# MA431 Spectral Graph Theory: Lecture 4

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#### 7 The Laplacian matrix and spectrum

Let G = (V, E) be a graph (recall that loops are not allowed by parallel edges are). Denote by  $\Delta(G)$  the diagonal matrix corresponding to the vertex degrees of G. That is, the rows and columns of  $\Delta(G)$  are indexed by V, and for each vertex  $u \in V$ , the *uu*-entry of  $\Delta(G)$  is equal to deg(u). Recall that A(G) is the adjacency matrix of G.

**Definition 7.1.** The Laplacian matrix of G is the real symmetric matrix  $\Delta(G) - A(G)$ .

An orientation of G is a directed graph D that is obtained from G by orienting every edge in an arbitrary direction. The *incidence matrix of* D is the  $0, \pm 1$  matrix whose rows and columns are indexed by the vertices and arcs, respectively, where column (v, u) is equal to  $e_u - e_v$ .

**Proposition 7.2.** Let *L* be the Laplacian matrix of *G*. Then

- 1.  $L = MM^{\top}$ , where M is the incidence matrix of any orientation of G,
- 2.  $L = \sum_{\{u,v\}\in E} (e_u e_v)(e_u e_v)^{\top}$ ,
- 3. for every  $x \in \mathbb{R}^V$ ,

$$x^{\top}Lx = \sum_{\{u,v\}\in E} (x_u - x_v)^2.$$

In particular, L is a positive semidefinite matrix.

Proof. Exercise.

**Definition 7.3.** The Laplacian spectrum of G is the spectrum of its Laplacian matrix. If G has n vertices, then its spectrum is denoted  $\lambda_1(G) \leq \cdots \leq \lambda_n(G)$ .<sup>1</sup>

For general graphs, the Laplacian spectrum and the spectrum are not related; for example, it is possible for two cospectral graphs to have different Laplacian spectra (see Exercises). For regular graphs, however, the situation is different:

**Theorem 7.4.** Let G be an n-vertex graph that is k-regular. If G has spectrum  $\theta_1 \ge \theta_2 \ge \cdots \ge \theta_n$ , then its Laplacian spectrum is  $k - \theta_1 \le k - \theta_2 \le \ldots \le k - \theta_n$ .

<sup>&</sup>lt;sup>1</sup>Note that for the Laplacian spectrum,  $\lambda_1$  denotes the least eigenvalue, while for the usual spectrum,  $\theta_1$  denotes the largest eigenvalue.

*Proof.* Let A := A(G), and let  $v_1, \ldots, v_n$  be eigenvectors of A with eigenvalues  $\theta_1, \ldots, \theta_n$ , respectively. Let  $L := \Delta(G) - A$  be the Laplacian matrix of G. As G is k-regular,  $\Delta(G) = kI$ , so L = kI - A. Subsequently,

$$Lv_i = (kI - A)v_i = (k - \theta_i)v_i,$$

implying in turn that  $v_1, \ldots, v_n$  are also eigenvectors of L with eigenvalues  $k - \theta_1, \ldots, k - \theta_n$ , as claimed.  $\Box$ 

Given that the Laplacian matrix is positive semidefinite, its eigenvalues are nonnegative. In fact, the least eigenvalue of the Laplacian spectrum is guaranteed to be 0:

**Proposition 7.5.** Let G be a graph with c connected components, let L be its Laplacian matrix, and let  $\lambda_1 \leq \cdots \leq \lambda_n$  be the Laplacian spectrum. Then the following statements hold:

- 1.  $L\mathbf{1} = \mathbf{0}$ , that is,  $\mathbf{1}$  is an eigenvector with eigenvalue 0. In particular,  $\lambda_1 = 0$ .
- 2. If Lx = 0, then x takes the same value on the vertices of each connected component of G.
- 3.  $\operatorname{rank}(L) = n c$ . Equivalently, the eigenvalue 0 of L has multiplicity c.

*Proof.* (1) follows immediately from the definition of the Laplacian matrix.

(2) Let x be a vector such that Lx = 0. By Proposition 7.2,

$$0 = x^{\top} L x = \sum_{\{u,v\} \in E} (x_u - x_v)^2$$

implying that  $x_u = x_v$  whenever u, v are adjacent. Thus x takes the same value on the vertices of each connected component, as required.

(3) Let  $V_1, \ldots, V_c$  be the vertex sets of the connected components of G. For each  $i \in [c]$ , let  $v_i \in \{0, 1\}^V$  be the incidence vector of  $V_i$ . We claim that  $v_1, \ldots, v_c$  is a basis for the null space of L, i.e.  $\{x : Lx = 0\}$ . Clearly,  $Lv_i = 0$ , and the  $v_i$  are linearly independent. Now choose a vector x such that Lx = 0. Then, by (2), x is a linear combination of  $v_1, \ldots, v_c$ . Thus,  $v_1, \ldots, v_c$  is a basis for the null space of L, implying in turn that rank(L) = n - c.

Given that the least Laplacian eigenvalue is zero, one may ask questions about the second least Laplacian eigenvalue of a graph. Fiedler [2] calls  $\lambda_2(G)$  the *algebraic connectivity of G*.

One can also get an upper-bound of n on the largest eigenvalue  $\lambda_n(G)$  of the Laplacian of a simple graph G – see Exercises).

#### 8 The Matrix-Tree Theorem

Let G = (V, E) be an *n*-vertex graph, and let L be the Laplacian matrix. Denote by T(G) the number of spanning trees of a graph – so if G is not connected, this number is zero. In this section, we prove Kirchhoff's

*Matrix-Tree Theorem*, which states that T(G) is equal to the determinant of any  $(n-1) \times (n-1)$  principal submatrix of L.

The Matrix-Tree Theorem is by and large a consequence of the Laplace (cofactor) expansion for the determinant, combined with a powerful *deletion-contraction* recursive formula for T(G). To elaborate on the latter, let G be a graph, and let e be an edge. The *deletion*  $G \setminus e$  is the graph obtained from G after removing the edge e. The *contraction* G/e is the graph obtained after identifying the ends of G, and deleting all the loops created.<sup>2</sup> Observe that contracting may create (additional) parallel edges.

**Lemma 8.1.** Let G be a graph. Then for every edge e,

$$T(G) = T(G/e) + T(G \setminus e).$$

*Proof.* The spanning trees of G can be separated into two groups, those that contain the edge e, and those that do not. The ones in the second group are precisely the spanning trees of  $G \setminus e$ . The ones are in the first group, however, are in correspondence with the spanning trees of G/e. More precisely, if T' is a spanning tree of G/e then  $T' \cup \{e\}$  is a spanning tree of G containing e, and if T is a spanning tree of G containing e then  $T - \{e\}$  is a spanning tree of  $G \setminus e$ . The formula above is an immediate consequence of this grouping of the spanning trees of G.

In the next lecture, we will prove the following theorem:

**Theorem 8.2** (Matrix-Tree Theorem). Let G be an n-vertex graph, and let L be its Laplacian matrix. Then T(G) is equal to the determinant of any  $(n-1) \times (n-1)$  principal submatrix of L.

### Acknowledgements

The presentation of  $\S7$  and  $\S8$  is inspired by [3], Chapter 13.

The Matrix-Tree Theorem dates back to the 1800s. Gustav Kirchhoff proved the "dual" of it in 1847 [4], but it was James Maxwell who stated the result explicitly in *A Treatise on Electricity and Magnetism*, *I* [5] (see Part II, Chapter 6, pp. 329-337). The theorem, as is, was stated and proved by Trent [6]. See also [1] for other references.

## References

- [1] S. Chaiken and D. Kleitman. Matrix tree theorems. Journal of Combinatorial Theory, Series A, 24:377–381, 1978.
- [2] M. Fiedler. A property of eigenvectors of nonnegative symmetric matrices and its application to graph theory. *Czechoslo-vak Mathematical Journal*, 25(4):619–633, 1975.
- [3] C. Godsil and G. Royle. Algebraic Graph Theory. Springer, New York, NY, 2000.

<sup>&</sup>lt;sup>2</sup>In general, loops are not deleted after edge contractions, but in our context we must.

- [4] G. Kirchhoff. Uber die auflösung der gleichungen, auf welche man bei der untersuchung der linearen verteilung galvanischer ströme gefuhrt wird. Ann. Phys. Chem., 72(497-508), 1847.
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- [6] H. M. Trent. Note on the enumeration and listing of all possible trees in a connected linear graph. *Proc. Nat. Acad. Sci.* U.S.A., 40:1004–1007, 1954.