

# **Network Analysis**

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**CS 448B: Visualization  
Fall 2018**

## **Announcements**

# Final project

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## New visualization research or data analysis

- Pose problem, Implement creative solution
- Design studies/evaluations

## Deliverables

- Implementation of solution
- 6-8 page paper in format of conference paper submission
- Project progress presentations

## Schedule

- Project proposal: **Mon 11/5**
- Project progress presentation: **11/12 and 11/14 in class (3-4 min)**
- Final poster presentation: **12/5 Location: Lathrop 282**
- Final paper: **12/9 11:59pm**

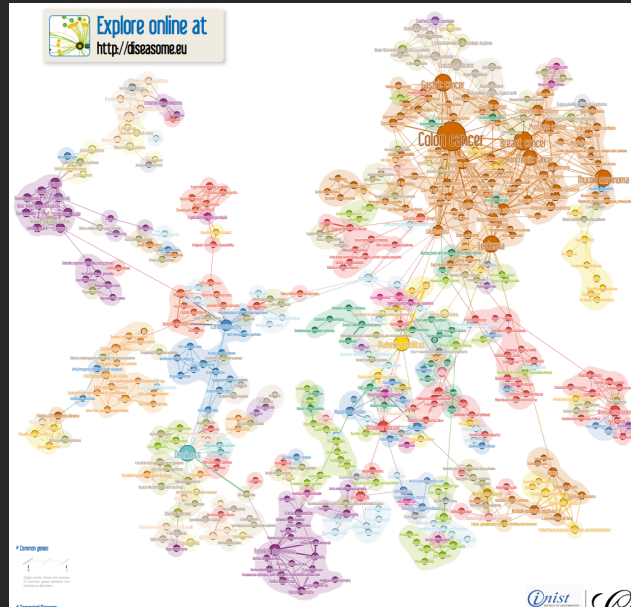
## Grading

- Groups of **up to 3 people**, graded individually
- Clearly report responsibilities of each member

# Network Analysis

\*Slides adapted from E. Adar's / L. Adamic's Network Theory and Applications course slides.

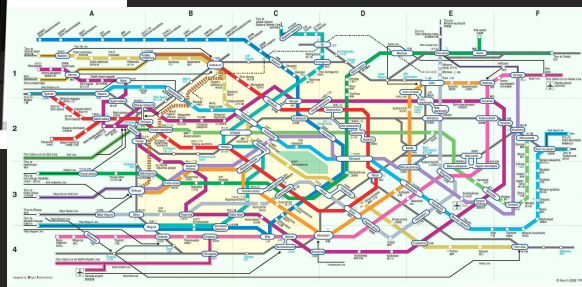
# Diseases

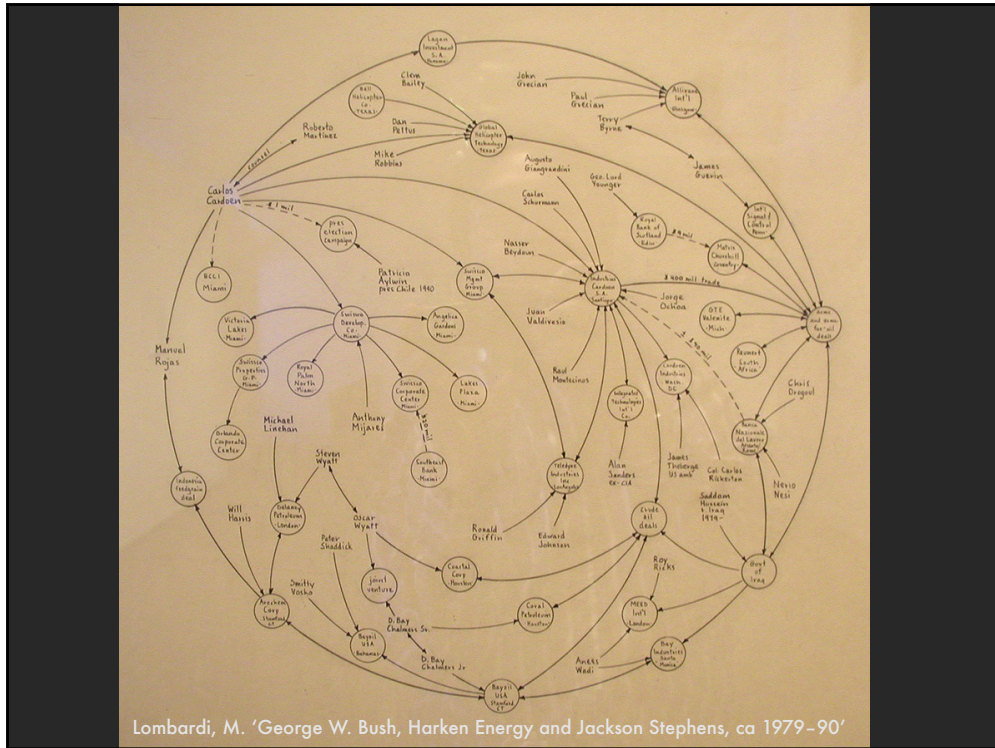


# Transportation

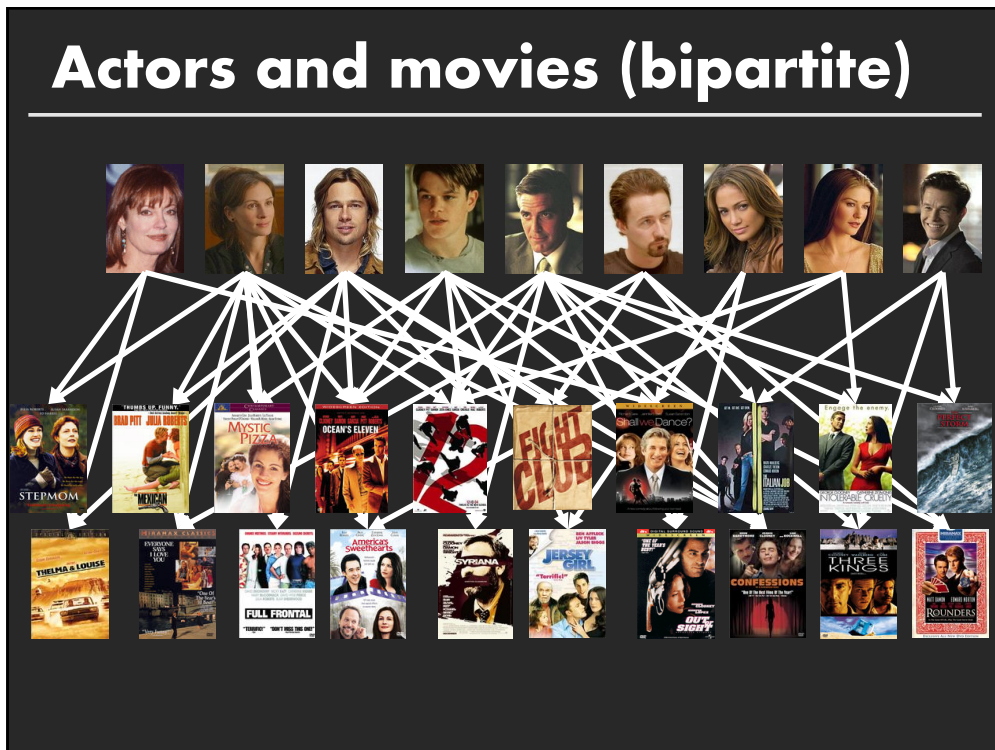


<http://www.lx97.com/maps/>








# Actors and movies (bipartite)



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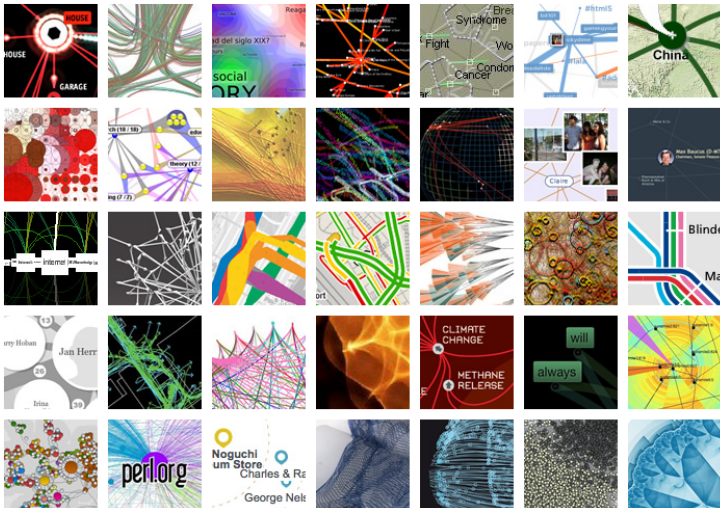
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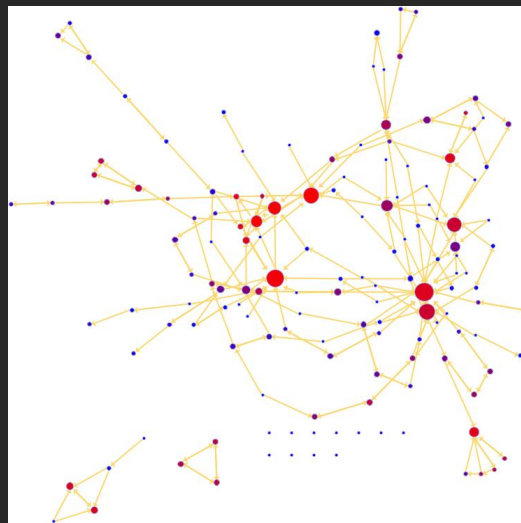
See All (714)

**VC Book** is now in progress

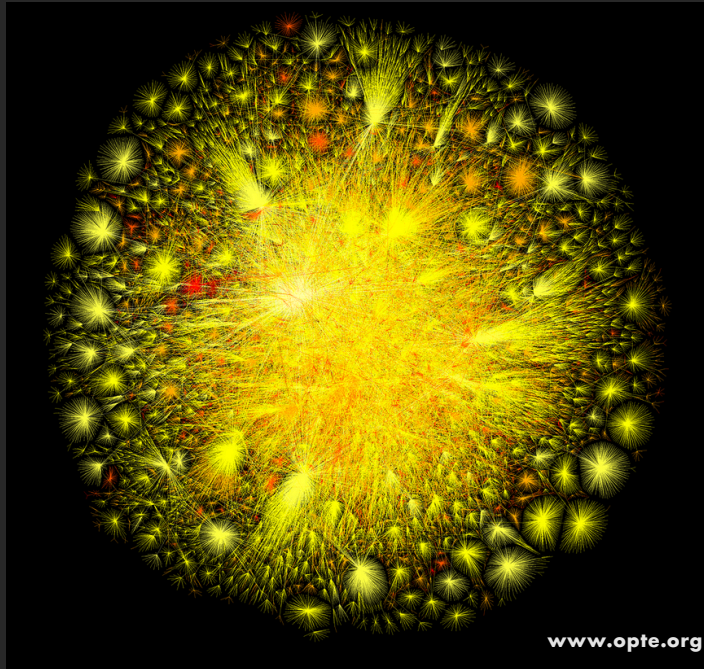


# Characterizing networks

## What does it look like?



Size?  
Density?  
Centralization?  
Clustering?  
Components?  
Cliques?  
Motifs?  
Avg. path length?  
...



## Topics

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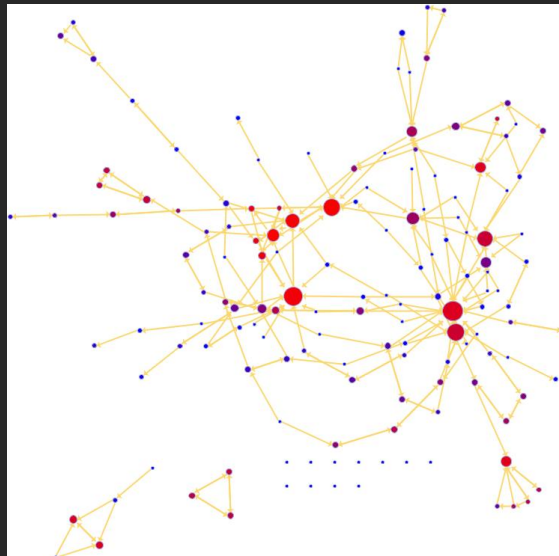
### Network Analysis

- Centrality / centralization
- Community structure
- Pattern identification
- Models

# Centrality

**How far apart are things?**

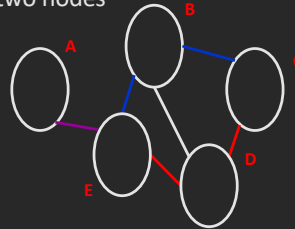
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# Distance: shortest paths

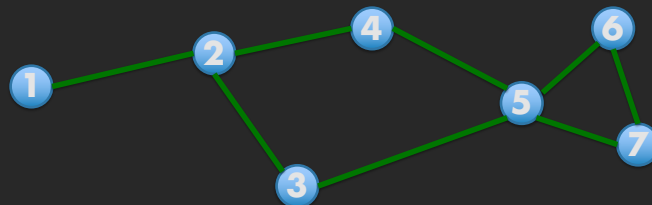
Shortest path (geodesic path)

- The shortest sequence of links connecting two nodes
- Not always unique
- A and C are connected by 2 shortest paths
  - A - E - B - C
  - A - E - D - C



# Distance: shortest paths

Shortest path from 2 to 3: 1

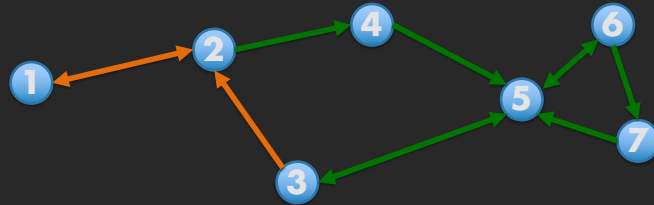


Diameter?

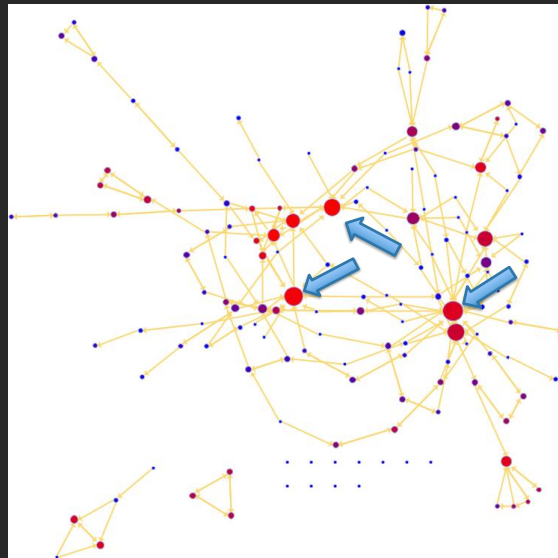


# Distance: shortest paths

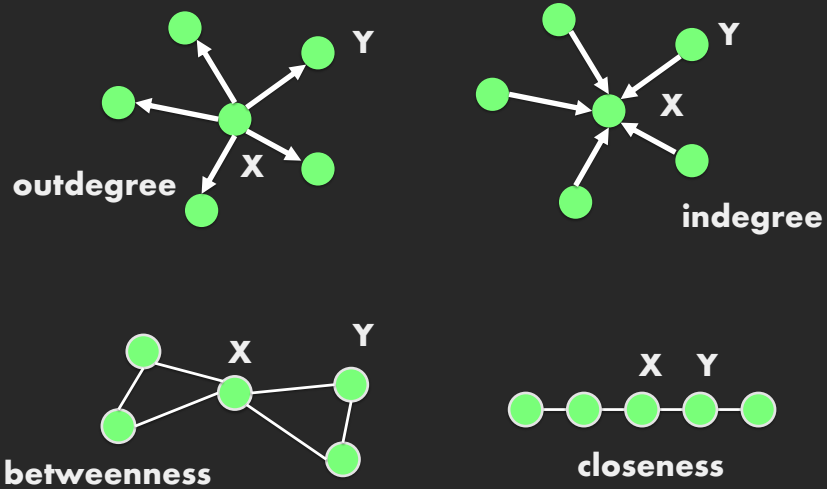
Shortest path from 2 to 3?



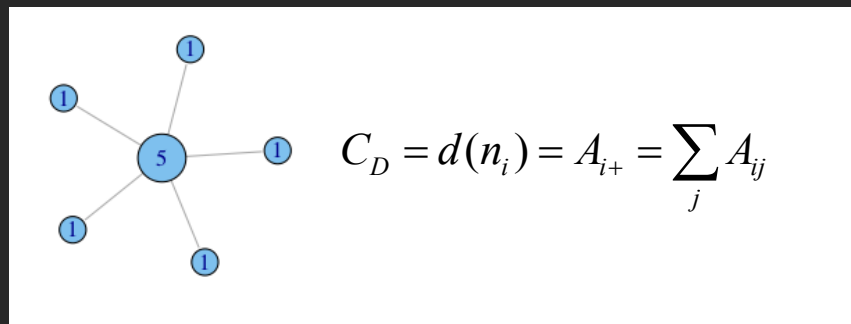
# Most important node?



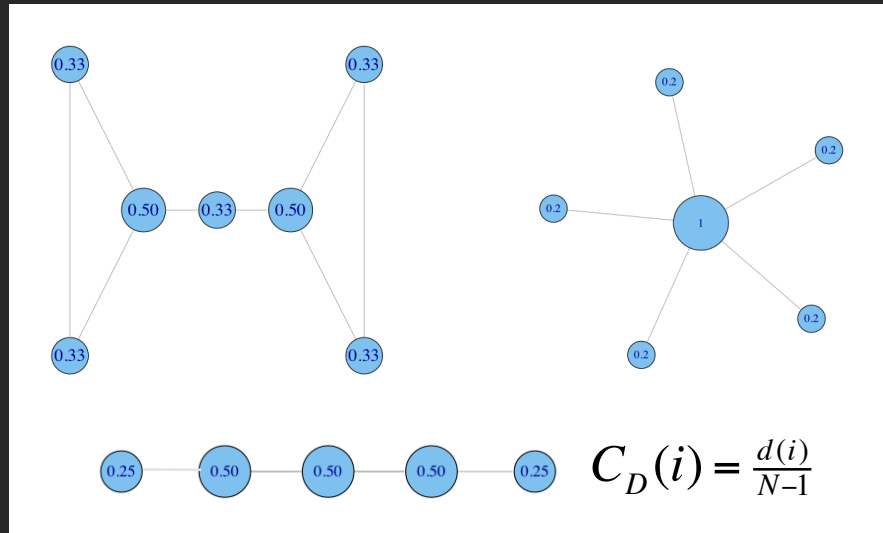
# Centrality



# Degree centrality (undirected)



## Normalized degree centrality



## When is degree not sufficient?

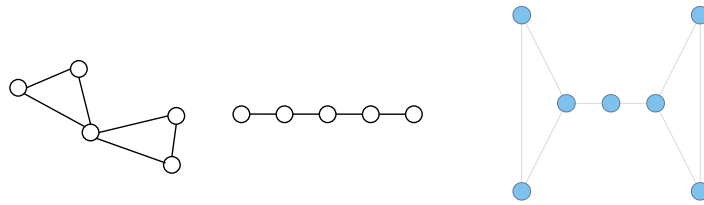
### Does not capture

Ability to broker between groups

Likelihood that information originating anywhere in the network reaches you

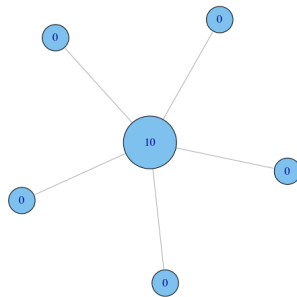
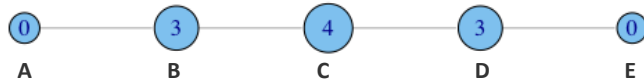
## Betweenness

Assuming nodes communicate using the most direct (shortest) route, how many pairs of nodes have to pass information through target node?



## Betweenness - examples

non-normalized:



## Betweenness: definition

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$$C_B(i) = \sum_{j,k \neq i, j < k} g_{jk}(i) / g_{jk}$$

$g_{jk}$  = the number of geodesics connecting  $jk$

$g_{jk}(i)$  = the number that node  $i$  is on.

Normalization:

$$C'_B(i) = C_B(i) / [(n-1)(n-2)/2]$$

number of pairs of vertices excluding the vertex itself

## When are $C_d$ , $C_b$ not sufficient?

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### Do not capture

Likelihood that information originating anywhere in the network reaches you

# Closeness: definition

Being close to the center of the graph

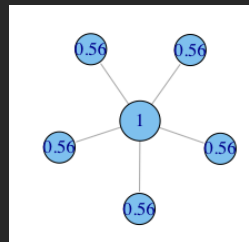
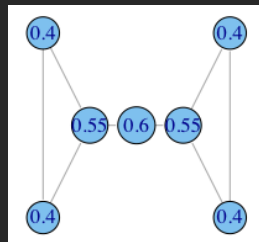
Closeness Centrality:

$$C_c(i) = \left[ \sum_{j=1, j \neq i}^N d(i, j) \right]^{-1}$$

Normalized Closeness Centrality

$$C'_c(i) = (C_c(i)) / (N - 1) = \frac{N - 1}{\sum_{j=1, j \neq i}^N d(i, j)}$$

# Examples - closeness



## Centrality in directed networks

Prestige  $\sim$  indegree centrality

Betweenness  $\sim$  consider directed shortest paths

Closeness  $\sim$  consider nodes from which target node can be reached

Influence range  $\sim$  nodes reachable from target node

Straight-forward modifications to equations for non-directed graphs



## Characterizing nodes

\* when they are not, there is likely something interesting about the network

\* suggest possible topologies and node positions to fit each square

	<b>Low Degree</b>	<b>Low Closeness</b>	<b>Low Betweenness</b>
<b>High Degree</b>		Node embedded in cluster that is far from the rest of the network	Node's connections are redundant - communication bypasses him/her
<b>High Closeness</b>	Node links to a small number of important/active other nodes.		Many paths likely to be in network; node is near many people, but so are many others
<b>High Betweenness</b>	Node's few ties are crucial for network flow	Rare. Node monopolizes the ties from a small number of people to many others.	

# Centralization – how equal

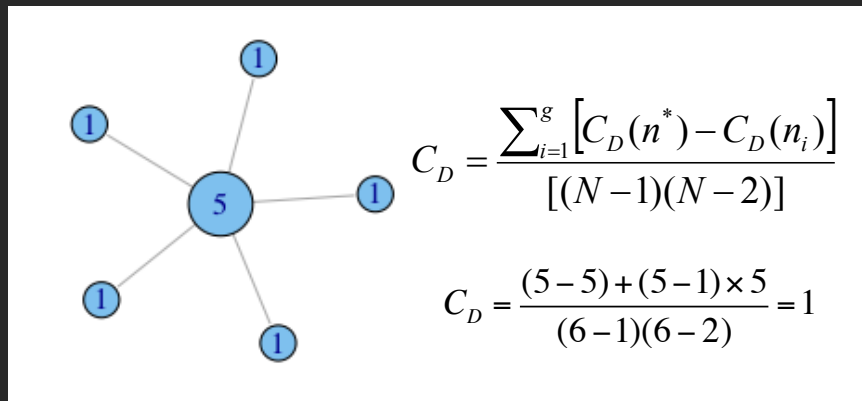
Variation in the centrality scores among the nodes

Freeman's general formula for centralization:

$$C_D = \frac{\sum_{i=1}^g [C_D(n^*) - C_D(i)]}{[(N-1)(N-2)]}$$

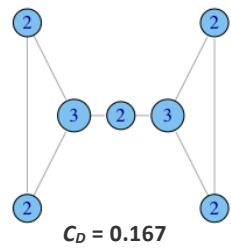
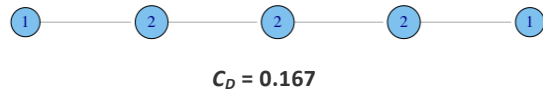
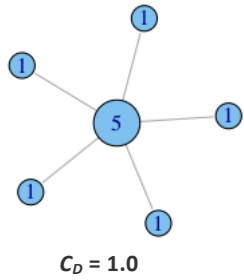
maximum value in the network

## Examples

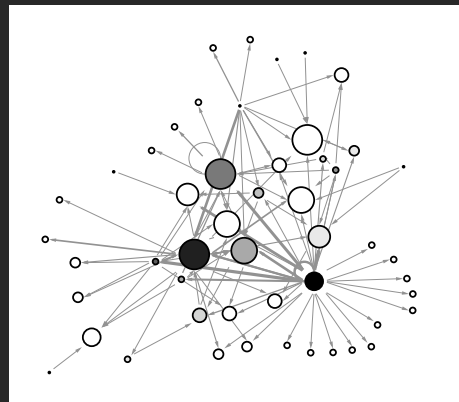
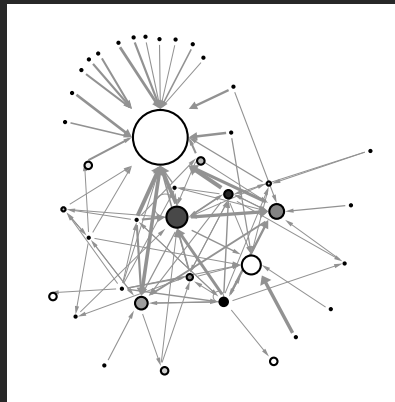




# Examples



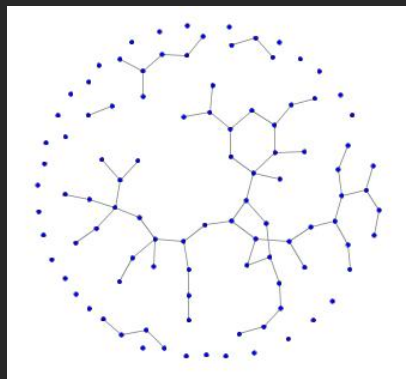
# Financial networks



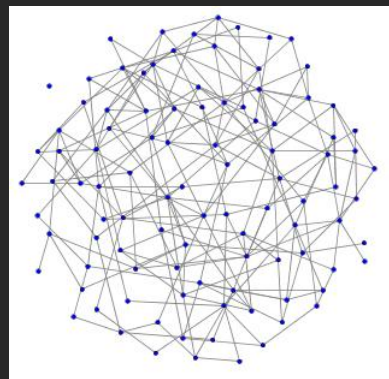
# Community Structure

## How dense is it?

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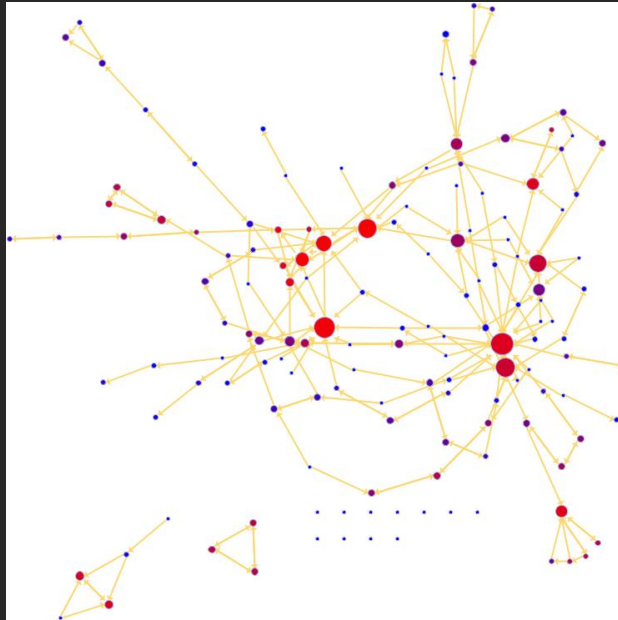
$$\text{density} = e / e_{\max}$$



Max. possible edges:

- Directed:  $e_{\max} = n * (n-1)$
- Undirected:  $e_{\max} = n * (n-1) / 2$

# Is everything connected?

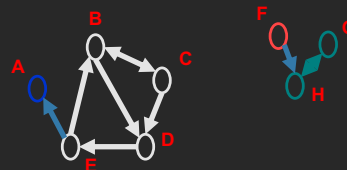


## Connected Components - Directed

### Strongly connected components

- Each node in component can be reached from every other node in component by following directed links

- BCDE
- A
- GH
- F



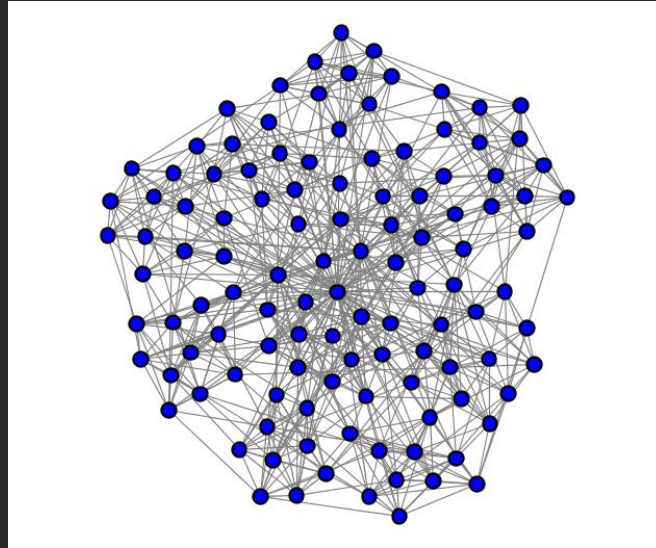
### Weakly connected components

- Each node can be reached from every other node by following links in either direction

- ABCDE
- GHF

## Community finding (clustering)

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## Hierarchical clustering

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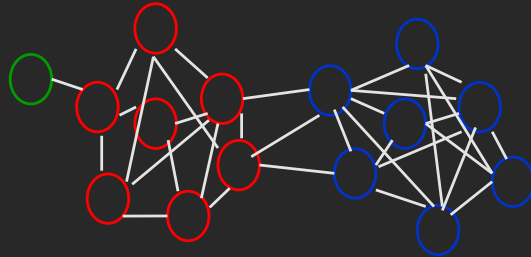
Process:

- Calculate affinity weights  $W$  for all pairs of vertices
- Start:  $N$  disconnected vertices
- Adding edges (one by one) between pairs of clusters in order of decreasing weight (use closest distance to compare clusters)
- Result: nested components



## Hierarchical clustering (path counts)

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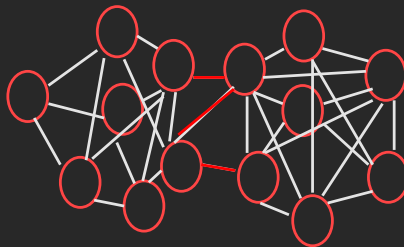


## Betweenness clustering

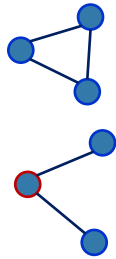
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Girvan and Newman 2002 iterative algorithm:

- Compute  $C_b$  of all edges
- Remove edge  $i$  where  $C_b(i) == \max(C_b)$
- Recalculate betweenness

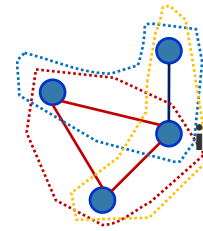


# Clustering coefficient



Local clustering coefficient:

$$C_i = \frac{\text{number of closed triplets centered on } i}{\text{number of connected triplets centered on } i}$$



Global clustering coefficient:

$$C_G = \frac{3 * \text{number of closed triplets}}{\text{number of connected triplets}}$$

$$C_i = 1/3 = 0.33$$

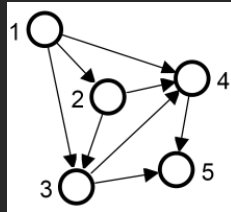
$$C_G = 3 * 1/5 = 0.6$$

# Pattern finding - motifs

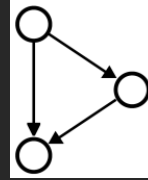
Define / search for a particular structure, e.g. complete triads



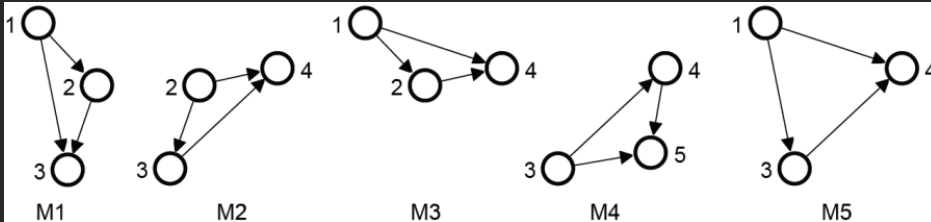
# Motifs can overlap in the network



graph



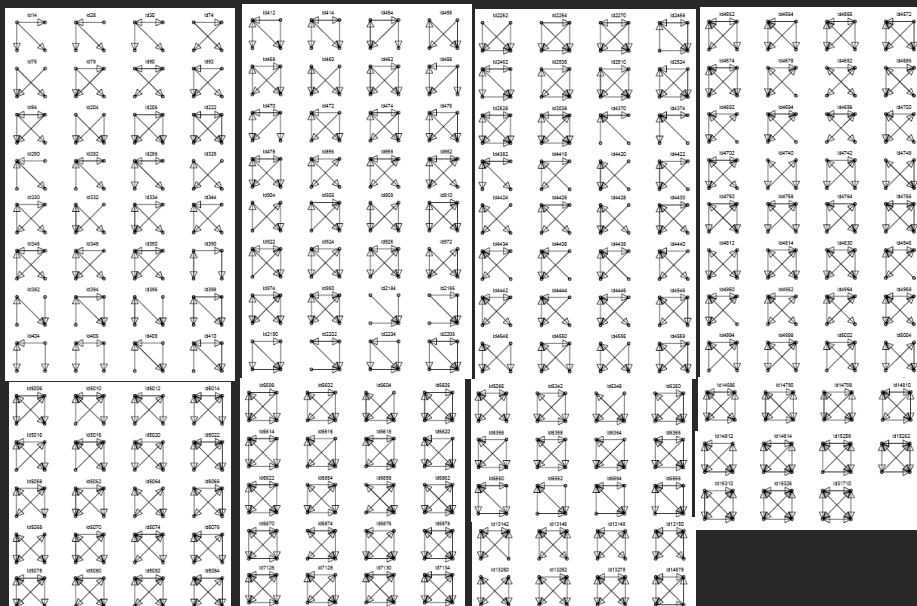
motif to be found



motif matches

[http://mavisto.ipk-gatersleben.de/frequency\\_concepts.html](http://mavisto.ipk-gatersleben.de/frequency_concepts.html)

## 4 node subgraphs

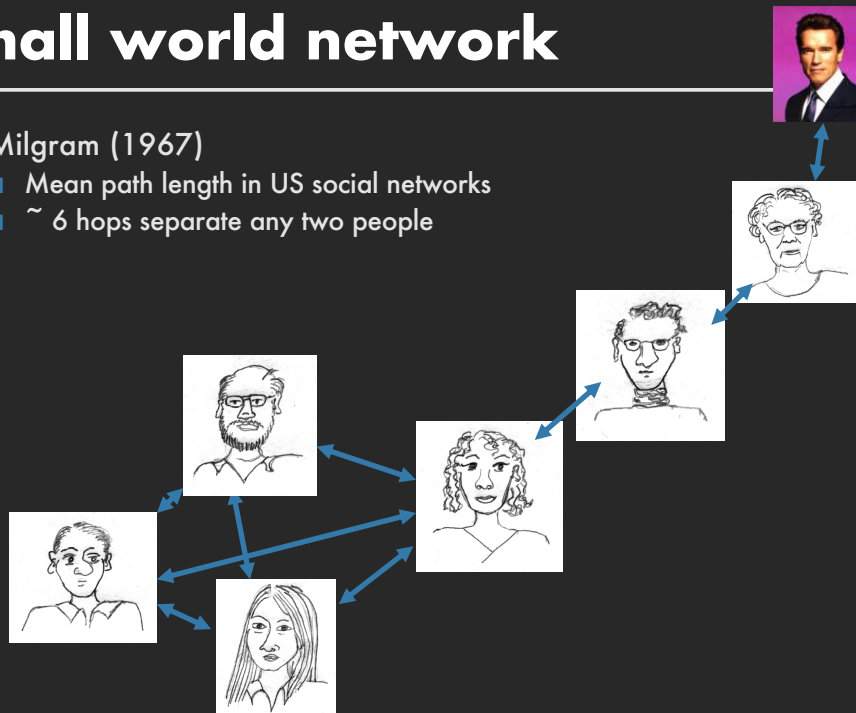


# Simulating network models

## Small world network

Milgram (1967)

- Mean path length in US social networks
- $\sim 6$  hops separate any two people

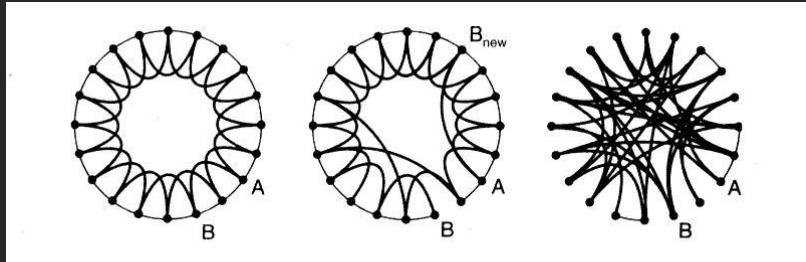




# Small world networks

Watts and Strogatz 1998

- a few random links in an otherwise structured graph make the network a small world



**regular lattice:**  
my friend's friend is  
always my friend

**small world:**  
mostly structured  
with a few random  
connections

**random graph:**  
all connections  
random

## Defining small world phenomenon

Pattern:

- high clustering
- low mean shortest path

$$C_{\text{network}} \gg C_{\text{random graph}}$$

$$l_{\text{network}} \approx \ln(N)$$

Examples

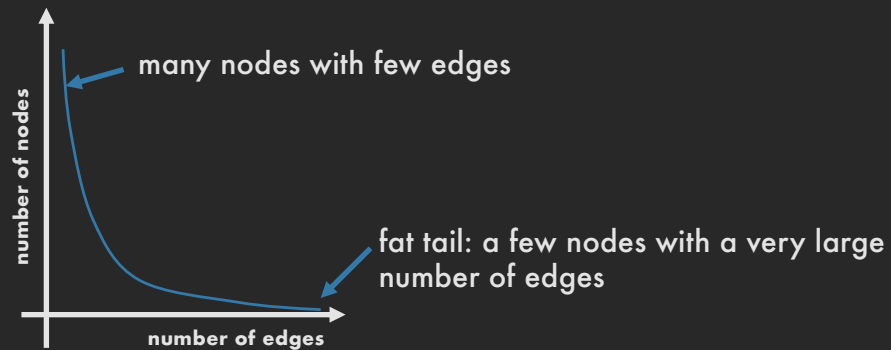
- neural network of *C. elegans*,
- semantic networks of languages,
- actor collaboration graph
- food webs

# Power law networks

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Many real world networks contain hubs: highly connected nodes

Usually the distribution of edges is extremely skewed



# Summary

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Structural analysis

- Centrality
- Community structure
- Pattern finding

→ Widely applicable across domains