



NETWORK SCIENCE

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Learning Goals

- Understand the context of network Science
- Explain main concepts and measurements
- Use mail tools
- Apply to specific context



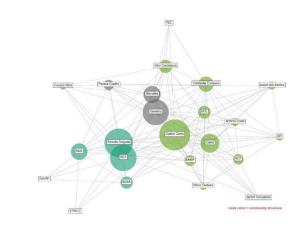
Agenda

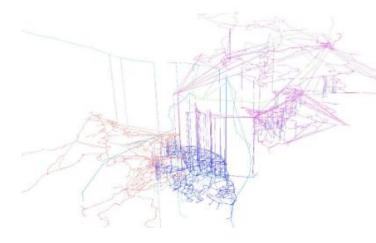
- Complex Network
- Network Elements
- Measurements
- Node Degree
- · Networks metrics
- Node metrics
- Tools
- Applications
- Challenges



Complex Network

- telecommunication networks
- computer networks
- biological networks
- cognitive and semantic networks
- social networks



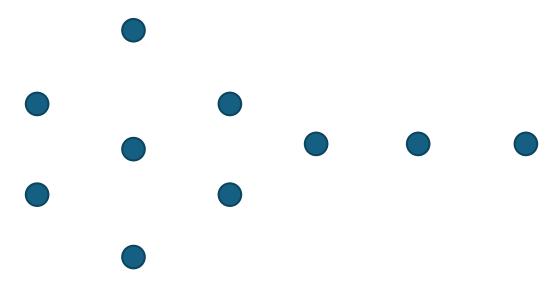




- "vertex" and "edge" (Mathematics)
- "nodes" and "connections" (or links) (Computer Science)
- "actors" (or "agents") and "relationships" (Sociology)
- "site" and "bond" (Physics)
- "dot" and "arcs" (or ties)

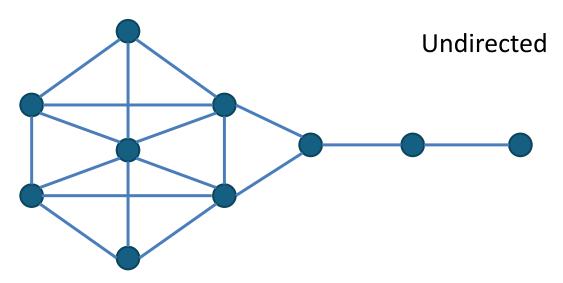


Vertex, nodes and actors



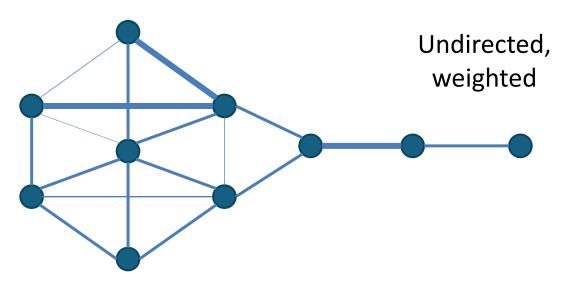


• Edges, arcs, links and relationships



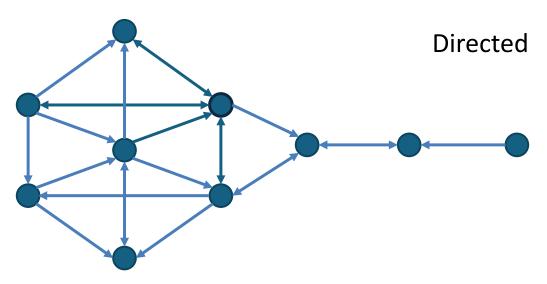


• Edges, arcs, links and relationships





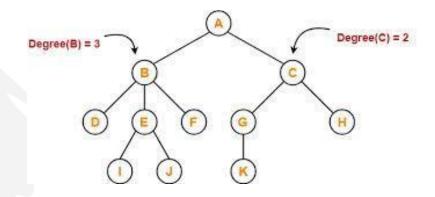
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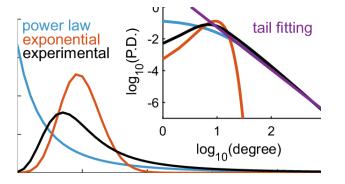


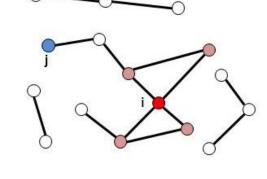


Node Degree

- number of connections it has to other nodes
- degree distribution is the probability distribution of these degrees over the whole network.



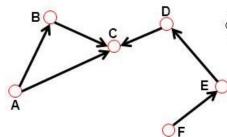




$$\langle k \rangle = \frac{1}{N} \sum_{i=1}^{N} k_i \qquad \langle k \rangle \square \frac{2L}{N}$$

N – the number of nodes in the graph

Directed

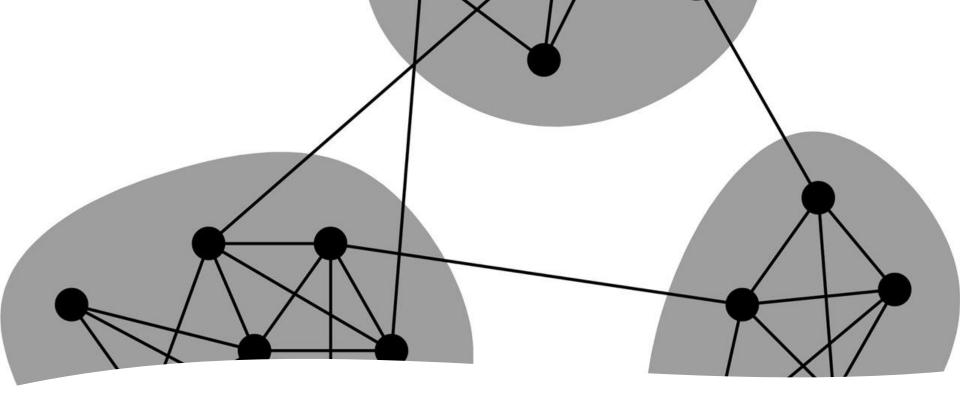


$$\left\langle k^{\,in} \right\rangle \equiv \frac{1}{N} \sum_{i=1}^{N} k^{\,in}_{i}, \quad \left\langle k^{\,out} \right\rangle \equiv \frac{1}{N} \sum_{i=1}^{N} k^{\,out}_{i}, \quad \left\langle k^{\,in} \right\rangle = \left\langle k^{\,out} \right\rangle$$

$$\langle k \rangle \Box \frac{L}{N}$$

Network metrics

Average Degree (average links per node)

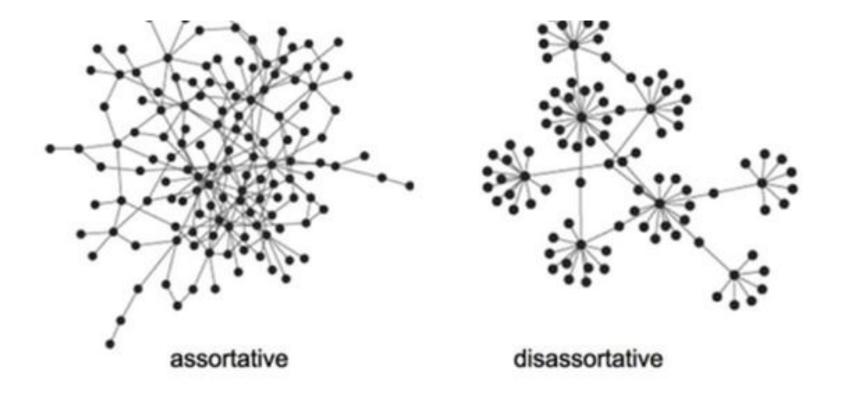


Network metrics

Modularity

- is one measure of the structure of networks or graphs.
- It was designed to measure the strength of division of a network into modules (also called groups, clusters or communities).
- Networks with high modularity. have dense connections between the nodes within modules but sparse connections between nodes in different modules



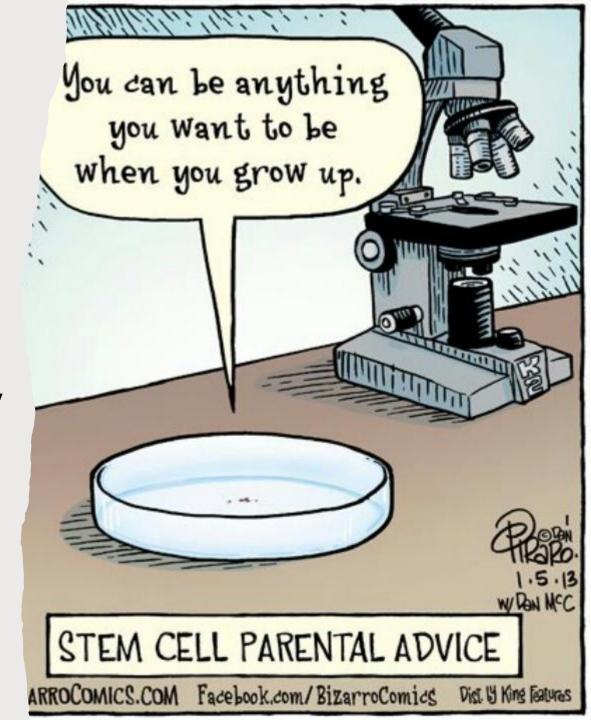


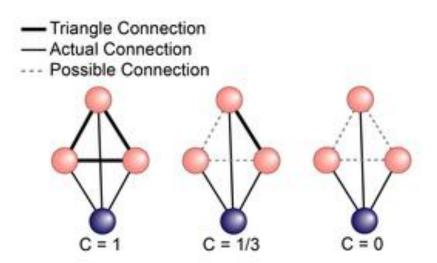
Network metrics

- Connected Components
- Social networks are highly assortative (homophily): high degree nodes connect to other high degree nodes
- technological are disassortative: high degree nodes connect to low degree nodes
- Assortative and disassortative mating



- Clustering Coefficient
- Centrality
- Closeness Centrality
- Betweenness Centrality
- Eigenvector Centrality

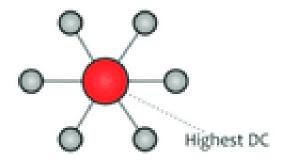




Node metrics

a clustering coefficient is a measure of the degree to which nodes in a graph tend to cluster together

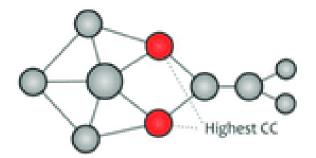
Degree centrality



Node metrics

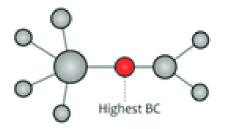
Centrality refers to indicators which identify the most important vertices within a graph

Closeness centrality



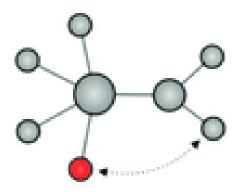
- In connected graphs there is a natural distance metric between all pairs of nodes, defined by the length of their shortest paths.
- The farness of a node is defined as the sum of its distances to all other nodes, and its closeness is defined as the reciprocal of the farness.
- Thus, the <u>more central a node is the lower its total</u> <u>distance to all other nodes.</u>

Betweenness centrality



- Betweenness centrality
- quantifies the number of times a node acts as a bridge along the shortest path between two other nodes.

Eigenvector centrality

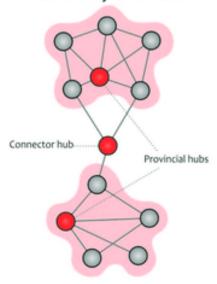


- Eigenvector centrality
- a node is connected to many nodes
 who themselves have high scores.

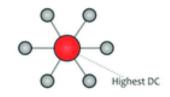
Node metrics

Centrality

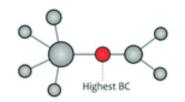
Centrality and hubs



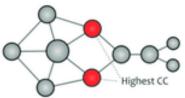
Degree centrality



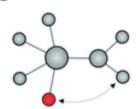
Betweenness centrality



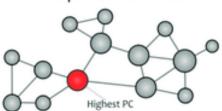
Closeness centrality



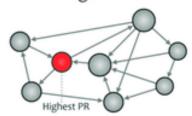
Eigenvector centrality



Participation coefficient



PageRank



Tools







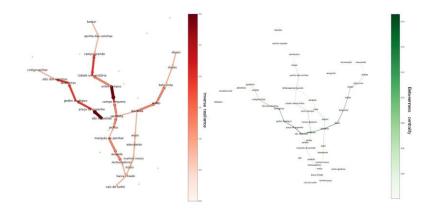


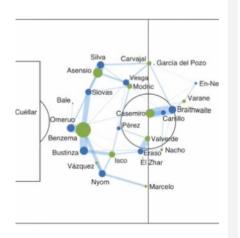


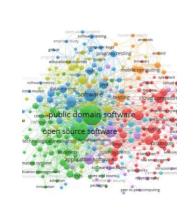


Applications

- . Crime study
- Gang and organized crime
- · Research collaboration
- Papers citations
- · Social media study
- . Team sports

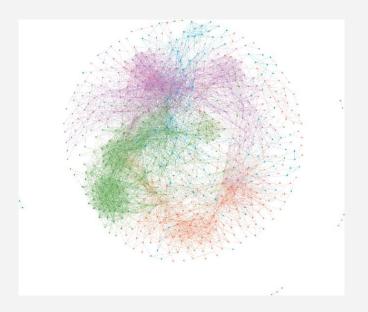






Applications

- Health and disease spread
- · Capital participation
- Equity investment
- · Internal communication
- Commercial relationship between enterprises

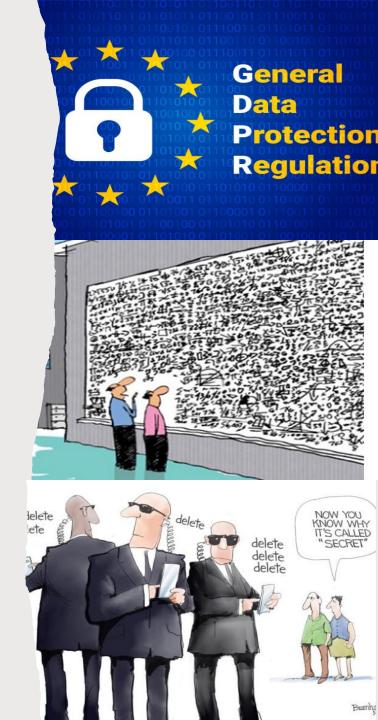




Challenges

- Internal information
- Secrecy
- · GDPR
- Conceptual Complexity
- Main value is obtained by integrate with other approaches

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Python

















Conclusions

- . Complex Network
- . Network Elements
- Measurements
- Node Degree
- Networks Overview
- Node Overview
- · Tools
- Applications
- Challenges



References

Aparicio, J., Arsenio, E., Henriques, R. (2023). Tex2net: A Package for Storytelling Using Network Models. In Proceedings of the 41st ACM International Conference on Design of Communication (SIGDOC '23). Association for Computing Machinery, New York, NY, USA, 119–125. https://doi.org/10.1145/3615335.3623022

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