Create a histogram count array,  $H$ , of size  $n$ , and initialize each  $H[i] = 0$ .
3. For each vertex,  $v$ , increment  $H[\deg(v)]$ .
4. Plot the values of  $H$  from 0 to  $n-1$  on a regular and log-log scale
5. If the values on the log-log plot form a straight line, determine its slope to find the exponent of the power law degree distribution

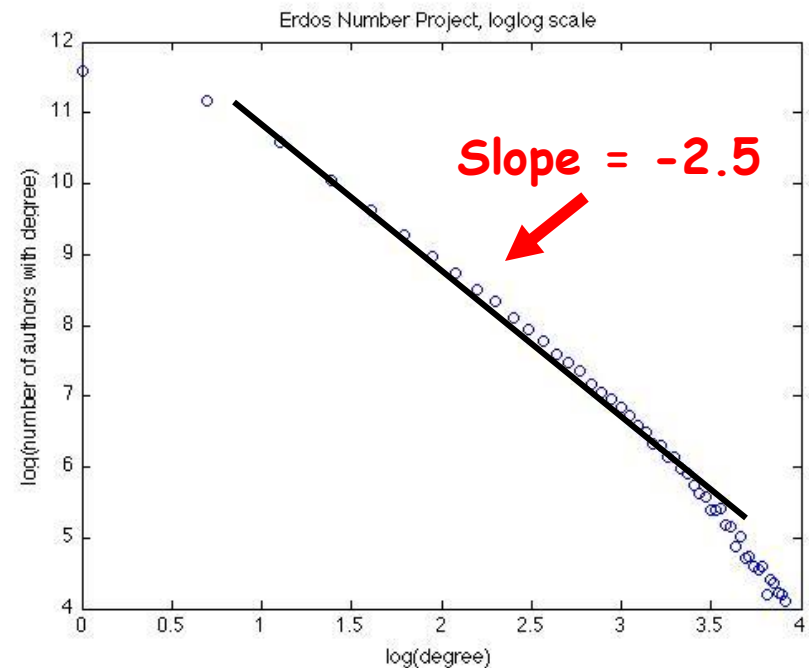
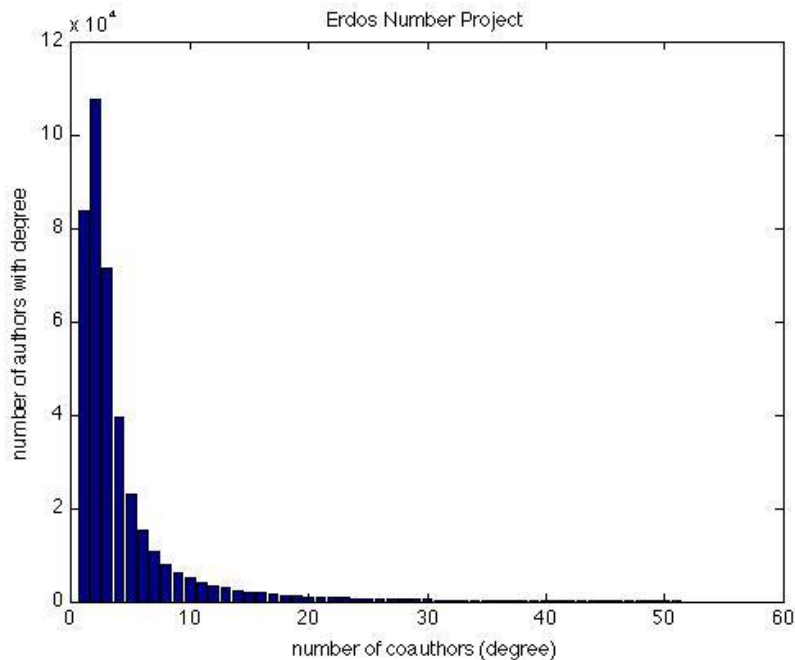
# Example 1

- Degree distribution without a long/heavy tail .
- Does not exhibit a power law.



# Example 2

- Degree distribution with a long/heavy tail .
- Does exhibit a power law, with exponent -2.5.





# Distance

- The **distance** between two vertices is the length of the shortest path connecting them.
  - This assumes the network has only a single connected component
  - If two vertices are in different components, their distance is infinite

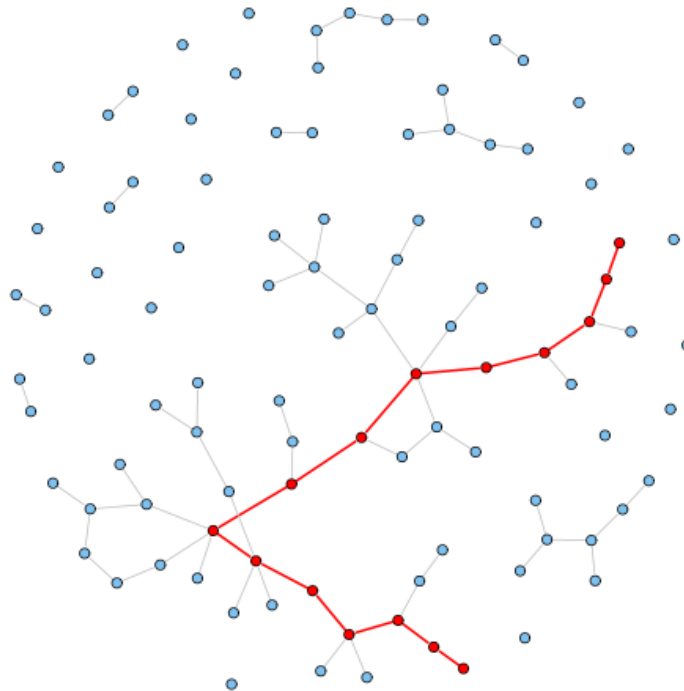
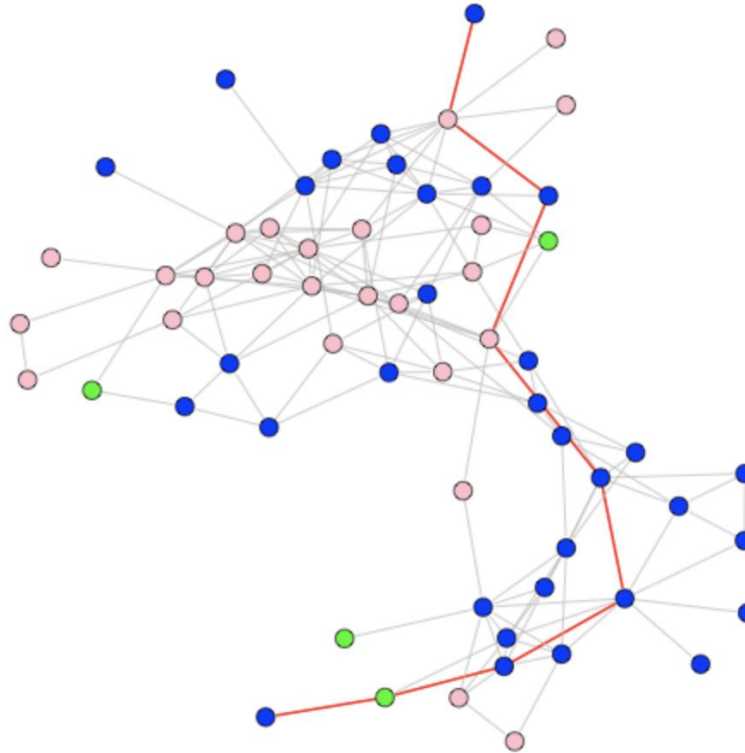


Image from

<https://www.sci.unich.it/~francesco/teaching/network/geodesic.html>

# Diameter

- The **diameter** of a network is the maximum distance between a pair of vertices in the network
  - It measures how near or far typical individuals are from each other



*The dolphin network with the **diameter** (the longest shortest path) highlighted in red. The diameter is 8 edges long.*

*[From https://users.dimi.uniud.it/~massimo.franceschet/bottlenose/bottlenose.html](https://users.dimi.uniud.it/~massimo.franceschet/bottlenose/bottlenose.html)*

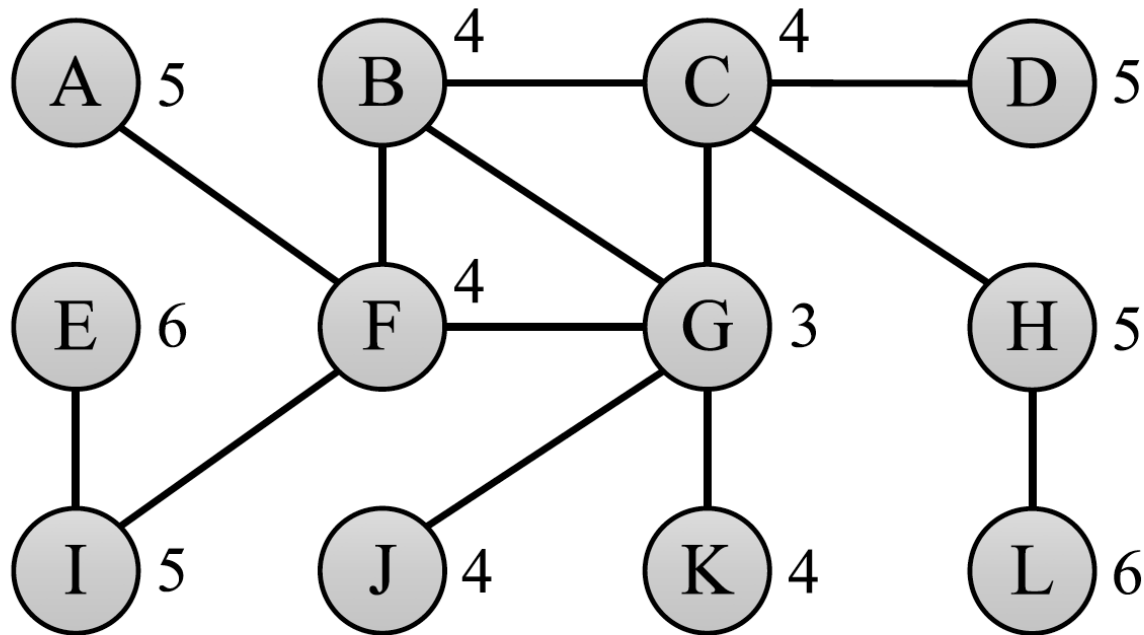
# Definitions

- Consider a connected undirected graph  $G = (V, E)$  with  $n = |V|$  nodes and  $m = |E|$  edges
- **Distance**  $d(v, w)$ : length of shortest path between nodes  $v, w \in V$
- **Diameter**  $D(G)$ : maximal distance (longest shortest path length) over all node pairs:  $\max_{v, w \in V} d(v, w)$
- **Eccentricity**  $e(v)$ : length of a longest shortest path from  $v$ :  $e(v) = \max_{w \in V} d(v, w)$
- **Diameter**  $D(G)$  (alternative definition): maximal eccentricity over all nodes:  $\max_{v \in V} e(v)$
- Eccentricity distribution: (relative) frequency  $f(x)$  of each eccentricity value  $x$

$$f(x) = \frac{|\{u \in V \mid e(u) = x\}|}{n}$$

# Example

- A graph with diameter 6
- Numbers next to nodes denote eccentricity values

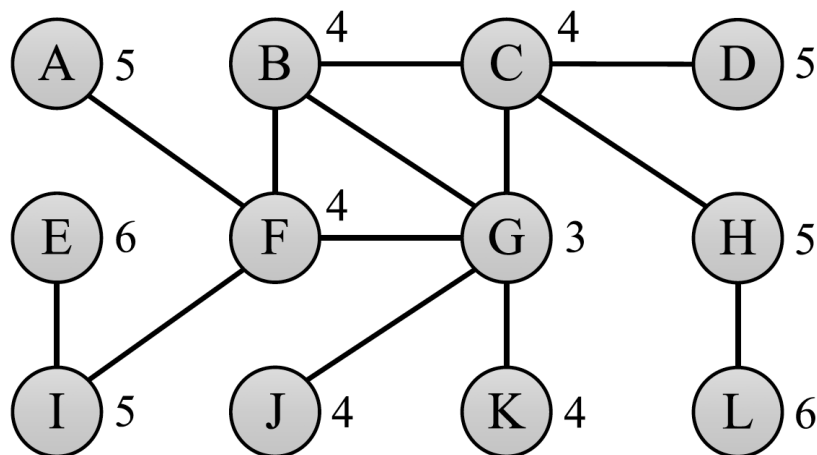


# Naïve Algorithm

- **Diameter** is equal to the largest value returned by an All Pairs Shortest Path (APSP) algorithm
- Brute-force: for each vertex  $v$ , execute a Breadth First Search (BFS) from  $v$  in  $O(m)$  time to find  $v$ 's **eccentricity**. Return the largest value found.
- Time complexity  $O(nm)$
- Problematic if  $n = 8$  million and  $m = 1$  billion.
  - If one BFS takes 6 seconds on a 3.4GHz machine, this brute-force algorithm takes 1.5 years to compute the diameter . . .

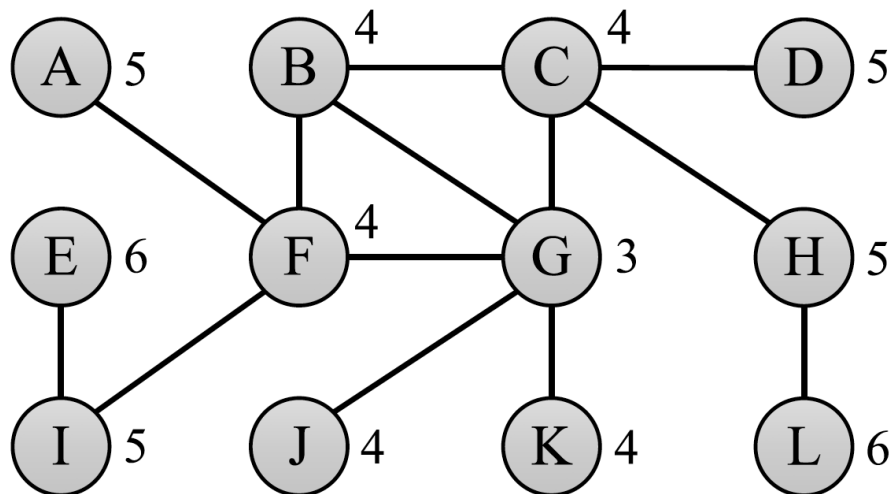
# Heuristic Idea 1

- If we can find one of the nodes in a diameter pair, we can compute the diameter with one more BFS.
1. Perform a BFS from a random sample of nodes, recording nodes with maximum found distance,  $d$ .
  2. Perform a BFS from all the far nodes (if small) or a random sample of this set (if large).



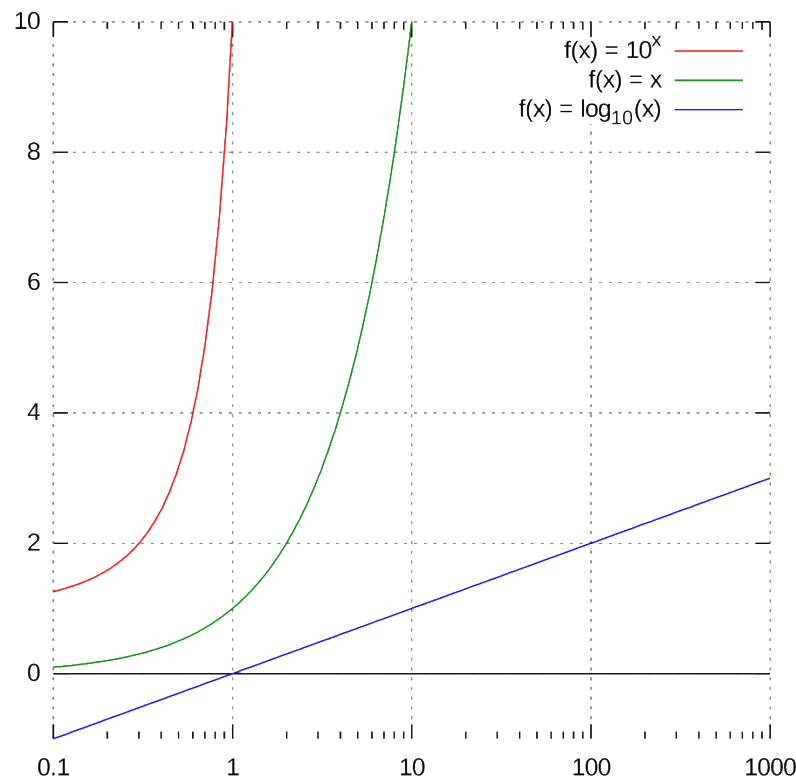
# Heuristic Idea 2

1. Let  $r$  be a random vertex and set  $D_{\max} = 0$ .
  2. Perform a BFS from  $r$ .
  3. Select the farthest node,  $w$ , in this BFS.
- If the distance from  $r$  to  $w$  is larger than  $D_{\max}$ , set  $D_{\max}$  to this distance, let  $r = w$ , and repeat the above two steps.



# Plot Results as a Function of n

- If the networks exhibit the **small world** phenomenon, then diameters are small.
- So plot diameters as a function of n on a lin-log scale:



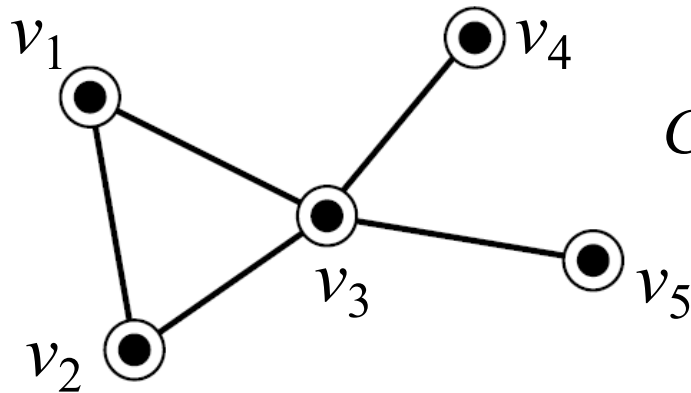
The log n function  
looks like a straight line



# Clustering Coefficient

- “friend of a friend is a friend”
- If  $a$  connects to  $b$ , and  $b$  to  $c$ , then with high probability  $a$  connects to  $c$ .
- Clustering coefficient  $C$ :

$C = 3 * \text{number of triangles} / \text{number of 2-edge paths}$



$$C = 3 * (1) / (1 + 1 + 6 + 0 + 0) = 3/8 = 0.375$$

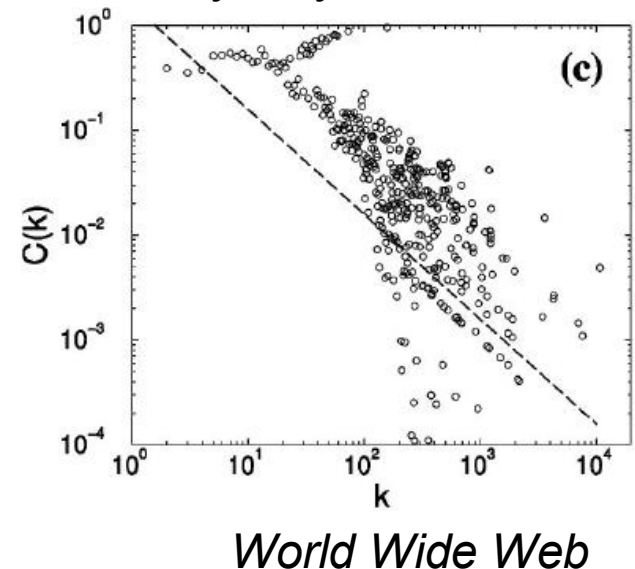
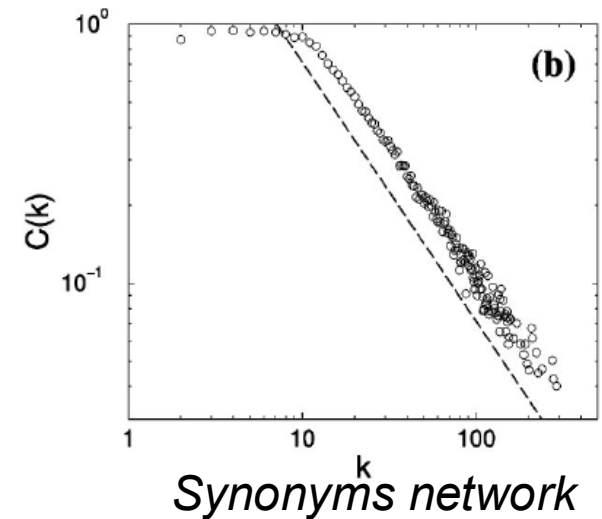
# Clustering Coefficient (2)

- Clustering coefficient might have a power law:

$$C(k) \sim k^{-1}$$

- It is speculated that in real networks:

$$C=O(1) \text{ as } n \rightarrow \infty$$

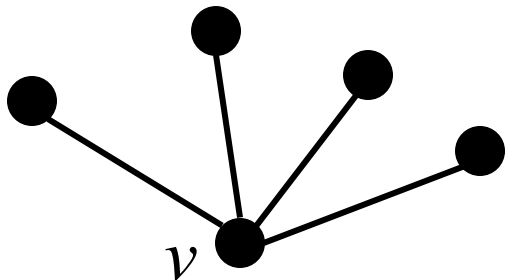


# Clustering Coefficient Algorithm

- Clustering coefficient  $C$ :

$$C = 3 * \text{number of triangles} / \text{number of 2-edge paths}$$

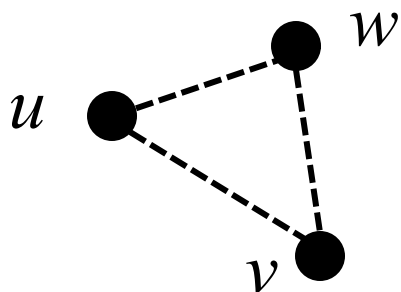
- Computing the **denominator** is **easy**:
  - For each vertex  $v$ , let  $\deg(v)$  denote its degree.
  - The number of paths of length 2 with  $v$  in the middle is  $\deg(v) \text{ choose } 2 = \deg(v)(\deg(v)-1)/2$ .
  - So, to get the denominator for  $C$ , sum up  $\deg(v)(\deg(v)-1)/2$  for all vertices,  $v$ , in  $G$ .



Number of 2-edge paths with  $v$  in the middle is  $4(3)/2 = 6$ .

# Counting Triangles

- To get the numerator for  $C$ , we need to count the number of triangles in the graph,  $G$ .
- Naïve algorithm:
  - For every triple,  $u, v, w$  in  $G$ , see if they form a triangle. If so, add 1 to a running count.
  - Running time is  $O(n^4)$  if  $G$  is represented with an adjacency list.

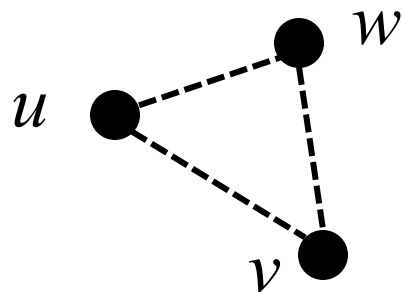


This is bad.



# Counting Triangles: Slight Improvement

- Put every edge,  $(v,w)$ , into a hash table,  $T$ , so we can do a lookup to see if an edge exists in  $O(1)$  expected time, i.e., with a  $\text{get}((v,w))$ .
- Slightly better naïve algorithm:
  - For every triple,  $u, v, w$  in  $G$ , see if they form a triangle. If so, add 1 to a running count.
  - Running time is  $O(n^3)$  expected if edges in  $G$  are stored in a hash table.



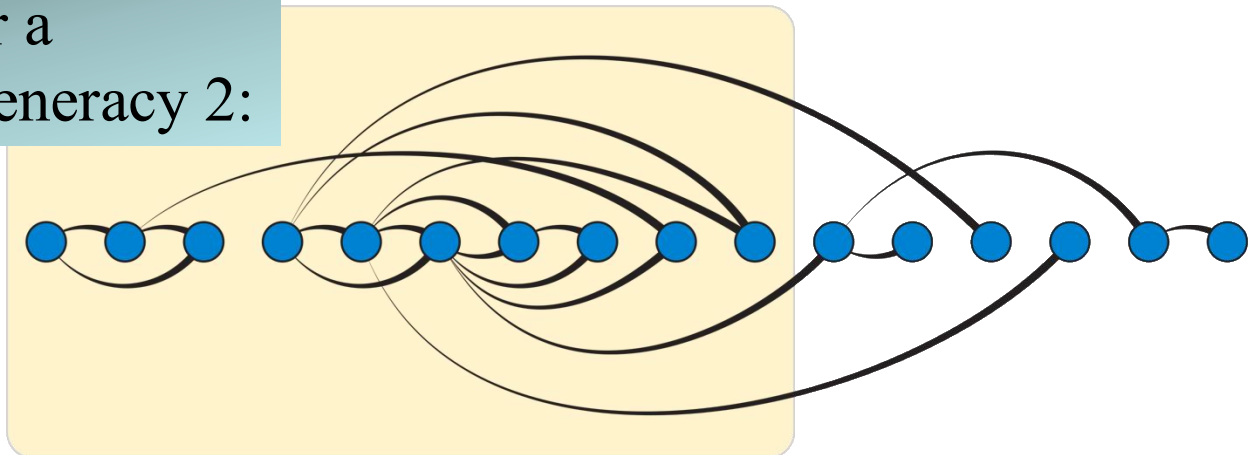
This is still bad.



# Graph Degeneracy

- The **degeneracy** of a graph is the smallest value of  $d$  for which every subgraph has a vertex of degree at most  $d$ .
- If a graph has degeneracy  $d$ , then there exists an ordering of the vertices of  $G$  in which each vertex has at most  $d$  neighbors that are earlier in the ordering.

An ordering for a graph with degeneracy 2:



# Real-World Graphs

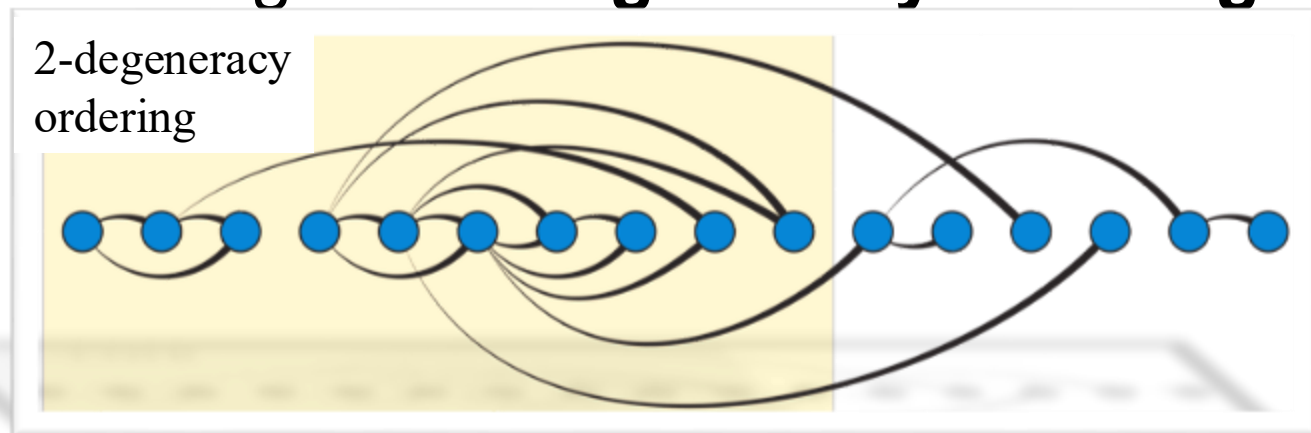
- Real-world graphs tend to have small degeneracy,  $d$ .

graph	n	m	d
zachary [48]	34	78	4
dolphins [35]	62	159	4
power [47]	4,941	6,594	5
polbooks [28]	105	441	6
adjnoun [29]	112	425	6
football [15]	115	613	8
lesmis [25]	77	254	9
celegensneural [47]	297	1,248	9
netscience [39]	1,589	2,742	19
internet [40]	22,963	48,421	25
condmat-2005 [38]	40,421	175,693	29
polblogs [4]	1,490	16,715	36
astro-ph [38]	16,706	121,251	56

graph	n	m	d
roadNet-CA [34]	1,965,206	2,766,607	3
roadNet-PA [34]	1,088,092	1,541,898	3
roadNet-TX [34]	1,379,917	1,921,660	3
amazon0601 [30]	403,394	2,443,408	10
email-EuAll [31]	265,214	364,481	37
email-Enron [24]	36,692	183,831	43
web-Google [2]	875,713	4,322,051	44
soc-wiki-Vote [33]	7,115	100,762	53
soc-slashdot0902 [34]	82,168	504,230	55
cit-Patents [18]	3,774,768	16,518,947	64
soc-Epinions1 [42]	75,888	405,740	67
soc-wiki-Talk [33]	2,394,385	4,659,565	131
web-berkstan [34]	685,231	6,649,470	201

# Degeneracy Ordering Algorithm

- Degeneracy can be computed by a simple **greedy algorithm**:
  - Repeatedly find and remove the vertex of smallest degree, adding it to the end of the list.
  - The degeneracy is then the highest degree,  $d$ , of any vertex at the moment it is removed.
  - The ordering is a  **$d$ -degeneracy ordering**.





# Linear-time Implementation

1. Initialize an output list,  $L$ , to be empty.
2. Compute a number,  $d_v$ , for each vertex  $v$  in  $G$ , which is the number of neighbors of  $v$  that are not already in  $L$ . Initially,  $d_v$  is just the degrees of  $v$ .
3. Initialize an array  $D$  such that  $D[i]$  contains a list of the vertices  $v$  that are not already in  $L$  for which  $d_v = i$ .
4. Let  $N_v$  be a list of the neighbors of  $v$  that come before  $v$  in  $L$ . Initially,  $N_v$  is empty for every vertex  $v$ .
5. Initialize  $k$  to 0.
6. Repeat  $n$  times:
  - Let  $i$  be the smallest index such that  $D[i]$  is nonempty.
  - Set  $k$  to  $\max(k, i)$ .
  - Select a vertex  $v$  from  $D[i]$ . Add  $v$  to the beginning of  $L$  and remove it from  $D[i]$ . Mark  $v$  as being in  $L$  (e.g., using a hash table,  $H_L$ ).
  - For each neighbor  $w$  of  $v$  not already in  $L$  (you can check this using  $H_L$ ):
    - Subtract one from  $d_w$
    - Move  $w$  to the cell of  $D$  corresponding to the new value of  $d_w$ , i.e.,  $D[d_w]$
    - Add  $w$  to  $N_v$

# Triangle Counting Algorithm

- Compute a  $d$ -degeneracy ordering of the vertices, e.g., using the algorithm of the previous slide.
- Process the vertices according to this ordering,  $L$ :  
For each vertex,  $v$ :  
For each pair of vertices,  $u$  and  $w$ , adjacent to  $v$  and earlier in the ordering, i.e.,  $u$  and  $w$  are in the list  $N_v$  from the degeneracy algorithm:  
If  $(u, w)$  is an edge in the graph, then add one to the triangle count.
- Running time is  $O(d^2n) = O(dm)$  expected, assuming edges are stored in a hash table.