

Complex Networks IV

Plan of the lectures

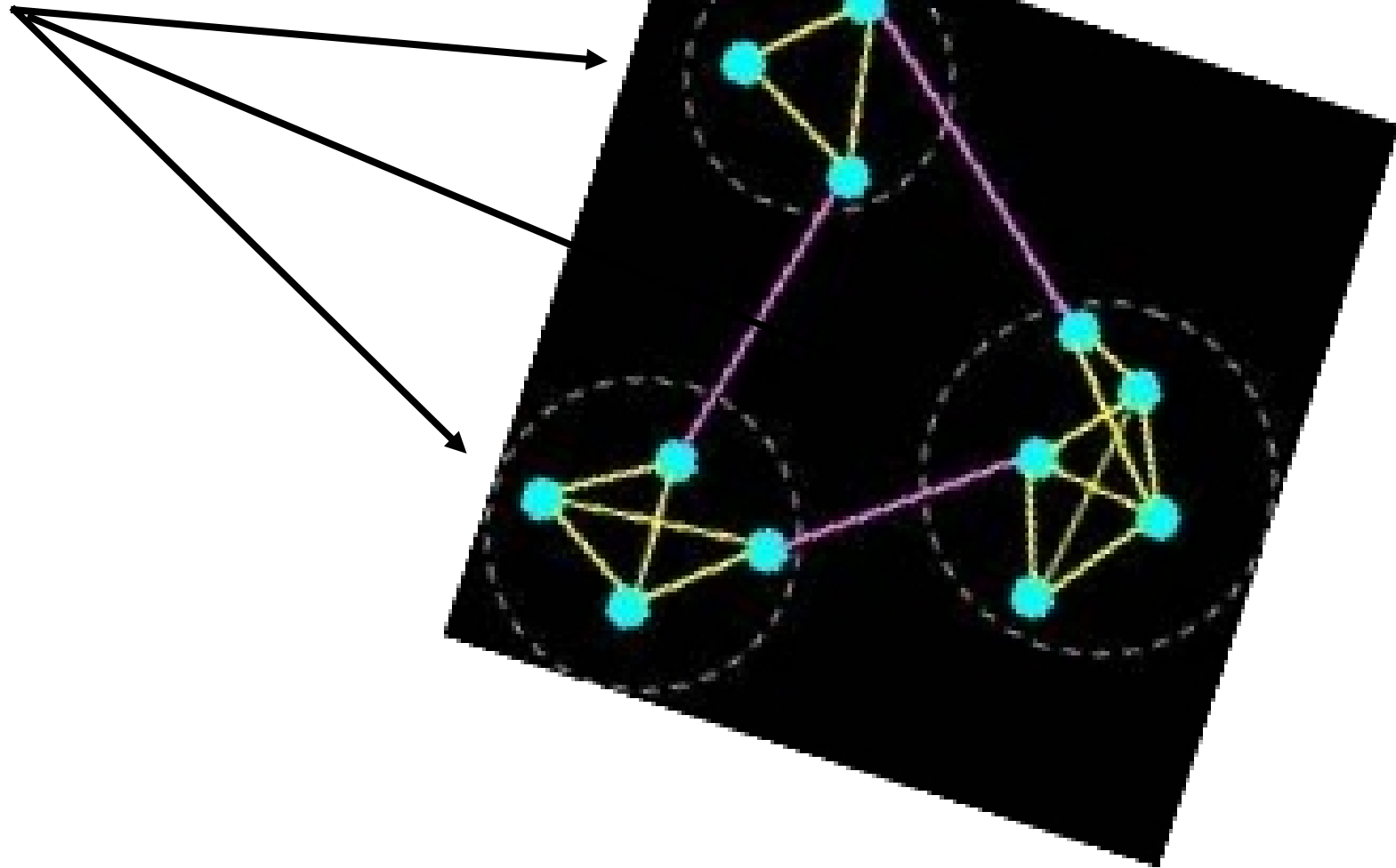
- I. Introduction
- II. Networks: basic definitions
- III. Models
- IV. **Community Detection**

Communities

Complex Networks typically show a high degree of modularity. These modules are called “communities”

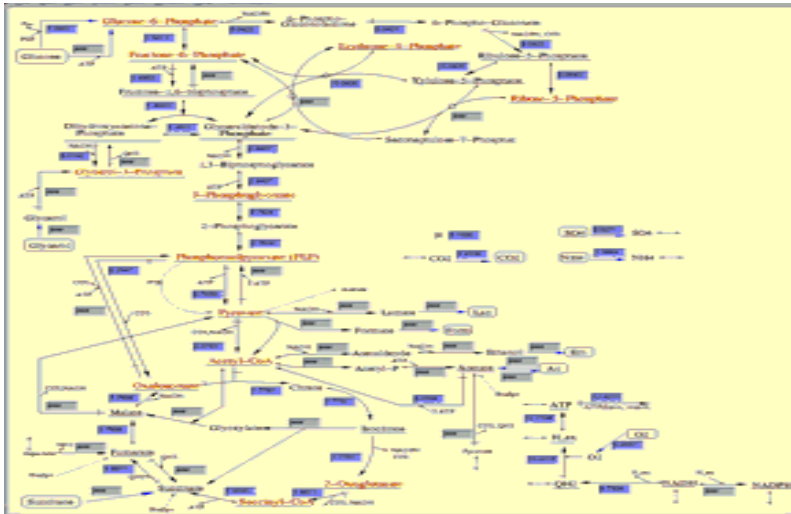
The goal of this lecture is to give a rigorous definition of modularity and a set of recipes to perform community detection.

“Community”



More link “within” a community than outside

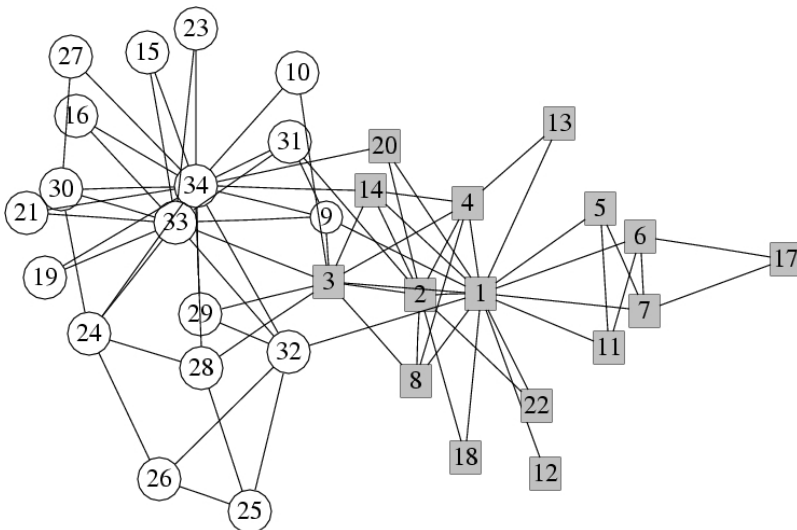
Metabolic



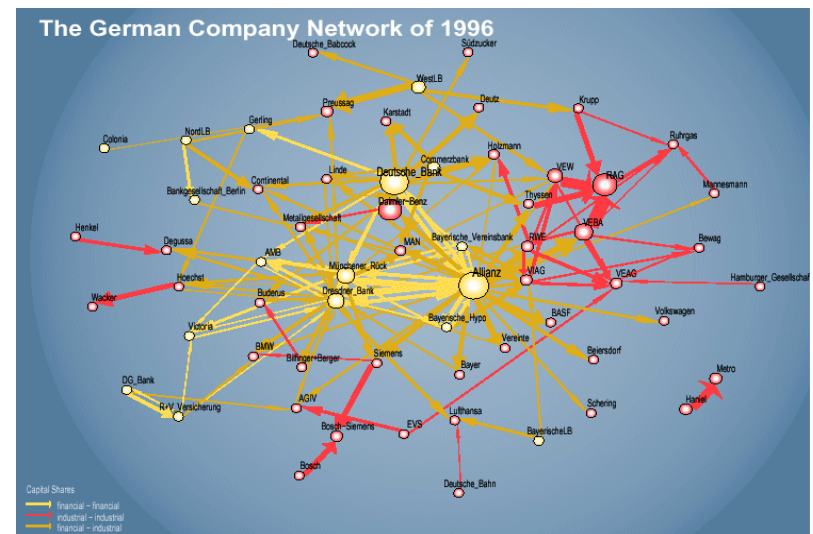
Protein-protein



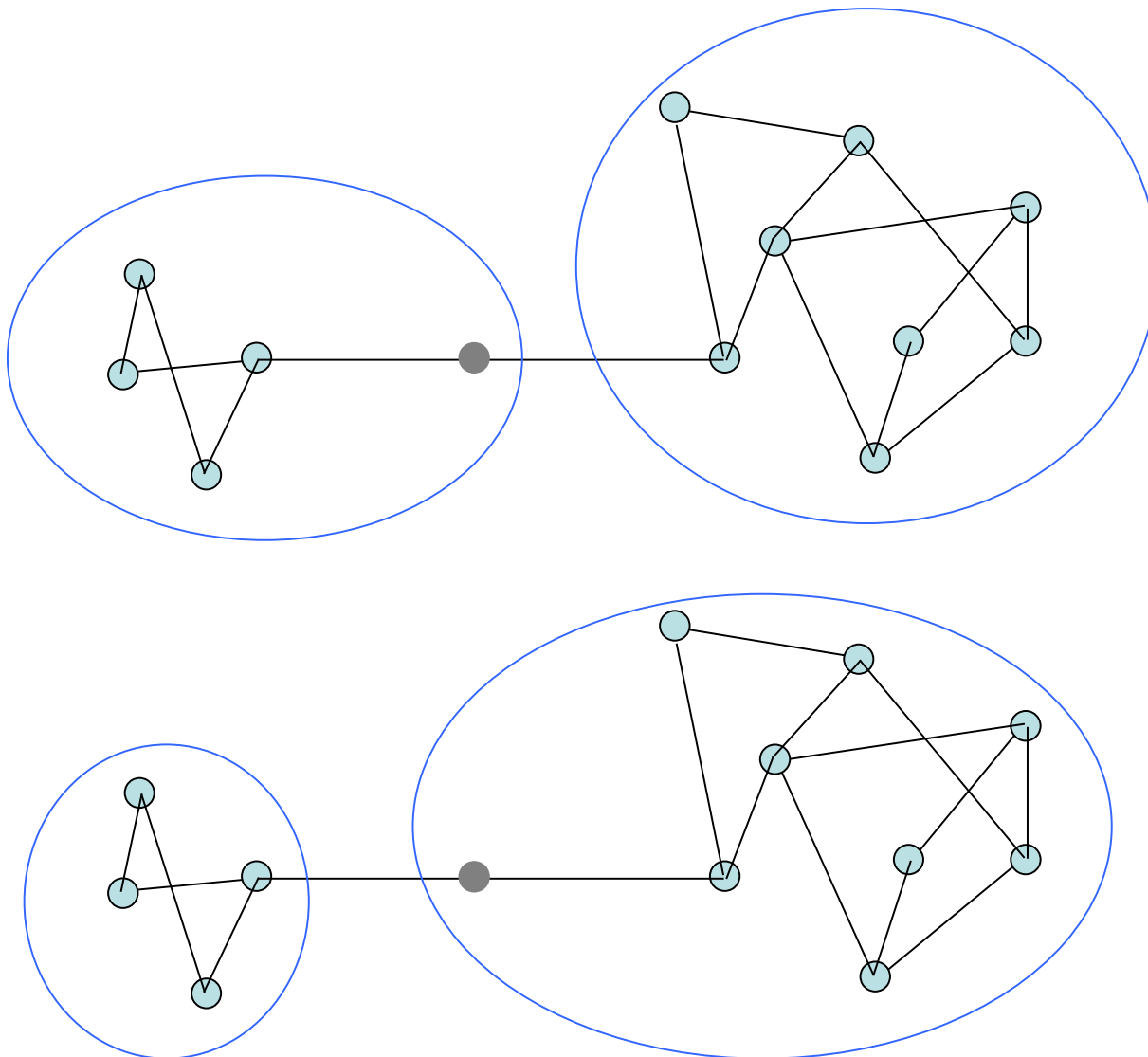
Social



Economical



How can we compare different partitions?



?

A few important remarks:

- 1) There are several different ways to perform community detection, **depending on the quantity one wants to optimize**
- 2) The problem is NP complete, there is no exact solution, only **heuristic approximations**
- 3) Finding communities is more and more difficult as the network becomes larger. One has to face a trade-off between efficiency and quality of the community organization.

Modularity

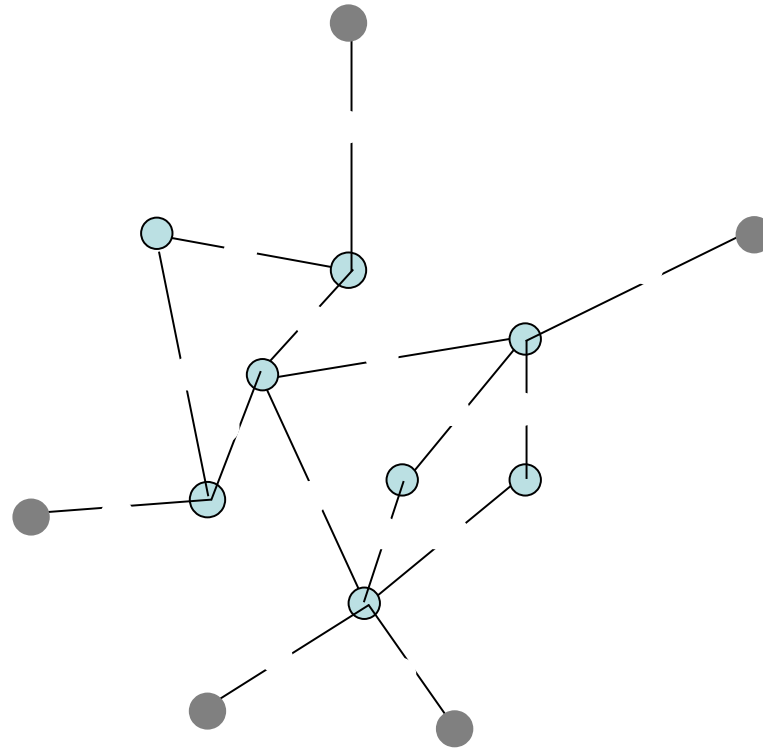
$$Q = \frac{1}{L} \sum_{i=1}^n \left(l_i - \frac{d_i^2}{4L} \right)$$

n = num. of communities according to a given partition

L = total number of links, d_i = number of nodes in module i

l_i = # of links in module i

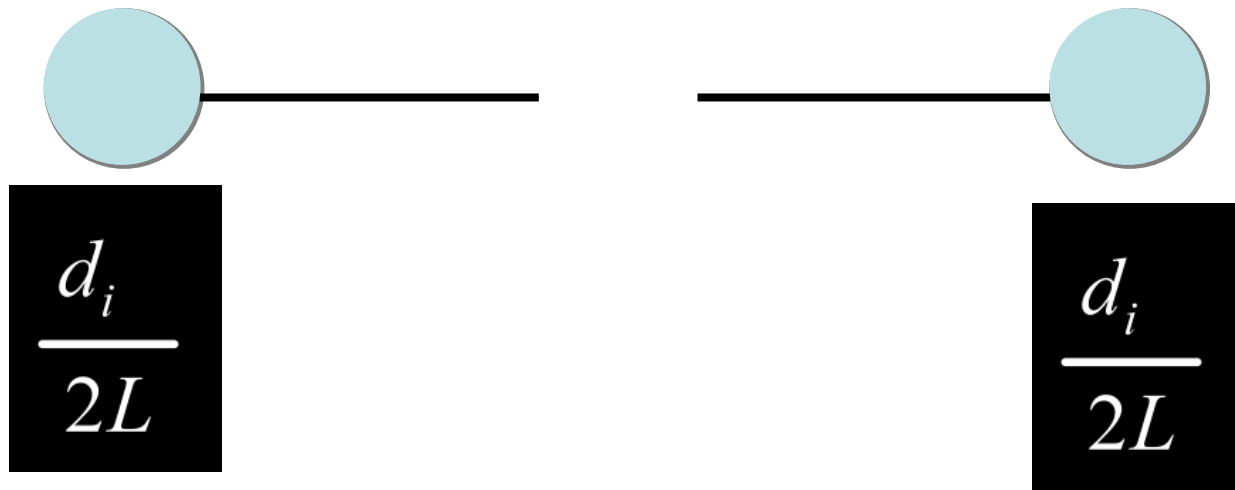
$\frac{d_i^2}{4L}$ = expected number of links in module i given the number of nodes.



$$\frac{d_i}{2L}$$

=

Probability that a link chosen at random ends in module i.



$$\frac{d_i}{2L} \cdot \frac{d_i}{2L} = \frac{d_i^2}{4L^2}$$

Probability that the link
is inside module i

$$\frac{d_i^2}{4L^2} \cdot L = \frac{d_i^2}{4L}$$

Expected number of links in
module i

Girvan-Newman algorithm

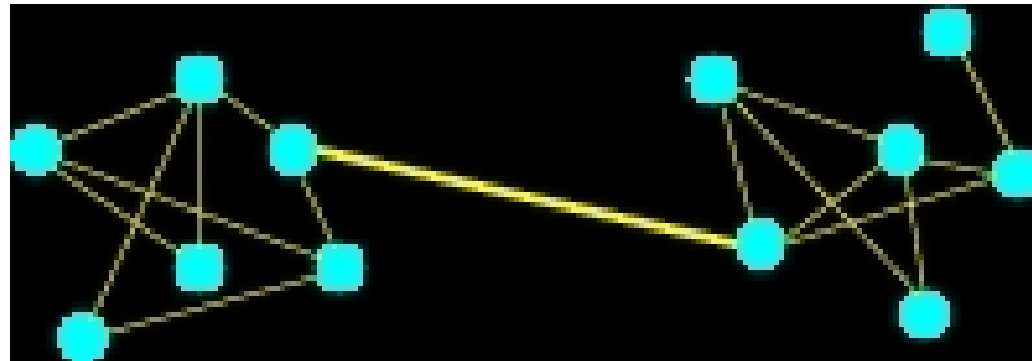
M. Girvan & M.E.J Newman,
PNAS 99, 7821-7826 (2002)

Method: cut the link which connect different communities
so as to isolate them.

How can we identify these links?

Betweenness!!

Link-betweenness: number of paths which must go through a given link



Iterative Algorithm:

1. Evaluate the link betweenness for each link
2. Eliminate the link with highest betweenness
3. Evaluate the link betweenness of all remaining links
4. GO TO 2

This algorithm gives a hierarchy of partitions,
which is the right one?

The best choice is the one with
highest modularity Q

M.E.J. Newman & M. Girvan, Phys. Rev. E
69, 026113 (2004)

General tools for network analysis

igraph

is open source and can be programmed in R, Python, Mathematica and C++ (also contains seven different community detection algorithms)
<https://igraph.org/>

Stanford Network Analysis Project (SNAP)

written in C++ with a Python interface
<http://snap.stanford.edu/>

NETWORKX

<https://networkx.org/documentation/stable/reference/introduction.html>

SCIKIT-NETWORK

<https://scikit-network.readthedocs.io/en/latest/>

Repositories

Networkrepository

Thousands of networks on different contexts (Bio-socio-economics...)

<https://networkrepository.com/index.php>

Networks

hundreds of benchmark networks

<https://networks.skewed.de/about>

Community detection tools

OSLOM

<http://www.oslom.org/>

INFOMAP

<https://www.mapequation.org/infomap/>

LOUVAIN

https://en.wikipedia.org/wiki/Louvain_method

<https://scikit-network.readthedocs.io/en/latest/tutorials/clustering/loouvain.html>

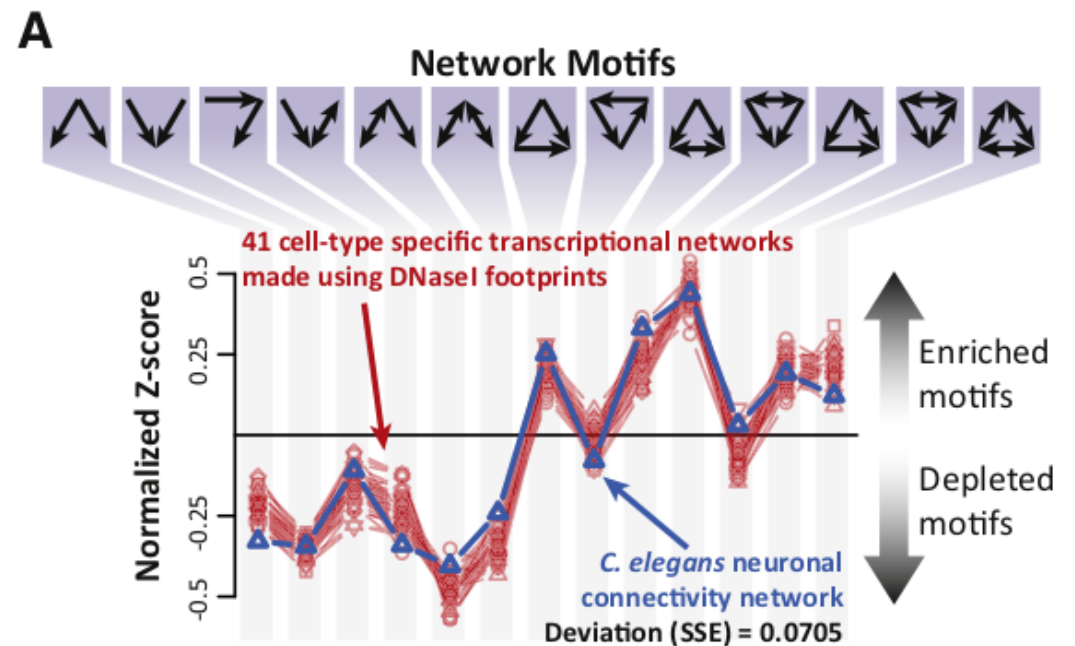
Motif detection tools

MFINDER

<https://www.weizmann.ac.il/mcb/UriAlon/download/network-motif-software>

Allows to estimate z-score with a reshuffling null model

$$z = \frac{x - \mu}{\sigma}$$



Useful References

- 1) M. Girvan; M. E. J. Newman (2002). "Community structure in social and biological networks". Proc. Natl. Acad. Sci. USA. 99 (12): 7821–7826. arXiv:cond-mat/0112110.
- 2) S. Fortunato (2010). "Community detection in graphs". Phys. Rep. 486 (3–5): 75–174. arXiv:0906.0612.
- 3) Peixoto, Tiago (2014) "Hierarchical Block Structures and High-Resolution Model Selection in Large Networks". Physical Review X. 4 (1): 011047. arXiv:1310.4377
- 4) Martin Rosvall; Carl T. Bergstrom (2007). "An information-theoretic framework for resolving community structure in complex networks". PNAS 104 (18): 7327–7331. arXiv:physics/0612035.
- 5) Milo et al. (2002) "Network Motifs: Simple Building Blocks of Complex Networks" Science 298, 824.