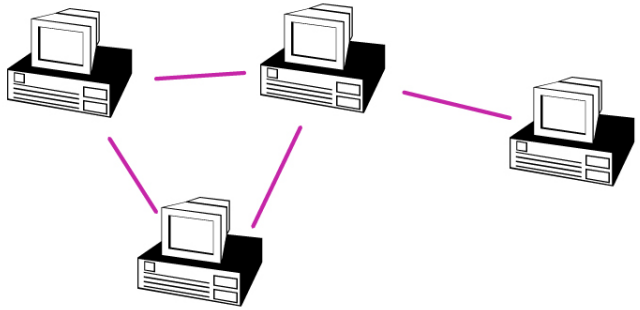
CptS 591: Elements of Network Science

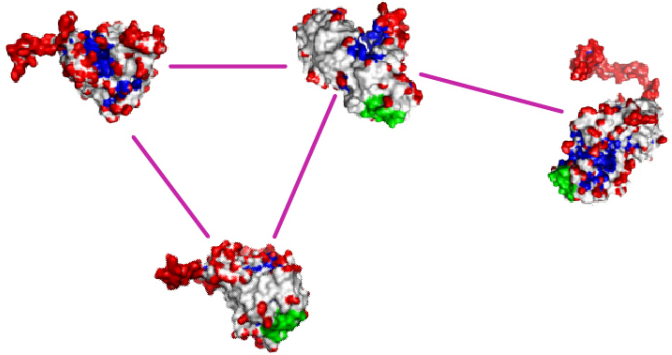
Basic Network Properties

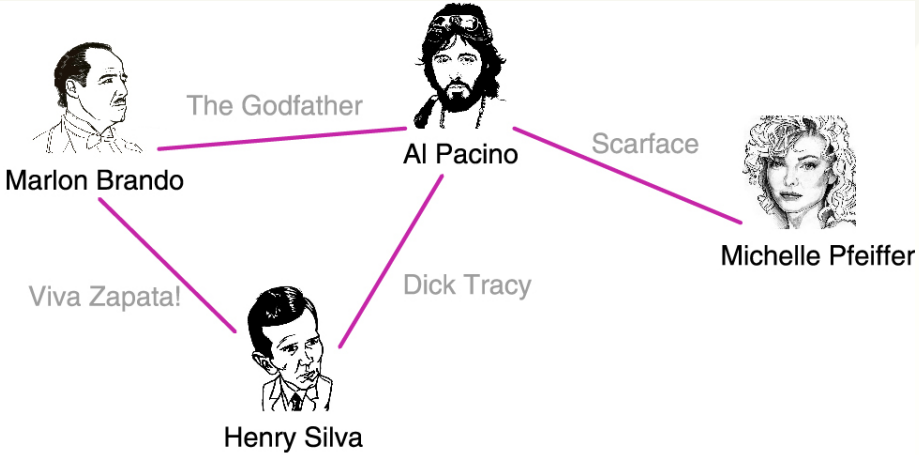
Outline

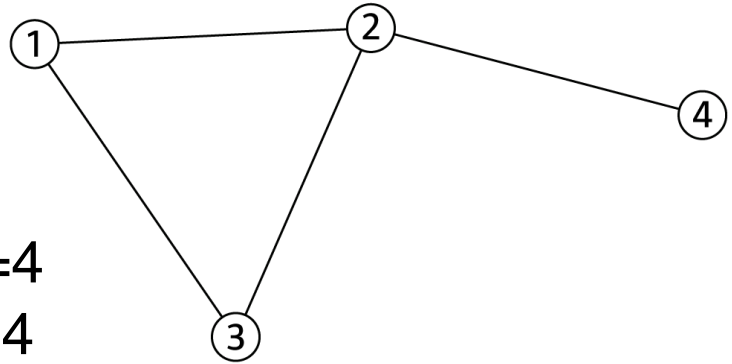
- Three central quantities:
  - Degree distribution
  - Network diameter and average path length
  - Clustering coefficient
- Associated side discussions:
  - Networks and Adjacency Matrices
  - Types of networks
- Note:
  - This material is adapted from slides of the "Graph Theory" section/chapter in the "Network Science" course/book by Albert-László Barabási.

A Common Language









N=4

L=4

Choosing a Proper Representation

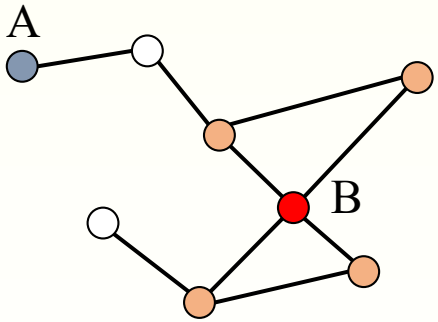
- The choice of the proper network representation determines our ability to use network science successfully.
- In some cases there is a unique, unambiguous representation.
- In other cases, the representation is by no means unique.
- For example, the way we assign the links between a group of individuals will determine the nature of the question we can study.

Basic Network Properties

Degree, degree distribution.

Node Degrees

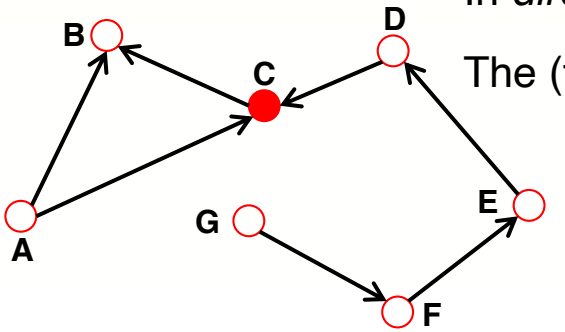
Undirected



Node degree: the number of links connected to the node.

$$k_A = 1 \qquad k_B = 4$$

Directed



In *directed networks* we can define an in-degree and out-degree.

The (total) degree is the sum of in- and out-degree.

$$k_C^{in} = 2 \qquad k_C^{out} = 1 \qquad k_C = 3$$

Source: a node with $k^{in} = 0$; Sink: a node with $k^{out} = 0$.

A Bit of Statistics

BRIEF STATISTICS REVIEW

Four key quantities characterize a sample of N values $x_1, \ldots, x_N$:

Average (mean):

$$\langle x \rangle = \frac{x_1 + x_2 + \ldots + x_N}{N} = \frac{1}{N}\sum_{i=1}^{N} x_i$$

The n^{th} moment:

$$\langle x^n \rangle = \frac{x_1^n + x_2^n + \ldots + x_N^n}{N} = \frac{1}{N}\sum_{i=1}^{N} x_i^n$$

Standard deviation:

$$\sigma_x = \sqrt{\frac{1}{N}\sum_{i=1}^{N}\left(x_i - \langle x \rangle\right)^2}$$

Distribution of x:

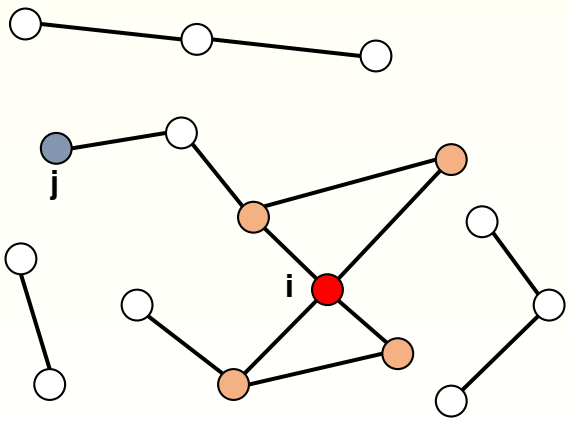
$$p_x = \frac{1}{N}\sum_i \delta_{x,x_i}$$

where p_x follows

$$\sum_i p_x = 1 \quad \left(\int p_x \, dx = 1\right)$$

Average Degree

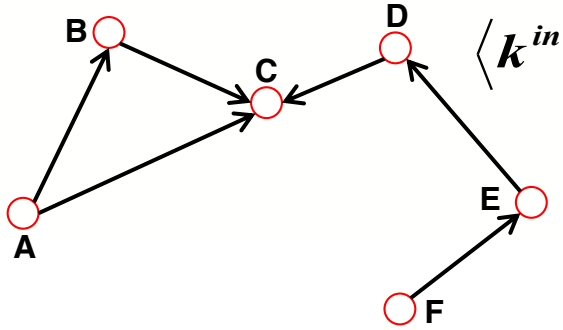
Undirected



$$\langle k \rangle \equiv \frac{1}{N}\sum_{i=1}^{N} k_i \qquad \langle k \rangle \equiv \frac{2L}{N}$$

N – the number of nodes in the graph

Directed



$$\langle k^{in} \rangle \equiv \frac{1}{N}\sum_{i=1}^{N} k_i^{in}, \quad \langle k^{out} \rangle \equiv \frac{1}{N}\sum_{i=1}^{N} k_i^{out}, \quad \langle k^{in} \rangle = \langle k^{out} \rangle$$

$$\langle k \rangle \equiv \frac{L}{N}$$

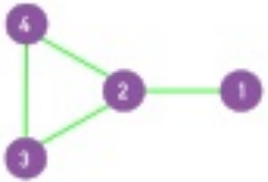
Average Degree

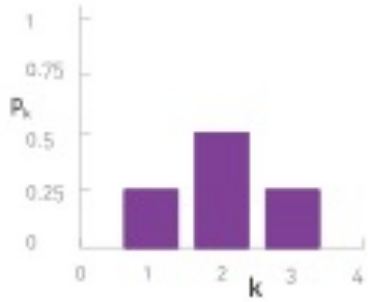
NETWORK	NODES	LINKS	DIRECTED UNDIRECTED	N	L	⟨k⟩
Internet	Routers	Internet connections	Undirected	192,244	609,066	6.33
WWW	Webpages	Links	Directed	325,729	1,497,134	4.60
Power Grid	Power plants, transformers	Cables	Undirected	4,941	6,594	2.67
Mobile Phone Calls	Subscribers	Calls	Directed	36,595	91,826	2.51
Email	Email addresses	Emails	Directed	57,194	103,731	1.81
Science Collaboration	Scientists	Co-authorship	Undirected	23,133	93,439	8.08
Actor Network	Actors	Co-acting	Undirected	702,388	29,397,908	83.71
Citation Network	Paper	Citations	Directed	449,673	4,689,479	10.43
E. Coli Metabolism	Metabolites	Chemical reactions	Directed	1,039	5,802	5.58
Protein Interactions	Proteins	Binding interactions	Undirected	2,018	2,930	2.90

Degree Distribution

Degree distribution

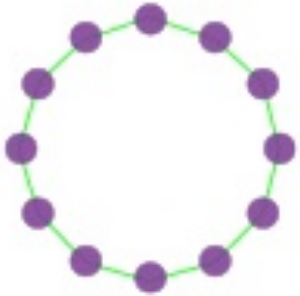
P(k): probability that a randomly chosen node has degree *k*

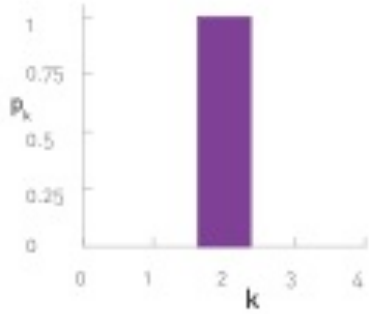




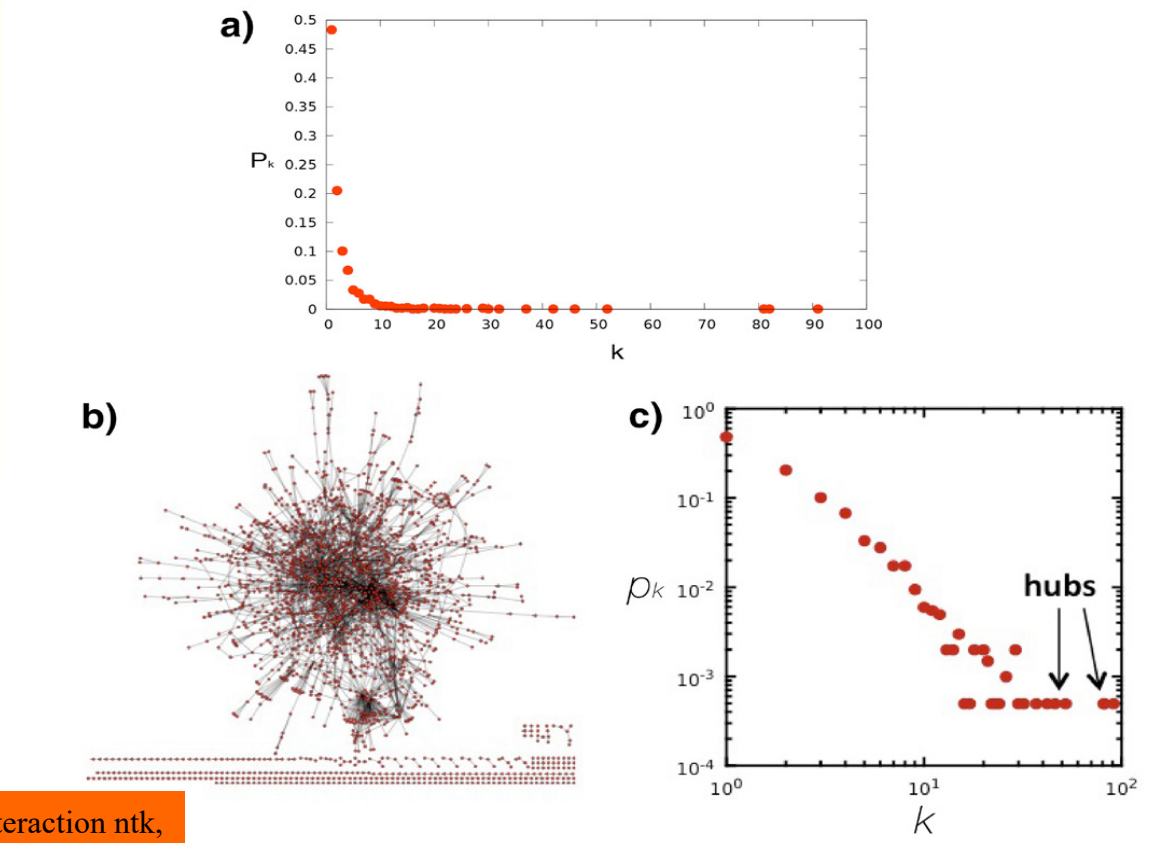
N_k = # nodes with degree k

P(k) = N_k / N ➔ plot





Degree Distribution



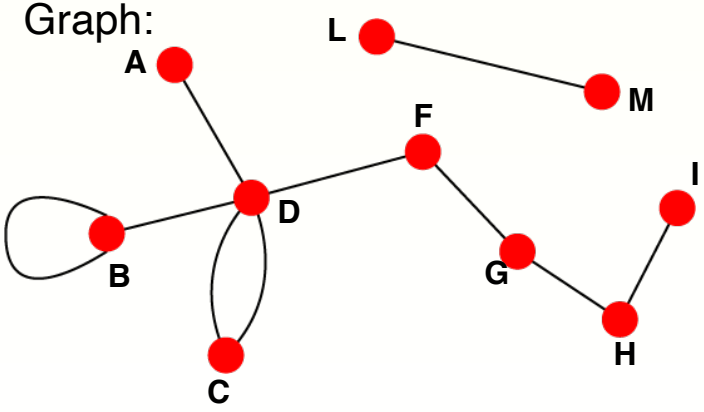
Protein interaction ntk, roughly 2k nodes.

Undirected vs Directed Networks

Undirected

Links: undirected (*symmetrical*)

Graph:

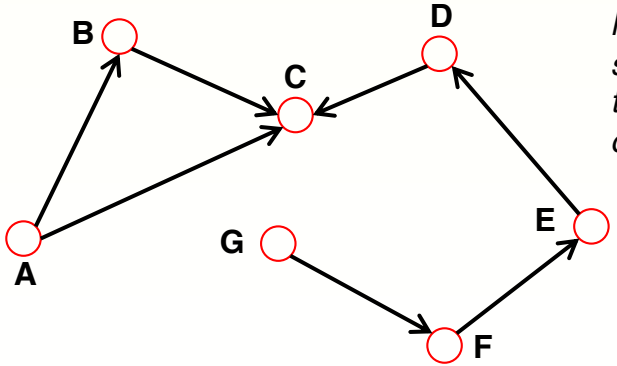


Undirected links :
coauthorship links
Actor network
protein interactions

Directed

Links: directed (*arcs*).

Digraph = directed graph:



An undirected link is the superposition of two opposite directed links.

Directed links :
URLs on the www
phone calls
metabolic reactions

(Reference) Networks

NETWORK	NODES	LINKS	DIRECTED UNDIRECTED	N	L
Internet	Routers	Internet connections	Undirected	192,244	609,066
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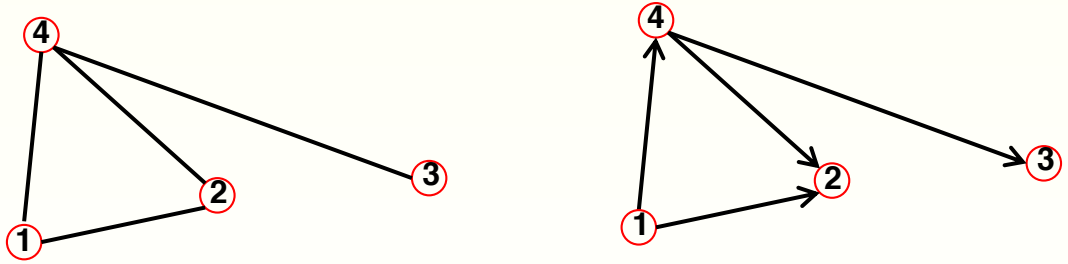
Networks and Matrices

Adjacency Matrices

Adjacency Matrix



A_{ij}=1 if there is a link between node *i* and *j*

A_{ij}=0 if nodes *i* and *j* are not connected to each other.

$$A_{ij} = \begin{pmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix} \qquad A_{ij} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

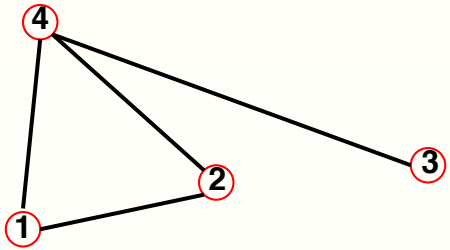
Note that for a directed graph (right) the matrix is not symmetric.

$A_{ij} = 1$ if there is a link pointing from node *j* and *i*

$A_{ij} = 0$ if there is no link pointing from *j* to *i*.

Adjacency Matrix and Node Degrees

Undirected



$$A_{ij} = \begin{pmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

$$k_i = \sum_{j=1}^{N} A_{ij}$$

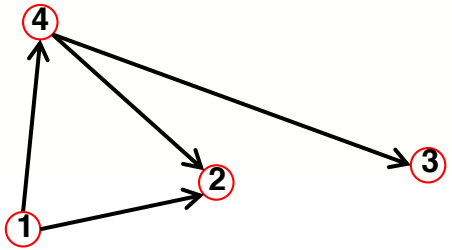
$$k_j = \sum_{i=1}^{N} A_{ij}$$

$$A_{ij} = A_{ji}$$

$$A_{ii} = 0$$

$$L = \frac{1}{2}\sum_{i=1}^{N} k_i = \frac{1}{2}\sum_{ij}^{N} A_{ij}$$

Directed



$$A_{ij} = \begin{pmatrix} 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 \end{pmatrix}$$

$$k_i^{in} = \sum_{j=1}^{N} A_{ij}$$

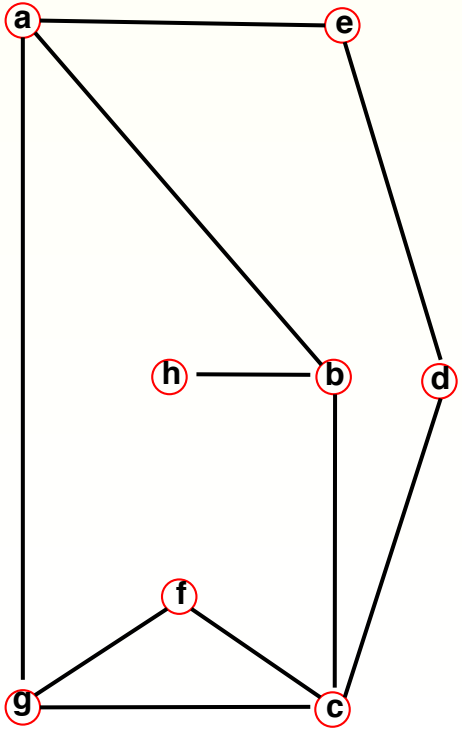
$$k_j^{out} = \sum_{i=1}^{N} A_{ij}$$

$$A_{ij} \neq A_{ji}$$

$$A_{ii} = 0$$

$$L = \sum_{i=1}^{N} k_i^{in} = \sum_{j=1}^{N} k_j^{out} = \sum_{i,j}^{N} A_{ij}$$

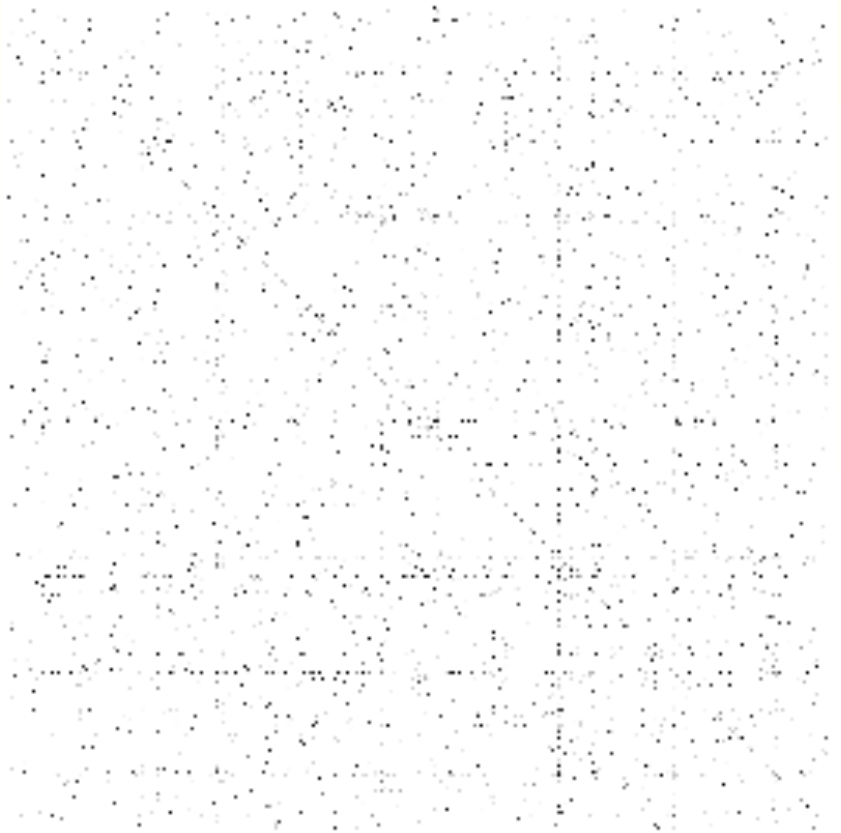
Adjacency Matrix



	a	b	c	d	e	f	g	h
a	0	1	0	0	1	0	1	0
b	1	0	1	0	0	0	0	1
c	0	1	0	1	0	1	1	0
d	0	0	1	0	1	0	0	0
e	1	0	0	1	0	0	0	0
f	0	0	1	0	0	0	1	0
g	1	0	1	0	0	0	0	0
h	0	1	0	0	0	0	0	0

The adjacency matrix can take far more complicated forms for a larger network

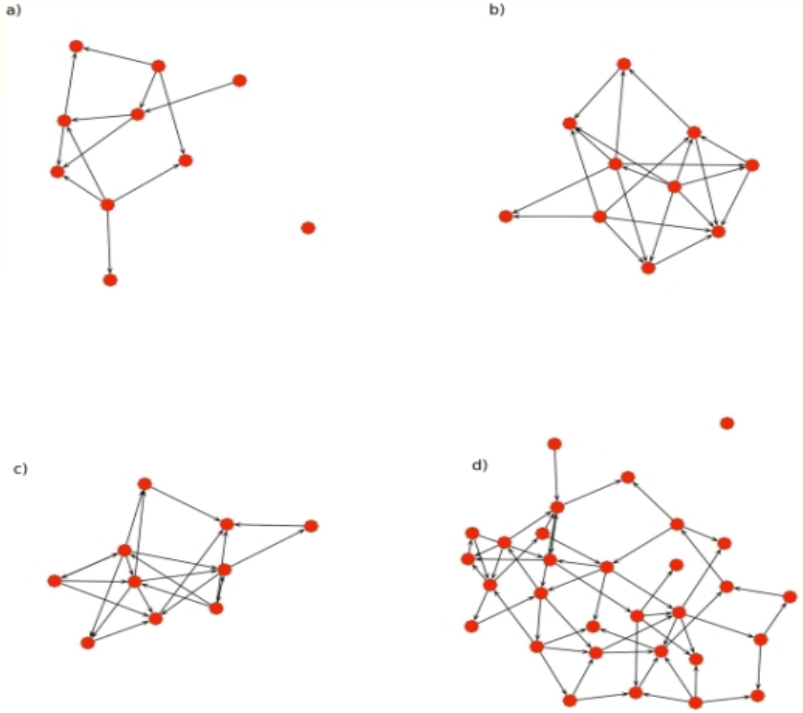
Adjacency Matrices are Sparse



Networks and Matrices

Sparseness

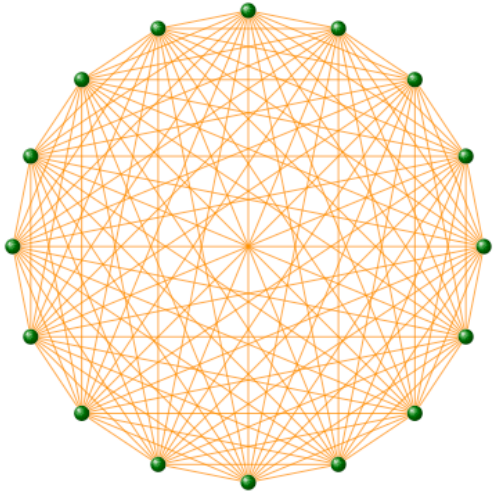
Real networks are sparse



Complete Graph (Clique)

The maximum number of links a network of N nodes can have is:

$$L_{\max} = \binom{N}{2} = \frac{N(N-1)}{2}$$



A graph with $L=L_{max}$ links is called a complete graph, and its average degree is **<k>=N-1**

Real World Networks Are Sparse

Most networks observed in real systems are sparse:

$$L \ll L_{max}$$

or

$$\langle k \rangle \ll N-1.$$

WWW (ND Sample): <k>=4.51	N = 325,729;	L=1.4 10^6	$L_{max}=10^{12}$
Protein (*S. Cerevisiae*): <k>=2.39	N =1,870;	L=4,470	$L_{max}=10^7$
Coauthorship (Math): <k>=3.9	N = 70,975;	L=2 10^5	$L_{max}=3$ x 10^{10}
Movie Actors: <k>=28.78	N = 212,250;	L=6 10^6	$L_{max}=1.8$ x10^{13}

(Source: Albert, Barabasi, RMP2002)

Weighted and Unweighted Networks

Weighted and Unweighted Networks

A weighted network is represented using an adjacency matrix in which the entries are real numbers, rather than just 0 and 1.

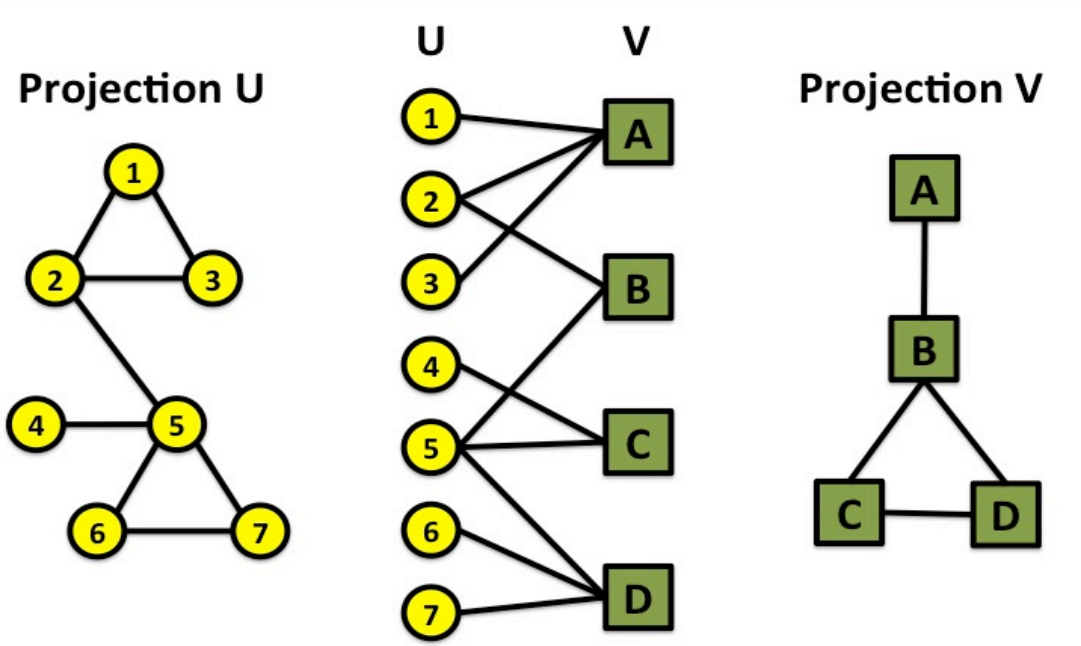
$$A_{ij} = w_{ij}$$

Network types

Bipartite Networks

Bipartite Graphs

A bipartite graph is a graph whose nodes can be divided into two disjoint sets U and V such that every link connects a node in U to one in V; that is, U and V are independent sets.



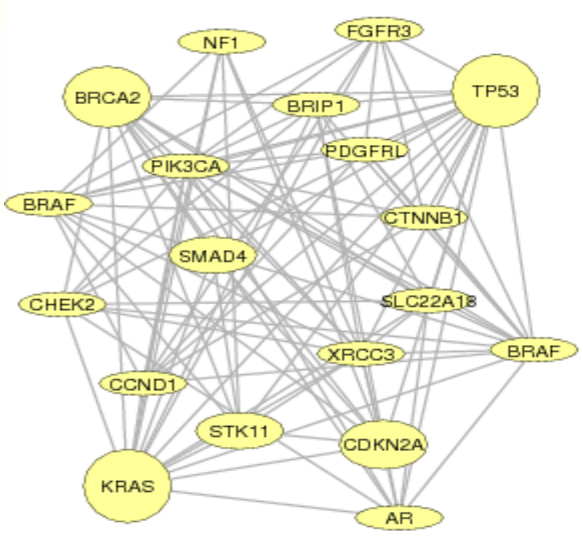
Examples:

Hollywood actor network

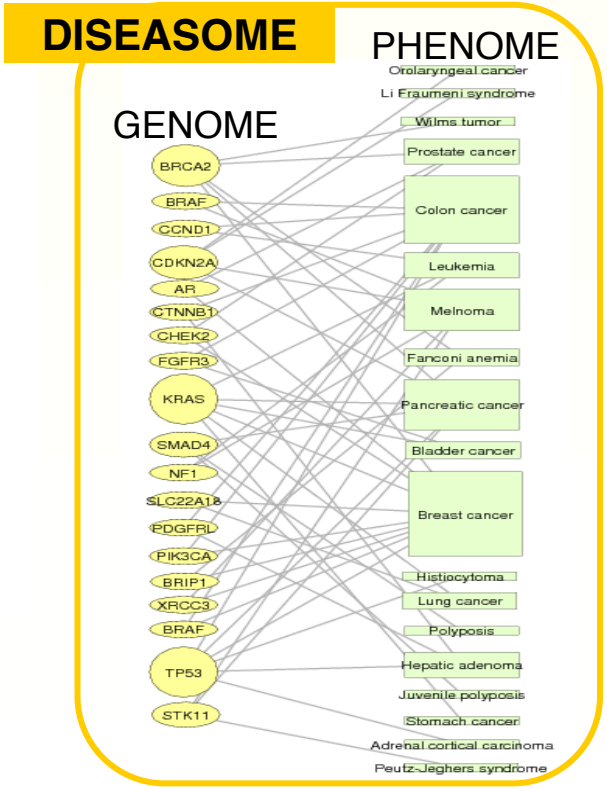
Collaboration networks

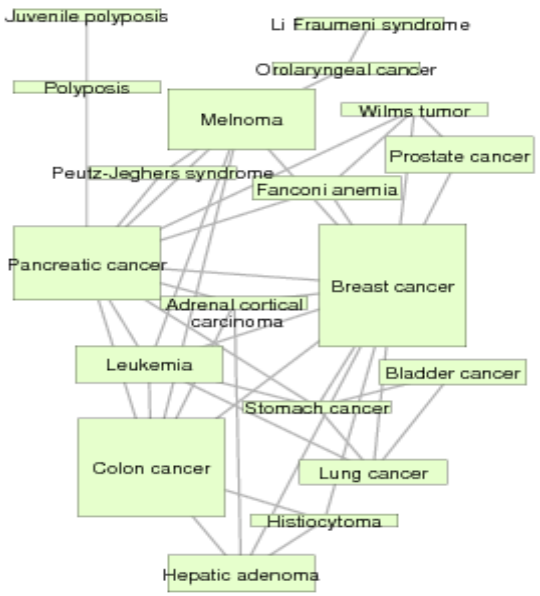
Disease network (diseasome)

Gene Network – Disease Network



Gene network

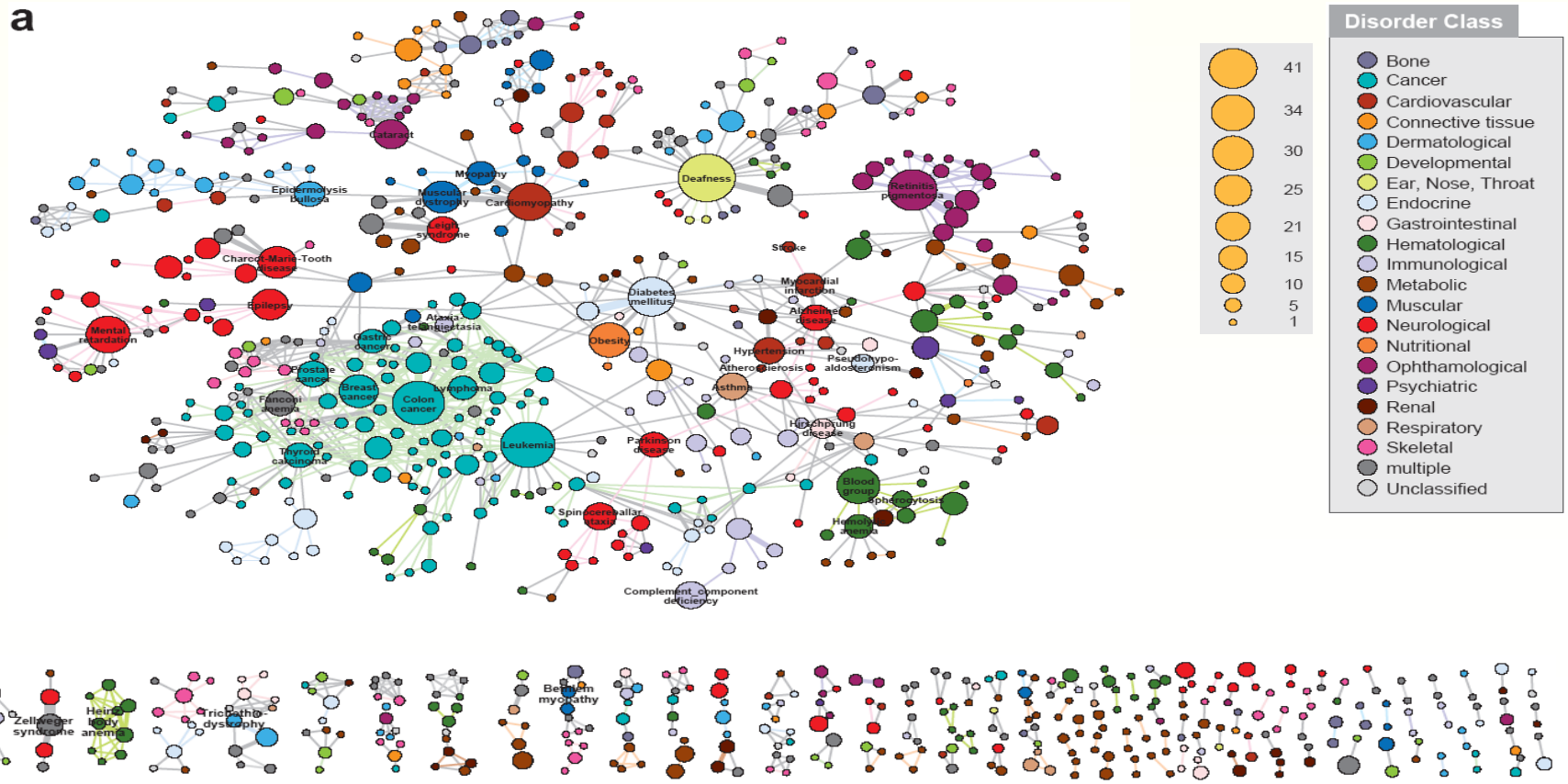




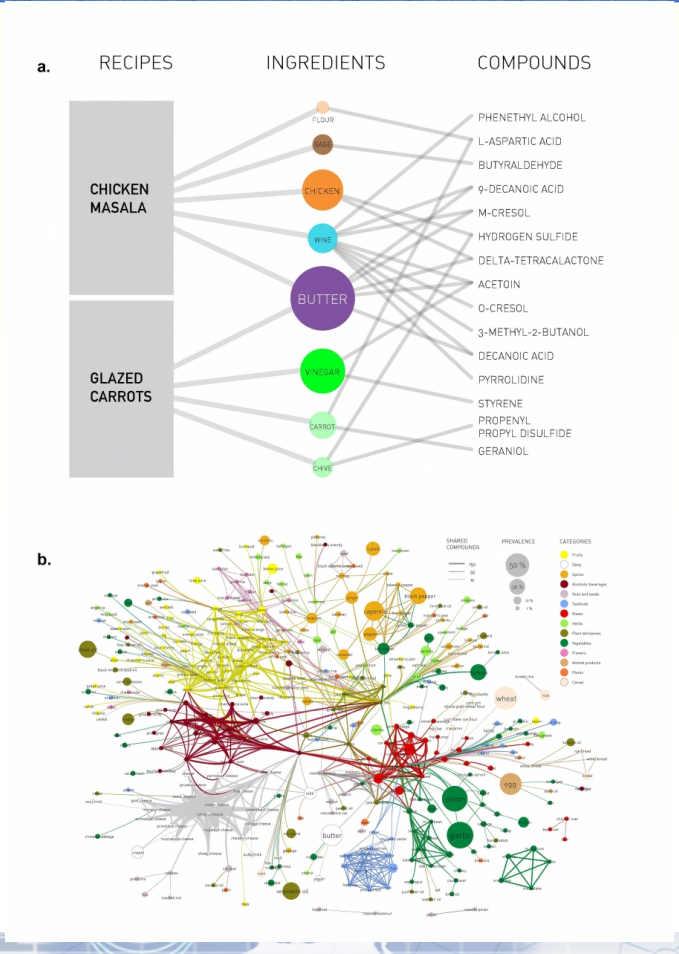
Disease network

Goh, Cusick, Valle, Childs, Vidal & Barabási, PNAS (2007)

Human Disease Network



Tripartite network: recipes, ingredients and compounds

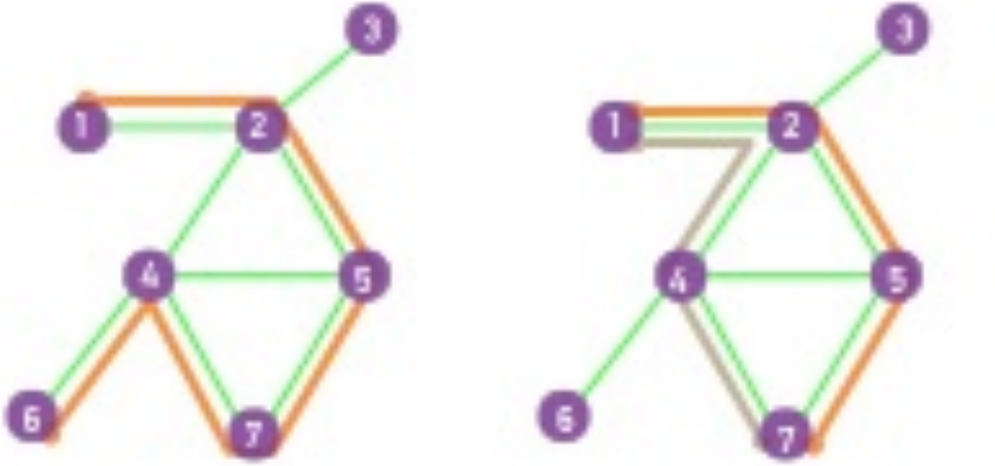


Paths

Paths (reminder)

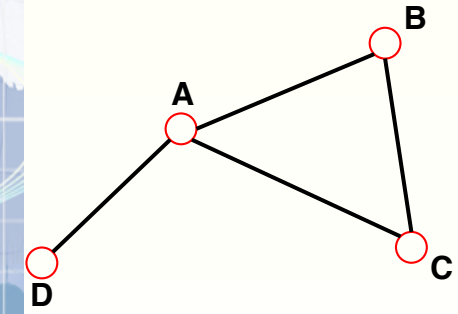
- A *path is* a sequence of nodes in which each node is adjacent to the next one
- P_{i_0,i_n} of length n between nodes i_0 and i_n is an ordered collection of $n+1$ nodes and n links

$$P_n = \{i_0, i_1, i_2, \ldots, i_n\} \qquad P_n = \{(i_0, i_1), (i_1, i_2), (i_2, i_3), \ldots, (i_{n-1}, i_n)\}$$



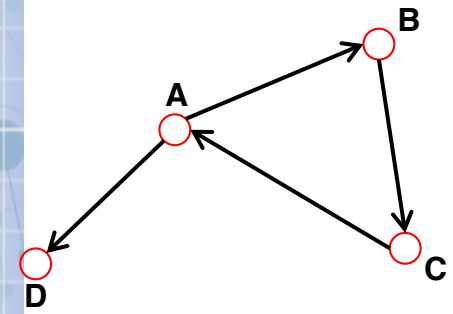
- In a directed network, the path can follow only the direction of an arrow.

Distance in a Graph
Shortest Path, Geodesic Path



The *distance (shortest path, geodesic path)* between two nodes is defined as the number of edges along the shortest path connecting them.

If the two nodes are disconnected, the distance is infinity.



In directed graphs each path needs to follow the direction of the arrows.

Thus in a digraph the distance from node A to B (on an AB path) is generally different from the distance from node B to A (on a BCA path).

Number of Paths Between Two Nodes
Adjacency Matrix

N_{ij},number of paths between any two nodes *i* and *j*:

Length n=1: If there is a link between *i* and *j*, then $A_{ij}=1$ and $A_{ij}=0$ otherwise.

Length n=2: If there is a path of length two between *i* and *j*, then $A_{ik}A_{kj}=1$, and $A_{ik}A_{kj}=0$ otherwise.

The number of paths of length 2:

$$N_{ij}^{(2)} = \sum_{k=1}^{N} A_{ik}A_{kj} = [A^2]_{ij}$$

Length n: In general, if there is a path of length *n* between *i* and *j*, then $A_{ik}\ldots A_{lj}=1$ and $A_{ik}\ldots A_{lj}=0$ otherwise.

The number of paths of length *n* between *i* and *j* is*

$$N_{ij}^{(n)} = [A^n]_{ij}$$

**holds for both directed and undirected networks.*

Network Diameter and Average Path

Diameter: $\boldsymbol{d_{max}}$ the maximum distance between any pair of nodes in the graph.

Average path length/distance, $\langle d \rangle$, for a connected graph:

where d_{ij} is the distance from node i to node j

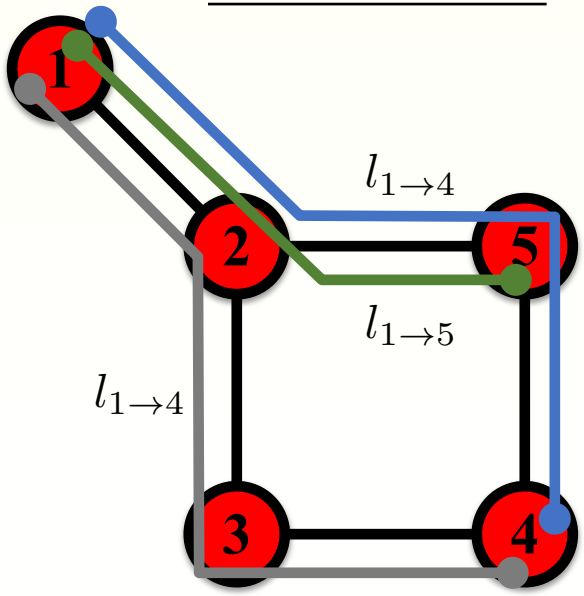
$$\langle d \rangle \equiv \frac{1}{2L_{\max}} \sum_{i,j \neq i} d_{ij}$$

In an *undirected graph* $d_{ij} = d_{ji}$, *so* we only need to count them once:

$$\langle d \rangle \equiv \frac{1}{L_{\max}} \sum_{i,j>i} d_{ij}$$

Paths: summary/example

Shortest Path



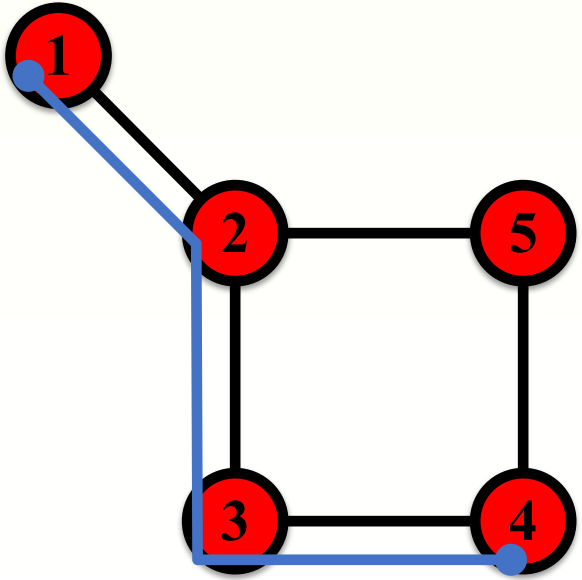
$$l_{1\to4} = 3$$

$$l_{1\to5} = 2$$

The path with the shortest length between two nodes (distance).

Paths: summary/example

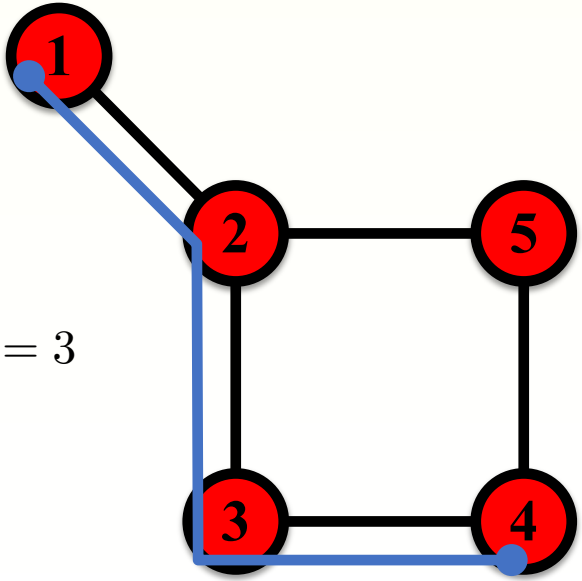
Diameter



$l_{1\to 4} = 3$

The longest shortest path in a graph

Average Path Length



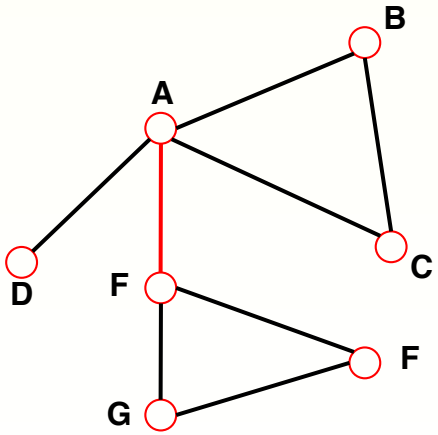
$$(l_{1\to 2} + l_{1\to 3} + l_{1\to 4} + \\ + l_{1\to 5} + l_{2\to 3} + l_{2\to 4} + \\ + l_{2\to 5} + l_{3\to 4} + l_{3\to 5} + \\ + l_{4\to 5})/10 = 1.6$$

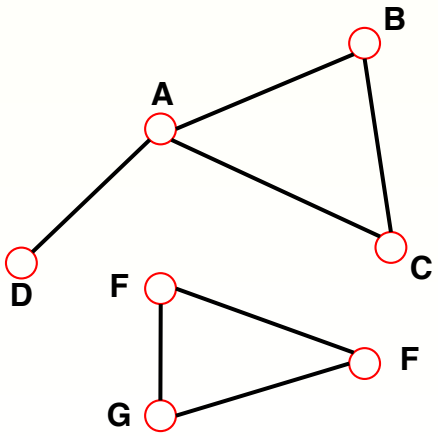
The average of the shortest paths for all pairs of nodes.

CONNECTEDNESS

Connectivity of undirected graphs (reminder)

Connected (undirected) graph: any two vertices can be joined by a path.
A disconnected graph is made up by two or more connected components.





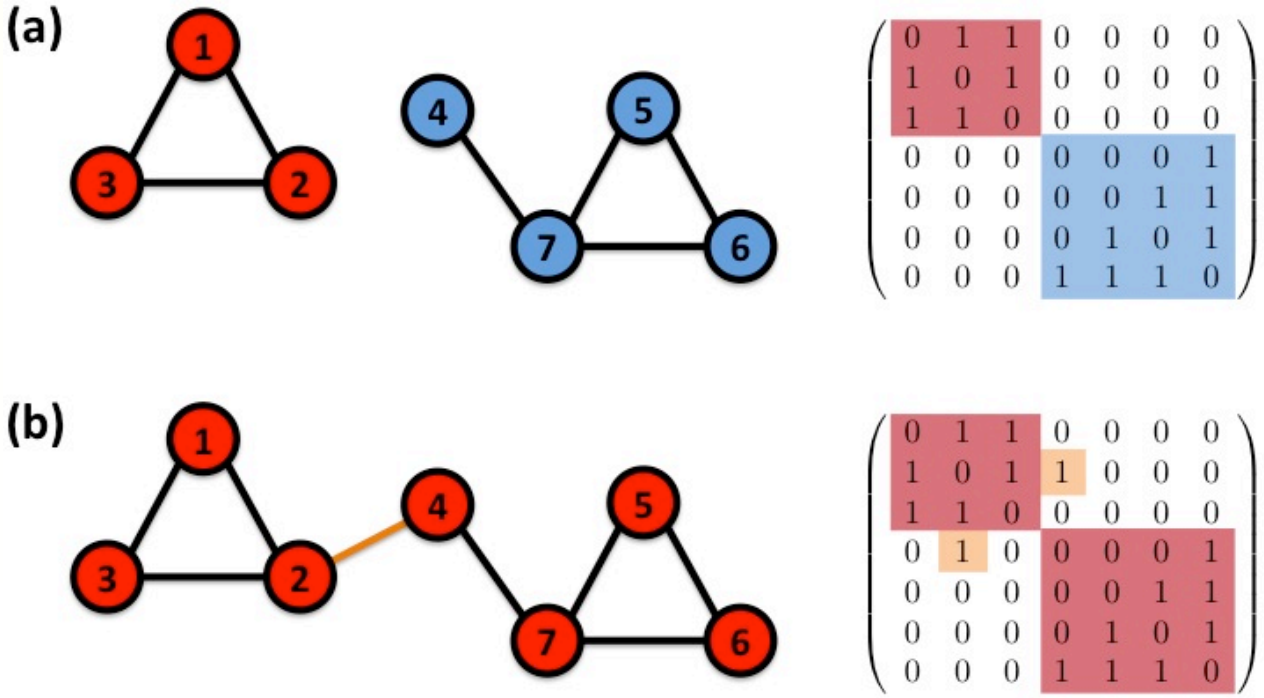
Largest Component:
Giant Component

The rest: **Isolates**

Bridge: if we erase it, the graph becomes disconnected.

Connectivity of Undirected Graphs
Adjacency Matrix

The adjacency matrix of a network with several components can be written in a block-diagonal form, so that nonzero elements are confined to squares, with all other elements being zero:

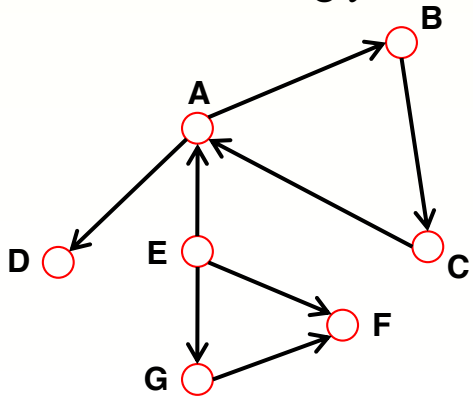


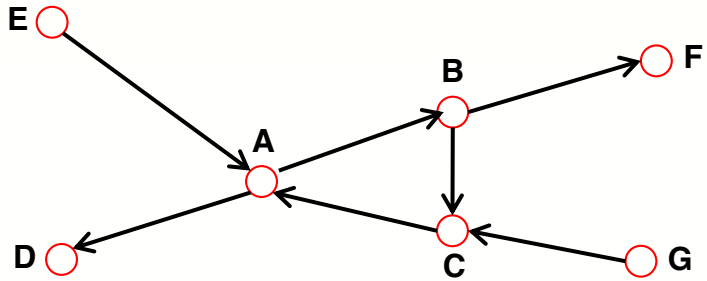
Connectivity of Directed Graphs

Strongly connected directed graph: has a path from each node to every other node and vice versa (e.g. AB path and BA path).

Weakly connected directed graph: it is connected if we disregard the edge directions.

Strongly connected components can be identified, but not every node is part of a nontrivial strongly connected component.





In-component: nodes that can reach the scc,

Out-component: nodes that can be reached from the scc.

Connected components: Algorithm

FINDING THE CONNECTED COMPONENTS OF A NETWORK

1. Start from a randomly chosen node i and perform a BFS (BOX 2.5). Label all nodes reached this way with $n = 1$.

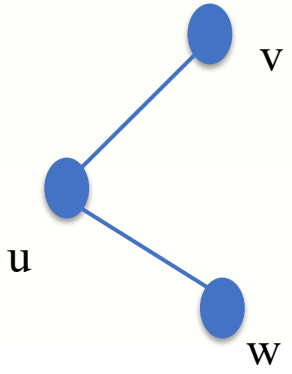
2. If the total number of labeled nodes equals N, then the network is connected. If the number of labeled nodes is smaller than N, the network consists of several components. To identify them, proceed to step 3.

3. Increase the label $n \rightarrow n + 1$. Choose an unmarked node j, label it with n. Use BFS to find all nodes reachable from j, label them all with n. Return to step 2.

Another basic property

Clustering coefficient

Clustering coefficient



- Measures the average probability that two neighbors of a vertex are themselves neighbors
- In effect it measures the density of triangles in the network
- It is of interest because in many cases it is found to have values different from what are expected on the basis of chance, because of "triadic closure"
- Two kinds (with many variations) of coefficients are used in the literature:
  - Global
  - Local

Global Clustering Coefficient

C = (number of closed paths of length two) /
(number of paths of length two)

0 <= C <= 1

no closed triads (e.g. trees)

All components are cliques

Alternatively,

C = (number of triangles) x 6 /
(number of paths of length two)

x 6 because each triangle is counted six times

Yet another expression,

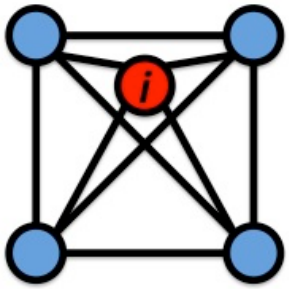
C = (number of triangles) x 3 /
(number of connected triplets)

Local clustering coefficient

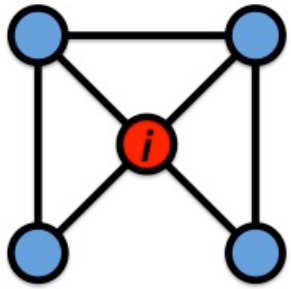
For each node i,

C_i = (number of pairs of neighbors of i that are connected) / (number of pairs of neighbors of i)

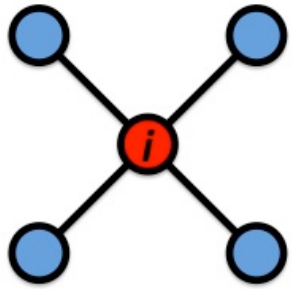
$$C_i = \frac{2e_i}{k_i(k_i - 1)}$$



$C_i = 1$



$C_i = 1/2$



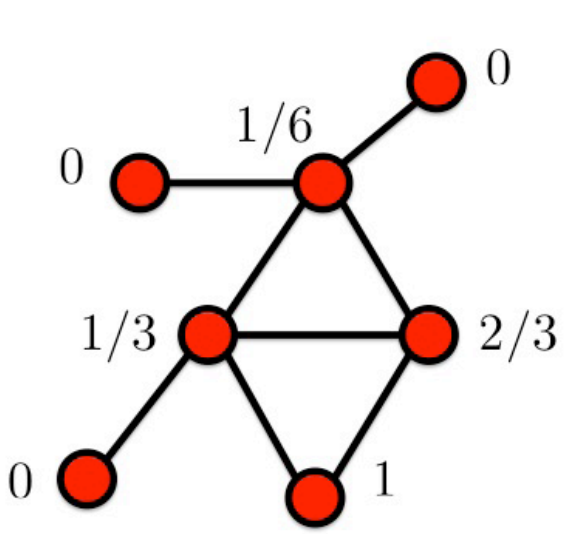
$C_i = 0$

$C_{WS} = C^{avg}$ = 1/n (sum of C_i), where n is number of nodes

$C^{avg} \neq C$

Watts & Strogatz, Nature 1998.

Example: global cc versus average local cc



$$\langle C \rangle = \frac{13}{42} \approx 0.310$$

$$C = \frac{3}{8} = 0.375$$

$$C_i = \frac{2e_i}{k_i(k_i - 1)}$$

Local clustering coefficient

Is interesting for several reasons

1) Empirically it is found to have rough dependence on degree:
   Vertices with higher degree have lower clustering coefficient on average

2) Can be used as an indicator of "structural holes"
   Structural holes: bad thing for efficient spread of information or other traffic
   Structural holes: can be good thing if we are measuring importance/influence
   lower local cc can mean higher centrality/importance

Summary of basic properties

Three Central Quantities in Network Science

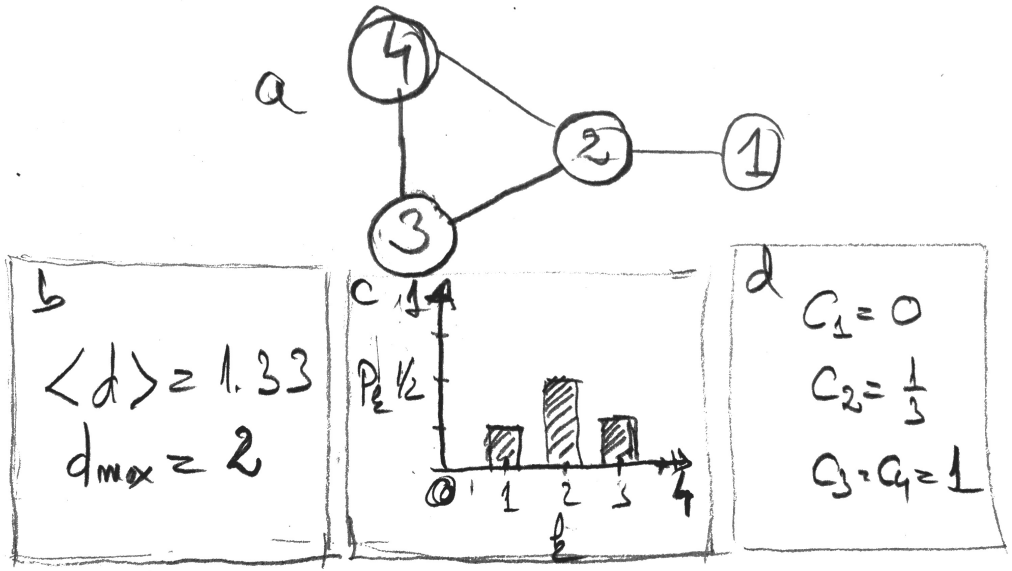
Degree distribution: **P(k)**

Path length: ***<d>***

Clustering coefficient:

$$C_i = \frac{2e_i}{k_i(k_i - 1)}$$

Three Central Quantities in Network Science

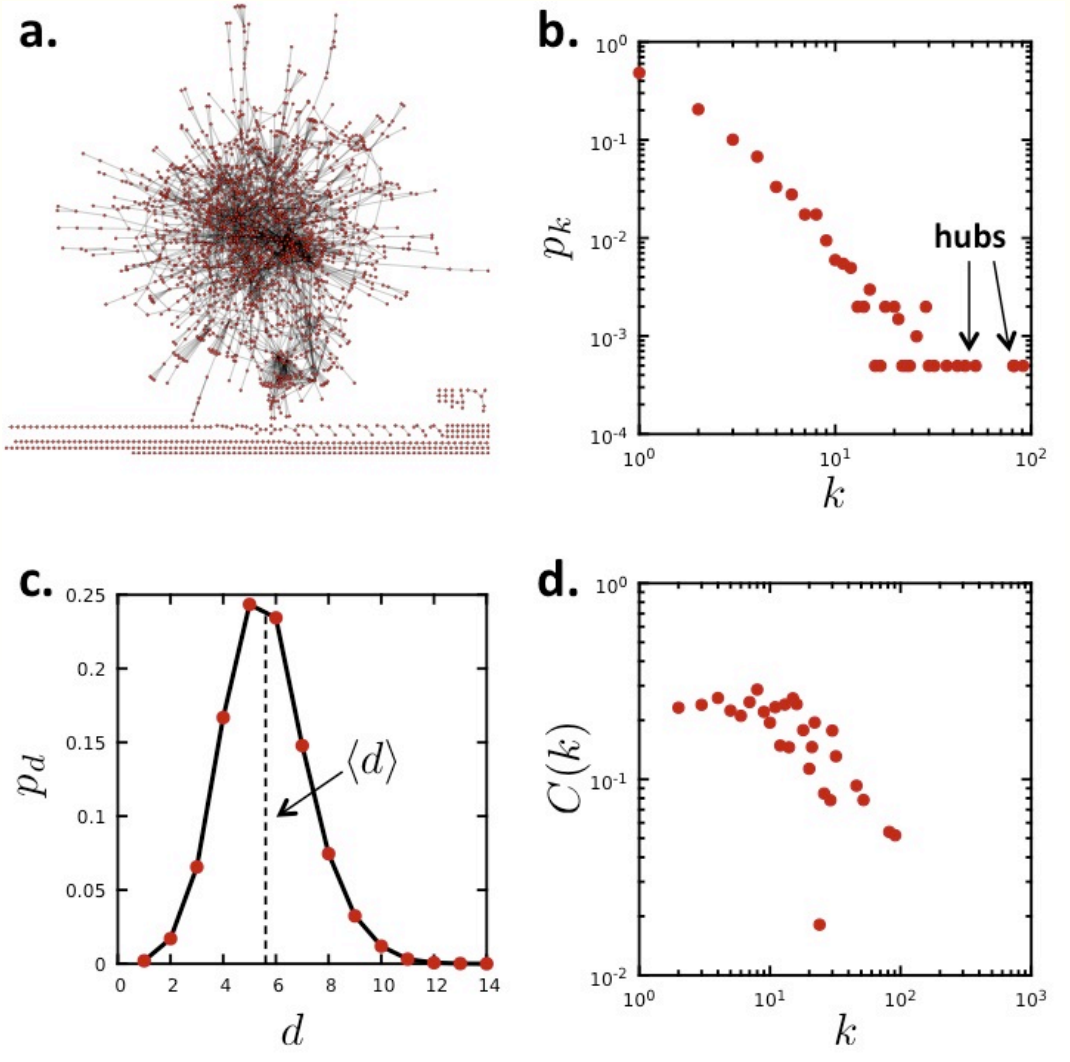


A. Degree distribution: p_k

B. Path length: $\langle d \rangle$

C. Clustering coefficient: $C_i = \frac{2e_i}{k_i(k_i - 1)}$

A CASE STUDY: PROTEIN-PROTEIN INTERACTION NETWORK



Undirected network

N=2,018 proteins as nodes

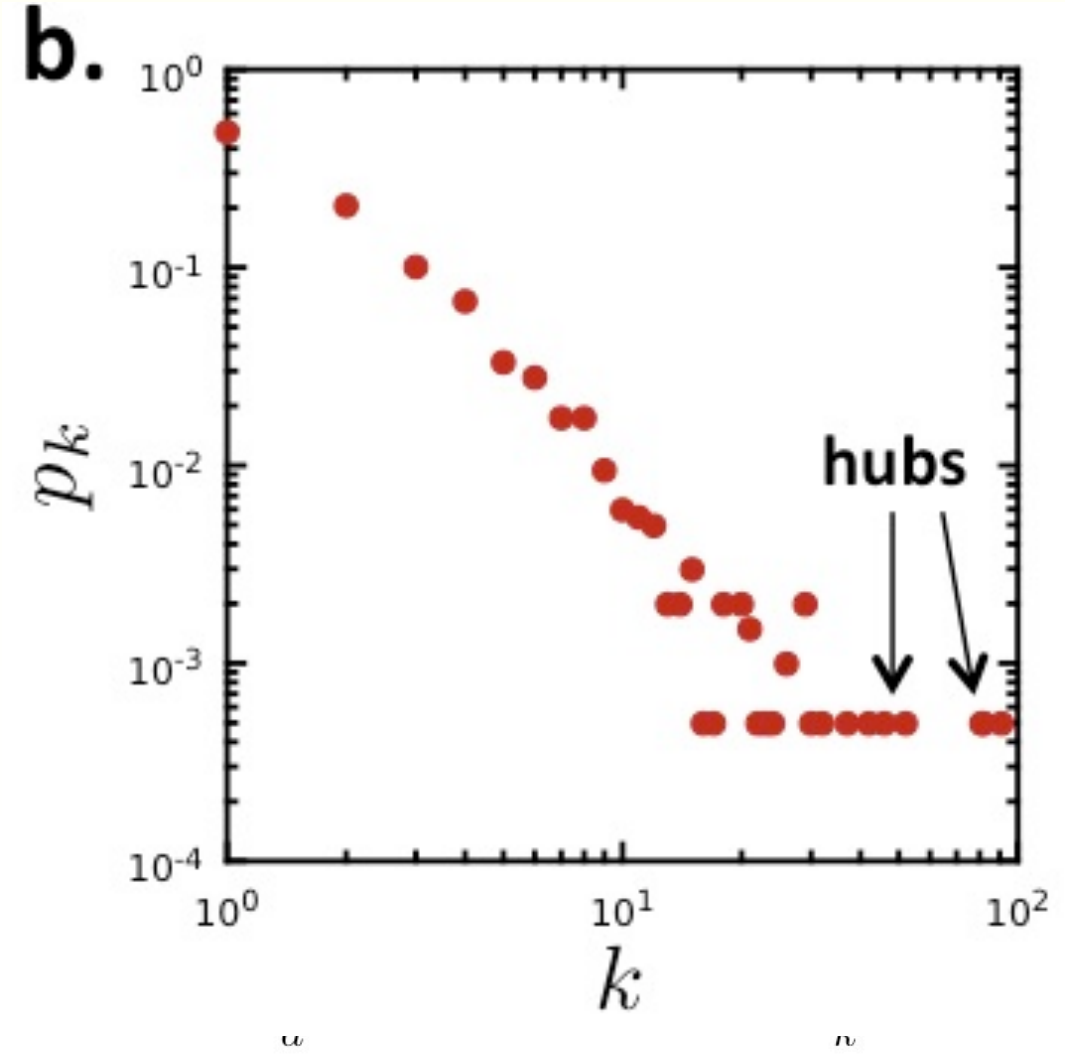
L=2,930 binding interactions as links.

Average degree <k>=2.90.

Not connected: 185 components

the largest (giant component) 1,647 nodes

A CASE STUDY: PROTEIN-PROTEIN INTERACTION NETWORK

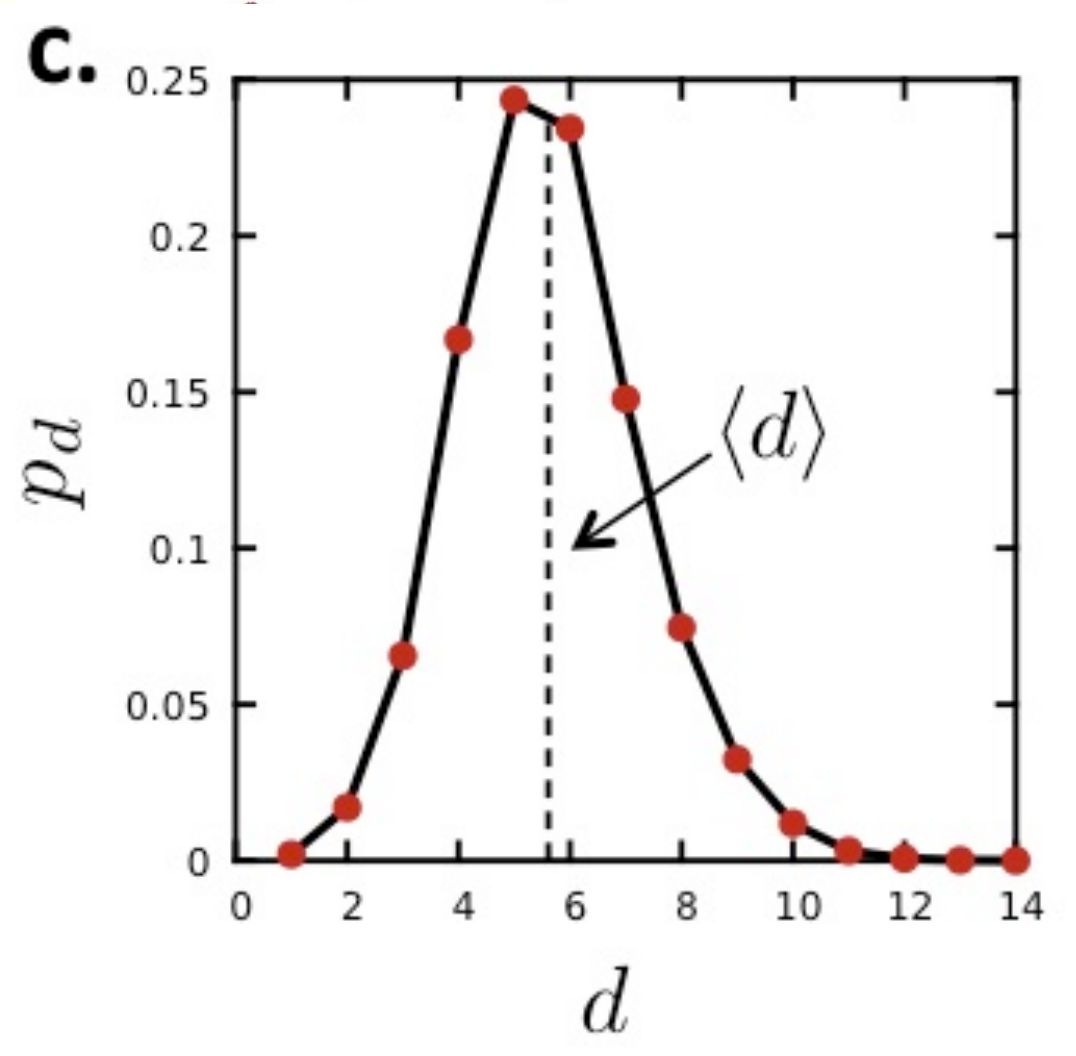


$\mathbf{p_k}$ is the probability that a node has degree **k**.

N_k = # nodes with degree k

$p_k = N_k / N$

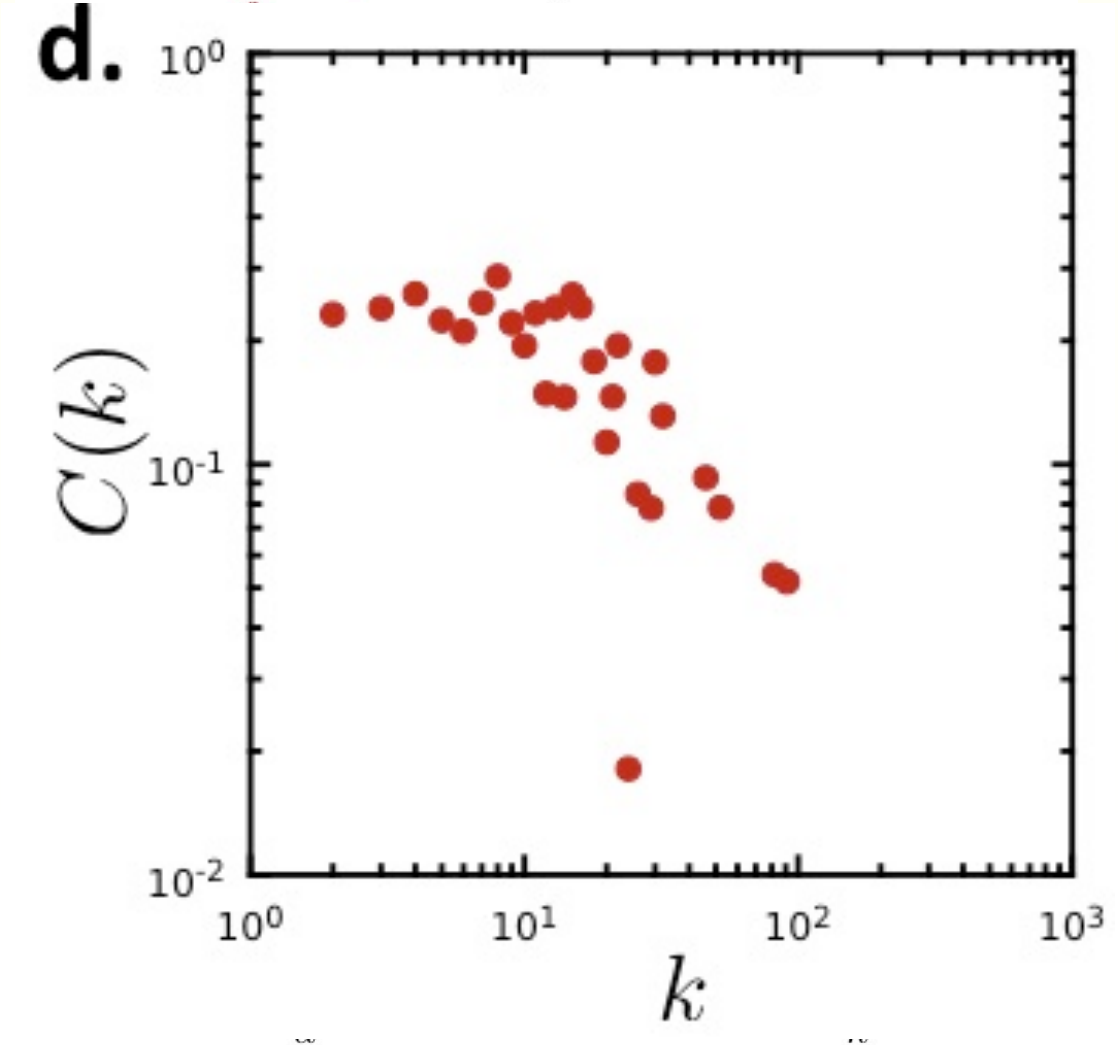
A CASE STUDY: PROTEIN-PROTEIN INTERACTION NETWORK

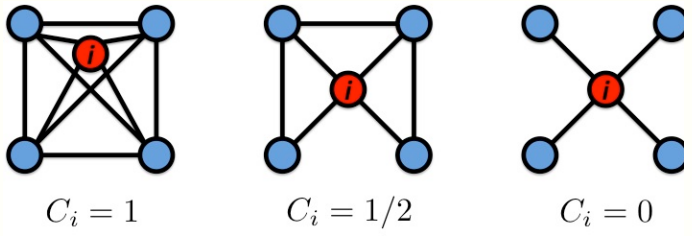


$d_{max}=14$

$\langle d \rangle=5.61$

A CASE STUDY: PROTEIN-PROTEIN INTERACTION NETWORK





$$C_i = \frac{2e_i}{k_i(k_i - 1)}$$

<C>=0.12

Further reading

Network Science book by Barabasi, Chapter 2:

http://networksciencebook.com/chapter/2