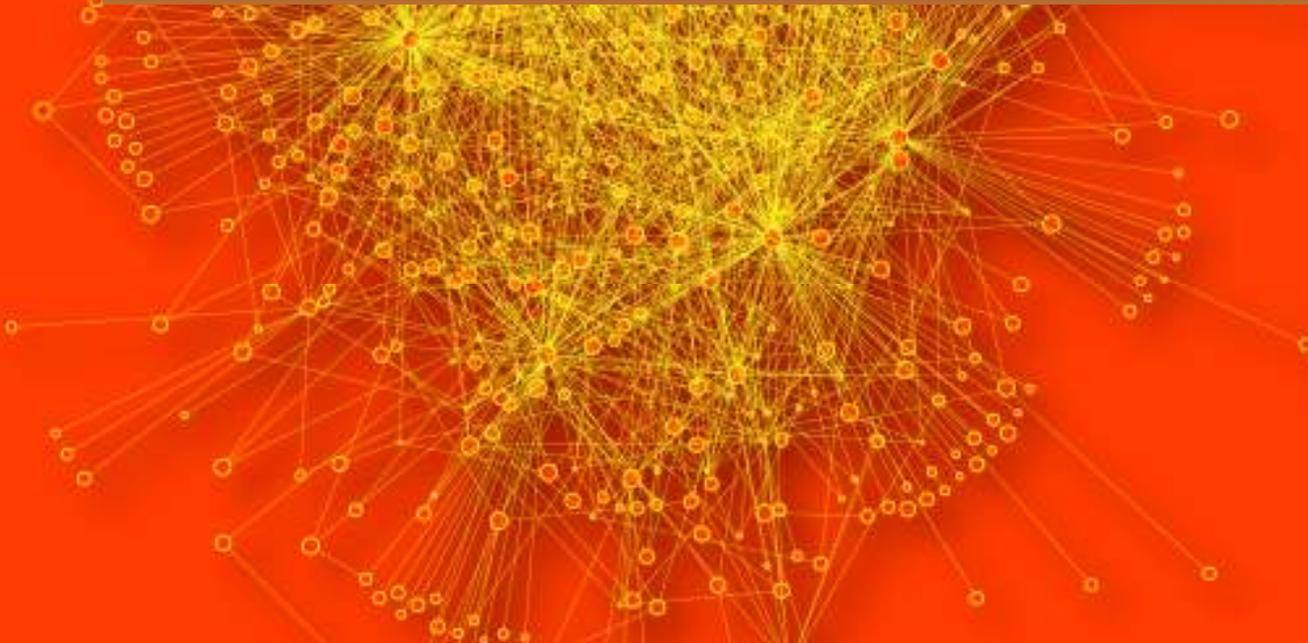


# CENTRALITY MEASURES

Measure the “importance”  
of a node in a network.





# Hollywood Revolves Around

Click on a name to see that person's table.

[Steiger, Rod](#) (2.678695)

[Lee, Christopher \(I\)](#) (2.684104)

[Hopper, Dennis](#) (2.698471)

[Sutherland, Donald \(I\)](#) (2.701850)

[Keitel, Harvey](#) (2.705573)

[Pleasence, Donald](#) (2.707490)

[von Sydow, Max](#) (2.708420)

[Caine, Michael \(I\)](#) (2.720621)

[Sheen, Martin](#) (2.721361)

[Quinn, Anthony](#) (2.722720)

[Heston, Charlton](#) (2.722904)

[Hackman, Gene](#) (2.725215)

[Connery, Sean](#) (2.730801)

[Stanton, Harry Dean](#) (2.737575)

[Welles, Orson](#) (2.744593)

[Mitchum, Robert](#) (2.745206)

[Gould, Elliott](#) (2.746082)

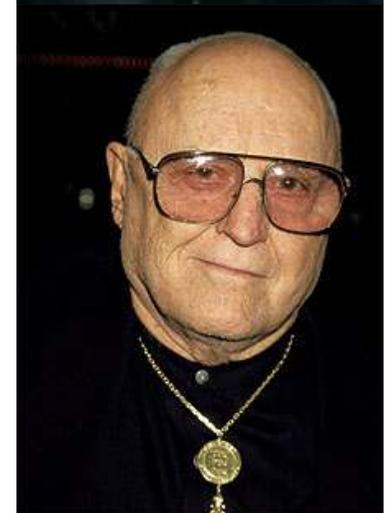
[Plummer, Christopher \(I\)](#) (2.746427)

[Coburn, James](#) (2.746822)

[Borgnine, Ernest](#) (2.747229)



Rod Steiger



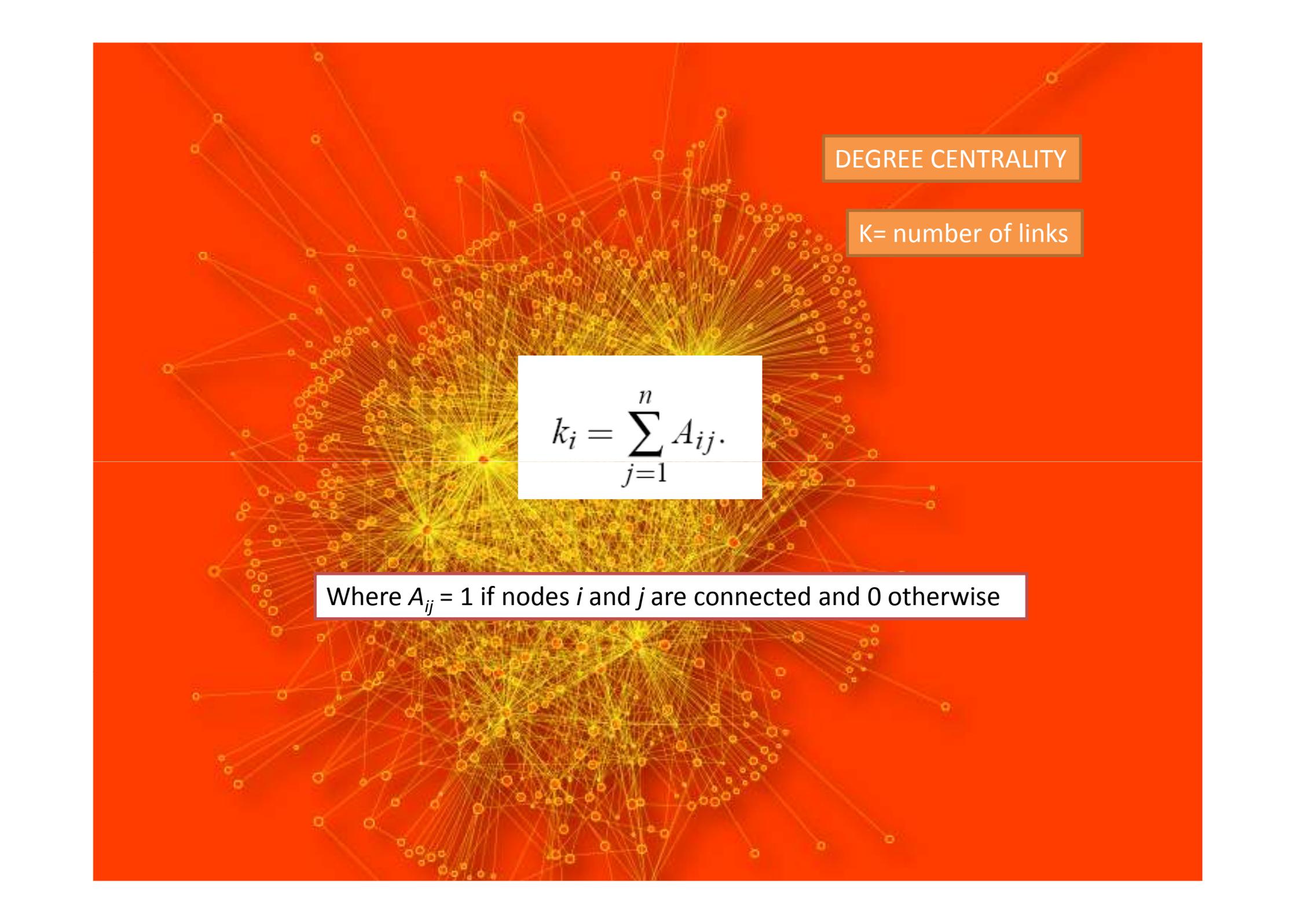
# Most Connected Actors in Hollywood

(measured in the late 90's)

Mel Blanc 759
Tom Byron 679
Marc Wallice 535
Ron Jeremy 500
Peter North 491
TT Boy 449
Tom London 436
Randy West 425
Mike Horner 418
Joey Silvera 410



XXX



DEGREE CENTRALITY

K= number of links

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

Where  $A_{ij} = 1$  if nodes  $i$  and  $j$  are connected and 0 otherwise

## BETWEENNESS CENTRALITY

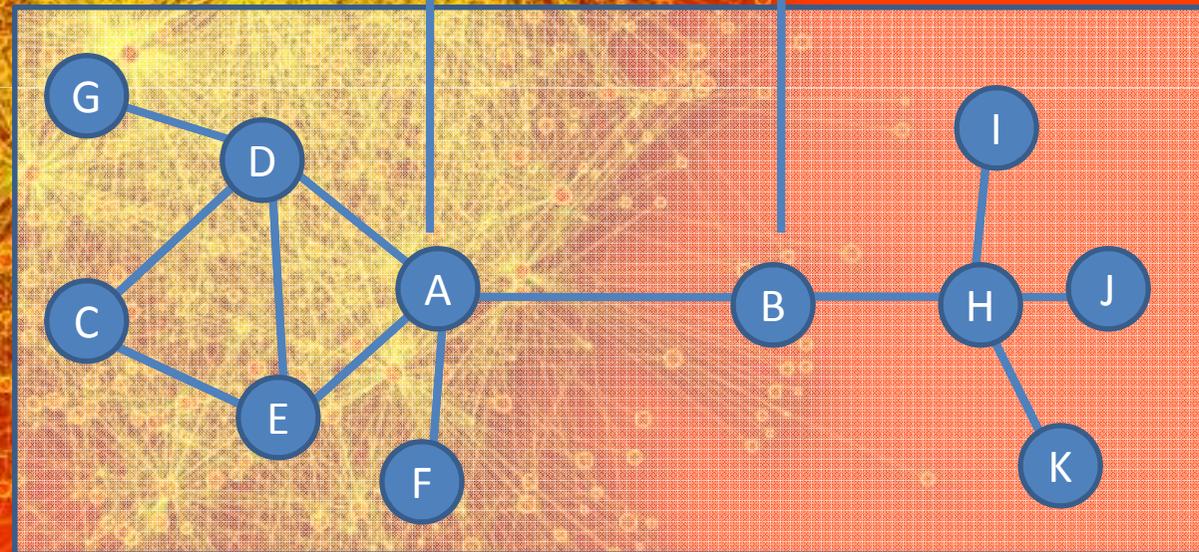
BC= number of shortest Paths that go through a node.

$$BC(G)=0$$

$$BC(D)=9+7/2=12.5$$

$$BC(A)=5*5+4=29$$

$$BC(B)=4*6=24$$



N=11

A set of measures of centrality based on  
betweenness

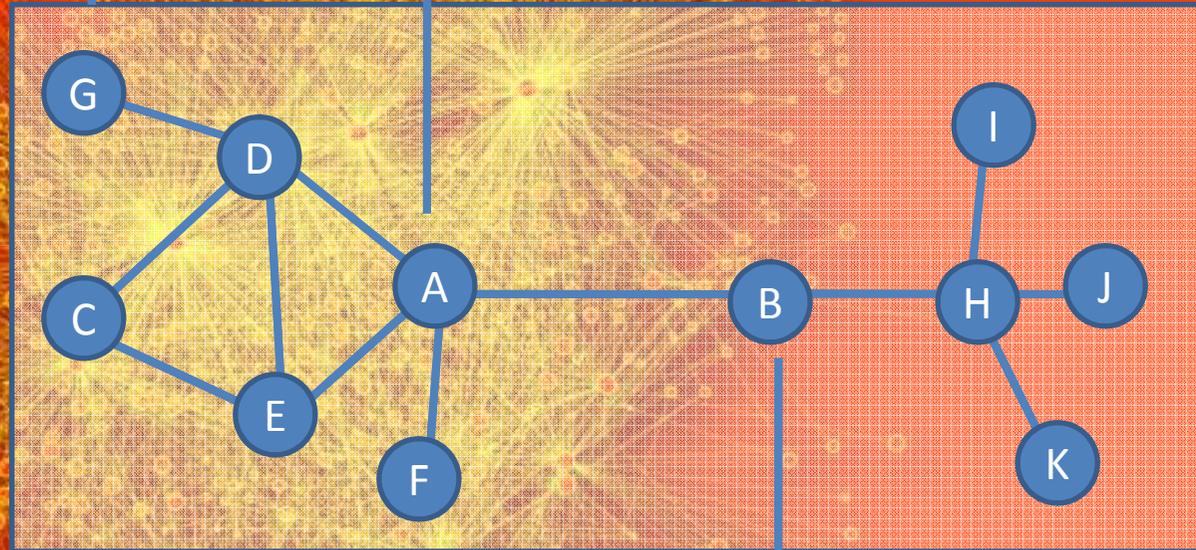
LC Freeman - Sociometry, 1977 - jstor.org

$$C(G) = \frac{1}{10}(1 + 2 \cdot 3 + 2 \cdot 3 + 4 + 3 \cdot 5)$$
$$C(G) = 3.2$$

$$C(A) = \frac{1}{10}(4 + 2 \cdot 3 + 3 \cdot 3)$$
$$C(A) = 1.9$$

## CLOSENESS CENTRALITY

C = Average Distance to neighbors



$$C(B) = \frac{1}{10}(2 + 2 \cdot 6 + 2 \cdot 3)$$
$$C(B) = 2$$

N=11

Consider the Adjacency Matrix  $A_{ij} = 1$  if node  $i$  is connected to node  $j$  and 0 otherwise. Now, measure the centrality of a node, as the sum over the centralities of all nodes....

$$x_i = \frac{1}{\lambda} \sum_{j \in M(i)} x_j = \frac{1}{\lambda} \sum_{j=1}^N A_{i,j} x_j$$

This is equivalent to eigenvalue problem:

$$\mathbf{Ax} = \lambda \mathbf{x}$$

Then the eigenvector centrality of node (i) is defined as:

$$x_i$$

where  $\lambda$  is the largest eigenvalue associated with  $\mathbf{A}$  and  $\mathbf{x}$  is its associated eigenvector.

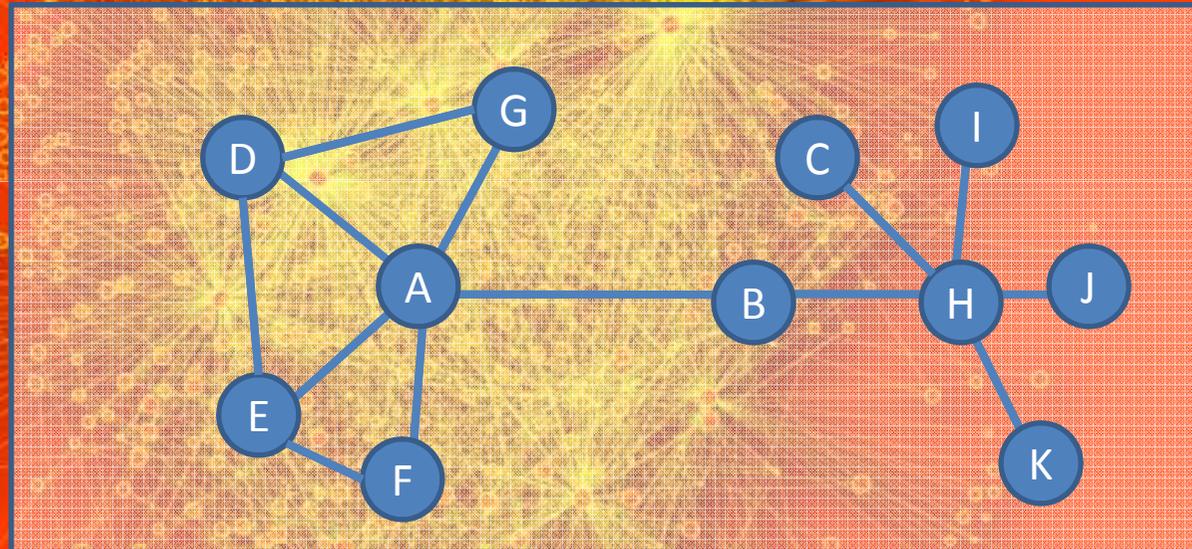


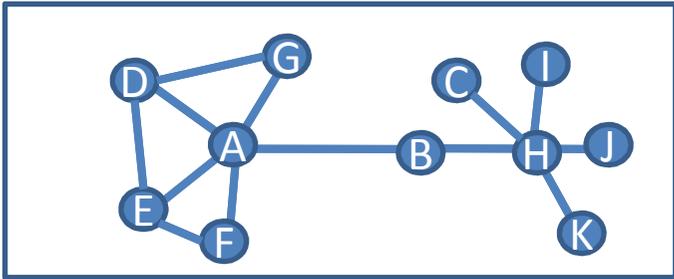
## Power and Centrality: A Family of Measures

Phillip Bonacich

*The American Journal of Sociology*, Vol. 92, No. 5 (Mar., 1987), 1170-1182.

$$c(\alpha, \beta) = \alpha \sum_{k=0}^{\infty} \beta^k R^{k+1} = \alpha(R1 + \beta R^2 1 + \beta^2 R^3 1 + \dots). \quad (5)$$





$$c(\alpha, \beta) = \alpha \sum_{k=0}^{\infty} \beta^k R^{k+1} \mathbf{1} = \alpha (R \mathbf{1} + \beta R^2 \mathbf{1} + \beta^2 R^3 \mathbf{1} + \dots). \quad (5)$$

$\downarrow$                        $\downarrow$                        $\downarrow$   
 degree    # of nodes    # of nodes  
                  two hops    Three hops  
                  away                   away

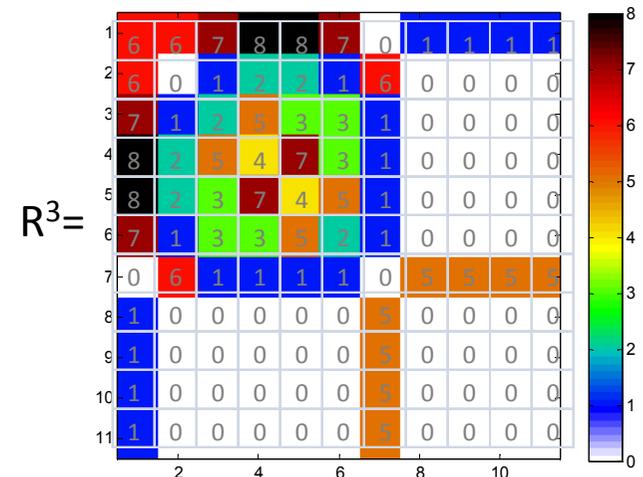
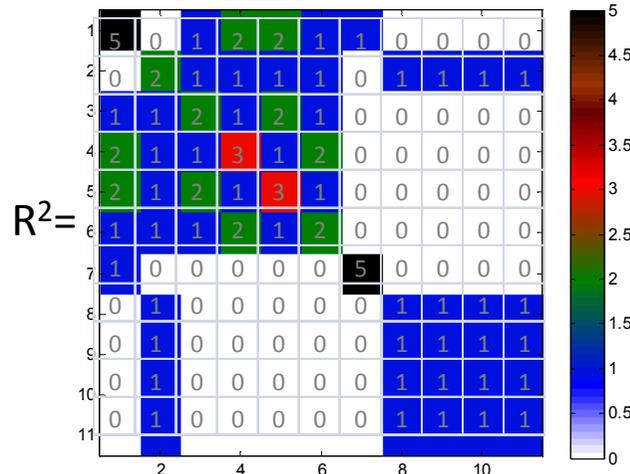
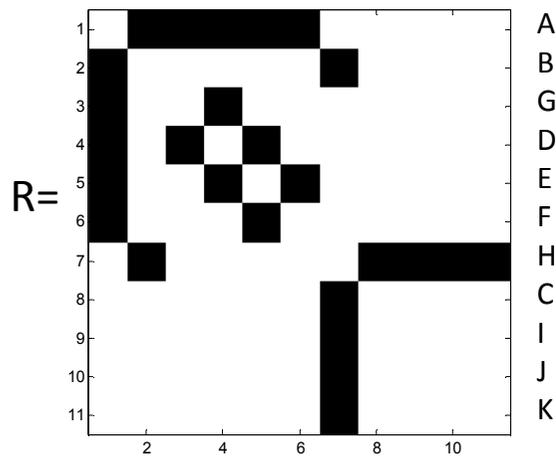
A=5	A=12	A=46
B=2	B=10	B=18
H=5	H=6	H=30
K=1	K=5	K=6
E=3	E=10	E=30

$\beta=0.5$  (after 3 iterations)

A → 22.5    K → 5  
 B → 11.5    E → 15.5  
 H → 15.5

$\beta=-0.5$  (after 3 iterations)

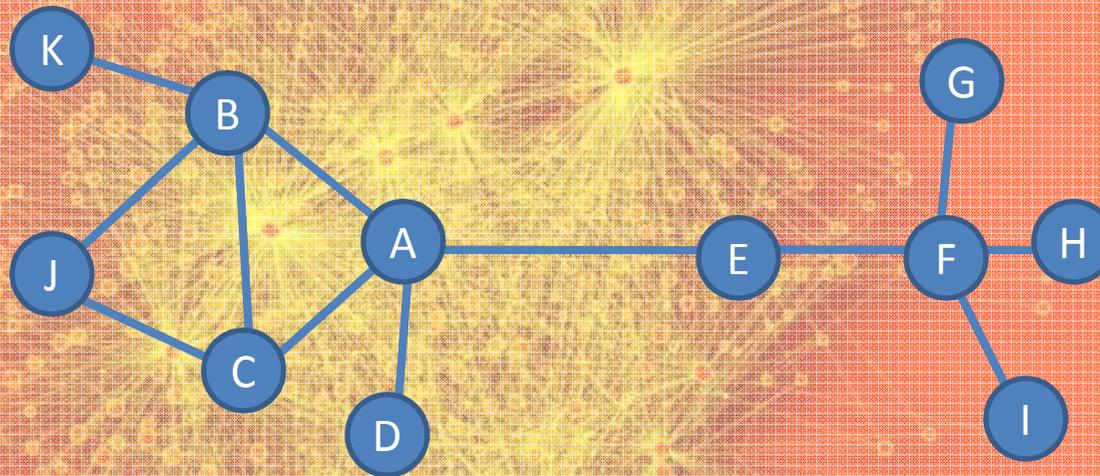
A → 10.5    K → 0  
 B → 1.5    E → 5.5  
 H → 9.5



## PAGE RANK

**PR**=Probability that a random walker with interspersed Jumps would visit that node.

**PR**=Each page votes for its neighbors.



$$PR(A) = PR(B)/4 + PR(C)/3 + PR(D) + PR(E)/2$$

A random surfer eventually stops clicking

$$PR(X) = (1-d)/N + d(\sum PR(y)/k(y))$$

## PAGE RANK

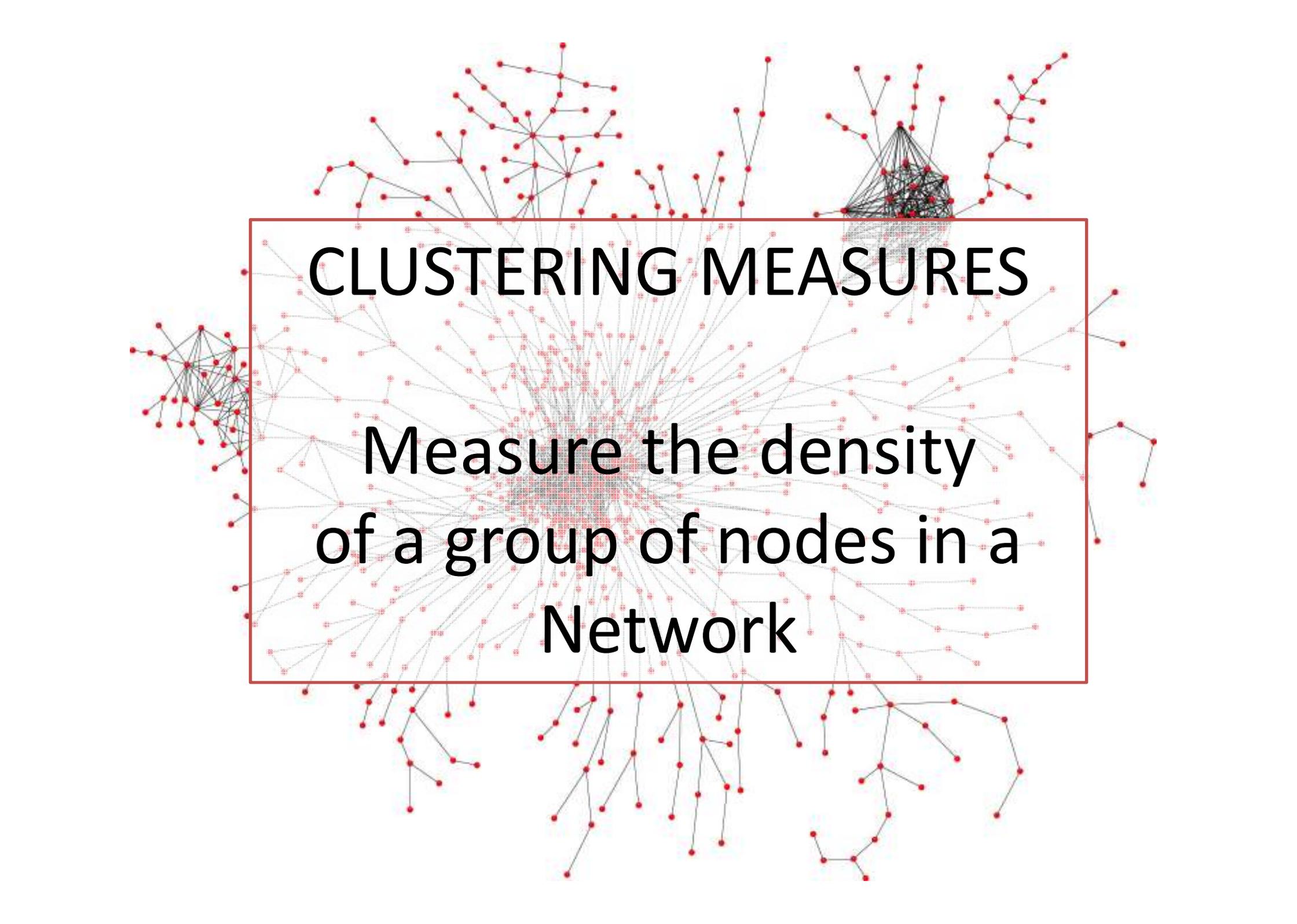
PR=Probability that a random Walker would visit that node.

PR=Each page votes for its neighbors.

$$\mathbf{R} = \begin{bmatrix} PR(p_1) \\ PR(p_2) \\ \vdots \\ PR(p_N) \end{bmatrix}$$

$$\mathbf{R} = \begin{bmatrix} (1-d)/N \\ (1-d)/N \\ \vdots \\ (1-d)/N \end{bmatrix} + d \begin{bmatrix} \ell(p_1, p_1) & \ell(p_1, p_2) & \cdots & \ell(p_1, p_N) \\ \ell(p_2, p_1) & \ddots & & \vdots \\ \vdots & & \ell(p_i, p_j) & \\ \ell(p_N, p_1) & \cdots & & \ell(p_N, p_N) \end{bmatrix} \mathbf{R}$$

$$\sum_{i=1}^N \ell(p_i, p_j) = 1,$$



# CLUSTERING MEASURES

Measure the density  
of a group of nodes in a  
Network

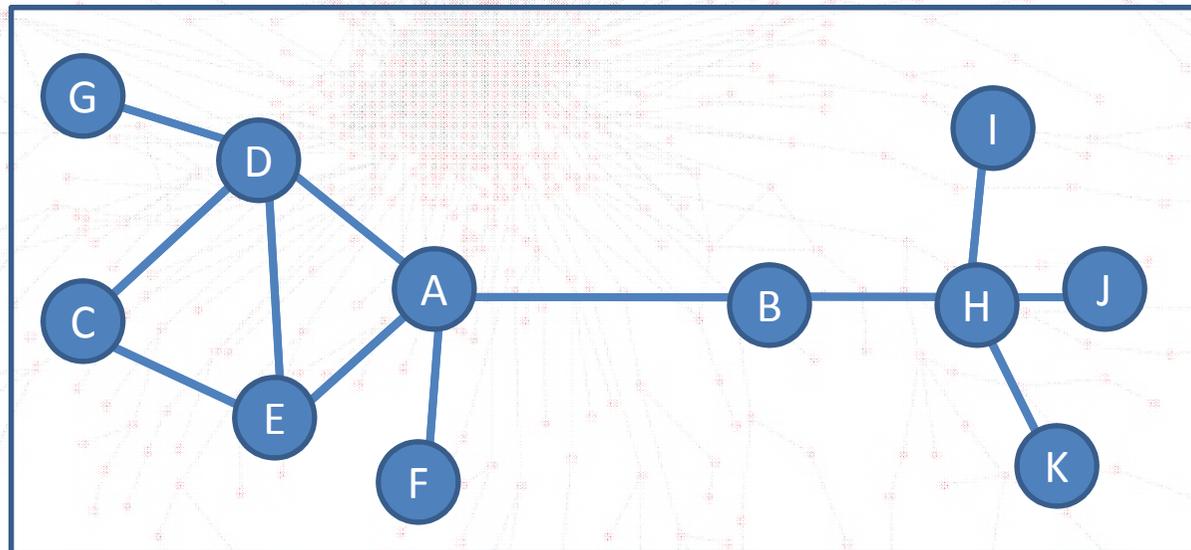
# Clustering Coefficient, Transitivity

$$C_i = 2\Delta / k(k-1)$$

$$C_A = 2/12 = 1/6$$

$$C_C = 2/2 = 1$$

$$C_E = 4/6 = 2/3$$



# Topological Overlap Mutual Clustering

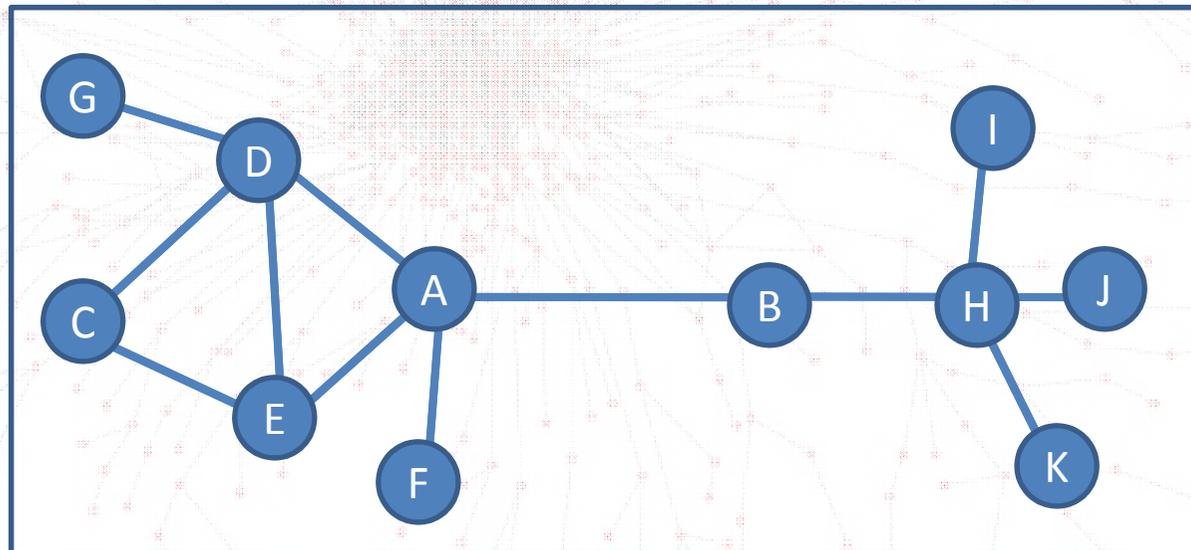
$$TO(A,B) = \text{Overlap}(A,B) / \text{NormalizingFactor}(A,B)$$

$$TO(A,B) = N(A,B) / \max(k(A), k(B))$$

$$TO(A,B) = N(A,B) / \min(k(A), k(B))$$

$$TO(A,B) = N(A,B) / (k(A) \times k(B))^{1/2}$$

$$TO(A,B) = N(A,B) / (k(A) + k(B))$$



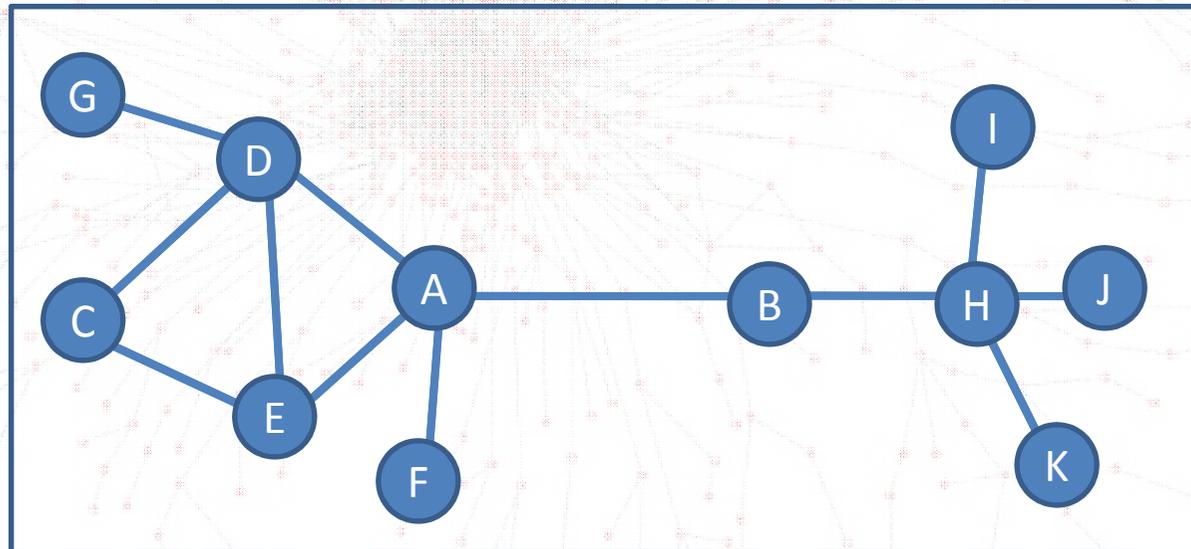
# Topological Overlap Mutual Clustering

$$TO(A,B) = N(A,B) / \max(k(A), k(B))$$

$$TO(A,B) = 0$$

$$TO(A,D) = 1/4$$

$$TO(E,D) = 2/4$$

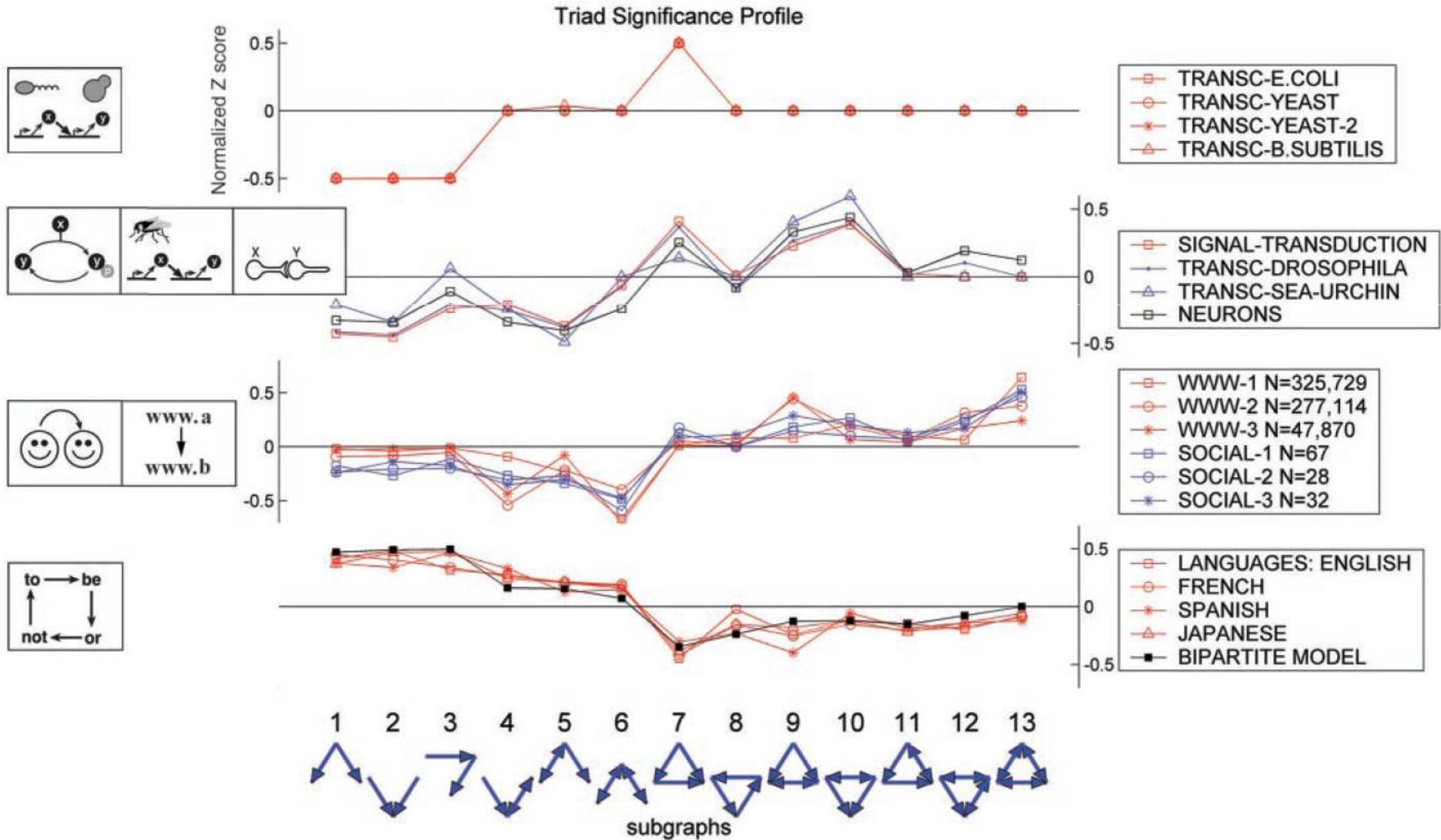


# Global Measures

The Distribution of any of the  
previously introduced measures

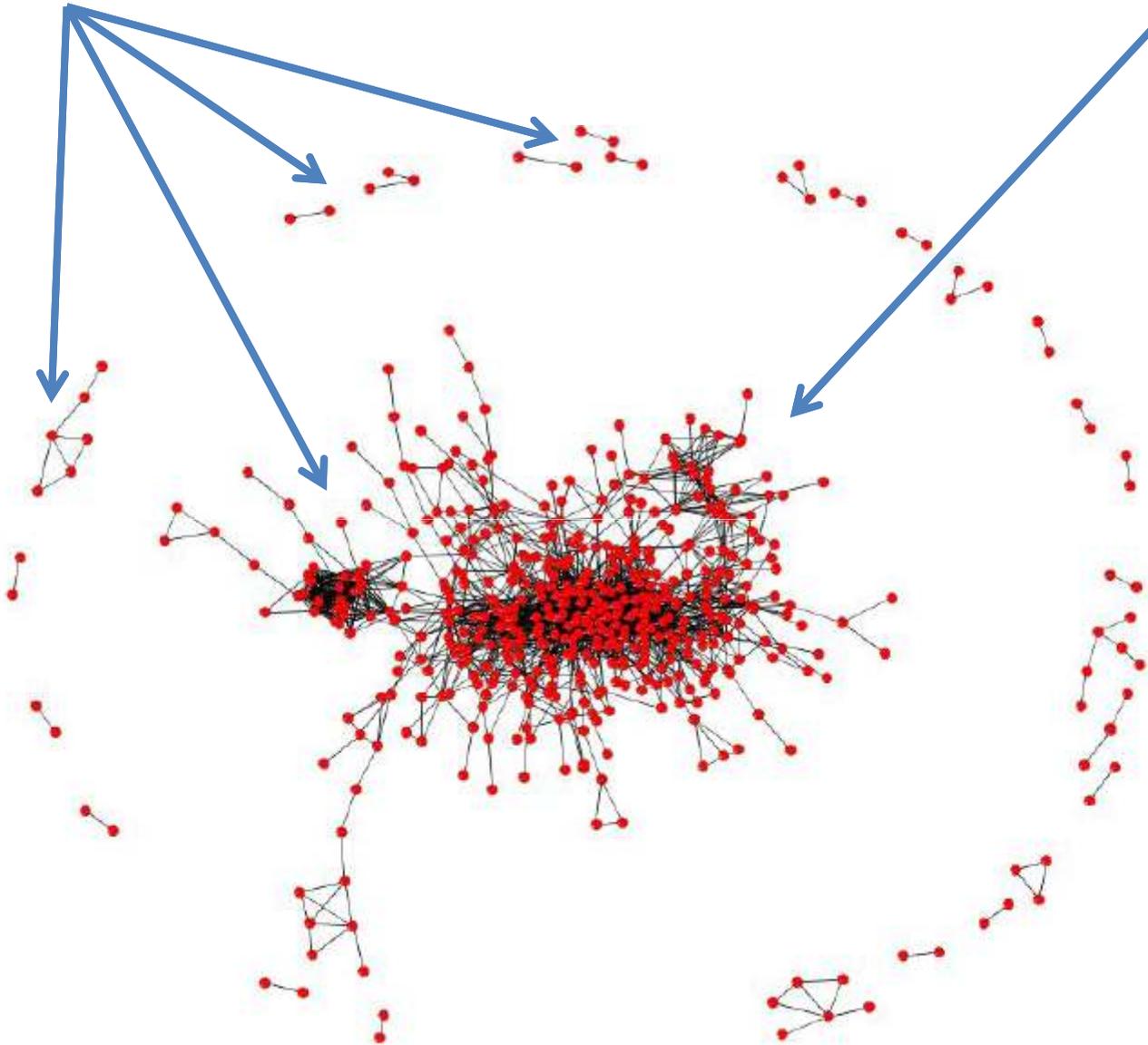
# Superfamilies of Evolved and Designed Networks

Ron Milo, Shalev Itzkovitz, Nadav Kashtan, Reuven Levitt, Shai Shen-Orr, Inbal Ayzenshtat, Michal Sheffer, Uri Alon\*



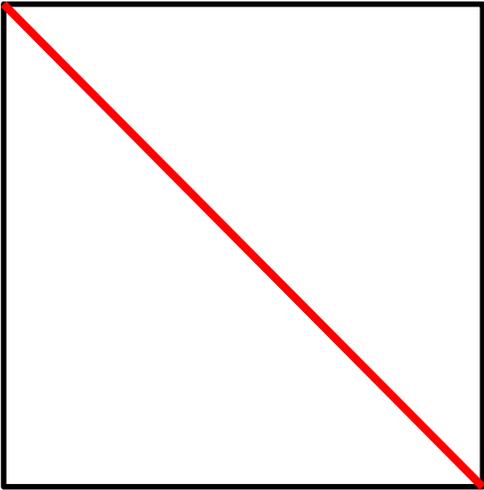
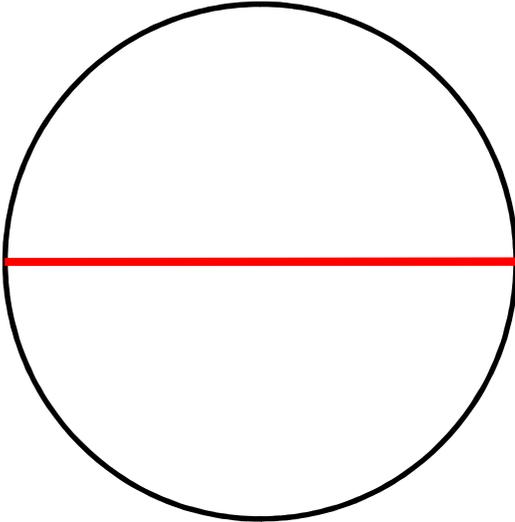
Components

Giant Component



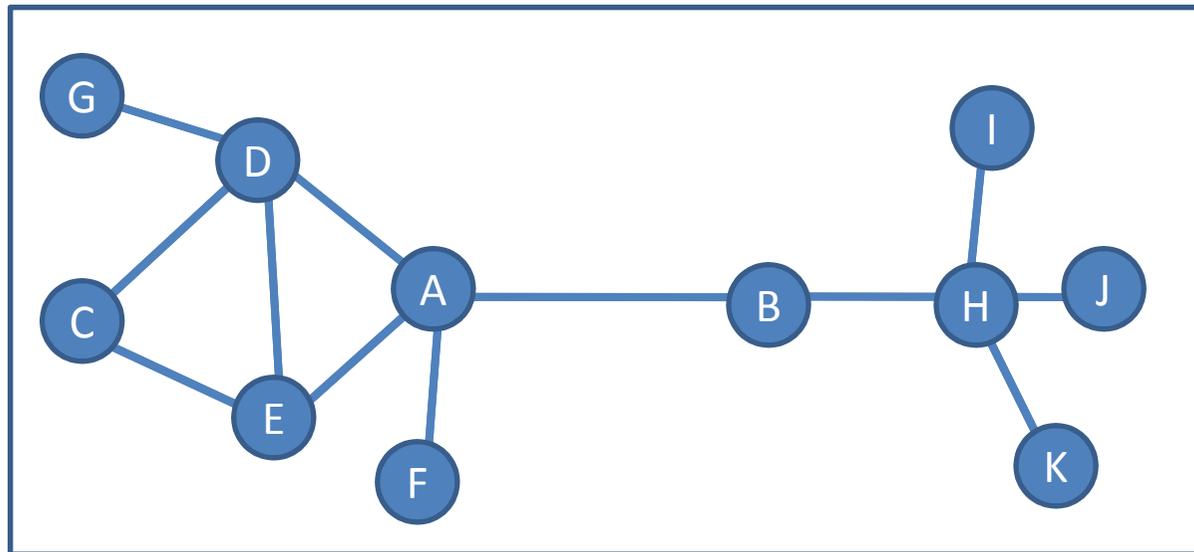
$$S = \frac{\text{NumberOfNodesInGiantComponent}}{\text{TotalNumberOfNodes}}$$

Diameter



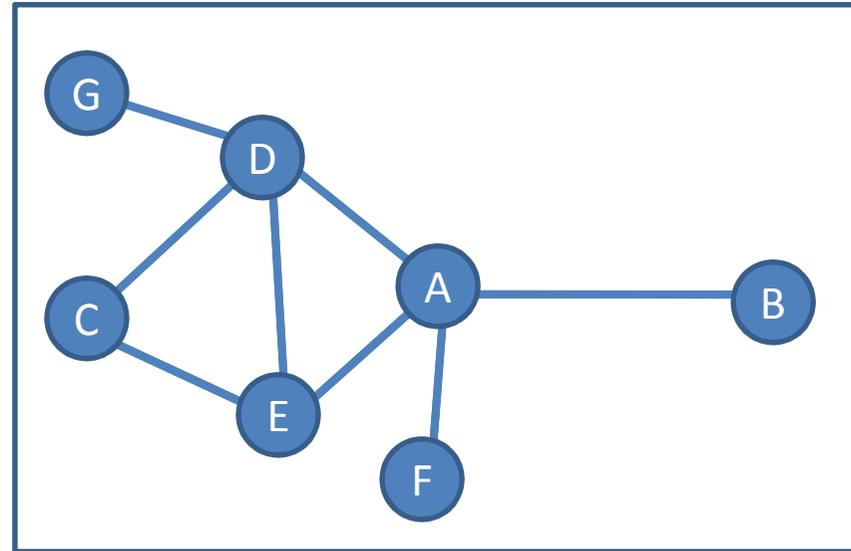
Diameter=Maximum Distance Between Elements in a Set

Diameter= $D(G,J)=D(C,J)=D(G,I)=\dots=5$



# Average Path Length

	A	B	C	D	E	F	G
A		1	2	1	1	1	2
B			3	2	2	2	3
C				1	1	3	2
D					1	2	1
E						2	2
F							3
G							



$$D(1)=8$$

$$D(2)=9$$

$$D(3)=4$$

$$L=(8+2 \times 9+3 \times 4)/(8+9+4) \quad L=1.8$$

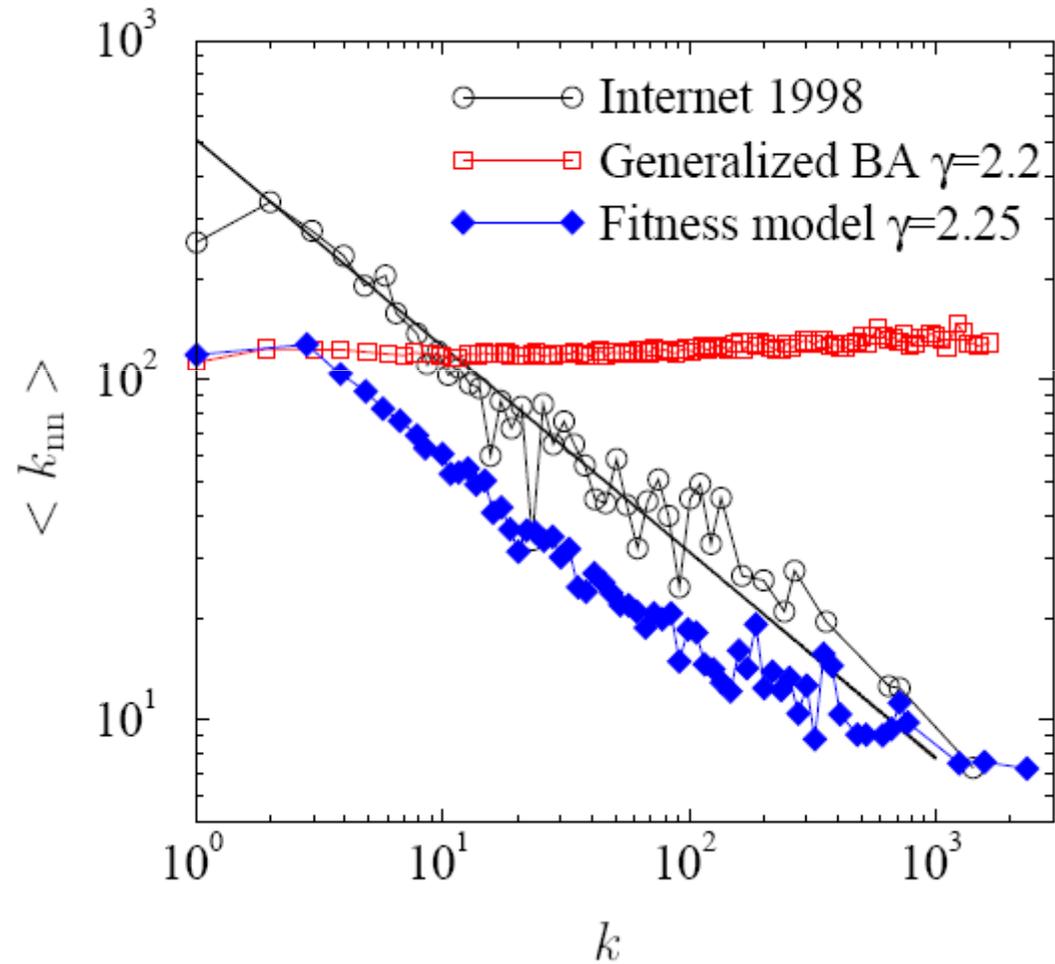
# Degree Correlations

## **Are Hubs Connected to Hubs?**

# Dynamical and correlation properties of the Internet

Romualdo Pastor-Satorras,<sup>1</sup> Alexei Vázquez,<sup>2</sup> and Alessandro Vespignani<sup>3</sup>

Phys. Rev. Lett. 87, 258701 (2001)



## Assortative Mixing in Networks

M. E. J. Newman

*Department of Physics, University of Michigan, Ann Arbor, MI 48109–1120*

*Santa Fe Institute, 1399 Hyde Park Road, Santa Fe, New Mexico 87501*

(Received 20 May 2002; published 28 October 2002)

A network is said to show assortative mixing if the nodes in the network that have many connections tend to be connected to other nodes with many connections. Here we measure mixing patterns in a variety of networks and find that social networks are mostly assortatively mixed, but that technological and biological networks tend to be disassortative. We propose a model of an assortatively mixed network, which we study both analytically and numerically. Within this model we find that networks percolate more easily if they are assortative and that they are also more robust to vertex removal.

the normalized correlation function is

$$r = \frac{1}{\sigma_q^2} \sum_{jk} jk(e_{jk} - q_j q_k), \quad (3)$$

which is simply the Pearson correlation coefficient of the degrees at either ends of an edge and lies in the range  $-1 \leq r \leq 1$ . For the practical purpose of evaluating  $r$  on

Network	$n$	$r$
Physics coauthorship (a)	52 909	0.363
Biology coauthorship (a)	1 520 251	0.127
Mathematics coauthorship (b)	253 339	0.120
Film actor collaborations (c)	449 913	0.208
Company directors (d)	7 673	0.276
Internet (e)	10 697	-0.189
World-Wide Web (f)	269 504	-0.065
Protein interactions (g)	2 115	-0.156
Neural network (h)	307	-0.163
Marine food web (i)	134	-0.247
Freshwater food web (j)	92	-0.276
Random graph (u)		0
Callaway <i>et al.</i> (v)		$\delta/(1 + 2\delta)$
Barabási and Albert (w)		0

What's a problem here?

