Autumn 2023 Exercise class: in 302 on October 10th

Exercises for Computational Tools for Data Science (02807)

WEEK 7: MINING SOCIAL NETWORK GRAPHS

For this exercise sheet we recommend installing and using the following Python libraries: *networkx, scikit-learn* and *matplotlib*.

For more information about *scikit-learn* and *matplotlib* see Exercise Sheet 6.

For information about *networkx* see: https://networkx.org/.

Exercise 1: Divisive clustering via Girvan–Newman

Implement a divisive hierarchical clustering algorithm via the Girvan–Newman method and compute modularity scores. Hence, you should implement the following parts:

- 1. Compute the betweenness centrality for every edge of a graph.
- 2. Implement a divisive hierarchical clustering algorithm based on removing edges with highest betweenness centrality and considering connected components as clusters.
- 3. Compute modularity for a clustering of a graph.

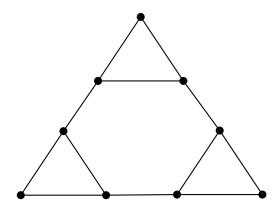
Test your implementations on the Karate-Club graph, found in *networkx.karate_club_graph*. Compare your implementations with the ones from *networkx*:

- $\bullet \ network x. algorithms. centrality. edge_betweenness_centrality$
- $\bullet \ network x. algorithms. community. girvan_newman$
- $\bullet \ network x. algorithms. community. modularity$

Use *matplotlib* to visualise the communities.

Exercise 2:

Consider the following graph G.



Compute the eigenvalues of the Laplacian matrix L(G) of G.

(To compute eigenvalues you could use numpy.linalg.eigvals (you should import numpy).) (If you use networkx, you could use networkx.linalg.laplacianmatrix to obtain the Laplacian matrix of G.)

Derive a bipartition of G via the second smallest eigenvalue of L(G).

Compare your result with *SpectralClustering* from *sklearn.cluster* for 2 clusters. Experiment with *SpectralClustering*, e.g. by using more clusters or running it on the Karate-club graph from *networkx.karate_club_graph*.