# Optimization and Computational Linear Algebra for Data Science Lecture 8: Graphs and Linear Algebra

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Warning: This material is not meant to be lecture notes. It only gathers the main concepts and results from the lecture, without any additional explanation, motivation, examples, figures...

## 1 Graphs

We start by a formal definition of a (simple non-oriented) graph:

### Definition 1.1 (Graph)

A graph G is defined as a pair  $V_G$ ,  $E_G$  where  $V = V_G$  is the set of vertices of G and  $E = E_G$  is the set of edges of G which is a subset of  $V \times V$ . Two vertices i, j are connected by an edge if  $\{i, j\} \in E$ . In such case we write  $i \sim j$  and say that i and j are neighboors.

#### Definition 1.2

The degree of a node  $i \in V$  is the number of its neighboors.

In this lecture we will only consider finite graphs, where V is finite. We let n = #V. One can assume (up to renaming the vertices) that  $V = \{1, \ldots, n\}$ .

#### Definition 1.3

We define the adjacency matrix  $A \in \mathbb{R}^{n \times n}$  of the graph G by

$$A_{i,j} = \begin{cases} 1 & \text{if } i \sim j \\ 0 & \text{otherwise.} \end{cases}$$

The degree matrix of G is defined by  $D = \text{Diag}(\deg(1), \dots, \deg(n))$ .

Notice that A is a symmetric matrix.

### 2 Graph Laplacian

#### Definition 2.1 (Graph Laplacian)

The Laplacian matrix of G is defined as

$$L = D - A$$
.

#### Proposition 2.1

The matrix L satisfies the following properties:

- 1. L is symmetric and positive semi-definite.
- 2. The smallest eigenvalue of L is 0, the corresponding eigenvector is the constant one vector  $(1,1,\ldots,1)$ .
- 3. L has n non-negative eigenvalues  $0 = \lambda_1 \le \lambda_2 \le \cdots \le \lambda_n$ .

#### Proposition 2.2

The graph G is connected if and only if  $\lambda_2 > 0$ .

More generally, one can show that the multiplicity of the eigenvalue 0 of L (i.e. the number of i such that  $\lambda_i = 0$ ) is equal to the number of connected components of L.

# 3 Spectral clustering with the graph Laplacian

### Proposition 3.1

Assume that G is connected. Let  $v_2$  be an eigenvector associated to  $\lambda_2$ , the second smallest eigenvalue of L. Let

$$W = \{i \mid v_2(i) \ge 0\}.$$

Then the subgraph induced by W is connected.

This indicates that

$$cut(G, v) = \sum_{i \sim j} \mathbb{1}(v_i \neq v_i) = \frac{1}{2} \sum_{i,j} A_{i,j} (1 - v_i v_j)$$

