

# Graph Theory

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## 1 Key Topics

Today, we continue our investigation of eigenvalues, eigenvectors, and M-matrices.

Recall that an eigenvalue of the matrix  $A \in \mathbb{C}^{n \times n}$  is a scalar  $\lambda \in \mathbb{C}$  such that  $\lambda I - A$  is singular, i.e., there exists a non-zero  $\mathbf{x} \in \mathbb{C}^n$  such that  $(\lambda I - A)\mathbf{x} = 0$ . The vector  $\mathbf{x}$  is an eigenvector of  $A$  corresponding to  $\lambda$ . Moreover, note that  $A\mathbf{x} = \lambda\mathbf{x}$ .

Also, recall that  $A \in \mathbb{R}^{n \times n}$  is an M-matrix if it can be written in the following form:

$$A = sI - B,$$

where  $s \geq \rho(B)$  and  $B \geq 0$ . We can bound the spectral radius of a matrix using Gerschgorin's theorem.

**Theorem 1.1.** *Let  $A \in \mathbb{C}^{n \times n}$ . Then for any eigenvalue  $\lambda$  of  $A$  there exists an integer  $k \in \{1, \dots, n\}$  such that*

$$|\lambda - a_{k,k}| \leq \sum_{j=1, j \neq k}^n |a_{k,j}|.$$

*Proof.* Let  $\lambda$  be an eigenvalue of  $A$  and let  $\mathbf{x}$  be a corresponding eigenvector. Then, for each  $i \in \{1, \dots, n\}$ , we have

$$\sum_{j=1}^n a_{i,j}x_j = \lambda x_i$$

Since  $\mathbf{x}$  is non-zero, there is an integer  $k$  such that  $0 < |x_k| = \max\{|x_i| : i \in \{1, \dots, n\}\}$ . For this  $k$ , we have

$$(\lambda - a_{k,k})x_k = \sum_{j=1, j \neq k}^n a_{k,j}x_j$$

Taking absolute values and applying the triangle inequality:

$$|\lambda - a_{k,k}| |x_k| \leq \sum_{j=1, j \neq k}^n |a_{k,j}| |x_j| \leq \sum_{j=1, j \neq k}^n |a_{k,j}| |x_k|.$$

Therefore,

$$|\lambda - a_{k,k}| \leq \sum_{j=1, j \neq k}^n |a_{k,j}|.$$

□

**Corollary 1.2.** *Let  $A \in \mathbb{C}^{n \times n}$ . Then,*

$$\rho(A) \leq \max \left\{ \sum_{j=1}^n |a_{i,j}| : i \in \{1, \dots, n\} \right\}.$$

## 1.1 Properties of M-Matrices

In this section, we will introduce several properties of M-matrices. Since all Laplacian matrices are symmetric M-matrices, that will be our focus.

**Proposition 1.3.** *Let  $A$  be a symmetric M-matrix. Then, the eigenvalues of  $A$  are non-negative.*

*Proof.* Since  $A$  is an M-matrix, it can be written in the form

$$A = sI - B,$$

where  $s \geq \rho(B)$ . Let  $\lambda$  be an eigenvalue of  $A$  and let  $\mathbf{x}$  be a corresponding eigenvector. Then,  $A\mathbf{x} = \lambda\mathbf{x}$ , which can be re-written as

$$B\mathbf{x} = (s - \lambda)\mathbf{x}.$$

Since  $\lambda$  is real, it follows that  $s - \lambda$  is real. Furthermore,  $s - \lambda \leq \rho(B)$ , which implies that  $\lambda \geq 0$ .  $\square$

The next property (see Corollary 1.5) will be proved using the Perron Frobenius theorem, which we state (in part) below. Note that a matrix  $A$  is *irreducible* if there is no permutation matrix  $P$  such that

$$PAP^T = \begin{bmatrix} B & C \\ 0 & D \end{bmatrix}.$$

For symmetric matrices,  $A$  is irreducible if there is no permutation matrix  $P$  such that

$$PAP^T = \begin{bmatrix} B & 0 \\ 0 & C \end{bmatrix},$$

i.e., the graph corresponding to  $A$  is connected (does not have more than one connected component). Also, note that an eigenvalue  $\lambda$  of the matrix  $A$  is *simple* if there is only one linearly independent eigenvector of  $A$  corresponding to  $\lambda$ .

**Theorem 1.4** (Perron Frobenius). *If  $A \geq 0$  is irreducible, then  $\rho(A)$  is a simple eigenvalue.*

**Corollary 1.5.** *Let  $A$  be a singular, irreducible M-matrix. Then, 0 is a simple eigenvalue of  $A$ .*

*Proof.* Since  $A$  is an M-matrix, it follows that

$$A = sI - B,$$

where  $s \geq \rho(B)$  and  $B \geq 0$ . Furthermore, since  $A$  is irreducible, it follows that  $B \geq 0$  is irreducible. Hence, the Perron Frobenius Theorem implies that  $\rho(B)$  is a simple eigenvalue.

Note that the eigenvalues of  $A$  are of the form  $s - \lambda$ , where  $\lambda$  is an eigenvalue of  $B$ . Since  $A$  is singular, it follows that  $0 = s - \rho(B)$  is an eigenvalue of  $A$ . Hence,  $s = \rho(B)$ . Furthermore, since  $\rho(B)$  is a simple eigenvalue of  $B$ , it follows that 0 is a simple eigenvalue of  $A$ .  $\square$

## 2 Exercises

- I. Let  $G = (V, E)$  be a graph and let  $L$  be the Laplacian matrix of  $G$ . Show that the multiplicity of 0 as an eigenvalue of  $G$  is equal to the number of connected components of  $G$ .