

Introduction to graphs and matrices

Spectral graph theory lies in the intersection of linear algebra and graph theory. It is a very broad subject, with applications in different areas of research, such as graph theory (of course), physics, statistics, computer science, data science, . . . There are too many topics and applications to fit into one course, and we have flexibility regarding which to cover. So, please talk to me about your interests!

Graphs

A graph is a pair $G(V, E)$, where V is a set of *vertices* (nodes) and $E \subseteq \binom{V}{2}$ a set of *edges* (links). In this course, we will almost always consider graphs to be *simple*, meaning that they have no self-loops or multiple edges, and that edges are undirected. Let $d(v)$ indicate the *degree* of a vertex $v \in V(G)$, i.e. the number of edges that are *incident* to it (contain v), or the number of vertices *adjacent* to v (that share an edge with v). A *path* is a graph of the form $P(V, E)$ with $V = \{v_0, \dots, v_k\}$ and $E = \{v_0v_1, \dots, v_{k-1}v_k\}$. A *cycle* is a graph of the form $C(V, E)$ with $V = \{v_0, \dots, v_k\}$ and $E = \{v_0v_1, \dots, v_{k-1}v_k, v_1v_k\}$. We say that a graph is *connected* if there exists a path in the graph between any pair of vertices.

We can store/communicate the information contained in a graph by, for example, drawing the graph, or listing the sets V and E , or maybe as a dictionary that keeps track of the set of neighbors of each vertex, . . . Here, however, we will focus on ways to store graphs as a matrix. As it turns out there are many ways to do this, each with their own advantages and applications.

Incidence matrix

We start with an *incidence matrix* A , which has a row for each vertex, and a column for each edge of G . We let $A_{ve} = 1$ if $v \in e$ and $A_{ve} = 0$ otherwise.

A famous result in graph theory is the so-called Handshake Lemma.

Lemma 1 (Handshake Lemma.). *For any graph G , we have*

$$2|E| = \sum_{v \in V} d(v).$$

Exercise 1. *Use the incidence matrix of a graph G to prove the Handshake Lemma.*

1 point

Exercise 2. *As a variation on the incidence matrix, consider the following. For each edge $e = uv$ and associated column in the matrix A , instead of putting a 1 in both positions u and v , set one of them to 1 and one to -1 (this choice is arbitrary, so this matrix is not uniquely defined). Suppose that a graph G is connected, and consider such an associated matrix A . Can you describe cycles, forests and spanning trees in G in terms of linear algebra on the columns of A ?*

2 points

Adjacency matrix

The *adjacency matrix* M (or often denoted as A) has a row for each vertex and a column for each vertex of G . We let $M_{uv} = 1$ if $uv \in E$ and $M_{uv} = 0$ otherwise.

Since we assume that graphs are simple, the adjacency matrix has only real-valued entries and is symmetric (and Hermitian), i.e. we have $M = M^T$. We will assume a few facts about symmetric matrices. You may try to prove these as an exercise.

Theorem 2. Let M be a real, symmetric matrix. Then,

- (i) for any eigenvalue μ of M , its algebraic multiplicity is equal to its geometric multiplicity;
- (ii) all eigenvalues of M are real;
- (iii) if \vec{x} and \vec{y} are eigenvectors of M with distinct eigenvalues λ and μ , respectively, then $\vec{x} \perp \vec{y}$. Therefore, there exists an orthonormal eigenbasis for the columnspan of M .

Exercise 3. Show that if two graphs G_1 and G_2 are isomorphic (the same under reordering of the vertices) then their adjacency matrices M_1 and M_2 are similar, i.e. find a matrix B such that $B^{-1}M_1B = M_2$. 1 point

Exercise 4. Let G and H be two graphs. What can you say about the spectrum of the disjoint union $G \cup H$? 1/2 point

Exercise 5. Suppose that G is a k -regular graph. Give one eigenvector and associated eigenvalue of M . 1/2 point

Exercise 6. Show that the eigenvalues of M are bounded by $\Delta(G)$ (the maximum degree of G). 1/2 point

Exercise 7. Think of M as an operator on the set of functions $g : V(G) \rightarrow \mathbb{R}$ (write g as a column vector of length n such that $g(v) = \vec{g}_v$). For a vertex v , what is $(M\vec{g})_v$? 1/2 point

Exercise 8. What does the matrix M tell us about walks in the graph? Write the following invariants in terms of the matrix M : $d(u, v)$ (distance between two vertices u and v), $\text{mboxecc}(v)$ (the eccentricity of a vertex v : maximum $d(u, v)$ over all $u \in V$) and $\text{diam}(G)$ (the diameter of the graph: maximum $d(u, v)$ over all $u, v \in V$). 2 points

Exercise 9. One more fact about symmetric matrices is that if μ_1, \dots, μ_k are the unique eigenvalues of M , then $(M - \mu_1 I)(M - \mu_2 I) \dots (M - \mu_k I) = 0$ (this is the minimal polynomial of M). Use this to show that M^l for any $l \geq 0$ can be written as a linear combination of $I, M, M^2, \dots, M^{k-1}$. Then, conclude that the number of unique eigenvalues of M must exceed the diameter of G . What does this tell us about the spectrum of the path graph P_n ? 2 points

Exercise 10. Find the eigenvalues of the adjacency matrix of K_n , the complete graph on n vertices. 1 point

Exercise 11. Find the eigenvalues of the adjacency matrix of K_{n_1, n_2} , the complete bipartite graph with partite sets of order n_1 and n_2 respectively. 2 points

Exercise 12. Find the eigenvalues of the adjacency matrix of C_n , the cycle graph on n vertices. Start with a directed cycle (not a symmetric matrix). 3 points