Chapter 1

Preliminaries

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

Definition 1.0.1. 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 1.0.2. When C is an $n \times n$ real matrix, the *spectral radius* $\rho(C)$ of C is defined by

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

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

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

Definition 1.0.3. 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 1.0.4. 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 1.0.5.** For a matrix $C = (c_{ij})$ and subsets α , β 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 1.0.6 (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 1.0.7. Let $C = (c_{ij})$, $C' = (c'_{ij})$, P and Q be $n \times n$ matrices. Assume that

(I) $PCQ \leq PC'Q$;

(IV) $v^T PQu > 0$.

- (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 λ' 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 λ is an eigenvalue of C with associated left eigenvector $v^T P$; and

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

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

Proof. 1.0.8. Multiplying the nonnegative vector u in theorem ?? assumption ??, where Qu is eigenvector of C', to the right of both terms of ??,

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

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

$$\lambda v^T P Q u = v^T P C Q u \le v^T P C' Q u = \lambda' v^T P Q u. \tag{1.3}$$

Now delete the positive term $v^T P Q u$ 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 $v_i \neq 0$. 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 (??).