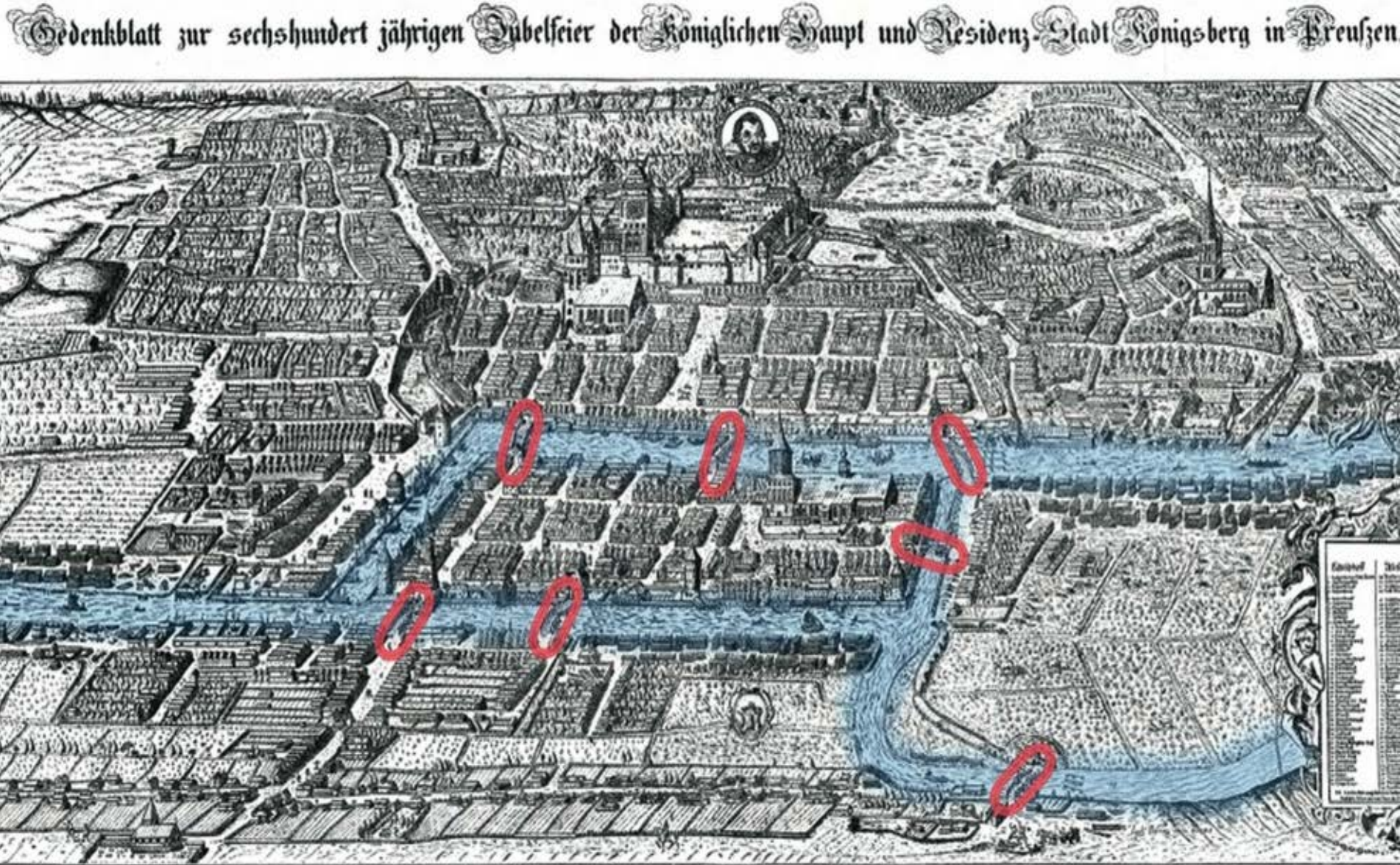


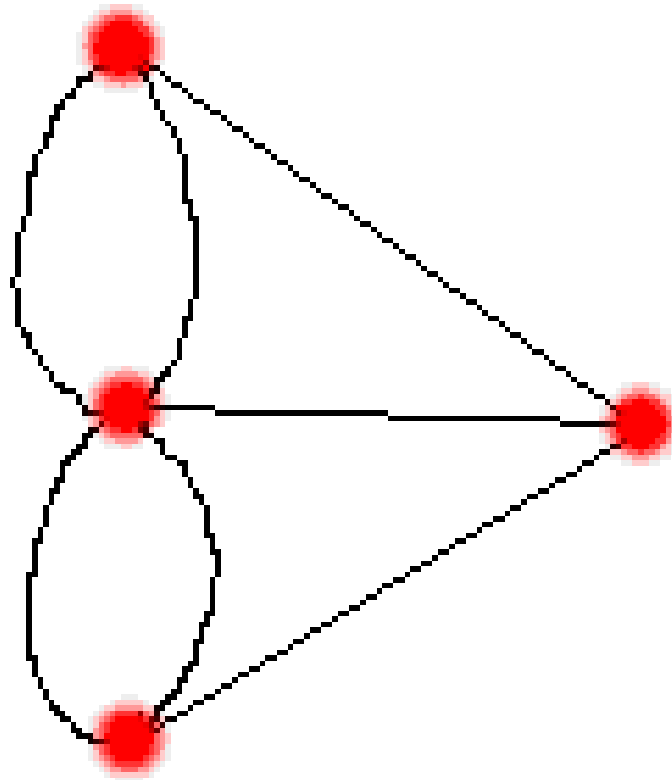
Network thinking in music



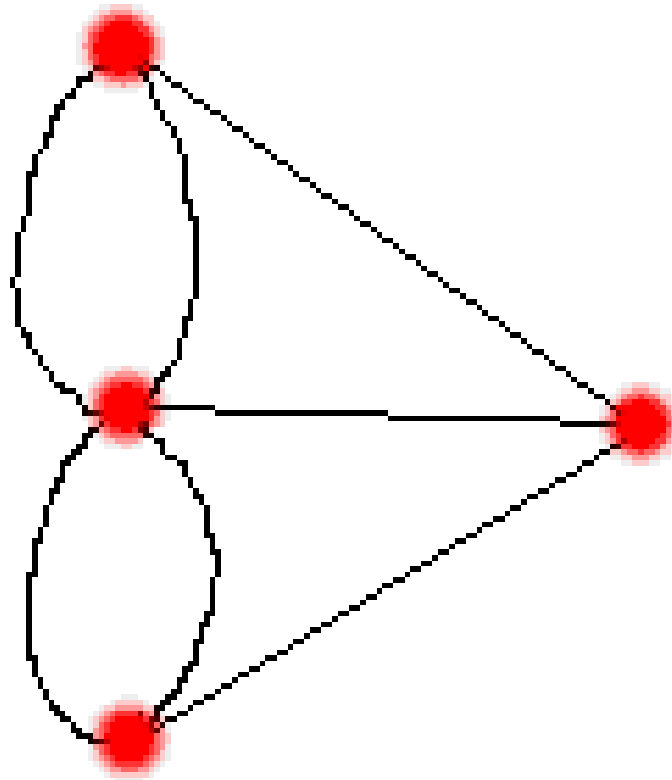
The seven bridges of Königsberg



Network thinking reduces complexity



Hidden properties of network structures



Graph or networks?



Graph and networks

- Networks are found in nature and society

Graph and networks

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- Graphs are the mathematical representation of these networks (a map)

Graph and networks

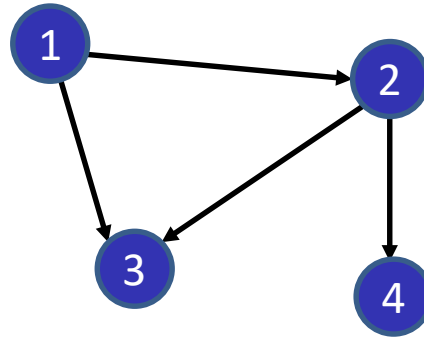
- Networks are found in nature and society
- Graphs are the mathematical representation of these networks (a map)
- In the literature, both are used interchangeably

An advice network

- Emma (1) helps Mason (2)
- Emma (1) helps William (3)
- Mason (2) helps William (3)
- Mason (2) helps Sophia (4)

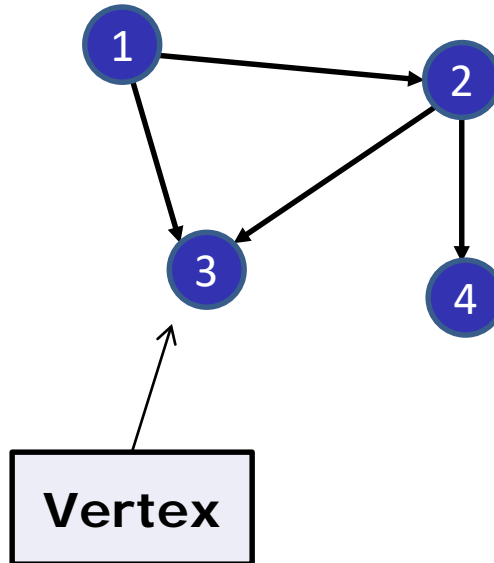
A graph

- Emma (1)
- Mason (2)
- William (3)
- Sophia (4)



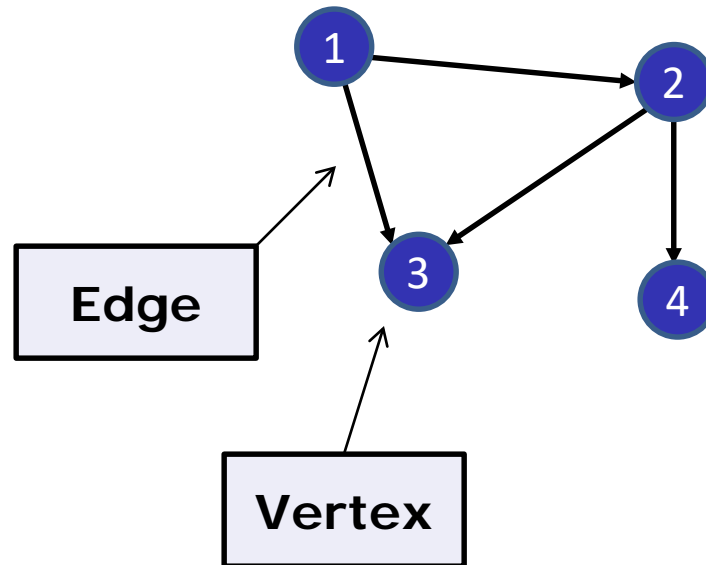
A graph

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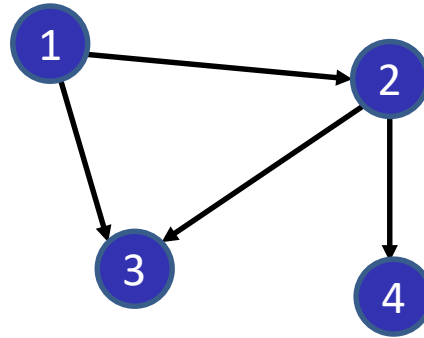
A graph

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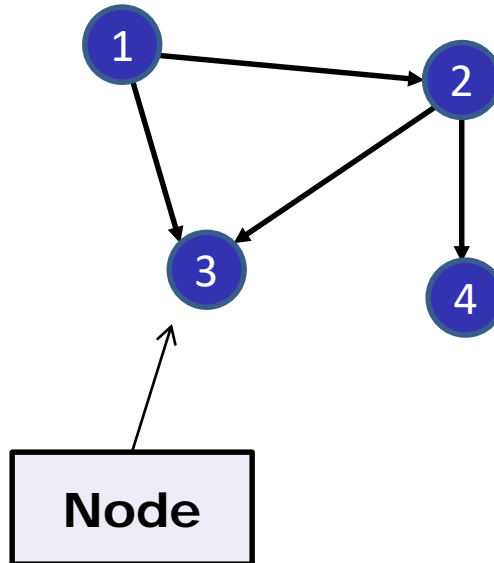
A network

- Emma (1)
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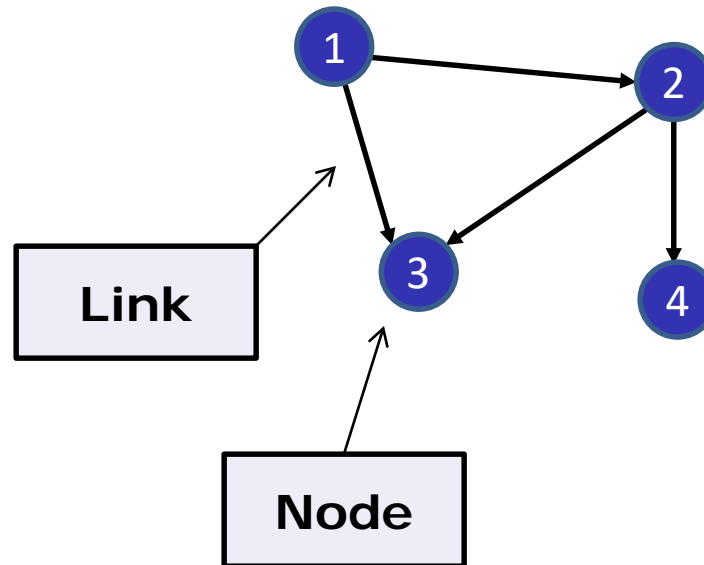
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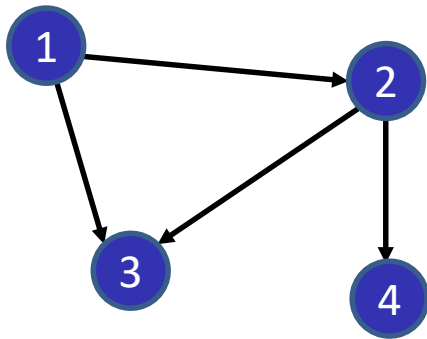


Network terms

- N = number of nodes (size of the network)
- $N = 4$
- The network is composed by the nodes $i = 1, 2, \dots, N$
- L = number of links
- $L = 4$
- The connection between Mason and William [*Mason (2) helps William (3)*] is denoted as (2,3)
- A graph might be denoted as G , its vertex set as $V(G)$, and its edge set as $E(G)$

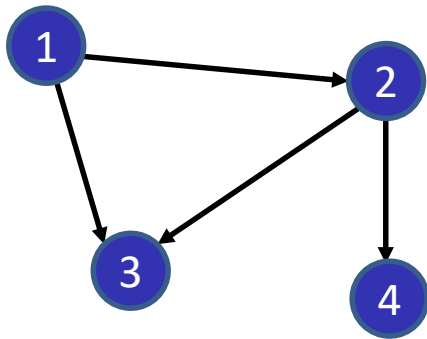
Different types of networks

Directed network

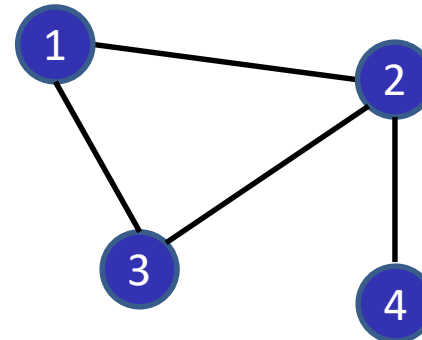


Different types of networks

Directed network

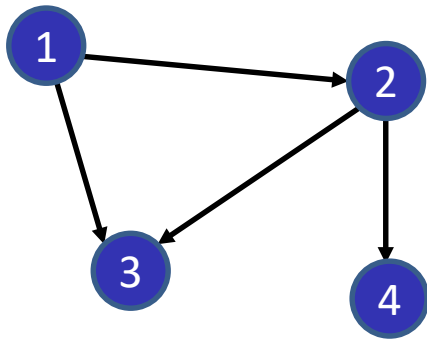


Undirected network

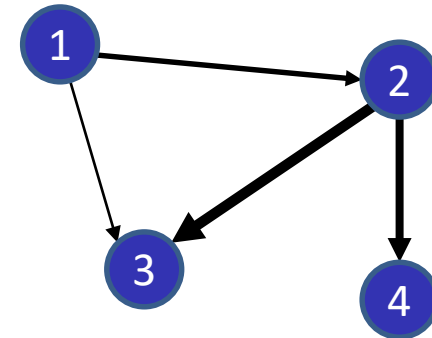


Different types of networks

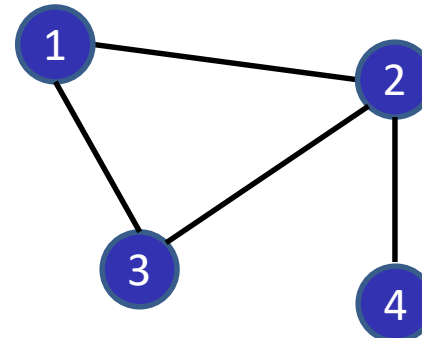
Directed network



Weighted network



Undirected network

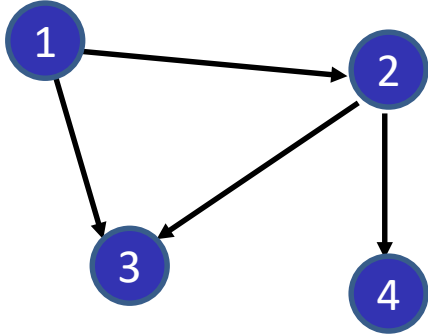


Real world networks

Network name	Nodes	Links	Direction of ties	N
Internet	Routers	Internet connections	No	200,000
WWW	WebPages	Hyperlinks	Yes	500,000
Friendship network	Individuals	Friendship	No	200
Actor network	Actors	Co-acting	No	200,000
Patent citations network	Patent documents	Citations	Yes	7,000,000
Co-invention network	Inventors	Co-patenting	No	200,000

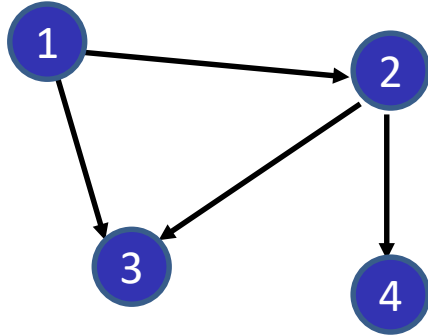
Network representations

Directed graph (digraph)



Network representations

Directed graph (digraph)

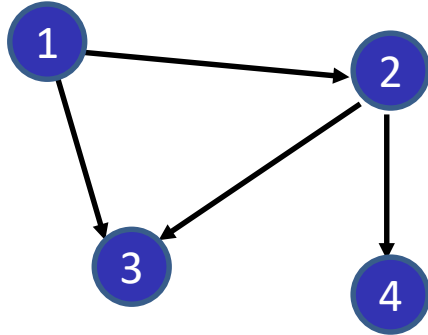


Edge list

Vertex	Vertex
1	2
1	3
2	3
2	4

Network representations

Directed graph (digraph)



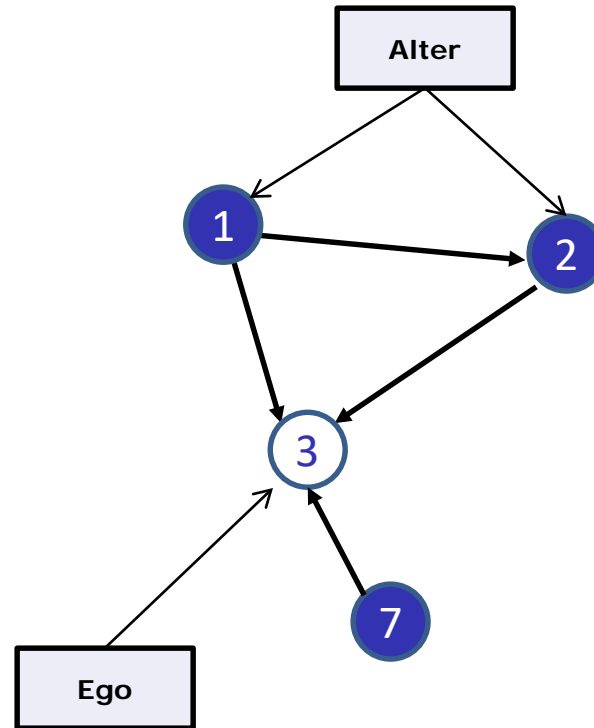
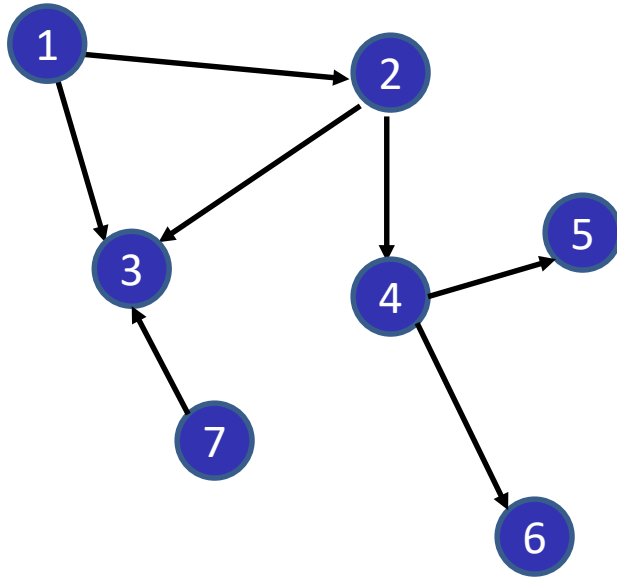
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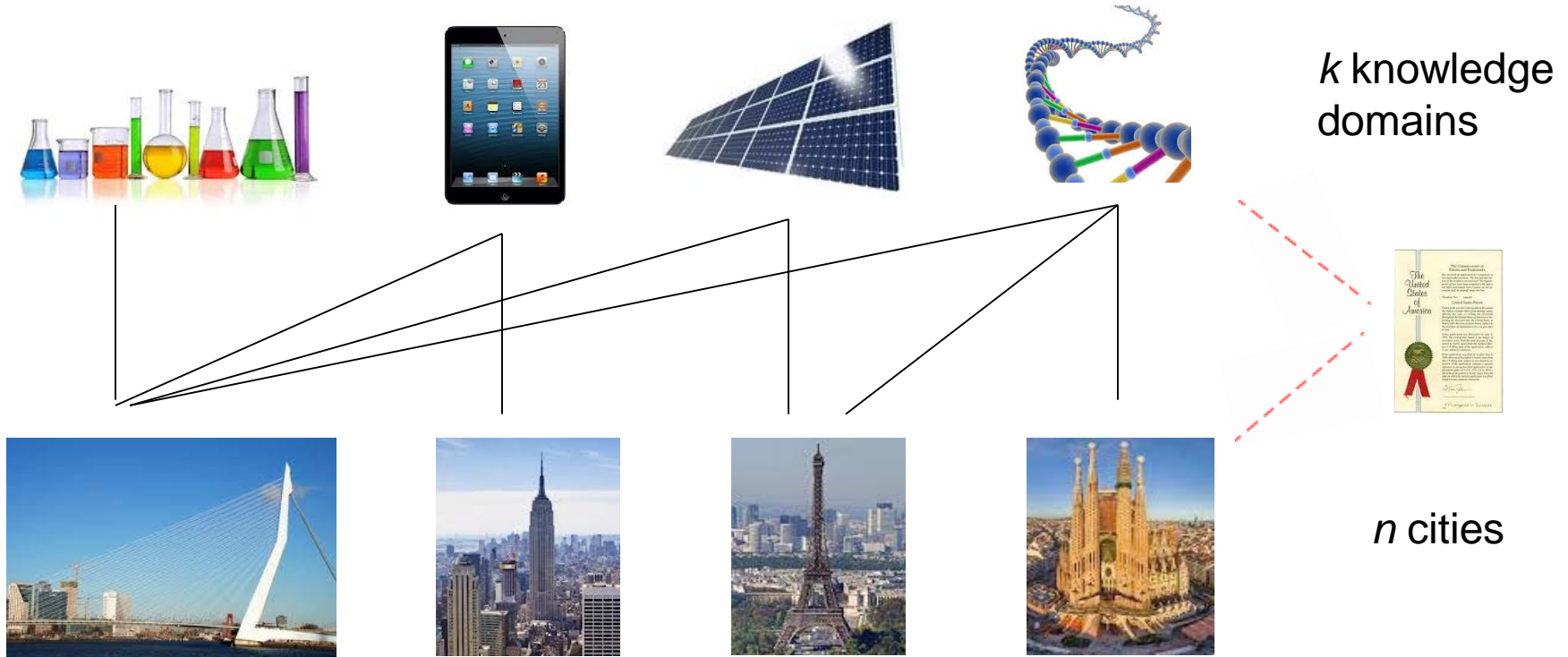
Adjacency matrix

Vertex	1	2	3	4
1	-	1	1	0
2	0	-	1	1
3	0	0	-	0
4	0	0	0	-

Ego networks and whole networks



Bipartite network (2-mode)



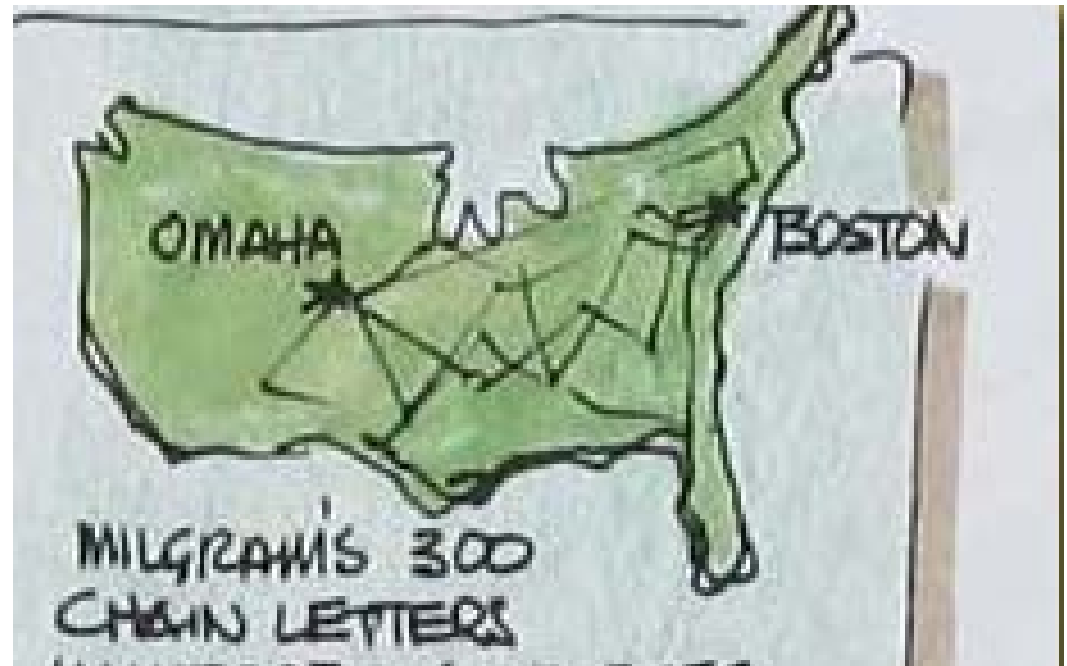
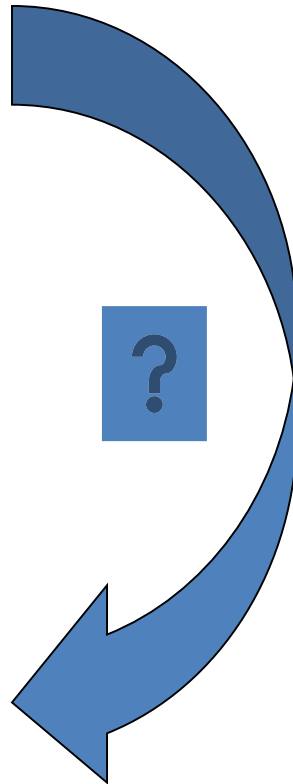
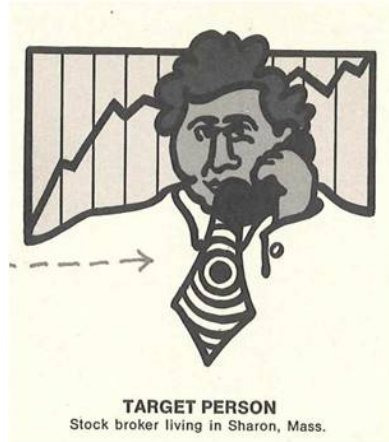
It is possible to formalize the data that connect **cities** to the **knowledge** they produce as a n by k bipartite **network** (two different sets of nodes)

A **link** between a city i and a knowledge domain j means that i produces knowledge in category j

Multiplex networks

- Emma (1) helps Mason (2)
- Emma (1) is friends with Mason (2)
- Emma (1) works with Mason (2)
- ...

The small world experiment of Milgram



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- Roughly a third of the letters eventually arrived at the target

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- Goal of reaching the target as rapidly as possible.
- Roughly a third of the letters eventually arrived at the target
- In a median of six steps (6 degrees of separation)

Small world Milgram – 2 lessons

- Short paths are ubiquitous in human networks (it is a small world after all)
- Individuals are good at finding these short paths (search process: what if no name? no occupation, etc...)

The Bacon number



The screenshot shows the Oracle of Bacon website. At the top, the title "THE ORACLE OF BACON" is centered in a large, white, serif font. To the left of the title is a classical marble bust of a man, and to the right is a modern portrait of actor Kevin Bacon. Below the title, on the left side, is a dark vertical menu with white text links: "Welcome", "Credits", "How it Works", "Contact Us", and "Other stuff »". In the center, there is a search interface with a text input containing "Kevin Bacon", a "to" label, another text input containing "sean connery", a "Find link" button, and a "More options >>" button. Below the search bar, there are three icons: a smartphone, the Android robot, and the Windows logo. To the right of these icons is a text block that reads: "Hey, smartphone and tablet users! Check out the Six Degrees app for [iOS](#), [Android](#), and [Windows Phone](#). Click the icons to the left for more details."

<https://oracleofbacon.org/>

Structural features of networks

- Real-world networks are characterized by:
 - (1) Small average path length

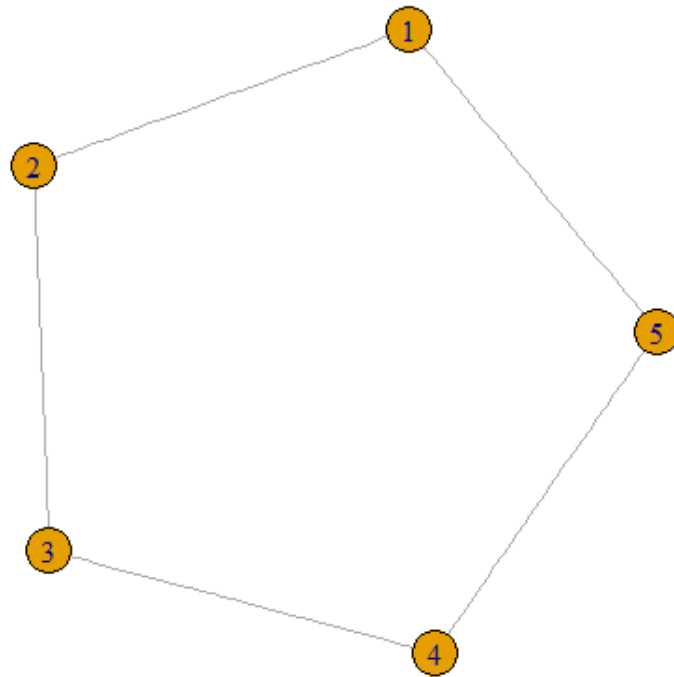
Structural features of networks

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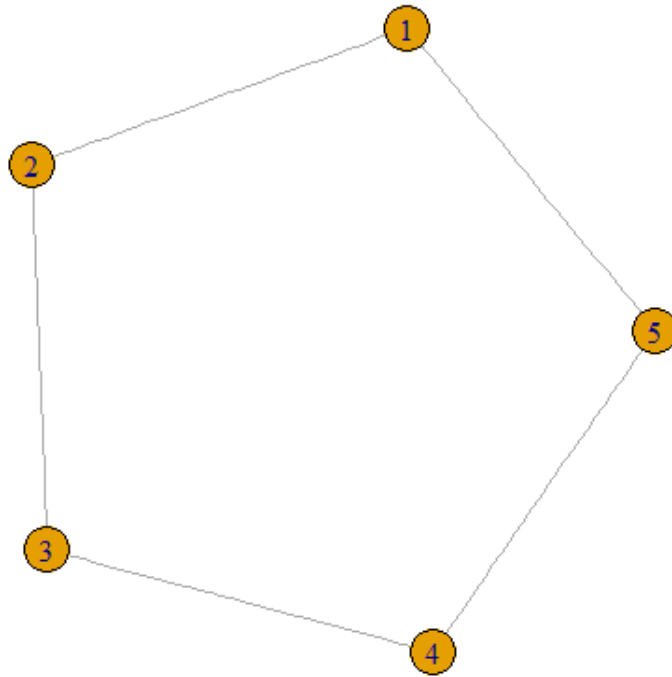
(1) Small average path length

$$l_s = \frac{1}{n.(n-1)} \cdot \sum_{i \neq j} d(v_i, v_j)$$

Compute average path length



Compute average path length

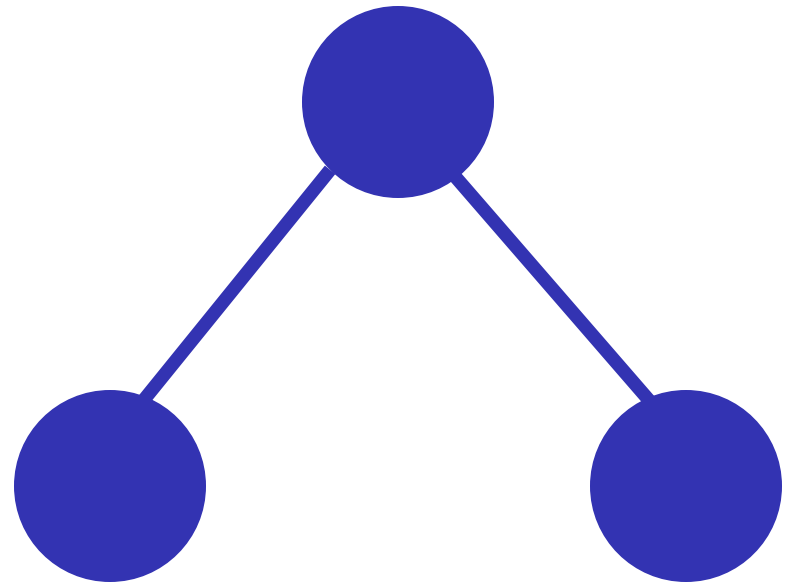


$$\text{Average path length} = (1+2+2+1)/4 = 1.5$$

Erdős–Rényi Random Graph model

- Early attempt to generate random graphs
- A random network consists of N nodes where each node pair is connected with the same probability p
- Path length in random graphs is short (similar to path length in real-world networks)
- Often used as a benchmark for network metrics
- As p increases the network tends to form a giant component (percolation)

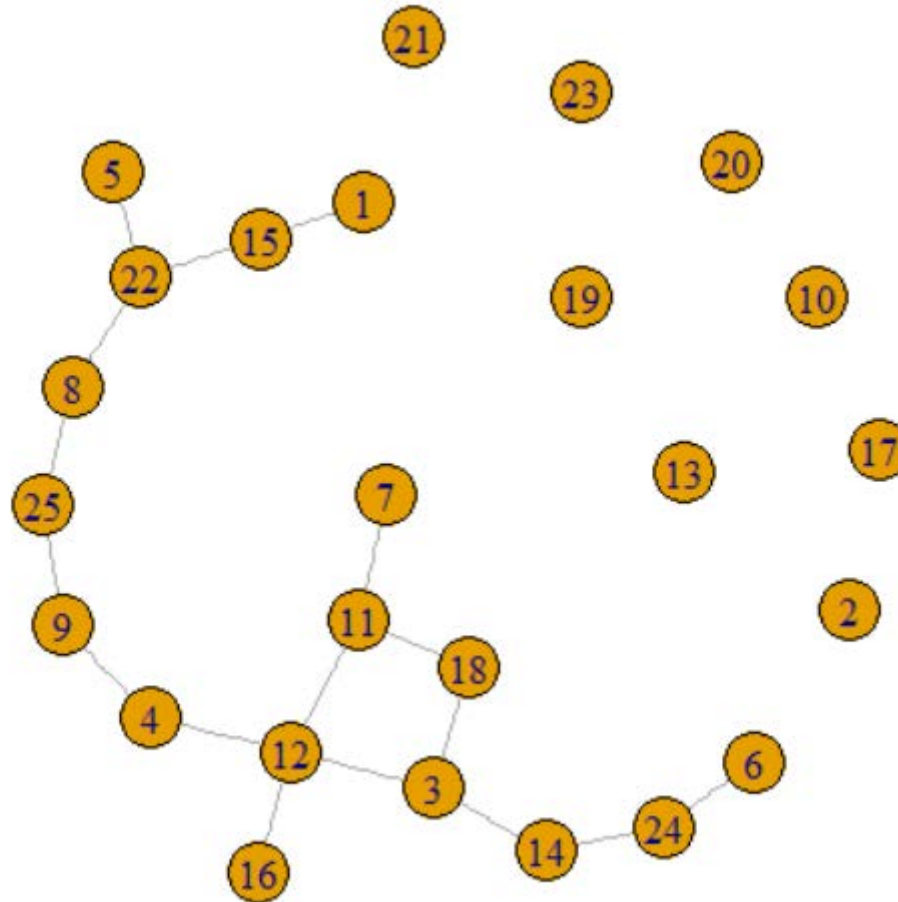
Clustering in networks



Structural features of networks

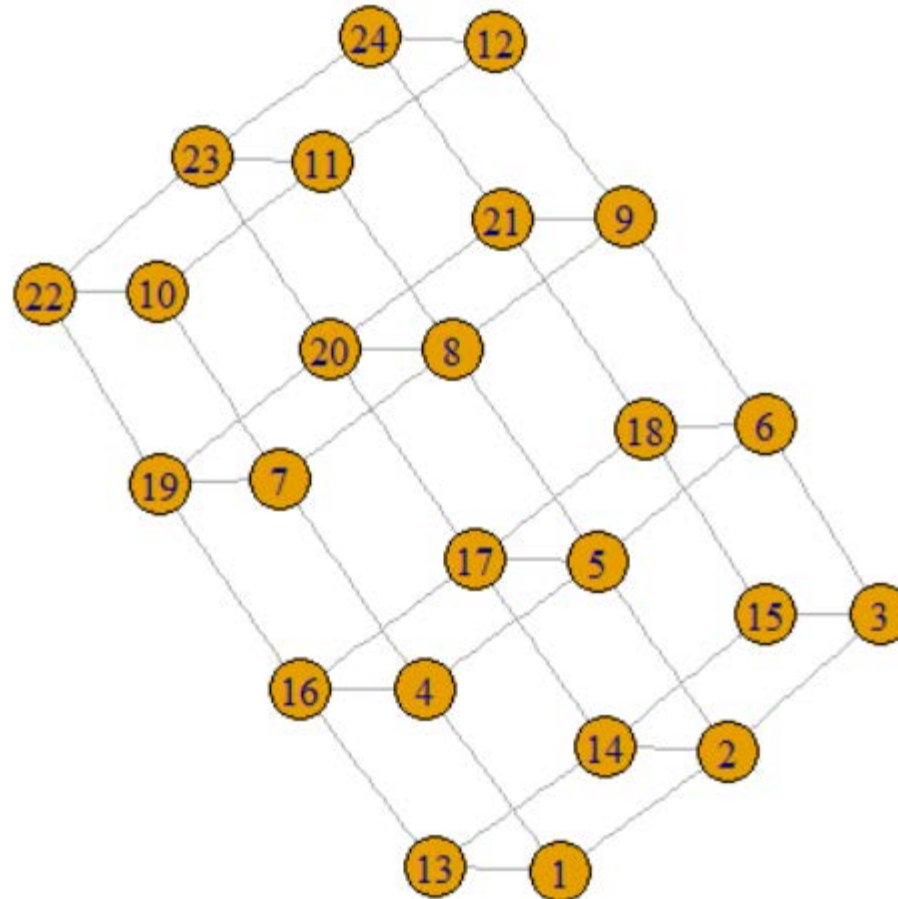
- Real-world networks are characterized by:
 - (1) Small average path length
 - (2) High clustering coefficient [equation]

Random network



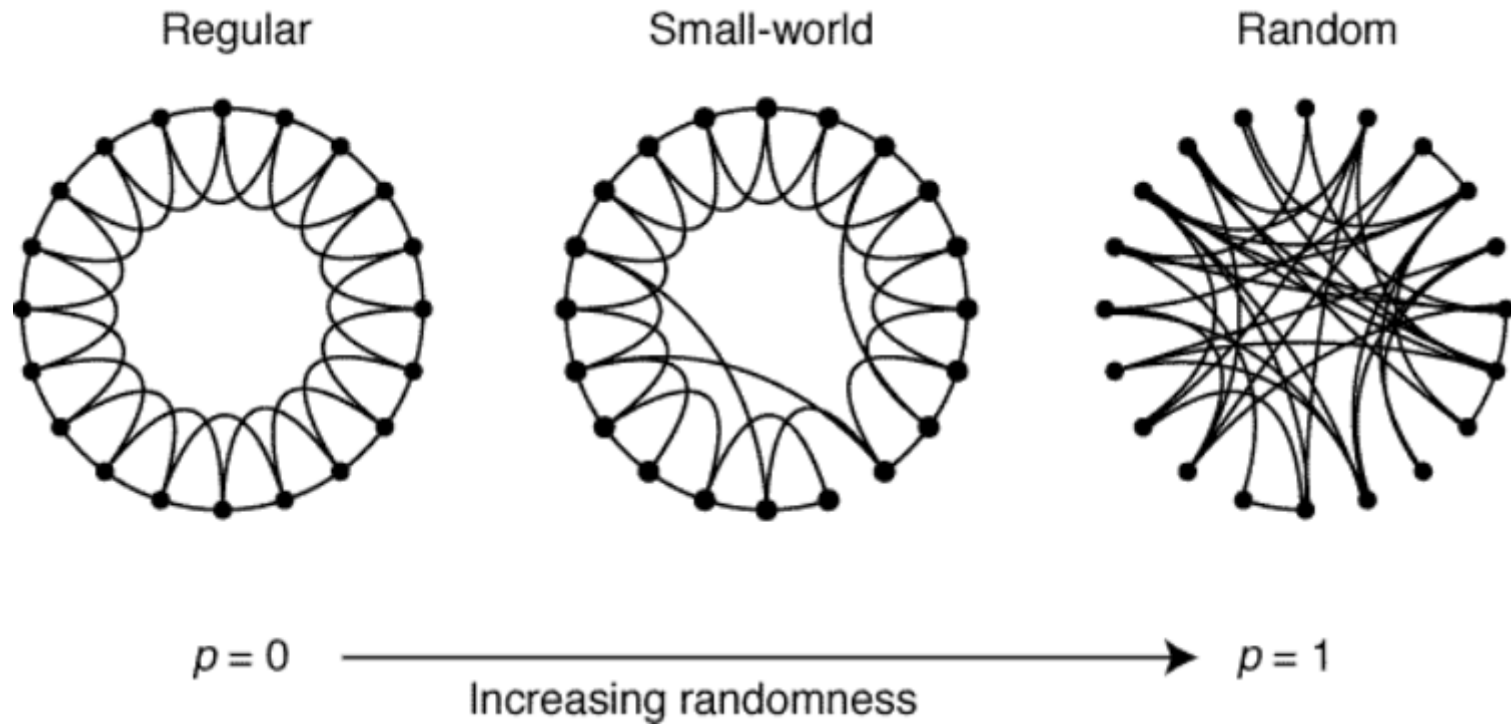
(1) Small average path length

Lattice



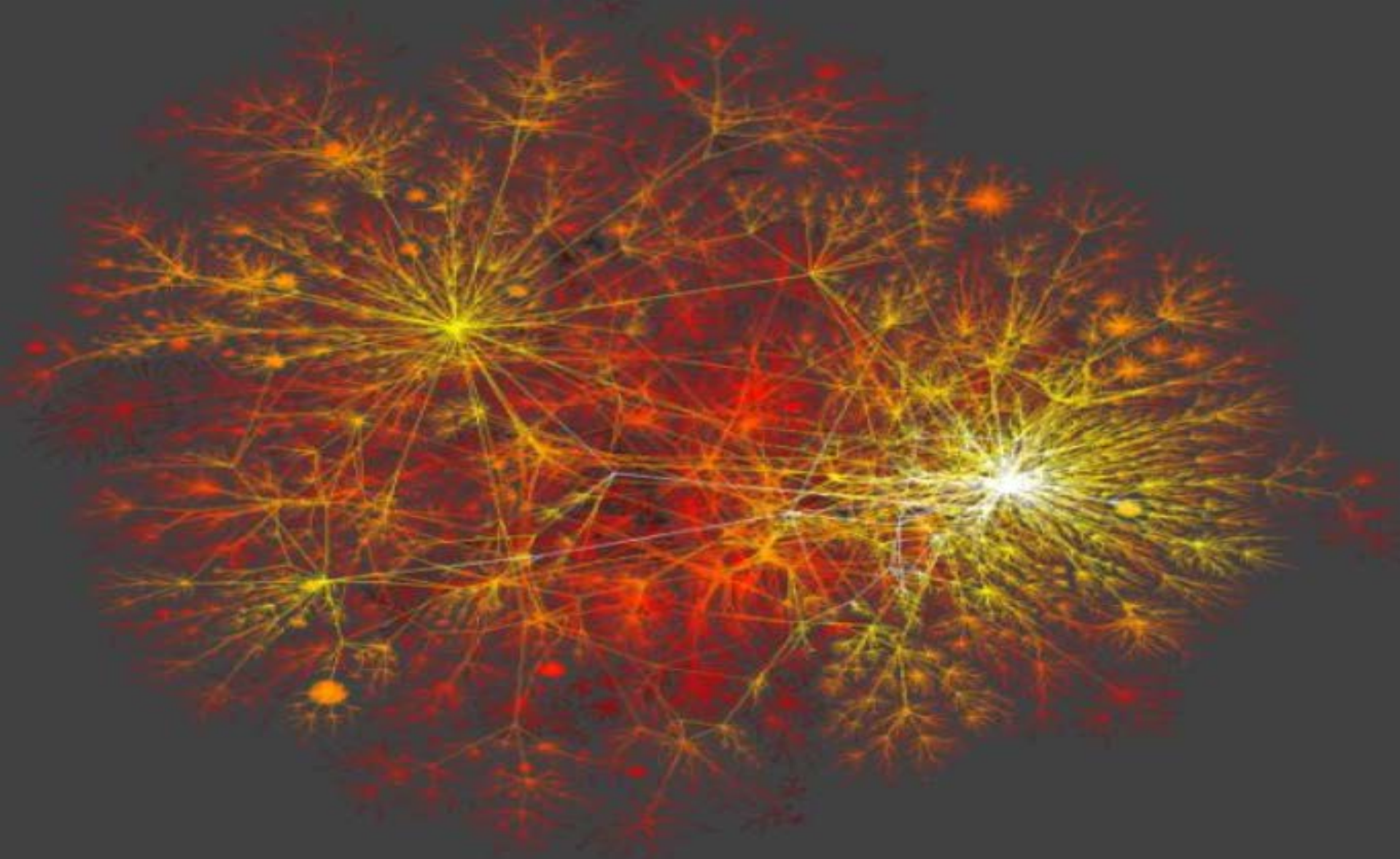
(2) High clustering coefficient

Small world model



Watts, D. J., & Strogatz, S. H. (1998). Collective dynamics of 'small-world' networks. *Nature*, 393(6684), 440-442.

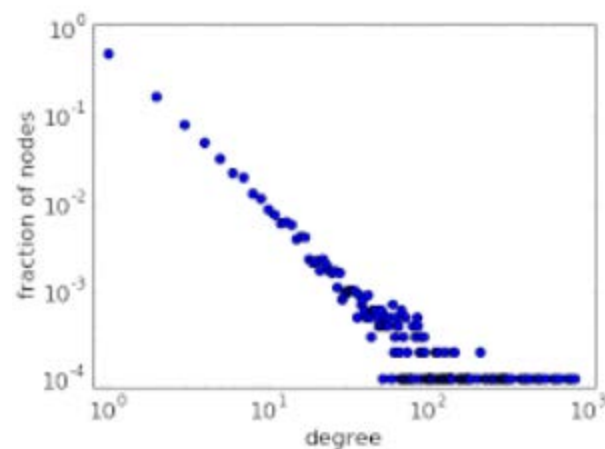
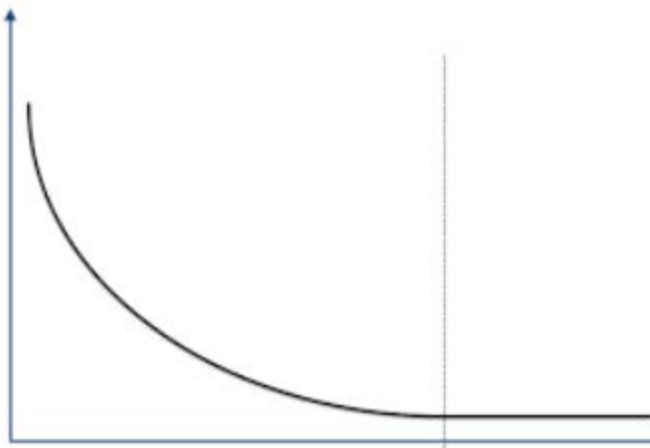
Structural inequality



Structural features of networks

- Real-world networks are characterized by:
 - (1) Small average path length
 - (2) High clustering coefficient
 - (3) Unequal degree distribution

Long tails and power law distribution



Barabási-Albert model

- A few hubs have a lot of connections
- Most of the nodes have very few
- Barabási-Albert model generates these type of networks by "preferential attachment"
- New network members prefer to make a connection to the more popular existing members

Implementation of the BA model

- The model starts with two nodes connected by an edge
- At each step, a new node is added
- A new node picks an existing node to connect to randomly, but with some bias
- A node's chance of being selected is proportional to the number of connections it already has (degree)

Complex networks

- Non-trivial topology
- Inherently dynamical properties – emergent behavior
- High structural heterogeneity