On approximate diagonalization of third order symmetric tensors by orthogonal transformations

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Abstract

In this paper, we study the approximate orthogonal diagonalization problem of third order symmetric tensors. We define several classes of approximately diagonal tensors, including the ones corresponding to the stationary points of this problem. We study the relationships between these classes, and other well-known objects, such as tensor Z-eigenvalue and Z-eigenvector. We also prove results on convergence of the cyclic Jacobi (or Jacobi CoM2) algorithm.

Keywords: symmetric tensors; orthogonally decomposable tensors; approximate tensor diagonalization; Jacobi-type algorithms; maximally diagonal tensors

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1. Introduction

Arrays with more than two indices have become more and more important in the last two decades because of their usefulness in various fields, including signal processing, numerical linear algebra and data analysis \cite{1, 2, 3, 4, 5}. Admitting a common abuse of language, we shall refer to them as tensors, being understood that we are considering the associated multilinear forms (and hence fully contravariant tensors) \cite{2}. Real symmetric matrices can be diagonalized by orthogonal transformations, which is a key property leading to the spectral decomposition. On the other hand, the orthogonal diagonalization of symmetric tensors has also been addressed, as an exact decomposition in \cite{6, 7, 8}, or as a low-rank approximation in \cite{9, 10}. In fact, the approximate orthogonal diagonalization of third and fourth order cumulant tensors is in the core of Independent Component Analysis \cite{9, 10, 11}, and finds many applications \cite{3}. However, the latter problem is much more difficult than the spectral decomposition of symmetric matrices since it is well known that not every symmetric tensor can be diagonalized by orthogonal transformations \cite{6, 7}.

Notation. Let $\mathbb{R}^{m \times n \times p}$ defined $\mathbb{R}^m \otimes \mathbb{R}^n \otimes \mathbb{R}^p$ be the linear space of third order real tensors and $S_n \subseteq \mathbb{R}^{n \times n \times n}$ be the set of symmetric ones, whose entries do not change.

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under any permutation of indices $[12, 13]$. Let $O_n \subseteq \mathbb{R}^{n \times n}$ be the orthogonal group. Let $SO_n \subseteq \mathbb{R}^{n \times n}$ be the special orthogonal group, that is, the set of orthogonal matrices with determinant 1. We denote by $\| \cdot \|$ the Frobenius norm of a tensor or a matrix, or the Euclidean norm of a vector. Tensor arrays, matrices, and vectors, will be respectively denoted by bold calligraphic letters, e.g. $\mathcal{A}$, with bold uppercase letters, e.g. $M$, and with bold lowercase letters, e.g. $u$; corresponding entries will be denoted by $A_{ijk}$, $M_{ij}$, and $u_i$. Operator $\bullet_p$ denotes contraction on the $p$th index of a tensor; when contracted with a matrix, it is understood that summation is always performed on the second index of the matrix. For instance, $[\mathcal{A} \bullet_1 M]_{ijk} = \sum_\ell A_{\ell jk} M_{i \ell}$. When contraction of a symmetric tensor is performed on vectors, the subscript $p$ can be omitted. For $\mathcal{A} \in S_n$ and a fixed set of indices $\{i, j\}$, $1 \leq i < j \leq n$, we denote by $\mathcal{A}^{(i,j)}$ the 2-dimensional subtensor obtained from $\mathcal{A}$ by allowing its indices to vary in $\{i, j\}$ only. Similarly for $1 \leq i < j < k \leq n$, we denote by $\mathcal{A}^{(i,j,k)}$ the 3-dimensional subtensor obtained by allowing indices of $\mathcal{A}$ to vary in $\{i, j, k\}$ only. The identity matrix of size $n$ is denoted by $I_n$, and its columns by $e_i$, $1 \leq i \leq n$, which form the canonical orthonormal basis.

Contribution. We formulate the approximate orthogonal symmetric tensor diagonalization problem as the maximization of diagonal terms $[14]$. More precisely, let $\mathcal{A} \in S_n$, $Q \in SO_n$, and $W = \mathcal{A} \bullet_1 Q^T \bullet_2 Q^T \bullet_3 Q^T$. This problem is to find

$$Q_\ast = \text{argmax}_{Q \in SO_n} f(Q),$$

where

$$f(Q) \overset{\text{def}}{=} \| \text{diag}\{W\} \|^2 = \sum_{i=1}^{n} W_{iii}^2.$$

Methods based on Jacobi rotations (e.g., the well-known Jacobi CoM2 algorithm [9, 10, 11]) are widely used in practice [3, 15] to solve problem (1). These methods aim at making a symmetric tensor as diagonal as possible by successive Jacobi rotations. They are particularly attractive due to the low computational cost of iterations. Other popular methods include Riemannian optimization methods [16] that alternate between descent steps and retractions. The above methods are typically known to converge (globally or locally) to stationary points [16, 17], though the convergence of the original Jacobi CoM2 method has not been studied.

The main goal of this paper is to quantify the notion of approximate diagonality, by introducing several classes of approximately diagonal tensors and studying the relationships between them. These classes include stationary diagonal tensors, Jacobi diagonal tensors, locally maximally diagonal tensors, maximally diagonal tensors, generally maximally diagonal tensors and pseudo diagonal tensors. We characterize (i) the class of Jacobi diagonal tensors by the stationary diagonal ratio, and (ii) the orbit of pseudo diagonal tensors by Z-eigenvalue and Z-eigenvectors. Moreover, we study (iii) the class of locally maximally diagonal tensors based on Riemannian Hessian. We show that this class is not equal to the class of Jacobi diagonal tensors, and thus Jacobi-type algorithms may converge to a saddle point of (2). We also study (iv) whether a symmetric tensor is maximally diagonal if and only if it is generally maximally diagonal. Several problems related to low rank orthogonal approximation are proved to be equivalent to the fact

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that these two classes are equal when the dimension is greater than 2. We present a
counterexample to these equivalent problems based on the decomposition of orthogonal
matrices. Moreover, we prove a result that can be seen as an orthogonal analogue of
the so-called Comon’s Conjecture \cite{18}. The second goal of this paper is to study the
convergence properties of the original Jacobi CoM2 algorithm \cite{11}.

Organization. The paper is organized as follows. In section 2 we recall basic
properties of the cost function, introduce notation for derivatives, and present the scheme
of Jacobi-type algorithms. In section 3 we define the classes of approximately diagonal
tensors, which are considered in this paper. Some basic relationships between these
classes are shown. The stationary diagonal ratio is introduced, and the orbit of pseudo
diagonal tensors is studied. In section 4 we study the class of locally maximally diagonal
tensors using Riemannian Hessian. In section 5 we study the relationship between
maximally diagonal tensors and generally maximally diagonal tensors. Section 6 contains
results on convergence of the Jacobi CoM2 algorithm. Finally, Appendix A contains
the remaining proofs.

2. Optimization problem: properties and algorithms

2.1. Riemannian gradient and stationary points

First, we recall that the Riemannian gradient of (2) \cite[§4.1]{17}, is, by definition,
\[
\text{Proj} \nabla f(Q) = Q\Lambda(Q),
\]
where \(\Lambda(Q)\) is the matrix with entries
\[
\Lambda(Q)_{k,l} = 3(W_{lll}W_{lkk} - W_{lkk}W_{kkk}).
\]
The matrix \(Q\) is a stationary point of (2) if and only if \(\text{Proj} \nabla f(Q) = 0\). A local
maximum point of (2), of course, is a stationary point. A reasonable local optimization
algorithm should at least converge to a stationary point.

2.2. Elementary rotations and Jacobi-type algorithms

Let \((i, j)\) be a pair of indices with \(1 \leq i < j \leq n\). We denote the Givens rotation (by
an angle \(\theta \in \mathbb{R}\)) matrix to be
\[
G^{(i,j,\theta)} = \begin{bmatrix}
1 & & & \\
& \ddots & & \\
& & \cos \theta & -\sin \theta \\
& & \sin \theta & \cos \theta \\
& & & \ddots \\
0 & & & 1
\end{bmatrix},
\]
i.e., the matrix defined by

\[
(G^{(i,j,\theta)})_{k,l} = \begin{cases} 
1, & k = l, k \notin \{i, j\}, \\
\cos \theta, & k = l, k \in \{i, j\}, \\
\sin \theta, & (k, l) = (j, i), \\
-\sin \theta, & (k, l) = (i, j), \\
0, & \text{otherwise}
\end{cases}
\]

for \(1 \leq k, l \leq n\).

Jacobi-type algorithms proceed by successive optimization of the cost function with respect to elementary rotations, summarized in the following scheme.

**Algorithm 1.** Input: \(A \in S_n\) and \(Q_0 = I_n\). Output: a sequence of iterations \(\{Q_k : k \in \mathbb{N}\}\).

- For \(k = 1, 2, \ldots \) until a stopping criterion is satisfied do:
  - Choose the pair \((i_k, j_k)\) according to a certain pair selection rule.
  - Compute the angle \(\theta_k^*\) that maximizes the function
    \[
    h_k(\theta) \overset{\text{def}}{=} f(Q_{k-1}G^{(i_k,j_k,\theta)}).
    \] (5)
  - Update \(Q_k = Q_{k-1}G^{(i_k,j_k,\theta_k^*)}\).

- End for

The algorithm is similar in spirit to block-coordinate descent. Important differences are: the coordinate system is changing at every iteration, and, for each elementary rotation, the global maximum is achieved. Recently, local and global convergence to stationary points \([19, 17]\) has been established for variants of Algorithm 1. Apart from Jacobi-type algorithms, Jacobi rotations are also very useful in computing the fast re-tractions \([16, \text{p. 58}]\) in Riemannian optimization methods \([16]\).

### 2.3. Directional derivatives

We introduce some useful notation that will be used throughout the paper.

**Definition 2.1.** Let \(A \in S_n\) and \(1 \leq i < j \leq n\). Define

\[
d_{i,j}(A) \overset{\text{def}}{=} A_{iii}A_{iij} - A_{ijj}A_{jjj},
\]

\[
\omega_{i,j}(A) \overset{\text{def}}{=} A_{ii}^2 + A_{jj}^2 - 3A_{ij}^2 - 3A_{ji}^2 - 2A_{iii}A_{ijj} - 2A_{iij}A_{jjj}.
\]

In order to simplify notation, we denote functions (5) with \(Q_{k-1} = I_n\) as

\[
\hat{h}_{i,j}(\theta) \overset{\text{def}}{=} \| \text{diag}\{A \bullet (G^{(i,j,\theta)}_1)^T \bullet (G^{(i,j,\theta)}_2)^T \bullet (G^{(i,j,\theta)}_3)^T \} \|^2
\]

for \(1 \leq i < j \leq n\). Then it holds that \([17, \text{Lemma 5.7}]\)

\[
\hat{h}_{i,j}^\prime(0) = 6d_{i,j}(A) \quad \text{and} \quad \hat{h}_{i,j}^\prime(0) = -6\omega_{i,j}(A). \quad (6)
\]
3. Classes of approximately diagonal tensors

3.1. Definitions of classes

In this subsection, we define several classes of third order symmetric tensors. Some of them are related to the stationary points of (2) or the points where Algorithm 1 may stop. For simplification, we look at the derivatives of (2) at \( Q = I_n \).

**Definition 3.1.** (i) Let \( \mathbf{A}, \mathbf{B} \in S_n \). Then \( \mathbf{A} \) is orthogonally similar \([13, 20]\) to \( \mathbf{B} \) if there exists \( Q \in O_n \) such that
\[
\mathbf{B} = \mathbf{A} \cdot Q \cdot Q \cdot Q.
\]

(ii) Let \( C \subseteq S_n \) be a subset. Define the orbit\(^1\) of \( C \) to be:
\[
O(C) \overset{\text{def}}{=} \{ \mathbf{A} \cdot Q \cdot Q \cdot Q, \mathbf{A} \in C, Q \in O_n \}.
\]

**Definition 3.2.** We denote by \( D_n \) the set of diagonal tensors in \( S_n \), and \( O(D_n) \) the set of orthogonally decomposable tensors (referred to as “odeco” in [6]). More precisely, any \( \mathbf{A} \in O(D_n) \) can be decomposed as
\[
\mathbf{A} = \sum_{k=1}^{n} \lambda_k \mathbf{u}_k \otimes \mathbf{u}_k \otimes \mathbf{u}_k
\]
where \( \lambda_k \in \mathbb{R} \) and \( \mathbf{u}_1, \cdots, \mathbf{u}_n \in \mathbb{R}^n \) form an orthonormal basis.

**Definition 3.3.** Let \( \mathbf{A} \in S_n \). The class of pseudo diagonal tensors is defined to be
\[
PD_n \overset{\text{def}}{=} \{ \mathbf{A} : A_{ij} = A_{ji} = 0, \text{ for any } 1 \leq i < j \leq n \}.
\]

**Remark 3.4.** It is clear that \( D_n \subseteq PD_n \) and \( O(D_n) \subseteq O(PD_n) \).

In section 3.4, we will give characterizations of \( PD_n \) and \( O(PD_n) \) from the perspective of tensor spectral theory. Besides, it is well known that \( O(D_n) \nsubseteq S_n \), that is, not every symmetric tensor can be diagonalized by orthogonal transformations \([6, 7]\).

**Definition 3.5.** Let \( \mathbf{A} \in S_n \).

(i) The class of stationary diagonal tensors is defined to be
\[
SD_n \overset{\text{def}}{=} \{ \mathbf{A} : d_{ij}(\mathbf{A}) = 0, \text{ for any } 1 \leq i < j \leq n \}.
\]

(ii) The class of Jacobi diagonal tensors is defined to be
\[
JD_n \overset{\text{def}}{=} \{ \mathbf{A} : 0 \in \text{argmax} \overline{h}_{ij}(\theta), \text{ for any } 1 \leq i < j \leq n \}.
\]

(iii) The class of locally Jacobi diagonal tensors is defined to be
\[
LJD_n \overset{\text{def}}{=} \{ \mathbf{A} : 0 \text{ is a local maximum point of } \overline{h}_{ij}(\theta), \text{ for any } 1 \leq i < j \leq n \}.
\]

\(^1\)Classically, the notion of orbit is defined for a single element (e.g., \( C \in S_n \)). In this paper, we use the word “orbit” as a shorthand for saying “the action of \( O_n \) on \( C \).”
Remark 3.6. From (4), it follows that $A \in SD_n$ if and only if $\text{Proj} \nabla f(I_n) = 0$ in (3). In other words, $A \in SD_n$ if and only if $I_n$ is a stationary point of (2). Moreover, it can be seen that Algorithm 1 stops at $A$ if $A \in JD_n$. This is the reason why we call the tensors in $SD_n$ and $JD_n$ stationary diagonal and Jacobi diagonal respectively.

Lemma 3.7. Let $A \in S_n$. The following are equivalent.

(i) $A \in JD_n$.

(ii) $A \in LJD_n$.

(iii) $d_{i,j}(A) = 0$ and $\omega_{i,j}(A) \geq 0$ for any $1 \leq i < j \leq n$.

Proof. (i)$\Rightarrow$(ii) is clear. (ii)$\Rightarrow$(iii) follows from (6). Let us prove (iii)$\Rightarrow$(i). We have

$$\bar{h}_{i,j}(\theta) - \bar{h}_{i,j}(0) = \frac{3}{(1 + x^2)^2} (2d_{i,j}(A)(x - x^3) - \omega_{i,j}(A)x^2)$$

for $x = \tan(\theta)$, any $1 \leq i < j \leq n$ by [17, Eq. (22)] (see also [15]). Note that $\bar{h}_{i,j}(\theta) - \bar{h}_{i,j}(0) \equiv 0$ if $d_{i,j}(A) = \omega_{i,j}(A) = 0$. If $d_{i,j}(A) = 0$ and $\omega_{i,j}(A) \geq 0$, then $\bar{h}_{i,j}(\theta)$ reaches its maximum value at $\theta = 0$, by (7). It follows that $A \in JD_n$.

Definition 3.8. Let $A \in S_n$ and $f$ be as in (2).

(i) The class of maximally diagonal tensors is defined to be

$$MD_n \overset{\text{def}}{=} \{ A : I_n \in \text{argmax} f(Q) \}.$$ 

(ii) The class of locally maximally diagonal tensors is defined to be

$$LMD_n \overset{\text{def}}{=} \{ A : I_n \text{ is a local maximum point of } f(Q) \}.$$ 

(iii) The class of generally maximally diagonal tensors is defined to be

$$GMD_n \overset{\text{def}}{=} \{ A : (I_n, I_{n}, I_{n}) \in \text{argmax} F(P, Q, R) \},$$

where

$$F(P, Q, R) \overset{\text{def}}{=} \| \text{diag}\{ A_{1} P^T_{1} Q_{2}^T R_{1}^T \} \|^2.$$ (8)

Remark 3.9. Note that $O_n \subseteq \mathbb{R}^{n \times n}$ is a compact submanifold and (2) is continuous. Since (2) takes the same maximum on $O_n$ and $SO_n$, we get that $O(MD_n) = S_n$. Note that $MD_n \subseteq LMD_n$. It follows that $O(LMD_n) = S_n$. In other words, for any $A \in S_n$, there exist $Q_*$ and $Q_{**}$ in $SO_n$ such that

$$A_{1} Q_{2} \cdots Q_{n} \in LMD_n \quad \text{and} \quad A_{1} Q_{**} \cdots Q_{**} \in MD_n,$$

respectively. How to find $Q_*$ or $Q_{**}$ is the goal of problem (1).
3.2. Basic relationships

The tensor classes defined in section 3.1 have the following relationships. The first row and column denote the corresponding orbits, i.e., arrows stand for the action of $O_n$.

\[
\begin{array}{cccc}
O(D_n) & \subseteq & O(PD_n) & \subseteq \leq S_n \\
\cap & \subseteq & PD_n & \subseteq \leq JD_n \\
GMD_n & \subseteq & \leq SD_n & \subseteq \leq S_n \\
\cap & \subseteq & MD_n & \subseteq \leq \mathcal{MD}_n & \subseteq \leq J\mathcal{D}_n
\end{array}
\]

Remark 3.10. Most of the above relationships are easy to get by Definition 3.5 and Definition 3.8. We only derive some of them for $S_2$, which are not obvious.

(i) Note that $SO_2$ coincides with the set of Givens rotations. We see that $MD_2 = LMD_2 = JD_2 = LJD_2$.

(ii) It will be shown that $GMD_2 = MD_2$ in Theorem 5.3. It follows that $GMD_2 = MD_2 = LMD_2 = JD_2 = LJD_2$.

(iii) $PD_n$ and $JD_n$ will be characterized in Remark 3.12 and Theorem 3.13. It follows by these characterizations that $PD_2 \not\subseteq JD_2$.

(iv) By Theorem 3.13, we see that $JD_2 \not\subseteq SD_2$.

(v) Note that $D_2 = PD_2$. We have that $O(PD_2) \not\subseteq S_2$ by Remark 3.4.

3.3. Stationary diagonal ratio

In this subsection, we define the stationary diagonal ratio for the tensors in $SD_n$, which can be used to characterize $JD$ and $PD_n$.

Definition 3.11. Let $A \in SD_n$ and $1 \leq i < j \leq n$. The stationary diagonal ratio, denoted by $\gamma_{ij}$, is defined as follows.

\[
\gamma_{ij} \equiv \begin{cases} 0, & \text{if } A^{(i, j)} = 0; \\
\infty, & \text{if } A_{iii} = A_{jjj} = 0 \quad \text{and} \quad A_{iij}^2 + A_{ijj}^2 \neq 0; \\
\text{otherwise, } \gamma_{ij} \text{ is the (unique) number such that } & \\
(A_{ijj}) = \gamma_{ij} (A_{iii}) (A_{ijj}).
\end{cases}
\]
Remark 3.12. Let \( \mathcal{A} \in \mathcal{S} \mathcal{D}_n \). Then \( \mathcal{A} \in \mathcal{PD}_n \) if and only if \( \gamma_{ij} = 0 \) for any \( 1 \leq i < j \leq n \).

Theorem 3.13. Let \( \mathcal{A} \in \mathcal{S} \mathcal{D}_n \). Then \( \mathcal{A} \in \mathcal{J} \mathcal{D}_n \) if and only if \( \gamma_{ij} \in [-1, 1/3] \) for any \( 1 \leq i < j \leq n \).

Proof. Note that \( \mathcal{A} \in \mathcal{J} \mathcal{D}_n \) if and only if \( d_{i,j}(\mathcal{A}) = 0 \) and \( \omega_{i,j}(\mathcal{A}) \geq 0 \) for any \( 1 \leq i < j \leq n \) by Lemma 3.7. We only need to show that \( \omega_{i,j}(\mathcal{A}) \geq 0 \) if and only if \( \gamma_{ij} \in [-1, 1/3] \). If \( \gamma_{ij} = \infty \), then \( \omega_{i,j}(\mathcal{A}) < 0 \). If \( \gamma_{ij} < \infty \), by Definition 3.11, we have that

\[
-\omega_{i,j}(\mathcal{A}) = (3\gamma_{ij}^2 + 2\gamma_{ij} - 1)(A_{ii}^2 + A_{jj}^2).
\]

It follows that \( \omega_{i,j}(\mathcal{A}) \geq 0 \) if and only if \( \gamma_{ij} \in [-1, 1/3] \).

3.4. Orbit of the pseudo diagonal tensors
3.4.1. Characterization

In this subsection, we characterize the orbit of pseudo diagonal tensors based on the Z-eigenvalue and Z-eigenvectors defined in [13].

Definition 3.14. Let \( \mathcal{A} \in \mathcal{S}_n \) and \( \lambda \in \mathbb{R} \). If \( \lambda \) satisfies

\[
\mathcal{A} \cdot u \cdot u = \lambda u
\]

for a unit vector \( u \in \mathbb{R}^n \). Then \( \lambda \) is called a Z-eigenvalue [13] of \( \mathcal{A} \). This vector is called the Z-eigenvector associated with \( \lambda \).

Remark 3.15. Let \( \mathcal{A}, \mathcal{B} \in \mathcal{S}_n \). If \( \mathcal{A} \) is orthogonally similar to \( \mathcal{B} \), then \( \mathcal{A} \) and \( \mathcal{B} \) have the same Z-eigenvalues [13, Thm 2.20]. In fact, if

\[
\mathcal{A} = \mathcal{B} \cdot Q^T \cdot Q \quad \text{and} \quad \mathcal{A} \cdot u \cdot u = \lambda u
\]

for \( \lambda \in \mathbb{R} \) and a unit vector \( u \in \mathbb{R}^n \), then \( \mathcal{B} \cdot (Qu) \cdot (Qu) = \lambda Qu \).

Theorem 3.16. Let \( \mathcal{A} \in \mathcal{S}_n \). We have two necessary and sufficient conditions below:

(i) \( \mathcal{A} \in \mathcal{PD}_n \) if and only if \( \{e_i : 1 \leq i \leq n\} \) is a set of Z-eigenvectors. This is equivalent to

\[
\mathcal{A} \cdot e_i \cdot e_i \cdot e_j = 0
\]

for any \( 1 \leq i \neq j \leq n \).

(ii) \( \mathcal{A} \in \mathcal{O}(\mathcal{PD}_n) \) if and only if there exists an orthonormal set of Z-eigenvectors \( \{u_i : 1 \leq i \leq n\} \). This is equivalent to

\[
\mathcal{A} \cdot u_i \cdot u_i \cdot u_j = 0
\]

for any \( 1 \leq i \neq j \leq n \). In this case, \( \mathcal{A} \cdot Q_1^T \cdot Q_2^T \cdot Q_3^T \in \mathcal{PD}_n \) for \( Q_3 = [u_1, \cdots, u_n] \).

Proof. (i) By definition, \( \mathcal{A} \in \mathcal{PD}_n \) if and only if \( \mathcal{A} \cdot e_i \cdot e_i \cdot e_j = 0 \) for any \( 1 \leq i \neq j \leq n \). But if \( \mathcal{A} \cdot e_i \cdot e_i \) is orthogonal to every \( e_j, j \neq i \), it must be collinear to \( e_i \), which means

\[
\mathcal{A} \cdot e_i \cdot e_i = \lambda e_i
\]

for some nonzero \( \lambda \), which turns out to yield \( \lambda = \mathcal{A} \cdot e_i \cdot e_i \cdot e_i \).

(ii) The second result follows from (i) and Remark 3.15.
3.4.2. Relationship with orthogonally decomposable tensors

Example 3.17. We present an example to show that $O(D_n) \nsubseteq O(PD_n)$ for $S_n$. Let
\[
A = e_1 \otimes e_2 + e_1 \otimes e_3 + e_2 \otimes e_1 + e_2 \otimes e_3 + e_3 \otimes e_1 + e_3 \otimes e_2.
\]
It is easy to see that $A \in O(PD_3)$. On the other hand, it is known [24, Prop. 3.1 and 4.3] that the symmetric tensor rank is
\[
srank(A) = 4,
\]
hence $A$ cannot be in $O(D_3)$ (otherwise it would have rank at most 3).

Proposition 3.18. (i) Let $A \in PD_n$. Then $A \in D_n$ if and only if $A \cdot e_i \cdot e_j \in \text{span}\{e_i, e_j\}$ for any $1 \leq i \neq j \leq n$.

(ii) Let $A \in O(PD_n)$. Let $\{u_i: 1 \leq i \leq n\}$ be the set of orthonormal $Z$-eigenvectors, proved to exist in Theorem 3.16 (ii). Then $A \in O(D_n)$ if and only if $A \cdot u_i \cdot u_j \in \text{span}\{u_i, u_j\}$ for any $1 \leq i \neq j \leq n$.

Proof. First note that $A \cdot e_i \cdot e_j \in \text{span}\{e_i, e_j\}$ for any $1 \leq i < j \leq n$ if and only if $A_{ijk} = 0$ for any $1 \leq i < j < k \leq n$. Then (i) is proved. Next, (ii) follows from (i) and Remark 3.15. \qed

4. Locally maximally diagonal tensors

Even if Givens rotations span $SO_n$, it is not obvious that a sequence of optimally chosen Givens rotations will find the optimal orthogonal transform in $SO_n$. In other words, we know that $LMD_n \subseteq LJD_n$, but the converse may not be true. This motivates the comparison between $LJD_n$ and $LMD_n$.

4.1. Riemannian Hessian

In this subsection, we study the conditions that a tensor in $S_n$ is locally maximally diagonal based on the Riemannian Hessian [16, 22, 23].

Lemma 4.1. Let $A \in S_n$ and $f$ be as in (2). Let $T_Q O_n$ be the tangent vector space at $Q$; it contains matrices of the form $Q \Delta$, where $\Delta$ are skew-symmetric matrices satisfying $\Delta^T = -\Delta$. We denote
\[
U = A \cdot Q^T, \quad V = A \cdot Q^T \cdot Q^T, \quad X = V \cdot (Q \Delta)^T, \quad Y = U \cdot (Q \Delta)^T \cdot (Q \Delta)^T, \quad Z = V \cdot (Q \Delta^2)^T.
\]
Let $\text{Hess}(f)(Q)$ be the Riemannian Hessian of $f$ at $Q$. Then $\text{Hess}(f)(Q)(\Delta_1, \Delta_2)$ is a bilinear form defined on $T_Q O_n$. We have:
\[
\text{Hess}(f)(Q)(Q \Delta, Q \Delta) = 6 \sum_j (3X_{jji}^2 + 2Y_{jii}W_{jii} - Z_{jji}W_{jij}).
\]
Proof. By eqn. (2.55), it can be calculated that

\[ \text{Hess}(Q)(Q\Delta, Q\Delta) = \sum_{i,j,k,l} \frac{\partial^2 f}{\partial Q_{ij} \partial Q_{kl}} (Q\Delta)_{ij} (Q\Delta)_{kl} + \frac{1}{2} \text{tr}((\nabla f(Q))^T Q\Delta^2 + \Delta(\nabla f(Q))^T Q\Delta) \]

\[ = 6 \sum_{i,j,k} (3V_{ijj} V_{kjj} + 2W_{ijj} U_{ikj})(Q\Delta)_{ij} (Q\Delta)_{kj} + \text{tr}((\nabla f(Q))^T Q\Delta^2) \]

\[ = 6 \sum_{i,j,k} (3V_{ijj} V_{kjj} + 2W_{ijj} U_{ikj})(Q\Delta)_{ij} (Q\Delta)_{kj} - 6 \sum_{i,j,k,l} Q_{ik} V_{ijj} W_{jjl} \Delta_{kl} \Delta_{jl} \quad (9) \]

\[ = 6 \sum_{j} (3X_{jjj}^2 + 2Y_{jjj} W_{jjj} + Z_{jjj} W_{jjj}). \]

\[ \square \]

Corollary 4.2. Let \( Q = I_n \) in Lemma 4.1. The tangent vector space \( T_{I_n} \mathcal{O}_n \) contains the skew symmetric matrices \( \Delta \). It follows by (9) that

\[ \text{Hess}(I_n)(\Delta, \Delta) = 6 \sum_{i,j,k} (3A_{i,ijj}A_{k,jkj} + 2A_{i,ijj}A_{k,ikj}) \Delta_{ij} \Delta_{kj} - 6 \sum_{i,j} A_{iii} A_{ii} \Delta_{ij} \Delta_{kj} \]

\[ = 6 \sum_{i,j} (3A_{i,ijj}^2 + 2A_{i,ijj}A_{ii} - A_{ii}^2) \Delta_{ij}^2 \]

\[ + 6 \sum_{i,j,k,l \neq i} (3A_{i,ijj}A_{k,jkj} + 2A_{i,ijj}A_{k,ikj} - A_{kii} A_{iii}) \Delta_{ij} \Delta_{kj} \]

Remark 4.3. Let \( \mathcal{A} \in \mathcal{D}_n \).

(i) If \( \text{Hess}(I_n)(\Delta, \Delta) < 0 \) for any \( \Delta \in T_{I_n} \mathcal{O}_n \setminus \{0\} \), then \( \mathcal{A} \in \mathcal{LMD}_n \).

(ii) If \( \mathcal{A} \in \mathcal{LMD}_n \), then \( \text{Hess}(I_n)(\Delta, \Delta) \leq 0 \) for any \( \Delta \in T_{I_n} \mathcal{O}_n \).

4.2. Euclidean Hessian matrix for \( \mathcal{D}_3 \)

Note that \( \mathcal{LMD}_n \subseteq \mathcal{D}_n \) and \( \mathcal{LMD}_n \) is corresponding to the local maximum point of (2). In this subsection, based on Corollary 4.2, we show how to determine whether \( \mathcal{A} \in \mathcal{D}_3 \) is locally maximally diagonal or not.

Definition 4.4. Let \( \mathcal{A} \in \mathcal{D}_3 \). Let \( \gamma_{12}, \gamma_{13} \) and \( \gamma_{23} \) be the stationary diagonal ratios introduced in Definition 3.11. Denote by

\[ a = A_{111}, \quad b = A_{222}, \quad c = A_{333} \quad \text{and} \quad g = A_{123}. \]

We define the Euclidean Hessian matrix of \( \mathcal{A} \) to be \( M_{\mathcal{A}} \)

\[ M_{\mathcal{A}} = \begin{bmatrix}
(3\gamma_{12}^2 + 2\gamma_{12} - 1)(a^2 + b^2) & 2ga + (3\gamma_{12}\gamma_{13} - \gamma_{23})bc & -2gb - (3\gamma_{23}\gamma_{12} - \gamma_{13})ca \\
2ga + (3\gamma_{12}\gamma_{13} - \gamma_{23})bc & (3\gamma_{12}^2 + 2\gamma_{12} - 1)(c^2 + a^2) & 2gc + (3\gamma_{13}\gamma_{23} - \gamma_{12})ab \\
-2gb - (3\gamma_{23}\gamma_{12} - \gamma_{13})ca & 2gc + (3\gamma_{13}\gamma_{23} - \gamma_{12})ab & (3\gamma_{23}^2 + 2\gamma_{23} - 1)(b^2 + c^2)
\end{bmatrix}. \]

Theorem 4.5. Let \( \mathcal{A} \in \mathcal{D}_3 \). If \( M_{\mathcal{A}} \) is negative definite, then \( \mathcal{A} \in \mathcal{LMD}_3 \). If \( \mathcal{A} \in \mathcal{LMD}_3 \), then \( M_{\mathcal{A}} \) is negative semidefinite.
Proof. Let
\[ \Delta = \begin{bmatrix} 0 & u & v \\ -u & 0 & w \\ -v & -w & 0 \end{bmatrix} \in T_{I_3}O_3. \]

Define \( \Phi(u, v, w) \overset{\text{def}}{=} \text{Hess}(I_3)(\Delta, \Delta) \). By Corollary 4.2 we have that
\[
\Phi(u, v, w) = 6\left[(3\gamma_{12}^2 + 2\gamma_{12} - 1)(a^2 + b^2)u^2 + (3\gamma_{13}^2 + 2\gamma_{13} - 1)(c^2 + a^2)v^2 \\
+ (3\gamma_{23}^2 + 2\gamma_{23} - 1)(b^2 + c^2)w^2 + 4g(auv + cvw - bwu) \\
+ 6(\gamma_{12}\gamma_{13}bcuv + \gamma_{13}\gamma_{23}abvw - \gamma_{23}\gamma_{12}cawu) - 2(\gamma_{23}bcuv + \gamma_{12}abvw - \gamma_{13}cawu)\right]
= 6\xi^T M_A \xi,
\]
where \( \xi = (u, v, w)^T \). By Remark 4.3, the proof is complete. \( \square \)

Example 4.6. Let \( A \in J\mathcal{D}_3 \) be such that \( A_{111} = A_{222} = A_{333} = 1, \gamma_{12} = \gamma_{13} = \gamma_{23} = \gamma \).
(i) Then
\[
\Phi(u, v, w) = 12(u^2 + v^2 + w^2)[(3\gamma^2 + 2\gamma - 1) - (3\gamma^2 - \gamma + 2A_{123})\frac{uw - uv - vw}{u^2 + v^2 + w^2}]
\]
for any \((u, v, w) \in \mathbb{R}^3 \setminus \{(0, 0, 0)\}\) by (10). Note that
\[
\frac{uw - uv - vw}{u^2 + v^2 + w^2} \in [-1/2, 1].
\]
Since \( 3\gamma^2 + 2\gamma - 1 \leq 0 \) by Theorem 3.13, it follows that \( A \in L\mathcal{M}\mathcal{D}_3 \) if
\[
3\gamma - 1 < A_{123} < -\frac{9}{2}\gamma^2 - 3\gamma + 1.
\]
Moreover, we have that \( A \not\in L\mathcal{M}\mathcal{D}_3 \), if
\[
A_{123} < 3\gamma - 1 \text{ or } A_{123} > -\frac{9}{2}\gamma^2 - 3\gamma + 1.
\]
(ii) If \( \gamma = 0 \), then
\[
\Phi(u, v, w) = -12(u^2 + v^2 + w^2)[1 + 2A_{123}\frac{uw - uv - vw}{u^2 + v^2 + w^2}],
\]
for any \((u, v, w) \in \mathbb{R}^3 \setminus \{(0, 0, 0)\}\). Note that
\[
\frac{uw - uv - vw}{u^2 + v^2 + w^2} \in [-1/2, 1].
\]
It follows that \( A \in L\mathcal{M}\mathcal{D}_3 \) if \( A_{123} \in (-1/2, 1) \). Moreover, if \( A_{123} \notin [-1/2, 1] \), then \( A \not\in L\mathcal{M}\mathcal{D}_3 \).

Example 4.7. Let \( A \in S_n \) with \( n > 3 \). Suppose that
\[
A^{(i,j,k)} \in L\mathcal{M}\mathcal{D}_3
\]
for any $1 \leq i < j < k \leq n$. It may be interesting to wonder whether it holds that
\[ A \in \mathcal{LMD}_n. \]

In fact, the answer is negative. Let $A \in \mathcal{PD}_4$ with
\[ A_{ijk} = \begin{cases} 1, & i = j = k, \\ 3/4, & i \neq j \neq k, \\ 0, & \text{otherwise}. \end{cases} \]

By Example 4.6 (ii), we see that $A(i,j,k) \in \mathcal{LMD}_3$ for any $1 \leq i < j < k \leq n$. Let
\[ \Delta_* = \begin{bmatrix} 0 & 1 & 1 & 1 \\ -1 & 0 & 0 & 0 \\ -1 & 0 & 0 & 0 \\ -1 & 0 & 0 & 0 \end{bmatrix} \in T_4O_4. \]

By Corollary 4.2, we get that $\text{Hess}(I_4)(\Delta_*, \Delta_*) = 18 > 0$. It follows that $A \not\in \mathcal{LMD}_4$.

5. Orbit of generally maximally diagonal tensors

5.1. Equivalent problem formulations

In this subsection, we first prove that the statement $O(\mathcal{GMD}_n) = S_n$ is equivalent to several other optimization problems in Proposition 5.2. Then we give a positive answer to these equivalent problems when the dimension is 2 in Theorem 5.3.

Proposition 5.1. Let $n \geq 2$. Then $O(\mathcal{GMD}_n) = S_n$ if and only if $\mathcal{GMD}_n = \mathcal{MD}_n$.

Proof. We only have to prove that $\mathcal{GMD}_n = \mathcal{MD}_n$ if $O(\mathcal{GMD}_n) = S_n$. In fact, if $A \in \mathcal{MD}_n$, there exists $Q_*$ such that
\[ A \cdot Q_1^T \cdot Q_2^T \cdot Q_3^T \in \mathcal{GMD}_n. \]

Let $f$ be as in (2) and $F$ be as in (8). It follows that
\[ f(I_n) \geq f(Q_*) = \max_{P, Q, R \in O_n} F(P, Q, R) \geq f(I_n). \]

Then we have that $A \in \mathcal{GMD}_n$. \qed

Proposition 5.2. Denote
\[ \mathcal{GO}(\mathcal{D}_n) \overset{\text{def}}{=} \{ A \cdot P_1^T \cdot Q_2^T \cdot R_3^T A \in \mathcal{D}_n, P, Q, R \in O_n \}. \]

The following statements are equivalent.
(i) $\mathcal{SMD}_n = \mathcal{MD}_n$.
(ii) For any $A \in S_n$,
\[ \max_{Q \in O_n} f(Q) = \max_{P, Q, R \in O_n} F(P, Q, R), \]

5. Orbit of generally maximally diagonal tensors
where \( f \) is as in \((2)\) and \( \mathcal{F} \) is as in \((3)\).

(iii) For any \( \mathbf{A} \in S_n \), it holds that

\[
\min_{\mathbf{u}_i \perp \mathbf{u}_j, \forall i \neq j, \mu_k \in \mathbb{R}} \|\mathbf{A} - \sum_{k=1}^{n} \mu_k \mathbf{u}_k \otimes \mathbf{u}_k \| = \min_{\mathbf{x}_i \perp \mathbf{y}_j, \forall i \neq j, \mathbf{z}_k \in \mathbb{R}} \|\mathbf{A} - \sum_{k=1}^{n} \lambda_k \mathbf{x}_k \otimes \mathbf{y}_k \otimes \mathbf{z}_k \|.
\]

(iv) For any \( \mathbf{A} \in S_n \) and for the Euclidean distance \( d \), it holds that

\[
d(\mathbf{A}, \mathcal{O}(\mathcal{D}_n)) = d(\mathbf{A}, \mathbb{S}\mathcal{O}(\mathcal{D}_n)).
\]

(v) Let \( \mathbf{A} \in S_n \). The best rank-\( n \) orthogonal approximation can always be chosen to be symmetric, that is, there exist \( \mathbf{\mu} \in \mathbb{R} \) and orthonormal basis \( \{\mathbf{u}_k, 1 \leq k \leq n\} \) such that

\[
\|\mathbf{A} - \sum_{k=1}^{n} \mu_k \mathbf{u}_k \otimes \mathbf{u}_k \| = \min_{\mathbf{x}_i \perp \mathbf{y}_j, \forall i \neq j, \lambda_k \in \mathbb{R}} \|\mathbf{A} - \sum_{k=1}^{n} \lambda_k \mathbf{x}_k \otimes \mathbf{y}_k \otimes \mathbf{z}_k \|.
\]

Proof. (i)\(\Leftrightarrow\)(ii). Suppose that (i) holds and \( Q_* = \arg \max_{\mathbf{Q} \in \mathbb{S}\mathcal{O}_n} f(\mathbf{Q}) \). Let

\[
\mathbf{W}_* = \mathbf{A} \ast Q_*^T \mathbf{Q}^T_\mathcal{S} Q^T_\mathcal{S}
\]

Then \( \mathbf{W}_* \in \mathcal{M}D_n \) and thus \( \mathbf{W}_* \in \mathbb{S}\mathcal{M}\mathcal{D}_n \). It follows that

\[
f(Q_*) = \max_{\mathbf{P}, \mathbf{Q}, \mathbf{R} \in \mathbb{S}\mathcal{O}_n} \mathcal{F}(\mathbf{P}, \mathbf{Q}, \mathbf{R}).
\]

If (ii) holds and \( \mathbf{A} \in \mathcal{M}D_n \), then \( \mathbf{I}_n = \arg \max_{\mathbf{Q} \in \mathbb{S}\mathcal{O}_n} f(\mathbf{Q}) \) and thus

\[
(\mathbf{I}_n, \mathbf{I}_n, \mathbf{I}_n) = \arg \max_{\mathbf{P}, \mathbf{Q}, \mathbf{R} \in \mathbb{S}\mathcal{O}_n} \mathcal{F}(\mathbf{P}, \mathbf{Q}, \mathbf{R}),
\]

which implies that \( \mathbf{A} \in \mathbb{S}\mathcal{M}\mathcal{D}_n \).

(ii)\(\Leftrightarrow\)(iii). By [24, Proposition 5.1], [24, (5.6)] and [24, (5.23)], we get that

\[
\max_{\mathbf{Q} \in \mathbb{S}\mathcal{O}_n} f(\mathbf{Q}) = \|\mathbf{A}\|^2 - \min_{\mathbf{u}_i \perp \mathbf{u}_j, \forall i \neq j, \mu_k \in \mathbb{R}} \|\mathbf{A} - \sum_{k=1}^{n} \mu_k \mathbf{u}_k \otimes \mathbf{u}_k \|^2,
\]

\[
\max_{\mathbf{P}, \mathbf{Q}, \mathbf{R} \in \mathbb{S}\mathcal{O}_n} \mathcal{F}(\mathbf{P}, \mathbf{Q}, \mathbf{R}) = \|\mathbf{A}\|^2 - \min_{\mathbf{x}_i \perp \mathbf{y}_j, \forall i \neq j, \mathbf{z}_k \in \mathbb{R}} \|\mathbf{A} - \sum_{k=1}^{n} \lambda_k \mathbf{x}_k \otimes \mathbf{y}_k \otimes \mathbf{z}_k \|^2.
\]

It follows that (ii)\(\Leftrightarrow\)(iii).

(iii)\(\Leftrightarrow\)(iv) is clear.

(iii)\(\Leftrightarrow\)(v). Note that \( \mathcal{O}(\mathcal{D}_n) \) is closed. There exist \( \mathbf{\mu} \in \mathbb{R} \) and an orthonormal basis \( \{\mathbf{u}_k^*, 1 \leq k \leq n\} \) such that

\[
\|\mathbf{A} - \sum_{k=1}^{n} \mu_k \mathbf{u}_k \otimes \mathbf{u}_k \| = \min_{\mathbf{v}_i \perp \mathbf{v}_j, \forall i \neq j, \mu_k \in \mathbb{R}} \|\mathbf{A} - \sum_{k=1}^{n} \mu_k \mathbf{v}_k \otimes \mathbf{v}_k \|. \]
Theorem 5.3. It holds that $\mathcal{MD}_2 = \mathcal{GMD}_2$.

Proof. We only need to prove that $A \in \mathcal{GMD}_2$ if $A \in \mathcal{MD}_2$. Let

\[ \mathcal{W} = A \cdot P^T \cdot Q^T \cdot R^T \]

with $P, Q, R \in \mathcal{SO}_2$. These rotations can be written as

\[ P = \frac{1}{\sqrt{1+x^2}} \begin{bmatrix} 1 & -x \\ x & 1 \end{bmatrix}, \quad Q = \frac{1}{\sqrt{1+y^2}} \begin{bmatrix} 1 & -y \\ y & 1 \end{bmatrix} \quad \text{and} \quad R = \frac{1}{\sqrt{1+z^2}} \begin{bmatrix} 1 & -z \\ z & 1 \end{bmatrix} \]

for $x, y, z \in \mathbb{R}$. Define

\[ F(x, y, z) \overset{\text{def}}{=} \| \text{diag} \{ \mathcal{W} \} \|^2 \]

as in (8). Denote

\[ a = A_{111}, \quad b = A_{112}, \quad c = A_{122}, \quad d = A_{222}, \]

and $\gamma = \gamma_{12}$ is the stationary diagonal ratio in Definition 3.11. Then $c = \gamma a$ and $b = \gamma d$ by definition. Moreover, $\gamma \in [-1, 1/3]$ by Theorem 3.13. It can be calculated that

\[ F(x, y, z) = a^2 + d^2 + \frac{(a^2 + d^2)(\gamma + 1)}{(1+x^2)(1+y^2)(1+z^2)} \sigma(x, y, z), \]

where

\[ \sigma(x, y, z) \overset{\text{def}}{=} (\gamma - 1)(x^2 + y^2 + z^2 + x^2y^2 + y^2z^2 + z^2x^2) \]

\[ + 2\gamma(x^2yz + xyz^2 + xy^2z + yz + zx). \]

Note that $F(0, 0, 0) = a^2 + d^2$. We only need to prove that $\sigma(x, y, z) \leq 0$ for any $x, y, z \in \mathbb{R}$. If $\gamma \in [0, 1/3]$, then $\gamma - 1 \leq -2\gamma$, and thus

\[ \sigma(x, y, z) \leq -2\gamma([x^2 + y^2 + z^2 + x^2y^2 + y^2z^2 + z^2x^2] \]

\[ - (x^2yz + xyz^2 + xy^2z + yz + zx)] \]

\[ = -\gamma[(x - y)^2 + (y - z)^2 + (z - x)^2 + (xy - yz)^2 + (yz - zx)^2 + (zx - xy)^2] \leq 0. \]

If $\gamma \in [-1, 0)$, then

\[ \sigma(x, y, z) = -(1 + \gamma)(x^2 + y^2 + z^2 + x^2y^2 + y^2z^2 + z^2x^2) \]

\[ + \gamma((x + y)^2 + (y + z)^2 + (z + x)^2 + (xy + yz)^2 + (yz + zx)^2 + (zx + xy)^2) \leq 0. \]

5.2. Symmetric tensors of dimension $n > 2$

In this subsection, we first present a counterexample to show that the equivalent problems in Proposition 5.1 and Proposition 5.2 have a negative answer when $n > 2$. Then we prove a related result, which can be seen as an orthogonal analogue of the Comon’s conjecture.
5.2.1. A counterexample

**Lemma 5.4.** Define

\[ \rho(Q) \overset{\text{def}}{=} Q_{11}^2 Q_{22}^2 + Q_{21}^2 Q_{22}^2 + Q_{31}^2 Q_{32}^2 \]  

for \( Q \in \mathbb{O}_3 \). We have that \( \rho(Q) < 1/12 \) for any \( Q \in S\mathbb{O}_3 \).

The proof of Lemma 5.4 can be found in Appendix A.

**Example 5.5.** Let \( A \) be as in Example 3.17. Let \( F \) be as in (8). Suppose that \( P^* = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, Q^* = \begin{bmatrix} 0 & 0 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}, R^* = \begin{bmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{bmatrix} \).

Then \( F(P^*, Q^*, R^*) = 3 \). However, easy calculations show that \( f(Q) = 36 \rho(Q) < 3 \), where the last inequality follows by Lemma 5.4. Thus, we see that

\[ f(Q) < 3 = F(P^*, Q^*, R^*) \]

for any \( Q \in \mathbb{O}_3 \). It follows that Proposition 5.2 (ii) has a negative answer when \( n > 2 \). Moreover, we have \( O(\text{SM}_{\mathbb{R}}) \subseteq S_n \) when \( n > 2 \) by Proposition 5.7.

**Remark 5.6.** It was proved that the best rank-1 approximation of any \( A \in S_n \) can always be chosen to be symmetric \[25, 26\]. Example 5.5 provides a counterexample to Proposition 5.2 (v) when \( n > 2 \). It will be interesting to study whether the best rank-\( p \) (\( 1 < p < n \)) orthogonal approximation can be chosen to be symmetric when \( n > 2 \), which can be seen as an orthogonal analogue of \[27, \text{Conjecture 8.7}\].

5.2.2. An orthogonal analogue of Comon’s conjecture

Although Proposition 5.2 (iii) has a negative answer by Example 5.5 when \( n > 2 \), we have the following result.

**Proposition 5.7.** Let \( A \in S_n \). Then for any \( p \)

\[ \min \| A - \sum_{k=1}^{p} \lambda_k x_k \otimes y_k \otimes z_k \| = 0 \]

implies

\[ \min_{u_k \perp u_j, 1 \neq j} \| A - \sum_{k=1}^{p} \mu_k u_k \otimes u_k \otimes u_k \| = 0. \]

**Proof.** Suppose that \( 1 \leq p \leq n \) and

\[ A = \sum_{k=1}^{p} \lambda_k x_k \otimes y_k \otimes z_k. \]
where $\lambda_k \in \mathbb{R} \setminus \{0\}$ and $x_i \perp x_j, y_i \perp y_j, z_i \perp z_j$ for any $i \neq j$. We assume that $\|x_k\| = \|y_k\| = \|z_k\| = 1$ without loss of generality. Note that $\mathbf{A}$ is symmetric. Then

$$\mathbf{A} \cdot z_k = \lambda_k x_k \otimes y_k$$

is a symmetric matrix for any $1 \leq k \leq p$. It follows that $x_k = \pm y_k$. In a similar way, we can prove that $y_k = \pm z_k$. The proof is complete. \qed

**Corollary 5.8.** Let $\mathbf{A} \in S_n$. Then we have

$$d(\mathbf{A}, \mathfrak{O}(D_n)) = 0 \Rightarrow d(\mathbf{A}, \mathfrak{O}(D_n)) = 0,$$

that is,

$$\mathfrak{O}(D_n) = S_n \cap \mathfrak{O}(D_n).$$

**Remark 5.9.** (i) Proposition 5.7 can be seen as an orthogonal analogue of the Comon’s conjecture [12, 27, 18], which conjectured that rank and symmetric rank of a symmetric tensor are equal, that is,

$$\min_{x_k, y_k, z_k \in \mathbb{R}^n, \lambda_k \in \mathbb{R}} \|\mathbf{A} - \sum_{k=1}^p \lambda_k x_k \otimes y_k \otimes z_k\| = 0 \Rightarrow \min_{u_k \in \mathbb{R}^n, \mu_k \in \mathbb{R}} \|\mathbf{A} - \sum_{k=1}^p \mu_k u_k \otimes u_k \otimes u_k\| = 0$$

for any $\mathbf{A} \in S_n$ and $p \in \mathbb{N}$ minimal.

(ii) An alternative proof of Corollary 5.8 can be found in [28, Proposition 32].

6. Convergence results for cyclic Jacobi algorithm

6.1. Cyclic Jacobi algorithm description

In this subsection, we recall the cyclic Jacobi algorithm (also called the Jacobi CoM2 algorithm) given in [10, 3], which is a special case of Algorithm 1.

**Algorithm 2.** Input: $\mathbf{A} \in S_n$ and $Q_0 = \mathbf{I}_n$.

Output: a sequence of iterations $\{Q_k : k \in \mathbb{N}\}$.

- For $k = 1, 2, \ldots$ until a stopping criterion is satisfied do:
  - Choose the pair $(i_k, j_k)$ according to the following cyclic-by-row rule
    $$\begin{align*}
    (1, 2) &\rightarrow (1, 3) \rightarrow \cdots \rightarrow (1, n) \rightarrow \\
    (2, 3) &\rightarrow \cdots \rightarrow (2, n) \rightarrow \\
    &\cdots \rightarrow \\
    (n-1, n) &\rightarrow \\
    (1, 2) &\rightarrow (1, 3) \rightarrow \cdots .
    \end{align*}$$
  - Compute the angle $\theta_k^* \in \mathbb{R}$ that maximizes the function $h_k(\theta)$ defined in [5].
  - Update $Q_k = Q_{k-1} \mathcal{G}^{(i_k, j_k, \theta_k^*)}$.

- End for
6.2. Derivatives and relations between them

In this subsection, we present some basic properties of Algorithm \textsuperscript{2}. More details can be found in \textsuperscript{10, 17}. We first give a definition.

Take the $k$-th iteration with pair $(i_k, j_k)$ in Algorithm \textsuperscript{2}. Let

\[
\mathcal{W}_{k}^{(k-1)} = A \cdot Q_{k-1}^T \cdot Q_{k-1}^T \cdot Q_{k-1}^T.
\]

By \textsuperscript{[5]}, we have that

\[
h_k(\theta) = \| \text{diag}\{ \mathcal{W}_{1}^{(k-1)} \cdot \mathcal{G}^{(i_k, j_k, \theta)} \cdot \mathcal{G}^{(i_k, j_k, \theta)} \cdot \mathcal{G}^{(i_k, j_k, \theta)} \} \|^2. \quad (14)
\]

Let $x = \tan(\theta)$, and define

\[
\tau_k : \mathbb{R} \rightarrow \mathbb{R} \quad \text{by} \quad \tau_k(x) \overset{\text{def}}{=} h_k(\arctan(x)).
\]

In the rest of this subsection, with some abuse of notation, we use a shorthand notation $d_k = d_{i_k, j_k}(\mathcal{W}_{k}^{(k-1)})$ and $\omega_k = \omega_{i_k, j_k}(\mathcal{W}_{k}^{(k-1)})$. It can be calculated that \textsuperscript{[17], Lemma 5.8]

\[
\begin{align*}
\tau_k(x) - \tau_k(0) & = \frac{3}{(1 + x^2)^2} (2d_k(x - x^3) - \omega_k x^2), \quad (15) \\
\tau'_k(x) & = \frac{6}{(1 + x^2)^3} (d_k(1 - 6x^2 + x^4) - \omega_k (x - x^3)), \quad (16) \\
\tau''_k(x) & = \frac{6}{(1 + x^2)^4} [2d_k(-9x + 14x^3 - x^5) - \omega_k (1 - 8x^2 + 3x^4)].
\end{align*}
\]

Remark 6.1. Denote by $x_k^* = \tan(\theta_k^*)$ the optimal point of $\tau_k(x)$. Note that $\tau'_k(x_k^*) = 0$. It follows by \textsuperscript{[10]} that

\[
d_k(1 - 6x_k^2 + x_k^4) - \omega_k(x_k^* - x_k^*) = 0. \quad (17)
\]

(i) If $x_k^* - x_k^* = 0$, we get that

\[
\omega_k = \frac{(1 - 6x_k^2 + x_k^4)}{x_k^4(1 - x_k^2)} d_k, \quad \text{and thus} \quad \tau_k(x_k^*) - \tau_k(0) = \frac{3x_k^*}{(1 - x_k^2)} d_k.
\]

(ii) If $1 - 6x_k^2 + x_k^4 = 0$, we get that

\[
d_k = \frac{x_k^4(1 - x_k^2)}{(1 - 6x_k^2 + x_k^4)} \omega_k,
\]

and thus

\[
\tau''_k(x_k^*) = \frac{-6\omega_k}{(1 - 6x_k^2 + x_k^4)},
\]

\[
\tau_k(x_k^*) - \tau_k(0) = \frac{3x_k^*}{(1 - 6x_k^2 + x_k^4)} \omega_k. \quad (18)
\]
6.3. Convergence properties

In this subsection we prove some results on the convergence properties of Algorithm 2.

**Proposition 6.2.** Suppose that $\mathcal{A} \in S_n$ and $\{Q_k : k \in \mathbb{N}\} \subseteq SO_n$ are the iterations of Algorithm 2. If $Q_k \to Q_*$ and

$$W^* = A_1 Q_*^T \cdot Q_*^T \cdot Q_*^T,$$

then $d_{i,j}(W^*) = 0$ and $\omega_{i,j}(W^*) \geq 0$ for any $1 \leq i < j \leq n$.

*Proof.* Fix any $1 \leq i < j \leq n$. We choose a subsequence $\{\ell_k\}$ such that
does not converge to $Q_*$ for any $\ell$. It follows by Proposition 6.2 and Lemma 3.7, we see that if the iterations of Algorithm 2 converge to $Q_*$, then the result follows from continuity of the function.

$$Q \mapsto A_1 Q_*^T \cdot Q_*^T \cdot Q_*^T.$$  \hfill (19)

**Proposition 6.3.** Let $\mathcal{A} \in S_n$ and $\{Q_k : k \in \mathbb{N}\} \subseteq SO_n$ be the iterations of Algorithm 2. Suppose that there are a finite number of accumulation points of $\{Q_k : k \in \mathbb{N}\}$.

(i) Let $Q_* \in SO_n$ be any accumulation point and

$$W^* = A_1 Q_*^T \cdot Q_*^T \cdot Q_*^T.$$

Then there exists $1 \leq i_* < j_* \leq n$ such that $d_{i_*,j_*}(W^*) = \omega_{i_*,j_*}(W^*) = 0$.

(ii) For any $1 \leq i_* < j_* \leq n$, there exists an accumulation point $Q_* \in SO_n$ such that

$$W^* = A_1 Q_*^T \cdot Q_*^T \cdot Q_*^T.$$

(iii) We have that the directional derivative of $W^*$ tends to zero:

$$h_{i_*,j_*}'(0) = 6d_{i_*,j_*}(W^*) \to 0.$$

The proof can be found in Appendix A.

**Corollary 6.4.** Suppose that $\mathcal{A} \in S_n$ and $\{Q_k : k \in \mathbb{N}\} \subseteq SO_n$ are the iterations of Algorithm 2. Let $Q_* \in SO_n$ be an accumulation point and

$$W^* = A_1 Q_*^T \cdot Q_*^T \cdot Q_*^T.$$

If $\omega_{i,j}(W^*) > 0$ for any $1 \leq i < j \leq n$, then either $Q_k \to Q_*$, or there exist an infinite number of accumulation points in the iterations.

**Remark 6.5.** By Proposition 6.2 and Lemma 3.7, we see that if the iterations of Algorithm 2 converge to $Q_*$, then

$$W^* = A_1 Q_*^T \cdot Q_*^T \cdot Q_*^T,$$

satisfies $W^* \in \mathcal{F}D_n$; in particular, $Q_*$ is a stationary point of $W^*$ by Remark 3.6. However, $\mathcal{LM}D_3 \not\subseteq \mathcal{F}D_3$ by Example 4.6. It follows that Algorithm 2 may converge to a saddle point of $W^*$. 

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7. Conclusions

In this paper, we studied several classes of third order approximately diagonal tensors, which are closely related to Jacobi-type algorithms and the approximate diagonalization problem (1). We believe that these classes provide a better understanding of problem (1) and behavior of optimization algorithms; some examples in this paper can be used as test cases for the algorithms. There are some open questions left for future research, such as the global convergence of Algorithm 2 for third (or higher) order symmetric tensors.

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Appendix A. Remaining proofs

Proof of Lemma 5.4 Since (12) is invariant with respect to changes of signs of the columns of $Q$, it suffices to prove the statement for $Q \in SO_3$.

Step 1. By [29, p. 10], any $Q \in SO_3$ can be decomposed as $Q^T = Q_1(x)Q_2(y)Q_3(z)$, where

$$Q_1(x) = \frac{1}{\sqrt{1+x^2}} \begin{bmatrix} \sqrt{1+x^2} & 0 & 0 \\ 0 & 1 & -x \\ 0 & x & 1 \end{bmatrix}, \quad Q_2(y) = \frac{1}{\sqrt{1+y^2}} \begin{bmatrix} 1 & -y & 0 \\ y & 1 & 0 \\ 0 & 0 & \sqrt{1+y^2} \end{bmatrix},$$

$$Q_3(z) = \frac{1}{\sqrt{1+z^2}} \begin{bmatrix} 1 & 0 & -z \\ 0 & \sqrt{1+z^2} & 0 \\ z & 0 & 1 \end{bmatrix}$$

for $x, y, z \in \mathbb{R}$. It can be calculated that

$$\rho(x, y, z) \overset{\text{def}}{=} \rho(Q) = \frac{1}{(1+x^2)(1+y^2)(1+z^2)} [y^4z^2 + y^2z^2](x^4 + 1) + 2\sqrt{x^2+1}y^3(z^3 - z)(x^3 - x) + (y^4z^4 - 4y^4z^2 + y^4 + y^2z^4 + y^2 + z^2)x^2].$$

If $x = 0$, then

$$\rho(0, y, z) = \frac{y^2z^2}{(1+y^2)(1+z^2)} \leq \frac{1}{16} < \frac{1}{12}.$$ 

The similar result holds if $z = 0$. Therefore, we only need to prove that $\rho(x, y, z) < 1/12$ in the case that $xz \neq 0$.

Step 2. Let $u = x - \frac{1}{x}$ and $v = z - \frac{1}{z}$. 

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We define
\[
\Phi(u, v, y) \overset{\text{def}}{=} \rho(x, y, z) = \frac{1}{(u^2 + 4)(v^2 + 4)(1 + y^2)^3}[(y^4 + y^2)u^2 + 2\sqrt{y^4 + 1}y^3 v u + (y^4 + y^2)(v^2 + 4) + 1 - 4y^4].
\]

Let \((u_*, v_*, y_*)\) be the maximal point. If \(y_* = 0\), then
\[
\Phi(u_*, v_*, 0) = \frac{1}{(u^2_* + 4)(v^2_* + 4)} \leq \frac{1}{16} < \frac{1}{12}.
\]
Now we prove that \(u_*^2 = v_*^2\) if \(y_* \neq 0\). Assume that \(u_*^2 \neq v_*^2\). By
\[
\frac{\partial \Phi}{\partial u}(u_*, v_*, y_*) = \frac{\partial \Phi}{\partial v}(u_*, v_*, y_*) = 0,
\]
we get that
\[
u_*[v_*^2(y_*^4 + y_*^2) + 1 - 4y_*^4] = -y_*^3\sqrt{1 + y_*^2}v_* (u_*^2 - 4), \tag{A.1} \\
v_*[u_*^2(y_*^4 + y_*^2) + 1 - 4y_*^4] = -y_*^3\sqrt{1 + y_*^2}u_* (v_*^2 - 4). \tag{A.2}
\]
If \(u_* = 0\), then \(v_* = 0\) by (A.1), which implies that \(u_*^2 = v_*^2\). Otherwise, if \(u_* \neq 0\), then \(v_* \neq 0\). It follows that
\[
u_*^2(u_*^2 - 4)[v_*^2(y_*^4 + y_*^2) + 1 - 4y_*^4] = v_*^2(u_*^2 - 4)[u_*^2(y_*^4 + y_*^2) + 1 - 4y_*^4]
\]
by (A.1) and (A.2). It can be calculated that
\[
(y_*^4 + y_*^2)u_*^2v_*^2(v_*^2 - u_*^2) = -4(1 - 4y_*^4)(u_*^2 - v_*^2).
\]
By the assumption that \(u_*^2 \neq v_*^2\), we have
\[
u_*^2v_*^2 = \frac{-4(1 - 4y_*^4)}{y_*^4 + y_*^2}. \tag{A.3}
\]
Moreover, by (A.1) and (A.2), we also get
\[
(1 - 4y_*^4)(u_*^2 - v_*^2) = -y_*^3\sqrt{1 + y_*^2}u_* v_* (u_*^2 - v_*^2),
\]
which implies that
\[
u_*v_* = \frac{1 - 4y_*^4}{-y_*^3\sqrt{1 + y_*^2}}. \tag{A.4}
\]
By (A.3) and (A.4), we get that \(1 - 4y_*^4 = 0\). It follows that \(u_*v_* = 0\) by (A.3), which contradicts the assumption that \(u_* \neq 0\). Therefore, we prove that \(u_*^2 = v_*^2\).

**Step 3.** Now we define
\[
\psi(u, y) \overset{\text{def}}{=} \Phi(u, \pm u, y) = \frac{2(y^4 + y^2 \pm \sqrt{y^2 + 1}y^3)u^2 + 4y^2 + 1}{(u^2 + 4)(1 + y^2)^3}, \tag{A.5}
\]
\[
\varphi(u, y) \overset{\text{def}}{=} \frac{4y^2u^2(1 + y^2) + 4y^2 + 1}{(u^2 + 4)(1 + y^2)^3}. \tag{A.6}
\]
Note that
\[
\varphi(u, y) = \frac{2(y^4 + y^2 + \sqrt{y^2 + 1} + y^4)u^2 + 4y^2 + 1}{(u^2 + 4)^3(1 + y^2)^3} \geq \psi(u, y)
\]
for any \(u, y \in \mathbb{R}\). It is enough to prove that \(\varphi(u, y) < 1/12\) for any \(u, y \in \mathbb{R}\). Let \((u_*, y_*)\) be the maximal point of \(\varphi(u, y)\). By
\[
\frac{\partial \varphi}{\partial u}(u_*, y_*) = \frac{\partial \varphi}{\partial y}(u_*, y_*) = 0,
\]
we have that
\[
\begin{align*}
y_*(4u_*^2y_*^4 - 4u_*^2 + 8y_*^6 - 1) &= 0, \quad \text{(A.7)} \\
u_*(2u_*^2y_*^4 + 2u_*^2y_*^2 - 8y_*^4 - 4y_*^2 + 1) &= 0. \quad \text{(A.8)}
\end{align*}
\]
If \(y_* = 0\), then \(u_* = 0\) by (A.8). If \(u_* = 0\), then \(y_* = 0\) or \(y_*^2 = 1/8\) by (A.7). It is easy to check that \(\varphi(u_*, y_*) < 1/12\) in all these cases.

Now we assume that \(y_* \neq 0\) and \(u_* \neq 0\). Then (A.8) can be rewritten as
\[
u_*^2 = \frac{8y_*^4 + 4y_*^2 - 1}{2y_*^2(y_*^2 + 1)}. \quad \text{(A.9)}
\]
By substituting (A.9) into (A.8), we get that
\[
\varphi(u_*, y_*) = \frac{4y_*^2}{16y_*^6 + 18y_*^4 + 11y_*^2 - 1}.
\]
Next, we substitute (A.9) into (A.7), and get that \(y^{**}\) should satisfy
\[
(1 - 8y_*^2)(2y_*^4 + 2y_*^2) = (4y_*^4 - 4)(8y_*^4 + 4y_*^2 - 1).
\]
After division by \((y_*^2 + 1)\), we have
\[
16y_*^6 - 11y_*^2 + 2 = 0, \quad \text{(A.10)}
\]
which is a 3rd degree polynomial equation in \(y_*^2\): there are two positive solutions of (A.10) given by positive roots of the polynomial, i.e., \(y_*^2 \approx 0.7162\) or \(y_*^2 \approx 0.1921\). Taking into account (A.10), we have have that
\[
\varphi(u_*, y_*) = \frac{4y_*^2}{28y_*^4 + 22y_*^2 - 1}.
\]
hence \(\varphi(u_*, y_*) \approx 0.076 < 1/12\) or \(\varphi(u_*, y_*) \approx 0.065 < 1/12\) in these two cases.

**Step 4.** Finally, we have that
\[
\rho(x, y, z) \leq \max_{u, v, y \in \mathbb{R}} \Phi(u, v, y) = \max_{u, y \in \mathbb{R}} \psi(u, y) \leq \varphi(u_*, y_*) < \frac{1}{12}
\]
for any \(x, y, z \in \mathbb{R}\), which completes the proof. \(\square\)
Proof of Proposition 6.3. Since there are a finite number of accumulation points, there exists \( \delta > 0 \) such that the \( \delta \)-neighborhoods of these accumulation points have positive distance to each other.

(i) Let \( Q_\ast \) be any accumulation point. Let \( \mathcal{L} \subseteq \mathbb{N} \) be a subsequence such the subsequence \( \{Q_\ell, \ell \in \mathcal{L}\} \) is located in the \( \delta \)-neighborhood \( \mathcal{N}(Q_\ast, \delta) \) and \( Q_\ell \to Q_\ast \) when \( \ell \in \mathcal{L} \) tends to infinity. Note that \( Q_\ast \) is not the unique accumulation point. There exists a pair \((i_\ast, j_\ast)\) such that it appears for an infinite number of times in the sequence of pairs

\[
\{(i_{\ell+1}, j_{\ell+1}), Q_{\ell+1} \notin \mathcal{N}(Q_\ast, \delta), \ell \in \mathcal{L}\}.
\]

Now we construct a subsequence \( \mathcal{P} \subseteq \mathcal{L} \) such that

\[
(i_{p+1}, j_{p+1}) = (i_\ast, j_\ast) \quad \text{and} \quad Q_{p+1} \notin \mathcal{N}(Q_\ast, \delta)
\]

for any \( p \in \mathcal{P} \). Note that the \( \delta \)-neighborhoods of different accumulation points have positive distance to each other. There exists \( \sigma > 0 \) such that \( |x_{p+1}^\ast| > \sigma \) for any \( p \in \mathcal{P} \). Note that \( |x_{p+1}^\ast| \leq 1 \) for any \( p \in \mathcal{P} \). There exists \( \zeta_0 \in [-1, 1] \) such that \( \sigma \leq |\zeta_0| \leq 1 \) and \( \zeta_0 \) is an accumulation point of \( \{x_{p+1}, p \in \mathcal{P}\} \). We assume that \( x_{p+1} \to \zeta_0 \) for simplicity.

Now we prove that

\[
d_{i_\ast, j_\ast}(W^{(p)}) \to 0 \quad \text{and} \quad \omega_{i_\ast, j_\ast}(W^{(p)}) \to 0
\]

when \( p \in \mathcal{P} \) tends to infinity, and thus get (i) by the continuity of (19).

Denote by

\[
\vartheta_p \overset{\text{def}}{=} 2d_{i_\ast, j_\ast}(W^{(p)})(x_{p+1}^\ast - (x_{p+1}^\ast)^3) - \omega_{i_\ast, j_\ast}(W^{(p)})(x_{p+1}^\ast)^2
\]

and

\[
M_p \overset{\text{def}}{=} \begin{bmatrix}
2(x_{p+1}^\ast - (x_{p+1}^\ast)^3) & -(x_{p+1}^\ast)^2 \\
1 - 6(x_{p+1}^\ast)^2 + (x_{p+1}^\ast)^4 & -x_{p+1}^\ast + (x_{p+1}^\ast)^3
\end{bmatrix}.
\]

By (15) and (17), we see that

\[
M_p \begin{bmatrix} d_{i_\ast, j_\ast}(W^{(p)}) \\ \omega_{i_\ast, j_\ast}(W^{(p)}) \end{bmatrix} = \begin{bmatrix} \vartheta_p \\ 0 \end{bmatrix}.
\]

Note that

\[
\det(M_p) = -(x_{p+1}^\ast)^2((x_{p+1}^\ast)^2 + 1) \to -\zeta_0^2(\zeta_0^2 + 1)^2 \neq 0
\]

when \( p \in \mathcal{P} \) tends to infinity. Then \( M_p \) is invertible when \( p \) is large enough. Note that \( \vartheta_p \to 0 \). It follows that

\[
\begin{bmatrix} d_{i_\ast, j_\ast}(W^{(p)}) \\ \omega_{i_\ast, j_\ast}(W^{(p)}) \end{bmatrix} = M_p^{-1} \begin{bmatrix} \vartheta_p \\ 0 \end{bmatrix} \to 0
\]

when \( p \in \mathcal{P} \) tends to infinity.

\footnote{We use a simplified notation for subsequences in order to avoid multilevel indices.}
(ii) Let \((i_*, j_*)\) be any pair. There exists an accumulation point \(Q_* \in \mathcal{O}_n\) such that, if \(\{Q_\ell, \ell \in \mathcal{L}\}\) is the subsequence of \(\{Q_k, k \in \mathbb{N}\}\) located in \(N(Q_*, \delta)\), then \((i_*, j_*)\) appears for an infinite number of times in the sequence of pairs \(\{(i_\ell+1, j_\ell+1), \ell \in \mathcal{L}\}\).

(a) If it appears for an infinite number of times in \(\{(i_\ell+1, j_\ell+1), Q_\ell+1 \not\in N(Q_*, \delta)\}\), then the result follows by the same reasoning as in (i).

(b) Otherwise, it appears for an infinite number of times in \(\{(i_\ell+1, j_\ell+1), Q_\ell+1 \in N(Q_*, \delta)\}\).

We construct the subsequence \(\{Q_p, p \in \mathcal{P}\}\) of \(\{Q_\ell, \ell \in \mathcal{L}\}\) such that \((i_p+1, j_p+1) = (i_*, j_*)\) and \(Q_{p+1} \in N(Q_*, \delta)\).

Note that \(Q_p \to Q_*\) and \(Q_{p+1} \to Q_*\) when \(p \in \mathcal{P}\) tends to infinity. We get that \(x^p_{\ell+1} \to 0\), and eventually from the proof of Theorem 6.2:

\[d_{i_*, j_*}(W^{(p)}) \to 0\quad \text{and} \quad \omega_{i_*, j_*}(W^{(p)}) \geq 0\]

when \(p \in \mathcal{P}\) is large enough. Then we prove (ii) by the continuity of (19).

(iii) Note that there exist a finite number of accumulation points and \(|x^*_k| \leq 1\) for any \(k > 1\). The sequence \(\{x^k_\ell, k\}\) has a finite number of accumulation points. Let \(\zeta_0\) be any one of them. Then \(\zeta_0 \in [-1, 1]\). There exists a subsequence \(\mathcal{L} \subseteq \mathbb{N}\) such that \(x^\ell_{\ell+1} \to \zeta_0\) when \(\ell \in \mathcal{L}\) tends to infinity.

Next, we have that

\[d_{i_\ell+1, j_\ell+1}(W^{(\ell)}) \to 0,\]

which follows by (17) if \(\zeta_0 = 0\), and by a reasoning similar to (i) if \(\zeta_0 \neq 0\). Finally, note that there exist a finite number of accumulation points in \(\{x^*_k\}\), hence the proof is complete.

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