

# Chapter 1

## Preliminaries

Let  $\mathbb{R}$  and  $\mathbb{C}$  denote the field of real numbers and complex numbers respectively.

**Definition 1.1.** Let  $C$  be an  $n \times n$  real matrix, and  $u \in \mathbb{R}^n$  be a nonzero column vector. The scalar  $\lambda \in \mathbb{C}$  is an *eigenvalue* of  $C$  corresponding to the *eigenvector*  $u$ , if  $Cu = \lambda u$ .

**Definition 1.2.** When  $C$  is an  $n \times n$  real matrix, the *spectral radius*  $\rho(C)$  of  $C$  is defined by

$$\rho(C) := \max\{ |\lambda| \mid \lambda \text{ is an eigenvalue of } C \},$$

where  $|\lambda|$  is the magnitude of complex number  $\lambda$ .

We are interested in spectral radius of the following matrix associated with a simple graph.

**Definition 1.3.** Given an undirected graph  $G$ , the *adjacency matrix* of  $G$  is the square matrix  $A = (a_{ij})$  indexed by vertices of  $G$ , and

$$a_{ij} = \begin{cases} 1, & \text{if } i \text{ is adjacent to } j, \\ 0, & \text{otherwise.} \end{cases}$$

**Definition 1.4.** Given an undirected graph  $G$ , the *spectral radius*  $\rho(G)$  of  $G$  is the spectral radius of the adjacency matrix of  $G$ .

We introduce a notation of submatrix, which is taken from some columns and some rows of a matrix.

**Definition 1.5.** For a matrix  $C = (c_{ij})$  and subsets  $\alpha, \beta$  of row indices and column indices of  $C$  respectively, We use  $C[\alpha|\beta]$  to denote the submatrix of  $C$  with size  $|\alpha| \times |\beta|$  that has entries  $c_{ij}$  for  $i \in \alpha$  and  $j \in \beta$ ,

The following theorem is a part of Perron-Frobenius theorem

**Theorem 1.6** (Peron Frobius theorem). *If  $C$  is a nonnegative square matrix, then spectral radius  $\rho(C)$  is an eigenvalue of  $C$  with a corresponding nonnegative right eigenvector and a corresponding nonnegative left eigenvector.*

The following theorem is from %. For completeness, we also provide the proof.

**Theorem 1.7.** *Let  $C = (c_{ij})$ ,  $C' = (c'_{ij})$ ,  $P$  and  $Q$  be  $n \times n$  matrices. Assume that*

- (i)  $PCQ \leq PC'Q$ ;
- (ii) *there exist a nonnegative column vector  $u = (u_1, u_2, \dots, u_n)^T$  and a scalar  $\lambda' \in \mathbb{R}$  such that  $\lambda'$  is an eigenvalue of  $C'$  with associated eigenvector  $Qu$ ;*
- (iii) *there exist a nonnegative row vector  $v^T = (v_1, v_2, \dots, v_n)$  and a scalar  $\lambda \in \mathbb{R}$  such that  $\lambda$  is an eigenvalue of  $C$  with associated left eigenvector  $v^T P$ ; and*
- (iv)  $v^T PQu > 0$ .

*Then  $\lambda \leq \lambda'$ . Moreover,  $\lambda = \lambda'$  if and only if*

$$(PC'Q)_{ij} = (PCQ)_{ij} \quad \text{for } 1 \leq i, j \leq n \text{ with } v_i \neq 0 \text{ and } u_j \neq 0. \quad (1.1)$$

**Proof. 1.8.** Multiplying the nonnegative vector  $u$  in theorem ?? assumption ??, where  $Qu$  is eigenvector of  $C'$ , to the right of both terms of ??,

$$PCQu \leq PC'Qu = \lambda' PQu. \quad (1.2)$$

Multiplying the nonnegative left eigenvector  $v^T$  of  $C$  for  $\lambda$  in assumption ?? to the left of all terms in (??), where  $v^T P$  is left eigenvector of  $C$  for  $\lambda$ , thus we have

$$\lambda v^T PQu = v^T PCQu \leq v^T PC'Qu = \lambda' v^T PQu. \quad (1.3)$$

Now delete the positive term  $v^T PQu$  by assumption ?? to obtain  $\lambda \leq \lambda'$  and finish the proof of the first part. Assume that  $\lambda = \lambda'$ , so the inequality in (??) is an

equality. Especially  $(PCQu)_i = (PC'Qu)_i$  for any  $i$  with  $v_i \neq 0$ . Hence,  $(PCQ)_{ij} = (PC'Q)_{ij}$  for any  $i$  with  $v_i \neq 0$  and any  $j$  with  $u_j \neq 0$ . Conversely, (??) implies

$$v^T PCQu = \sum_{i,j} v_i (PCQ)_{ij} u_j = \sum_{i,j} v_i (PC'Q)_{ij} u_j = v^T PC'Qu,$$

so  $\lambda = \lambda'$  by (??).