



Social Data Management

Degree Correlations

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Hubs Connecting to Hubs

Previously, we talked about **hubs**, i.e., high degree nodes in social networks.

We saw that new nodes tend to connect to high degree hubs.

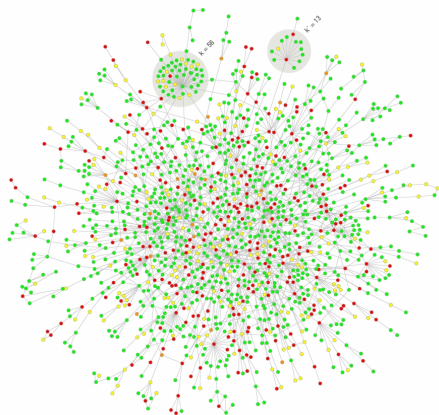
The question we wish to answer now is: **Is this behaviour general? Do nodes connect to other nodes of similar degree?**

Correlations in Social Networks

In **social networks**, hubs tend to connect to other hubs, e.g., celebrities dating other celebrities.

Correlations in Other Networks

In **other networks**, hubs tend to connect to very small degree nodes.



Protein interaction network

Degree Correlations

In a **random model**, we can consider that the probability of a node having degree k to another node having degree k' :

$$p_{k,k'} = \frac{k \cdot k'}{2L} \quad (1)$$

- $p_{k,k'}$ predicts that high degree nodes *should* be more likely to connect to other high degree nodes
- in the protein-interaction network, this does not happen: the probability that a high degree node connects to a low degree node is higher than predicted

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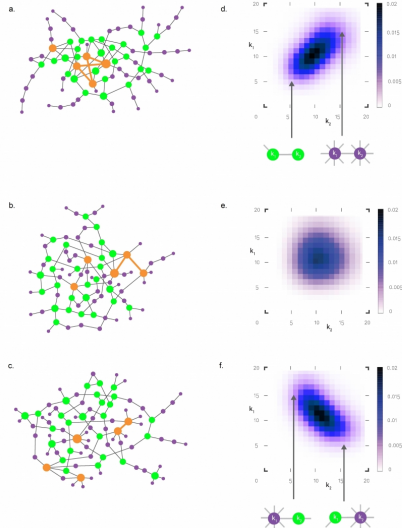
Impact of Degree Correlations

Assortativity and Disassortativity

Depending on how the hubs link to each other, we have 3 types of networks:

1. **Neutral networks:** the wiring is random, links between hubs corresponds to the ones expected by chance as in Eq. 1
2. **Assortative networks:** in which nodes tend to connect to other nodes of similar degree
3. **Disassortative network:** networks in which hubs avoid other hubs

Assortativity and Disassortativity



Assortativity and Disassortativity

The information is captured in a **degree correlation matrix** e where e_{ij} encodes the probability that a node of degree i connects to a node of degree j .

Taking a *random network* we know that the probability that there is a k degree node at the end of a randomly selected link:

$$q_k = \frac{k p_k}{\langle k \rangle},$$

giving

$$e_{ij} = q_i q_j.$$

A network exhibits **degree correlations** if it deviates from e_{ij} .

Perfect Assortativity

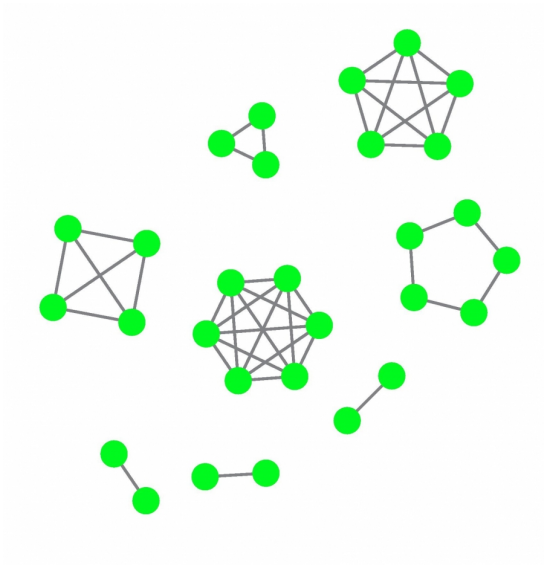


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Correlation Function

Degree correlation function:

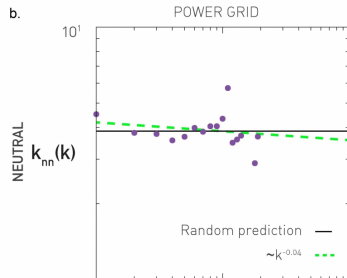
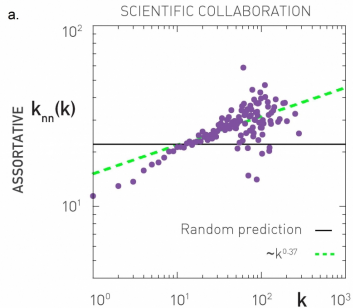
$$k_{nn}(k) = \sum_{k'} k' P(k'|k), \quad (2)$$

where $P(k'|k)$ is the conditional probability that following a link from degree k node we reach a node of degree k'

Depending on the network we have:

- in **neutral networks**, $k_{nn}(k) = \frac{\langle k^2 \rangle}{\langle k \rangle}$ – independent of the node's degree and only dependent on the global characteristics of the network;
- in **assortative networks** hubs tend to connect to other hubs – the higher the degree k is, the higher the avg. degree of the neighbours; and
- in **disassortative networks**, k_{nn} decreases with k .

Correlation Function



Approximating the Correlation Function

The above figures suggest a function of the form:

$$k_{nn}(k) = ak^{\mu}. \quad (3)$$

The sign of the **correlation exponent** μ characterizes the type of network:

- $\mu > 0$ assortative networks
- $\mu = 0$ neutral networks
- $\mu < 0$ disassortative networks

Degree Correlation Coefficient

We can also capture using a single value, the **degree correlation coefficient**:

$$r = \sum_{jk} \frac{jk(e_{jk} - q_j q_k)}{\sigma^2}, \quad (4)$$

where

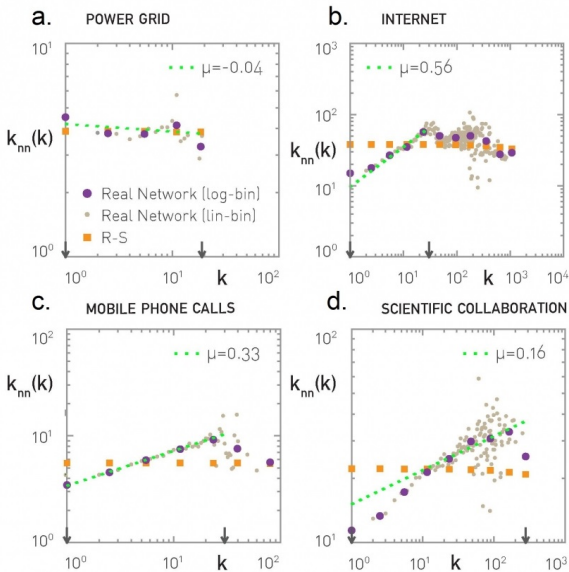
$$\sigma^2 = \sum_k k^2 q_k - \left(\sum_k k q_k \right)^2.$$

This is equivalent to the **Pearson correlation coefficient** between the degrees of the nodes on each link.

$r \in [-1, 1]$ also characterizes the type of network:

- $r < 0$ assortative networks
- $r = 0$ neutral networks
- $r > 0$ disassortative networks

Correlations in Real Networks



Correlations in Real Networks

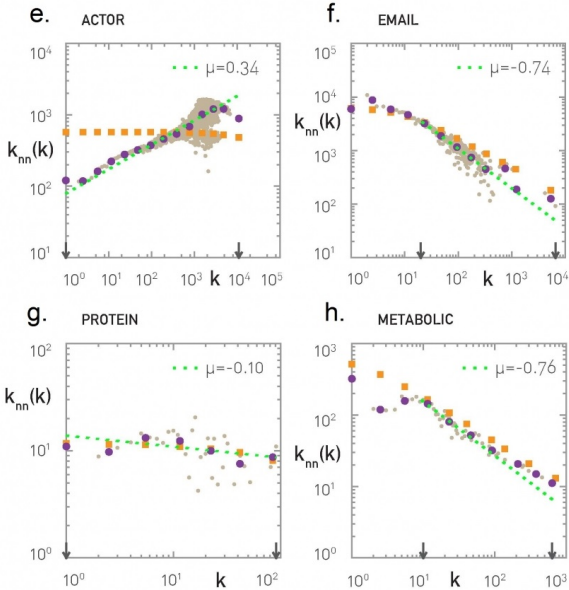


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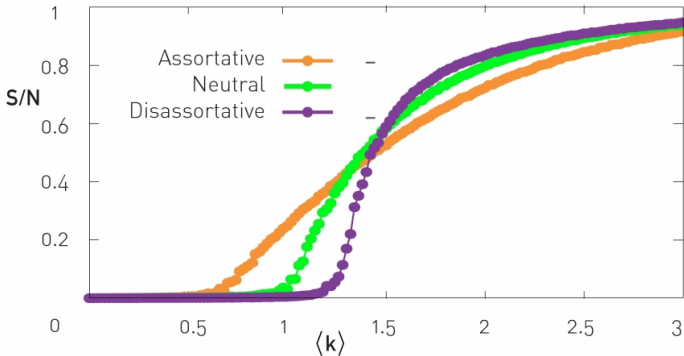
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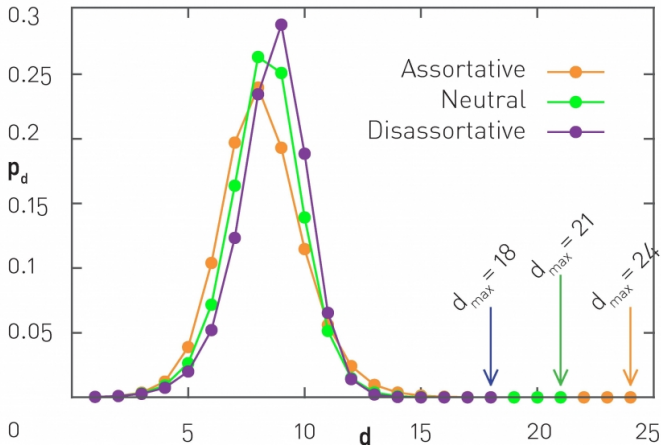
Giant Component

How does the assortativity influence the network? Depending on $\langle k \rangle$, the size of the giant component appears at different time steps – influence on **network robustness**



Other Consequences

- **average path length** is lower in assortative networks
- degree correlations influence stability (perturbations, stimuli)
- they influence greatly the cost of the **vertex cover** problem



Acknowledgments

Figures in slides 5, 9, 11, 14, 17, 18, 20, and 21 taken from the book “Network Science” by A.-L. Barabási. The contents is partly inspired by the flow of Chapter 9 of the same book.

<http://barabasi.com/networksciencebook/>

References i



Park, J. and Newman, M. E. J. (2003).

Origin of degree correlations in the internet and other networks.

Phys. Rev. E, 68.



Pastor-Satorras, R., Vázquez, A., and Vespignani, A. (2001).

Dynamical and correlation properties of the internet.

Phys. Rev. L., 87(25).