Complex Networks Complexity, Complex Systems (graphs)

What are they, history, examples, pictorials, current problems
How are they related to Physics, Sociology, Economics.....

graphs ~1930s

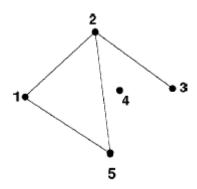
networks ~ 1998

Panos Argyrakis
University of Thessaloniki

Characteristics of networks:

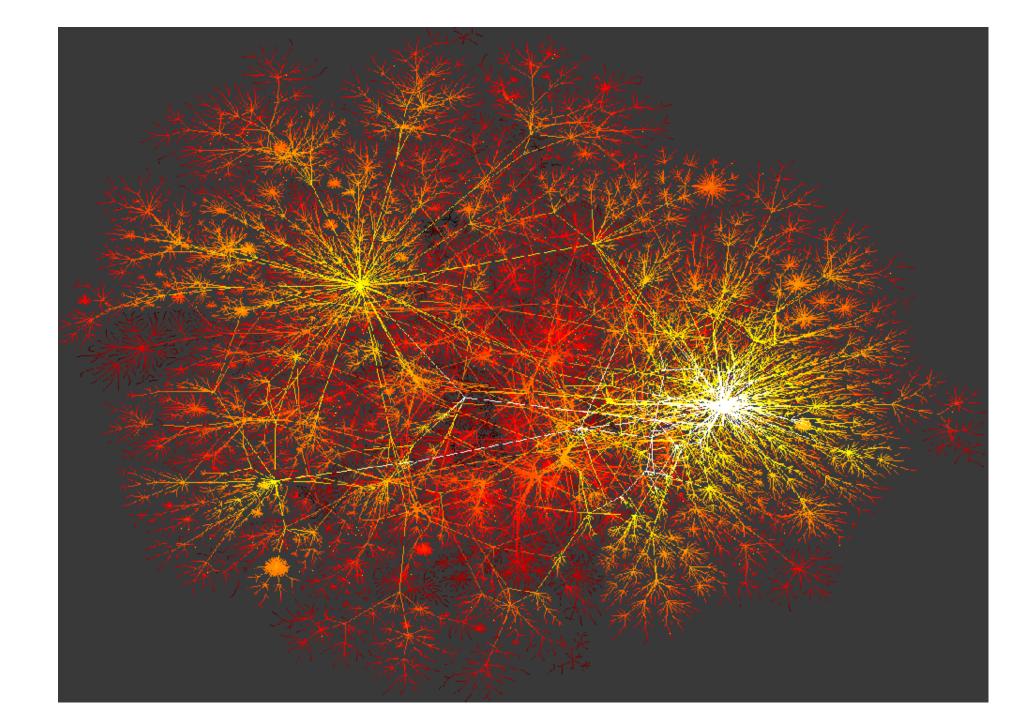
Structures that are formed by two distinct entities:

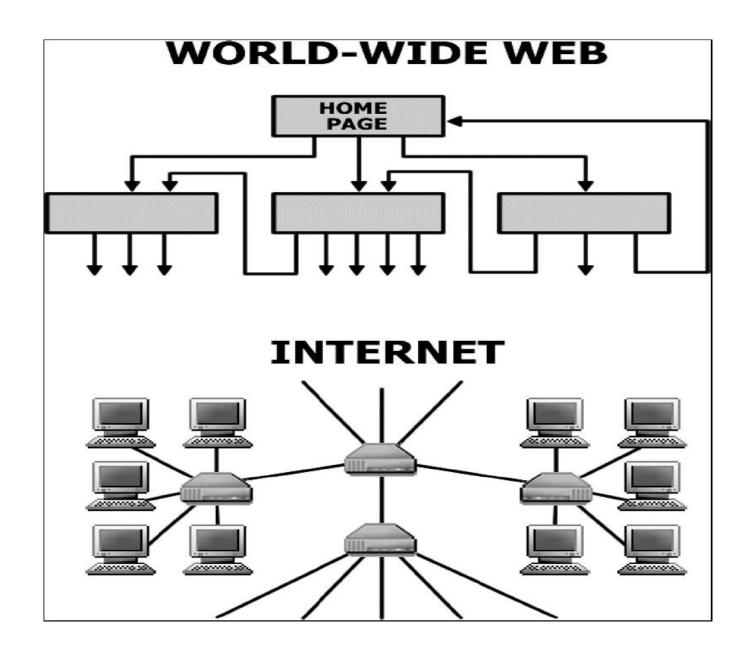
nodes, sites, vertices, etc. connections, edges, links, synapses, etc.



N= 5 nodes, n=4 connections

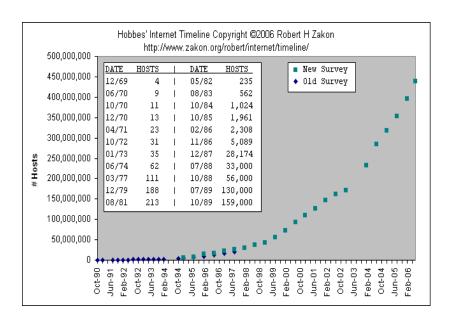
- They can be simple structures or very complicated (belonging to the class of complex systems)
- How did it all start?



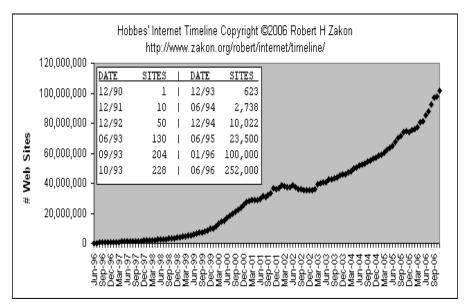


Internet and WWW

1970 – 10 H/Y 1990 – 1.75*10⁵ H/Y σήμερα – 1.2*10⁹ H/Y



1990 - 1 web site $1996 - 10^5$ web sites $σήμερα - 10^9$ web sites



Number of units worldwide

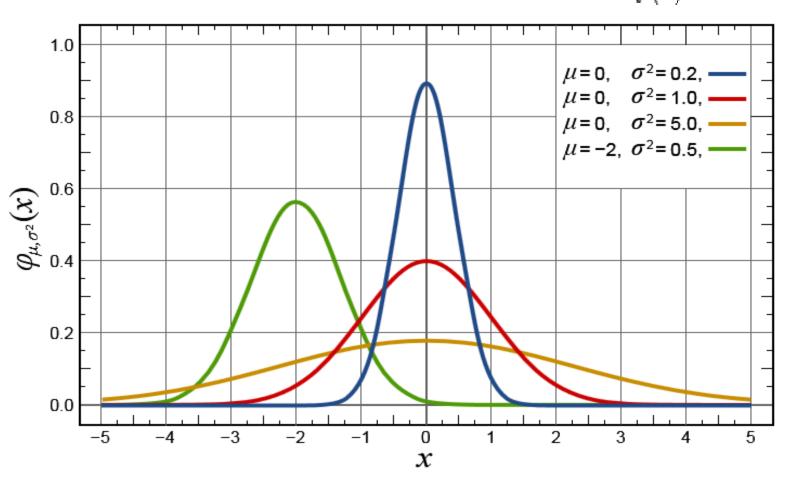
- Population of Earth
- Number of computers
- Facebook accounts
- Number of cell phones

- ~ 7 billion
- ~ 1.2 billion
- ~ 1.1 billion
- ~ 8 billion in 2016

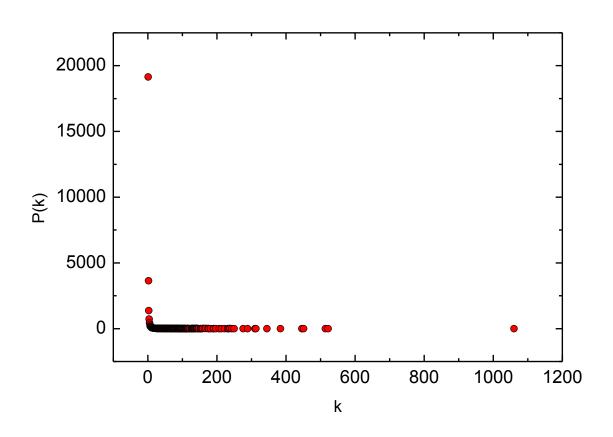
Connectivity between nodes in the Internet

- •How are the different nodes in the Internet connected?
- No real top-down approach ever existed
- ·Has been characterized by a large degree of randomness
- •Question of interest: What is the probability distribution function (PDF) of connectivities of all nodes?
- (i.e. how many connections does each node have)?
- Naïve approach: Normal distribution (Gaussian)
- Experimental result?
- •Studied ~1999

Gaussian distribution $f(x) = ae^{-\frac{(x-b)^2}{2c^2}}$



Probability distribution P(k) of the No. of connections (degree distribution)



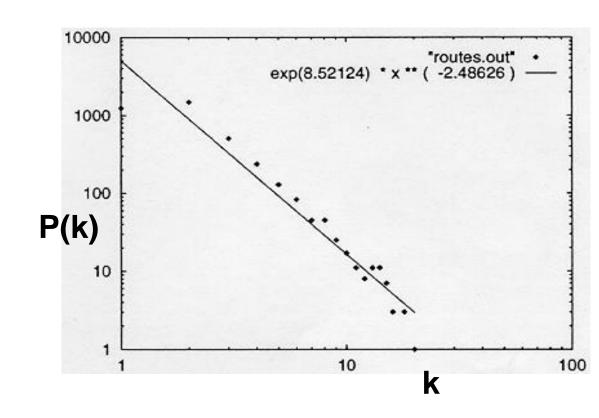
INTERNET BACKBONE

Nodes: computers, routers

Links: physical lines

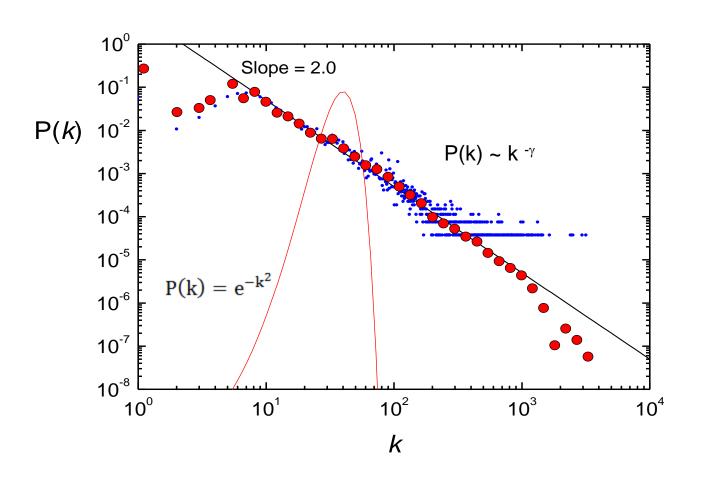
$$P(k) \sim k^{-\gamma}$$

k=number of connections of a node P(k)= the distribution of k

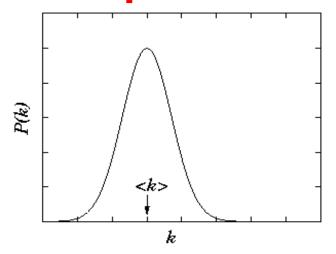


(Faloutsos, Faloutsos and Faloutsos, 1999)

Degree distribution in a scale-free network



What did we expect?



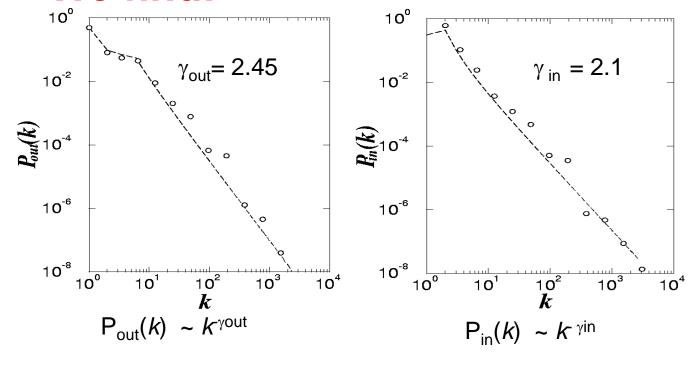
$$\langle k \rangle \sim 6$$

$$P(k=500) \sim 10^{-99}$$

$$N_{WWW} \sim 10^9$$

$$\Rightarrow$$
 N(k=500)~10⁻⁹⁰

We find:



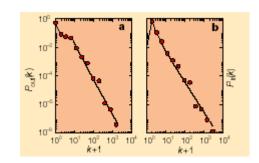
$$P(k=500) \sim 10^{-6}$$

$$N_{WWW} \sim 10^9$$

 $\Rightarrow N(k=500) \sim 10^3$

WWW

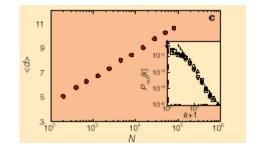
It has a power-law



Diameter of WWW:

$$<$$
d> $>=0.35 + 2.06 logN$

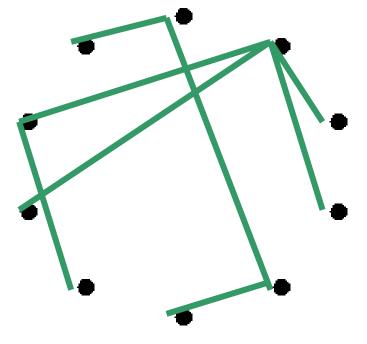
For: $N=8x10^8 \rightarrow <d>=18.59$



[compare to a square of same size: edge length=30000

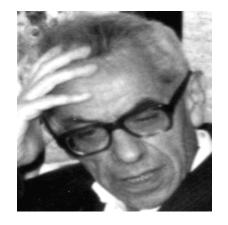
For a cube: edge length=1000]

The model of Erdös-Rényi (1960)



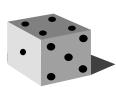
Connections
with
probability p
p=1/6
N=10

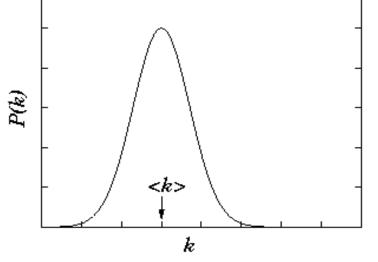
 $\langle k \rangle \sim 1.5$



Pál Erdös (1913-1996)

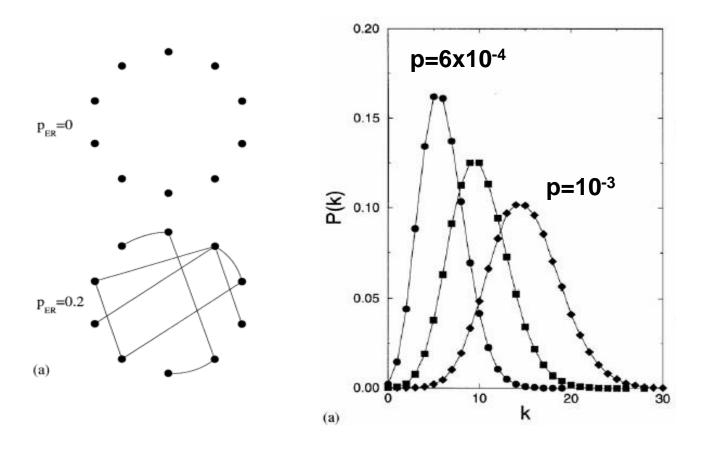
Poisson distribution



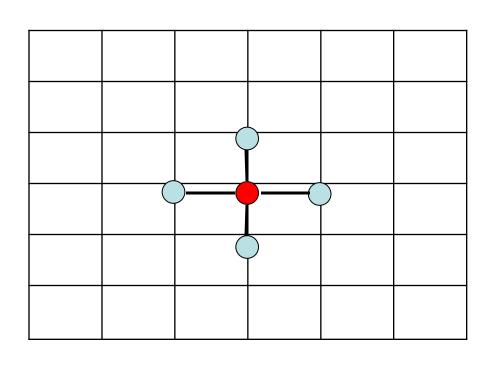


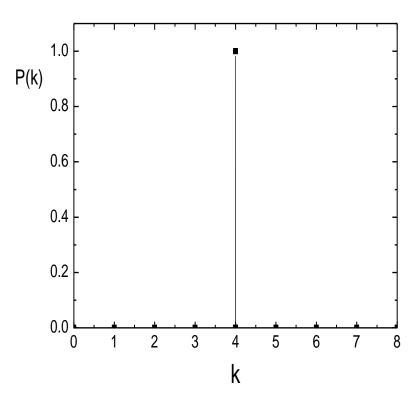
- Democratic
- Random

The Erdos-Renyi model

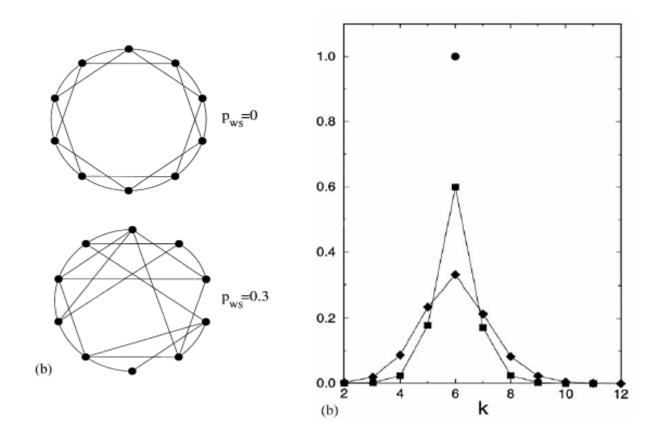


Regular network

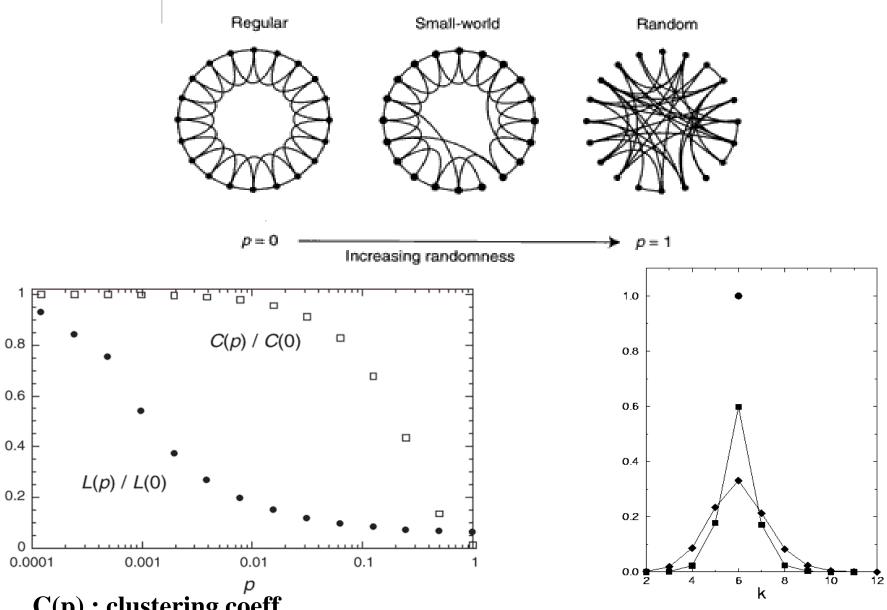




Small world network



The Watts-Strogatz model

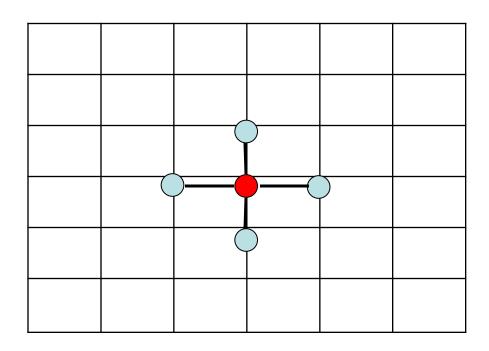


C(p): clustering coeff.

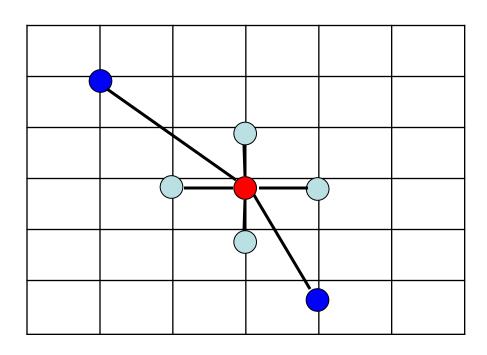
(Watts and Strogatz, Nature **393**, 440 (1998))

L(p): average path length

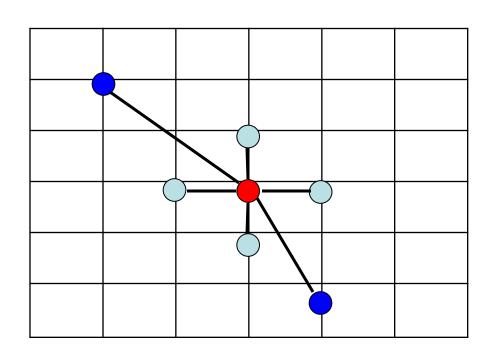
Small world network

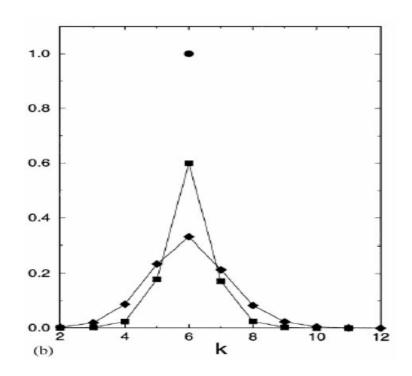


Small world network



Small-world network





Most real networks have similar internal structure:

Scale-free networks

Why?

What does it mean?

Scale-free networks

(1) The number of nodes is not pre-determined

The networks continuously expand with the addition of new nodes

Example:

WWW: addition of new pages

Citation: publication of new articles

(2) The additions are not uniform

A node that already has a large number of connections is connected with larger probability than another node.

Example:

WWW: new topics usually go to well-known sites

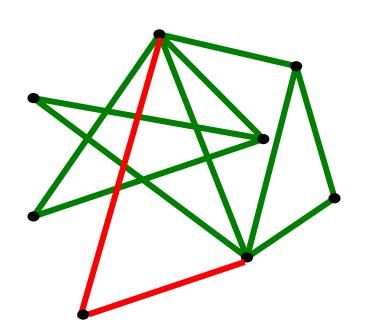
(CNN, YAHOO, NewYork Times, etc)

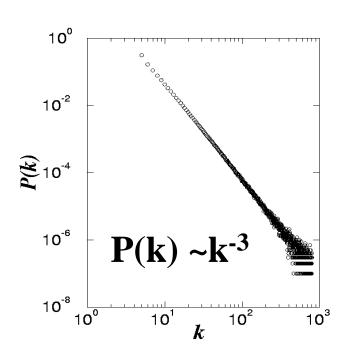
Citation: papers that have a large number of references are more

probable to be refered again

Scale-free networks

- (1) Growth: At every moment we add a new node with *m connections* (which is added to the already existing nodes).
- (2) Preferential Attachment: The probability Π that a new node will be connected to node i depends on the number k_i , the number of connections of this node $\Pi(k_i) = \frac{k_i}{\sum_j k_j}$



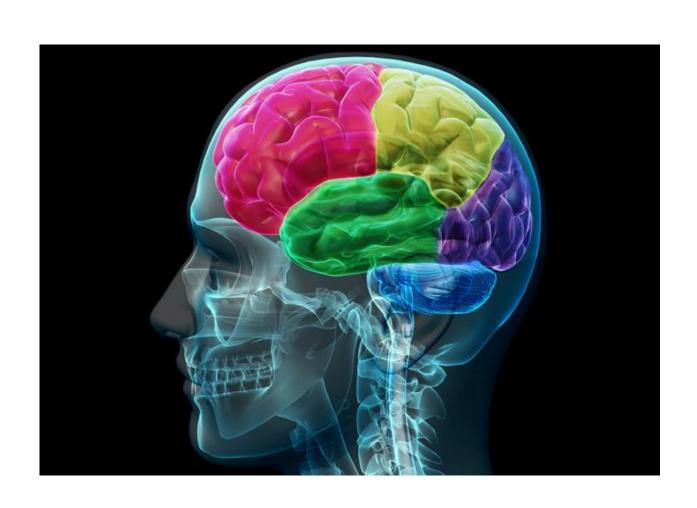


A.-L.Barabási, R. Albert, Science **286**, 509 (1999)

Network categories:

- (1) Random network (Erdos-Renyi)
- (2) Network on a regular lattice
- (2) Small world network (Strogatz-Watts)
- (3) Network with a power-law

Human Brain



Cells: Neurons



Human Brain:

- Nodes: cells=neurons
- Links: dendrites
- No. of nodes: 10,000,000,000, i.e. 10 billion
- All neurons are different, i.e. no two neurons are identical
- Each neuron has ~k= 10,000 connections (synapses)
- 10 billion nodes x 10,000 links each = total 10^{+14} links
- 100 trillion connections

Dendrimer molecules

Dendrimers

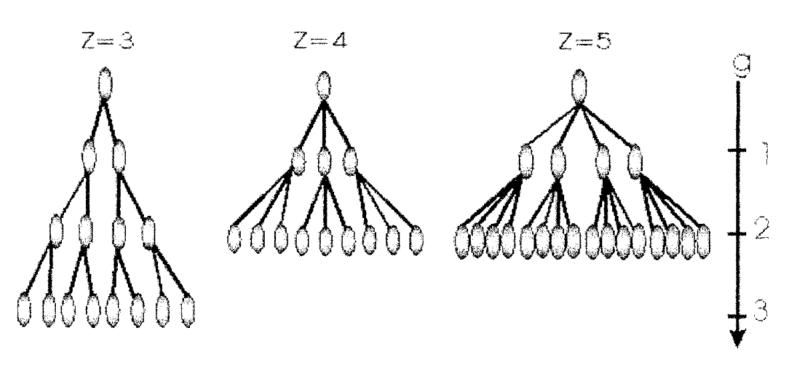
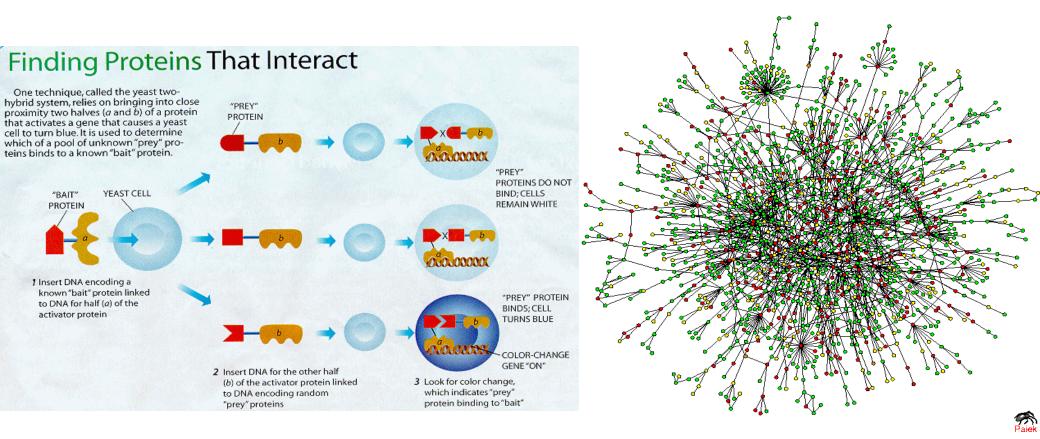


Fig. 1. Schematic of three dendrimers, with z = 3, 4, and 5. The generation order, g, is as shown, g = 3 (for z = 3), and g = 2 (for z = 4 and 5).

Yeast protein network

Nodes: proteins

Links: physical interactions (binding)

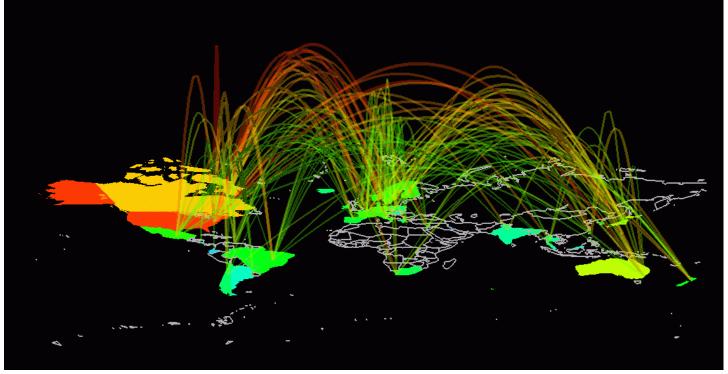


P. Uetz, et al. Nature 403, 623-7 (2000).

Communication networks

The Earth is developing an electronic nervous system, a network with diverse <u>nodes</u> and <u>links</u> are

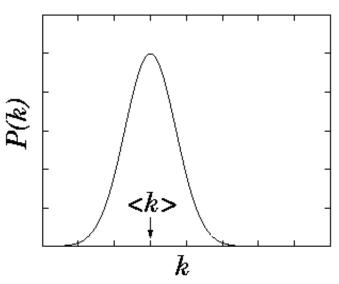
-computers -phone lines
-routers -TV cables
-satellites -EM waves



Communication networks: Many non-identical components with diverse connections between them.

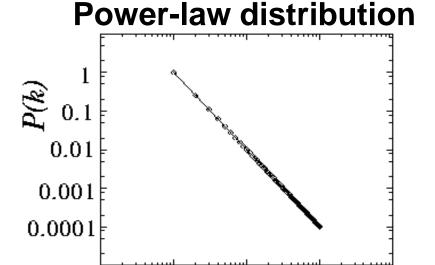
What does it mean?

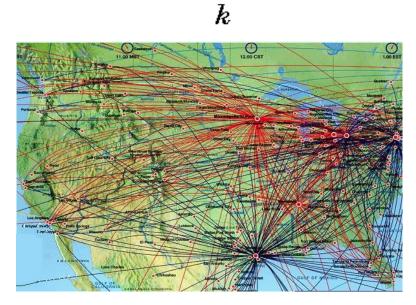
Poisson distribution





Exponential Network





10

100

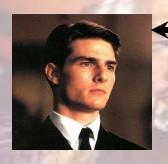
1000

Scale-free Network

ACTOR CONNECTIVITIES

cast jointly





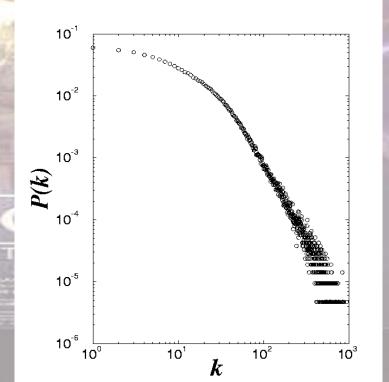
Days of Thunder (1990)

Far and Away (1992) Eyes Wide Shut (1999)



N = 212,250 actors $\langle \mathbf{k} \rangle = 28.78$

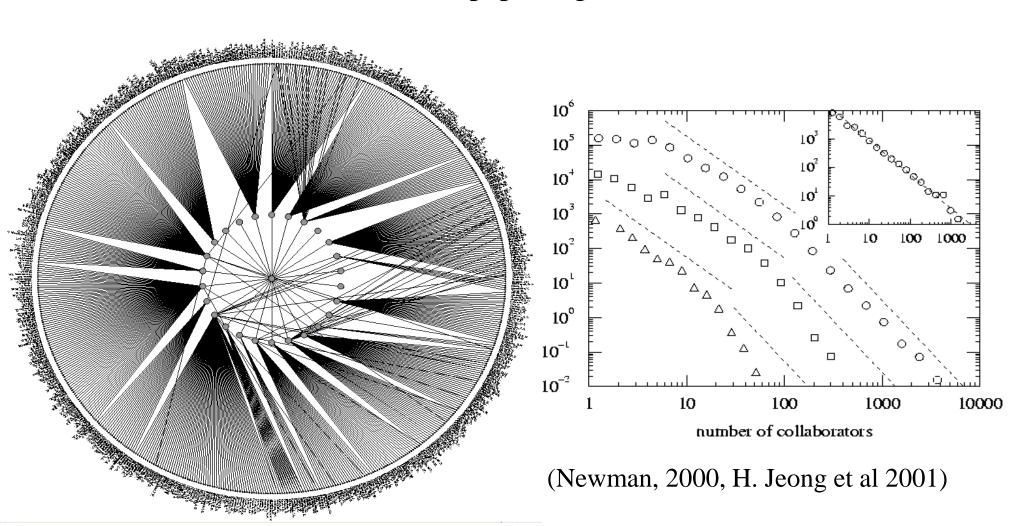
 $P(k) \sim k^{-\gamma}$



SCIENCE COAUTHORSHIP

Nodes: scientist (authors)

Links: write paper together



SCIENCE CITATION INDEX

1,000 Most Cited Physicists, 1981-June 1997

Out of over 500,000 Examined (see http://www.sst.nrel.gov)

Author name		Institute	Country	Field	avg.	total art.	total cites	rank by total cit
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^{*} citation total may be skewed because of multiple authors with the same name

Society

Nodes: individuals

<u>Links</u>: social relationship (family/work/friendship/etc.)

S. Milgram (1967)

John Guare



How many (n) connections are needed so that an individual is connected with any other person in the world?

N=6 billion people

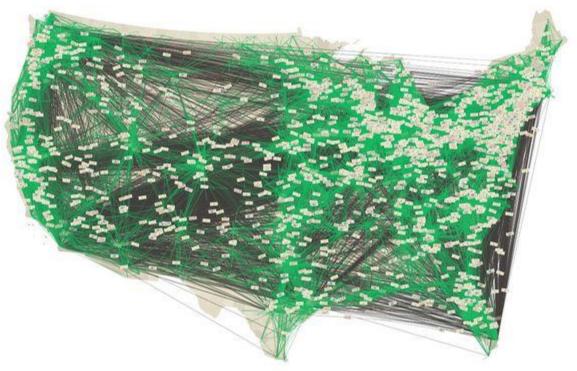
Result: n~6

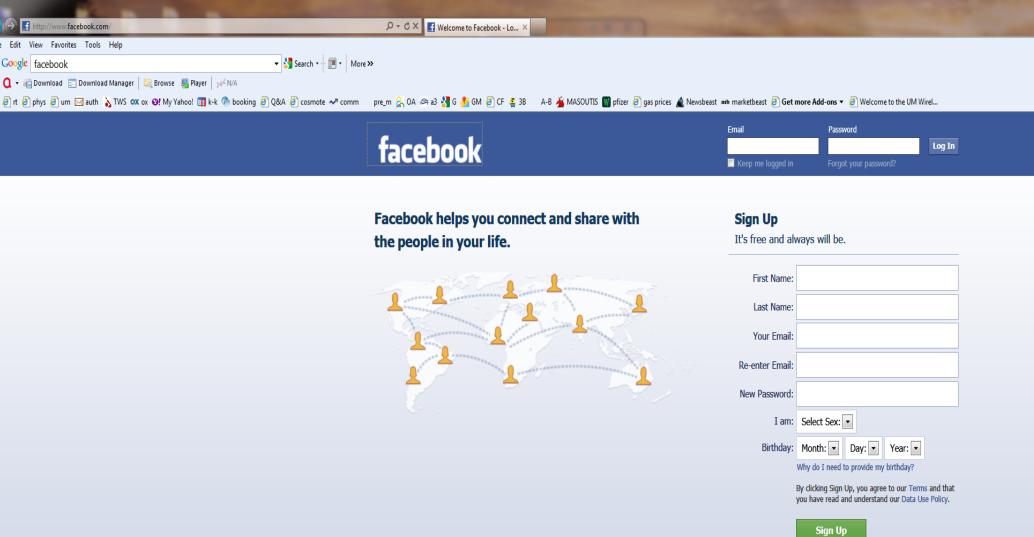
Conclusion: We live in a small world

Six Degrees of Separation!!

Where is George?







Create a Page for a celebrity, band or business.

English (US) Ελληνικό Español Português (Brasil) Français (France) Deutsch Italiano العربية (育成 中文(简体) ...

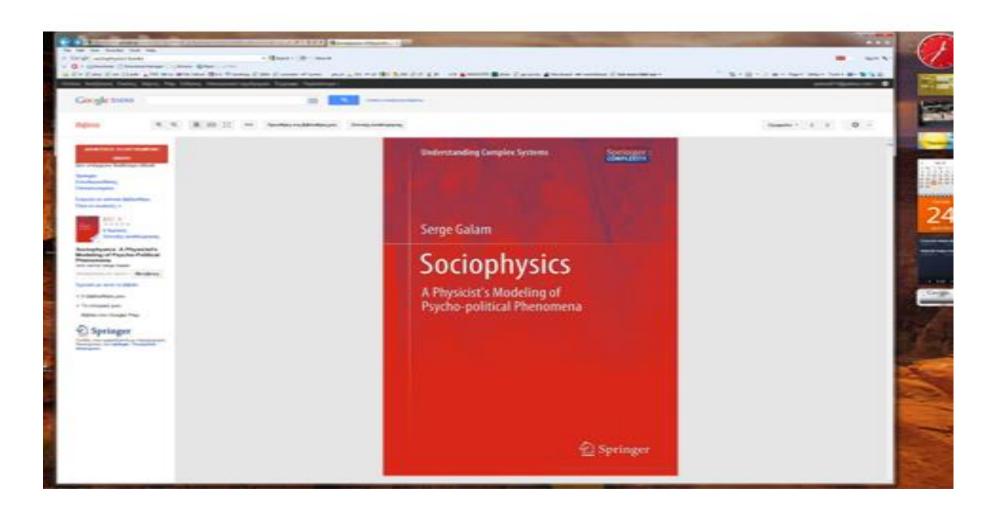
Facebook:

- •1,000,000,000 registered users
- •50,000,000 active users
- •5,000,000 generate 95% of the traffic

Questions needing answers:

- How many people communicate?
- •How many connections does one have?
- •How often does he communicate?
- •How long does it last?

New books:



Real-world phenomena related to communication patterns

- Crowd behavior: strategies to evacuate people and stop panic
- Search strategies: efficient networks for searching objects and people
- Traffic flow: optimization of collective flow
- Dynamics of collaboration: human relationship networks such as collaboration, opinion propagation and email networks
- Spread of epidemics: efficient immunization strategies
- Patterns in economics and finance: dynamic patterns in other disciplines, such as Economics and Finance, and Environmental networks

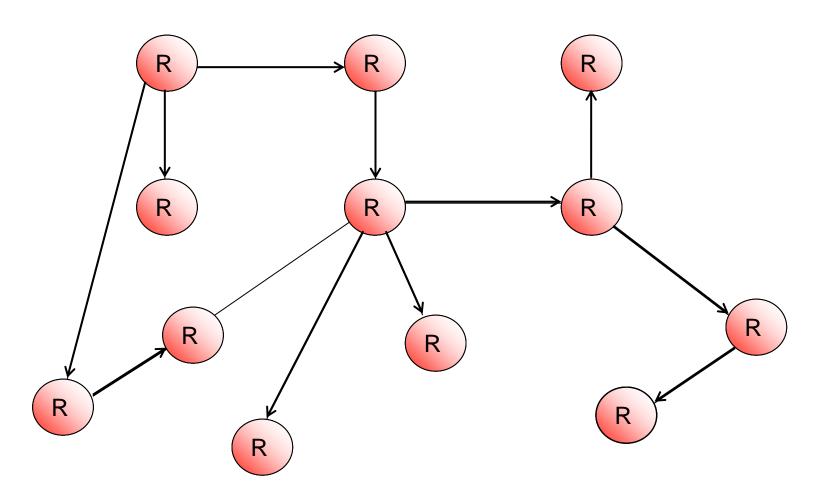
Spreading phenomena

- Starting at a random node
- Moving to nearest neighbors
 - use: Breadth First Search Algorithm
- SIR (susceptible-infected-removed/recovered)
- SIS (susceptible-infected-susceptible)
- SIRS (susceptible-infected-refractorysusceptible)
- Applications: Forest fires, epidemics, rumor spreading, virus spreading, etc.

The SIR model

- SIR (Susceptible, Infected, Removed),
- q=probability of infection
- Initially all nodes are susceptible (S)
- Then, a random node is infected (I)
- The virus is spread in the network, all I nodes become R
- This process continues until the virus either
 - has been spread in the entire network, or
 - has been totally eliminated
- M=infected mass
- Duration

SIR model





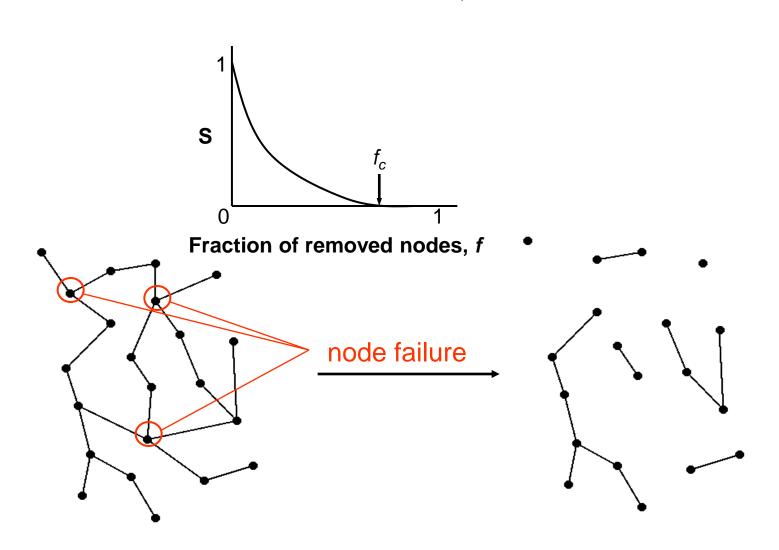


Immunization p_c =critical point for transmission

- Complete immunization of populations is not feasible (p_c can not be made significantly lower than 1)
- Try to find an efficient method of immunization
- We seek strategic ways of immunizing,
 e.g. try to isolate via immunization the largest possible
 portion of the population instead
 - New strategies may be needed

Robustness of networks

Complex systems maintain their basic functions even under errors and failures (cell \rightarrow mutations; Internet \rightarrow router breakdowns)



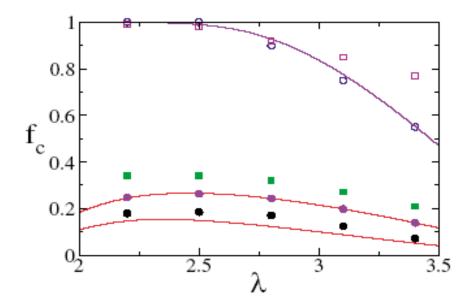
Attack tolerance

- The stability of the networks under failure or attack is very important
- In general, the integrity is destroyed after a critical percentage p_c of the nodes has been removed (no giant cluster).
- Scale-free networks are: extremely robust under random failure (p_c =0.99), very vulnerable under targeted attacks (p_c =0.07).
- The existence of a spanning cluster is based on the criterion

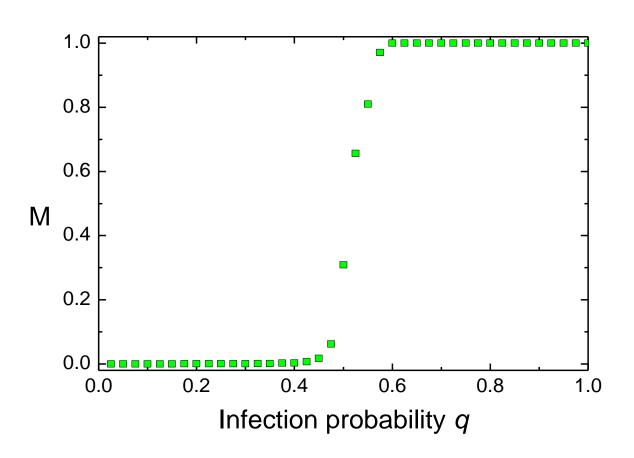
$$\kappa \equiv \frac{\left\langle k^2 \right\rangle}{\left\langle k \right\rangle} = 2$$

Acquaintance immunization

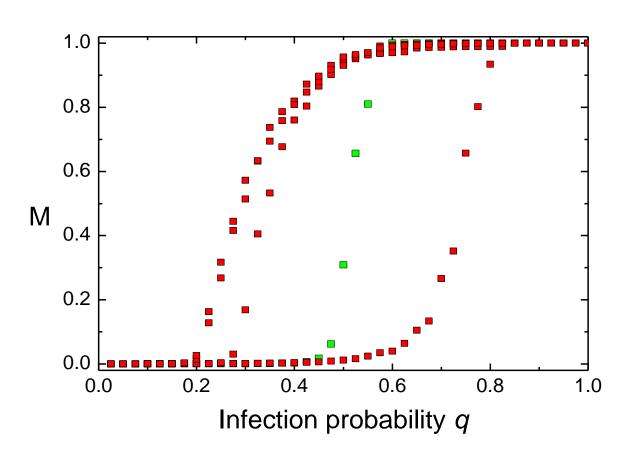
- Suggested by Cohen, Havlin, ben-Avraham (PRL, 2003)
- Strategy: Randomly choose a node and immunize a random neighbor of this node.



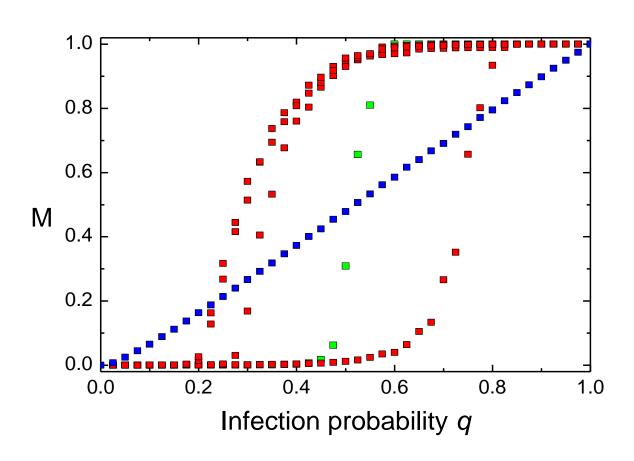
Comparison of the 3 network types



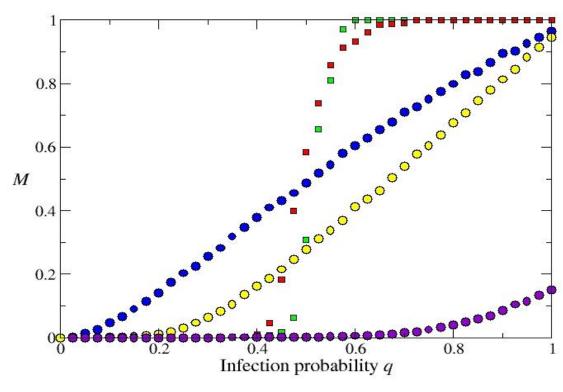
Comparison of the 3 network types



Comparison of the 3 network types



Comparison of different network types

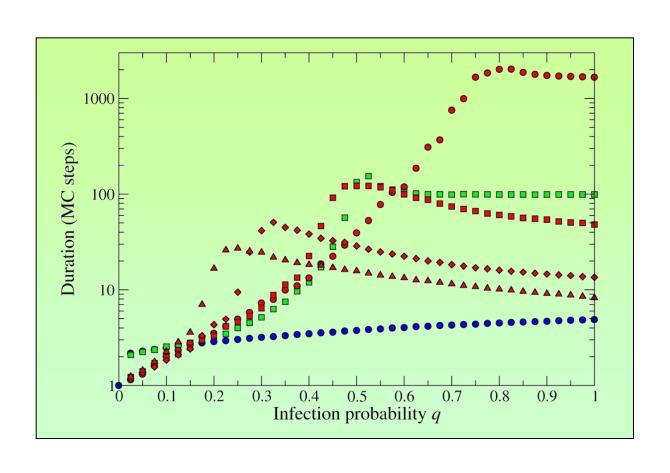


Small-world network

Lattice

- **Scale-free network**, γ =2.0
- \bigcirc Scale-free network, γ =2.5
- **Scale-free network**, γ **=3.0**

Duration of epidemics



DISEASE SPREADING:

The dynamical consequences of a scale-free network are drastically different than those on lattice and small-world networks.

There is no threshold in the infected mass as a function of the infection transmission probability.

The starting point of the disease is important, since it determines whether the disease will spread or die out.

The spreading is rapid and manifests the small diameter of the network.

ATTACK TOLERANCE:

The tolerance of a network depends on its connectivity.

The random node removal is the marginal case where the critical threshold moves from p_c =0 to p_c =1.

A small bias in the probability of selecting nodes either retains or destroys the compactness of the cluster.

SIR model:

- The network with a power-law structure is a more realistic representation than all the other network categories
- It does NOT show critical behavior

What is Econophysics?

"Econophysics is the application of typical methods from physics to the study of the financial markets, seen as a complex system."

H.E. Stanley, Boston University, Boltzmann Medal 2004:

"For his influential contribution to several areas of statistical physics..."

However...it is nothing really that new!

E. Majorana, **Scientia**, Vol. 36, 58 (1942) "On the value of the statistical laws in physics and in social sciences." Physica A, Vol. 285, p. 1 (2000)

Exotic statistical physics with applications to biology, medicine and economics.



Textbooks in Econophysics

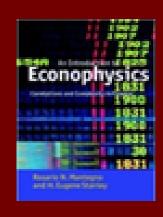
J.P. Bouchaud and M. Potters — Theory of Financial Risk and Derivative Pricing: from Statistical Physics to Risk Management, Cambridge University Press (2003)



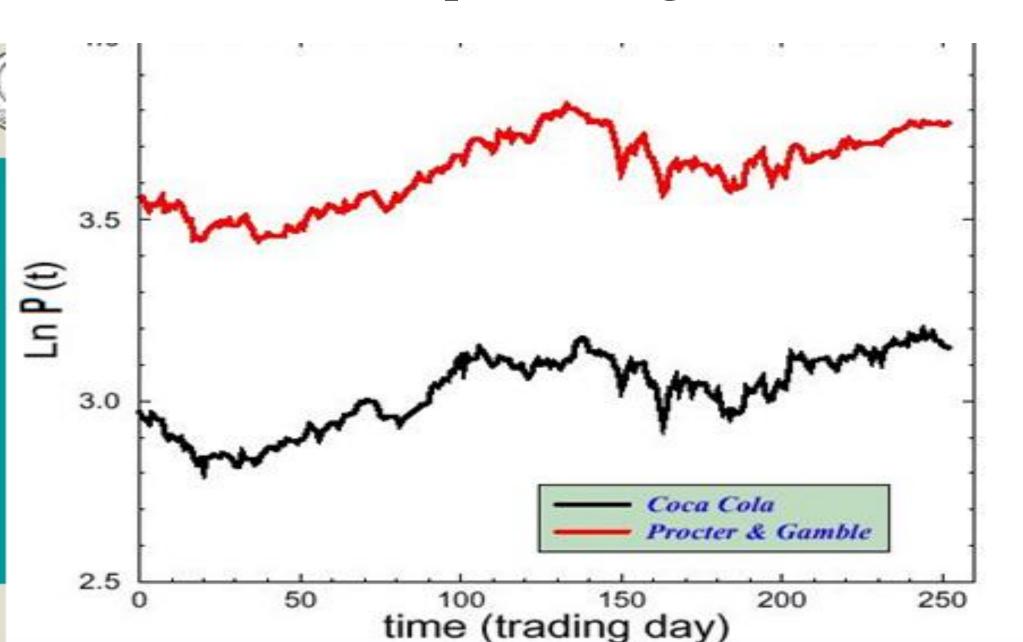


J. Voit – The Statistical Mechanics of Financial Markets, Springer (2005)

R.N. Mantegna and H.E. Stanley – An Introduction to Econophysics: Correlations and Complexity in Finance, Cambridge University Press (2000)



Stock price changes



Financial Data

Price Index

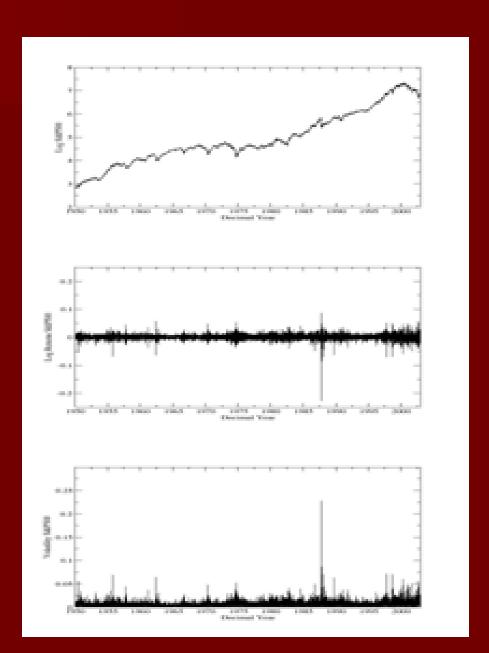
Logarithmic Price Returns

$$R(t) = \ln[P(t+1)] - \ln[P(t)]$$

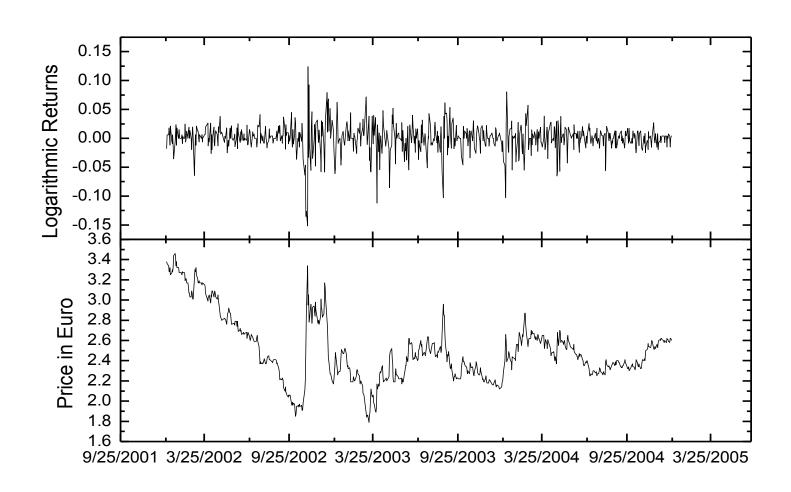
Volatility

$$v(t) = |R(t)|$$

Standard & Poor 500 (S&P500) data set from 3/1/1950 to 18/7/2003. N=13468.



Financial Time Series



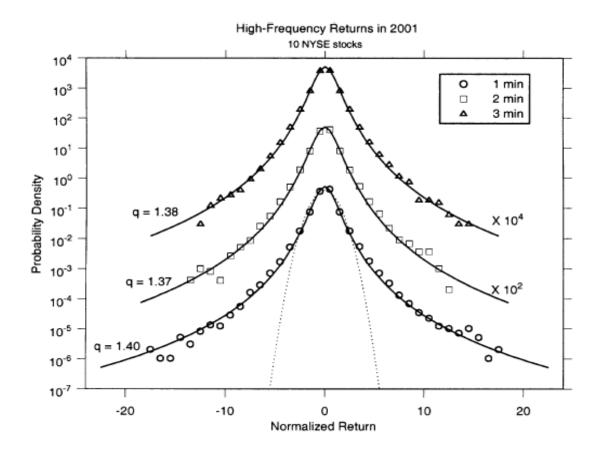
Theoretical Economics is dominated by pure mathematics:

- Lemma/theorem style is required
- Little effort to compare theoretical predictions to "experimental data" say, price record from real stock markets
- Bulk of papers are inaccessible and of no interest to "experimentalists" - practitioners of the field

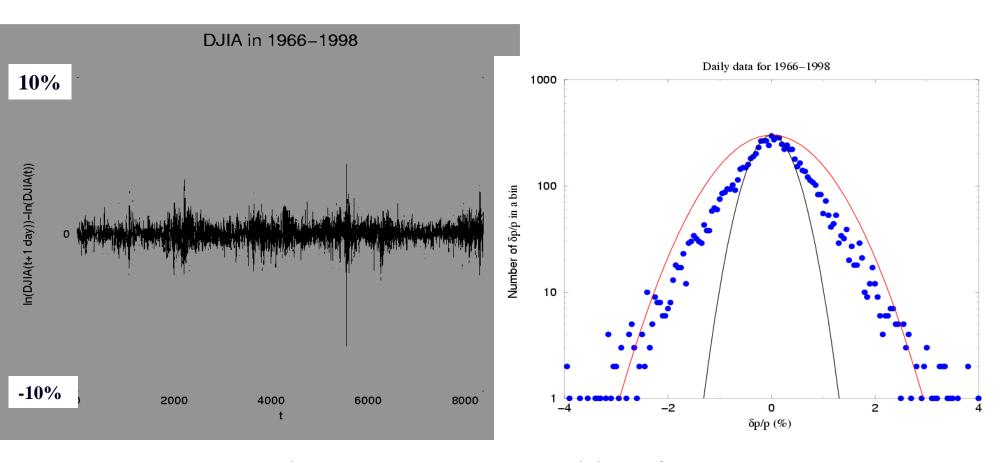
George Soros on theoretical economics

"Existing theories about the behavior of stock prices are remarkably inadequate. They are of so little value to the practitioner that I am not even fully familiar with them. The fact that I could get by without them speaks for itself."

G. Soros, "Alchemy of Finance" 1994



Quick experiment: free data from www.nyse.com/marketinfo/nysestatistics.html

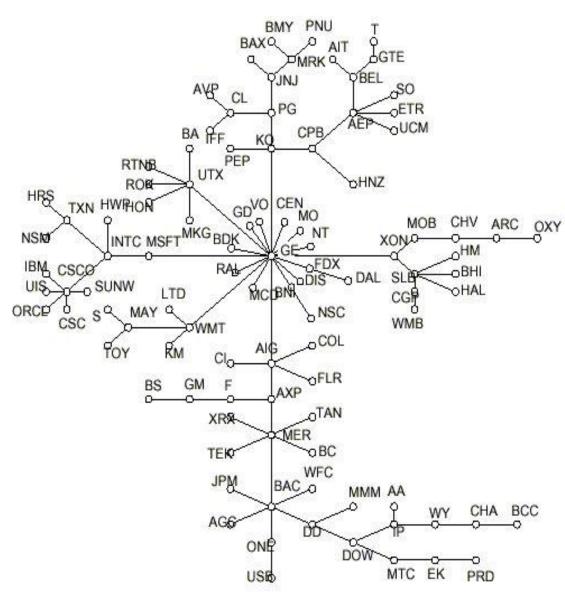


In a gaussian world the probability of the October 1987 crash would be 10⁻¹³⁵!

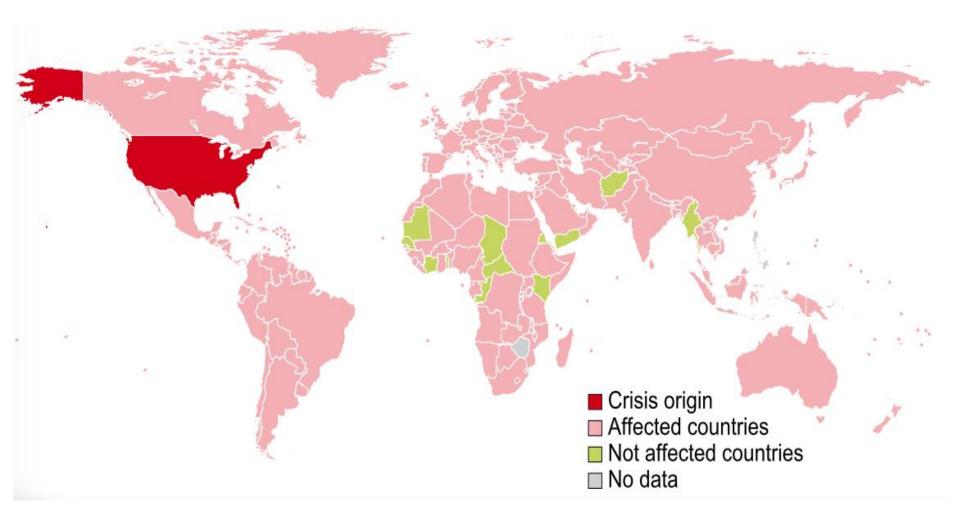
Filtering of the correlation matrix

•Minimum Spanning Tree of the 100 most capitalized US stocks in 1998 (R.N.Mantegna).

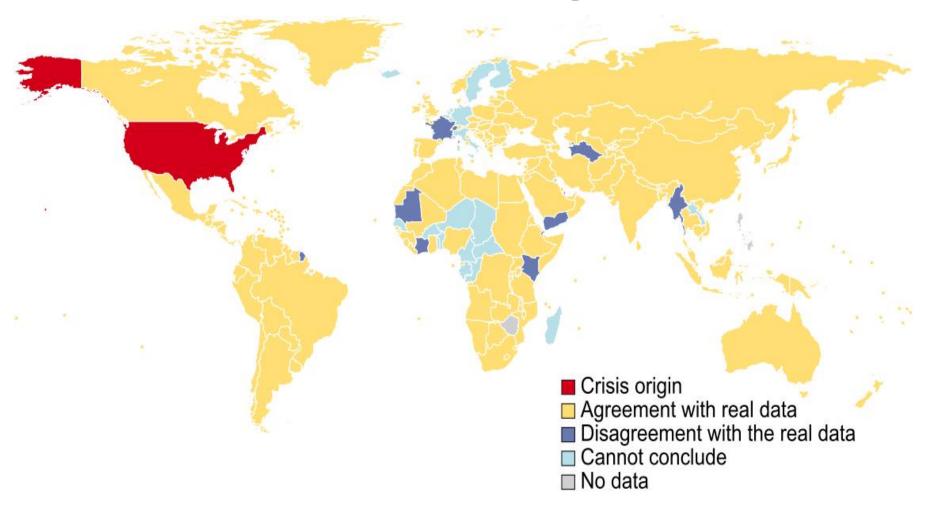
•From n(n-1)/2 connections only n-1 survive.



Countries affected by the 2008 economic crisis



Model results: a crisis starts in US and spreads through CON



Summary- Conclusions

- Networks is a new area in sciences which started from Physics, but pertains to ALL sciences today
- It shows rich dynamical behavior
- It has many-many applications in everyday life
- Will influence directly the way we live and act