A NOTE ON SIMULTANEOUSLY DIAGONALIZABLE MATRICES

ABRAHAM BERMAN¹ AND ROBERT J. PLEMMONS²

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Abstract. Consider the functional $f(U) = \sum_{i=1}^{n} \max_{j} \{ (U^T M_j U)_{ii} \}$ over orthogonal matrices U, where the collection of n-by-n symmetric matrices M_j are pairwise commutative, and thus simultaneously diagonalizable. Selecting an orthogonal matrix U which maximizes f(U) has applications in adaptive optics. A proof is given here that any orthogonal matrix Q which simultaneously diagonalizes the M_j maximizes the function f.

1. Introduction

Recently, there has been a growing interest in adaptive optics, i.e., methods to overcome the effects of distortion in imaging through a medium, such as the atmosphere [6, 4, 7]. In adaptive optics, the following optimization problem arises [4]. The problem involves maximizing the functional

$$f(U) = \sum_{i=1}^{n} \max_{j} \{ (U^{T} M_{j} U)_{ii} \}$$
(1)

over all *n*-by-*n* orthogonal matrices *U*. In this note we show that if the matrices M_j are simultaneously diagonalizable, then any orthogonal matrix *Q* which simultaneously diagonalizes the M_j maximizes *f*.

2. An Optimization Result

Recall that a set of real symmetric matrices can be simultaneously diagonalized by an orthogonal matrix if and only if they are pairwise commutative, e.g., [5], Theorems 1.3.19 and 2.3.3. For such matrices we state and prove our main result.

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THEOREM. Suppose $\{M_j\}$, $1 \leq j \leq k$, is a collection of pairwise commuting matrices. Let Q be any orthogonal matrix which simultaneously diagonalizes the M_j . Then Q maximizes the functional given in (1).

Proof. Let $B_j = U^T M_j U$, where U is an arbitrary orthogonal matrix. We can rewrite B_j using the orthogonal matrix $V = Q^T U$:

$$B_j = U^T M_j U = (QV)^T M_j (QV) = V^T Q^T M_j QV = V^T D_j V ,$$

where $D_j = Q^T M_j Q$ is the diagonalization of M_j using Q. Observe that

$$f(U) = \sum_{i=1}^{n} \max_{j} \{ (B_j)_{ii} \}$$

and that

$$f(Q) = \sum_{i=1}^{n} \max_{j} \{ (D_j)_{ii} \}.$$

We have to show that $f(Q) \ge f(U)$. To do this we define matrices $D = \text{diag}(d_{ss})$ and $B = (b_{st})$ as follows: $d_{ss} = \max_j(D_j)_{ss}$ and $b_{st} = \max_j(B_j)_{st}$. For example, d_{11} is the largest element of $\{(D_1)_{11}, (D_2)_{11}, \dots, (D_k)_{11}\}$. In other words,

$$D = \max_{1 \leq j \leq k} \{D_j\}$$
, and $B = \max_{1 \leq j \leq k} \{B_j\}$,

where the maxima are evaluated *elementwise*. We will use \tilde{d} , \tilde{d}_j , \tilde{b} , and \tilde{b}_j to represent the column vectors of the diagonals of D, D_j , B, and B_j , respectively. For example, \tilde{b}_1 is the column vector whose entries are the diagonal elements of B_1 . Finally, we let ebe a column vector of ones. With this notation, the traces of D and B can be written in the form $e^T \tilde{d}$ and $e^T \tilde{b}$. So, proving the theorem is equivalent to establishing the inequality

$$e^T \tilde{d} \geqslant e^T \tilde{b} . \tag{2}$$

Let $S = V \circ V$, the Hadamard product of V with itself; namely,

$$s_{ij} = v_{ij}^2$$

A matrix which can be expressed as $S = V \circ V$, where V is orthogonal is called orthostochastic. Every orthostochastic matrix is doubly stochastic, i. e., $s_{ij} \ge 0$ and $e^T S = e^T S^T = e^T$, e. g., [1].

The doubly stochastic matrix S^T can be used to relate \tilde{b}_j and \tilde{d}_j :

$$\tilde{b}_j = (V \circ V)^T \tilde{d}_j = S^T \tilde{d}_j, \quad 1 \le j \le k .$$
(3)

By definition, $\tilde{d} \ge \tilde{d}_i$ for all *j*. Since S^T is nonnegative, we can write

$$S^T \tilde{d} \ge S^T \tilde{d}_j = \tilde{b}_j, \quad 1 \le j \le k.$$
 (4)

This inequality holds elementwise; thus, we have

$$\left[S^{T}\tilde{d}\right]_{i} \geqslant \left[\tilde{b}_{j}\right]_{i}$$

for all values of *i*. Therefore,

$$S^T \tilde{d} \ge \max_i \tilde{b}_i = \tilde{b}$$
 (5)

Here again, the maximum is taken elementwise. After multiplying both sides of inequality (5) by e^T and noting that $e^T S^T = e^T$, we obtain inequality (2), thereby proving the theorem.

3. Remarks

The note is concluded with some remarks on the theorem and on the computational aspects of the problem.

• The converse of the theorem is not true; i.e., an orthogonal matrix Q can maximize f(U) without simultaneously diagonalizing the pairwise commuting M_j . In fact, let M_1, \dots, M_k be *n*-by-*n* diagonal matrices such that $M_1 \ge M_j$ for all *j*. Then for all *n*-by-*n* orthogonal matrices U, $f(U) = \text{trace}(M_1)$. But, in the case where one of the matrices has distinct diagonal entries, the only matrices that simultaneously diagonalize the (diagonal matrices) M_j are products of permutation and signature matrices.

• Although the theorem establishes the existence of an orthogonal matrix Q that maximizes the functional f(U) where the M_j are pairwise commutative, the process of computing the simultaneous diagonalizer Q can be quite nontrivial. A comprehensive study of numerical methods for simultaneous diagonalization of a single pair of commuting matrices is given in [2].

• The work on adaptive optics that motivated our discussion concerns real matrices. A similar theorem can be stated for Hermitian matrices M_j and a unitary matrix Q. Other recent applications of simultaneous diagonalization and related algorithms can be found in multivariate statistics [3].

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Abraham Berman Department of Mathematics Technion–Israel Institute of Technology Haifa 32000 Israel e-mail: berman@tx.technion.ac.il

Robert J. Plemmons Department of Computer Science Wake Forest University Winston-Salem, NC 27109 e-mail: plemmons@mthcsc.wfu.edu URL: http://www.mthcsc.wfu.edu/~plemmons/

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