Notes on Matrix Analysis

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Abstract

The note is based on [Tao11].

1 The Geometry of Eigenvalues

1.1 mini-max formula of eigenvalue

Theorem 1.1 (Courant-Fischer min-max theorem). Let A be an $n \times n$ Hermitian matrix. Then we have

$$\lambda_i(A) = \sup_{\dim(V) = i} \bar{f}(V) \stackrel{\text{def}}{=} \sup_{\dim(V) = i} \inf_{v \in V : |v| = 1} v^* A v \tag{1.1}$$

1.2 Partial Trace

Definition 1.2 (Partial Trace). Given an $n \times n$ Hermitian matrix **A** and an m-dimensional subspace V of \mathbb{C}^n , we difine the partial trace

$$\operatorname{tr}(\mathbf{A} \mid_{V}) \stackrel{\text{def}}{=} \sum_{i=1}^{m} v_{i}^{*} \mathbf{A} v_{i}, \tag{1.2}$$

where $v_1, \ldots v_m$ is any orthonormal basis of V.

Theorem 1.3 (Extremal partial trace). Given an $n \times n$ Hermitian matrix **A**. Then for any $1 \le k \le n$, one has

$$\sum_{j=n-k+1}^{n} \lambda_j(\mathbf{A}) \le \operatorname{tr}(\mathbf{A} \mid_V) \le \sum_{j=1}^{k} \lambda_j(\mathbf{A}). \tag{1.3}$$

It has a direct corollary as follows, since $a_{i_1i_1} + \cdots + a_{i_ki_k} = \operatorname{tr}(\mathbf{A} \mid_{\langle e_{i_1}, \dots, e_{i_k} \rangle})$.

Corollary 1.4 (Schur-Horn inequalities).

$$\sum_{j=n-k+1}^{n} \lambda_j(\mathbf{A}) \le a_{i_1 i_1} + \dots + a_{i_k i_k} \le \sum_{j=1}^{k} \lambda_j(\mathbf{A}). \tag{1.4}$$

The result can also be derived by an application of the Birkhoff's theorem for doubly stochastic matrices, implying that $\operatorname{diag}(\mathbf{A})$ lies in the permutahedron of $\lambda_1 \dots \lambda_n$.

Theorem 1.5 (Schur-Horn theorem, 1954). The orbit of diag($\mathbf{U}\mathbf{A}\mathbf{U}^*$) is exactly the permutahedron of $\lambda_1 \dots \lambda_n$.

proof sketch. It suffices to consider the n=2 case.

Since we provide a geometric description for the sum of top-k eigenvalues, we can naturally prove the Ky Fan inequality.

Theorem 1.6 (Ky Fan inequality).

$$\sum_{j=1}^{k} \lambda_j(\mathbf{A} + \mathbf{B}) \le \sum_{j=1}^{k} \lambda_j(\mathbf{A}) + \sum_{j=1}^{k} \lambda_j(\mathbf{B})$$
(1.5)

Proof. Since

$$\operatorname{tr}(\mathbf{A} + \mathbf{B} \mid_{V}) = \operatorname{tr}(\mathbf{A} \mid_{V}) + \operatorname{tr}(\mathbf{B} \mid_{V}) \le \sum_{j=1}^{k} \lambda_{j}(\mathbf{A}) + \sum_{j=1}^{k} \lambda_{j}(\mathbf{B})$$
(1.6)

holds for any subspace V. Let V maximize the left hand side give the desired result.

The extremal partial trace provides a geometric description the sum of top-k eigenvalues. Such description can be generalized to the sum of any k eigenvalues

Theorem 1.7 (Wielandt minimax formula). Let $1 \leq i_1 < \dots i_k \leq n$ be integers. Define a partial flag to be a nested collection $V_1 \subset \dots \subset V_k$ of subspaces of \mathbb{C}^n such that $\dim(V_j) = i_j$ for all $1 \leq j \leq k$. Define the associated Schubert variety $X(V_1, \dots, V_k)$ to be the collection of all k-dimensional subspaces W such that $\dim(W \cap V_j) \leq j$. Then for any $n \times n$ Hermitian matrix \mathbf{A} ,

$$\lambda_{i_1}(\mathbf{A}) + \dots + \lambda_{i_k}(\mathbf{A}) = \sup_{(V_1, \dots, V_k) \in \mathcal{V}_{i_1, \dots, i_k}} \inf_{W \in X(V_1, \dots, V_k)} \operatorname{tr}(\mathbf{A} \mid_W). \tag{1.7}$$

Proof. Part 1: We prove that

$$\lambda_{i_1}(\mathbf{A}) + \dots + \lambda_{i_k}(\mathbf{A}) \ge \inf_{W \in X(V_1, \dots, V_k)} \operatorname{tr}(\mathbf{A} \mid_W)$$
(1.8)

for $\forall (V_1, \ldots, V_k) \in \mathcal{V}_{i_1, \ldots, i_k}$. First we simplify the case to $i_1 = 1$. Second we find $W' \in X(V_2, \ldots, V_k)$ that minimize $\operatorname{tr}(\mathbf{A} \mid_{W'})$ which is at most $\lambda_{i_2}(\mathbf{A}) + \cdots + \lambda_{i_k}(\mathbf{A})$ (using induction on k). Thirdly, we check and find the minimal j such that $\dim(W' \cap V_j) < j$, then we add $v \in \mathcal{P}_{W'^{\perp}}V_j$ into W' to be W, which is the construction we require.

Part 2: Let **A** be diagonalized and $V_j = \langle e_1, \dots, e_{i_j} \rangle$. Now we prove that

$$\operatorname{tr}(\mathbf{A} \mid_{W}) \ge \lambda_{i_1}(\mathbf{A}) + \dots + \lambda_{i_k}(\mathbf{A})$$
 (1.9)

for $\forall W \in X(V_1, \dots, V_k)$. All we have to do is construct an orthogonal basis of W such that $v_j \in W \cap V_j \cap \operatorname{span}(v_1, \dots, v_{j-1})^{\perp}$, that yields 1.9.