

# A GENERALIZED COURANT-FISCHER MINIMAX THEOREM

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## 1. INTRODUCTION

A useful tool for analyzing the spectrum of an Hermitian matrix is the *Courant-Fischer Minimax Theorem* [2].

**Theorem 1.1.** (*Courant-Fischer Minimax Theorem*) *Suppose that  $S \in \mathbb{C}^{n \times n}$  is an Hermitian matrix, then*

$$\lambda_k(S) = \min_{\dim(U)=k} \max_{\substack{x \in U \\ x \neq 0}} \frac{x^* S x}{x^* x}$$

and

$$\lambda_k(S) = \max_{\dim(V)=n-k+1} \min_{\substack{x \in V \\ x \neq 0}} \frac{x^* S x}{x^* x}$$

where  $\lambda_k(S)$  is the  $k$ 'th largest eigenvalue of  $S$ .

The goal of this short communication is to present a generalization of Theorem 1.1, which we refer to as the *Generalized Courant-Fischer Minimax Theorem*. We now state the theorem, and we give a proof in the next section.

**Theorem 1.2.** (*Generalized Courant-Fischer Minimax Theorem*) *Suppose that  $S \in \mathbb{C}^{n \times n}$  is an Hermitian matrix and that  $T \in \mathbb{C}^{n \times n}$  is an Hermitian positive semidefinite matrix such that  $\text{null}(T) \subseteq \text{null}(S)$ . For  $1 \leq k \leq \text{rank}(T)$  we have*

$$\lambda_k(S, T) = \min_{\substack{\dim(U) = k \\ U \perp \text{null}(T)}} \max_{x \in U} \frac{x^* S x}{x^* T x}$$

and

$$\lambda_k(S, T) = \max_{\substack{\dim(V) = \text{rank}(T) - k + 1 \\ V \perp \text{null}(T)}} \min_{x \in V} \frac{x^* S x}{x^* T x}.$$

## 2. PROOF

We begin by stating and proving a generalization of the Courant-Fischer Theorem for pencils of Hermitian positive definite matrices.

**Theorem 2.1.** *Let  $S, T \in \mathbb{C}^{n \times n}$  be Hermitian matrices. If  $T$  is also positive definite then*

$$\lambda_k(S, T) = \min_{\dim(U)=k} \max_{x \in U} \frac{x^* S x}{x^* T x}$$

and

$$\lambda_k(S, T) = \max_{\dim(V)=n-k+1} \min_{x \in V} \frac{x^* S x}{x^* T x}.$$

*Proof.* Let  $T = L^* L$  be the Cholesky factorization of  $T$ . Let  $U$  be some  $k$ -dimensional subspace of  $\mathbb{C}^n$ , let  $x \in U$ , and let  $y = Lx$ . Since  $T$  is Hermitian positive definite (hence nonsingular), the subspace  $W = \{Lx : x \in U\}$  has dimension  $k$ . Similarly, for any  $k$ -dimensional subspace  $W$ , the subspace  $U = \{L^{-1}x : x \in W\}$  has dimension  $k$ . We have

$$\frac{x^* S x}{x^* T x} = \frac{x^* L^* L^{-*} S L^{-1} L x}{x^* L^* L x} = \frac{y^* L^{-*} S L^{-1} y}{y^* y}.$$

By applying the Courant-Fischer to  $L^{-*} S L^{-1}$ , we obtain

$$\begin{aligned} \lambda_k(L^{-*} S L^{-1}) &= \min_{\dim(W)=k} \max_{y \in W} \frac{y^* L^{-*} S L^{-1} y}{y^* y} \\ &= \min_{\dim(U)=k} \max_{x \in U} \frac{x^* S x}{x^* T x}. \end{aligned}$$

The generalized eigenvalues of  $(S, T)$  are exactly the eigenvalues of  $L^{-*} S L^{-1}$  so the first equality of the theorem follows. The second equality can be proved using a similar argument.  $\square$

Before proving the generalized version of the Courant-Fischer Minimax Theorem we show how to convert an Hermitian positive semidefinite problem to an Hermitian positive definite problem.

**Lemma 2.2.** *Let  $S, T \in \mathbb{C}^{n \times n}$  be Hermitian matrices. Assume that  $T$  is also a positive semidefinite and that  $\text{null}(T) \subseteq \text{null}(S)$ . For any  $Z \in \mathbb{C}^{n \times \text{rank}(T)}$  with  $\text{range}(Z) = \text{range}(T)$ , the determined generalized eigenvalues of  $(S, T)$  are exactly the generalized eigenvalues of  $(Z^* S Z, Z^* T Z)$ .*

*Proof.* We first show that  $Z^* T Z$  has full rank. Suppose that  $Z^* T Z v = 0$ . We have  $T Z v \in \text{null}(Z^*)$ . Therefore,  $T Z v \perp \text{range}(Z) = \text{range}(T)$ . Obviously  $T Z v \in \text{range}(T)$ , so we must have  $v = 0$ . Since  $\text{null}(Z^* T Z) = \{0\}$ , the matrix  $Z^* T Z$  has full rank.

Suppose that  $\lambda$  is a determined eigenvalue of  $(S, T)$ . We will show that it is a determined eigenvalue of  $(Z^* S Z, Z^* T Z)$ . The pencil  $(Z^* S Z, Z^* T Z)$  has exactly  $\text{rank}(Z^* T Z)$  determined eigenvalues. We will show that  $Z^* T Z$  is full rank, so the pencil  $(Z^* S Z, Z^* T Z)$  has exactly  $\text{rank}(T)$  eigenvalues. Since the pencil  $(S, T)$  has exactly  $\text{rank}(T)$  determined eigenvalues, each of them an eigenvalue of  $(Z^* S Z, Z^* T Z)$ , this will conclude the proof.

Now let  $\mu$  be an eigenvalue of  $(Z^* S Z, Z^* T Z)$ . It must be determined, since  $Z^* T Z$  has full rank. Let  $y$  be the corresponding eigenvector,  $Z^* S Z y = \mu Z^* T Z y$ , and let  $x = Z y$ . Now there are two cases. If  $\mu = 0$ , then  $S Z y = S x = 0$  (since  $Z^*$  has full rank and at least as many columns as rows). The vector  $x$  is in  $\text{range}(Z) = \text{range}(T)$ ,  $T x \neq 0$ . This implies that  $\mu = 0$  is also a determined eigenvalue of  $(S, T)$ .

If  $\mu \neq 0$ , the analysis is a bit more difficult. Clearly,  $TZy \in \text{range}(T) = \text{range}(Z)$ . But  $\text{range}(Z) = \text{range}(Z^{*+})$  [1, Proposition 6.1.6.vii], so  $Z^{*+}Z^*TZy = TZy$  [1, Proposition 6.1.7]. We claim that  $SZy \in \text{range}(Z)$ . If it is not, it  $Zy$  must be in  $\text{null}(T) \subseteq \text{null}(S)$ , but  $\mu$  would have to be zero. Therefore, we also have  $Z^{*+}Z^*SZy = SZy$ , so by multiplying  $Z^*SZy = \mu Z^*TZy$  by  $Z^{*+}$  we see that  $\mu$  is an eigenvalue of  $(S, T)$ .  $\square$

We are now ready to prove Theorem 1.2, the generalization of the Courant-Fischer Minimax Theorem. The technique is simple: we use Lemma 2.2 to reduce the problem to a smaller-sized full-rank problem, apply Theorem 2.1 to characterize the determined eigenvalues in terms of subspaces, and finally show a complete correspondence between the subspaces used in the reduced pencil and subspaces used in the original pencil.

*Proof. (Theorem 1.2)* Let  $Z \in \mathbb{C}^{n \times \text{rank}(T)}$  have  $\text{range}(Z) = \text{range}(T)$ . We have

$$\lambda_k(S, T) = \lambda_k(Z^*SZ, Z^*TZ) = \min_{\dim(W) = k} \max_{x \in W} \frac{x^*Z^*SZx}{x^*Z^*TZx}$$

and

$$\lambda_k(S, T) = \lambda_k(Z^*SZ, Z^*TZ) = \max_{\dim(W) = \text{rank}(T) - k + 1} \min_{x \in W} \frac{x^*Z^*TZx}{x^*Z^*SZx}.$$

The leftmost equality in each of these equations follows from Lemma 2.2 and the rightmost one follows from Theorem 2.1.

We now show that for every  $k$ -dimensional subspace  $U \subseteq \mathbb{C}^n$  with  $U \perp \text{null}(T)$ , there exists a  $k$ -dimensional subspace  $W \subseteq \mathbb{C}^{\text{rank}(T)}$  such that

$$\left\{ \frac{x^*Sx}{x^*Tx} : x \in U \right\} = \left\{ \frac{y^*Z^*SZy}{y^*Z^*TZy} : y \in W \right\},$$

and vice versa. The validity of this claim establishes the min-max side of the theorem.

We first need to show that  $k \leq \text{rank}(T)$ . This is true because every vector in  $U$  is in  $\text{range}(T)$ , so its dimension must be at most  $\text{rank}(T)$ .

Define  $W = \{y \in \mathbb{C}^{\text{rank}(T)} : Zy \in U\}$ . Let  $b_1, \dots, b_k$  be a basis for  $U$ . Because  $U \perp \text{null}(T)$ ,  $b_j \in \text{range}(T)$ , so there is a  $y_j$  such that  $Zy_j = b_j$ . Therefore, dimension of  $W$  is at most  $k$ . Now let the vectors  $y_i$ 's be a basis of  $W$  and define  $b_i = Zy_i$ . The  $b_i$ 's span  $U$ , so there are at most  $k$  of them, so the dimension of  $W$  is at least  $k$ . Therefore, it is exactly  $k$ .

Every  $x \in U$  is orthogonal to  $\text{null}(T)$ , so it must be in  $\text{range}(T)$ . There exist a  $y \in \mathbb{C}^{\text{rank}(T)}$  such that  $Zy = x$ . So we have  $x^*Sx/x^*Tx = y^*Z^*SZy/y^*Z^*TZy$ . Combining with the fact that  $y \in W$ , we have shown inclusion of one side. Now suppose  $y \in W$ . Define  $x = Zy \in U$ . Again we have  $x^*Sx/x^*Tx = y^*Z^*SZy/y^*Z^*TZy$ , which shows the other inclusion.

Now we will show that for every  $k$ -dimensional subspace  $W$  there is a subspace  $U$  that satisfies the claim. Define  $U = \{Zy : y \in W\}$ . Because  $Z$  has full rank,  $\dim(U) = k$ . Also,  $U \subseteq \text{range}(Z) = \text{range}(T)$  so  $U \perp \text{null}(T)$ . The equality of the Raleigh-quotient sets follows from taking  $y \in W$  and  $x = Zy \in U$  or vice versa.

## REFERENCES

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