

## NOTE ON DELETING A VERTEX AND WEAK INTERLACING OF THE LAPLACIAN SPECTRUM\*

ZVI LOTKER<sup>†</sup>

**Abstract.** The question of what happens to the eigenvalues of the Laplacian of a graph when we delete a vertex is addressed. It is shown that

$$\lambda_i - 1 \leq \lambda_i^v \leq \lambda_{i+1},$$

where  $\lambda_i$  is the  $i$ th smallest eigenvalues of the Laplacian of the original graph and  $\lambda_i^v$  is the  $i$ th smallest eigenvalues of the Laplacian of the graph  $G[V - v]$ ; i.e., the graph obtained after removing the vertex  $v$ . It is shown that the average number of leaves in a random spanning tree  $\mathcal{F}(G) > \frac{2|E|e^{-\frac{1}{\alpha}}}{\lambda_n}$ , if  $\lambda_2 > \alpha n$ .

**Key words.** Spectrum, Random spanning trees, Cayley formula, Laplacian, Number of leaves.

**AMS subject classifications.** 05C30, 34L15, 34L40.

**1. Introduction.** Given a graph  $G = (V, E)$  with  $n$  vertices  $V = \{1, \dots, n\}$  and  $E$  edges, let  $A$  be the adjacency matrix of  $G$ , i.e.  $a_{i,j} = 1$  if vertex  $i \in V$  is adjacent to vertex  $j \in V$  and  $a_{i,j} = 0$  otherwise. The *Laplacian* matrix of graph  $G$  is  $L = D - A$ , where  $D$  is a diagonal matrix where  $d_{i,i}$  is equal to the degree  $d_i$  of vertex  $i$  in  $G$ . The Laplacian of a graph is one of the basic matrices associated with a graph. The spectrum of the Laplacian fully characterizes the Laplacian (for more detail see [1]). Since  $L$  is symmetric and positive semidefinite, its eigenvalues are all nonnegative. We denote them by  $\lambda_1 \leq \dots \leq \lambda_n$ . One of the elementary operations on a graph is deleting a vertex  $v \in V$ , we denote the graph obtained from deleting the node  $v$  by  $G[V - v]$ , and the Laplacian Matrix of  $G[V - v]$  by  $L^v$ . Finally let  $\lambda_1^v \leq \dots \leq \lambda_{n-1}^v$  be the eigenvalues of  $L^v$ . A well known theorem in Algebraic Graph theory is the *interlacing* of Laplacian spectrum under addition/deletion of an edge; see for example [1, Thm. 13.6.2]) quoted next.

**THEOREM 1.1.** *Let  $X$  be a graph with  $n$  vertices and let  $Y$  be obtained from  $X$  by adding an edge joining distinct vertices of  $X$  then*

$$\lambda_{i-1}(L(Y)) \leq \lambda_i(L(X)) \leq \lambda_i(L(Y)),$$

for all  $i = 1, \dots, n$ , (we assume that  $\lambda_0 = -\infty$ ).

We remark that the eigenvalues of adjacency matrices  $A(G)$  and  $A(G[V - v])$  also interlace; see, for example, [1, Thm. 9.1.1]. A natural question is whether we get a similar behavior for the Laplacian when we add/delete a vertex. In this note we study this question.

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<sup>†</sup>Ben Gurion University, Communication Systems Engineering, Beer Sheva, Israel (zvilo@cse.bgu.ac.il).

**Related Work.** This work uses two theorems from Matrix Analysis. The first is Cauchy's Interlacing theorem which states that the eigenvalues of a Hermitian matrix  $A$  of order  $n$  interlace the eigenvalues of the principal submatrix of order  $n - 1$ , obtained by removing the  $i$ th row and the  $i$ th column for each  $i \in \{1, \dots, n\}$ .

**THEOREM 1.2.** *Let  $A$  be a Hermitian matrix of order  $n$  and let  $B$  be a principal submatrix of  $A$  of order  $n - 1$ . Then the eigenvalues of  $A$  and  $B$  are interlacing i.e.  $\lambda_1(A) \leq \lambda_1(B) \leq \lambda_2(A) \leq \dots \leq \lambda_{n-1}(B) \leq \lambda_n(A)$ .*

Proof of this theorem can be found in [2].

The second theorem we use is the Courant-Fischer Theorem. This theorem is an extremely useful characterization of the eigenvalues of symmetric matrices.

**THEOREM 1.3.** *Let  $L$  be a symmetric matrix. Then*

1. *the  $i$ th eigenvalue  $\lambda_i$  of  $L$  is given by*

$$\lambda_i = \min_U \max_{x \in U} \frac{x^t L x}{x^t x};$$

2. *the  $(n - i + 1)$ st eigenvalue  $\lambda_{n-i+1}$  of  $L$  is given by*

$$\lambda_{n-i+1} = \max_U \min_{x \in U} \frac{x^t L x}{x^t x},$$

where  $U$  ranges over all  $i$  dimensional subspaces.

Proof of this theorem can be found in [3, p. 186]. Let  $v \in V$  be a vertex. Let  $P$  be the principal submatrix after we delete the row and column that correspond to the vertex  $v$  of the Laplacian. Denote the eigenvalues of  $P$  by  $\rho_1 \leq \dots \leq \rho_{n-1}$ .

**2. Weak Interlace for the  $L, L^v$ .** In this section we show a weak interlacing connection between the  $L$  and  $L^v$ . Since  $L$  is a symmetric matrix we can use Cauchy's interlacing theorem. The next corollary simply applies this theorem for  $L$  and  $P$ .

**COROLLARY 2.1.**  $\lambda_1 \leq \rho_1 \leq \dots \leq \rho_{n-1} \leq \lambda_n$ .

The next lemma uses the Courant-Fischer Theorem in order to prove weak interlacing for  $L, P$ .

**LEMMA 2.2.** *For all  $i = 1, \dots, n - 1$ ,  $\rho_i \leq \lambda_i^v + 1$*

*Proof.* Let  $I_v = P - L^v$ . Note that  $I_v$  is a  $(0,1)$  diagonal matrix whose  $j$ th diagonal entry is 1 if and only if  $j$  is connected to  $v$  in  $G$ . Fix  $i \in \{1, \dots, n - 1\}$ . Using the Courant-Fischer Theorem it follows that

$$\rho_{n-i+1} = \max_U \min_{x \in U} \left\{ \frac{x^t P x}{x^t x} : U \subseteq \mathbb{R}^n, \dim(U) = i, x \in U = \text{span}(U) \right\},$$

where  $x^t$  is the transpose of  $x$ . Substituting  $L^v + I_v$  in  $P$  it follows that

$$\rho_{n-i+1} = \max_U \min_{x \in U} \left\{ \frac{x^t (L^v + I_v) x}{x^t x} : U \subseteq \mathbb{R}^n, \dim(U) = i, x \in U = \text{span}(U) \right\}.$$

Using standard calculus we get

$$\rho_{n-i+1} \leq \max_U \min_{x \in U} \left\{ \frac{x^t L^v x}{x^t x} : U \subseteq \mathbb{R}^n, \dim(U) = i, x \in U = \text{span}(U) \right\}$$

$$\begin{aligned}
 &+ \max_U \min_{x \in U} \left\{ \frac{x^t I_v x}{x^t x} : U \subseteq \mathbb{R}^n, \dim(U) = i, x \in U = \text{span}(U) \right\} \\
 &\leq \lambda_{n-i+1}^v + 1. \quad \square
 \end{aligned}$$

We now use the previous lemma to get a lower bound on  $\lambda_i^v$ .

LEMMA 2.3. *For all  $v = 1, \dots, n$  and for all  $i = 1, \dots, n - 1$ ,*

$$\lambda_i - 1 \leq \lambda_i^v.$$

*Proof.* Fix  $i \in \{1, \dots, n - 1\}$ . From Lemma 2.2 it follows that  $\rho_i \leq \lambda_i^v + 1$ . Now this lemma follows from substituting the conclusion of Corollary 2.1 into the previous inequality  $\lambda_i \leq \rho_i \leq \lambda_i^v + 1$ .  $\square$

The next lemma provides an upper bound on  $\lambda_i^v$ .

LEMMA 2.4. *For all  $v = 1, \dots, n$  and for all  $i = 1, \dots, n - 1$ ,*

$$\lambda_i^v \leq \lambda_{i+1}.$$

*Proof.* We prove this lemma by induction on  $d_v$ , the degree of the node  $v$ . If the degree is  $d_v = 0$ , then by removing the node  $v$  we reduce the multiplicity of the small eigenvalues, which is 0. Formally  $\lambda_i^v = \lambda_{i+1}$  for  $i = 1, \dots, n - 1$ . Therefore the lemma holds in this case. For the induction step, suppose that the statement holds for  $d_v = k$  and consider the case  $d_v = k + 1$ . Since  $d_v > 0$  it follows that there exists an edge  $e$  connecting the vertex  $v$  to some other node  $u$ . Denote the graph obtained by removing the edge  $e$  from the graph  $G$  by  $X$ . Let  $\sigma_1 \leq \dots \leq \sigma_{n-1}$  be the eigenvalues of the Laplacian of the graph  $X$ . From Theorem 1.1 it follows that  $\sigma_i \leq \lambda_i$  for all  $i = 1, \dots, n$ . Using induction we obtain that  $\lambda_{i-1}^v \leq \sigma_i \leq \lambda_i$ , for all  $i = 2, \dots, n$ .  $\square$

Now we present our main theorem.

THEOREM 2.5. *For all  $v = 1, \dots, n$  and for all  $i = 1, \dots, n - 1$ ,*

$$\lambda_i - 1 \leq \lambda_i^v \leq \lambda_{i+1}.$$

*Proof.* The proof is a direct consequence of Lemmas 2.3 and 2.4.  $\square$

We remark that both inequalities above are tight. To see that, we show there exist graphs such that  $\lambda_i - 1 = \lambda_i^v$ . Consider the graph  $K_n$ . It is well known that the eigenvalues of  $K_n$  are  $0, n, \dots, n$ , where the multiplicity of the eigenvalue  $n$  is  $n - 1$  and 0 is a simple eigenvalue. Now removing a vertex from  $K_n$  produces the graph  $K_{n-1}$ . Again the eigenvalues of  $K_{n-1}$  are  $0, n - 1, \dots, n - 1$ , where the multiplicity of the eigenvalue  $n - 1$  is  $n - 2$  and 0 is a simple eigenvalue. To see that there are graphs that satisfy  $\lambda_i^v = \lambda_{i+1}$ , consider the graph without any edges.

**3. Application to average leafy trees.** In this section we use the weak interlacing Theorem 2.5 to obtain a bound on the average number of leaves in a random spanning tree  $\mathcal{F}(G)$ . Our bound is useful when  $\lambda_2 > \alpha n$ , for fixed  $\alpha > 0$  and  $|E| = O(n^2)$ . We call such a graph a *dense expander*; in this case we show that the bound is linear in the number of vertices.

It is well known that the smallest eigenvalue of  $L$  is 0 and that its corresponding eigenvector is  $(1, 1, \dots, 1)$ . If  $G$  is connected, all other eigenvalues are greater than 0. Let  $P^v$  denote the submatrix of  $L$  obtained by deleting the  $v$ th row and  $v$ th column. Then, by the Matrix Tree Theorem, for each vertex  $v \in V$  we have  $t(G) = |\det(P^v)|$ , where  $t(G)$  is the number of spanning trees of  $G$ . One can rephrase the Matrix Tree Theorem in terms of the spectrum of the Laplacian matrix. The next theorem appears in [1, p. 284]; it connects the eigenvalues of the Laplacian of  $G$  and  $t(G)$ .

**THEOREM 3.1.** *Let  $G$  be a graph on  $n$  vertices and let  $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$  be the eigenvalues of the Laplacian of  $G$ . Then the number of labeled spanning trees in  $G$  is  $\frac{1}{n} \prod_{i=2}^n \lambda_i$ .*

Let  $G$  be a graph. Using the previous theorem it is possible to define the following probability space:  $\Omega(G) = \{T : T \text{ is a spanning tree in } G\}$ . On this set we take a spanning tree in a uniform probability. We are interested in finding the average number of leaves in a random spanning tree. Let  $T$  be a random spanning tree taken from  $\Omega(G)$  with the uniform distribution. Denote by  $\mathcal{F}(G)$  the expected number of leaves in  $T$ . Using the matrix theorem we can get a formula to compute the average number of leaves in a random spanning tree.

**LEMMA 3.2.**

$$\mathcal{F}(G) = \sum_{v \in V} \frac{nd_v \prod_{i=2}^{n-1} \lambda_i^v}{(n-1) \prod_{i=2}^n \lambda_i}$$

*Proof.* The number of trees that have vertex  $v$  as a leaf is  $\frac{d_v \prod_{i=2}^{n-1} \lambda_i^v}{n-1}$ . The lemma follows by summing over all vertices and dividing by the total number of trees.  $\square$

The weak interlacing theorem enables us to bound the average number of leaves in a dense expander graph. More precisely, we show that  $\mathcal{F}(G) = O(n)$ .

**THEOREM 3.3.** *Let  $G$  be a graph. If  $\lambda_2 > \alpha n$ , then the average number of leaves in  $T$  is bigger than  $\frac{2|E|e^{-\frac{1}{\alpha}}}{\lambda_n}$ .*

*Proof.*

$$\begin{aligned} \mathcal{F}(G) &= \sum_{v \in V} \frac{nd_v \prod_{i=2}^{n-1} \lambda_i^v}{(n-1) \prod_{i=2}^n \lambda_i} \\ &\geq \sum_{v \in V} \frac{nd_v \prod_{i=2}^{n-1} (\lambda_i - 1)}{(n-1) \prod_{i=2}^n \lambda_i} \\ &= \sum_{v \in V} \frac{nd_v \prod_{i=2}^{n-1} \frac{\lambda_i - 1}{\lambda_i}}{(n-1) \lambda_n} \\ &= \sum_{v \in V} \frac{nd_v \prod_{i=2}^{n-1} (1 - \frac{1}{\lambda_i})}{(n-1) \lambda_n} \\ &\geq \sum_{k \in V} \frac{nd_k (1 - \frac{1}{\lambda_2})^n}{(n-1) \lambda_n} \end{aligned}$$

$$\begin{aligned} &\geq \frac{2|E|e^{\frac{-n}{\lambda_2}}}{\lambda_n} \\ &\geq \frac{2|E|e^{\frac{-1}{\alpha}}}{\lambda_n}. \quad \square \end{aligned}$$

**COROLLARY 3.4.** *For any constant  $\alpha > 0$ , if  $\lambda_2 > \alpha n$ , and  $|E| = O(n^2)$ , then the average number of leaves in  $T$  is  $O(n)$ .*

**Conclusion.** In this paper we proved a weak interlacing theorem for the Laplacian. Using this theorem we showed that in a dense expander the average number of leaves is  $O(n)$ . A natural open question is to show that the average number of leaves in a random tree is an approximation to the maximal spanning leafy tree.

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