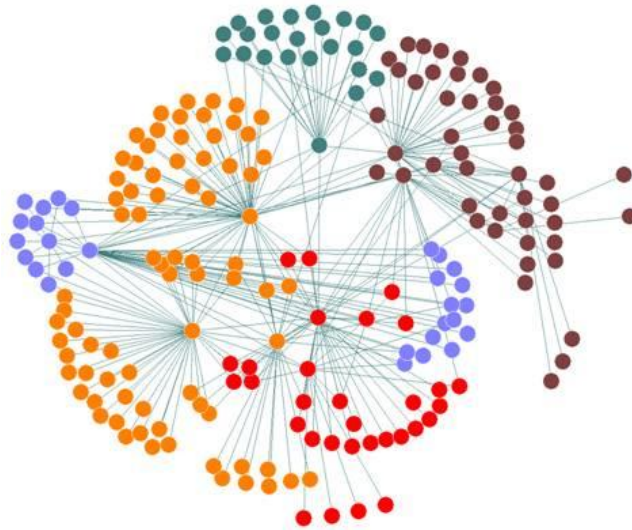




Algorithms and Applications in Social Networks



2023/2024, Semester A

Slava Novgorodov

Lesson #4

- Similarity in networks
- Network structure
- Communities
- Detection of communities

Similarity in Networks

Properties of Network

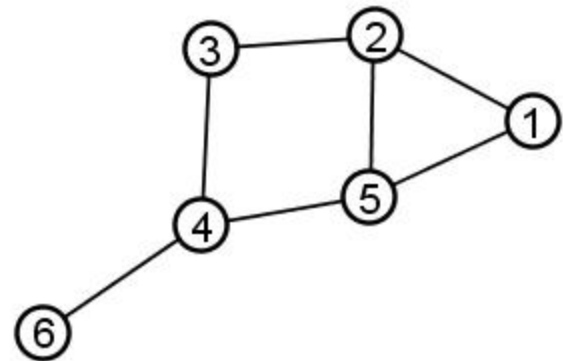
- Global (per network):
 - Average clustering coefficient
 - Degree distribution
 - Diameter
 - ...
- Local (per node):
 - Degree
 - Node centrality
 - ...
- Pairwise:
 - Similarity
 - ...

Similarity

- Jaccard Similarity: $[N(v) = \text{Neighbors}(v)]$

$$J(a, b) = \frac{|N(a) \cap N(b)|}{|N(a) \cup N(b)|}$$

$$J(2, 4) = \frac{|\{3, 5, 6\} \cap \{1, 3, 5\}|}{|\{3, 5, 6\} \cup \{1, 3, 5\}|} = \frac{2}{4} = 0.5$$



Similarity

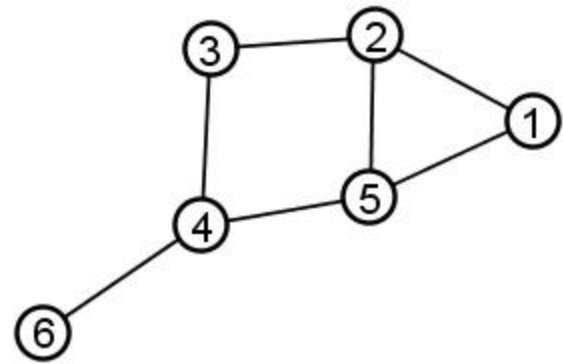
- Cosine Similarity:

$$\cos(a,b) = (a * b) / (|a| * |b|)$$

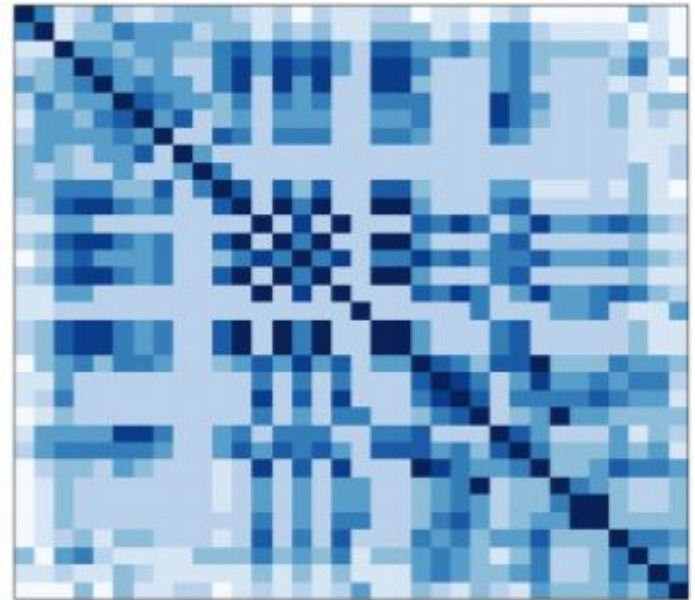
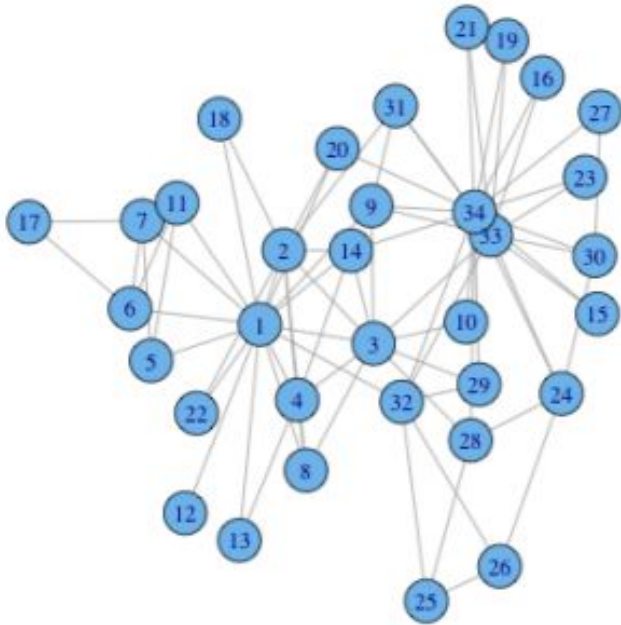
$$(2) = (1, 0, 1, 0, 1, 0)$$

$$(4) = (0, 0, 1, 0, 1, 1)$$

$$\cos((2), (4)) = 2 / 3$$



Similarity



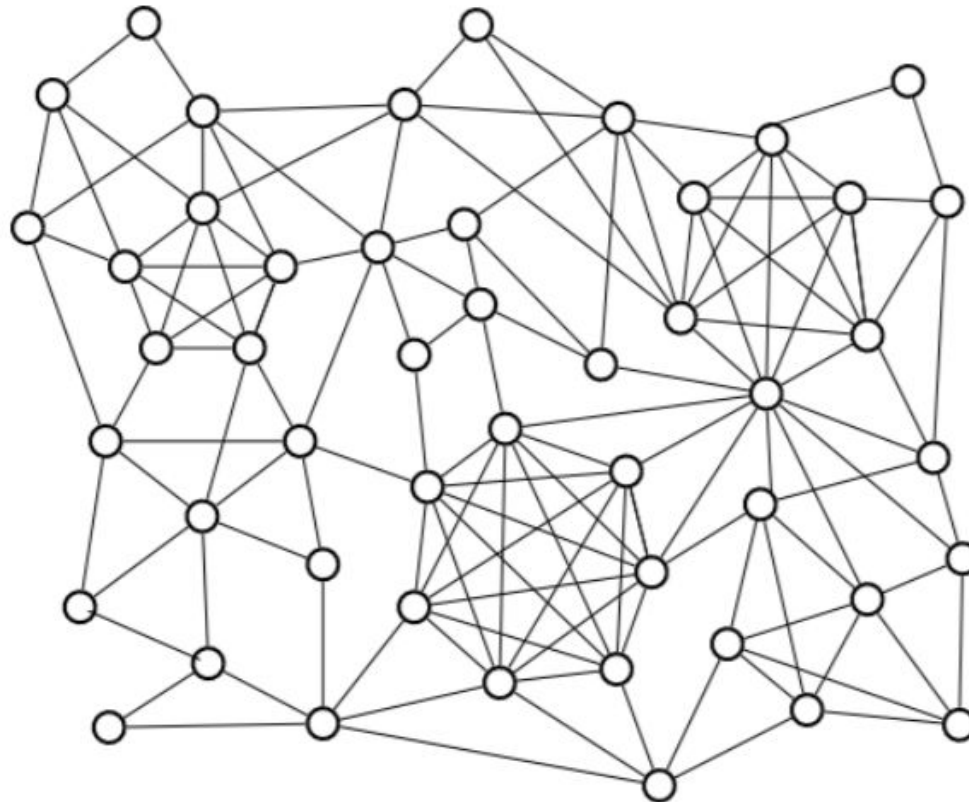
Network Structure

Network Structure



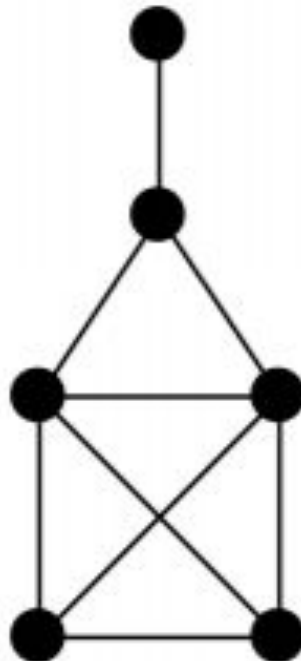
Graph cliques

- Clique is a complete subgraph.
- Cliques can overlap



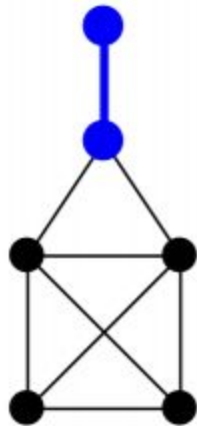
Graph cliques

- **Maximal clique** – a clique that cannot be extended by adding more nodes
- **Maximum clique** – a clique of a maximum size



Graph cliques

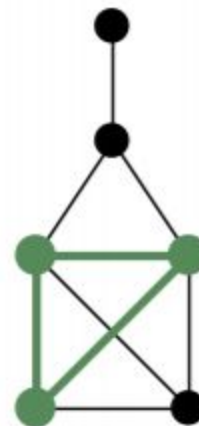
- **Maximal clique** – a clique that cannot be extended by adding more nodes
- **Maximum clique** – a clique of a maximum size



Maximal



Maximal
& Maximum



Not maximal



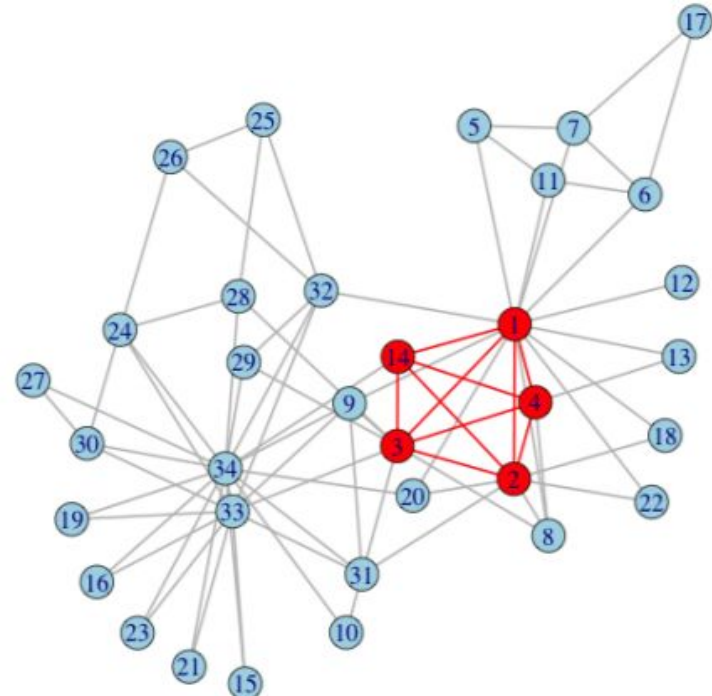
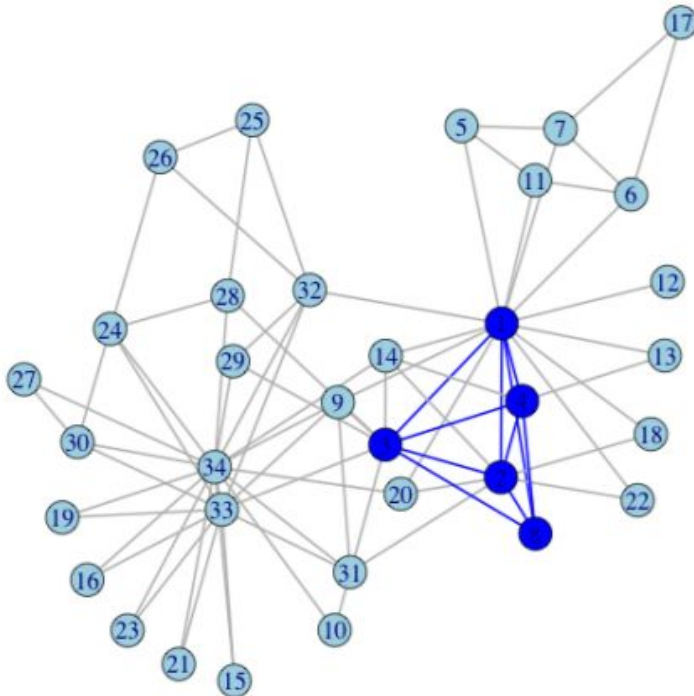
Not clique

Graph cliques

Maximal cliques:

Size: 2 3 4 5

#: 11 21 2 2



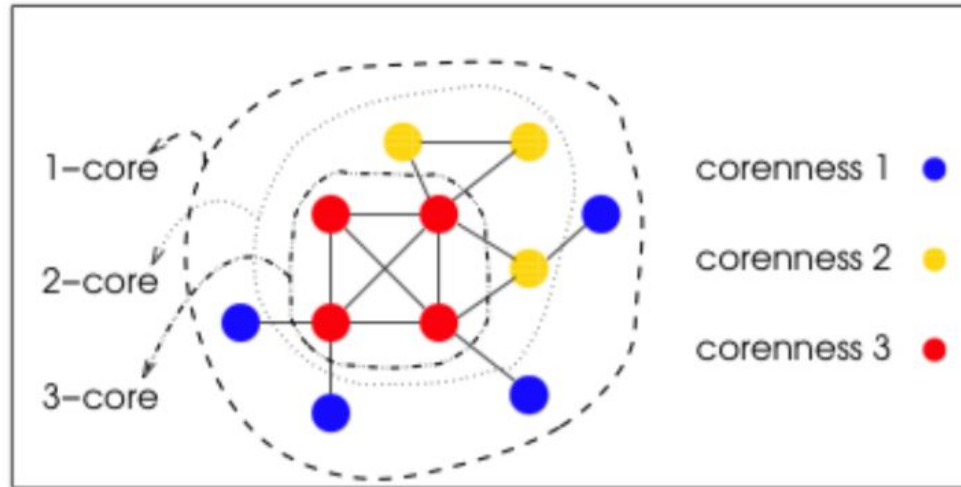
Graph cliques

Computation of cliques:

- Finding a clique of size k : $O(n^k k^2)$
- Finding maximum clique: $O(3^{n/3})$
- Easier in sparse graphs...

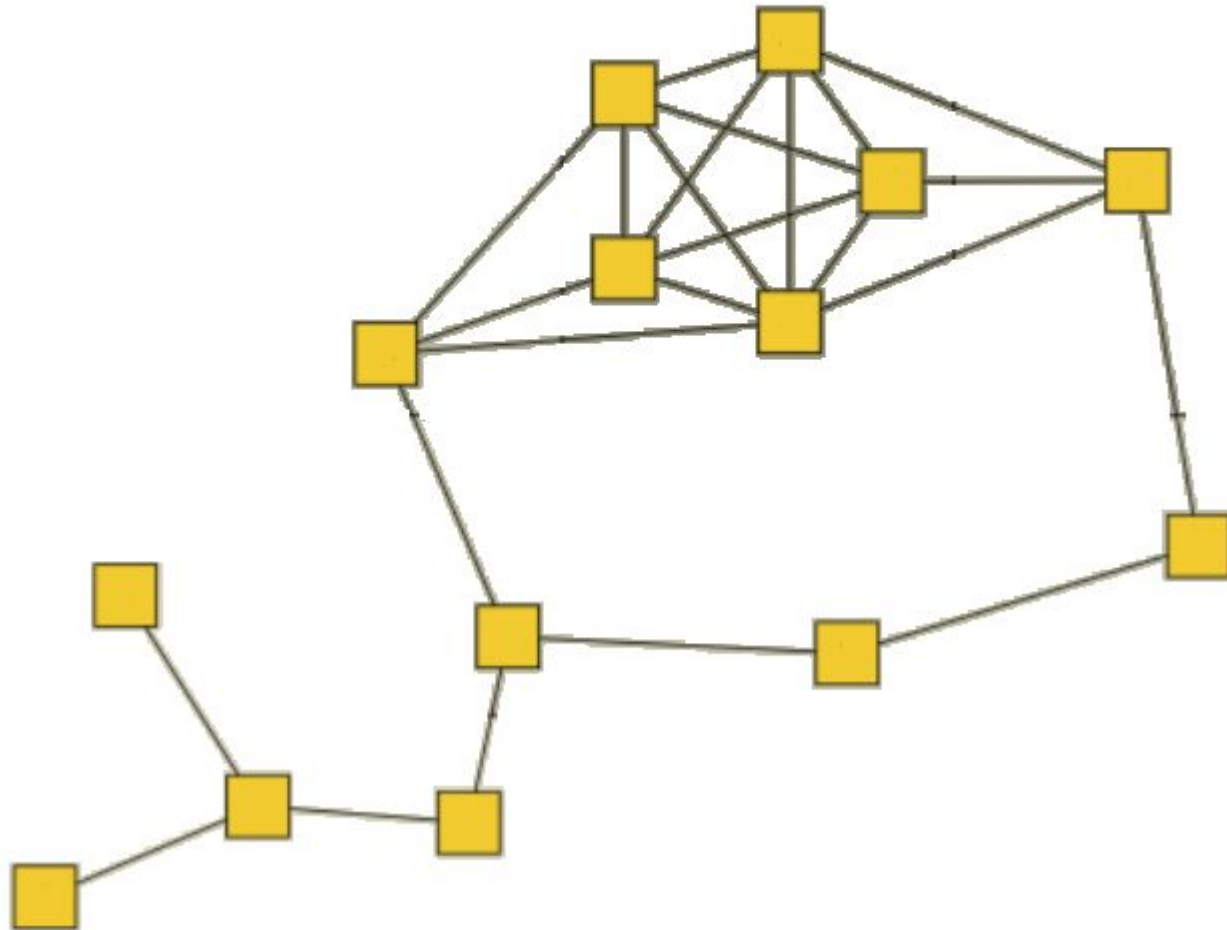
Graph Core

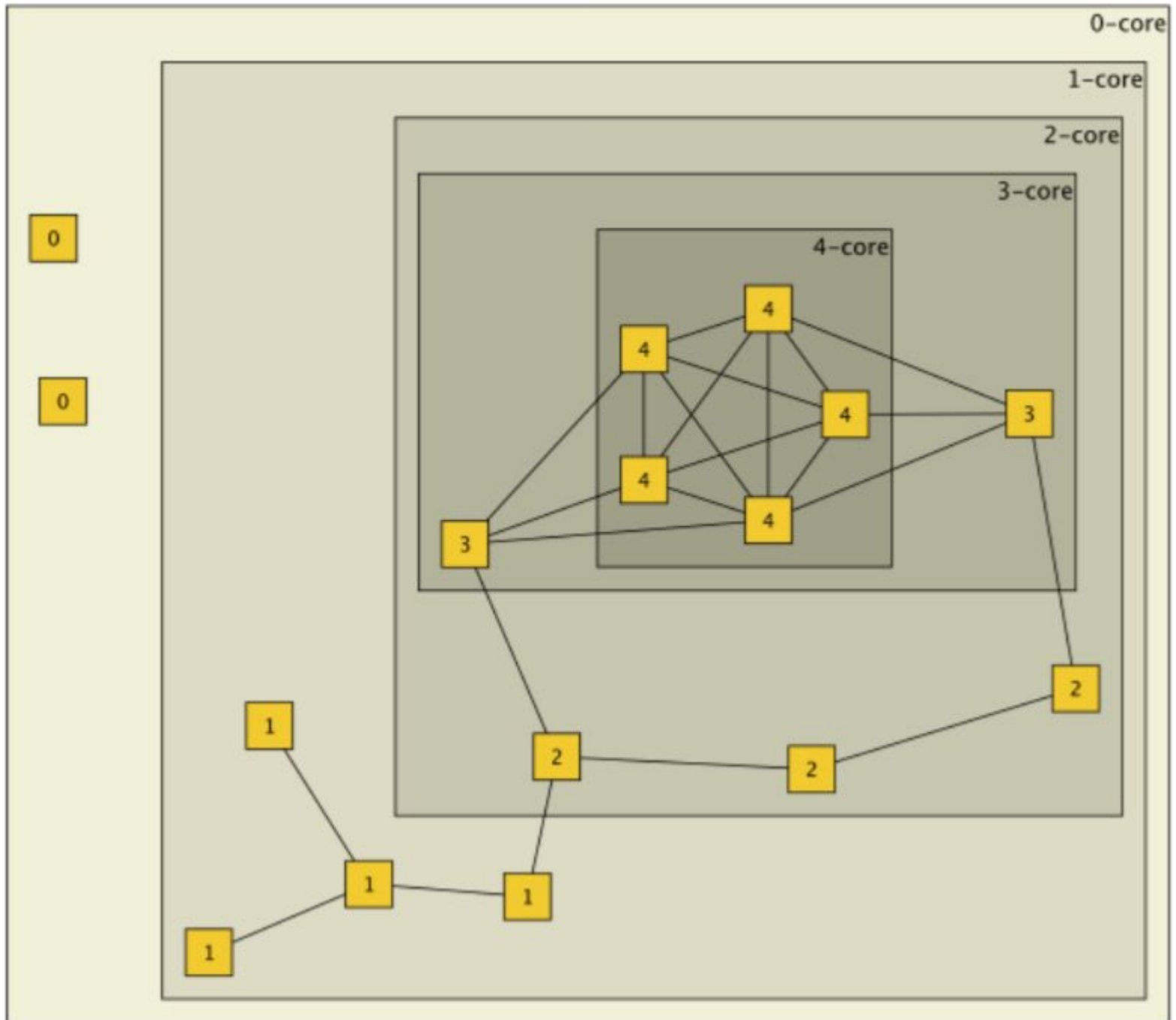
- A **k-core** is the largest subgraph S such as each node is connected to at least k nodes in S



- Every node in k -core has degree $\geq k$
- $(k+1)$ -core is always a subgraph of k -core
- Core number of node is the highest “ k ” of the k -core that contains this node

Graph Core - Example





k-core decomposition

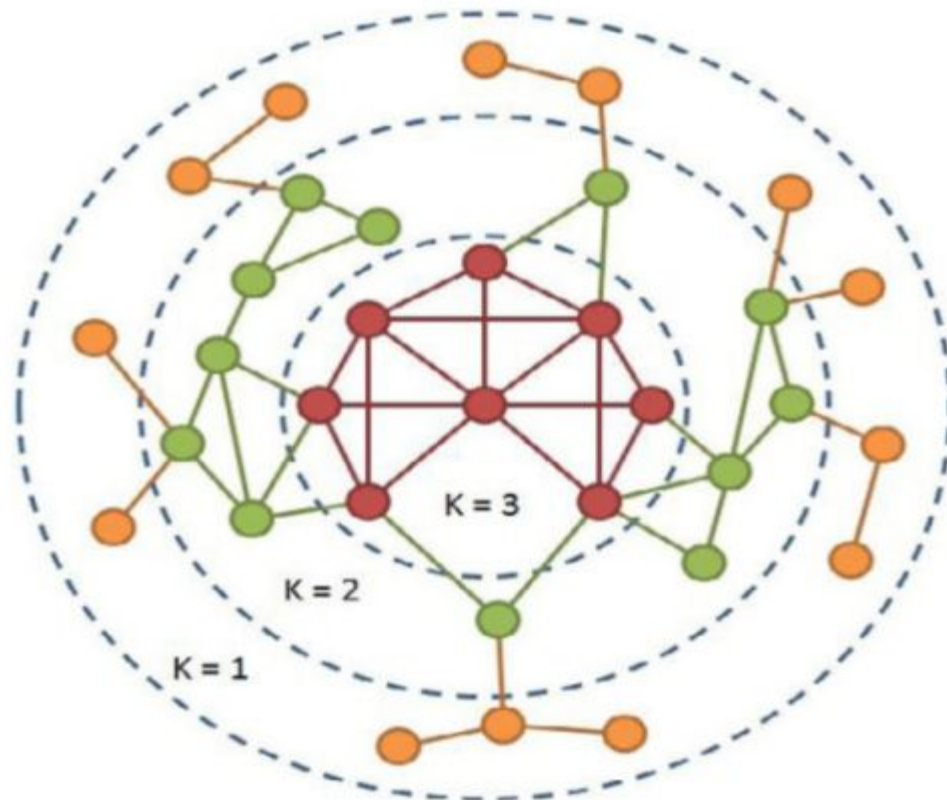
Input: Graph $G(V, E)$

Output: $\text{core}[v]$ – core number for each node

- 1: Store degree of v in $\text{degree}[v]$
- 2: Sort V by $\text{degree}[v]$ (ascending)
- 3: **For each** v **in** V :
- 4: $\text{core}[v] = \text{degree}[v]$
- 5: **For each** u **in** $\text{neighbors}(v)$:
- 6: **If** $\text{degree}[u] > \text{degree}[v]$:
- 7: $\text{degree}[u] -= 1$
- 8: Sort V by $\text{degree}[v]$

k-core decomposition

Recursively delete all nodes with degree less than k ,
the remaining graph is k -core



k-core decomposition

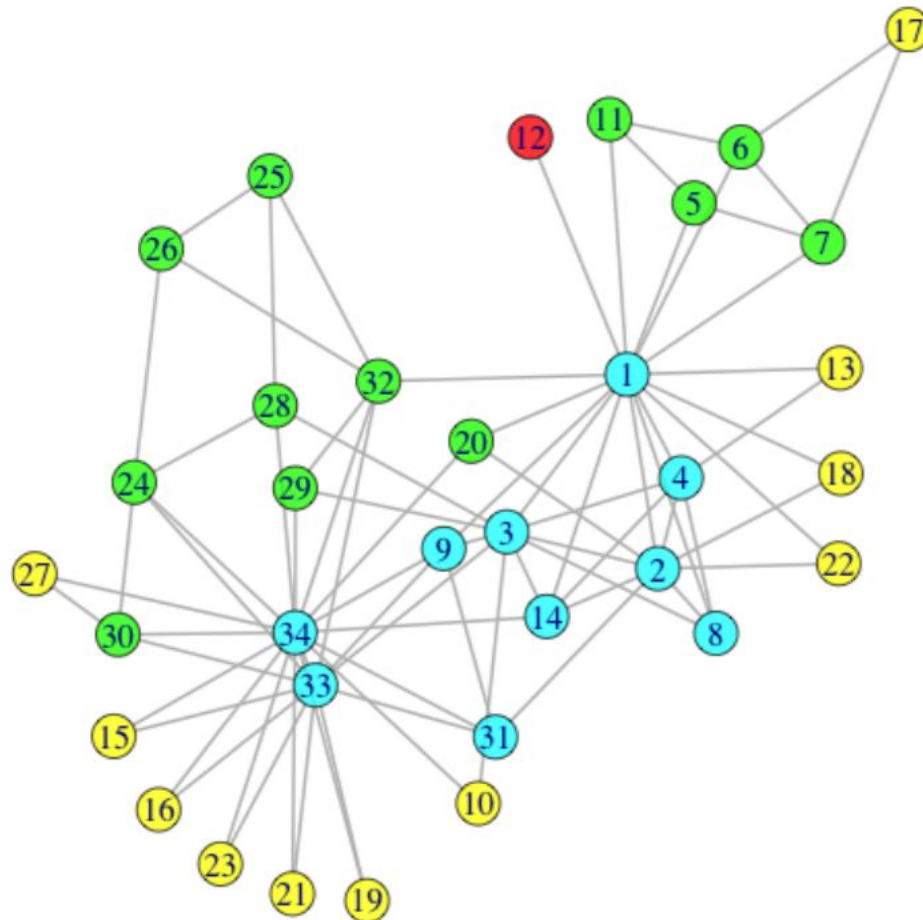
Zachary karate club k-core decomposition:

Red – 1

Yellow – 2

Green – 3

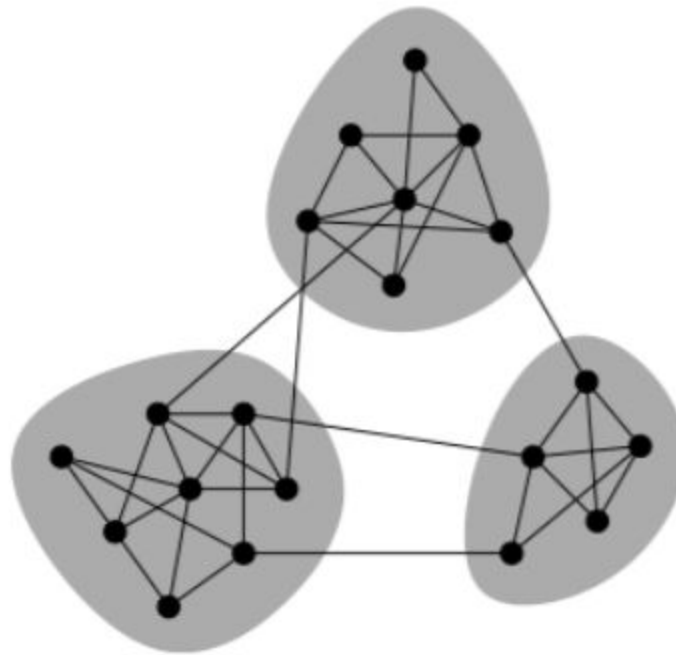
Blue – 4



Network Communities

Community

Network Communities are group of vertices such that vertices inside the group connected with many more edges than between groups

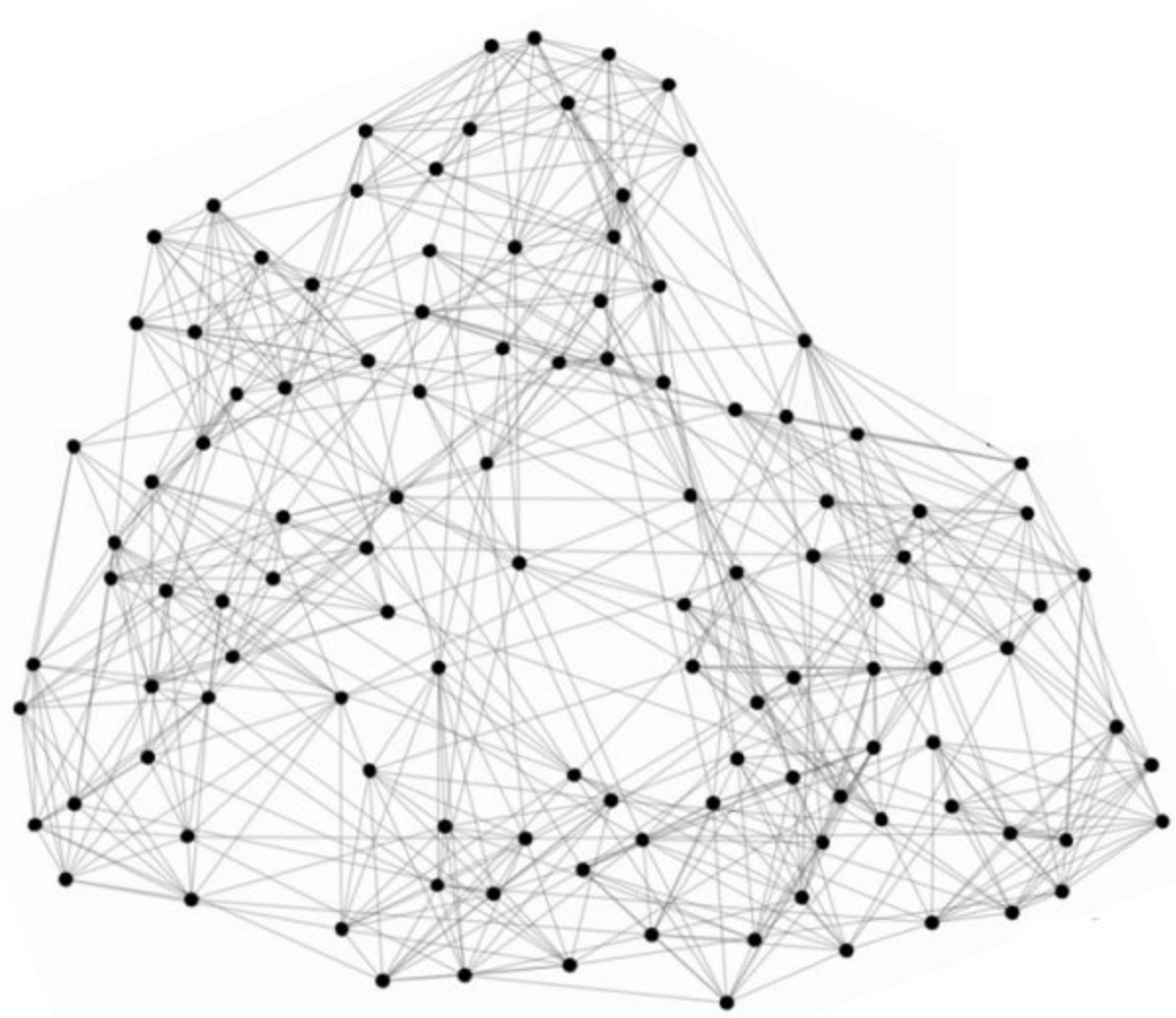


Community Types

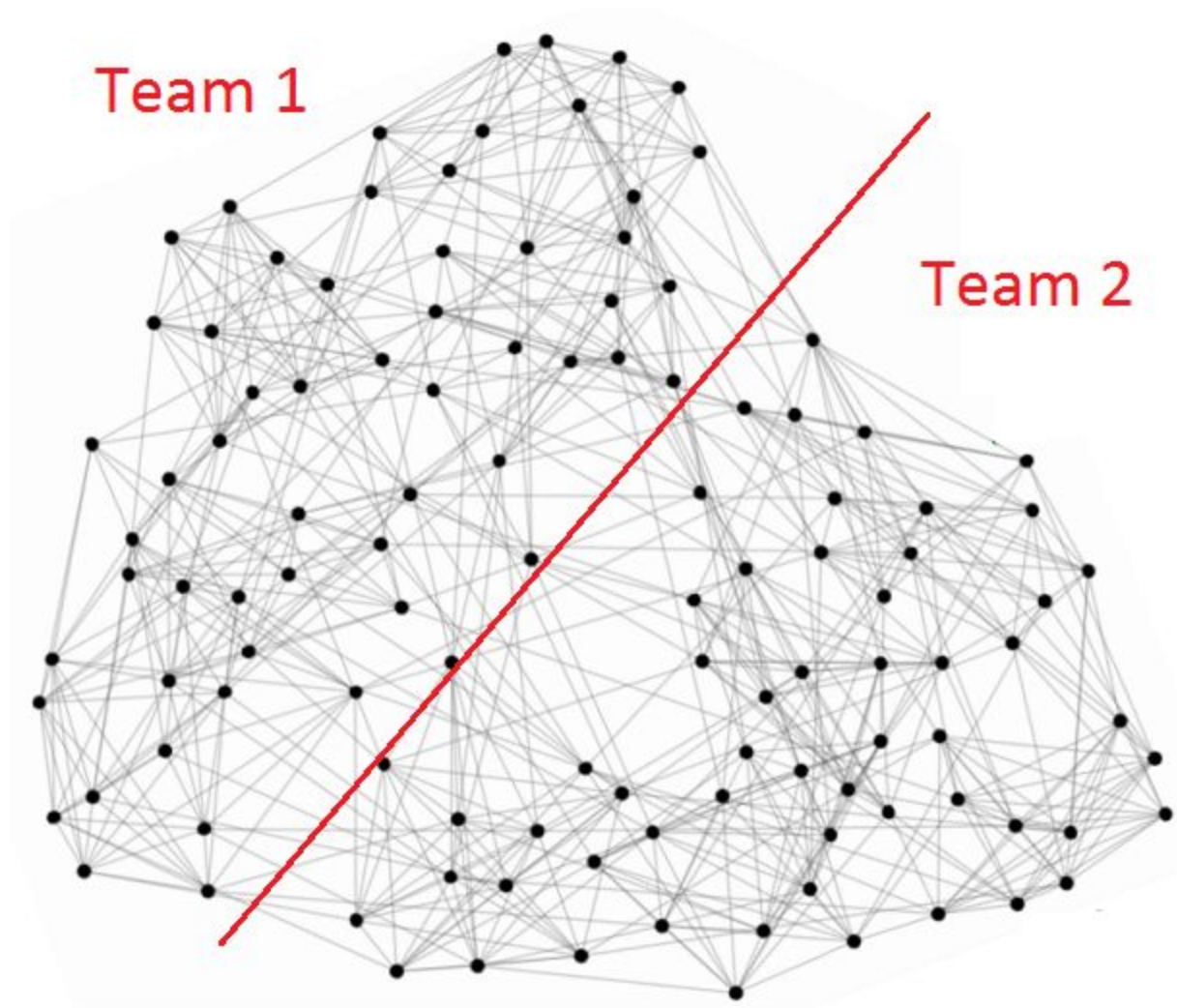
Communities can be either:

- Non-Overlapping
 - Country of residence
 - Working place
 - Favorite football club
- Overlapping
 - Types of friends (friends from school, army, work)
 - Groups of interest

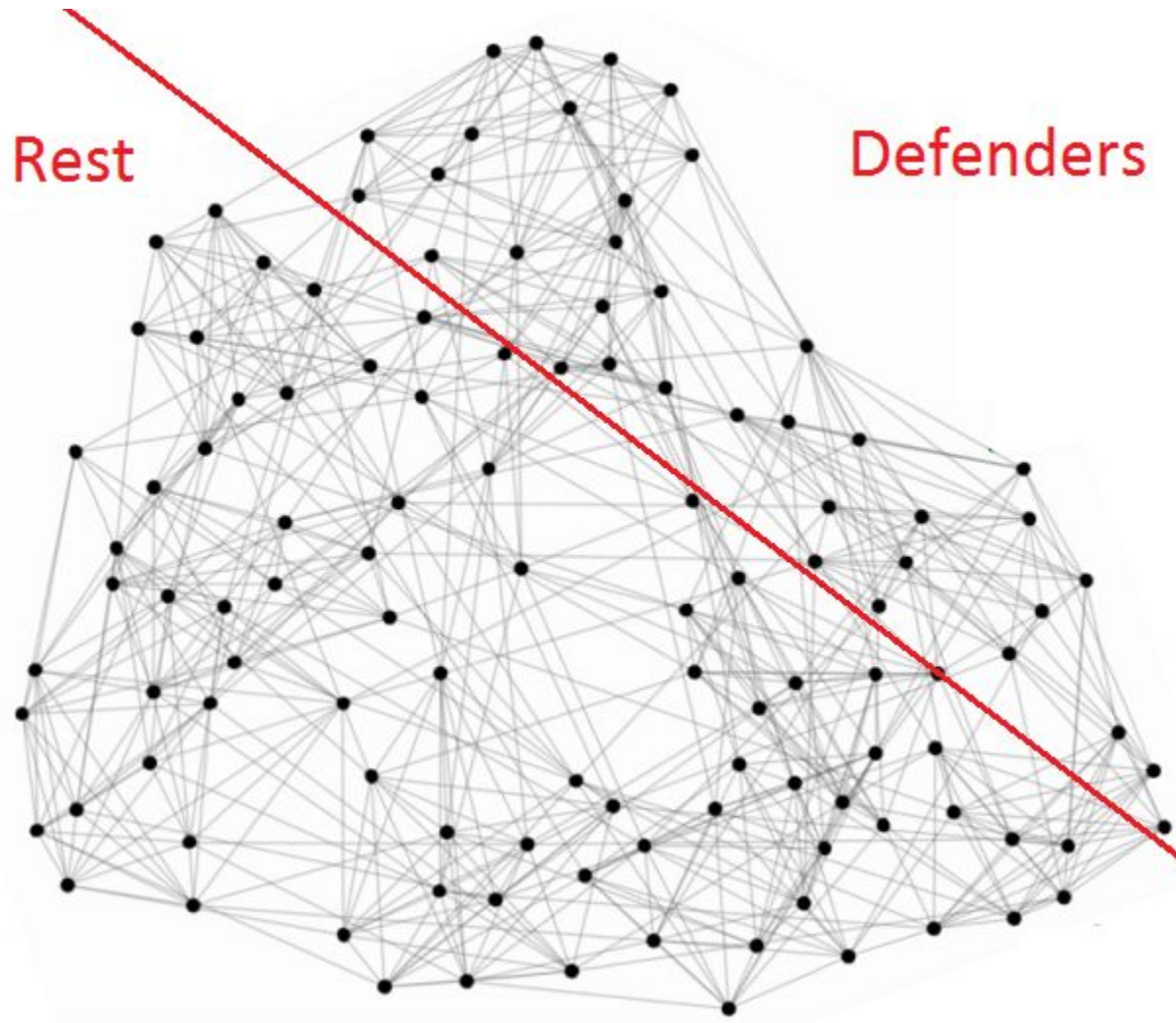
Example – Soccer teams



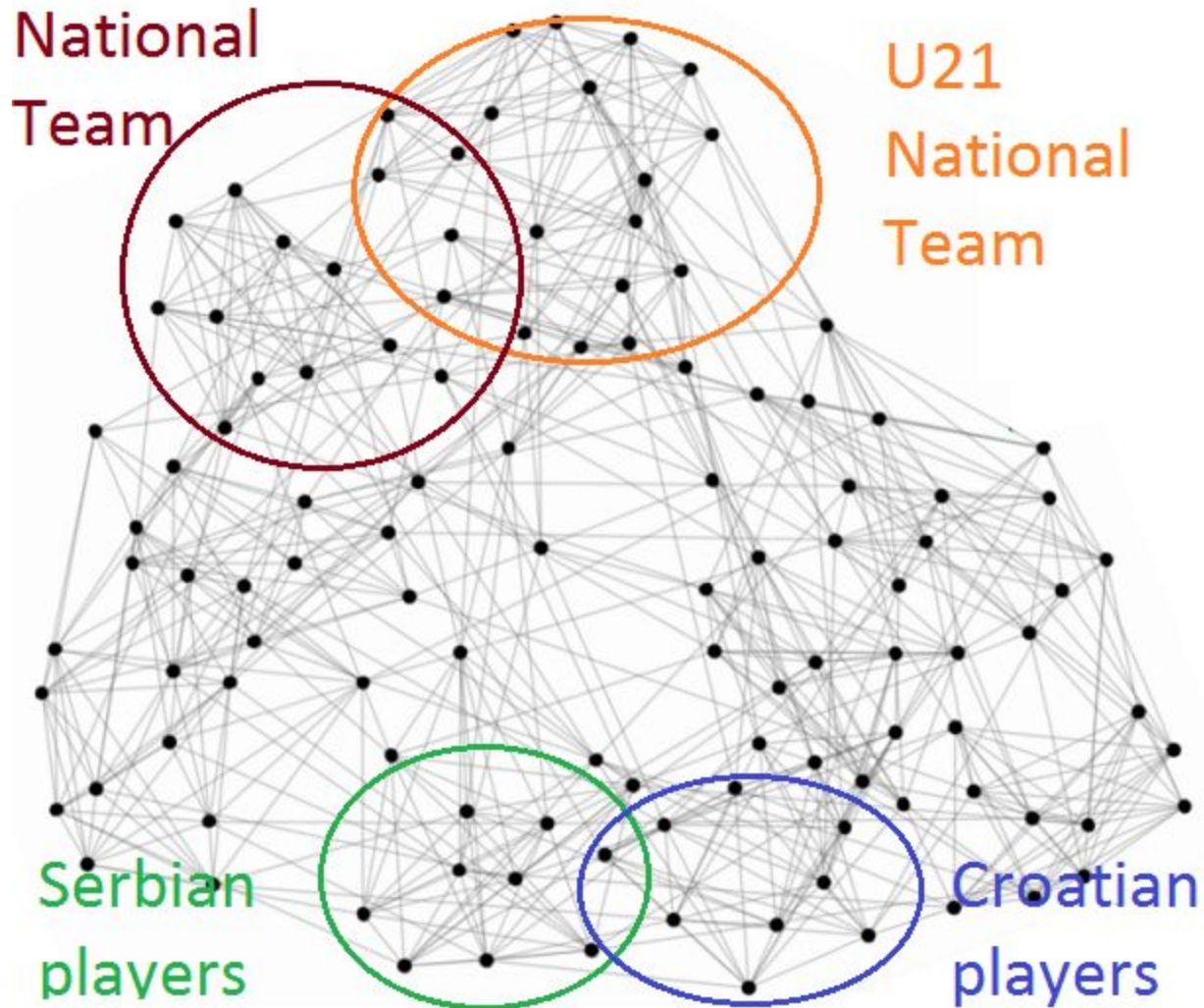
Example – Non-overlapping



Example – Non-overlapping



Example – Overlapping



Example



Example



© Twitter:themichaelowen

What makes community?

- Mutuality of edges.
 - Almost everyone in the group has edge to one another
- Compactness.
 - Closeness or reachability of group members in small number of steps, not necessarily adjacency
- Density of edges.
 - High frequency of edges within the group
- Separation.
 - Higher frequency of ties among group members compared to non-members

Community density

- Network density: $\rho = \frac{m}{n(n-1)/2}$
- Inter-community density: $\delta_{int}(C) = \frac{m_s}{n_s(n_s-1)/2}$
- External edges density: $\delta_{ext}(C) = \frac{m_{ext}}{n_c(n-n_c)}$

Community has: $\delta_{int} > \rho, \delta_{ext} < \rho$

n_s – nodes in S , m_s – edges in S , m_{ext} – edges from S outside

Modularity

Modularity:

$$Q = \frac{1}{2m} \sum_{ij} \left(A_{ij} - \frac{k_i k_j}{2m} \right) \delta(c_i, c_j), = \sum_u (e_{uu} - a_u^2)$$

e_{uu} - fraction of edges within community u

$a_u = \sum_v e_{uv}$ - fraction of ends of edges attached to nodes in u

$\delta_{ij} = \begin{cases} 0 & \text{if } i \neq j, \\ 1 & \text{if } i = j. \end{cases}$ - Kronecker Delta

Higher modularity score – better community

Single community – $Q = 0$

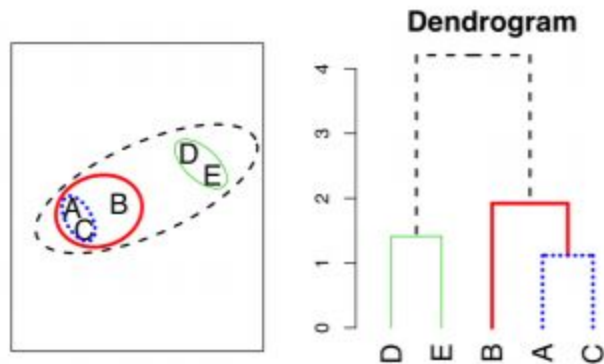
Detection of Communities

Similarity Based Clustering

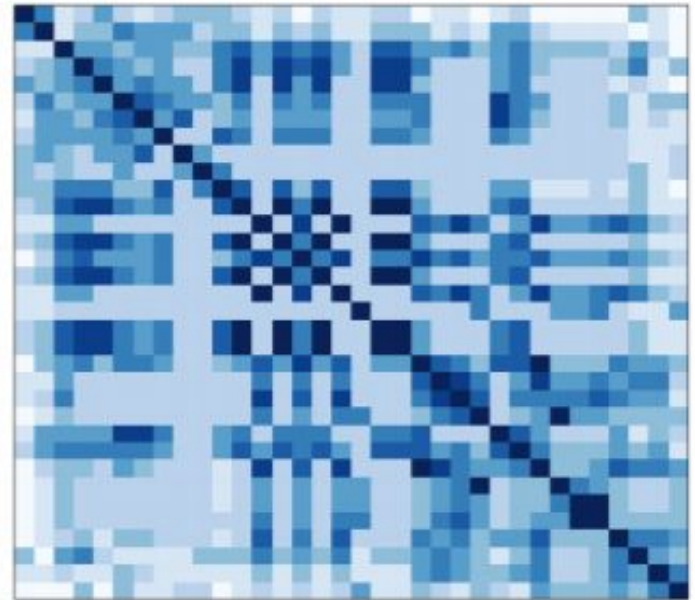
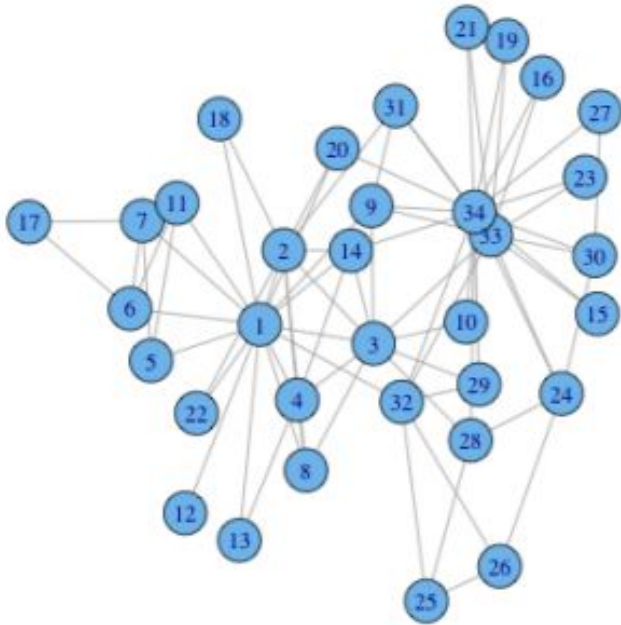
- Define similarity between every two nodes:
 - Jaccard
 - Cosine
 - ...
- Build similarity matrix (compute all-pairs similarity)
- Group together nodes with high similarity

Agglomerative clustering

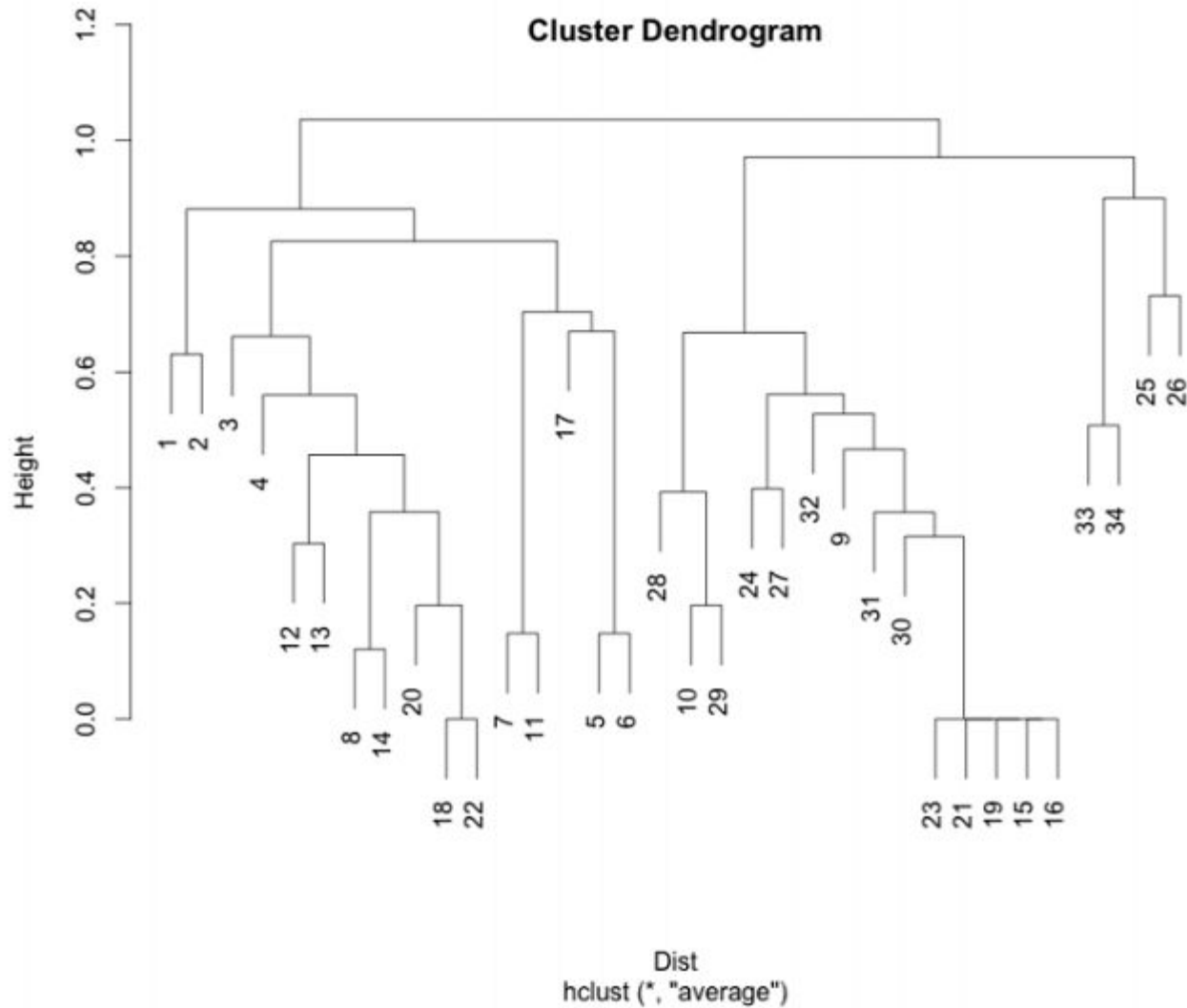
- Assign each vertex to a group of its own
- Find two groups with the highest similarity and join them in a single group
- Calculate similarity between groups:
 - single-linkage clustering (most similar in the group)
 - complete-linkage clustering (least similar in the group)
 - average-linkage clustering (mean similarity between groups)
- Repeat until all joined into single group



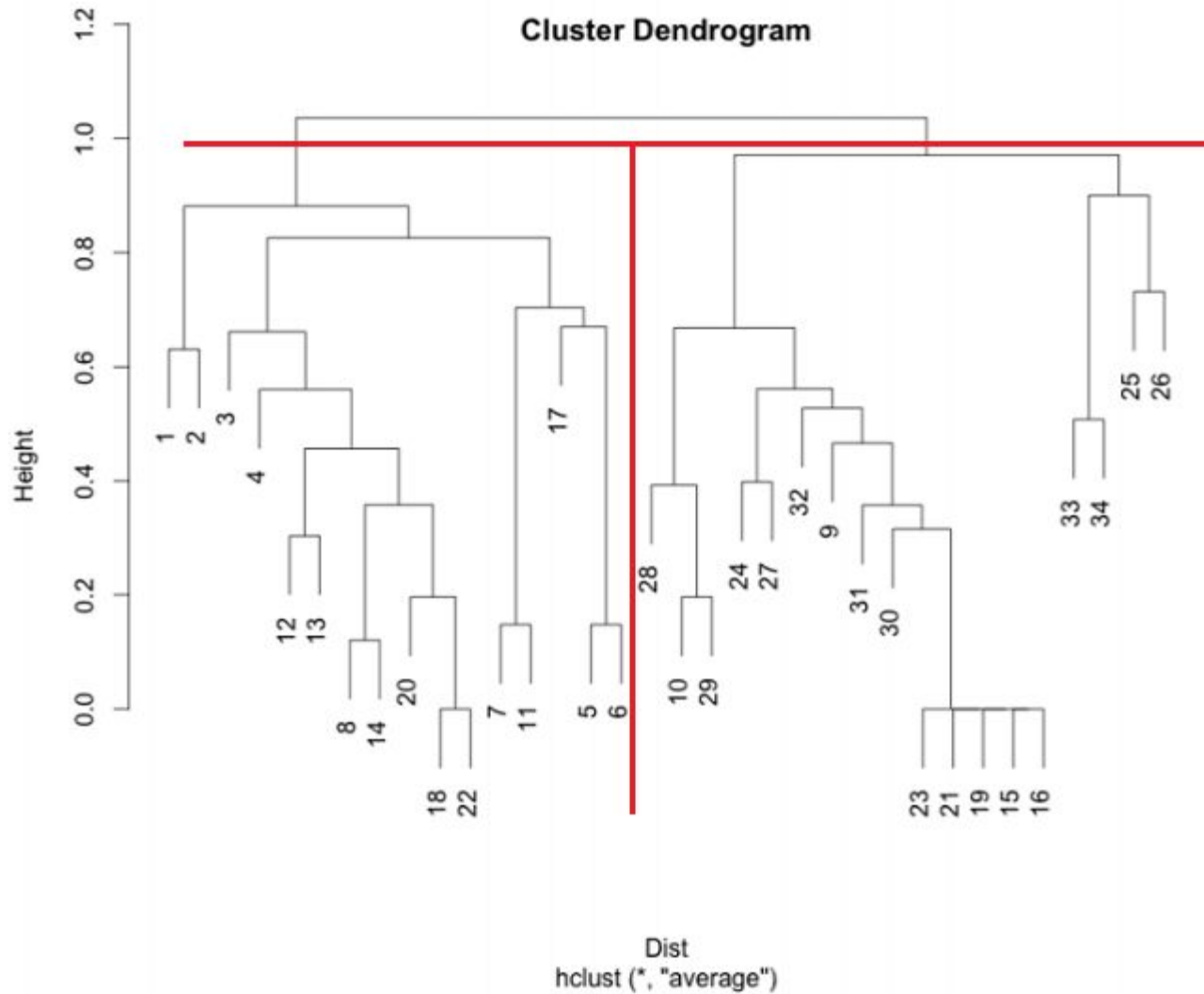
Similarity



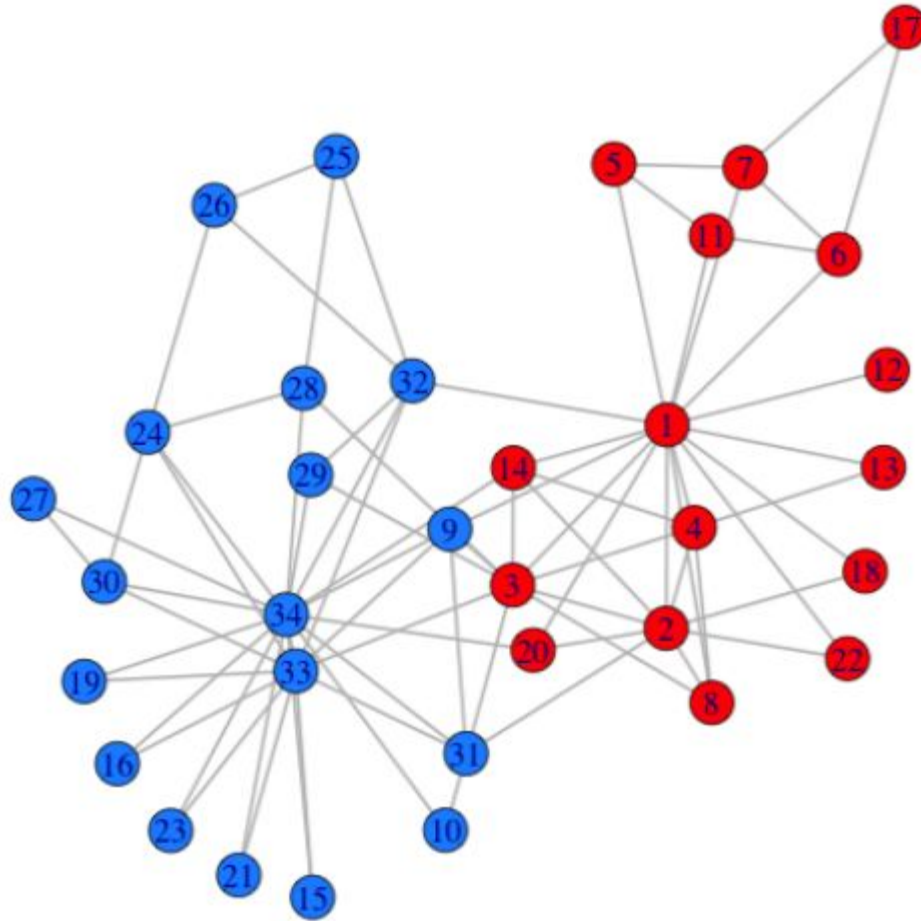
Karate club example



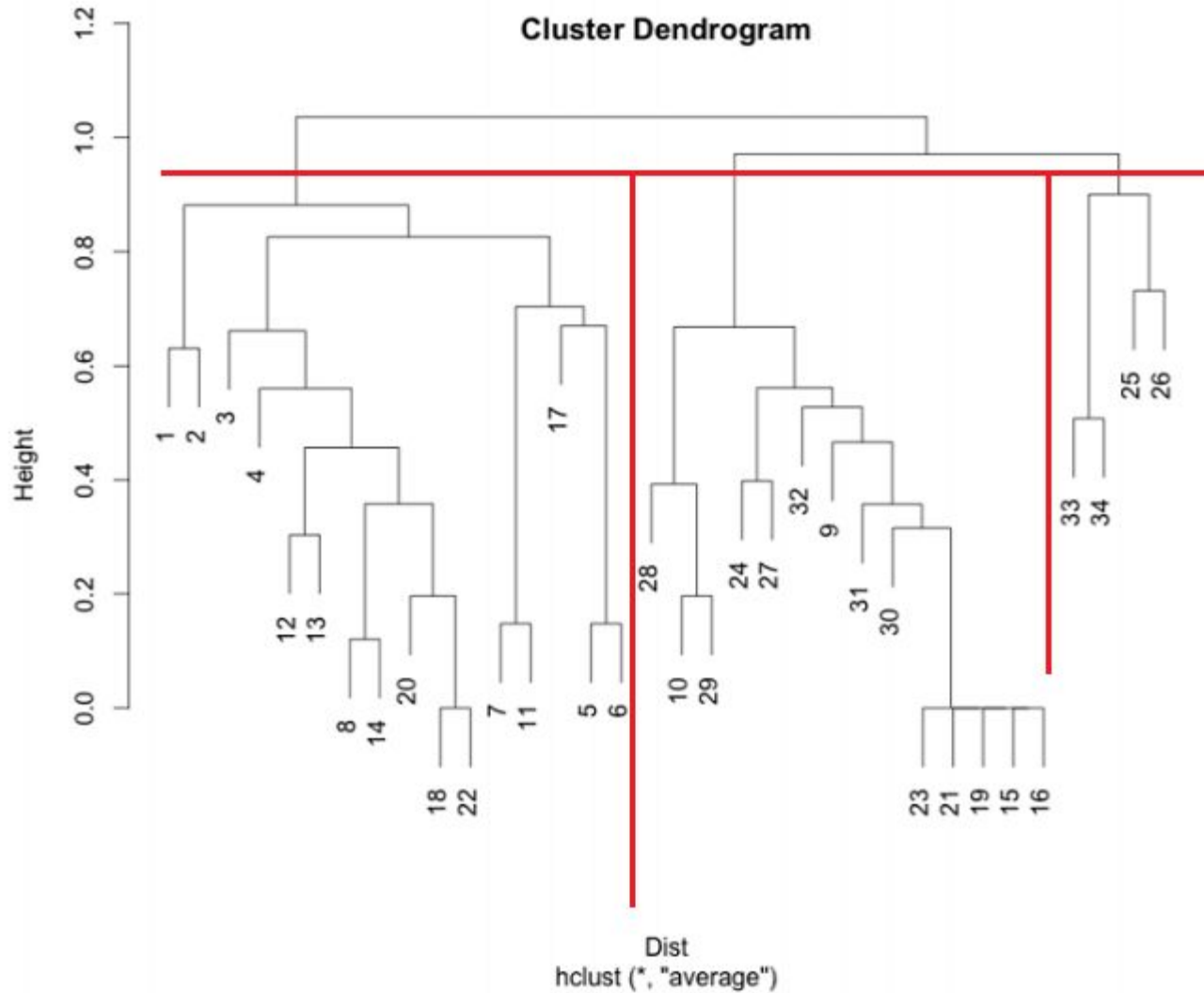
Karate club example



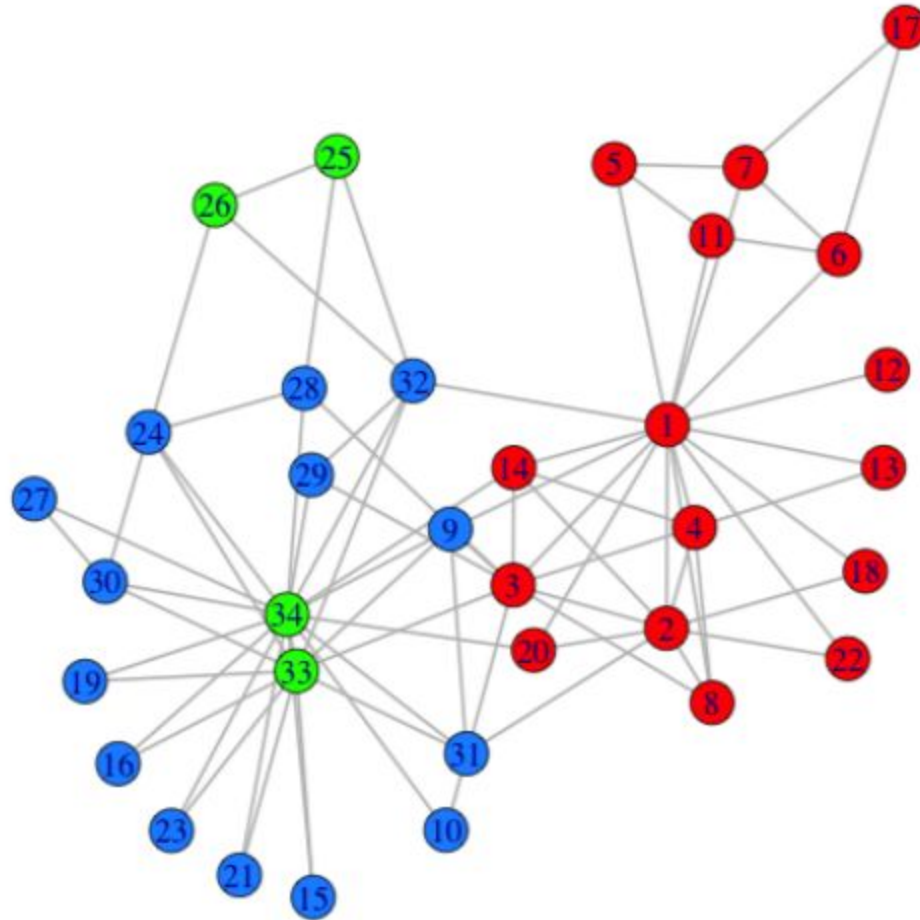
Karate club example



Karate club example



Karate club example



Graph Partitioning

Combinatorial problem:

- Number of ways to divide n nodes into n_1, n_2 nodes ($n_1+n_2 = n$), bi-partitioning

$$\# \text{ of combinations} = n! / (n_1! * n_2!)$$

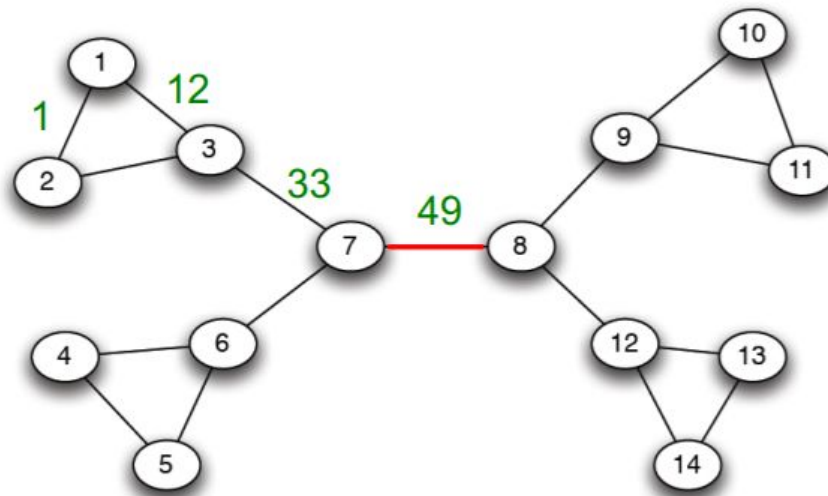
Community Detection

- Setup:
 - Undirected sparse graph ($m \ll n^2$)
 - Non-overlapping communities
 - Each community should be connected
 - Exact solution NP-hard, hence we are using heuristics / approximate algorithms

Edge Betweenness

- Number of shortest paths going via edge e

$$C_B(e) = \sum_{s \neq t} \frac{\sigma_{st}(e)}{\sigma_{st}}$$



Newman-Girvan algorithm

Algorithm: Newman-Girvan, 2004

Input: graph $G(V,E)$

Output: Dendrogram

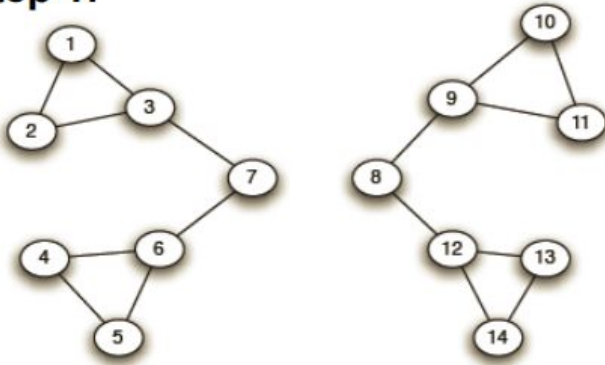
repeat

 For all $e \in E$ compute edge betweenness $C_B(e)$;
 remove edge e_i with largest $C_B(e_i)$;

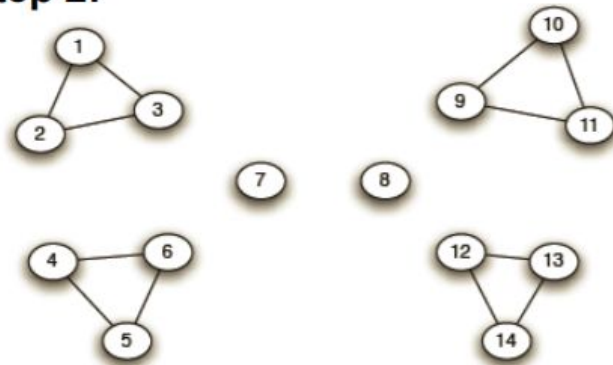
until *edges left*;

Step-by-step

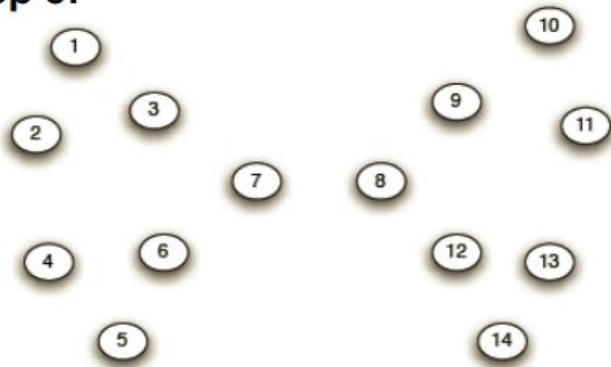
Step 1:



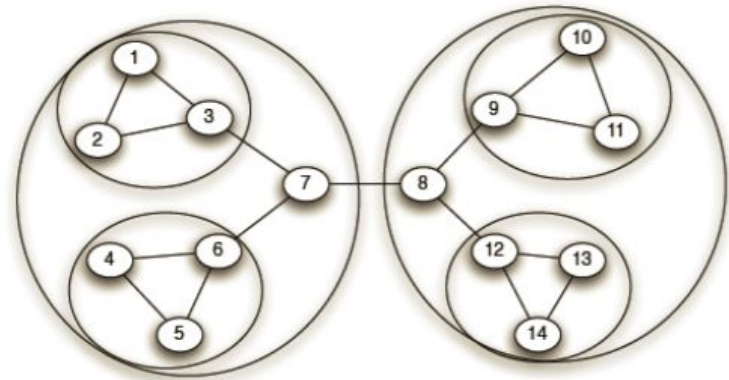
Step 2:



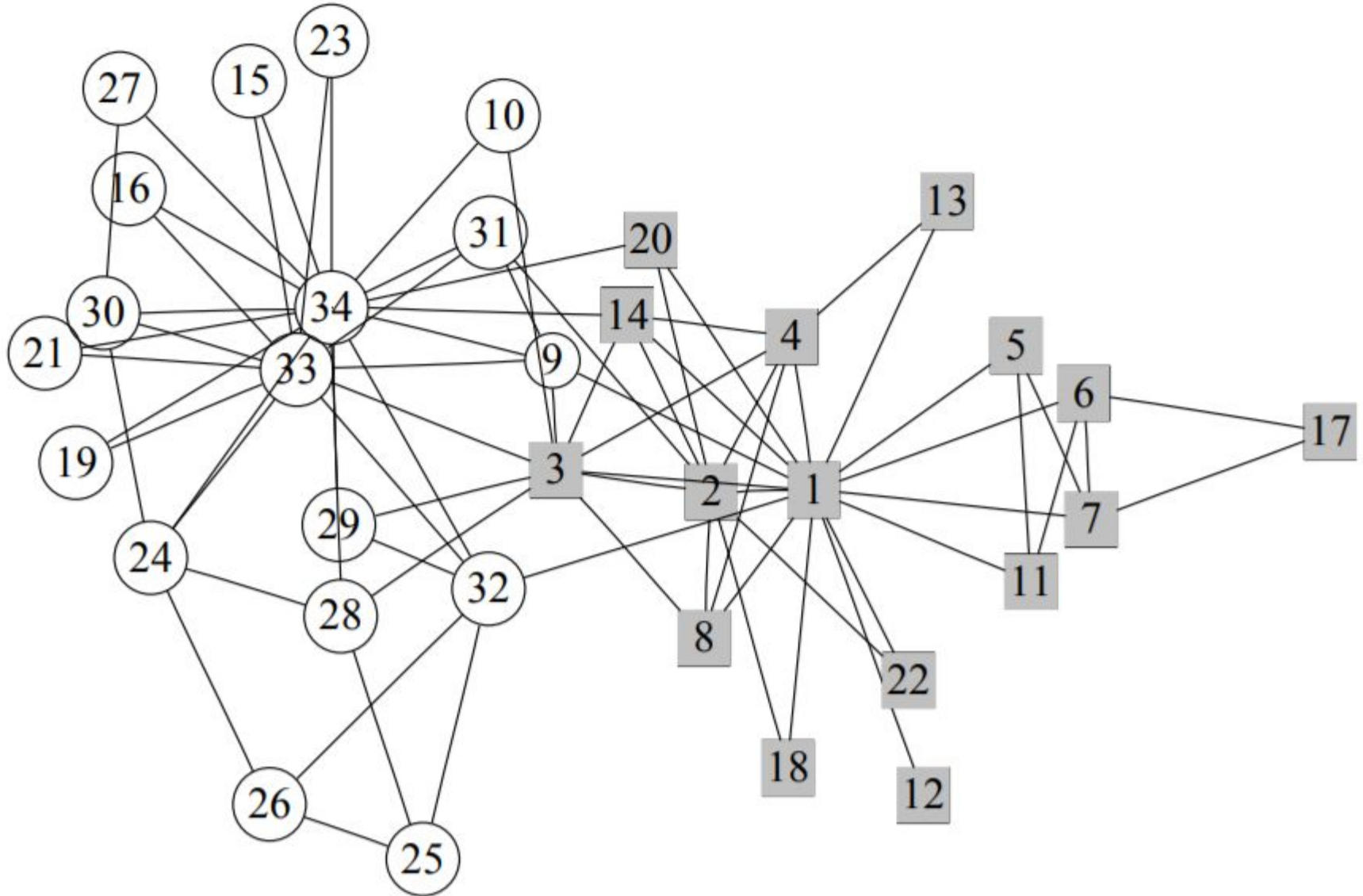
Step 3:



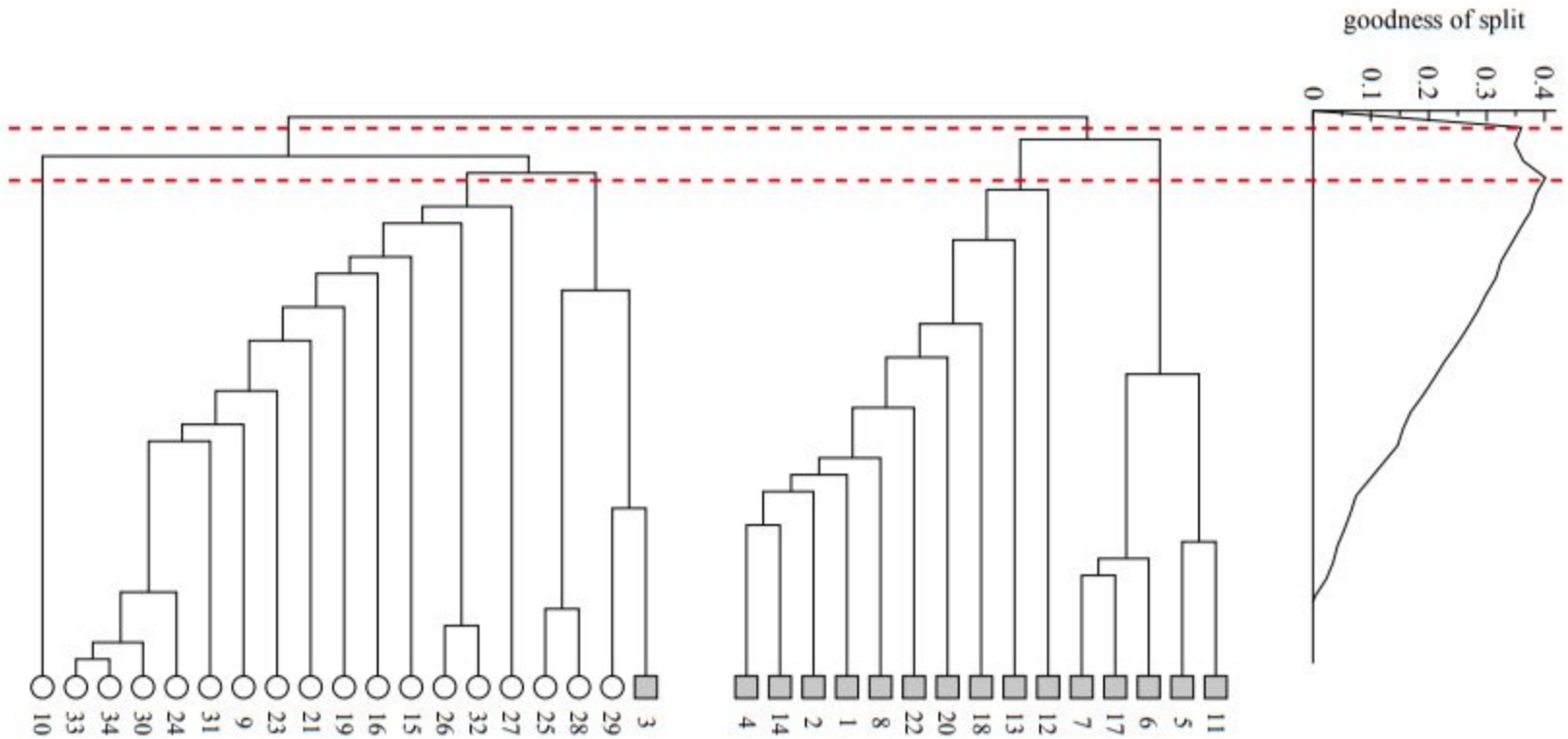
Hierarchical network decomposition:



Karate club example

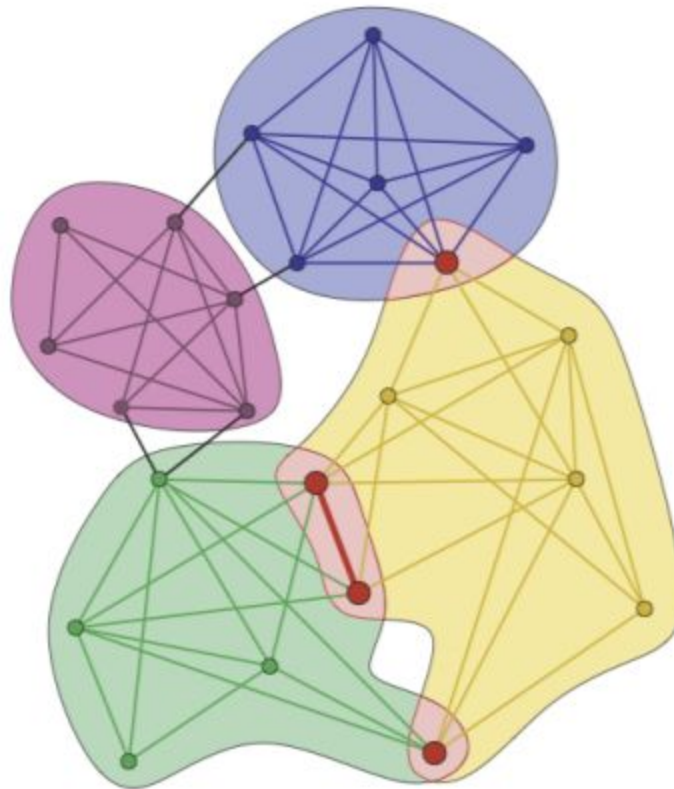



Karate club example



Overlapping Communities

- Next lesson we will extend the community detection task to find overlapping communities





Thank you!
Questions?