

Lecture 12: Largest eigenvalue and coloring

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We will start exploring connections between eigenvalues of Laplacian and combinatorial properties of the graph from this lecture. In today's lecture, we look at the connection of eigenvalues with bipartiteness and coloring. We will finish by showing interlacing property of eigenvalues when an edge is deleted.

Like last lecture, assume that L_G is the Laplacian of a graph $G = (V, E)$. Remember that L_G is symmetric and hence has n real eigenvalues. Let us call them, in the *ascending* order, $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$.

Similarly, assume that A_G is the adjacency matrix of a graph $G = (V, E)$. Remember that A_G is symmetric and hence has n real eigenvalues. Let us call them, in the *descending* order, $\mu_1 \geq \mu_2 \geq \dots \geq \mu_n$.

We ordered the eigenvalues of the Laplacian and the adjacency matrix this way because these eigenvalues are related for a d -regular graph.

Exercise 1. Show that $\lambda_k = d - \mu_k$, for a d -regular graph.

If the graph is not connected, then the adjacency (or Laplacian) matrix is block diagonal. Hence, the eigenvalues of the complete graph is just the collection of eigenvalues of all the connected components. So, we assume that our graph is connected.

To remind you, a k -coloring of a graph is a map $C : v \rightarrow [k]$, such that, no two adjacent vertices have the same color,

$$(i, j) \in E \Rightarrow C(i) \neq C(j).$$

We are generally interested in the minimum number of colors needed to color a graph.

Exercise 2. Show that if you can color G with k colors then you can color it with $l > k$ colors too.

The minimum number of colors needed to color a graph G is called the *chromatic number*, $\chi(G)$, of the graph.

Exercise 3. Show that the chromatic number of a bipartite graph is 2.

Finding the chromatic number of a graph is an NP-hard problem.

1 Perron Frobenius theorem

We know that every entry of an adjacency matrix A_G is non-negative. Perron Frobenius theorem gives us,

- the maximum eigenvalue of A_G has the largest absolute value, i.e., $\mu_1 \geq |\mu_n|$,
- the largest eigenvector is entry-wise non-negative.

Proof. You will prove the second property in the assignment. For the first property, we only need to show that for all unit vectors v ,

$$|v^T A v| \leq \mu_1.$$

Let u be the vector with absolute values of v as coordinates, $u_i = |v_i|$. Then,

$$|v^T A v| = \left| \sum_{(i,j) \in E} A_{(i,j)} v_i v_j \right| \leq \left| \sum_{(i,j) \in E} A_{(i,j)} u_i u_j \right| = u^T A u.$$

This follows because $A_{(i,j)}$ is non-negative. But we know that the maximum value of $x^T A x$ is μ_1 for all unit vectors x . So,

$$u^T A u \leq \mu_1.$$

□

If the graph is d -regular, we can convert this statement for the eigenvalues of the Laplacian.

Exercise 4. Show that the maximum eigenvalue of L_G is less than or equal to $2d$.

2 Bipartiteness and λ_n

Let us assume that we have a d -regular graph $G = (V, E)$. From the previous discussion we know that $\lambda_n \leq 2d$.

It turns out that λ_n captures the bipartiteness of the graph. In other words, we can prove the following theorem.

Theorem 1. *Given a connected d -regular graph $G = (V, E)$, the largest eigenvalue λ_n is $2d$ if and only if G is bipartite.*

Proof. We will prove the easy direction first. By definition of bipartite, we can divide the vertex set V into two parts, say A and B , such that, there are no edges inside A and inside B . Arrange the vertices so that all vertices of A come before all vertices of B . Then, A_G looks like,

$$\begin{pmatrix} 0 & A_1 \\ A_2 & 0 \end{pmatrix}$$

Here, every row of A and B sum to d . Then, the vector v , such that,

$$v_i = \begin{cases} 1 & i \in A \\ -1 & i \in B, \end{cases}$$

is an eigenvector of A_G with eigenvalue $-d$. In other words, there is an eigenvector of L_G with eigenvalue $2d$.

Exercise 5. Prove the above statement.

For the other direction, we need to prove that $\lambda_n < 2d$ if the graph is not bipartite. We will prove that $M = 2dI - L_G$ is positive semidefinite and there is an eigenvalue 0 then the graph is bipartite.

We will use the quadratic form of L for a unit vector x ,

$$x^T M x = x^T (2dI) x - x^T L_G x = 2d \sum_i x_i^2 - x^T L_G x. \quad (1)$$

From the previous lecture, we know the quadratic form of L_G ,

$$x^T L_G x = \sum_{(i,j) \in E} (x_i - x_j)^2 = \sum_i d x_i^2 - 2 \sum_{(i,j) \in E} x_i x_j.$$

Substituting this value in Eqn. 1,

$$x^T M x = d \sum_i x_i^2 + \sum_{(i,j) \in E} x_i x_j = \sum_{(i,j) \in E} (x_i + x_j)^2.$$

This shows that M is positive semidefinite and the maximum eigenvalue of L_G could be at most $2d$.

The quadratic form, $x^T M x$, can only be zero if $x_i = -x_j$ whenever $(i, j) \in E$. So, the negative and positive coordinates of x give the bipartition of G .

□

3 Coloring and eigenvalues of the adjacency matrix

Most of the content of this section is taken from Dan Spielman's course notes.

Let us first see the connection between μ_1 and degrees in the graph. For a d -regular graph, we know that $\mu_1 = d$. You will show in the assignment that μ_1 is less than the maximum degree for a general graph.

Theorem 2. Define d' to be the average degree of a graph G . Then,

$$\mu_1 \geq d'.$$

Let d'' to be the maximum average degree over any subgraph of G . Even then,

$$\mu_1 \geq d''.$$

Proof. Again, we will use the fact that μ_1 maximizes the quadratic form $x^T A_G x$ over all unit vectors x . In other words, $x^T A x$ for any unit vector x is a lower bound on μ_1 .

To prove the first assertion, choose x to be,

$$\frac{1}{\sqrt{n}}(1 \ 1 \cdots 1).$$

The quadratic form becomes $\sum_{i,j} A_{i,j} x_i x_j = \frac{1}{n} \sum_{i,j} A_{i,j}$.

Exercise 6. Show that $x^T A x$ is the average degree of graph G .

This proves that d' is a lower bound on μ_1 .

To prove that d'' is a lower bound on μ_1 , we will show that for every subgraph of G , average degree is a lower bound on μ_1 . This lower bound is achieved by using x to be the normalized indicator vector for the subgraph of G .

Exercise 7. Convince yourself that if x_S is the indicator vector for a subgraph S of G , then $x_S^T A_G x_S$ is the average degree of S .

□

Coming back to coloring, you will show in the assignment that you can always color a graph using *maximum degree plus one* number of colors. If there is a clique of size k in the graph then you need at least k colors.

Exercise 8. Show that if a graph G has a clique of size k then $\mu_1 \geq k - 1$.

The next theorem, known as Wilf's theorem, shows that you can color a graph with $\lfloor \mu_1 \rfloor + 1$ colors.

Theorem 3 (Wilf). Let G be a graph with adjacency matrix A_G and μ_1 be the maximum eigenvalue of A_G . Then, G can be colored with $\lfloor \mu_1 \rfloor + 1$ colors.

Proof. We will prove the theorem by induction. Since μ_1 is more than the average degree by Thm. 2, there exists a vertex with less than or equal to $\lfloor \mu_1 \rfloor$ number of neighbours. Call this vertex v .

Look at the subgraph G' on vertex set $V - v$. We will show that it can be colored with $\lfloor \mu_1 \rfloor + 1$ number of colors. Then, vertex v can also be colored with these colors. This is because v 's neighbours are one less than the total number of colors, leaving one color for the vertex v itself.

To prove that G' can be colored with $\lfloor \mu_1 \rfloor + 1$ colors, using induction, we only need to prove that the maximum eigenvalue of $A_{G'}$ is less than the maximum eigenvalue of A_G .

This follows simply because $A_{G'}$ is a submatrix of A_G . This fact is given as an assignment.

□

4 Eigenvalues with edge deletion

Let L_G be the Laplacian of a graph G . We are interested in the eigenvalues of $L_{G'}$ where G' is the graph with one edge deleted from G .

Note 1. Previously we talked about subgraph when a vertex was deleted. This time, we are deleting an edge.

We will use Courant-Fischer theorem, covered in the last class, to relate the eigenvalues of L_G and $L_{G'}$. For your reference, the theorem is given below.

Theorem 4 (Courant-Fischer). *Let M be a symmetric $n \times n$ matrix with eigenvalues $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$. Then,*

$$\lambda_k = \min_{S: \dim(S)=k} \max_{x \in S: \|x\|=1} x^T M x.$$

Here, S is a subspace of \mathbb{R}^n . Also,

$$\lambda_k = \max_{S: \dim(S)=n-k+1} \min_{x \in S: \|x\|=1} x^T M x.$$

Suppose we delete an edge (i, j) from G to obtain G' . Say, v be the vector with 1 at i -th position, -1 at the j -th position and 0 otherwise. The important thing to notice is that $L_G = L_{G'} + vv^T$.

Exercise 9. Prove the above statement.

The following theorem gives the relationship between the eigenvalues of M and $M' = M \pm vv^T$, addition/subtraction by a rank one matrix, using the Courant-Fischer theorem.

Theorem 5. *Let $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_n$ be the eigenvalues of M and $\eta_1 \leq \eta_2 \leq \dots \leq \eta_n$ be the eigenvalues of $M' = M \pm vv^T$ for some vector v . Then,*

$$\lambda_i \leq \eta_{i+1}.$$

Before we start the proof, we need a lemma about subspaces.

Lemma 1. *Suppose we are given a subspace S of dimension k and a vector v . For any vector $x \in S$ and $x \perp v$, we can always find a subspace S' of dimension $k - 1$, such that,*

$$x \in S' \text{ and } v \perp S'.$$

Proof. If $v \perp S$ then any S' containing x will work.

Otherwise, let v_1, \dots, v_n be an orthonormal basis of the entire space, such that, v_1, \dots, v_k is a basis of S and v_{k+1}, \dots, v_n are perpendicular to S .

Then, v can be written as $v = \sum_{i=1}^n \alpha_i v_i$. Now, construct a basis w_1, \dots, w_k by Gram-Schmidt decomposition, such that, w_1 is the unit vector in the direction of $\sum_{i=1}^k \alpha_i v_i$.

The following exercise completes the proof.

Exercise 10. Show that the space spanned by w_2, \dots, w_k is the required subspace S' . That is, $x \in \text{span}(w_2, \dots, w_k)$ and $v \perp w_i$ for all $i \neq 1$.

□

Proof of Thm. 5. From Courant-Fischer theorem,

$$\eta_{i+1} = \min_{S: \dim(S)=i+1} \max_{x \in S: \|x\|=1} x^T (M + vv^T) x.$$

If we restrict the feasible space of a maximization problem, the optimal value decreases.

$$\eta_{i+1} \geq \min_{S: \dim(S)=i+1} \max_{x \in S: \|x\|=1, x \perp v} x^T (M + vv^T) x = \min_{S: \dim(S)=i+1} \max_{x \in S: \|x\|=1, x \perp v} x^T M x$$

Look at the space S which minimizes the right hand side of the previous equation with vector x . Let $S' \subseteq S$ be the subspace of dimension i from Lem. 1, such that, v is perpendicular to S' and x is in S' . This shows that we can just optimize over all subspaces S' of dimension i with v being perpendicular to S' .

$$\eta_{i+1} \geq \min_{S': \dim(S')=i, v \perp S'} \max_{x \in S': \|x\|=1} x^T M x.$$

Again, we notice that if feasible space is increased than we get a higher value for a minimization problem.

$$\eta_{i+1} \geq \min_{S': \dim(S')=i} \max_{x \in S': \|x\|=1} x^T M x = \lambda_i.$$

The last equality follows from Courant-Fischer theorem. \square

By interchanging M and M' , Thm. 5 shows that $\eta_i \leq \lambda_{i+1}$. Hence, Thm. 5 implies that the eigenvalues of L_G and $L_{G'}$ are interlaced with each other.

5 Assignment

Exercise 11. Suppose M' is a submatrix of M . Show that the maximum eigenvalue of M' is less than the maximum eigenvalue of M . Similarly, show that the minimum eigenvalue of M' is bigger than the minimum eigenvalue of M .

Exercise 12. Suppose A_G is the adjacency matrix of a connected graph. Show that the eigenvector corresponding to the highest eigenvalue is non-negative. Then, show that it is strictly positive.

Exercise 13. Show that if a graph G is not bipartite then $\mu_1 = -\mu_n$.

Exercise 14. Show that the largest eigenvalue of the adjacency matrix is smaller than the maximum degree.

Exercise 15. Let d_m be the maximum degree in a graph. Prove that you can color a graph with $d_m + 1$ colors.