



**EECS6414:**

**Data Analytics & Visualization**

# Information Networks

# COMMUNICATIONS OF THE ACM

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OF THE

09/2021 VOL.64 NO.09

?

Issue of  
Sep 2021

## The Future Is Big Graphs

Managing IT Professional Turnover

An Internet of Things Service Roadmap

Whose Smartphone Is It?

Q&A with ACM Computing Prize Winner David Silver



Graphs

**what is a network or  
a graph?**



**networks are  
ubiquitous**



**World economy**



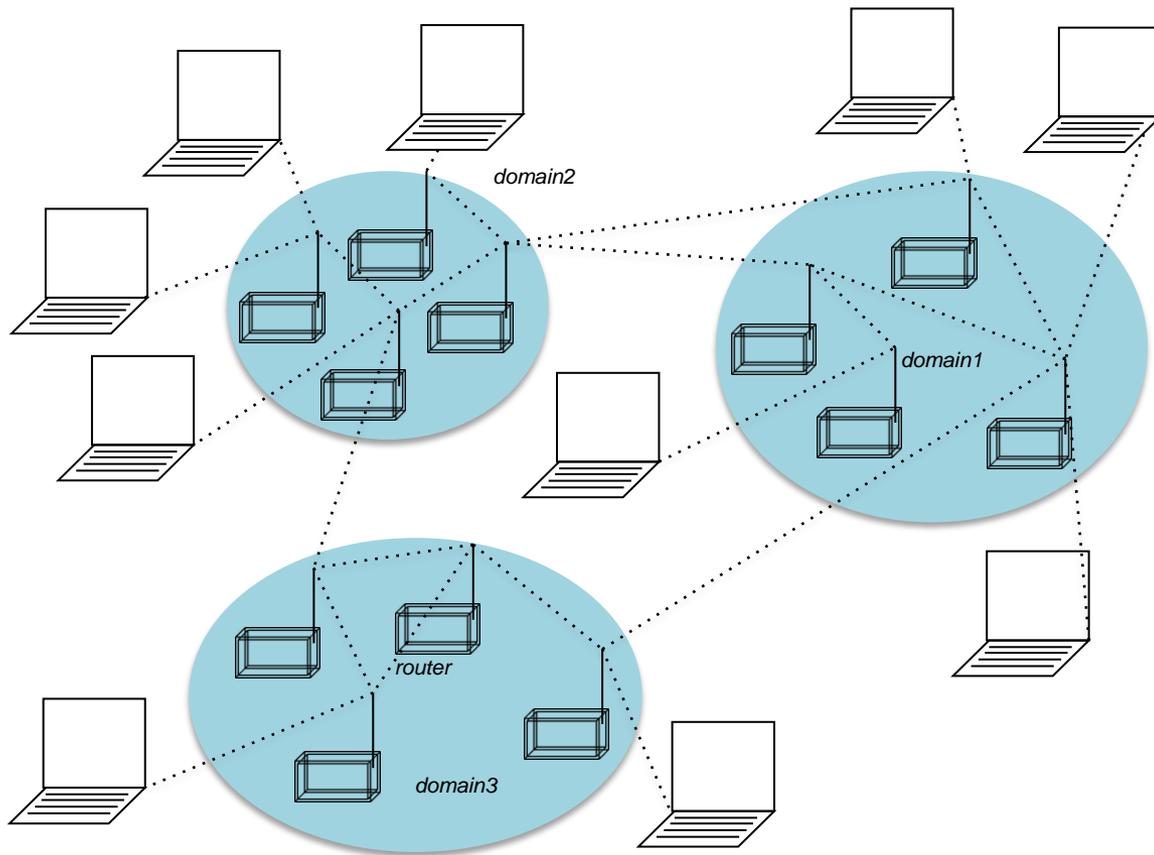
**Human cell**



# Railroads



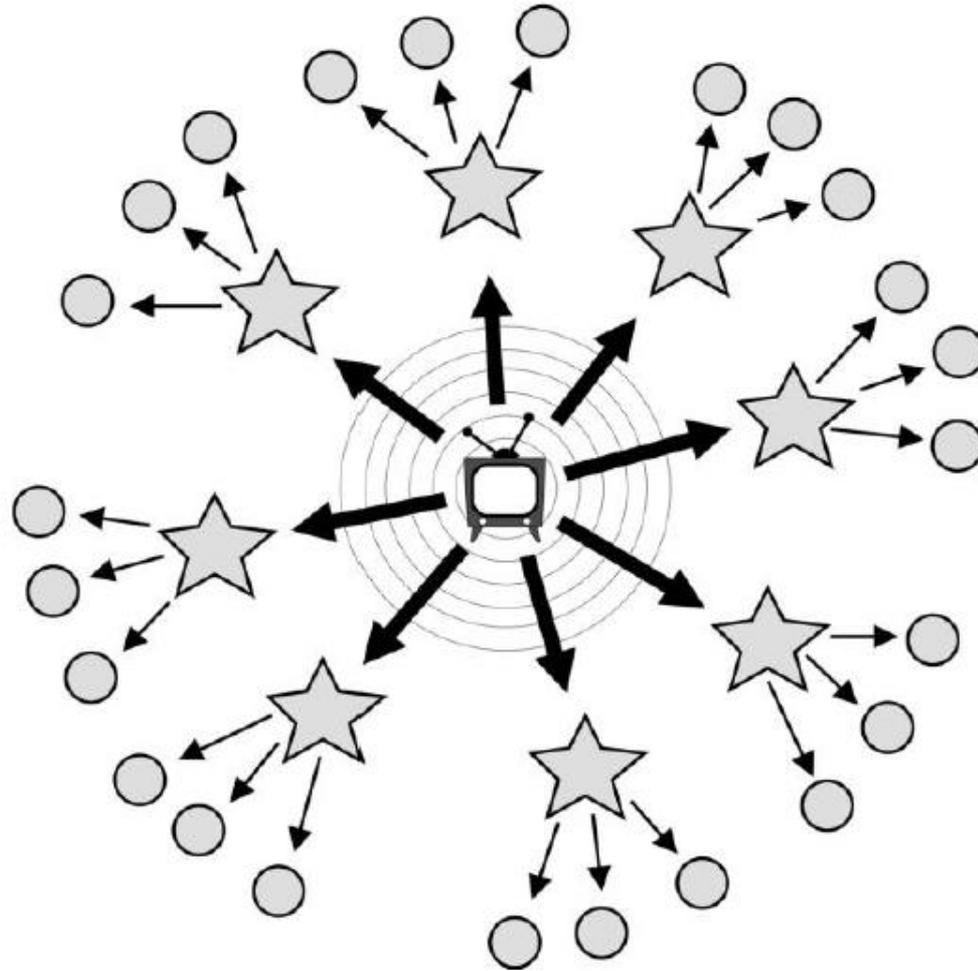
**Brain**



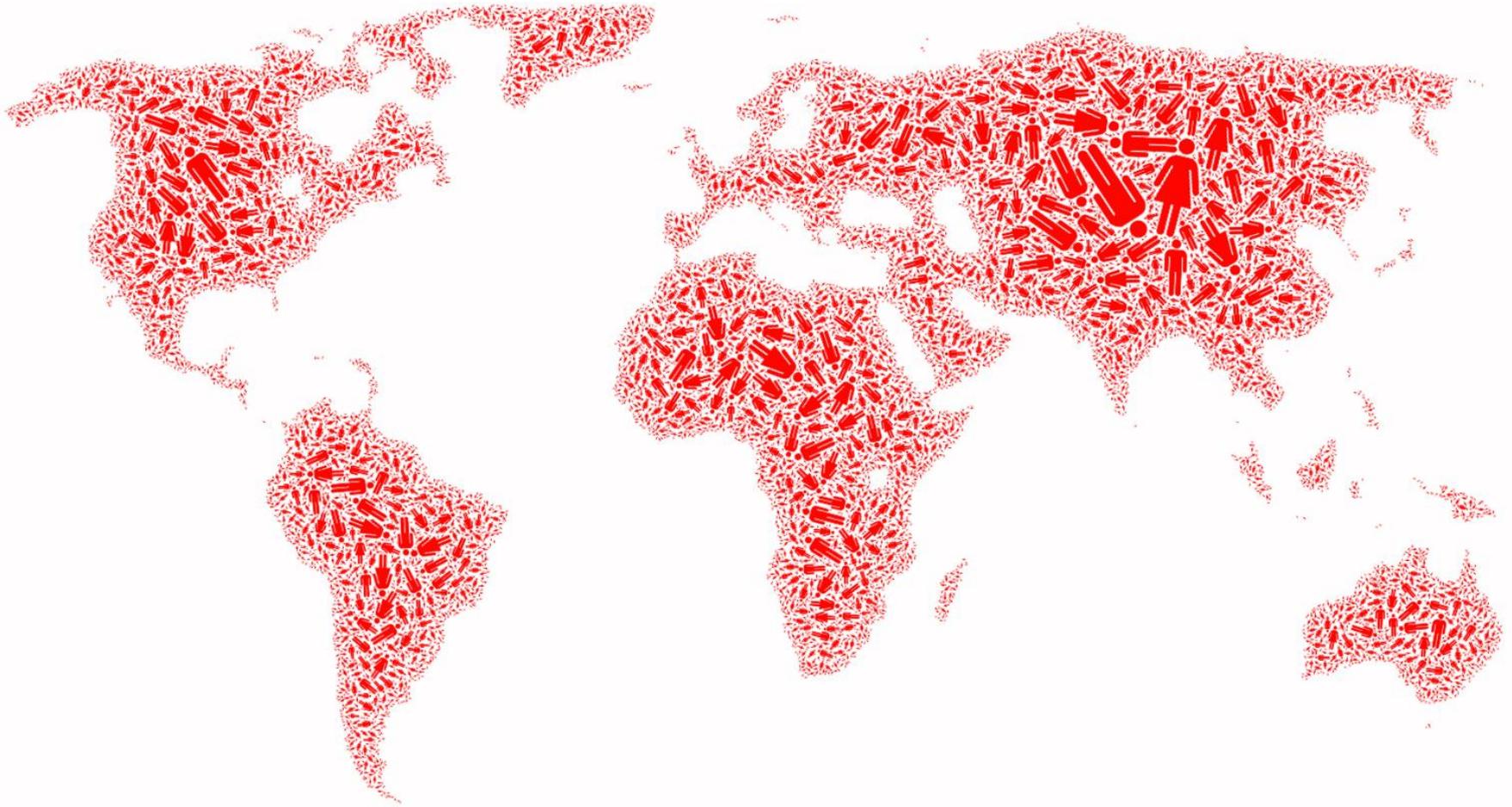
# Internet



**Friends & Family**

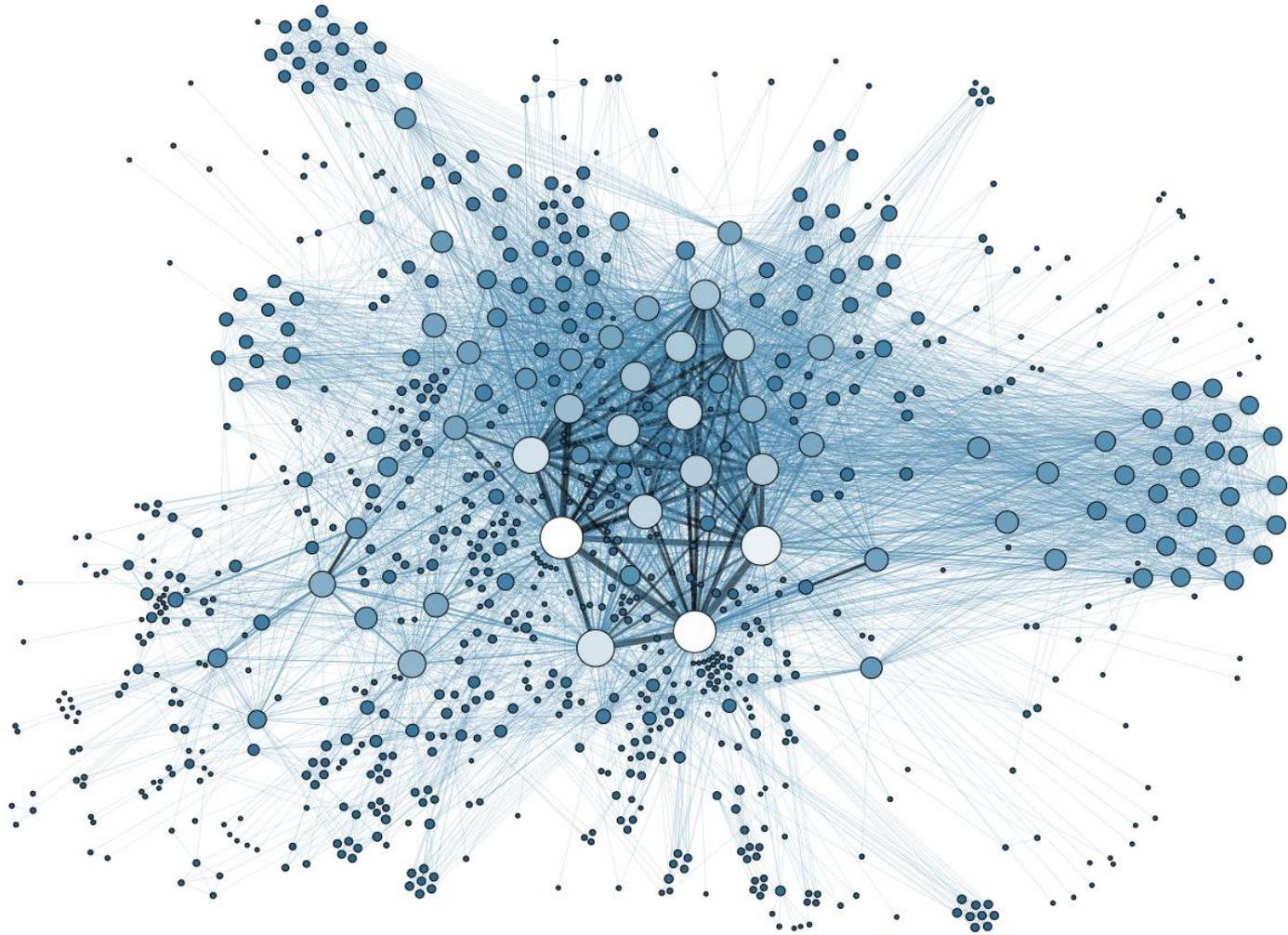


# Media & Information



**Society**

**What do the  
following things  
have in common?**



**Complex systems that can be modeled as  
Networks!**

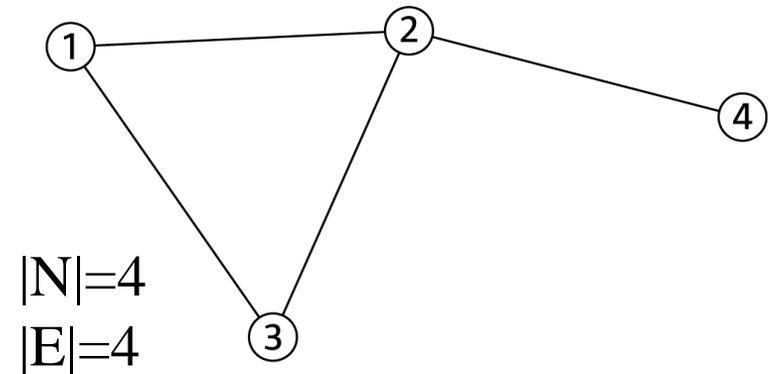
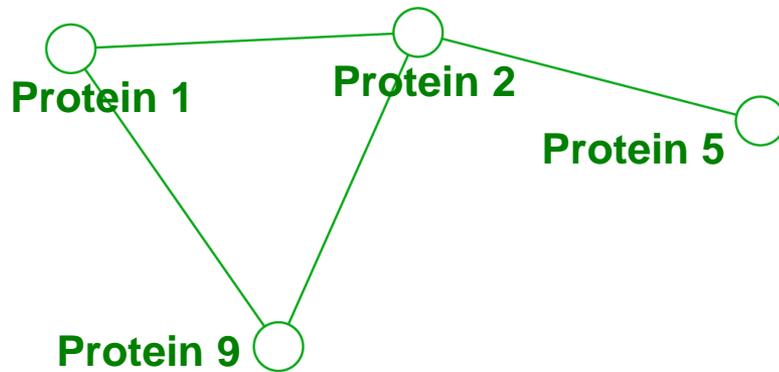
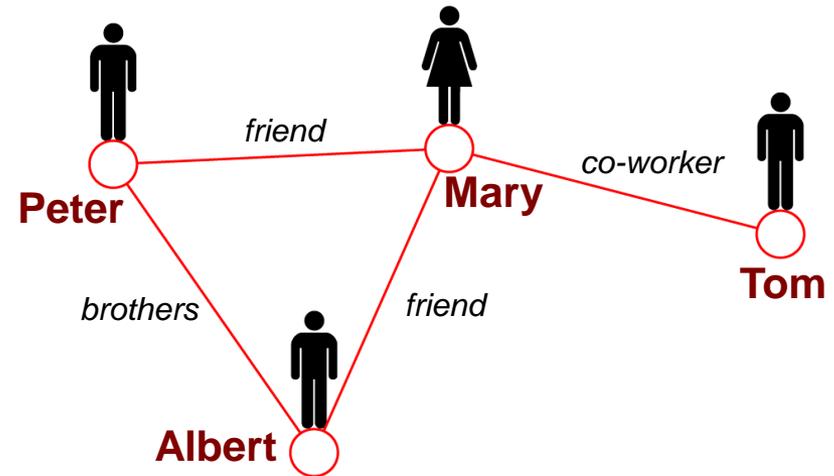
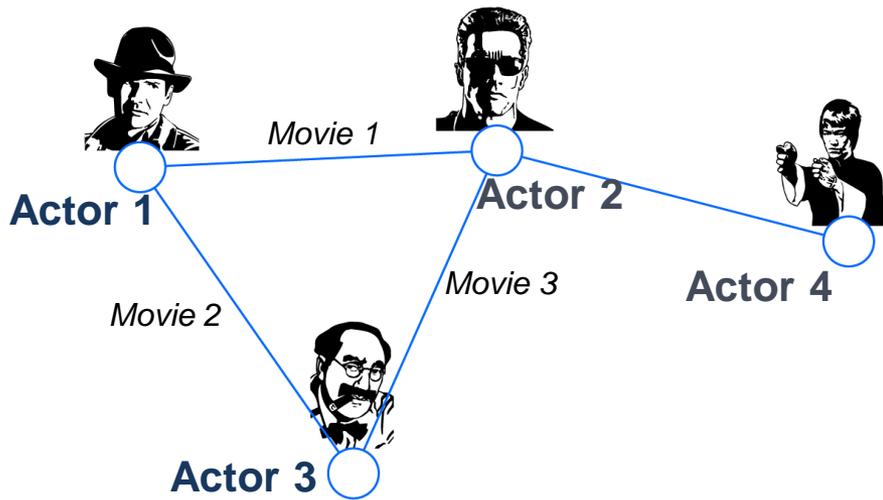
# Networks!

Behind many systems there is an intricate wiring diagram, **a network**, that defines the **interactions** between the components

**We will never understand these systems unless we understand the networks behind them!**

**But, why should we  
care about networks?  
Why now?**

# Why Networks?



Universal language for describing complex data

# Networks: Why Now?



Age and size of networks

CS!!

# Networks: Size Matters

- **Network data: Orders of magnitude**
  - **436-node** network of email exchange at a corporate research lab [Adamic-Adar, SocNets '03]
  - **43,553-node** network of email exchange at an university [Kossinets-Watts, Science '06]
  - **4.4-million-node** network of declared friendships on a blogging community [Liben-Nowell et al., PNAS '05]
  - **240-million-node** network of communication on Microsoft Messenger [Leskovec-Horvitz, WWW '08]
  - **800-million-node** Facebook network [Backstrom et al. '11]

**How can we *study*  
networks?**

# Network Analysis

**network analysis** helps to reveal the *underlying dynamics* of these systems, not easily observable before

**what do we study in  
networks?**

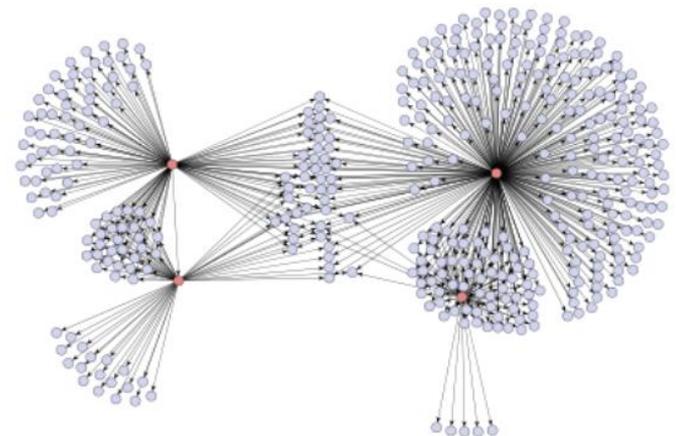
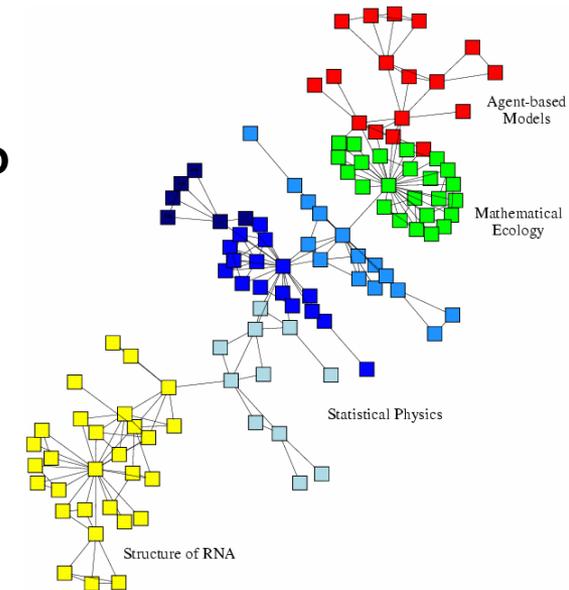
# Networks: Structure & Process

## ■ Structure and evolution

- What is the structure of a network?
- Why and how did it become to have such structure?

## ■ Processes and dynamics

- Networks provide “skeleton” for spreading of information, behavior, diseases



**how do we reason  
about networks?**

# Reasoning About Networks

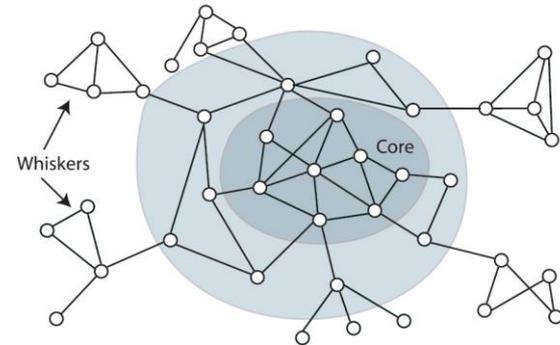
- **Empirical studies/properties:** Study network data to find organizational principles
- **Mathematical models:** Probabilistic, graph theory
- **Algorithms:** Methods for analyzing graphs

# Properties

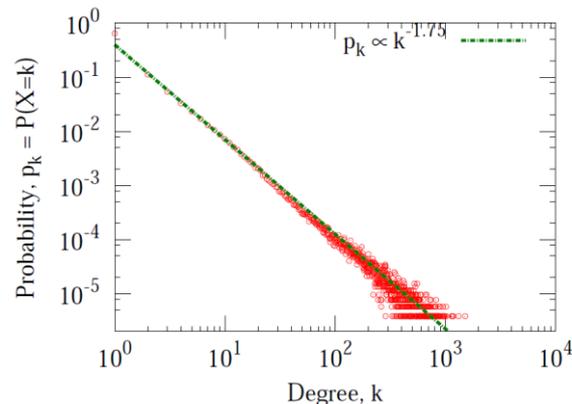
- Six degrees of separ.



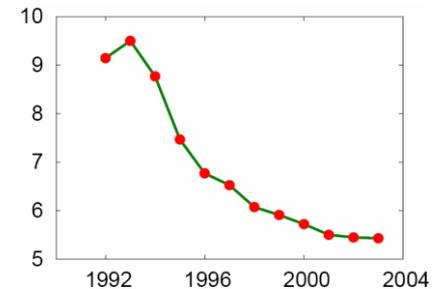
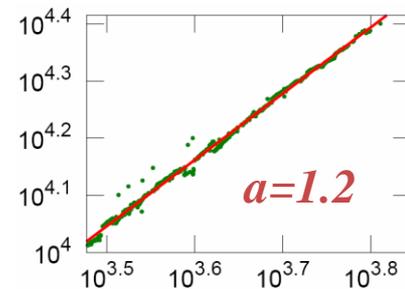
- Strength of weak ties



- Power-law degrees

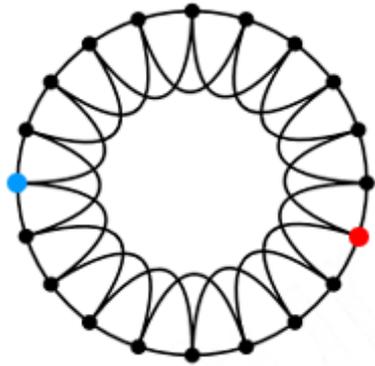


- Densif. power law,  
Shrinking diameter

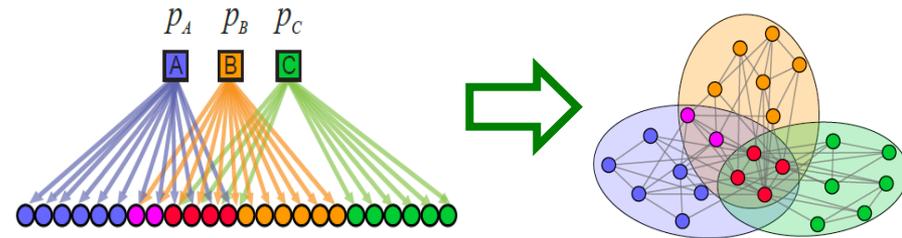


# Models

## ■ Erdős-Renyi model



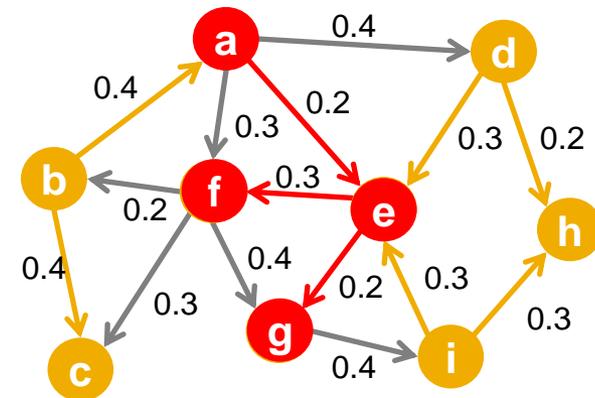
## ■ Community model



## ■ Small-world model

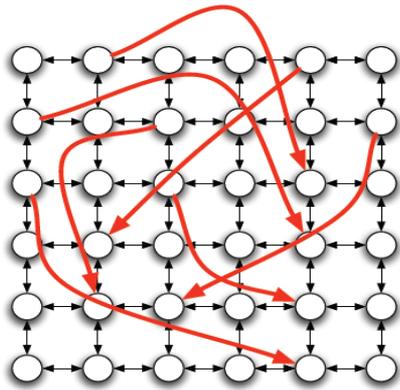


## ■ Cascade model

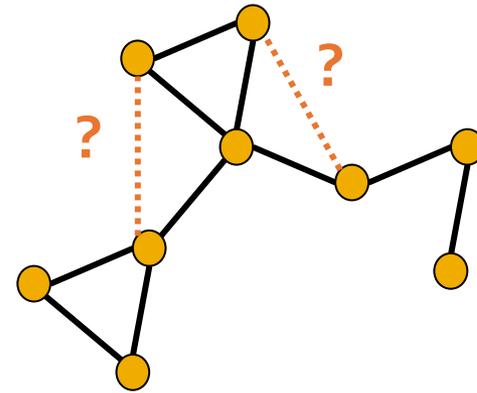


# Algorithms

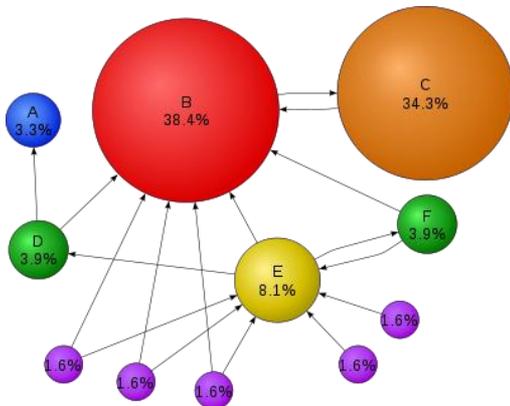
## ■ Decentralized search



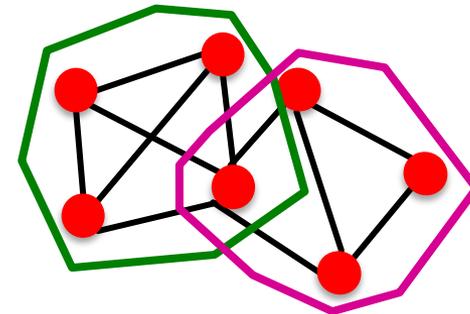
## ■ Link prediction



## ■ Link analysis



## ■ Community detection



# Map of Superpowers

## Properties

Small diameter,  
Edge clustering

Scale-free

Strength of weak ties,  
Core-periphery

Densification power  
law,  
Shrinking diameters

Information virality,  
reproductive number

## Models

Small-world model,  
Erdős-Renyi model

Preferential  
attachment, Copying  
model

Community-affiliation  
Graph Model

Microscopic model of  
evolving networks

Independent cascade  
model, Game theoretic  
model, SIR

## Algorithms

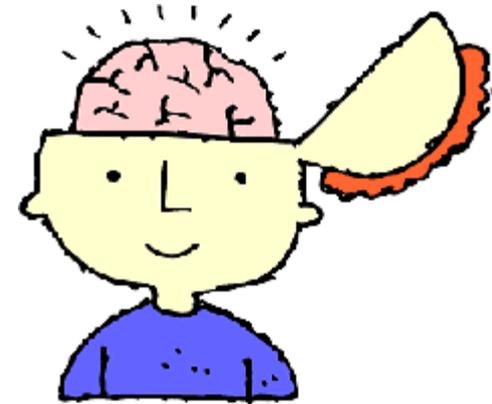
Decentralized search

PageRank, Hubs and  
authorities

Community detection:  
Girvan-Newman,  
Modularity

Link prediction,  
Supervised random  
walks

Influence maximization,  
Outbreak detection, LIM

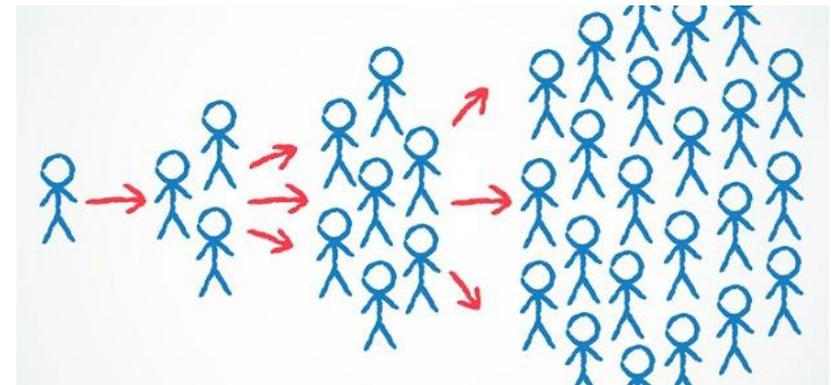


# Applying Our Superpowers

- **Social media analytics**

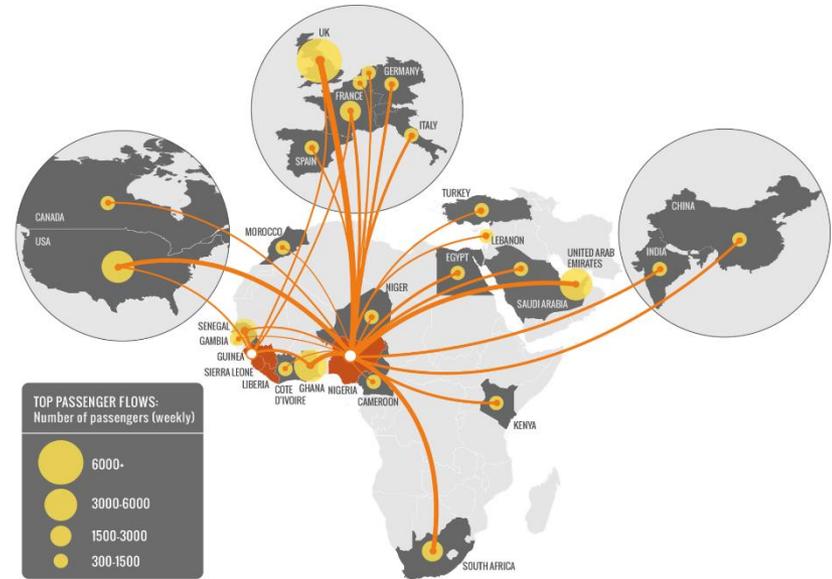


- **Viral marketing**

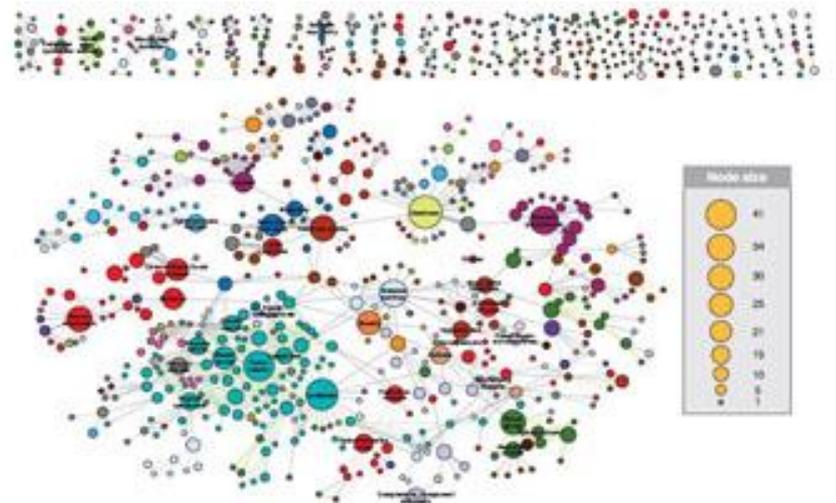


# Applying Our Superpowers

- Predicting epidemics:  
Ebola



- Drug design



# **examples of network studies**

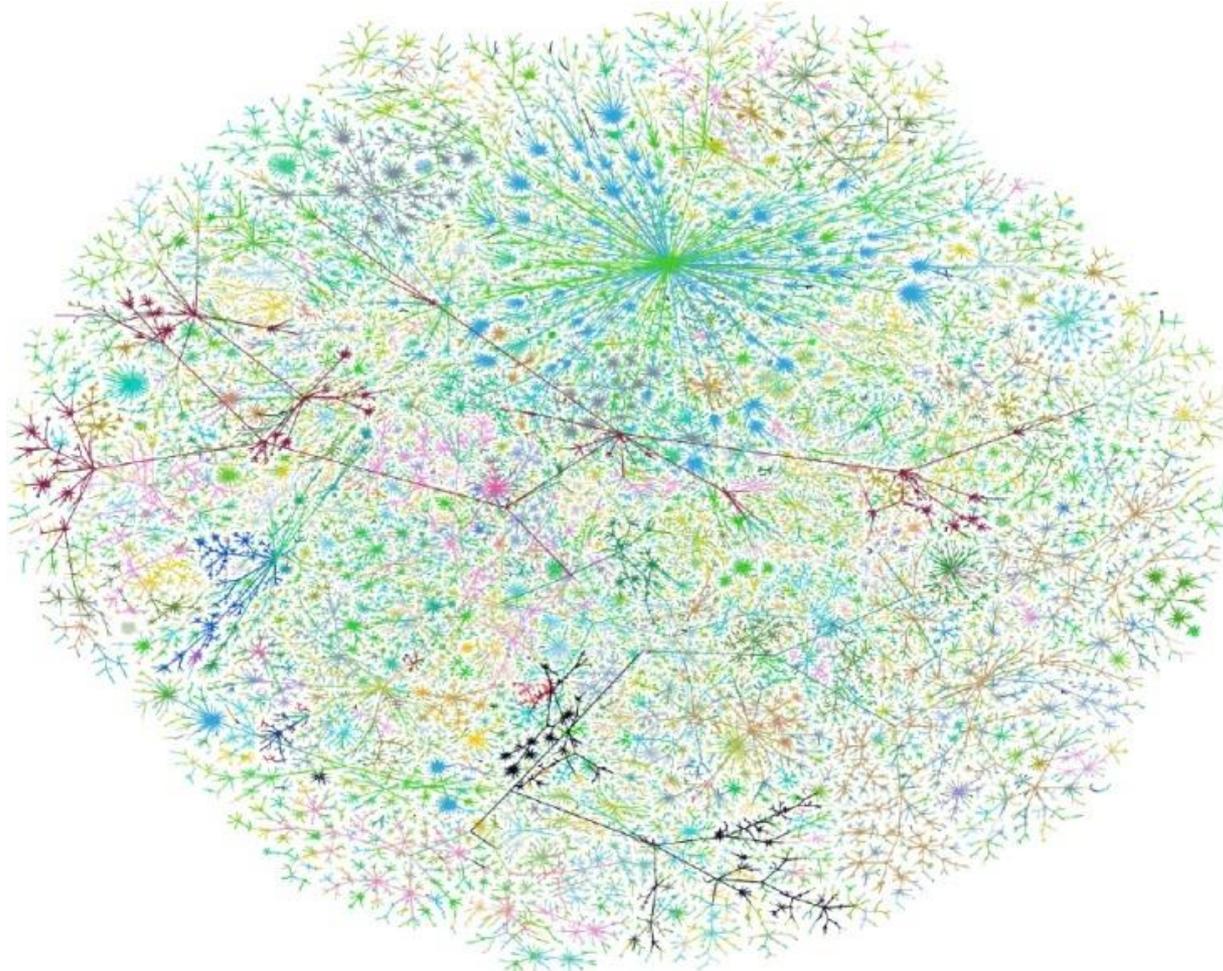
# Networks: Social



## Facebook social graph

4-degrees of separation [Backstrom-Boldi-Rosa-Ugander-Vigna, 2011]

# Networks: Communication

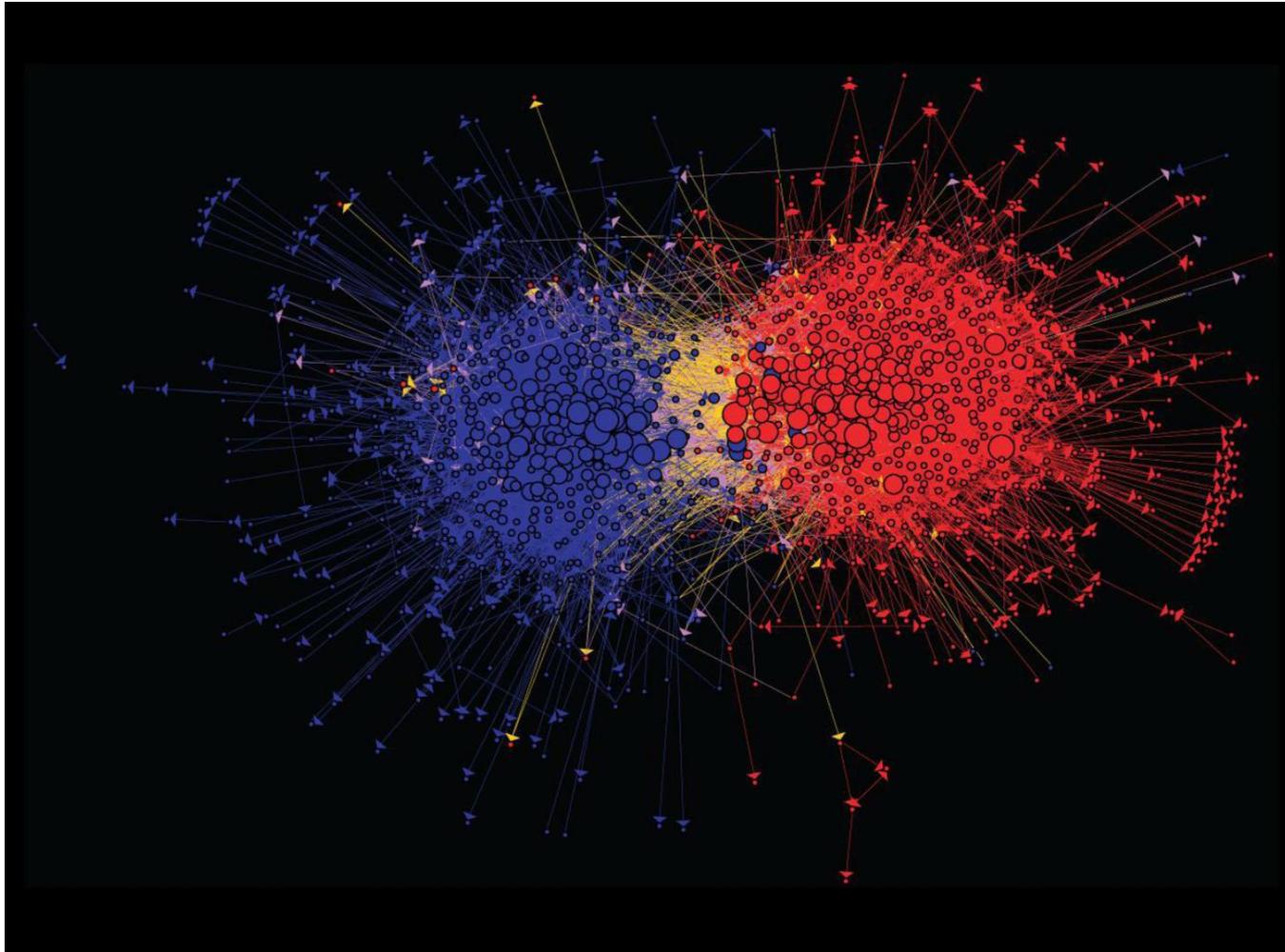


**Graph of the Internet (Autonomous Systems)**

Power-law degrees [Faloutsos-Faloutsos-Faloutsos, 1999]

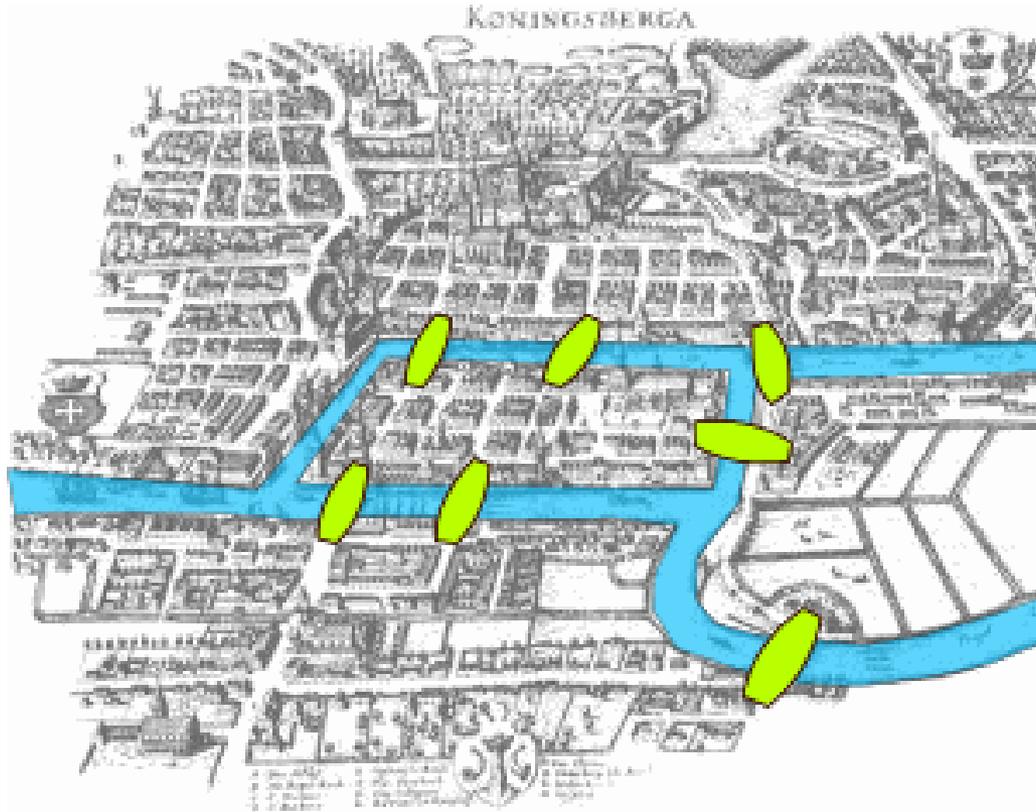
Robustness [Doyle-Willinger, 2005]

# Networks: Media



**Connections between political blogs**  
Polarization of the network [Adamic-Glance, 2005]

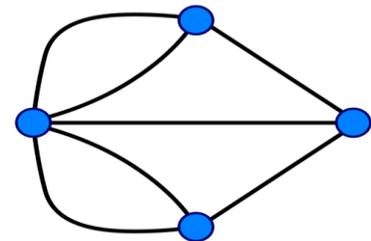
# Networks: Infrastructure



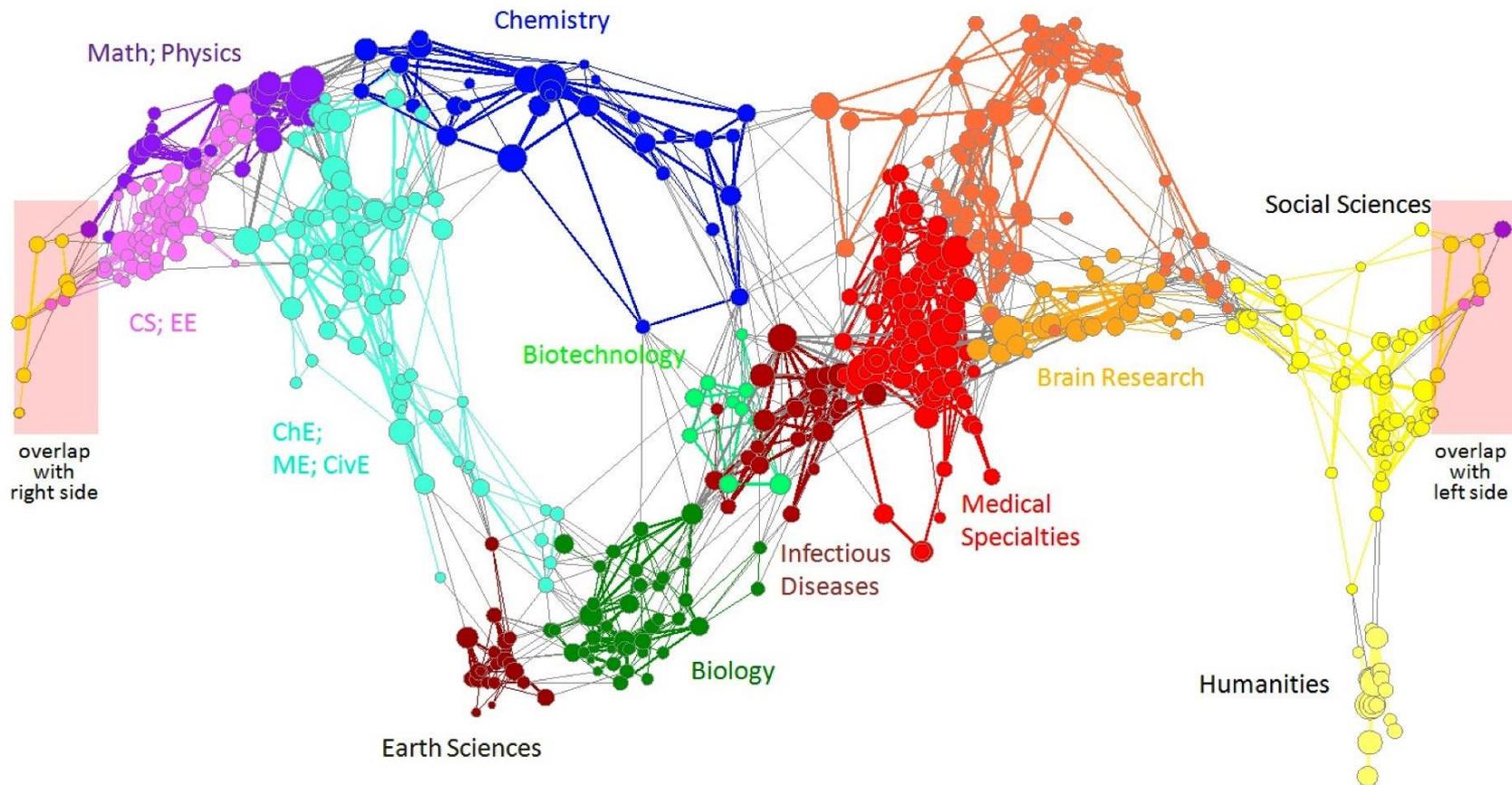
## Seven Bridges of Königsberg

[Euler, 1735]

Return to the starting point by traveling each link of the graph once and only once.



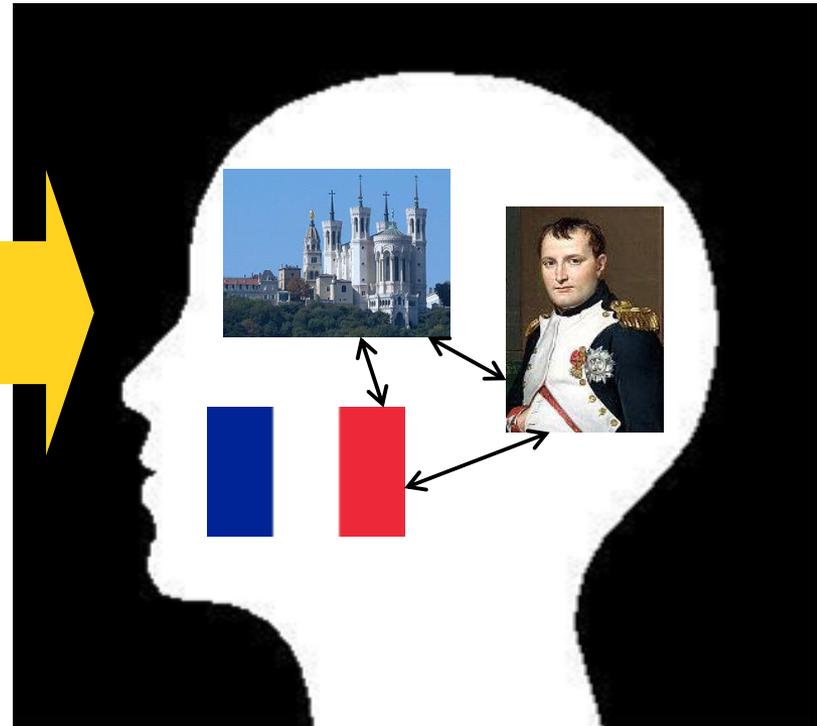
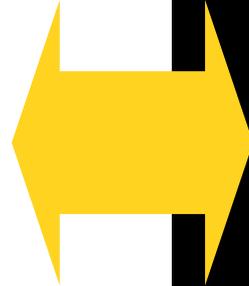
# Networks: Citation



Citation networks and Maps of science

[Börner et al., 2012]

# Networks: Knowledge



Understand how humans  
navigate Wikipedia

Get an idea of how  
people connect concepts

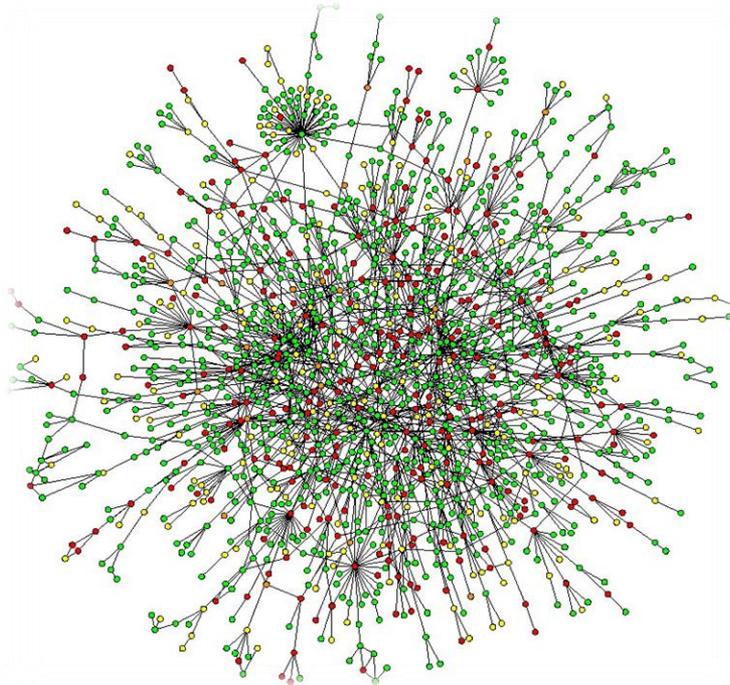


# Networks: Brain



Human brain has between  
~100 billion neurons, ~1,000 trillion synapses  
[Sporns, 2011]

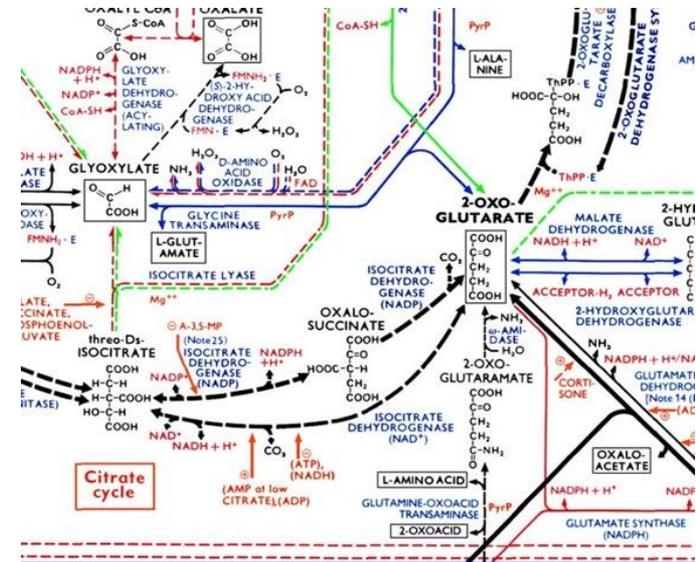
# Networks: Biology



## Protein-Protein Interaction Networks:

Nodes: Proteins

Edges: 'physical' interactions



## Metabolic networks:

Nodes: Metabolites and enzymes

Edges: Chemical reactions

# Web – The Lab for Humanity



LIVEJOURNAL



facebook



The Web is a  
“laboratory” for  
understanding the  
pulse of humanity.



WIKIPEDIA

**examples of network  
analysis impact**

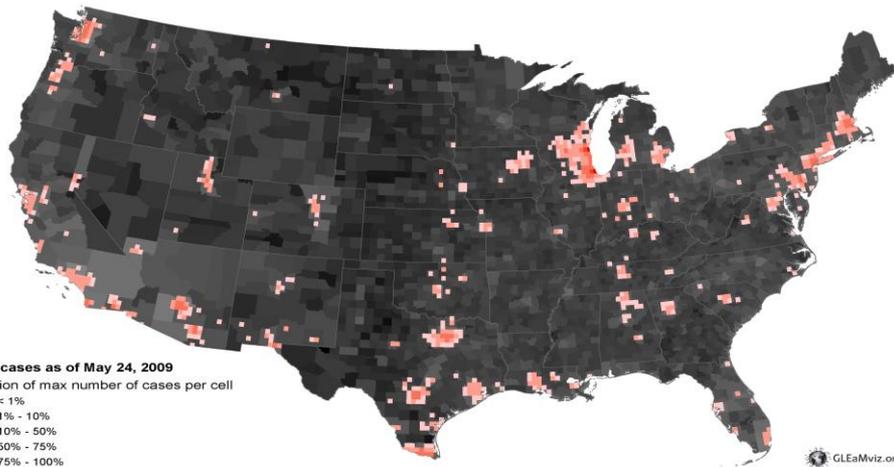
# Networks: Impact



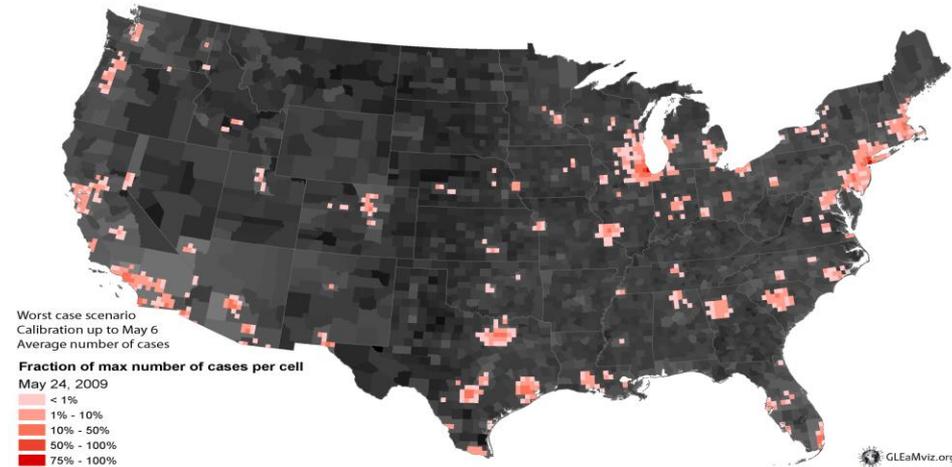
- **Google (Australia?)**  
Market cap: \$1700 billion
- **Cisco (Greece?)**  
Market cap: \$230 billion
- **Meta (Taiwan?)**  
Market cap: \$770 billion

# Networks: Impact

## ■ Predicting epidemics



Real

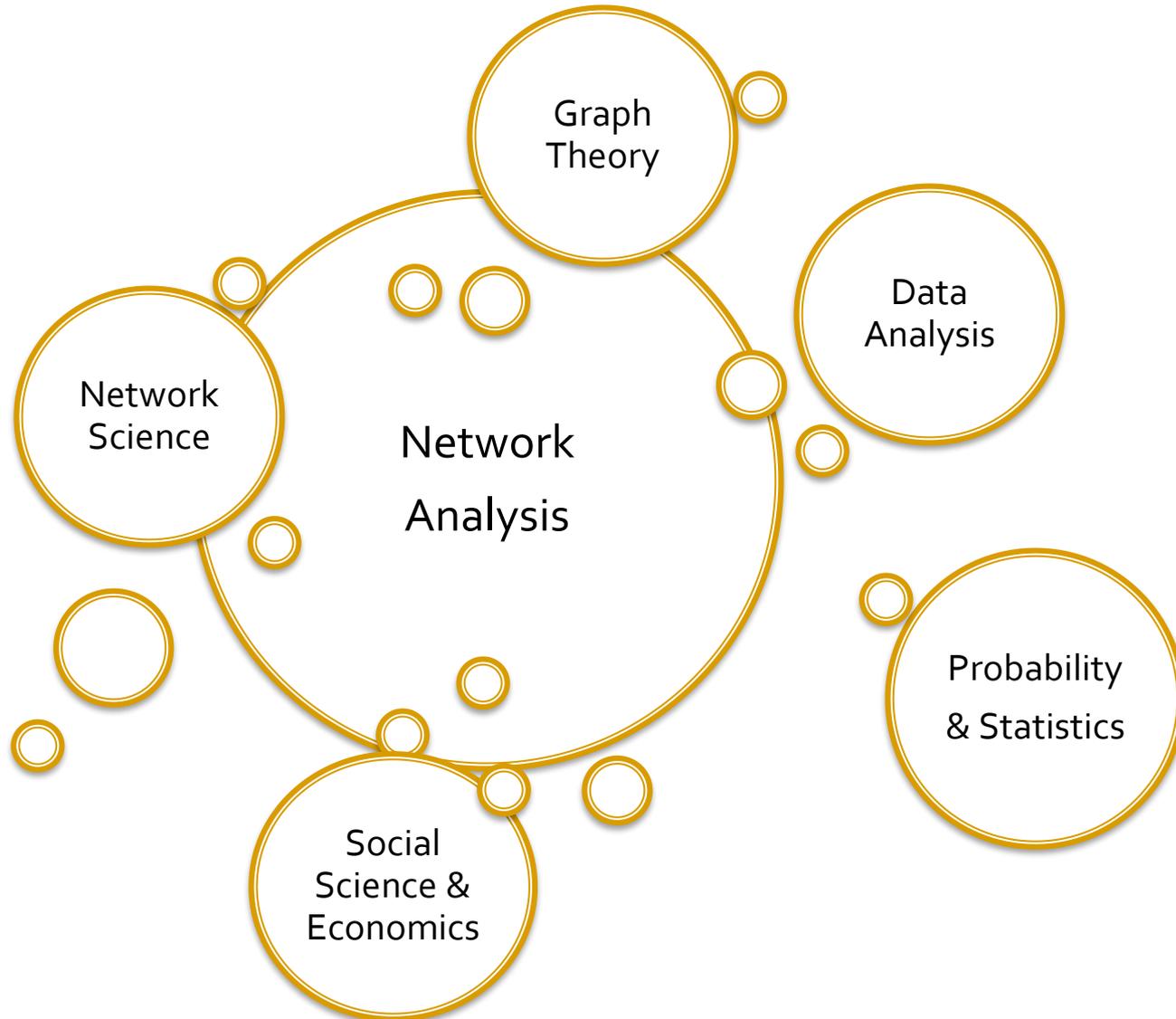


Predicted

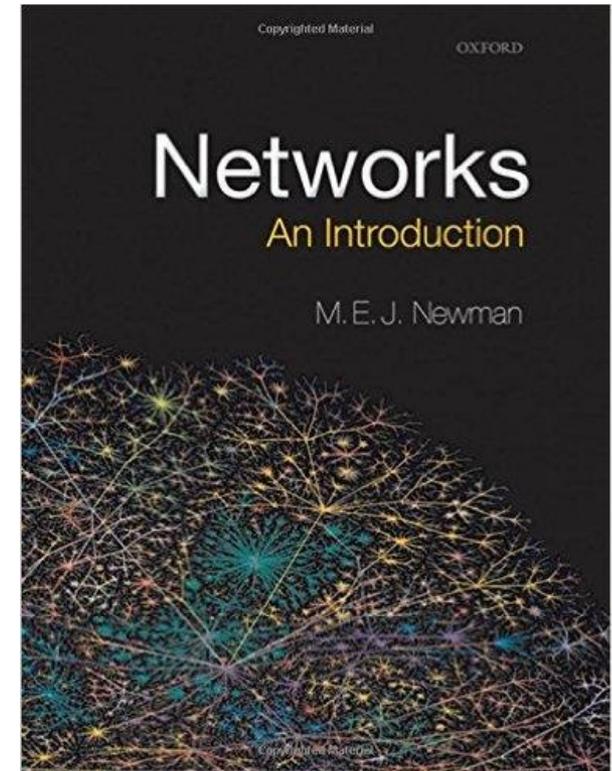
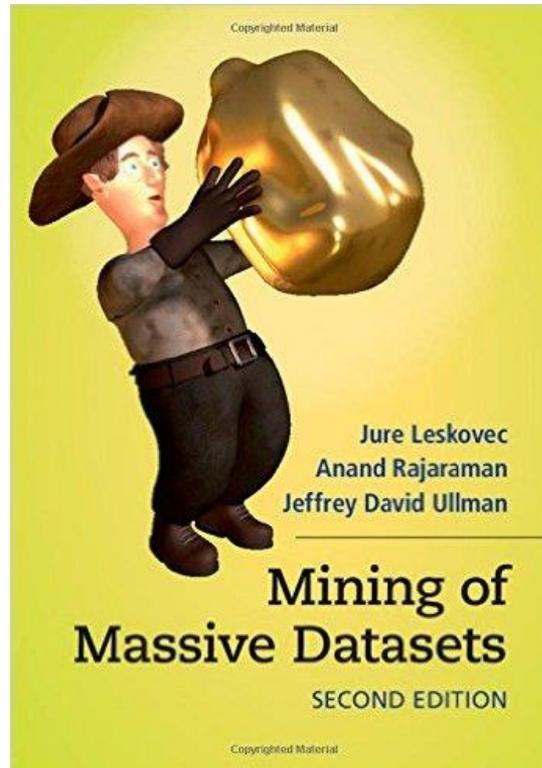
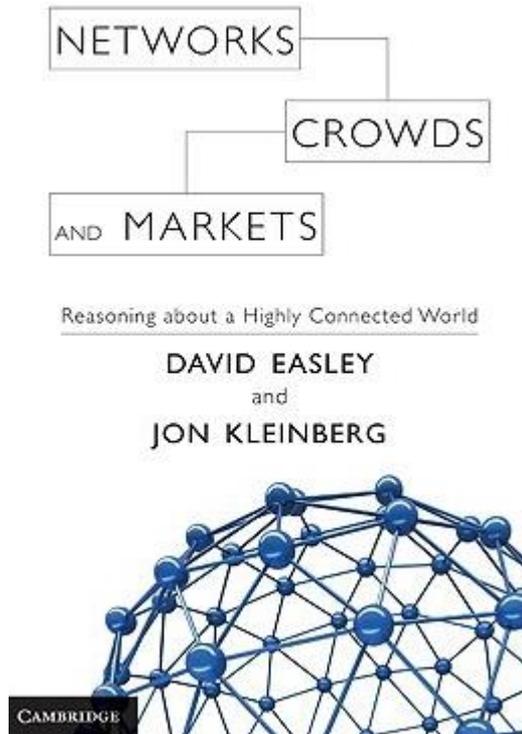
# Networks Really Matter

- If you want to understand the spread of diseases, **can you do it without social networks?**
- If you want to understand the structure of the Web, **it is hopeless without working with the Web's topology**
- If you want to understand dissemination of news or evolution of science, **it is hopeless without considering the information networks**

# Intellectual Content



# “Suggested” Textbooks



# Network Analysis Tools

- **Highly recommend SNAP:**
  - **SNAP C++:** more challenging but more scalable
  - **SNAP.PY:** Python ease of use, most of C++ scalability
- Other tools include:
  - **NetworkX**
  - **JUNG**
  - **iGraph**
  - **GraphX**
  - ...

# Example Research Questions/ Topics

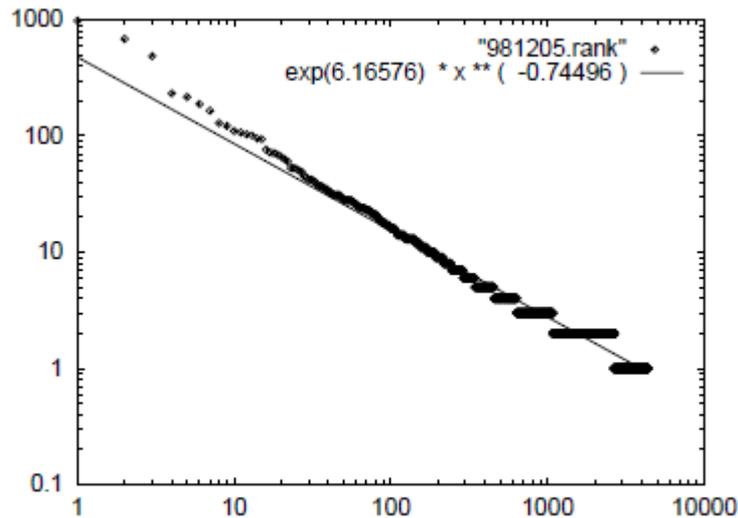
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# Topics

- Measuring real networks
- Modeling the evolution of networks
- Identifying important nodes in the graph
- Finding communities in graphs
- Link prediction and recommendation
- Modeling information cascades in networks
- ...

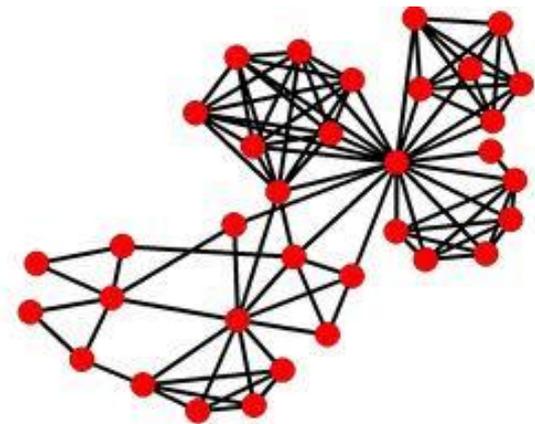
# Understanding Large Graphs

- What does a network **look like**?
  - Measure different properties to understand the structure



(a) Int-12-98

degree of nodes



Triangles in the graph

# Modeling Real Networks

- Real life networks are **not** “random”
- Can we define a **model** that **generates** graphs with statistical properties similar to those in real life?
- The **rich-get-richer** model

We need to accurately model the mechanisms that govern the evolution of networks (for prediction, simulations, understanding)

# Ranking Nodes on the Web

- Is my home page as important as the facebook page?
- We need algorithms to compute the **importance of nodes** in a graph
- The **PageRank** Algorithm
  - A success story of network use



It is impossible to create a web search engine without understanding the web graph

# Link Prediction

- Given a snapshot of a social network at time  $t$ , we seek to accurately **predict** the edges that will be added to the network during the interval from time  $t$  to a given future time  $t'$ .
- Applications
  - Accelerate the growth of a social network (e.g., Facebook, LinkedIn, Twitter)
  - Maximize information cascades

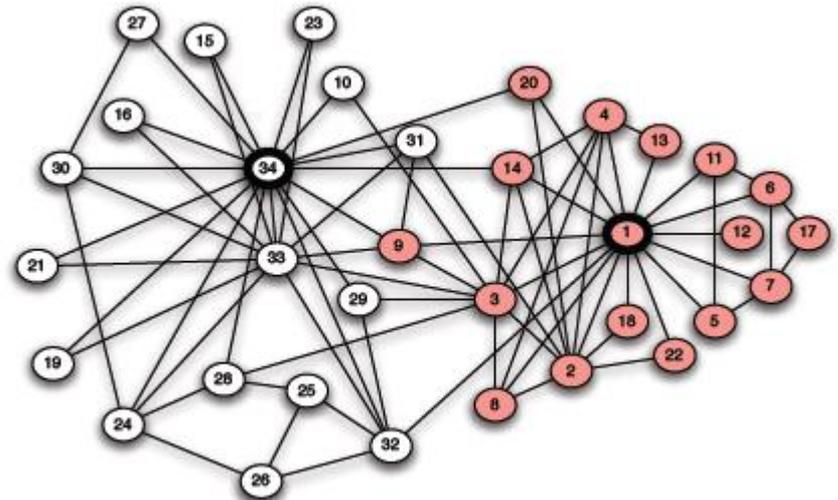
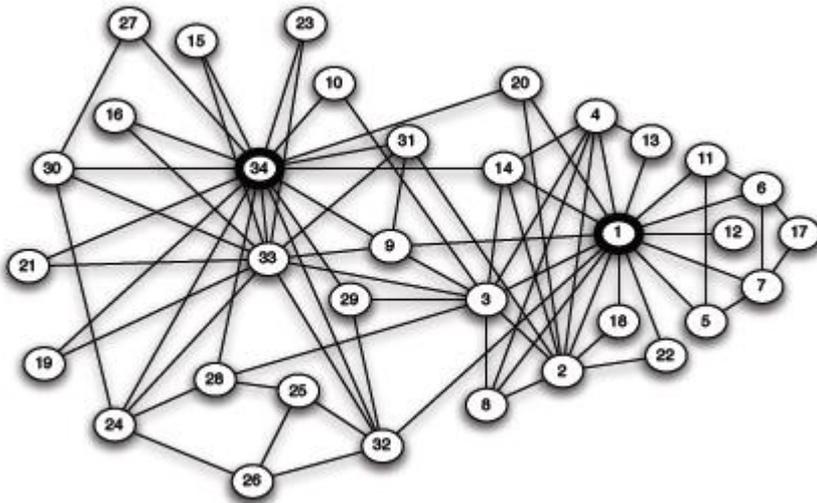


How do we predict future links?

# Clustering and Communities

- What is **community**?
  - “Cohesive subgroups are subsets of actors among whom there are relatively strong, direct, intense, frequent, or positive ties.” [Wasserman & Faust '97]

Karate club example [W. Zachary, 1970]



# Information/Virus Cascade

- How do **viruses spread** between individuals?  
How can we stop them?
- How does **information propagates** in social and information networks? What items become **viral**? Who are the **influencers** and trend-setters?
- We need **models and algorithms** to answer these questions

Online advertising relies heavily on online social networks and word-of-mouth marketing. There is currently need for models for understanding the spread of Covid-19 virus.

# Mining Social Media

- **Social Media** (Twitter, Facebook, Instagram) have supplanted the traditional media sources
  - Information is generated and disseminated by users
- Interesting problems:
  - Automatically detect events using Twitter
    - Earthquake response
    - Crisis detection and management
  - Sentiment mining
  - Track the evolution of events: socially, geographically, over time
  - ...

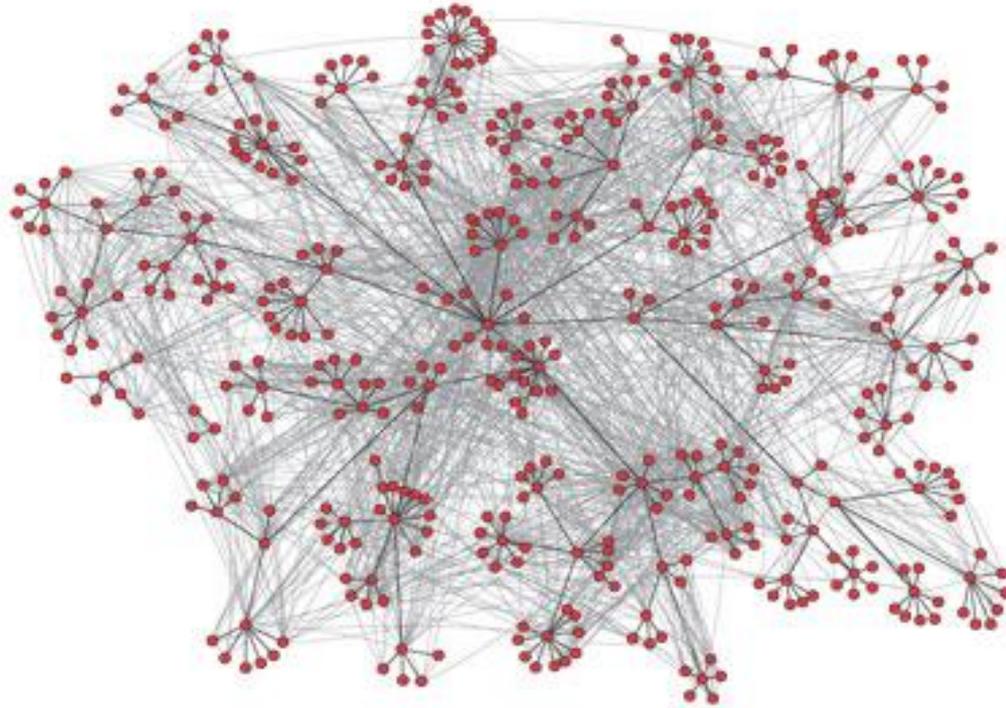
# Research in Graph Mining

- **Current hot research topics:**
  - Graph representation learning
  - Graph neural networks
  - Graph attention mechanisms
  - Graph generative models
  - Graph classification, clustering, anomaly detection
  - Dynamic graph analysis and mining
- **Relevant research conferences**
  - Data Mining: KDD, ICDM, WSDM, WWW, ...
  - ML: ICML, NeurIPS, ECML/PKDD, ...

**Example Topic:**  
**Structure of the Web Graph**

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# Structure of Networks?



Network is a collection of objects where some pairs of objects are connected by links

**What is the structure of the network?**



# Networks or Graphs?

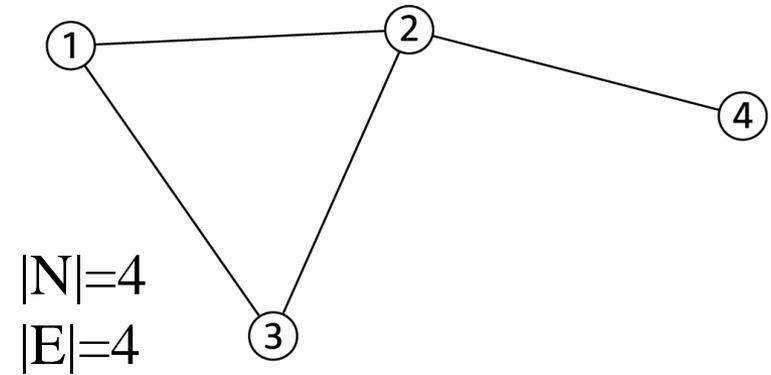
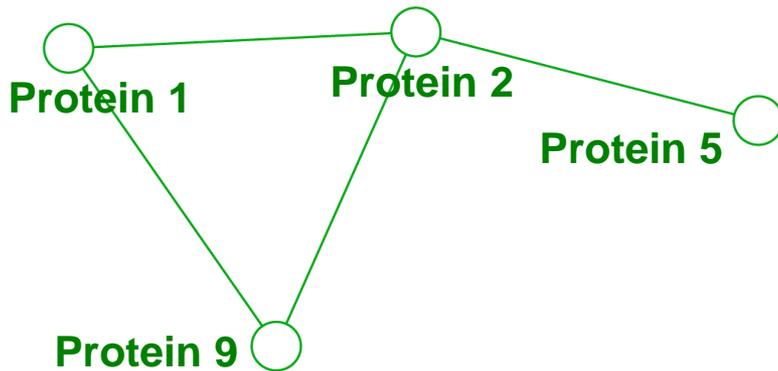
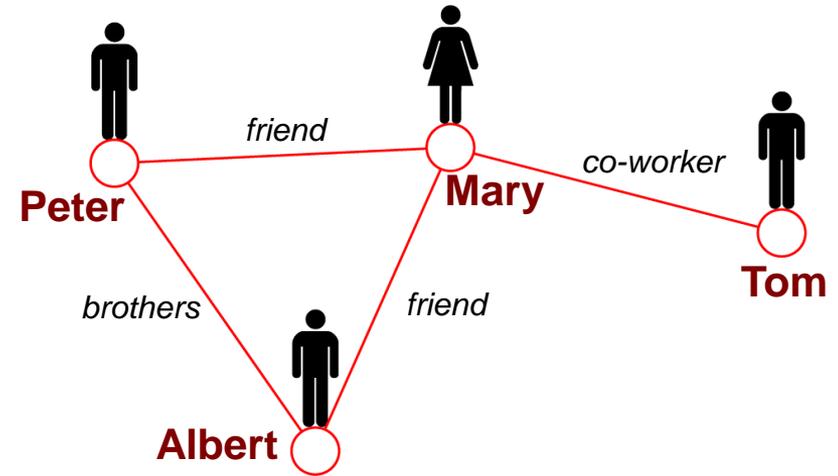
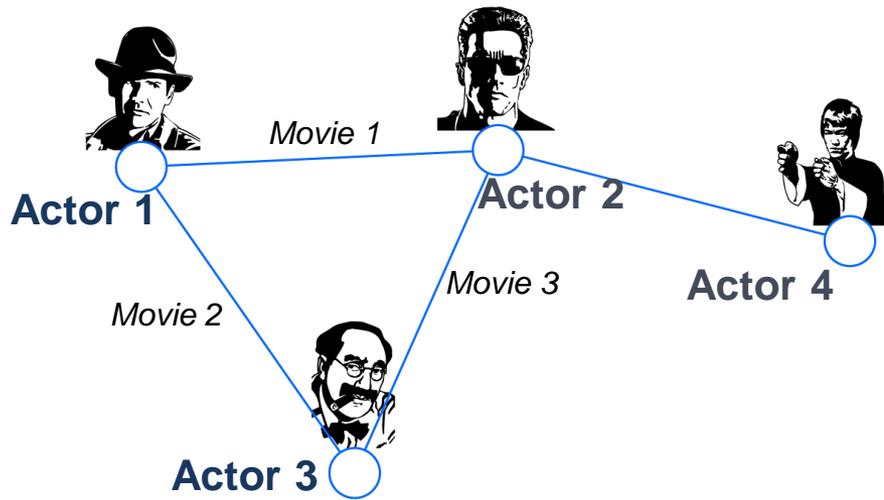
- **Network** often refers to real systems
  - Web, Social network, Metabolic network

**Language:** Network, node, link
- **Graph** is mathematical representation of a network
  - Web graph, Social graph (a Facebook term)

**Language:** Graph, vertex, edge

We will try to make this distinction whenever it is appropriate, but in most cases we will use the two terms interchangeably

# Networks: Common Language

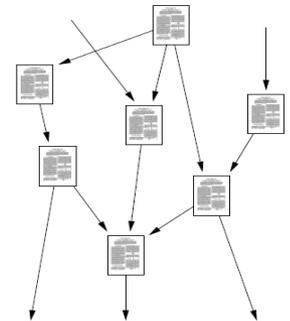
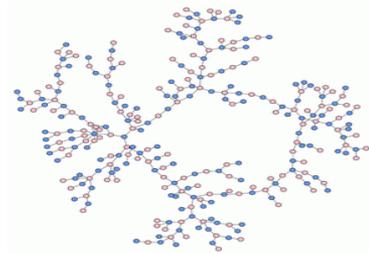


# Choosing Proper Representation

- **How to build a graph:**
  - What are nodes?
  - What are edges?
- **Choice of the proper network representation of a given domain/problem determines our ability to use networks successfully:**
  - In some cases there is a unique, unambiguous representation
  - In other cases, the representation is by no means unique
  - The way you assign links will determine the nature of the question you can study

# Choosing Proper Representation

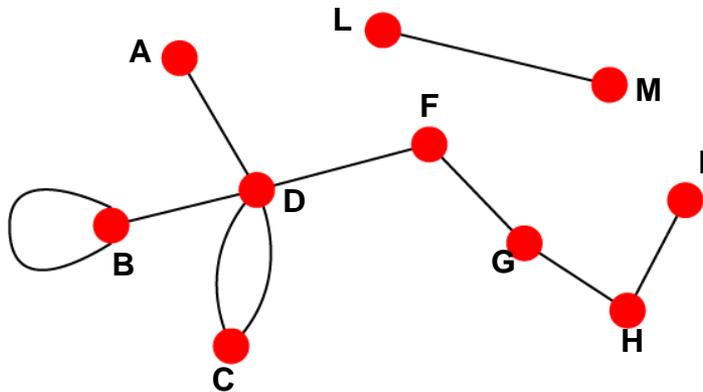
- If you connect individuals that work with each other, you will explore a **professional network**
- If you connect those that have a sexual relationship, you will be exploring **sexual networks**
- If you connect scientific papers that cite each other, you will be studying the **citation network**
- **If you connect all papers with the same word in the title, you will be exploring what?** It is a network, nevertheless



# Undirected vs. Directed Networks

## Undirected

- **Links:** undirected (symmetrical, reciprocal)

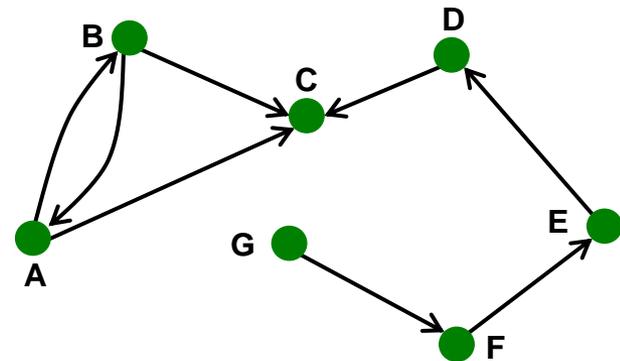


- **Examples:**

- Collaborations
- Friendship on Facebook

## Directed

- **Links:** directed (arcs)

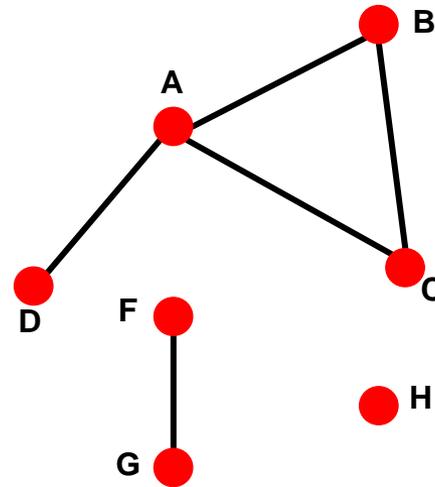
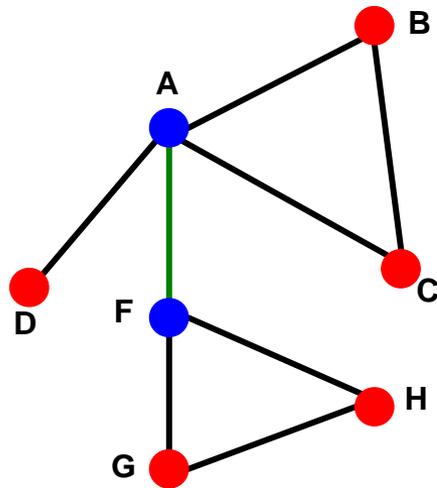


- **Examples:**

- Phone calls
- Following on Twitter

# Connectivity of Graphs

- **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**

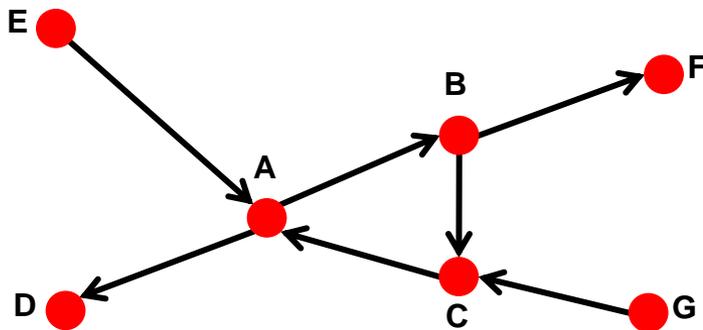
Isolated node (node H)

**Bridge edge:** If we erase it, the graph becomes disconnected.

**Articulation point:** If we erase it, the graph becomes disconnected.

# Connectivity of Directed Graphs

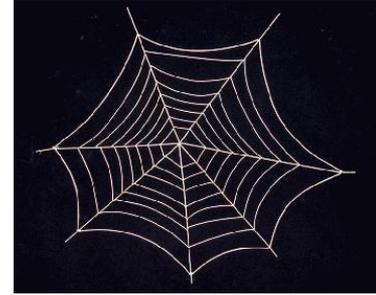
- **Strongly connected directed graph**
  - has a path from each node to every other node and vice versa (e.g., A-B path and B-A path)
- **Weakly connected directed graph**
  - is connected if we disregard the edge directions



Graph on the left is connected but not strongly connected (e.g., there is no way to get from F to G by following the edge directions).

# Web as a Graph

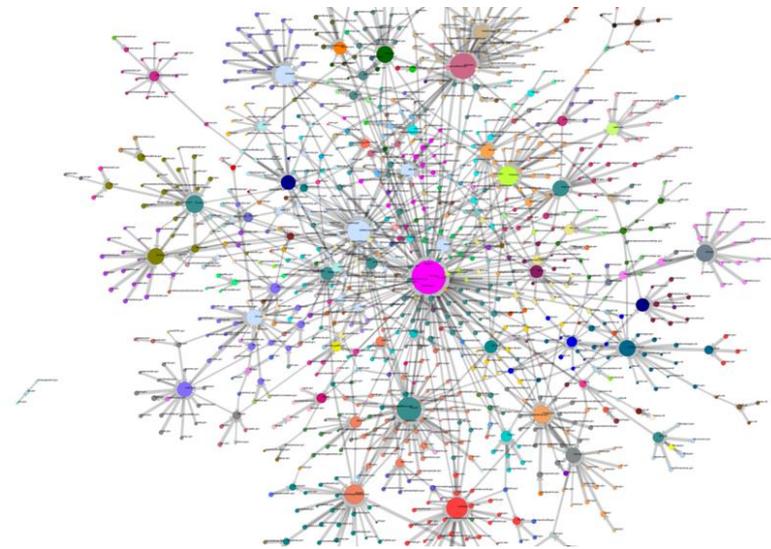
- **Q: What does the Web “look like”?**
- **Here is what we will do next:**
  - We will take a real system (i.e., the Web)
  - We will represent the Web as a graph
  - We will use language of graph theory to reason about the structure of the graph
  - Do a computational experiment on the Web graph
  - **Learn something about the structure of the Web!**



# Web as a Graph

Q: What does the Web “look like” at a global level?

- **Web as a graph:**
  - Nodes = web pages
  - Edges = hyperlinks
  - **Side issue:** What is a node?
    - Dynamic pages created on the fly
    - “dark matter” – inaccessible database generated pages



# The Web as a Graph

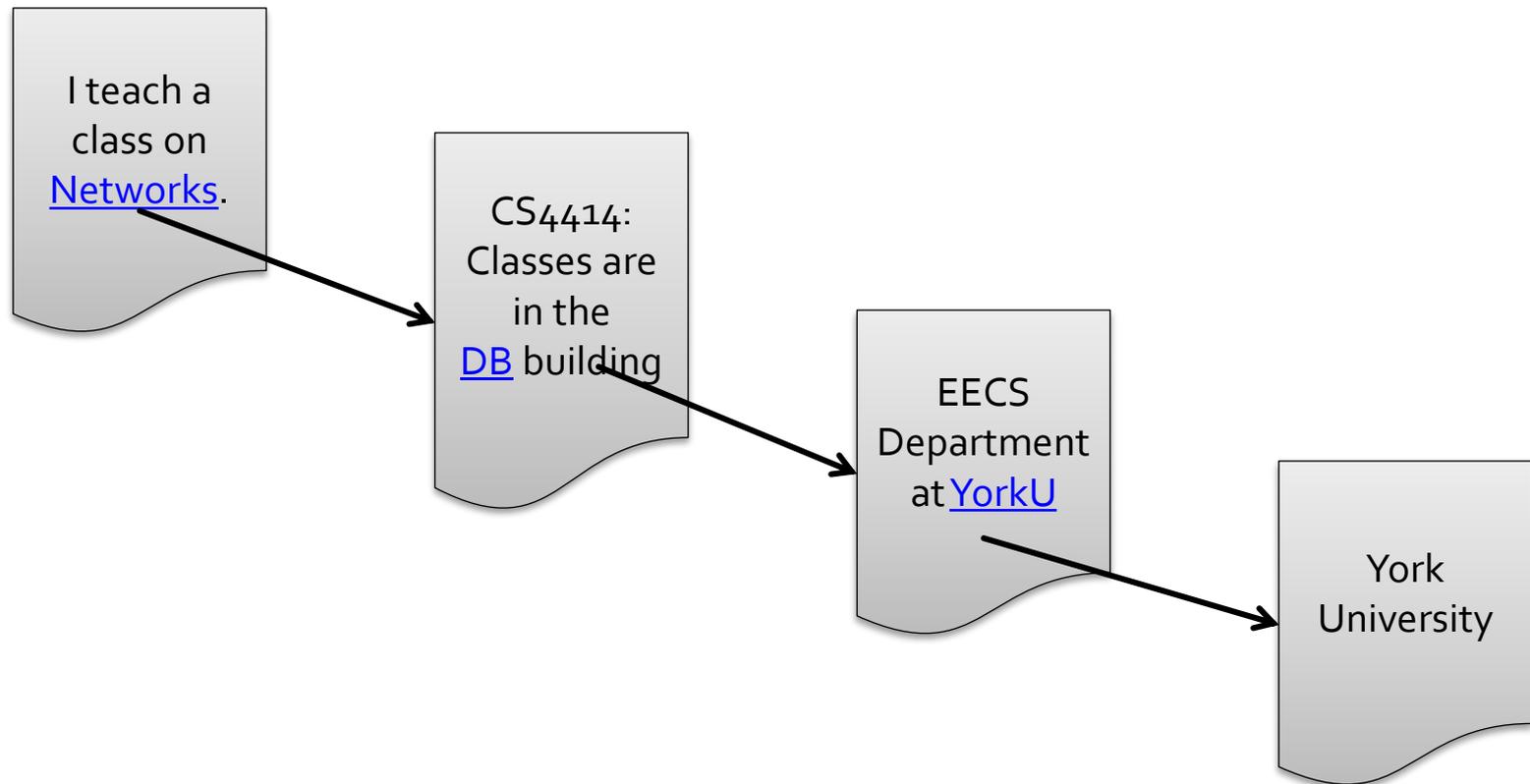
I teach a  
class on  
[Networks](#).

CS4414:  
Classes are  
in the  
[DB](#) building

EECS  
Department  
at [YorkU](#)

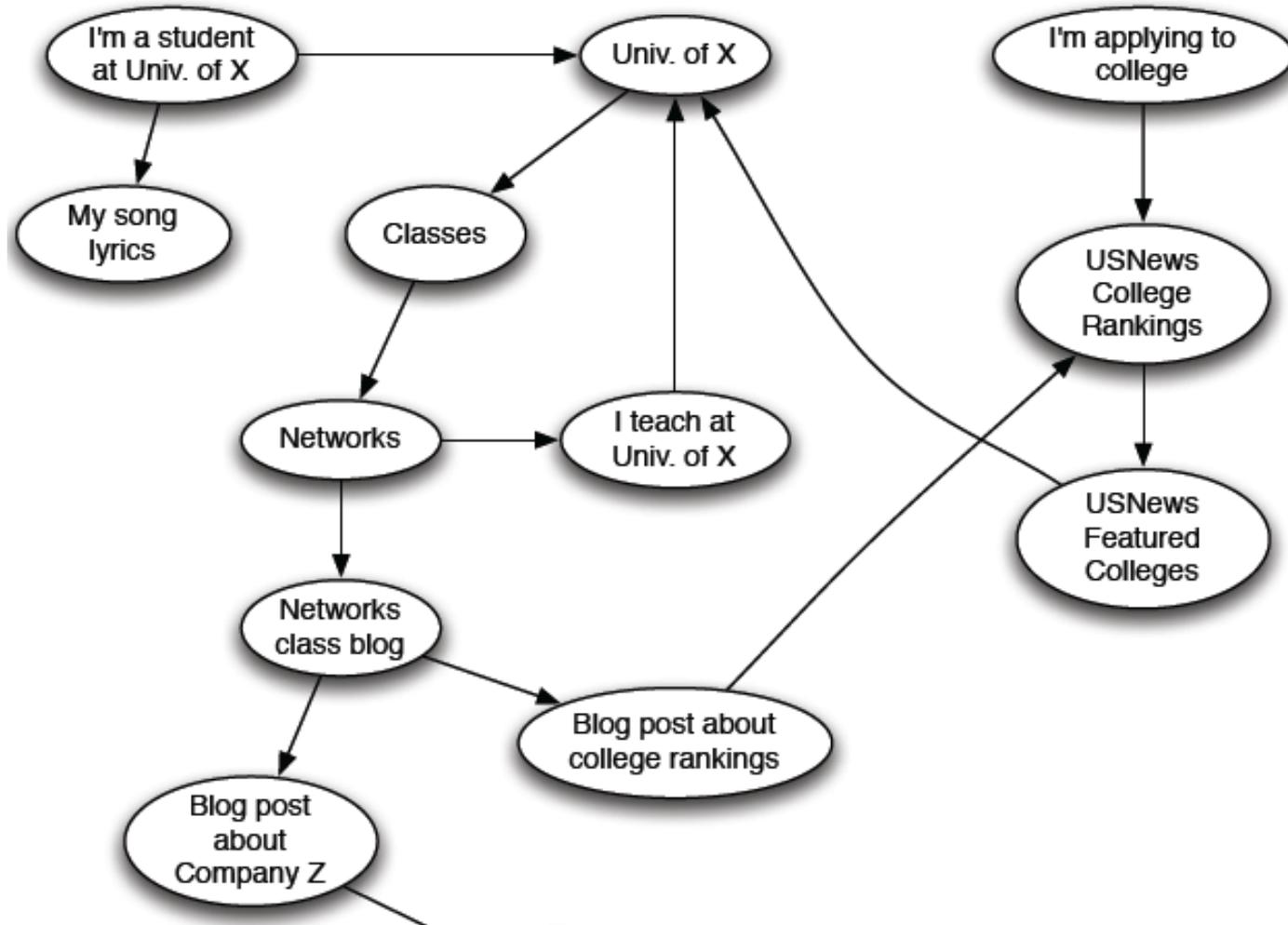
York  
University

# The Web as a Graph



- In early days of the Web links were **navigational**
- Today many links are **transactional**

# The Web as a Directed Graph

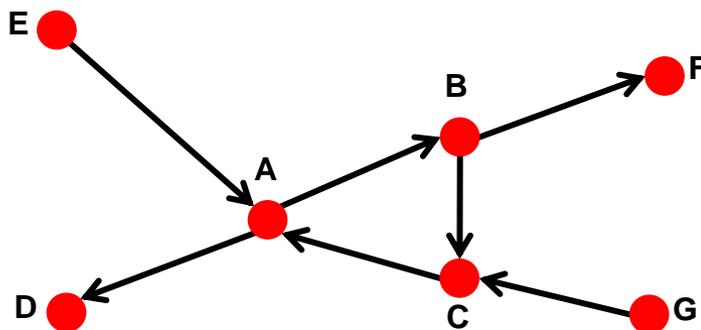


# What Does the Web Look Like?

- How is the Web linked?
- What is the “map” of the Web?

Web as a directed graph [Broder et al. 2000]:

- What nodes can reach  $v$ ?
- Given node  $v$ , what other nodes are reached by  $v$ ?



$$In(v) = \{w \mid w \text{ can reach } v\}$$

$$Out(v) = \{w \mid v \text{ can reach } w\}$$

For example:

$$In(A) = \{A, B, C, E, G\}$$

$$Out(A) = \{A, B, C, D, F\}$$

# Directed Graphs

- Two types of directed graphs:

- Strongly connected:

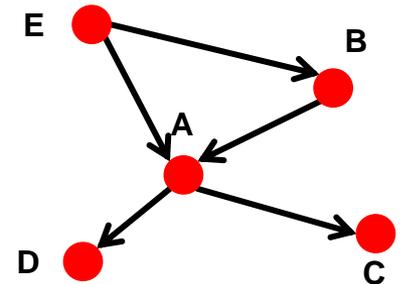
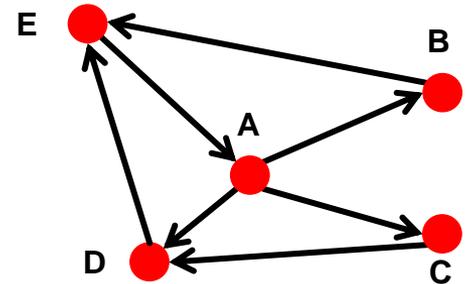
- Any node can reach any node via a directed path

$$In(A) = Out(A) = \{A, B, C, D, E\}$$

- DAG – Directed Acyclic Graph:

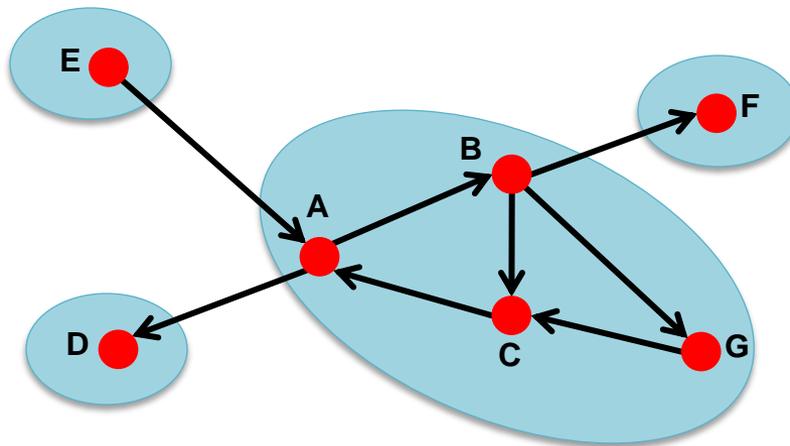
- Has no cycles: if  $u$  can reach  $v$ , then  $v$  can not reach  $u$

- Any directed graph can be expressed in terms of these two types!



# Strongly Connected Component

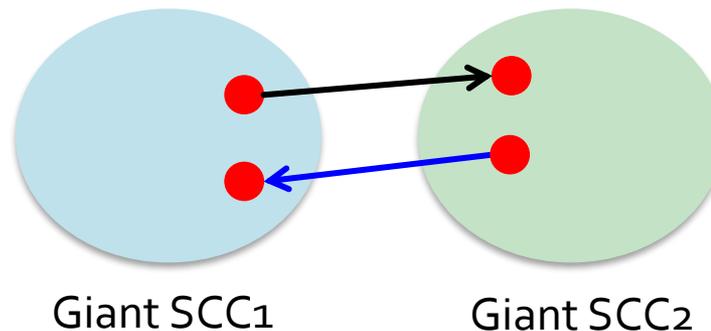
- **Strongly connected component (SCC)** is a set of nodes  $S$  so that:
  - Every pair of nodes in  $S$  can reach each other
  - There is no larger set containing  $S$  with this property



Strongly connected components of the graph:  
 $\{A, B, C, G\}$ ,  $\{D\}$ ,  $\{E\}$ ,  $\{F\}$

# Graph Structure of the Web

- **There is a single giant SCC**
  - That is, there won't be two SCCs
- **Heuristic argument:**
  - It just takes 1 page from one SCC to link to the other SCC
  - If the 2 SCCs have millions of pages the likelihood of this not happening is very very small



# Structure of the Web

- **Broder et al., 2000:**

- Altavista crawl from October 1999

- 203 million URLs

- 1.5 billion links

- Computer: Server with 12GB of memory

- **Undirected version of the Web graph:**

- 91% nodes in the largest weakly conn. component

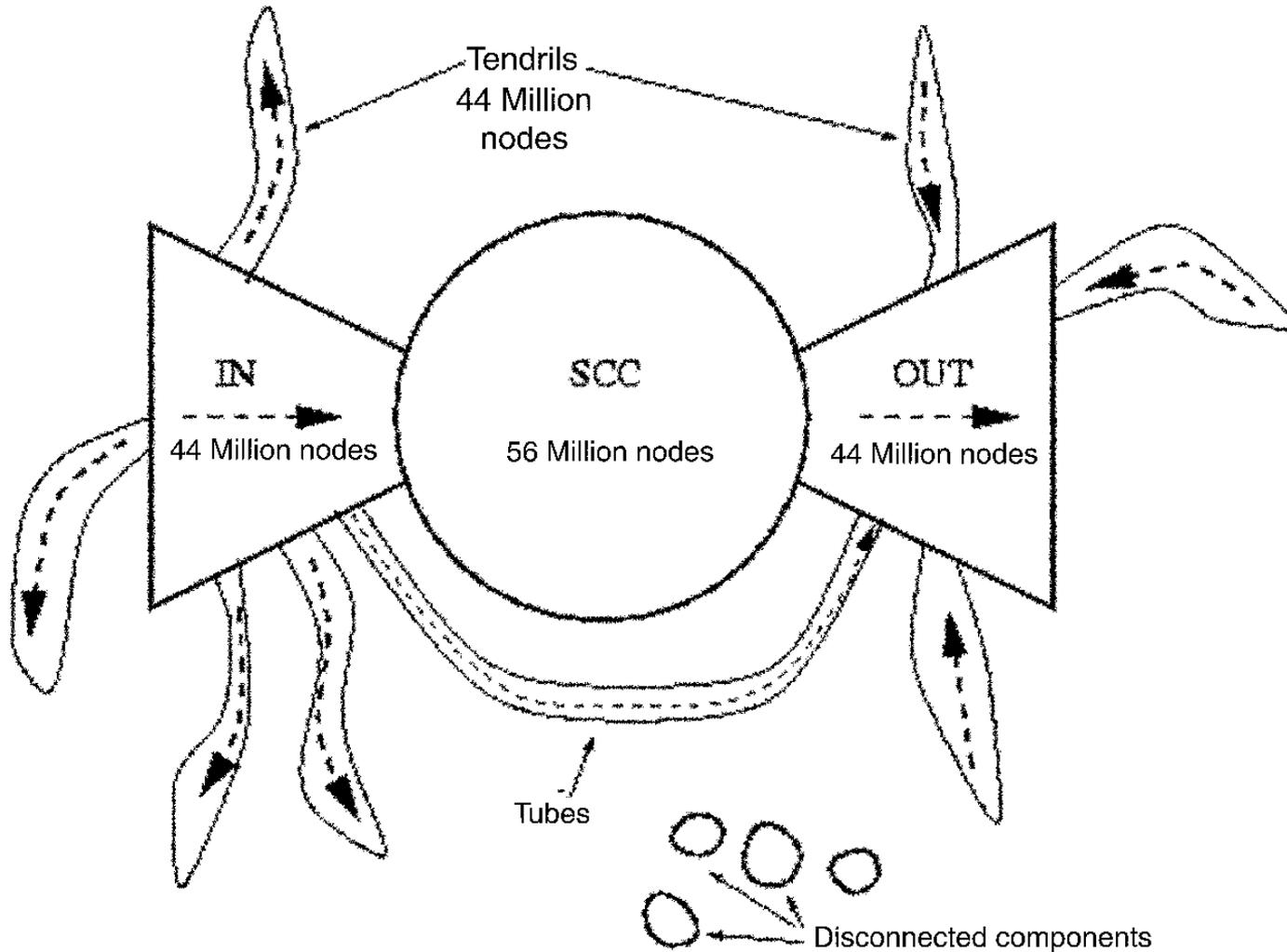
- **Are hubs making the web graph connected?**

- Even if they deleted links to pages with in-degree  $>10$   
WCC was still  $\approx 50\%$  of the graph

# Structure of the Web

- **Directed version of the Web graph:**
  - **Largest SCC:** 28% of the nodes (56 million)
  - Taking a random node  $v$ 
    - $\text{Out}(v) \approx 50\%$  (100 million)
    - $\text{In}(v) \approx 50\%$  (100 million)
- **What does this tell us about the conceptual picture of the Web graph?**

# Bow-tie Structure of the Web



**203 million pages, 1.5 billion links** [Broder et al. 2000]

# What did We Learn/Not Learn ?

- **What did we learn:**
  - Some conceptual organization of the Web (i.e., the bowtie)
- **What did we not learn:**
  - **Treats all pages as equal**
    - Google's homepage == my homepage
  - **What are the most important pages**
    - How many pages have  $k$  in-links as a function of  $k$ ?  
The degree distribution:  $\sim k^{-2}$
    - Link analysis ranking -- as done by search engines (PageRank)
  - **Internal structure inside giant SCC**
    - Clusters, implicit communities?
  - **How far apart are nodes in the giant SCC:**
    - Distance = # of edges in shortest path
    - Avg = 16 [Broder et al.]