



### Networks

- Network Analysis
  - Applications
  - Network Properties
- Network Models
  - Random-Graph Models
  - Growing Random Models
  - Strategic Network Formation
- Network Structure & Dynamics
  - Diffusion through Networks
  - Search on Networks
  - Social Influence Models
  - Networked Markets

Bibliography



berzal@acm.org



### Network Analysis



Networks permeate our lives.

Networks play a central role in determining

- the transmission of information about job opportunities,
- how diseases spread,

- which products we buy,
- our likelihood of succeeding professionally,

### Network Analysis

As a field of study...

 How relationships between parts give rise to the collective behaviors of a system and how the system interacts and forms relationships with its environment (complex systems).

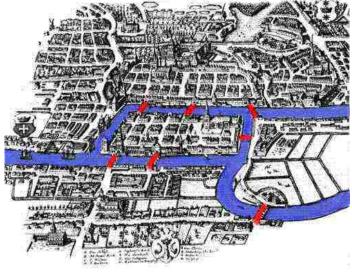
 Common principles, algorithms and tools that govern network behavior (network science).





### Network Analysis

#### **Origins: Graph Theory**



The Seven Bridges of Könisberg (Leonhard Euler, 1736)



#### Networks as graphs "on steroids"...

- **Objects**: Graph vertices.
  - Objects can be of different kinds.
  - Objects can be labeled.
  - Objects can have attributes
- Links between objects: Graph edges.
  - Links can be of different kinds.
  - Links can be directed (arcs) or undirected (edges).
  - Links can have attributes.







### Network Analysis



#### A formal definition of network

[Ted G. Lewis: "Network Science," 2009]

#### G(t) = { N(t), L(t), f(t) : J(t) }

where

- t = time (simulated or real)
  - N = nodes (a.k.a. vertices or "actors")
  - L = links (a.k.a. edges)
  - f = topology (connections through links)
  - J = behavior of nodes and links (algorithm)

### Network Analysis



#### An interdisciplinary field: Complex systems

("networks of heterogeneous components that interact")

- Physics: Nonlinear dynamics & chaos.
  Dynamical systems that are highly sensitive to initial conditions (a.k.a. butterfly effect).
- Economics: Markets.
  Spontaneous (or emergent) order as the result of human action, but not the execution of any human design [Austrian perspective].
- Information theory: Complex adaptive systems. (focus on the ability to change and learn from experience).



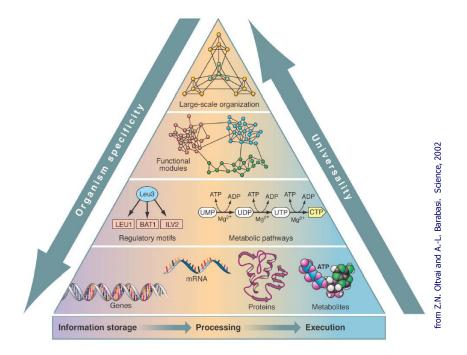


- "Cheminformatics": Chemical compounds.
- "Bioinformatics": Protein networks & bio-pathways
- Software Engineering: Program analysis...
- Network flow analysis (transport, workflows...)
- Semi-structured databases, e.g. XML
- Knowledge management: Ontologies & semantic nets
- Computer-aided design (CAD): IC design...
- Geographic information systems (GIS) & cartography
- Social networks, e.g. Web
- Economic networks, e.g. markets



### Applications

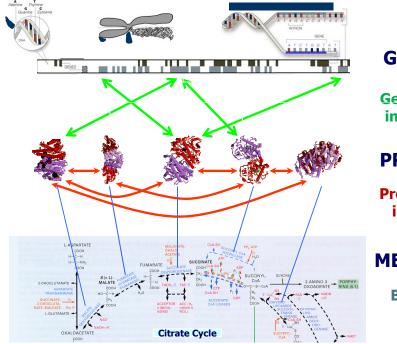
#### "Life complexity pyramid"







#### **Biological networks**



#### GENOME

Gene-protein interactions

#### PROTEOME

Protein-protein interactions

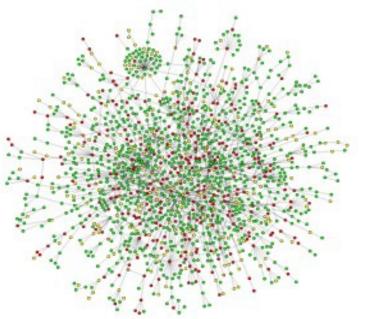
#### **METABOLISM**

Biochemical reactions



## Applications

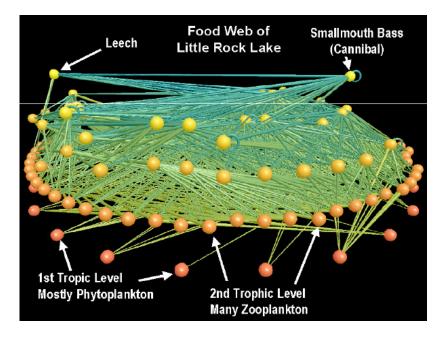
#### Yeast protein interaction network







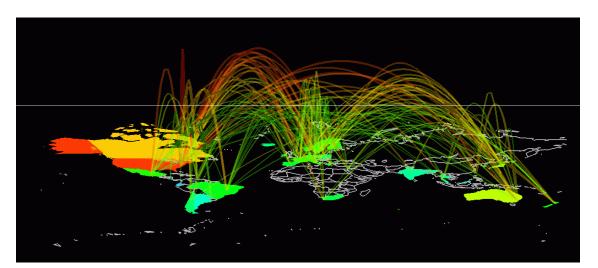
#### Ecological network: Trophic relationships in a food web.





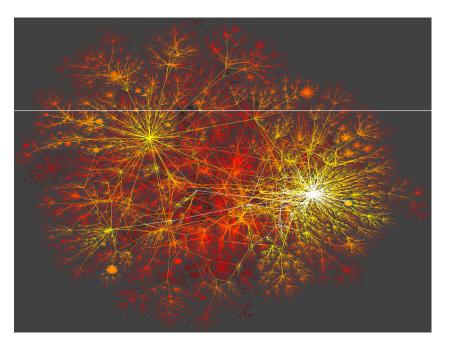
# Applications

#### Telecommunication network



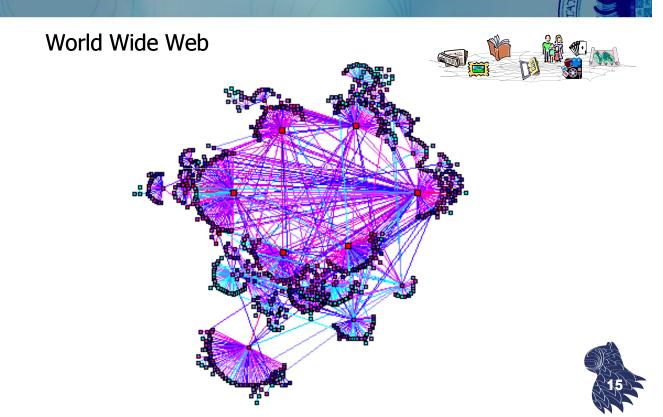


#### Internet



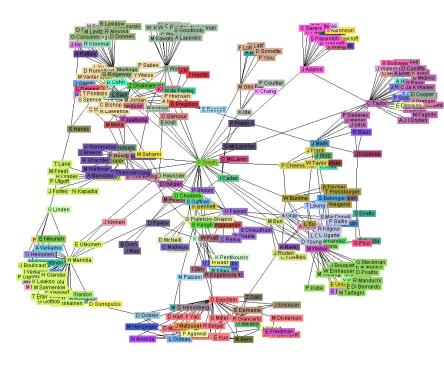


# Applications



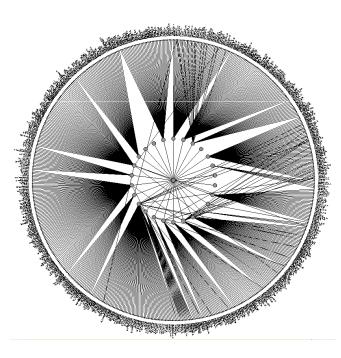


#### Social network: Bibliographic network (coauthors)





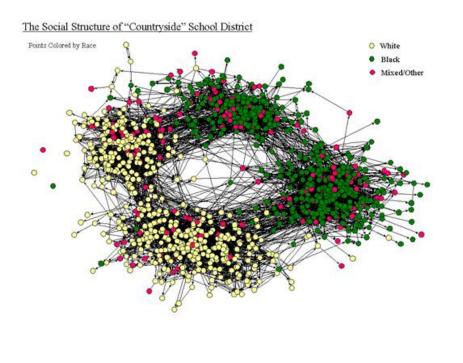
#### Social network: Bibliographic network (coauthors)







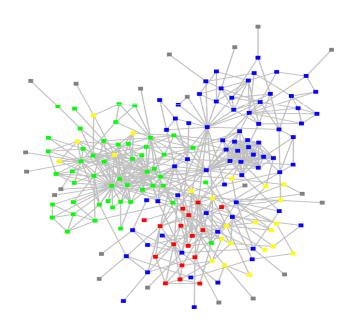
#### **Social network**: FOAF ("friend of a friend")





# Applications

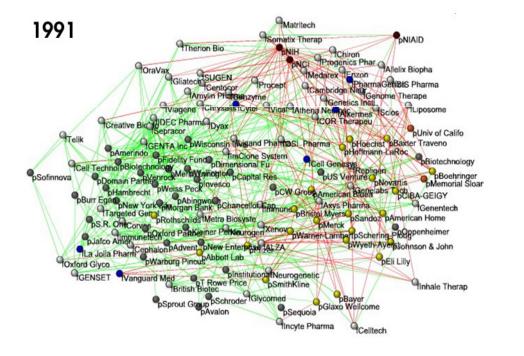
#### Social network: Organization







#### Social network: US Biotech Industry



### **Network Properties**

#### **Common network features:**

- Large scale.
- Continuous evolution.
- Distribution (nodes decide their connections).
- Interactions only through existing links.





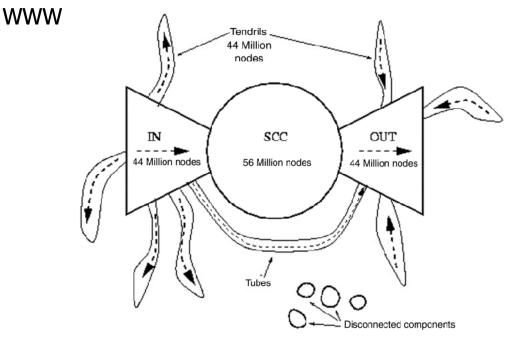


#### Some interesting structural properties:

- Connected components: How many? Of what size?.
- Network diameter: Average distance, worst case...
- Node degree distribution
  & existence of "hubs" (heavily-connected nodes).
- Groupings (balance between local and large-distance connections, as well as their roles).

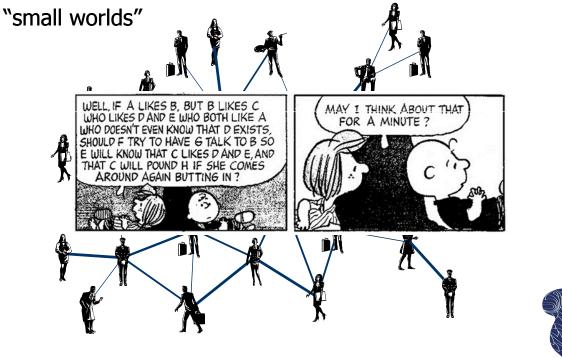
### **Network Properties**

#### **Network Connectivity**





#### **Network Diameter**



### **Network Properties**

#### **Clustering coefficient**

- nbr(u) Neighbors of the node u in the network.
- k Number of neighbors of u, i.e. |nbr(u)|.
- max(u) Maximum number of links among the neighbors of u, e.g. k\*(k-1)/2.

Clustering coefficient for the node u: c(u) = (#links among neighbors of u) / max(u)

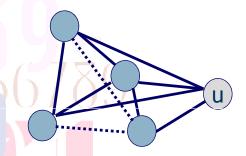
Clustering coefficient for the graph G: C = average of c(u) for every node in G







#### **Clustering coefficient**



#### 0 <= c(u) <= 1

Similarity of u neighbors to a clique (complete graph).

Informal interpretation: "My friends tend to be friends among them."



### **Network Properties**

Path length (L):



#### Clustering coefficient for some real networks

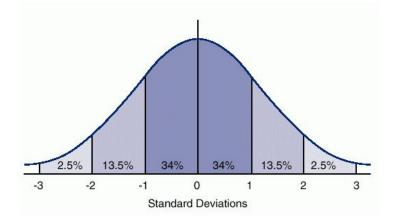
Network	N	С	<b>C</b> <sub>rand</sub>	L
WWW	15 <mark>3</mark> 127	0.1078	0.00023	3.1
Internet	3015-6209	0.18-0.30	0.001	3.7-3.76
Actor	225226	0.79	0.00027	3.65
Coauthorship	529 <mark>0</mark> 9	0 <mark>.4</mark> 3	0.00 <mark>018</mark>	5.9
Metabolic	282	0.32	0.0 <mark>2</mark> 6	2.9
Foodweb	134	0.22	0.0 <mark>6</mark>	2.43
C. elegance	282	0.28	0.05	2.65
12	040	970	2	
istering co	efficier	n <mark>t (C</mark> ):		C>(

L<L<sub>rand</sub>



#### Node degree distribution

Normal distribution Parameters: Average & deviation

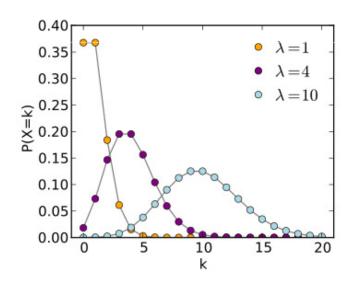




### **Network Properties**

#### Node degree distribution

Poisson distribution Single parameter:  $\lambda$  (mean & deviation)

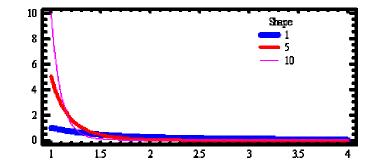






#### Node degree distribution

Pareto distribution (a.k.a. "power law") Single parameter:  $\alpha$ 



**P(x)** ~ x<sup>-α</sup>



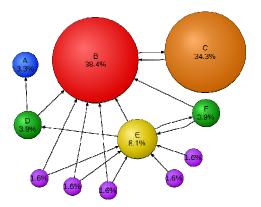
# **The Pareto principle (the "80-20 rule"):** 20% of the population controls 80% of the wealth.

### **Network Properties**

#### Node degree distribution

Hubs Small number of nodes with a very high degree.





 Hubs appear with power laws (P(x) ~ x<sup>-α</sup>), but not with normal/binomial/Poisson distributions.







#### Node degree distribution

Log-log plot

#### Pareto distribution

- $\log(\Pr[X = x]) = \log(1/x^{\alpha}) = -\alpha \log(x)$
- Linear,  $-\alpha$  slope.

#### Normal distribution

- $\log(\Pr[X = x]) = \log(a \exp(-x^2/b)) = \log(a) x^2/b$
- Nonlinear, concave around the average.

#### Poisson distribution

- $\log(\Pr[X = x]) = \log(\exp(-\lambda) \lambda^{x}/x!)$
- Nonlinear.

# **Network Properties**

#### Node degree distribution

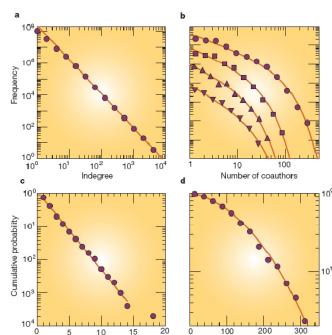
Log-log plot

a WWW power law

#### **b** Coauthorship networks power law with exponential cutoff

c Power grid exponential

d Social network Gaussian



Transmission lines



Acquaintances



"Natural" networks tend to have...

- One (or a few) connected components.
  - Independent of network size.
- A small diameter ("six degrees of separation").
  - Constant, logarithmically increasing, or even decreasing with network size.
- High clustering ("communities").
  - Much larger than expected from a random network (and, even so, with a small diameter!).
- A mixture of connections.
  - Local vs. "long-distance" connections

Do they share some "universal" features?

### **Network Models**

- Random networks.
- Random-biased networks.
- Small-world networks.
- Scale-free networks.
- Hierarchical & modular networks.
- Affiliation networks.







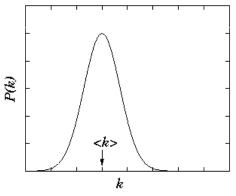


#### **Random Networks**

Erdös-Rényi model

- Small number of connected components (typically one).
- Low clustering coefficient.
- Poisson distribution.



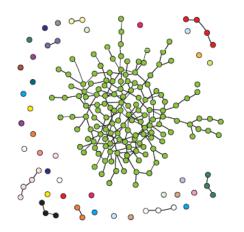


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### **Network Models**

#### **Random Networks**

Erdös-Renyi model



Number of links

Principal component

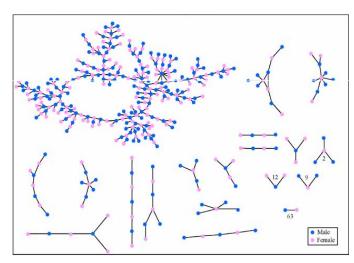






#### **Random Networks**

Example: Romantic relationships in the Add Health data set.





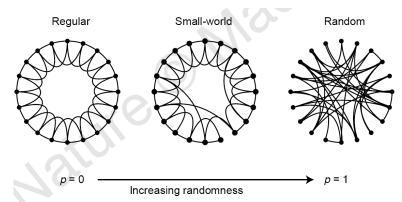
Peter S. Bearman, James Moody & Katherine Stovel: "Chains of Affection: The Structure of Adolescent Romantic and Sexual Networks" American Journal of Sociology, 110(1):44–91, July 2004

### **Network Models**

#### **Small-World Networks**

Watts & Strogatz model

- Small number of connected components (typically one).
- Small diameter.
- Poisson distribution.
- High clustering coefficient.



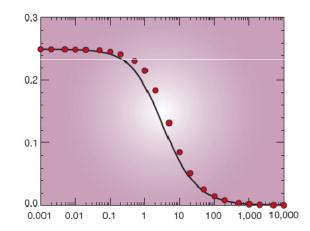






#### **Small-World Networks**

Watts & Strogatz model



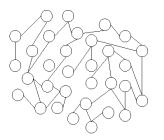
Average path length, normalized by system size, plotted as a function of the average number of shortcuts.



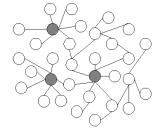
#### **Scale-Free Networks**

Barabási & Albert model

- Small number of connected components (typically one).
- Small diameter.
- Pareto distribution.
- Small clustering coefficient.
- Hubs.



(a) Random network





(b) Scale-free network





#### **Scale-Free Networks**

Barabási & Albert model

"Natural" interpretation of the model:

Variable number of nodes: Network grows as new nodes are added.

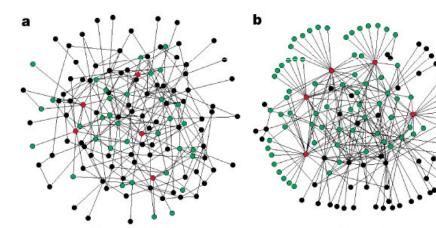
#### Preferential attachment:

The more connected a node is, the more likely it is to receive new links ("rich get richer" or Matthew effect).



#### **Scale-Free Networks**

Barabási & Albert model



Exponential model... ... without hubs.

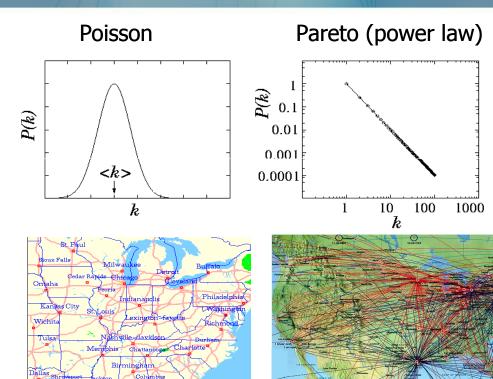
Scale-free model... ... with hubs.











### **Network Models**

#### **Scale-Free Networks**

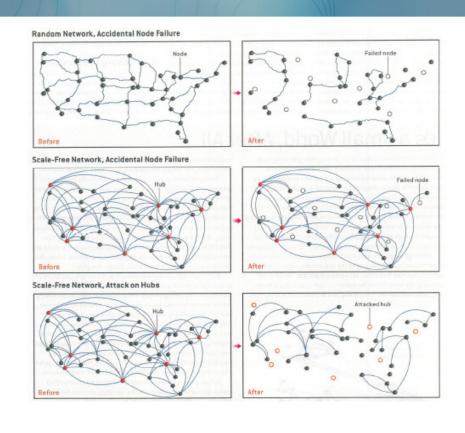
Features

- Self-organization traits: Links are not random (a feature found in many complex systems).
- Tolerance to random attacks, which easily disrupt random networks but not scale-free networks.
- Vulnerability to targeted attacks: "Hubs" are essential to maintain connectedness.





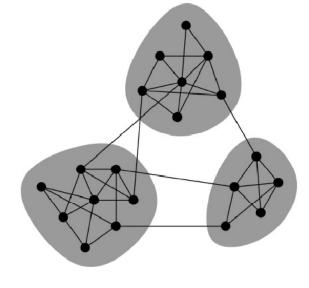




### Network Models

#### **Hierarchical/Modular Networks**

- Hierarchical organization.
- Hubs.
- Cliques.

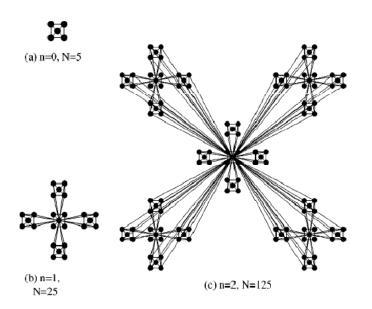






# A PARTICIPAL DESISTER

#### **Hierarchical/Modular Networks**

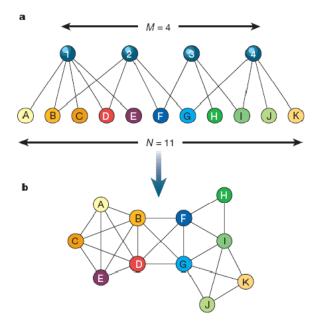




### **Network Models**

#### **Affiliation Networks**

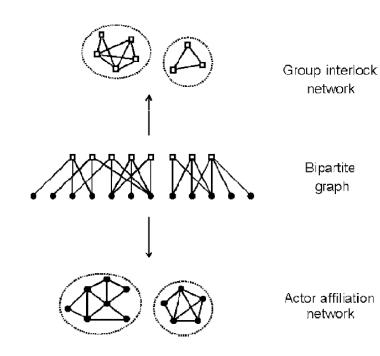
Bipartite graph to model social interactions:



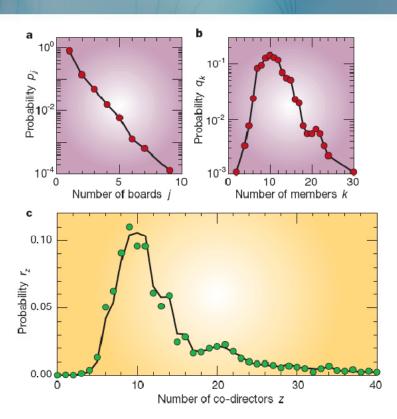




#### **Affiliation Networks**



Network Models











# Network Structure & Dynamics

The countless ways in which network structures affect our lives make it critical to understand:

1. How network structures affect behavior.

2. Which network structures are likely to emerge.



# Network Structure & Dynamics

A complex system is a system composed of interconnected parts that, as a whole, exhibit one or more properties (behavior) not obvious from the properties of the individual parts (i.e. emergence).



### Network Structure & Dynamics

#### **Research problems**

- Search on networks (with partial local information)
- Diffusion problems: epidemics, social contagion (ideas, fads, products...)
- Analysis of network properties e.g. robustness/vulnerability

# Network Structure & Dynamics

#### From an algorithmic point of view...

- Objects:
  - Ranking (HITS, PageRank...).
  - Classification & anomaly detection.
  - Clustering & community analysis.
  - Object identification (e.g. "entity resolution").
- Links:
  - Link prediction.
- Graphs:
  - Subgraph detection.
  - Graph classification.
  - Graph generation models.

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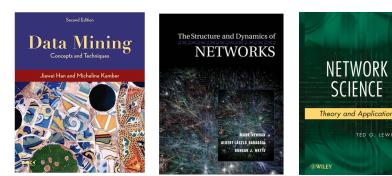


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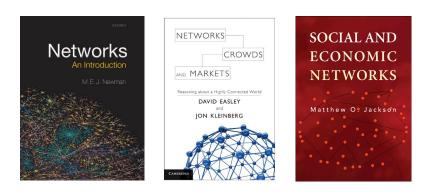




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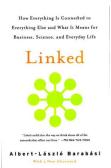


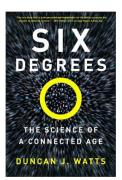
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Albert-László Barabási Author of LINKED



