MADR 2025 - Project tasks 6

Spectral clustering

Consider the following versions of the spectral clustering algorithm:

• Algorithm 1: unnormalized spectral clustering

input: a similarity graph G = (V, E, W), number of clusters k

- 1. compute L = D W and its first k eigenvectors u_1, \dots, u_k ,
- 2. let y_1,\dots,y_n be the rows of the matrix containing u_1,\dots,u_k as its columns,
- 3. cluster the points $y_1,\dots,y_n\in\mathbb{R}^k$ to C_1,\dots,C_k by the k-means algorithm

 $output \hbox{: clusters } A_1, \ldots, A_k \hbox{: } A_i = \{j \in V : y_j \in C_i\}.$

- Algorithm 2: normalized spectral clustering
- $\begin{array}{l} 1. \ \ \text{compute} \ \mathcal{L} = D^{-1/2}LD^{-1/2} \ \text{and its first} \ k \ \text{eigenvectors} \ w_1, \ldots, w_k, \\ 2. \ \ \text{let} \ \phi_1, \ldots, \phi_n \ \text{be the rows of the matrix containing} \ D^{-1/2}w_1, \ldots, D^{-1/2}w_k \ \text{as its columns,} \\ 3. \ \ \text{cluster the points} \ \phi_1, \ldots, \phi_n \in \mathbb{R}^k \ to \ C_1, \ldots, C_k \ \text{by the k-means algorithm,} \end{array}$

 $output \hbox{: clusters } A_1, \ldots, A_k \hbox{: } A_i = \{j \in V : y_j \in C_i\}.$

• Algorithm 3: normalized spectral clustering for k=2 clusters, without k-means

input: a similarity graph G = (V, E, W), number of clusters k

- 1. compute \mathcal{L} and its second eigenvector w_2 ,
- 2. set $y := D^{-1/2}w_2$ and order its coordinates,
- 3. find A_{opt} minimizing the Cheeger's cut among A_1, \dots, A_k (as defined during the lectures)

output: clusters A_{opt}, A_{opt}^C .

• Implement the algorithms yourself.

- Use your implementations for several different data sets of your choice. Try different numbers of clusters to see the effect that this has.
- Compare the performances of the normalized and unnormalized algorithms on the same data sets.
- Can you provide an example in which Algorithm 3 (the one with some theoretical guarantees) is superior to Algorithm 2 (with two clusters)?