

M1 CompuPhys 2022-2023  
Introduction to Python Language

# ***Network Analysis***



Célestin Coquidé  
PostDoc at UTINAM (Observatory building office n°13)

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## III Centrality Measures:

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- III.2) Drawing networks

## IV Data and network Analysis with Python:

- IV.1) Packages of interest
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# Lectures and Practical Works

Lectures

26/10 1.5 h

28/10 1.5 h

Practicing

17/11 3h<sup>1</sup>

24/11 3h<sup>2</sup>

<sup>1</sup>: Handling of different packages + Exercises

<sup>2</sup>: Preparation for a graded homework

# I Introduction







ARTWORK: TAMAR COHEN, ANDREW J BUBOLTZ, 2011, SILK SCREEN ON A PAGE FROM A HIGH SCHOOL YEARBOOK, 8.5" X 12"

DATA

## Data Scientist: The Sexiest Job of the 21st Century

by Thomas H. Davenport and D.J. Patil

FROM THE OCTOBER 2012 ISSUE

Forbes

Dites Oui à une réduction des coûts pouvant atteindre 38%\*, avec HPE Flexible Capacity.

Évoluez avec l'IT hybride

[\\* Plus d'information](#)

Leadership / #IfOnlyKnew  
SEP 21, 2017 @ 03:45 PM 15,659

## Becoming A Data Scientist: The Skills That Can Make You The Most Money



Karsten Strauss, FORBES STAFF  
[FULL BIO](#)



listings. The ten skills that appeared most often as prerequisites for the job, and the percentage of job listings in which they appeared, were:

1. Python (72%)
2. R (64%)
3. SQL (51%)
4. Hadoop (39%)
5. Java (33%)

URGENT

## Data surge: Demand for data scientists set to rise in the coming years

Craig Guillot CTW Features Sep 17, 2017 Updated Sep 28, 2017



PROMOTION

### Carriers Needed

Earn extra cash! Earn \$300 per week!



PROMOTION WEB ONLY

### Special Report: addiction in NW

## M CAMPUS

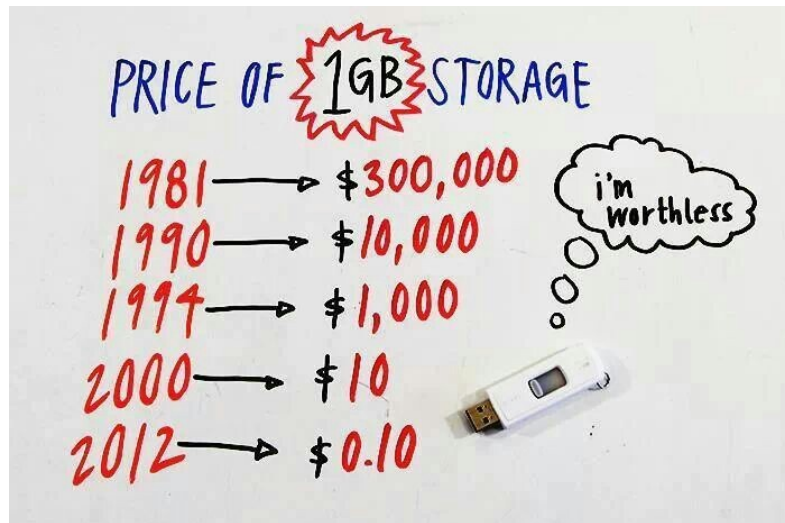
## Le « data scientist », nouvelle coqueluche des recruteurs

D'après notre baromètre en partenariat avec LinkedIn, les entreprises courtisent ces « scientifiques des données » pour leurs compétences en termes de transformation numérique

# Data science

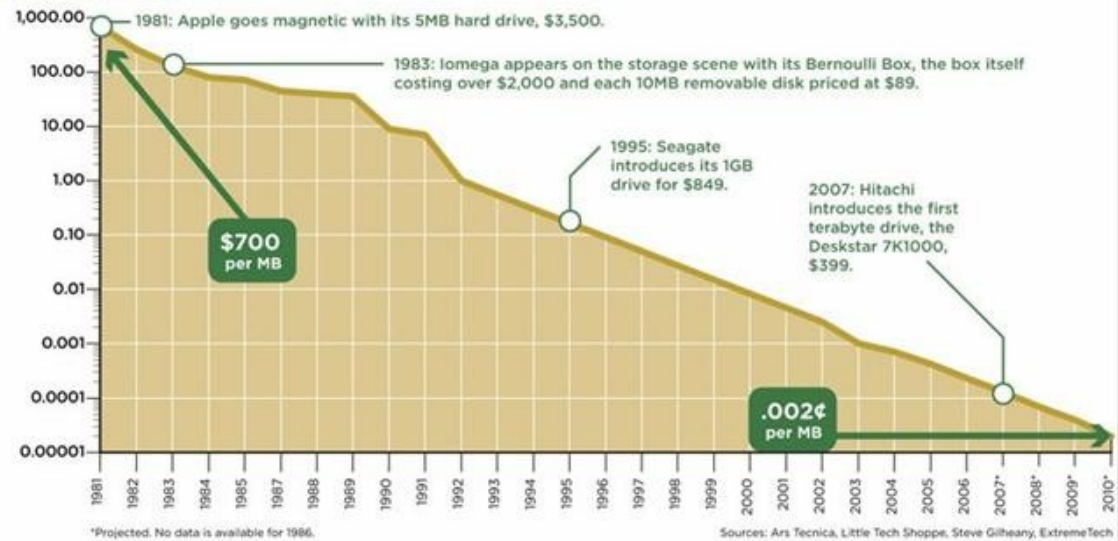
- A new discipline
- Few amount of book with a global and unified sight

# Evolutions in data

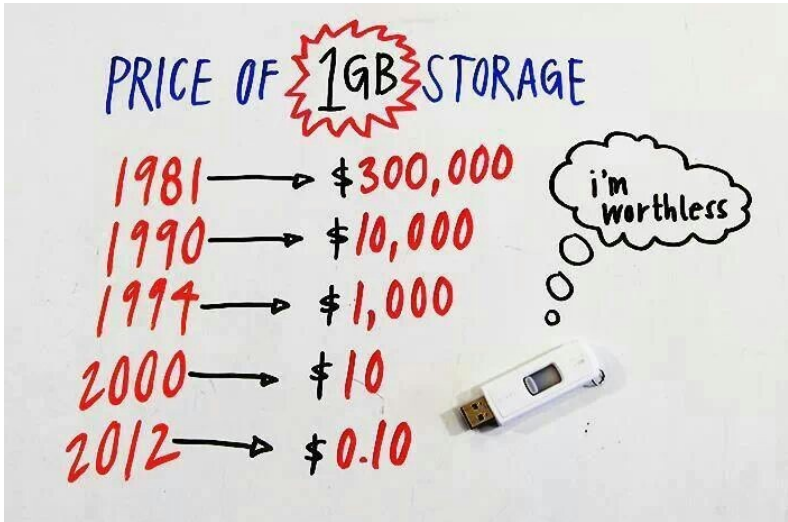


## STORAGE: FROM HIGHWAY ROBBERY TO RUNAWAY BARGAIN

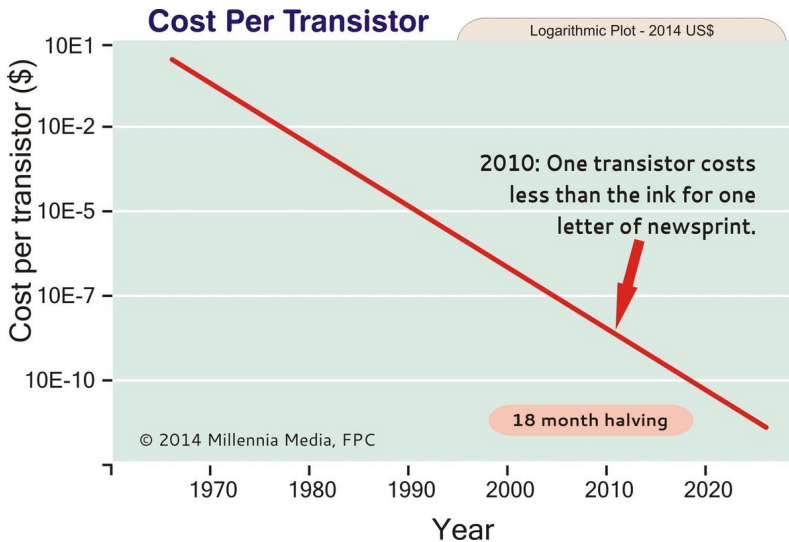
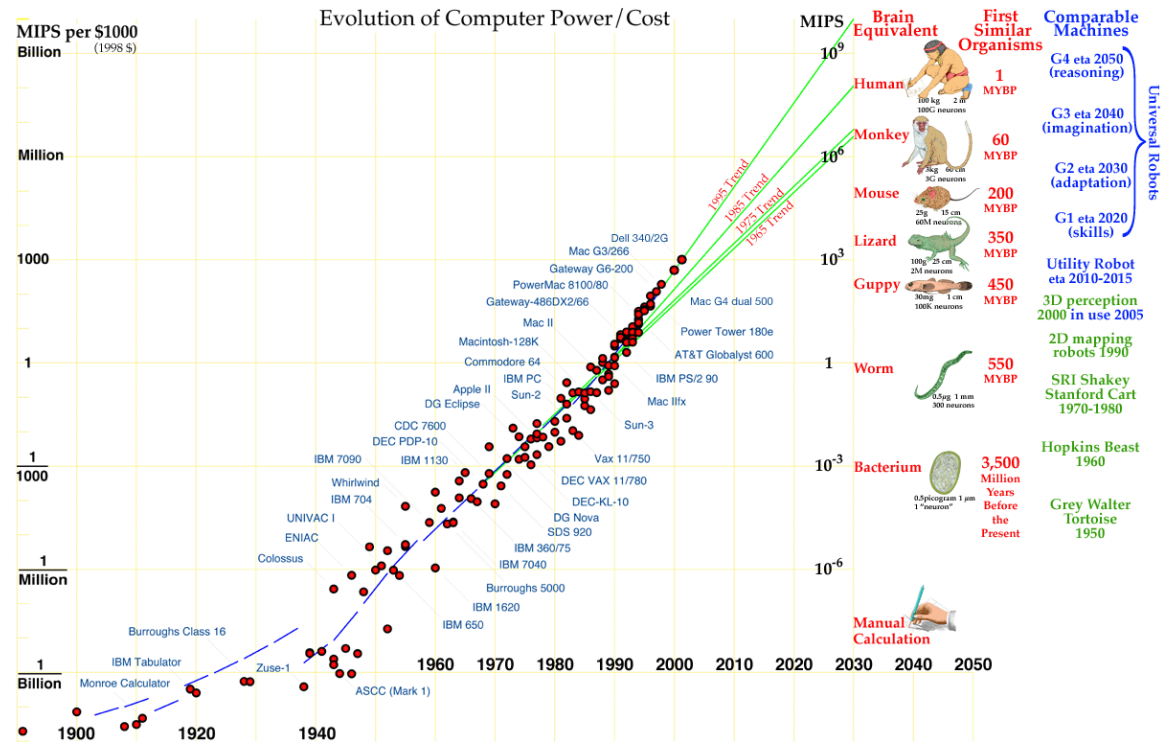
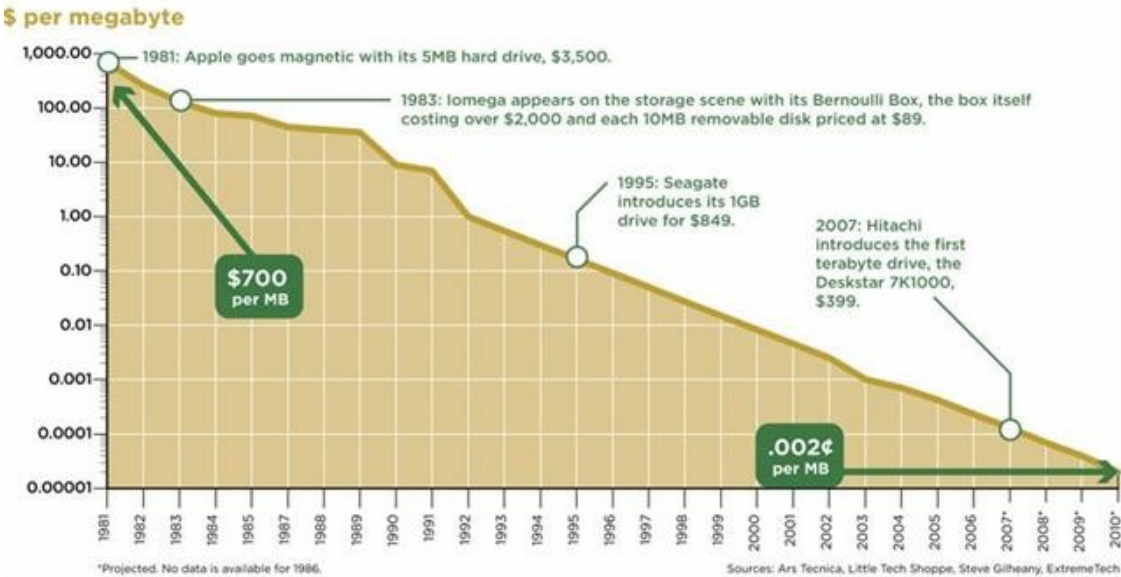
\$ per megabyte



# Evolutions in data



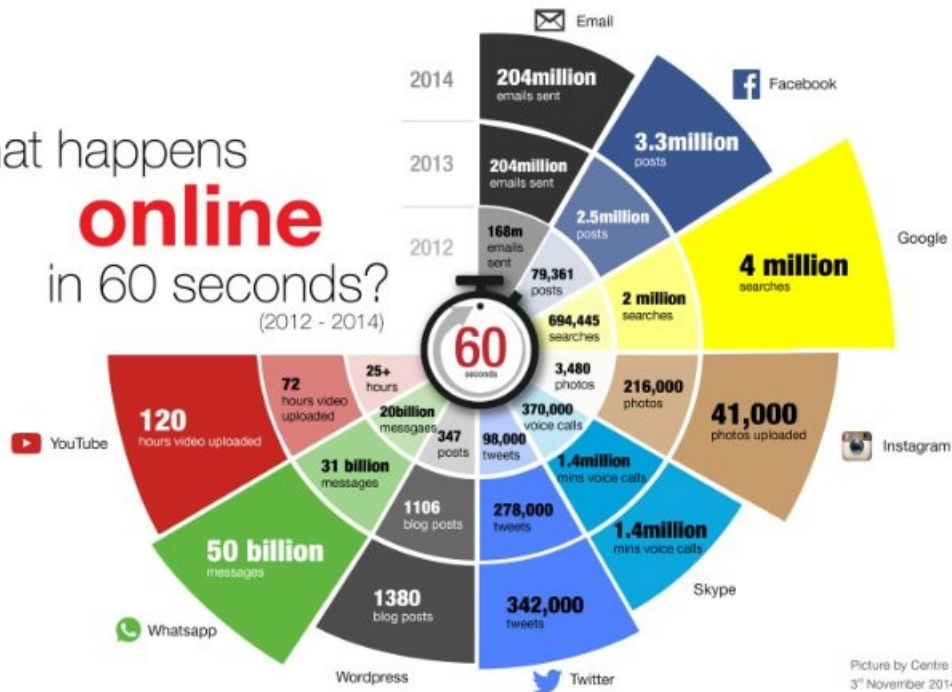
## STORAGE: FROM HIGHWAY ROBBERY TO RUNAWAY BARGAIN





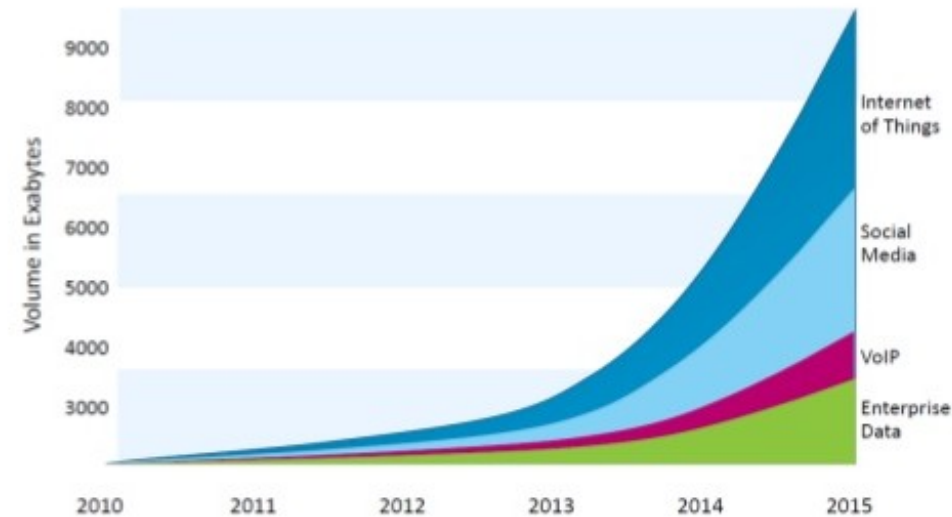
# Evolutions in data

What happens  
**online**  
in 60 seconds?  
(2012 - 2014)



Picture by Centre for  
3<sup>rd</sup> November 2014

## Amount of Data Generated is Exploding

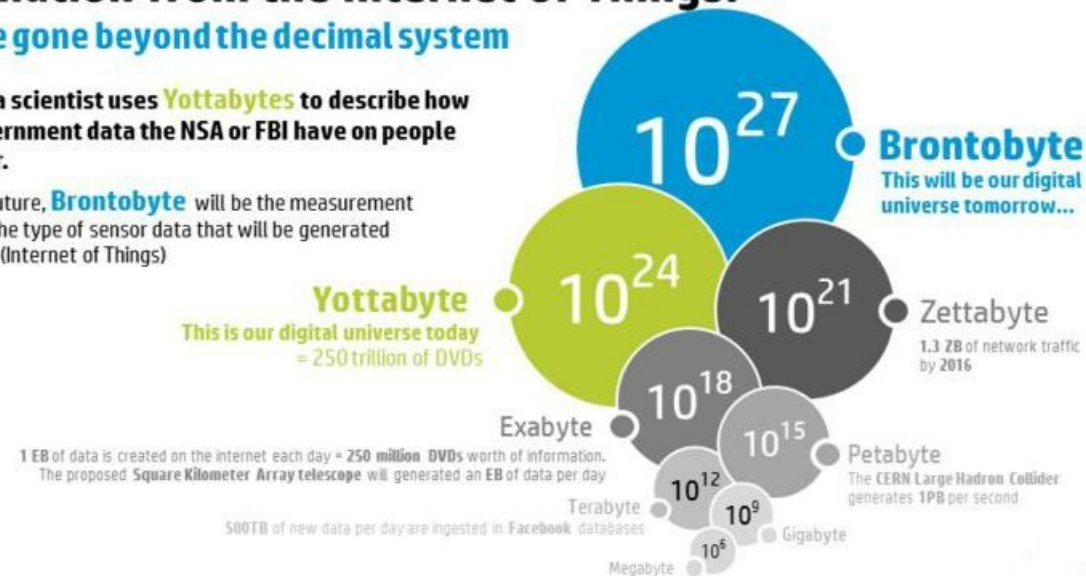


Source: IBM Global Technology Outlook

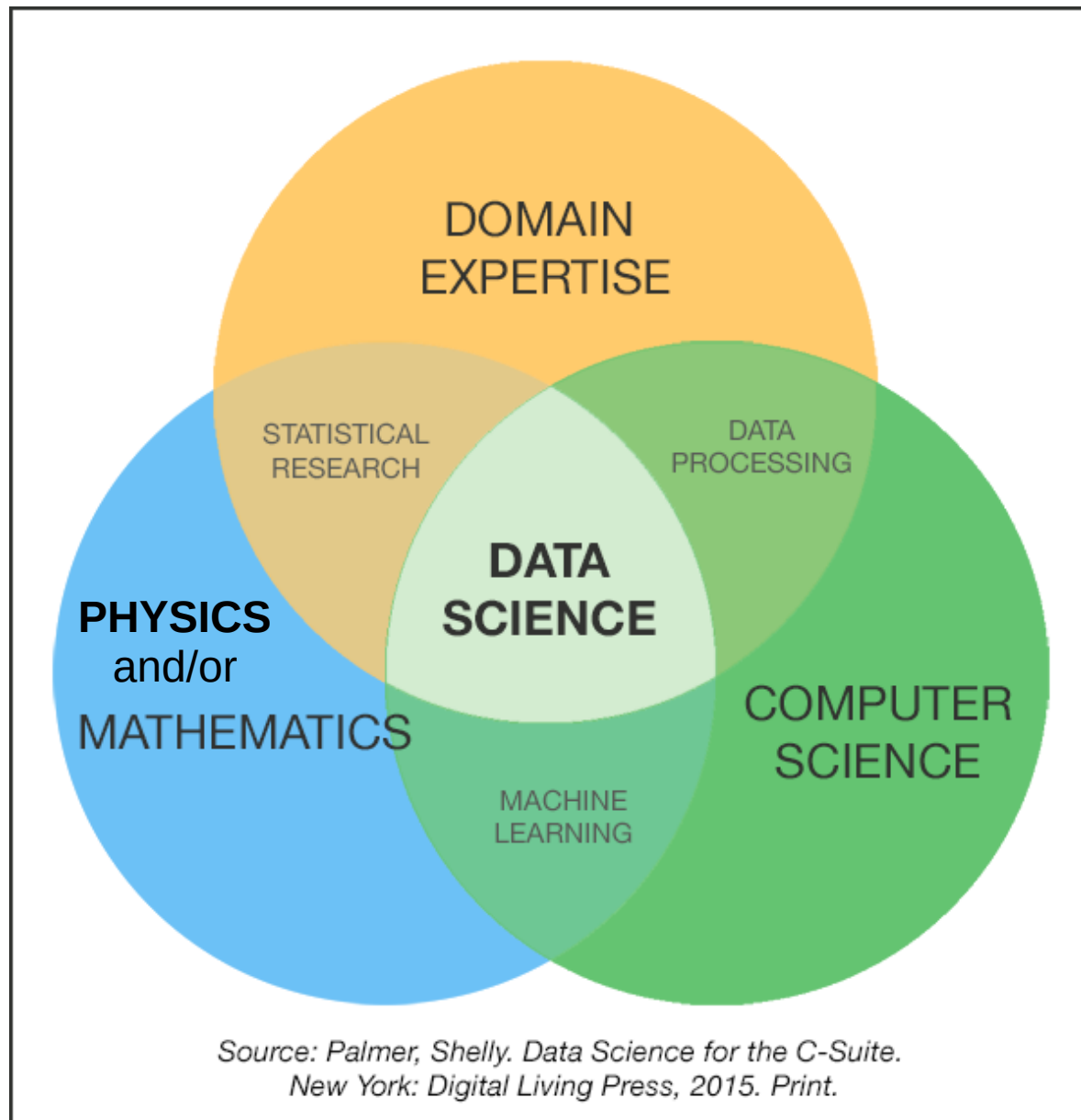
## Information from the Internet of Things: We have gone beyond the decimal system

Today data scientist uses **Yottabytes** to describe how much government data the NSA or FBI have on people altogether.

In the near future, **Brontobyte** will be the measurement to describe the type of sensor data that will be generated from the IoT (Internet of Things)



# Multidisciplinary research field



# Different types of data

- Linear data  
ex: Correlation matrix between proteins
- Sequential data (implicit time)  
ex: Geographical itineraries
- Temporal data  
ex: Wall Street market

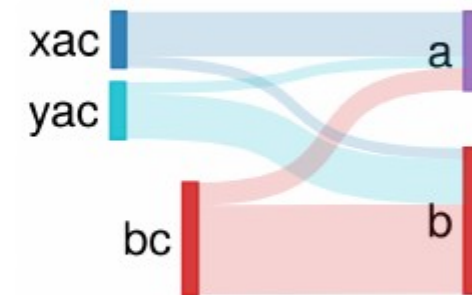
# Different types of data

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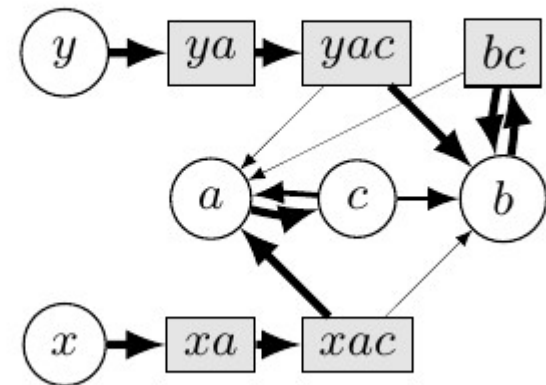
(a) Input

y	a	c	b	c
b	c	b	c	a
x	a	c	a	
y	a	c	b	
x	a	c	a	c
b	c	b		
...				

(b)  $P_x$  for  $x \in \{xac, yac, bc\}$



(c) VON network



Example of sequential data and its analysis through variable-order Markov chains



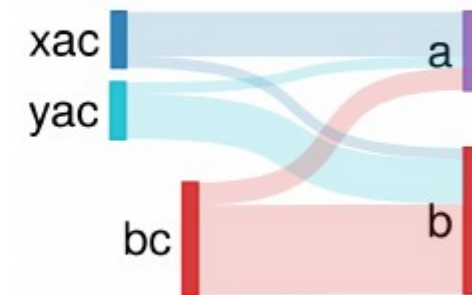
# Different types of data

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(a) Input

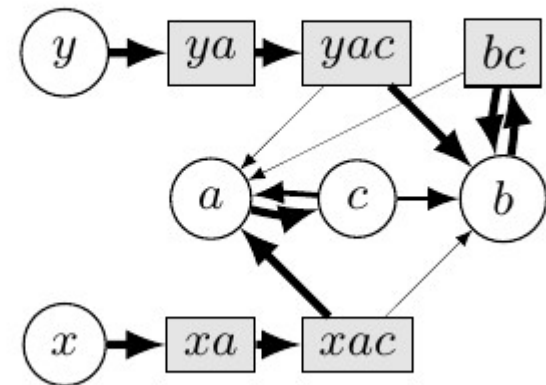
y	a	c	b	c
b	c	b	c	a
x	a	c	a	
y	a	c	b	
x	a	c	a	c
b	c	b		
...				

(b)  $P_x$  for  $x \in \{xac, yac, bc\}$



Network  
representation

(c) VON network

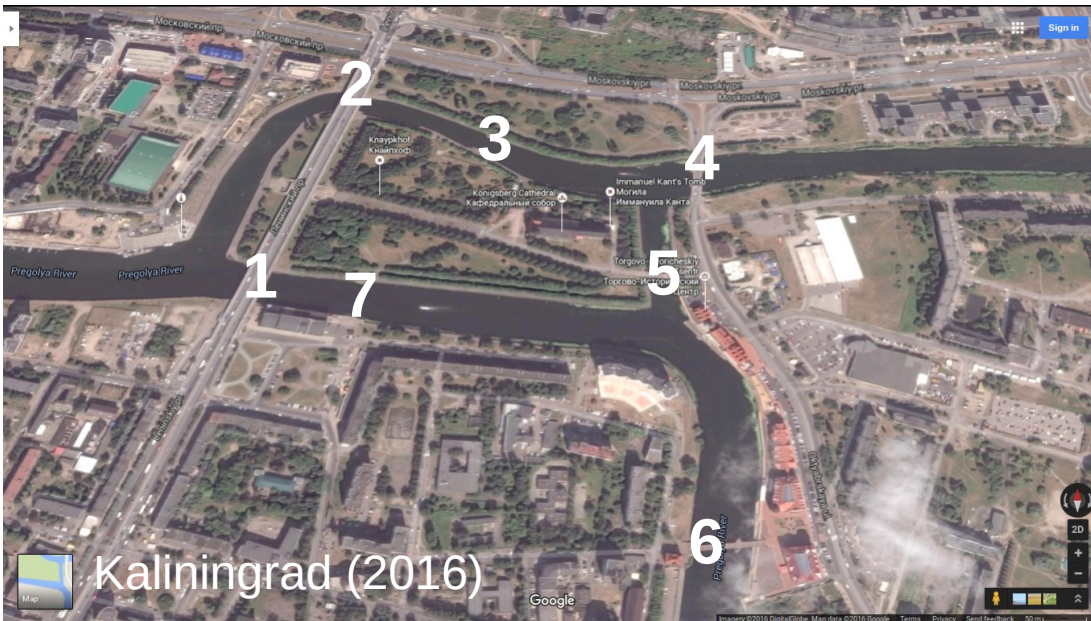


Example of sequential data and its analysis through variable-order Markov chains

# II Network Representations



# Graph theory



First paper on graph theory :

## The Seven Bridges of Königsberg Euler 1736

Is there a path connecting each territory passing by every bridge (but just once) ?



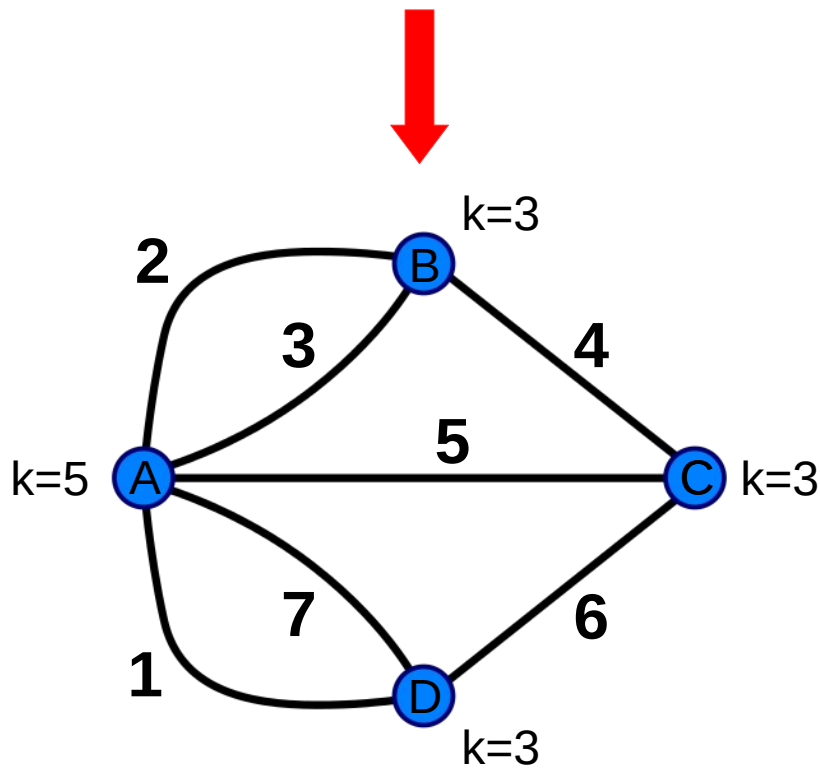
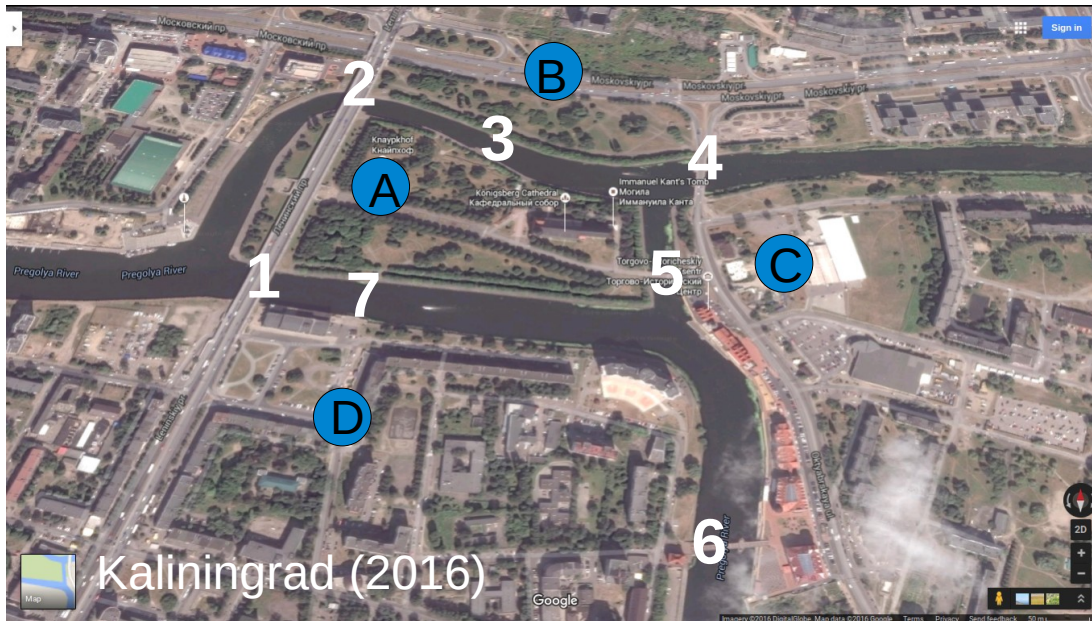
# Graph theory

First paper on graph theory :

## The Seven Bridges of Königsberg Euler 1736

Is there a path connecting each territory passing by every bridge (but just once) ?

Answer : **No**



$G=(V,E)$

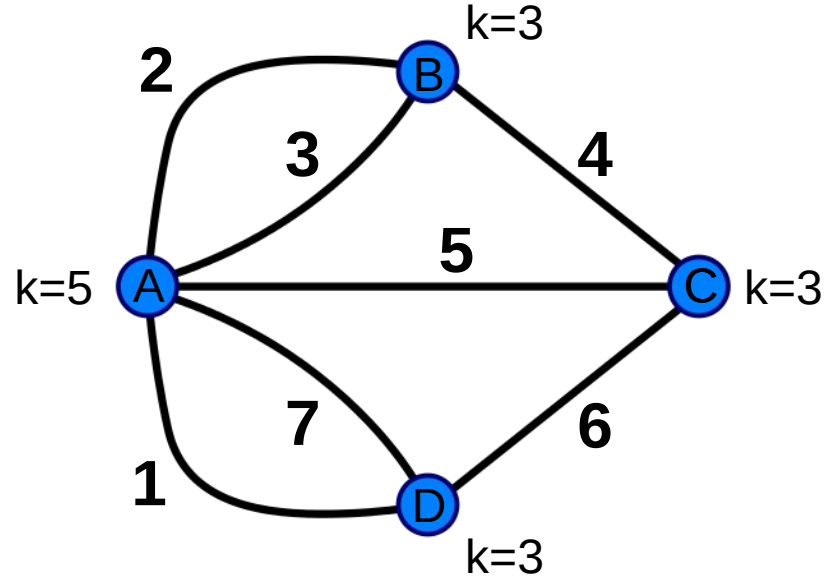
$V$  : set of vertices (nodes)

$E$  : set of edges (links)

$k$  : vertex degree



# Network representation



Path = sequence of links

$$L_1(A,C) = [(A,D), (D,C)]$$

$$L_2(A,C) = [(A,C)]$$

$$L_3(A,C) = [(A,B), (B,C)]$$

$$L_4(A,C) = [(A,B), (B,A), (A,C)]$$

...

Length 2

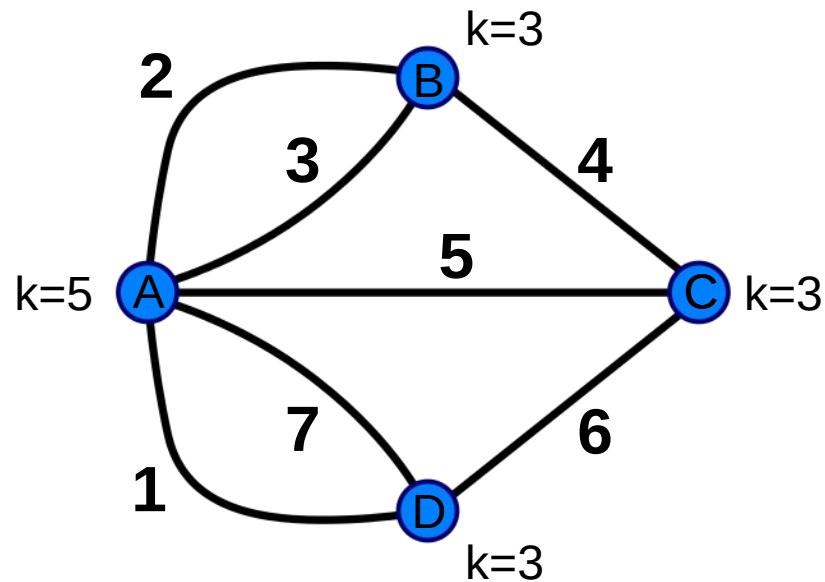
Length 1 = *shortest path*

Length 2

Length 3

...

# Network representation



No  
Eulerian path

**Eulerian path** = Sequence of links without duplicates representing a closed loop containing all nodes

# A trivial model: Random networks

(Erdős-Rényi random graph model from '59)

Random Network



Model 1 :

$$\mathbf{G}=\mathbf{G}(n,M)$$

Graphs with  $n$  vertices and  $M$  links

Model 2 :

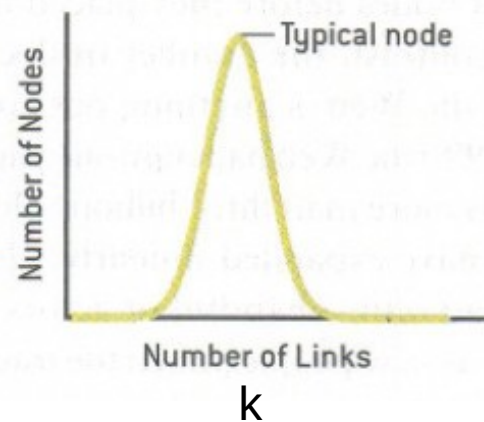
$$\mathbf{G}=\mathbf{G}(n,p)$$

Graphs with  $n$  vertices and linking probability  $p$ .

Number of expected links =  $pn(n-1)/2$

$$n \rightarrow \infty$$

Bell Curve Distribution of Node Linkages



**Properties studied :**

Connectedness

Graph diameter

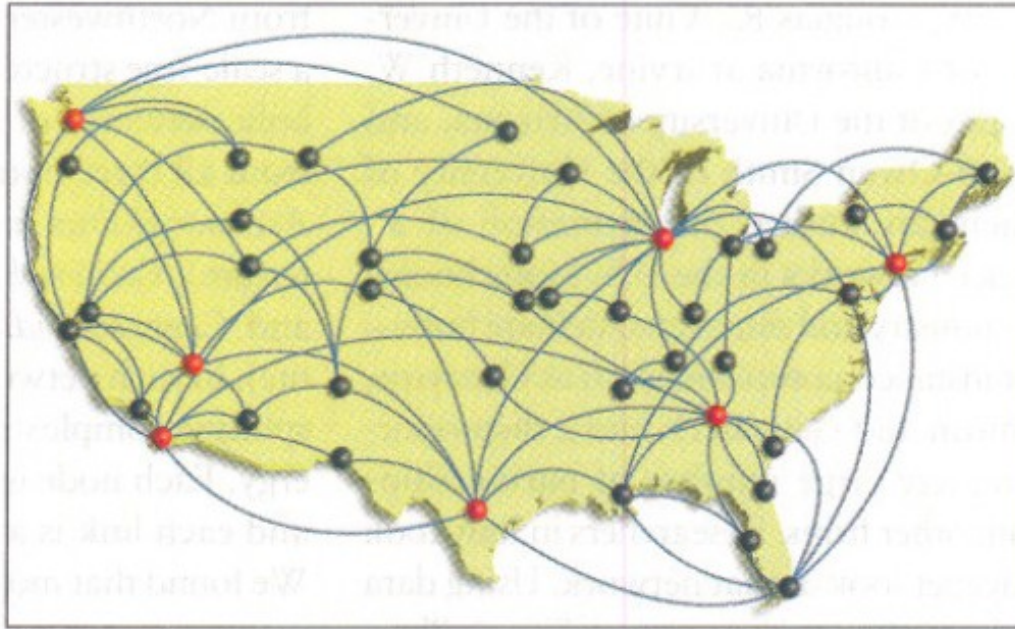
Subgraphs

...

# More complex model: Scale-free network

(Barabasi-Albert model from '99)

Scale-Free Network



Preferential attachment :

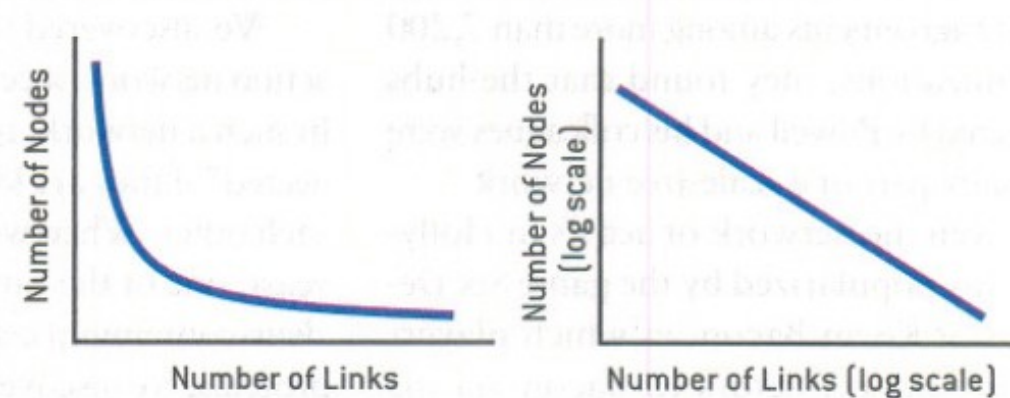
$$\mathbf{G} = \mathbf{G}(m_0, \vec{p})$$

Initial graphs with  $m_0$  vertices and preferential attachment vector  $\vec{p}$

$$p_i = \frac{k_i}{\sum_{j=1}^{m_0} k_j}$$

Where indexes  $i$  and  $j$  denote vertex “ $i$ ” and “ $j$ ” and  $k$  is a number of link.

Power Law Distribution of Node Linkages



**Properties studied :**

Connectedness

Graph diameter

Subgraphs

...



# Random graphs

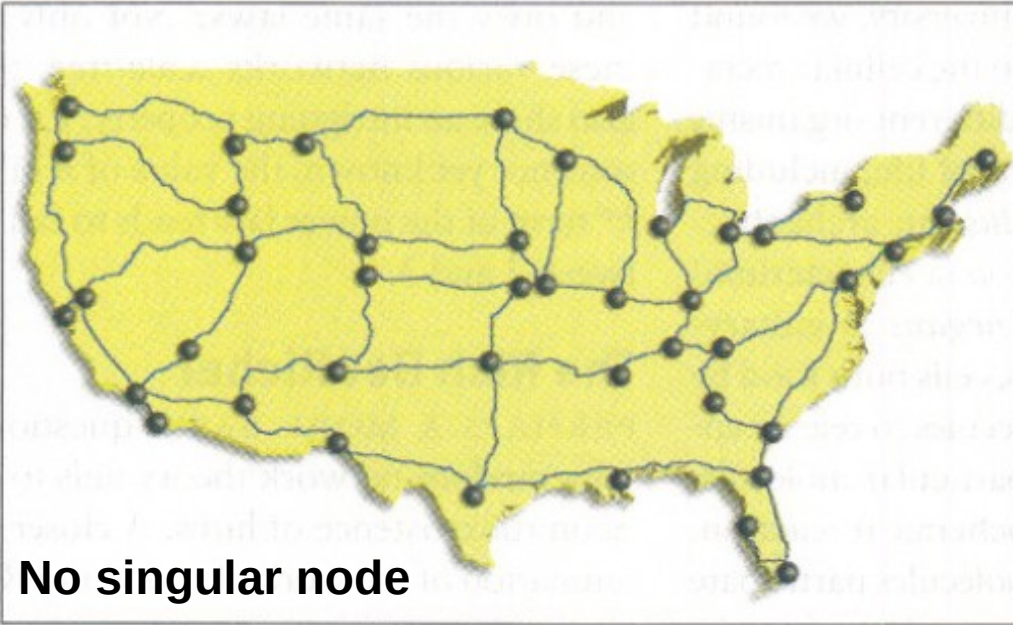
(Erdős-Rényi random graph model from '59)

VS

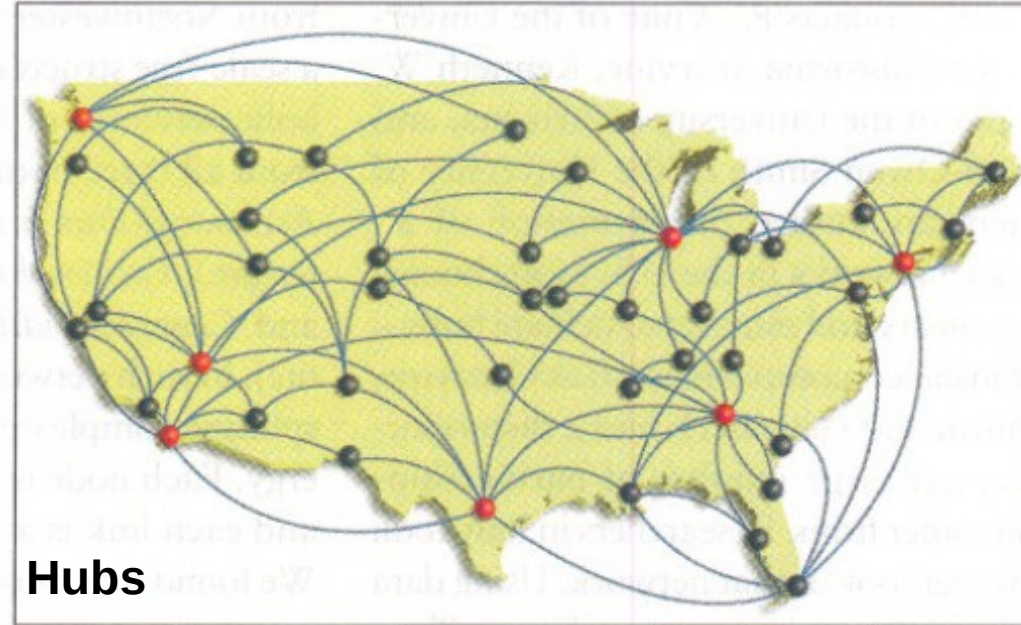
# (Real) complex networks

(A.-L. Barabási, R. Albert '99)

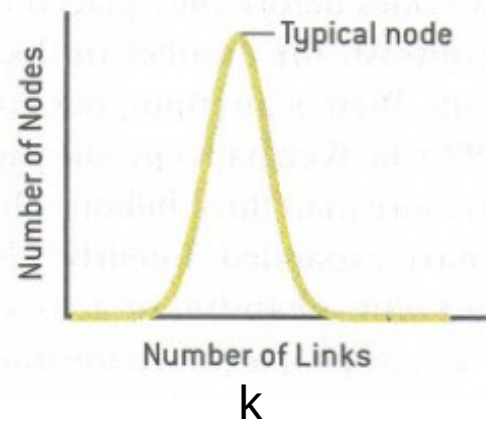
Random Network



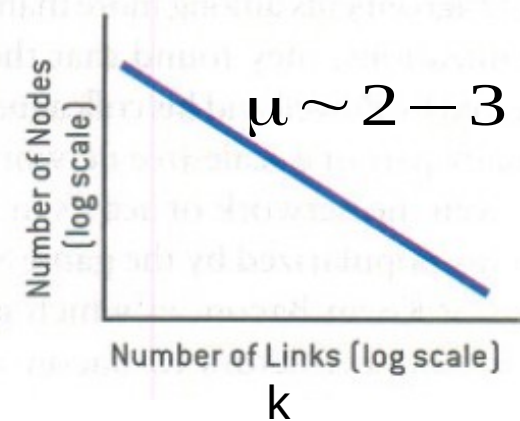
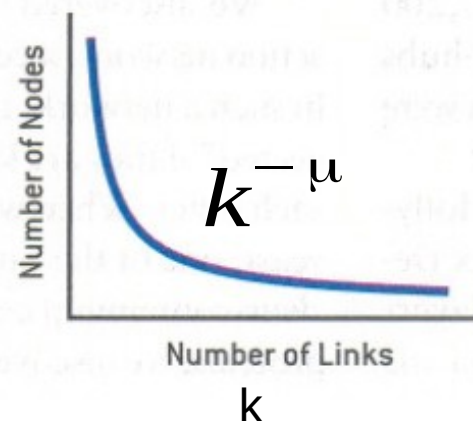
Scale-Free Network



Bell Curve Distribution of Node Linkages



Power Law Distribution of Node Linkages

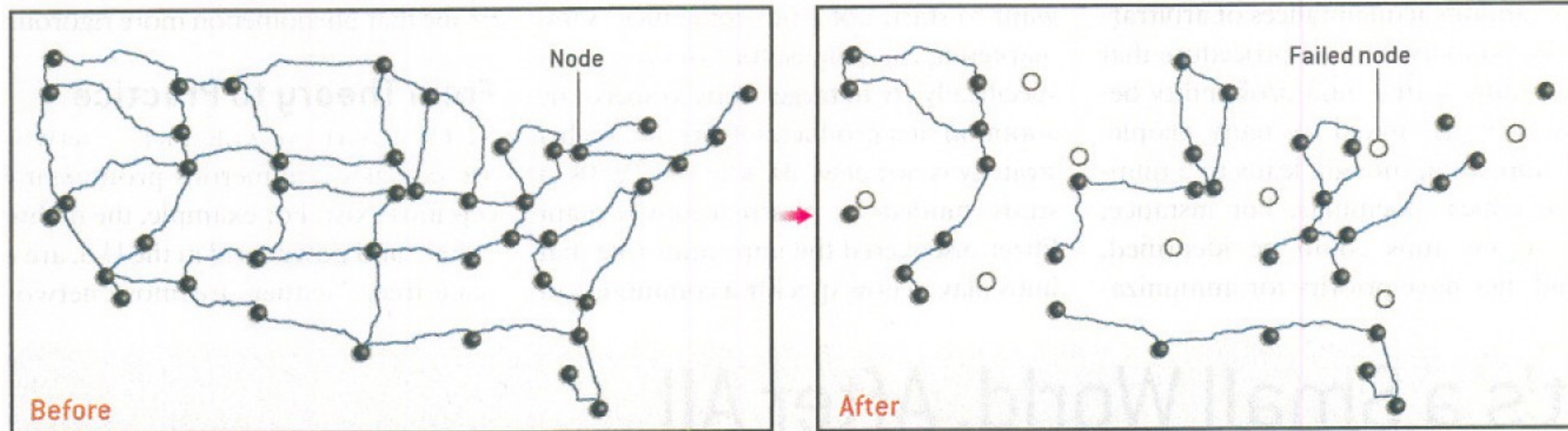


# Random graphs

# VS

# (Real) complex networks

Random Network, Accidental Node Failure



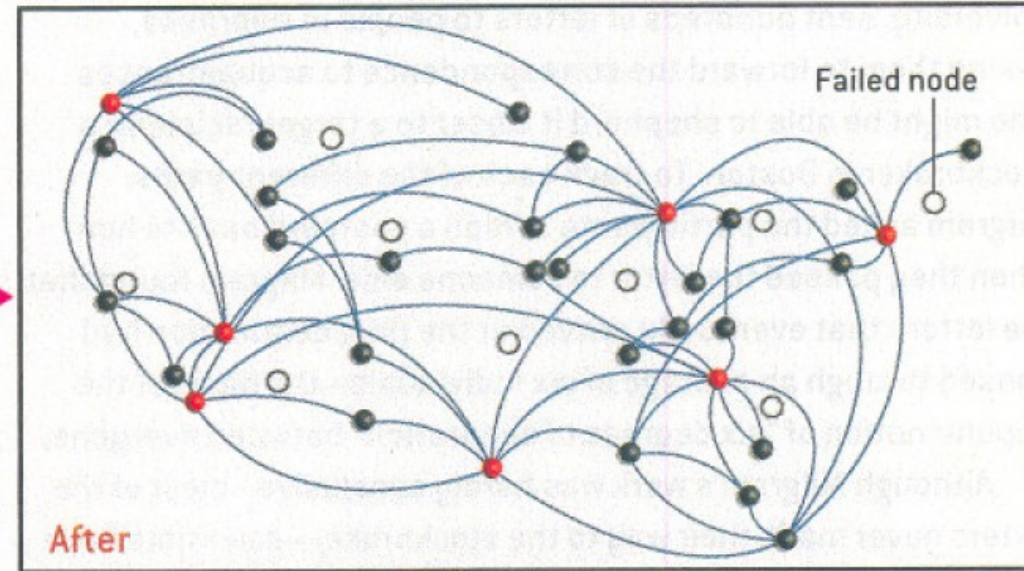
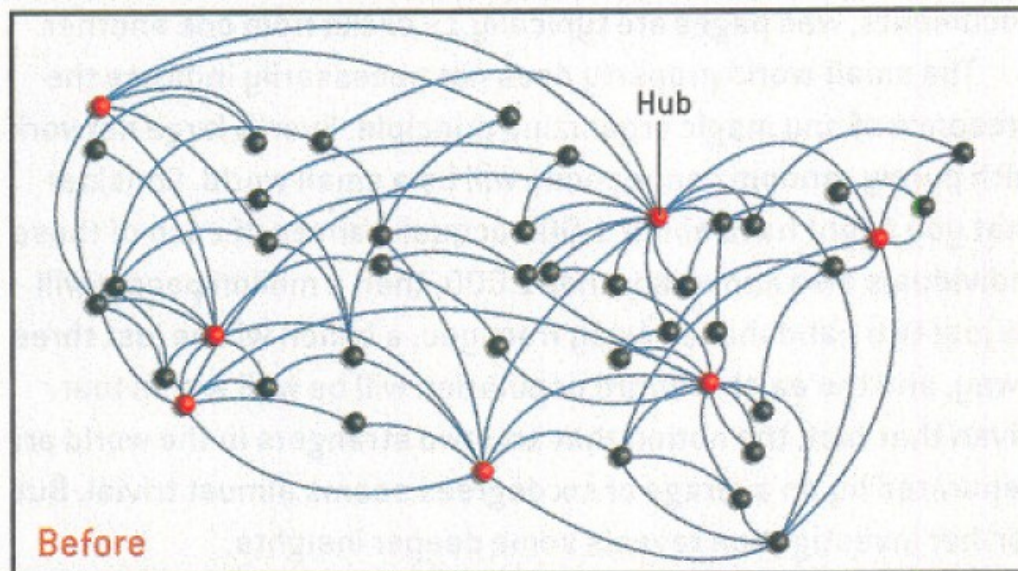


# Random graphs

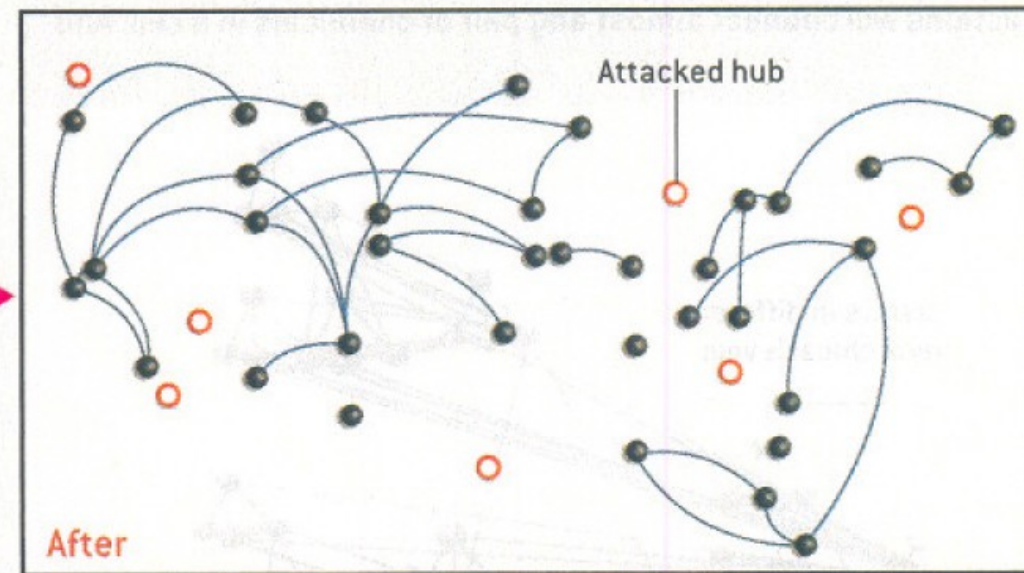
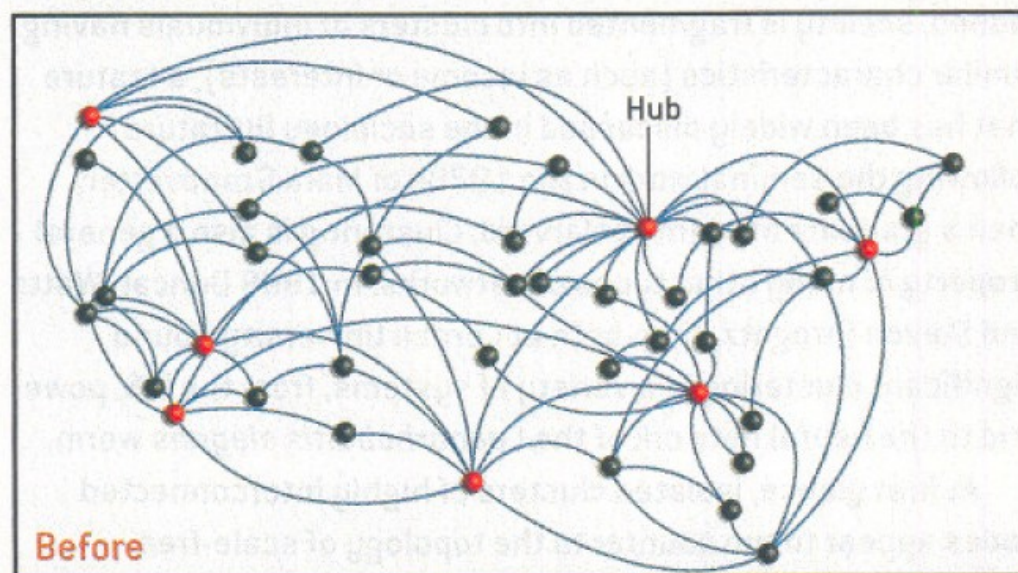
VS

# (Real) complex networks

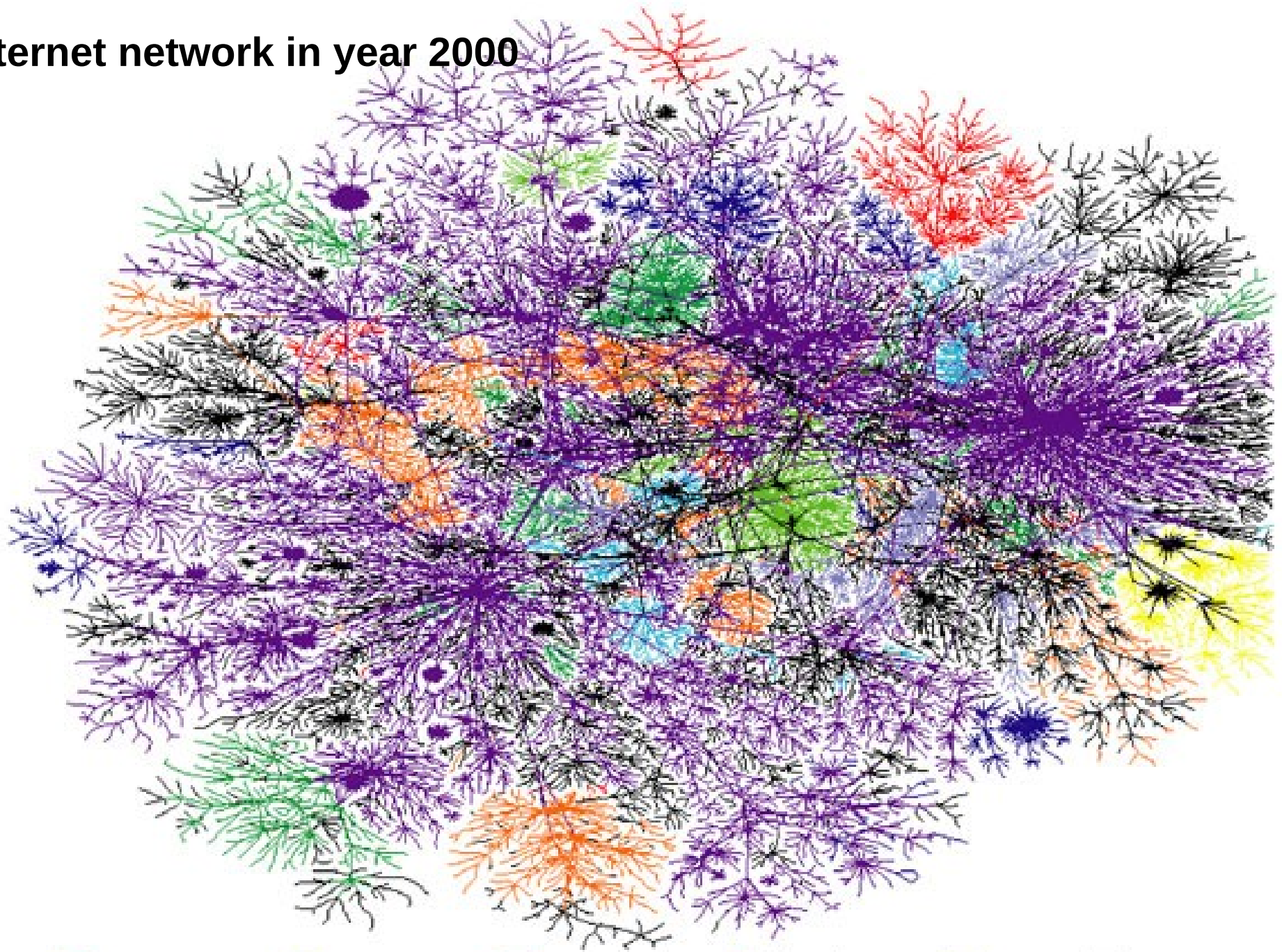
Scale-Free Network, Accidental Node Failure



Scale-Free Network, Attack on Hubs



## Internet network in year 2000





# Real scale-free networks

## *Examples of Scale-Free Networks*

NETWORK	NODES	LINKS
Cellular metabolism	Molecules involved in burning food for energy	Participation in the same biochemical reaction
Hollywood	Actors	Appearance in the same movie
Internet	Routers	Optical and other physical connections
Protein regulatory network	Proteins that help to regulate a cell's activities	Interactions among proteins
Research collaborations	Scientists	Co-authorship of papers
Sexual relationships	People	Sexual contact
World Wide Web	Web pages	URLs

# Real scale-free networks

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## Construction property

"The rich get richer"

or

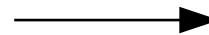
Matthew effect (Mt 25:29 NT)

or

Preferential attachment

## Type of links:

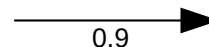
Directed



Undirected



Weighted(D/U)





# Scale-free networks

## Examples of Scale-Free Networks

NETWORK	NODES	LINKS
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## Construction property

"The rich get richer"  
or  
Matthew effect (Mt 25:29 NT)  
or  
Preferential attachment

## What about directionality ?

In fact any network can be considered as directed.

Now there is a difference  
between  
**node outdegree**  
And  
**node indegree**

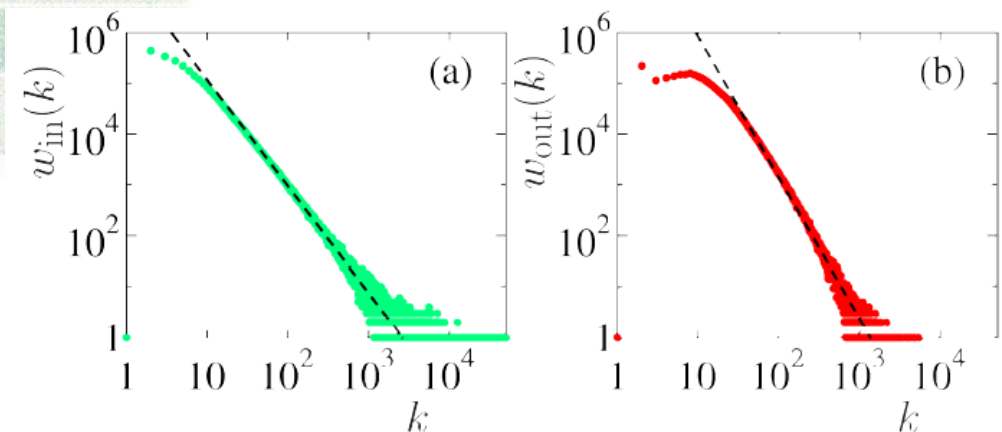
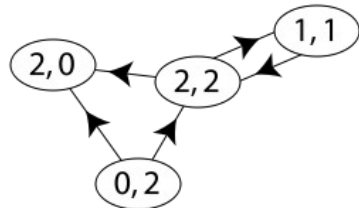
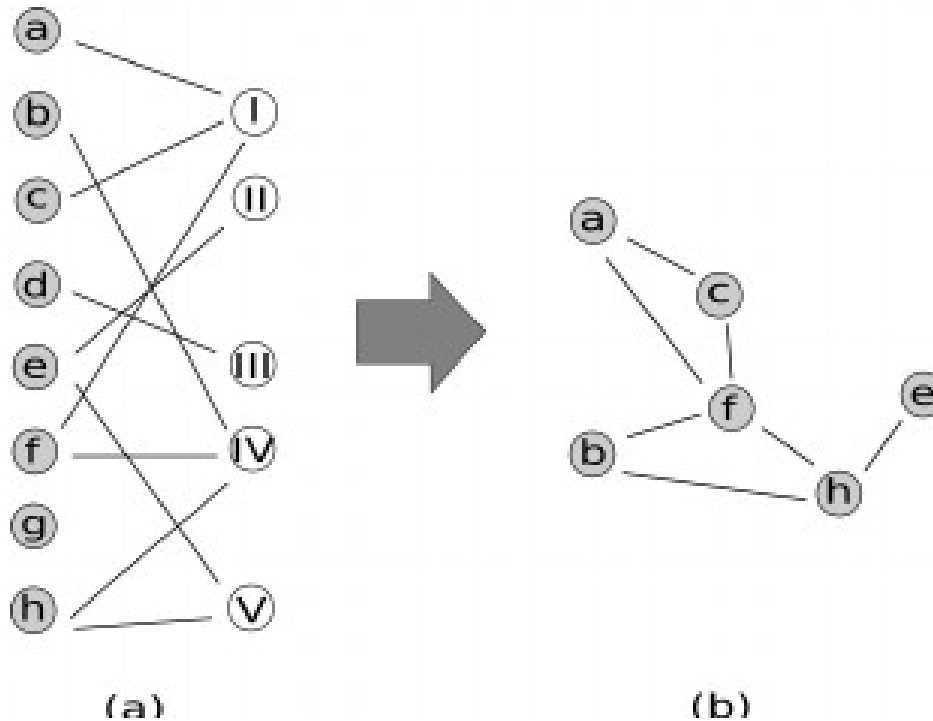


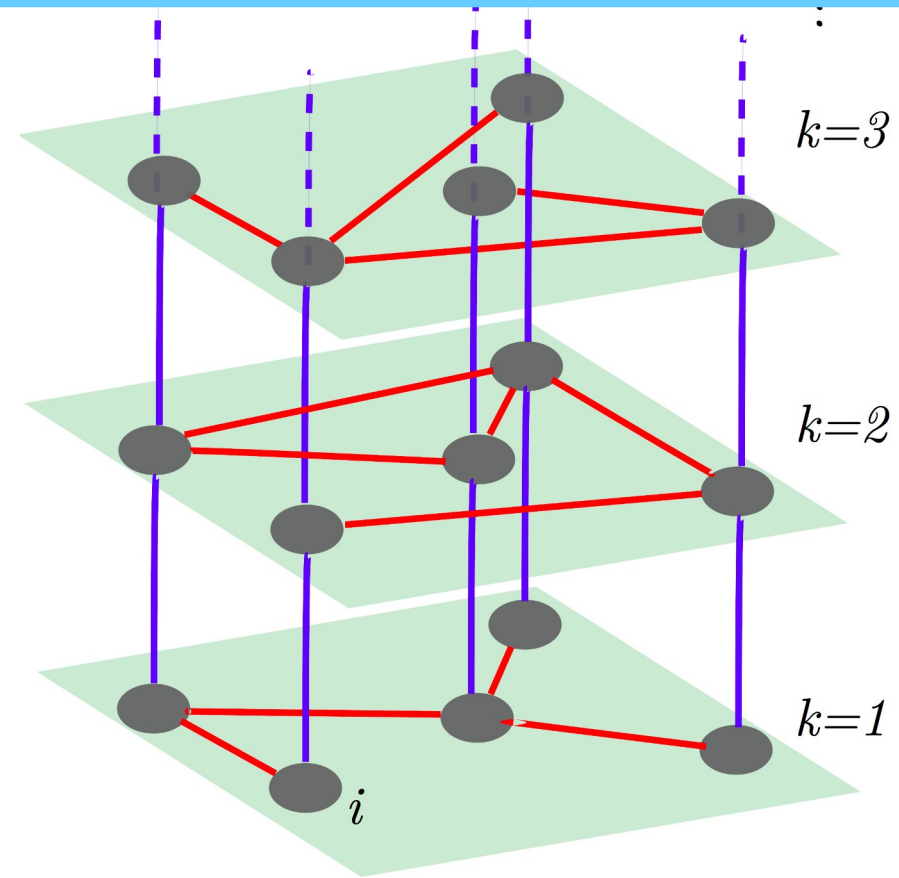
FIG. 2 (Color online) Distribution  $w_{in,out}(k)$  of number of incoming (a) and outgoing (b) links  $k$  for  $N = 3282257$  Wikipedia English articles (Aug 2009) of Fig. 1 with total number of links  $N_\ell = 71012307$ . The straight dashed fit line shows the slope with  $\mu_{in} = 2.09 \pm 0.04$  (a) and  $\mu_{out} = 2.76 \pm 0.06$  (b). After (Zhirov *et al.*, 2010).

# Large number of network models

Bipartite networks (ex: election network)  
Multi-layer networks (ex: Wikipedia network)  
Higher-order networks (Sequential data repr.)  
Temporal networks (Time series)  
Trees



Bipartite network and its projection



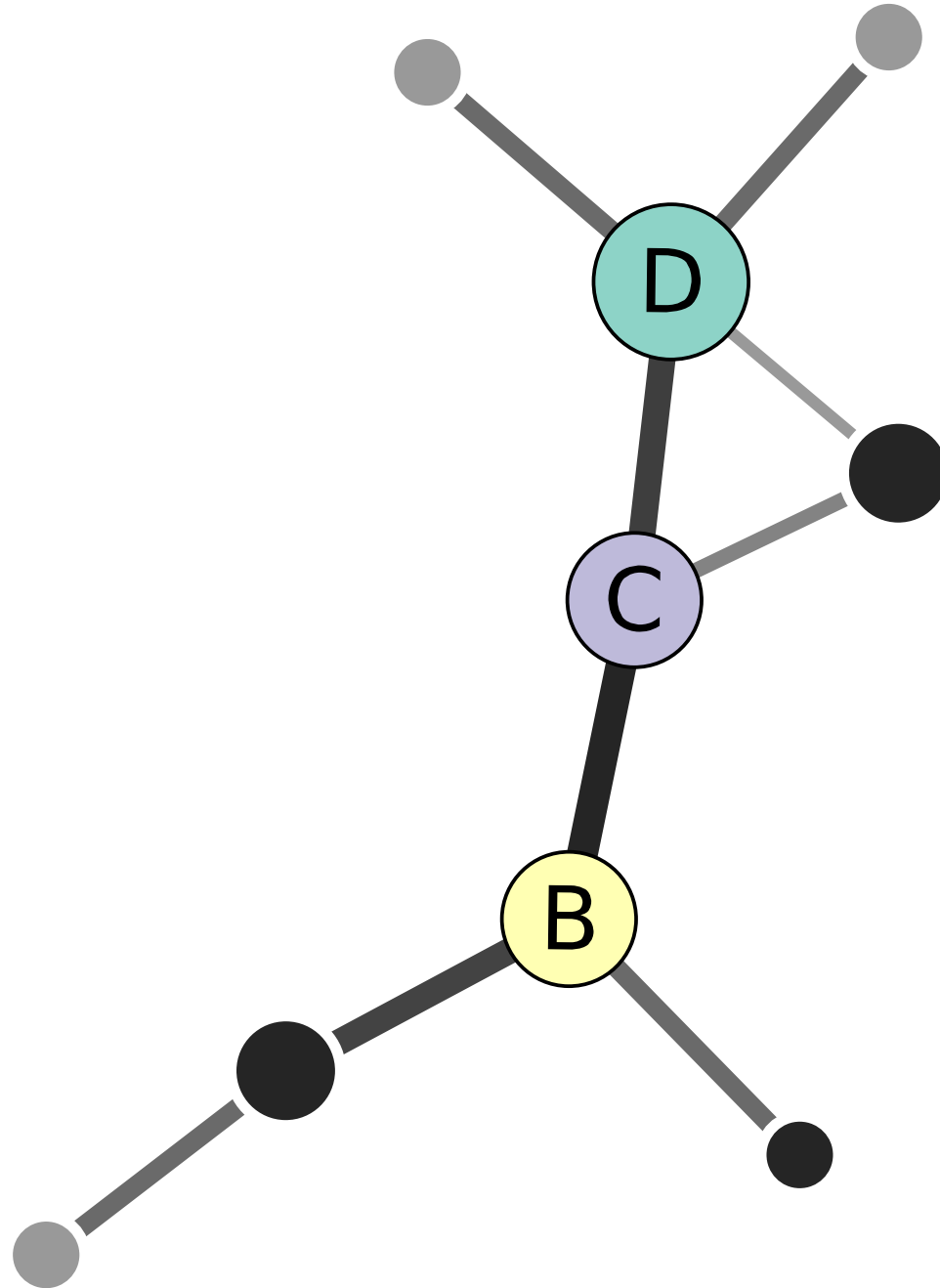
Multi-layer network

# III Centrality Measures



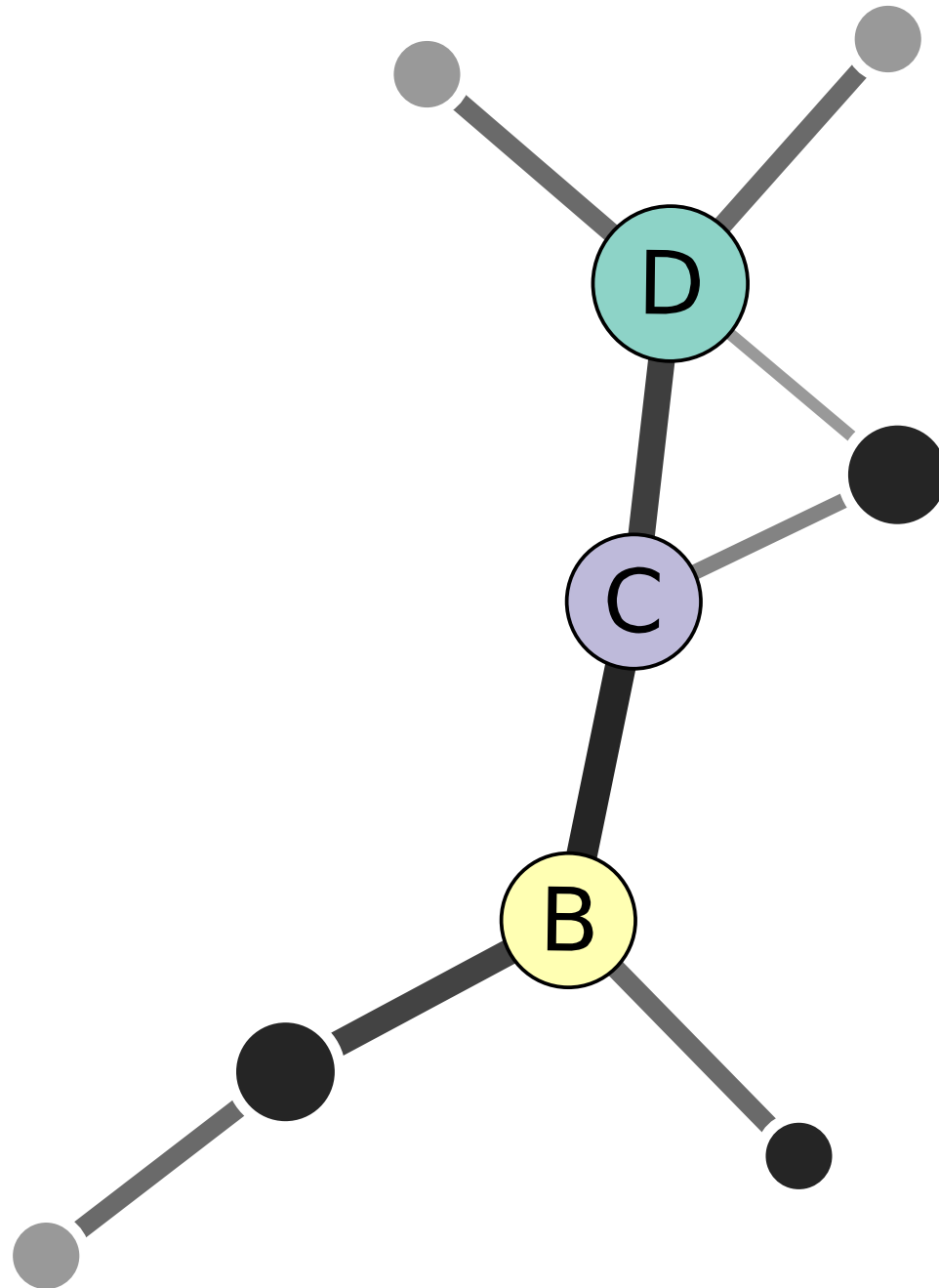


# What is a Centrality measure?



# What is a Centrality measure?

Number of Link?  
Efficiency of path?  
Number of path?

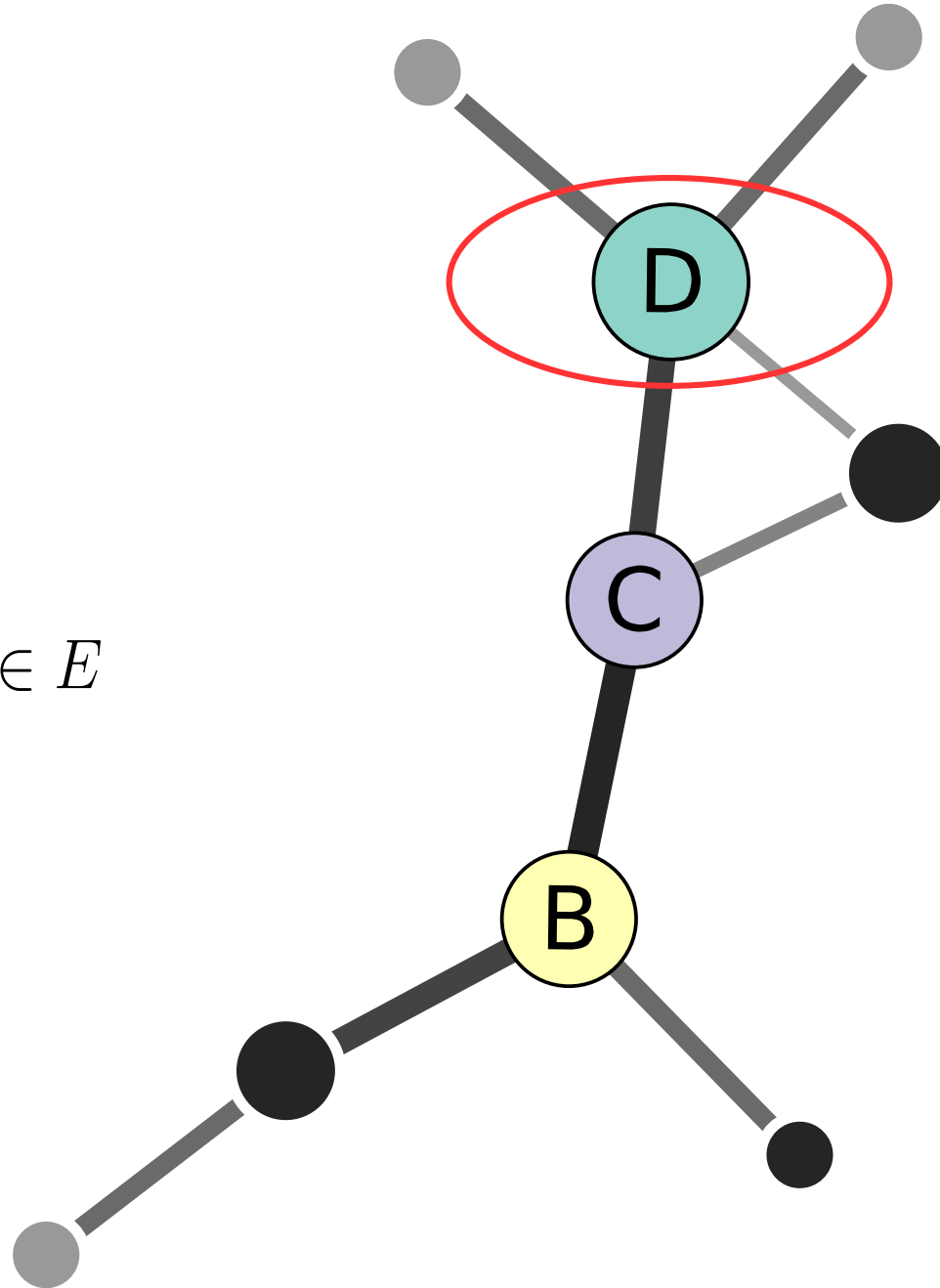


# What is a Centrality measure?

Degree (K)

$$K(i) = \sum_{j \in V} x_{ij}$$

$$x_{ij} = \begin{cases} 1 & \text{if } x_{ij} \in E \\ 0 & \text{else} \end{cases}$$

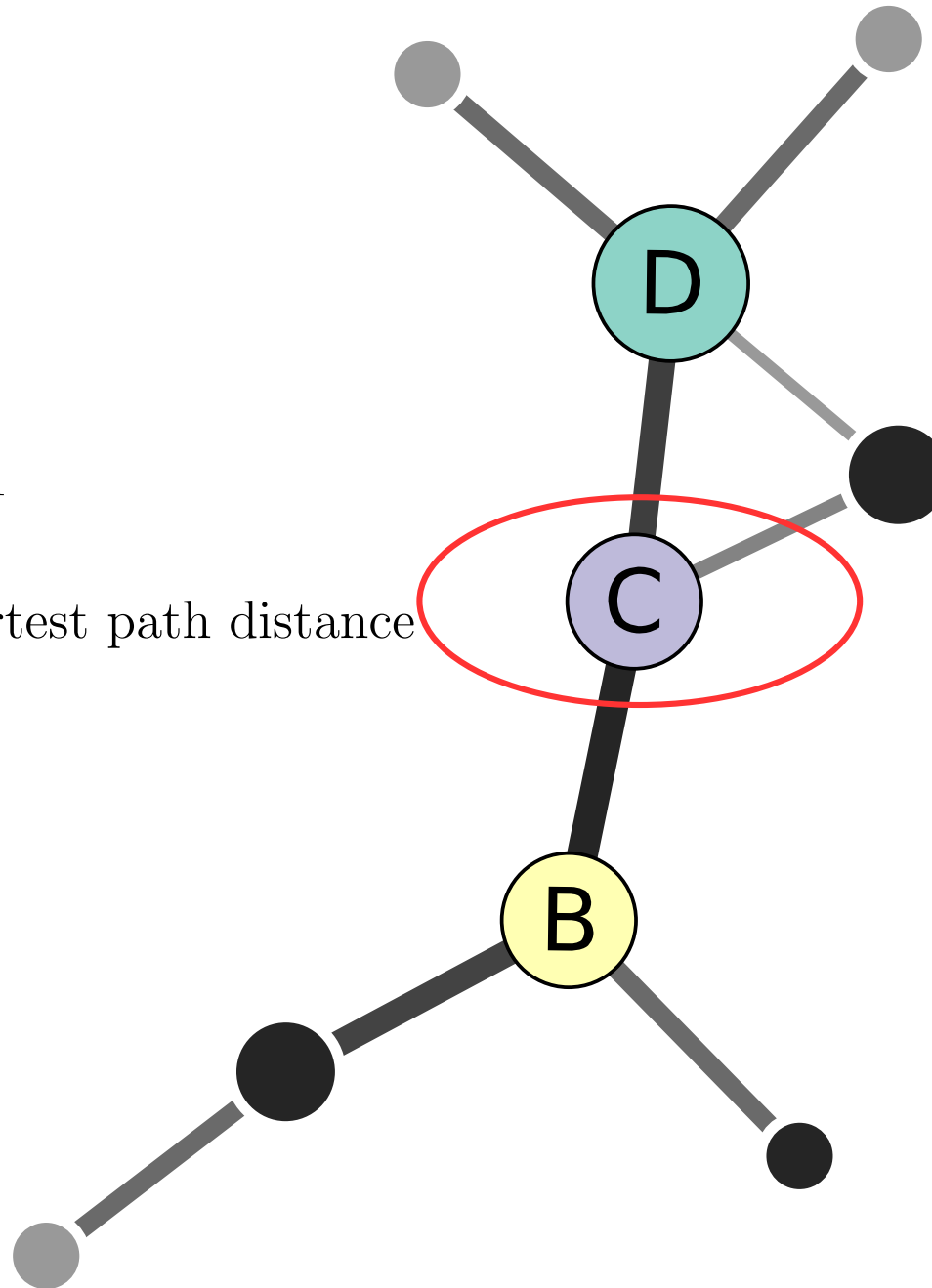


# What is a Centrality measure?

## Closeness

$$C(i) = \left( \sum_{j \in V} \text{dist}(i, j) \right)^{-1}$$

where  $\text{dist}(i, j)$  is the shortest path distance

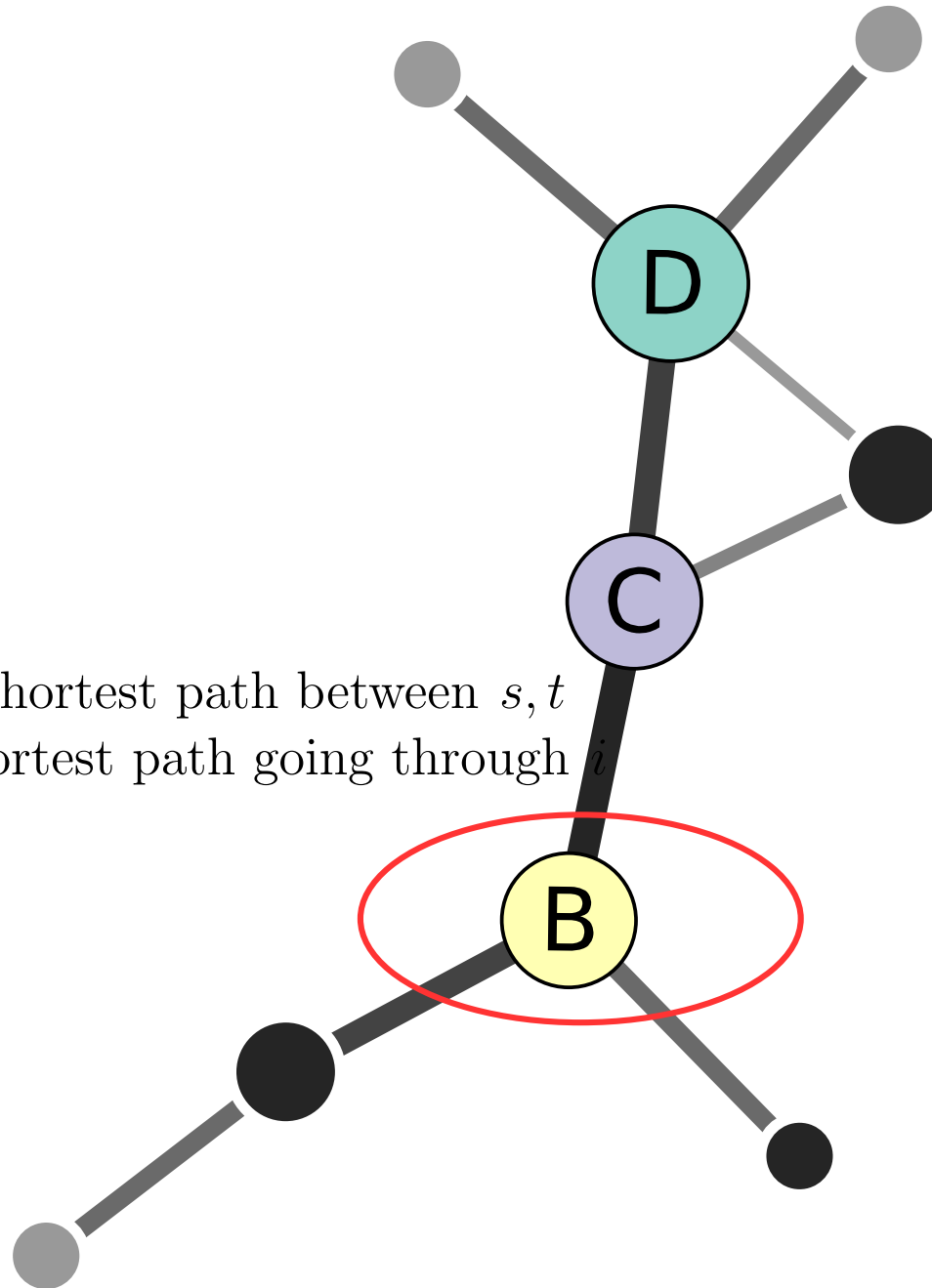


# What is a Centrality measure?

## Betweenness

$$B(i) = \sum_{s,t \in V} \frac{\sigma_{st}(i)}{\sigma_{st}}$$

where  $\sigma_{st}$  is the number of shortest path between  $s, t$  and  $\sigma_{st}(i)$  the number of shortest path going through  $i$



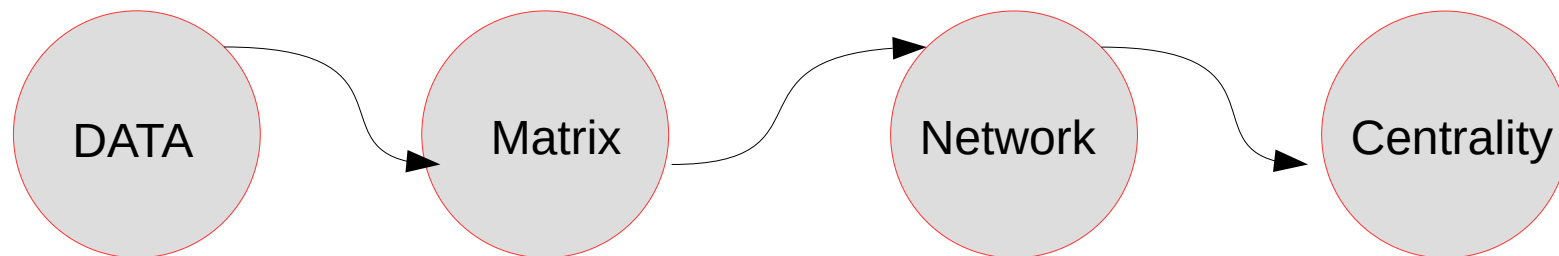


# What is a Centrality measure?

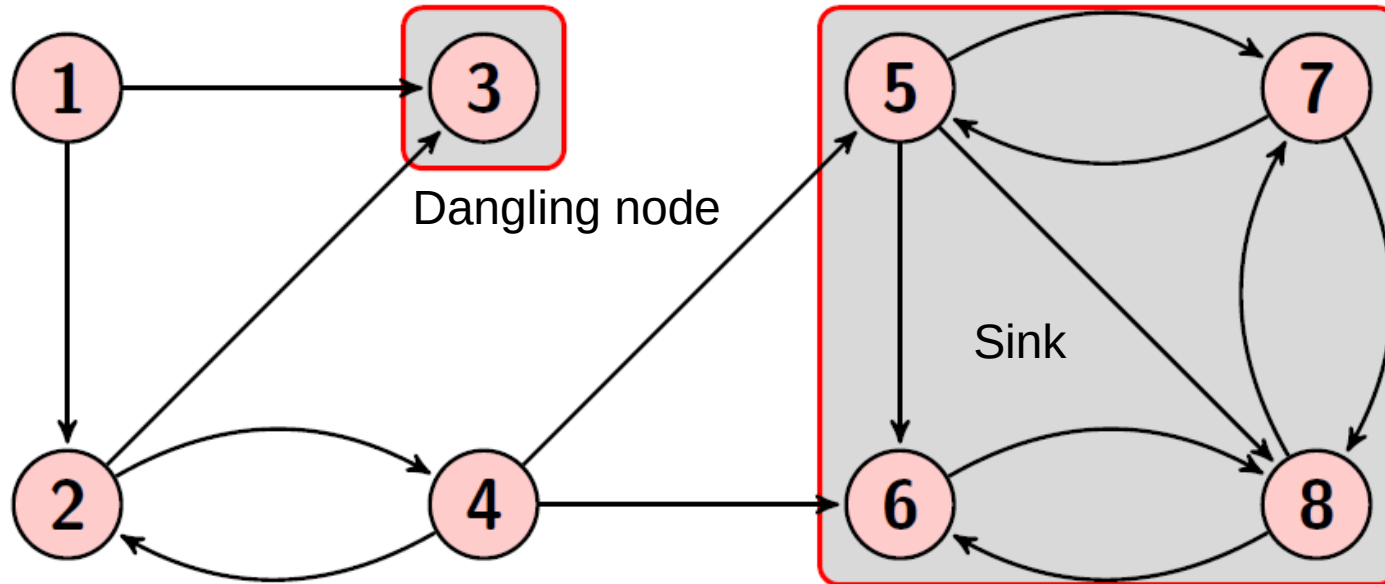
The method you use depends on what you are investigating

For directed network such that the World Wide Web, the eigenvector centrality is used

Ex: Google's research engine algorithm



# Modeling Random Walk on a network

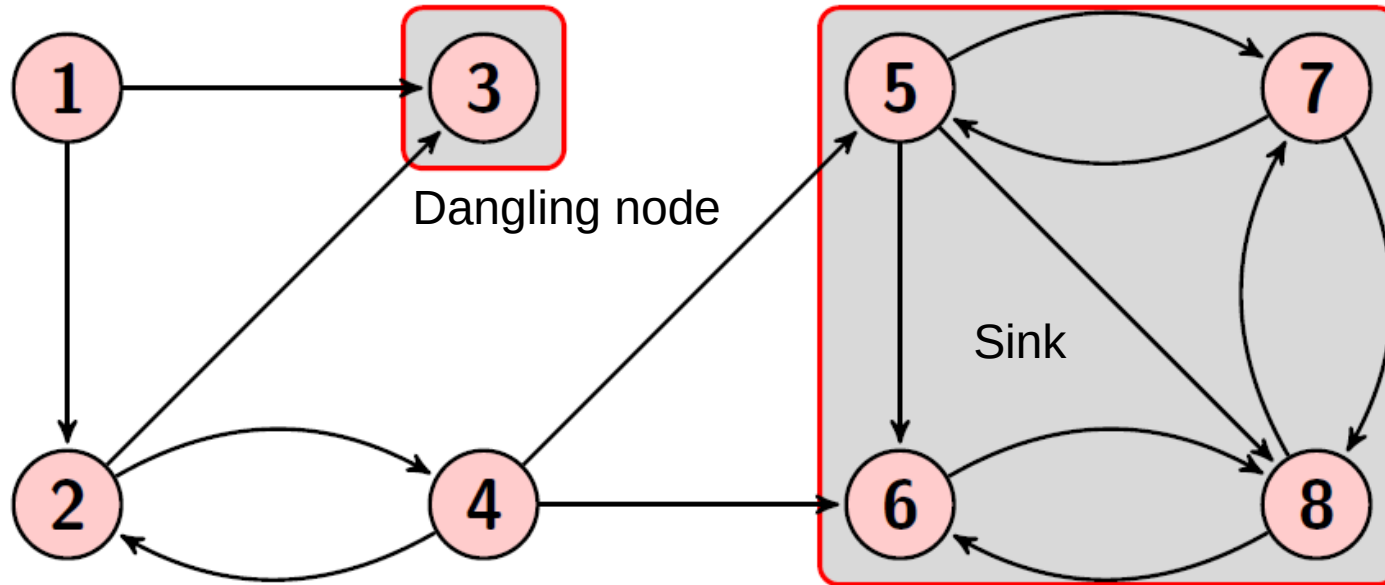


Adjacency matrix :

$$A_{ij} = \begin{cases} 1 & \text{if } j \rightarrow i \\ 0 & \text{otherwise} \end{cases}$$

$$A = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 & 1 & 0 & 0 \end{pmatrix}$$

# The Stochastic Matrix

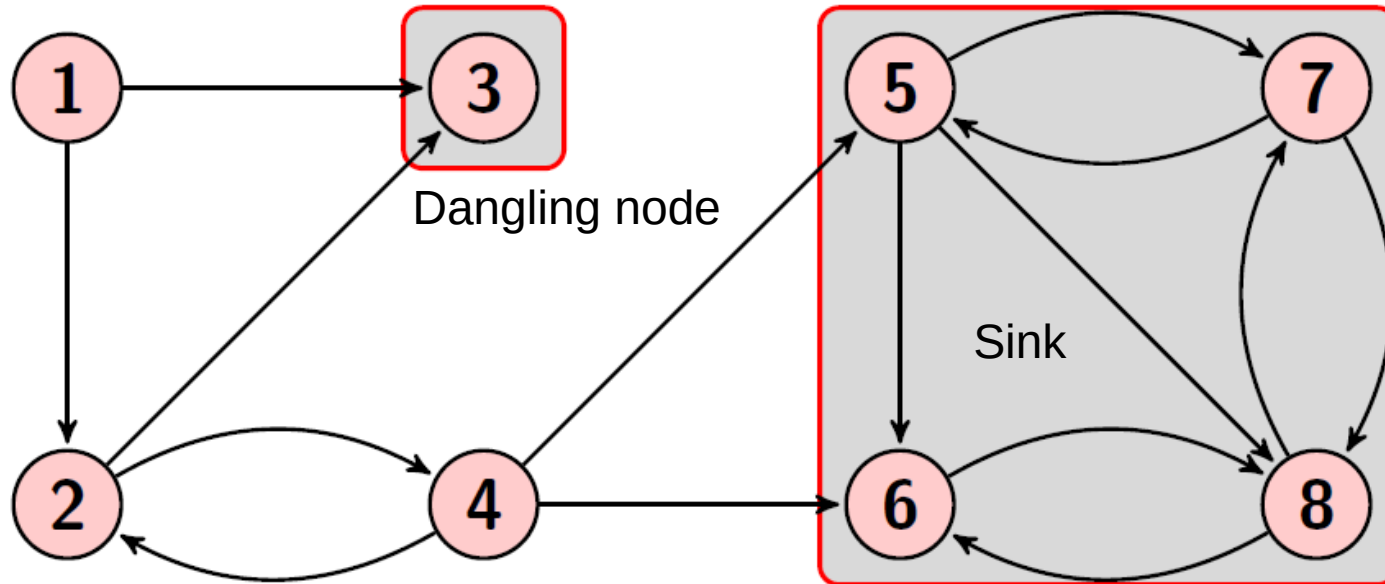


Stochastic matrix :

$$S_{ij} = \begin{cases} \frac{1}{N} & \text{if } j \text{ is a dangling node} \\ \frac{A_{ij}}{\sum_{i=1}^N A_{ij}} & \text{otherwise} \end{cases}$$

$$S = \begin{pmatrix} 0 & 0 & 1/8 & 0 & 0 & 0 & 0 & 0 \\ 1/2 & 0 & 1/8 & 1/3 & 0 & 0 & 0 & 0 \\ 1/2 & 1/2 & 1/8 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1/2 & 1/8 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/8 & 1/3 & 0 & 0 & 1/2 & 0 \\ 0 & 0 & 1/8 & 1/3 & 1/3 & 0 & 0 & 1/2 \\ 0 & 0 & 1/8 & 0 & 1/3 & 0 & 0 & 1/2 \\ 0 & 0 & 1/8 & 0 & 1/3 & 1 & 1/2 & 0 \end{pmatrix}$$

# The Google Matrix



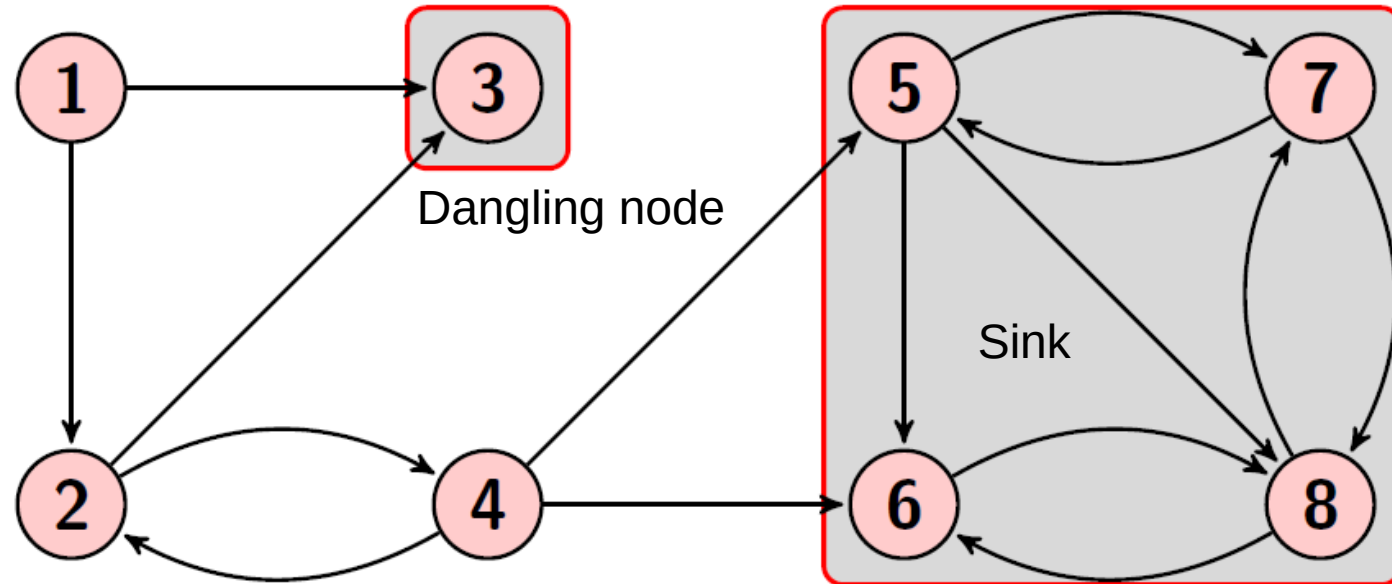
Google matrix :

$$G_{ij} = \alpha S_{ij} + (1 - \alpha)/N$$

$\alpha \in [0, 1]$  is the damping factor

$$G = \begin{pmatrix} 1/40 & 1/40 & 1/8 & 1/40 & 1/40 & 1/40 & 1/40 & 1/40 \\ 17/40 & 1/40 & 1/8 & 7/24 & 1/40 & 1/40 & 1/40 & 1/40 \\ 17/40 & 17/40 & 1/8 & 1/40 & 1/40 & 1/40 & 1/40 & 1/40 \\ 1/40 & 17/40 & 1/8 & 1/40 & 1/40 & 1/40 & 1/40 & 1/40 \\ 1/40 & 1/40 & 1/8 & 7/24 & 1/40 & 1/40 & 17/40 & 1/40 \\ 1/40 & 1/40 & 1/8 & 7/24 & 7/24 & 1/40 & 1/40 & 17/40 \\ 1/40 & 1/40 & 1/8 & 1/40 & 7/24 & 1/40 & 1/40 & 17/40 \\ 1/40 & 1/40 & 1/8 & 1/40 & 7/24 & 33/40 & 17/40 & 1/40 \end{pmatrix}$$

# The Google matrix



$$A = \begin{pmatrix} 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 & 1 & 0 \end{pmatrix}$$

$$S = \begin{pmatrix} 0 & 0 & 1/8 & 0 & 0 & 0 & 0 & 0 \\ 1/2 & 0 & 1/8 & 1/3 & 0 & 0 & 0 & 0 \\ 1/2 & 1/2 & 1/8 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1/2 & 1/8 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/8 & 1/3 & 0 & 0 & 1/2 & 0 \\ 0 & 0 & 1/8 & 1/3 & 1/3 & 0 & 0 & 1/2 \\ 0 & 0 & 1/8 & 0 & 1/3 & 0 & 0 & 1/2 \\ 0 & 0 & 1/8 & 0 & 1/3 & 1 & 1/2 & 0 \end{pmatrix}$$

$$\alpha = 0.8$$



# Fundamentals of Google research engine

$$G = \alpha S + \frac{1 - \alpha}{N} e e^T \quad e^T = (1, 1, \dots, 1)$$

If the spectrum of the stochastic matrix  $S$  is  $\{1, \lambda_1, \lambda_2, \dots, \lambda_N\}$ , then the spectrum of the Google matrix  $G = \alpha S + (1 - \alpha) e v^T$  is  $\{1, \alpha \lambda_1, \alpha \lambda_2, \dots, \alpha \lambda_N\}$ , where  $v^T$  is a probability vector.

## PageRank probability vector $P$

$$P = GP$$

$P$  is the eigenvector associated to the largest eigenvalue, ie 1

For very large network (such as WWW), no way to directly diagonalize  $G \rightarrow$  **Powermethod**

$$P = \lim_{n \rightarrow \infty} G^n v_0 \quad \forall v_0$$

## Interpretation

$$P_i = \sum_{j \in B_i} \frac{P_j}{k_{out}(j)}$$

The more a node is pointed by important node, the more it is important

Algorithm at the hearth of Google™ search engine.

(Brin & Page cofounders)

# PageRank – Measure of nodes centrality

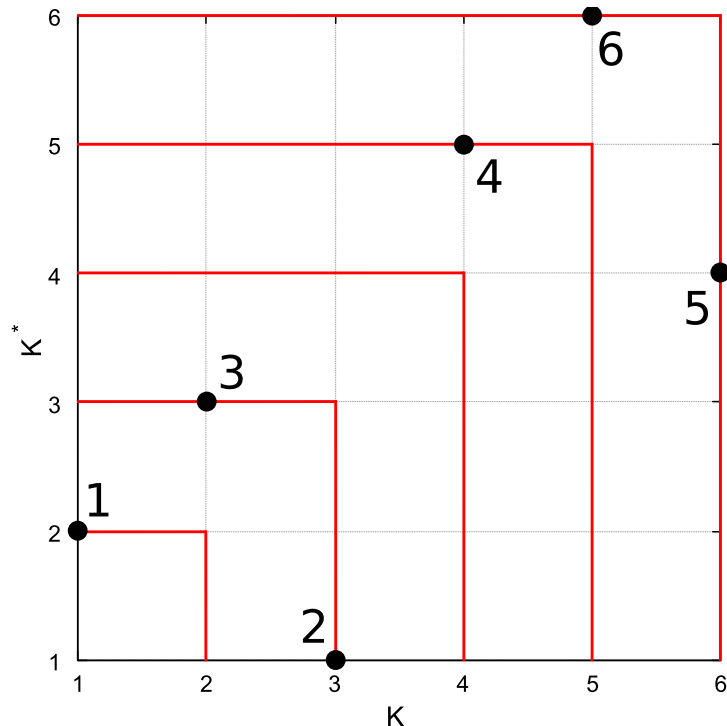
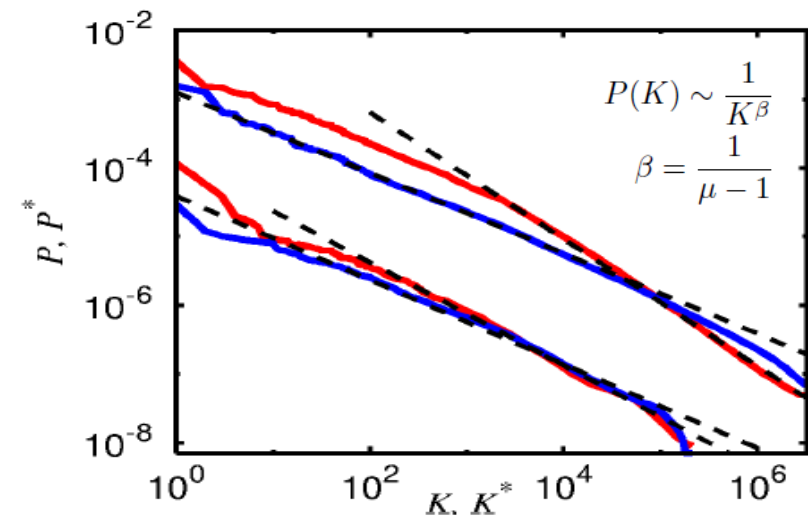
## PageRank

Measures the importance of a node as the property to be pointed by other important nodes (influent node).

## CheiRank

Same as PageRank but considering the adjacency matrix of the inverted network.

Measures the importance of a node as the property to point to other important nodes = communicability



P : PageRank vector

K : PageRank

P\* : CheiRank vector

K\* : CheiRank

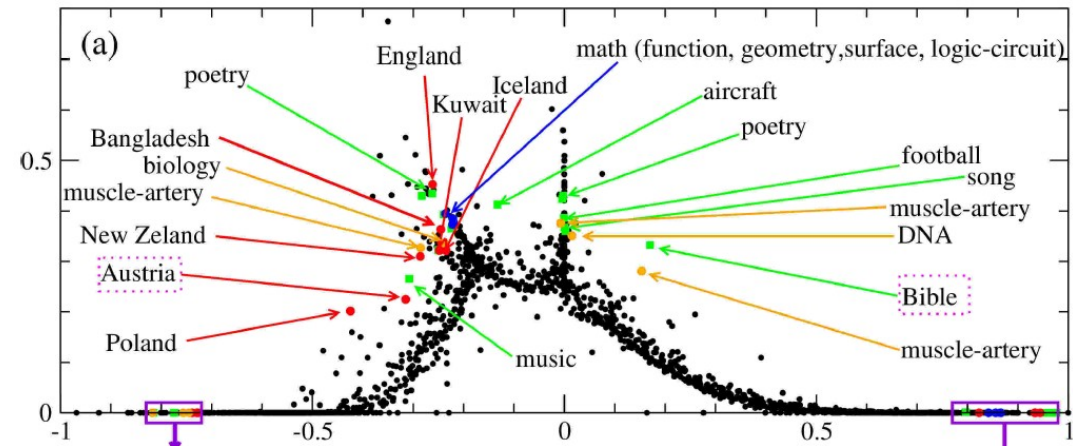
## 2DRank

Mix *PageRank* and CheiRank, measures the propension of a node to be pointed and to point towards other nodes.

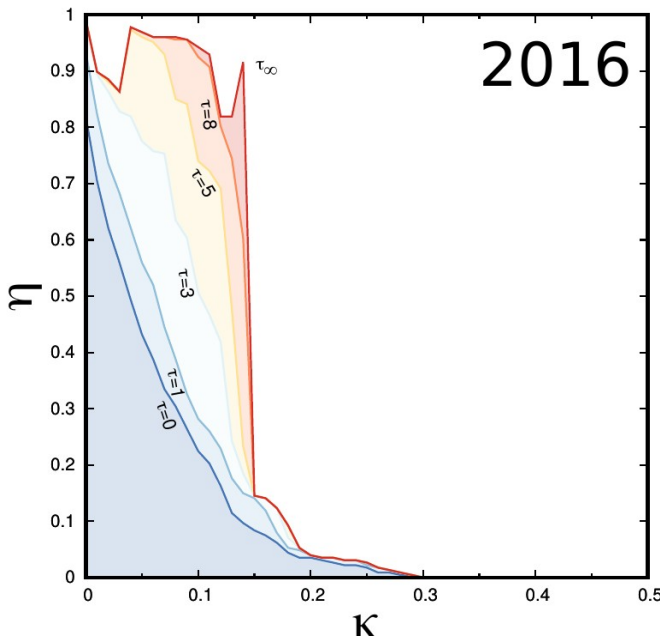
# Plenty of applications in scientific research

Non exhaustive list of applications :

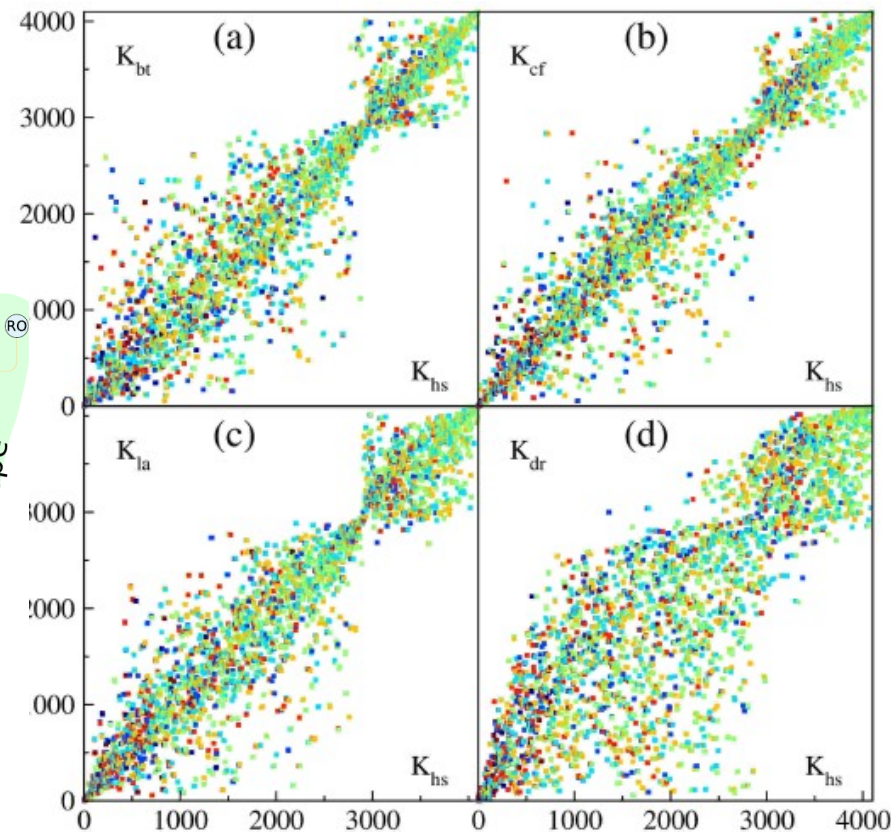
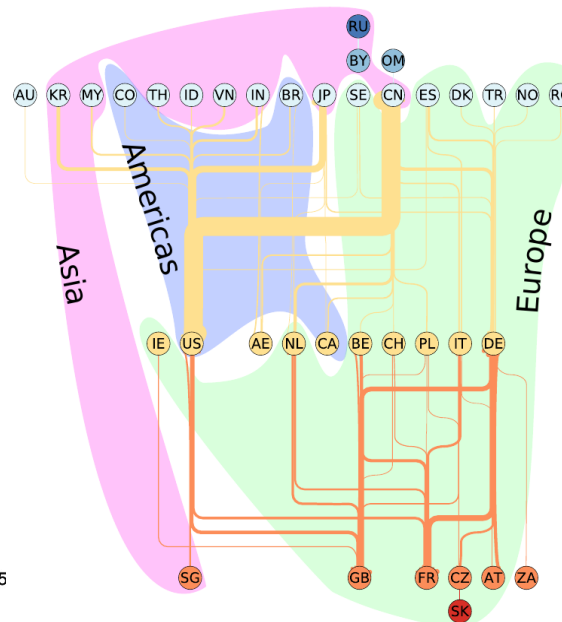
- Linux Kernel networks
- Wikipedia networks
- Twitter network
- World trade network
- Brain neural network
- DNA sequences
- Networks of Game Go moves
- Opinion formation on directed networks
- Contagion on networks
- ...



Spectrum of Google matrix of English Wikipedia



Crisis contagion on World trade network



DNA sequence analysis

# IV Data and Network analysis with Python





# Networkx and Pandas



NetworkX  
Network Analysis in Python

Creation and modification of network objects

- From generative models

- From data

Algorithm ready to use

- Clustering

- Centrality

Network visualization

Documentation and guide available at <https://networkx.org/>

Installing with **`pip install networkx`**

Using with **`import networkx as nx`**



Data manipulation tools  
CSV to Python Dictionary

Data visualization and analysis tools

Documentations and guide available at <https://pandas.pydata.org/>

Installing with **pip install pandas**

Using with **import pandas as pd**

# Numpy



Scientific computing for Python

Linear algebra

Documentations and guide available at <http://numpy.org/>

Installing with **`pip install numpy`**

Using with **`import numpy as np`**



Powerful Matplotlib object to generate graphics

PDF, SVG, JPG, GIF and other output formats

Documentations and guide available at <https://matplotlib.org/>

Installing with **`pip install matplotlib`**

Using with **`import matplotlib.pyplot as plt`**

# Tutorial

Create a python script taking as input: *links.dat* and *nodes.dat*

Generate outputs A and B

A: Network representation in PDF

B: File consisting in the list of nodes with: degree, closeness and betweenness

In case of B, try you own function for degree, betweenness and closeness, before using networkx's ones