

How tightly connected are communities?

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08 July 2013

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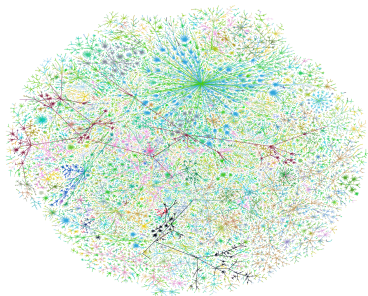
① Motivation

② Communities

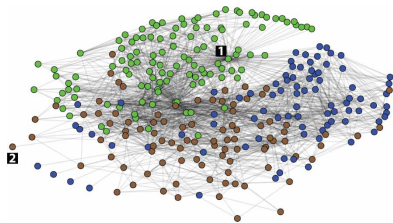
③ Experimental results

④ Conclusion

Networks are everywhere



(a) internet at level of AS [6]



(b) neural network [1]

Figure 1: Technological and biological networks.

Motivations for studying communities include. . .

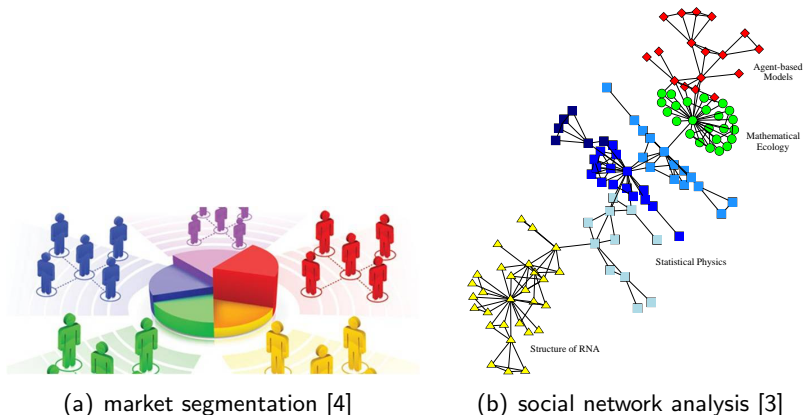
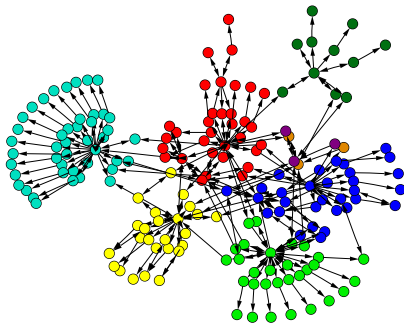


Figure 2: Communities in economic and social networks.

Motivations for studying communities include. . .



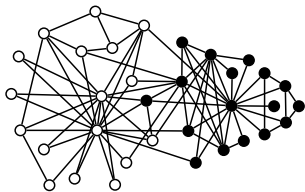
(a) organize computing clusters [2]



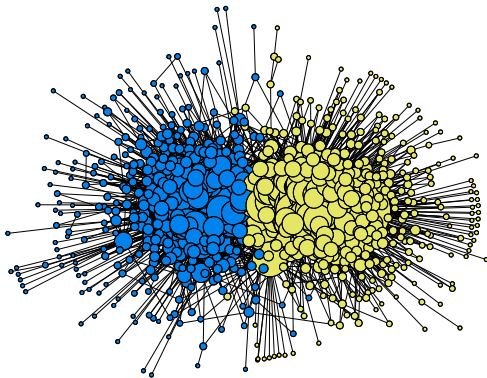
(b) analyze structure of WWW [7]

Figure 3: Communities in technological and information networks.

Communities are tightly inter-connected nodes



(a) karate club [8]



(b) political blogs [5]

Figure 4: Real-world networks with known community structures.

Hypotheses of community detection

- ➊ Better connection between nodes *within* a community.
- ➋ Few connection between nodes in *different* communities.

Research question

- How well inter-connected are nodes within a community?
- Define a compactness ratio that accounts for shortest paths between distinct pairs of nodes:

$$W^* = 1 - \frac{W(C)}{W(T)}$$

where

$$W(C) = \sum_{i < j} d_{ij}$$

and d_{ij} is the length of the shortest path between nodes i, j .

Results

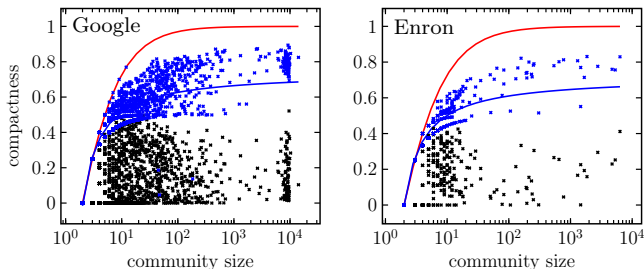


Figure 5: Compactness of communities in networks.

- black dot – compactness W^* of a community
- blue dot – ideal compactness $W_{K_n}^*$, i.e. when a community has all possible edges
- blue curve – model of the ideal compactness
- red curve – bound on both W^* and $W_{K_n}^*$

Shortcut links

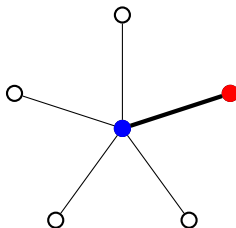


Figure 6: A node with many links (blue) connecting to a node with few links (red). The bold edge acts as a shortcut between the red node and other nodes.

Removing a shortcut edge is expected to reduce the compactness by a rate proportional to

$$-1/f_e^k$$

for some $k > 0$, where f_e is the fraction of edges removed.

Changes in compactness

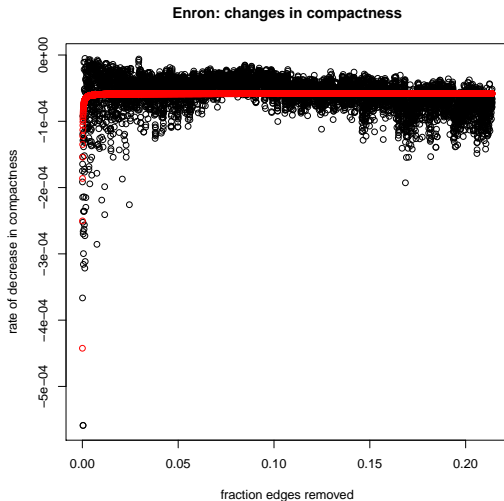


Figure 7: Largest community in the Enron email network. Change in compactness as a function of the fraction of edges removed.

Changes in compactness

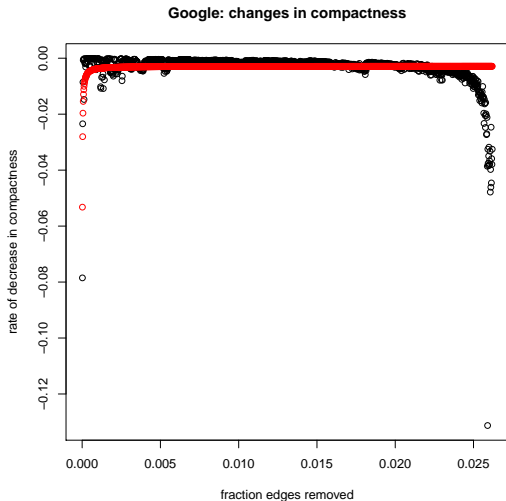


Figure 8: Largest community in the Google web graph. Change in compactness as a function of the fraction of edges removed.

Conclusion

- Address a fundamental hypothesis of community detection: Nodes within a community are better inter-connected among themselves.
- Compactness: Measure how well inter-connected are nodes in a community.
- Shortcut edges are important for increasing the inter-connectedness of nodes in a community.

Thank you



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