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Sharp Z-Eigenvalue Inclusion Set-Based Method for Testing the Positive Definiteness of Multivariate Homogeneous Forms

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Abstract. In this paper, we establish a sharp *Z*-eigenvalue inclusion set for even-order real tensors by *Z*-identity tensor and prove that new *Z*-eigenvalue inclusion set is sharper than existing results. We propose some sufficient conditions for testing the positive definiteness of multivariate homogeneous forms via new *Z*-eigenvalue inclusion set. Further, we establish upper bounds on the *Z*-spectral radius of weakly symmetric nonnegative tensors and estimate the convergence rate of the greedy rank-one algorithms. The given numerical experiments show the validity of our results.

1. Introduction

Consider the following multivariate homogeneous forms with spherical constraint:

$$f_{\mathcal{A}}(x) = \mathcal{A}x^{m} = \sum_{i_{1}, i_{2}, \dots, i_{m}=1}^{n} a_{i_{1}i_{2}\dots i_{m}} x_{i_{1}} x_{i_{2}} \dots x_{i_{m}}$$

$$s.t. \quad x^{\top}x = 1,$$
(1)

where $x \in \mathbb{R}^n, m, n \ge 2, f_{\mathcal{A}}(x)$ is a multivariate homogeneous form of degree *m* with *n* variables, and $\mathcal{A} \in \mathbb{R}^{[m,n]}$ is an *m*-order *n*-dimensional real tensor with entries [12, 14]

$$a_{i_1...i_m} \in \mathbb{R}, i_j \in \mathbb{N} = \{1, ..., n\}, j = 1, ..., m.$$

Clearly, the critical points of (1) satisfy the following equations for some $\lambda \in \mathbb{R}$:

$$\mathcal{A}x^{m-1} = \lambda x \text{ and } x^{\top}x = 1, \tag{2}$$

where $(\mathcal{A}x^{m-1})_i = \sum_{i_2,...,i_m \in \mathbb{N}} a_{ii_2...i_m} x_{i_2} \dots x_{i_m}$. The real number λ and the real vector x satisfying with (2) are

called Z-eigenvalue and Z-eigenvector, respectively.

The multivariate homogeneous form $f_{\mathcal{A}}(x)$ is positive definite, which plays important roles in signal processing [15] and the stability study of nonlinear autonomous systems via Lyapunov's direct method in

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automatic control [3, 4, 13]. Note that $f_{\mathcal{A}}(x)$ is positive definite if and only if tensor \mathcal{A} is positive definite, and that an even-order real symmetric tensor is positive definite if and only if all of its Z-eigenvalues are positive [14]. Some effective algorithms for finding Z-eigenvalue and the corresponding eigenvector have been implemented [5–9, 11, 16, 18, 21–26], but it is difficult to compute all the Z-eigenvalues and judge the positive definiteness of an even-order real symmetric tensor. Very recently, Li et al. [10] proposed Gershgorin-type Z-eigenvalue inclusion set with parameters by Z-identity tensor, which can identify the positive-definiteness of an even-order real symmetric tensor. It is remarkable that Brauer-type inclusion set is tighter than Gershgorin-type inclusion set [20]. As a continuation of the article [20], we shall establish sharp Brauer-type Z-eigenvalue localization set and propose some sufficient conditions for the positive definiteness of multivariate homogeneous forms.

To end this section, we introduce Z-identity tensor in [8, 10] and important results proposed in [10].

Definition 1.1. Assume that *m* is even. We call $I_Z \in \mathbb{R}^{[m,n]}$ a *Z*-identity tensor if

$$I_Z x^{m-1} = x, \ x^\top x = 1, \ \forall x \in \mathbb{R}^n.$$

It is worth noting that the even-order *n* dimension *Z*-identity tensor is not unique in general. For instance, each even tensor in the following is a Z-identity tensor:

Case I: $(I_Z)_{iii_2i_2...i_ki_k} = 1$, $\forall k \in N$ and m = 2k; Case II (Property 2.4 of [8]): $(I_Z)_{i_1i_2...i_m} = \frac{1}{m!} \sum_{p \in II_m} \delta_{i_p(1)} \delta_{i_p(2)} \dots \delta_{i_p(m-1)} \delta_{i_p(m)}$, where δ is the standard Kronecker,

i.e.,

$$\delta_{ij} = \begin{cases} 1, & \text{if } i = j, \\ 0, & \text{otherwise.} \end{cases}$$

Lemma 1.2. (Theorem 2 of [10]) Let $\mathcal{A} = (a_{i_1i_2...i_m}) \in \mathbb{R}^{[m,n]}$ and $I_Z \in \mathbb{R}^{[m,n]}$ be a Z-identity tensor with m being even. Let $\sigma_Z(\mathcal{A})$ be the set of all Z-eigenvalues of \mathcal{A} . For any real vector $\alpha = (\alpha_1, \ldots, \alpha_n)^{\mathsf{T}} \in \mathbb{R}^n$, then

$$\sigma_{Z}(\mathcal{A}) \subseteq \mathcal{G}(\mathcal{A}, \alpha) = \bigcup_{i \in \mathbb{N}} \mathcal{G}_{i}(\mathcal{A}, \alpha) := \{z \in \mathbb{R} : |z - \alpha_{i}| \leq R_{i}(\mathcal{A}, \alpha_{i})\},\$$

where $R_i(\mathcal{A}, \alpha_i) = \sum_{i_2...i_m \in \mathbb{N}} |a_{ii_2...i_m} - \alpha_i(I_Z)_{ii_2...i_m}|$. Furthermore, $\sigma_Z(\mathcal{A}) \subseteq \bigcap_{\alpha \in \mathbb{R}^n} \mathcal{G}(\mathcal{A}, \alpha)$.

2. A sharp Z-eigenvalue inclusion set for even-order real tensors

In this section, we establish new Z-eigenvalue inclusion set for even-order tensors. To this end, we define

$$\Theta_{j} = \left\{ (i_{2}, i_{3}, \dots, i_{m}) : i_{k} = j \text{ for some } k \in \{2, \dots, m\}, \text{ where } j, i_{2}, \dots, i_{m} \in N \right\},$$

$$\overline{\Theta}_{j} = \left\{ (i_{2}, i_{3}, \dots, i_{m}) : i_{k} \neq j \text{ all any } k \in \{2, \dots, m\}, \text{ where } j, i_{2}, \dots, i_{m} \in N \right\},$$

$$r_{i}^{\Theta_{j}}(\mathcal{A}, \alpha_{i}) = \sum_{\{i_{2}, \dots, i_{m}\} \in \Theta_{j}} |a_{ii_{2}\dots i_{m}} - \alpha_{i}(I_{Z})_{ii_{2}\dots i_{m}}|, r_{i}^{\overline{\Theta}_{j}}(\mathcal{A}, \alpha_{i}) = \sum_{\{i_{2}, \dots, i_{m}\} \in \overline{\Theta}_{j}} |a_{ii_{2}\dots i_{m}} - \alpha_{i}(I_{Z})_{ii_{2}\dots i_{m}}|.$$

Obviously, $R_i(\mathcal{A}, \alpha_i) = r_i^{\Theta_j}(\mathcal{A}, \alpha_i) + r_i^{\overline{\Theta}_j}(\mathcal{A}, \alpha_i).$

Theorem 2.1. Let $\mathcal{A} = (a_{i_1i_2...i_m}) \in \mathbb{R}^{[m,n]}$ and $I_Z \in \mathbb{R}^{[m,n]}$ be a Z-identity tensor with m being even. For any real vector $\alpha = (\alpha_1, \ldots, \alpha_n)^{\mathsf{T}} \in \mathbb{R}^n$, then

$$\sigma_{Z}(\mathcal{A}) \subseteq \mathfrak{O}(\mathcal{A}, \alpha) = \bigcup_{i \in N} \bigcap_{j \in N, j \neq i} \mathfrak{O}_{i,j}(\mathcal{A}, \alpha),$$

where $\mathcal{O}_{i,j}(\mathcal{A}, \alpha) = \left\{ z \in \mathbb{R} : (|\lambda - \alpha_i| - r_i^{\overline{\Theta}_j}(\mathcal{A}, \alpha_i)) | \lambda - \alpha_j| \le r_i^{\Theta_j}(\mathcal{A}, \alpha_i) \mathbb{R}_j(\mathcal{A}, \alpha_j) \right\}$. Furthermore, $\sigma_Z(\mathcal{A}) \subseteq \bigcap_{i \in \mathbb{N}^n} \mathcal{O}(\mathcal{A}, \alpha)$.

Proof. Let (λ, x) be a *Z*-eigenpair of \mathcal{A} and $I_Z \in \mathbb{R}^{[m,n]}$ be a *Z*-identity tensor, i.e.,

$$\mathcal{A}x^{m-1} = \lambda x = \lambda I_Z x^{m-1}, \ x^{\mathsf{T}}x = 1.$$
(3)

Assume $|x_t| = \max_{i \in N} |x_i|$, then $0 < |x_t|^{m-1} \le |x_t| \le 1$. On one hand, taking the *t*-th equation from (3), for any $j \in N$, $j \ne t$, we have

$$\sum_{i_2,\dots,i_m\in N}\lambda(I_Z)_{ii_2\dots i_m}x_{i_2}\dots x_{i_m} = \sum_{i_2,\dots,i_m\in N}a_{ii_2\dots i_m}x_{i_2}\dots x_{i_m}.$$
(4)

Hence, for any real number α_t , it follows that

$$(\lambda - \alpha_t)x_t = \sum_{i_2, \dots, i_m \in N} (\lambda - \alpha_t)(I_Z)_{i_2 \dots i_m} x_{i_2} \dots x_{i_m} = \sum_{i_2, \dots, i_m \in N} (a_{i_1 \dots i_m} - \alpha_t(I_Z)_{i_1 \dots i_m}) x_{i_2} \dots x_{i_m}$$

$$= \sum_{\{i_2, \dots, i_m\} \in \Theta_j} (a_{i_1 \dots i_m} - \alpha_t(I_Z)_{i_1 \dots i_m}) x_{i_2} \dots x_{i_m} + \sum_{\{i_2, \dots, i_m\} \in \overline{\Theta}_j} (a_{i_1 \dots i_m} - \alpha_t(I_Z)_{i_1 \dots i_m}) x_{i_2} \dots x_{i_m}$$
(5)

Taking modulus in (5) and using the triangle inequality give

$$\begin{aligned} |\lambda - \alpha_t||x_t| &\leq \sum_{\{i_2, \dots, i_m\} \in \Theta_j} |a_{ti_2 \dots i_m} - \alpha_t(I_Z)_{ti_2 \dots i_m}||x_{i_2}| \dots |x_{i_m}| + \sum_{\{i_2, \dots, i_m\} \in \overline{\Theta}_j} |a_{ti_2 \dots i_m} - \alpha_t(I_Z)_{ti_2 \dots i_m}||x_{i_2}| \dots |x_{i_m}| \\ &\leq r_t^{\Theta_j}(\mathcal{A}, \alpha_t)|x_j| + r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t)|x_t|, \end{aligned}$$

$$(6)$$

$$\left(|\lambda - \alpha_t| - r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t)\right)|x_t| \le r_t^{\Theta_j}(\mathcal{A}, \alpha_t)|x_j|.$$
(7)

On the other hand, for $t \neq j \in N$, taking the *j*-th equation from (3), we obtain

$$(\lambda - \alpha_j)x_j = \sum_{i_2, \dots, i_m \in N} (\lambda - \alpha_j)(I_Z)_{ji_2 \dots i_m} x_{i_2} \dots x_{i_m} = \sum_{i_2, \dots, i_m \in N} (a_{ji_2 \dots i_m} - \alpha_j(I_Z)_{ji_2 \dots i_m}) x_{i_2} \dots x_{i_m}.$$
(8)

Taking modulus in (8) and using the triangle inequality, one has

$$|\lambda - \alpha_j| |x_j| \le R_j(\mathcal{A}, \alpha_j) |x_t|.$$
(9)

If $|x_i| = 0$, by (7), we obtain

$$|\lambda - \alpha_t| \le r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t)$$

Thus, $\lambda \in \mathcal{O}_{t,i}(\mathcal{A}, \alpha) \subseteq \mathcal{O}(\mathcal{A}, \alpha)$.

Otherwise, $|x_i| > 0$. Multiplying (7) with (9) yields

$$\left(|\lambda - \alpha_t| - r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t)\right)|\lambda - \alpha_j||x_j||x_t| \le r_t^{\Theta_j}(\mathcal{A}, \alpha_t)R_j(\mathcal{A}, \alpha_j)|x_j||x_t|,$$

equivalently,

$$(|\lambda - \alpha_t| - r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t))|\lambda - \alpha_j| \le r_t^{\Theta_j}(\mathcal{A}, \alpha_t)R_j(\mathcal{A}, \alpha_j),$$

which implies $\lambda \in \mathcal{O}_{t,j}(\mathcal{A}, \alpha)$. From the arbitrariness of j, we have $\lambda \in \bigcup_{i \in N} \bigcap_{j \in N, j \neq i} \mathcal{O}_{i,j}(\mathcal{A}, \alpha)$. Further, $\sigma_Z(\mathcal{A}) \subseteq \mathcal{O}_{i,j}(\mathcal{A}, \alpha)$. $\bigcap_{\alpha \in \mathbb{R}^n} \mho(\mathcal{A}, \alpha) \text{ by the arbitrariness of } \alpha. \square$

Corollary 2.2. Let $\mathcal{A} = (a_{i_1i_2...i_m}) \in \mathbb{R}^{[m,n]}$ with *m* being even. For any real vector $\alpha = (\alpha_1, ..., \alpha_n)^\top \in \mathbb{R}^n$, then

 $\mho(\mathcal{A},\alpha)\subseteq \mathcal{G}(\mathcal{A},\alpha).$

Proof. For any $\lambda \in \mathcal{O}(\mathcal{A}, \alpha)$, without loss of generality, there exists $t \in N$ such that $\lambda \in \mathcal{O}_{t,s}(\mathcal{A})$, that is,

$$\left(|\lambda - \alpha_t| - r_t^{\Theta_s}(\mathcal{A}, \alpha_t)\right)|\lambda - \alpha_s| \le r_t^{\Theta_s}(\mathcal{A}, \alpha_t)R_s(\mathcal{A}, \alpha_s), \ \forall s \ne t.$$

$$(10)$$

Next, the following argument is divided into two cases.

Case I: $r_t^{\Theta_s}(\mathcal{A}, \alpha_t)R_s(\mathcal{A}, \alpha_s) = 0$. Since $|\lambda - \alpha_s| \ge 0$, from (10), we deduce $|\lambda - \alpha_t| - r_t^{\Theta_s}(\mathcal{A}, \alpha_t) \le 0$. Further, it holds that

$$|\lambda - \alpha_t| \leq r_t^{\overline{\Theta}_s}(\mathcal{A}, \alpha_t) \leq R_t(\mathcal{A}, \alpha_t),$$

i.e., $\lambda \in \mathcal{G}_t(\mathcal{A}, \alpha)$. So, we have $\mathcal{O}_{t,s}(\mathcal{A}, \alpha) \subseteq \mathcal{G}_t(\mathcal{A}, \alpha)$.

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Case II: $r_t^{\Theta_s}(\mathcal{A}, \alpha_t)R_s(\mathcal{A}, \alpha_s) > 0$. Then dividing both sides by $r_t^{\Theta_s}(\mathcal{A}, \alpha_t)R_s(\mathcal{A}, \alpha_s)$ in (10), we obtain

$$\frac{|\lambda - \alpha_t| - r_t^{\Theta_s}(\mathcal{A}, \alpha_t)}{r_t^{\Theta_s}(\mathcal{A}, \alpha_t)} \cdot \frac{|\lambda - \alpha_s|}{R_s(\mathcal{A}, \alpha_s)} \le 1,$$
(11)

which implies

$$\frac{|\lambda - \alpha_t| - r_t^{\Theta_s}(\mathcal{A}, \alpha_t)}{r_t^{\Theta_s}(\mathcal{A}, \alpha_t)} \le 1$$
(12)

or

$$\frac{|\lambda - \alpha_s|}{R_s(\mathcal{A}, \alpha_s)} \le 1.$$
(13)

If (12) holds, then we have $|\lambda - \alpha_t| - r_t^{\overline{\Theta}_s}(\mathcal{A}, \alpha_t) \le r_t^{\Theta_s}(\mathcal{A}, \alpha_t)$, i.e,

$$\lambda - \alpha_t | \leq r_t^{\overline{\Theta}_s}(\mathcal{A}, \alpha_t) + r_t^{\Theta_s}(\mathcal{A}, \alpha_t) = R_t(\mathcal{A}, \alpha_t).$$

So, $\lambda \in \mathcal{G}_t(\mathcal{A}, \alpha)$. Otherwise, (13) holds, we can verify $\lambda \in \mathcal{G}_s(\mathcal{A}, \alpha)$.

From the above two cases, we can get $\mathcal{O}_{t,s}(\mathcal{A}, \alpha) \subseteq \mathcal{G}_t(\mathcal{A}, \alpha) \cup \mathcal{G}_s(\mathcal{A}, \alpha)$. Thus, $\mathcal{O}(\mathcal{A}, \alpha) \subseteq \mathcal{G}(\mathcal{A}, \alpha)$ for a given parameter α . \Box

Next, we give a numerical comparison between Theorem 2.1 and Theorem 2 of [10].

Example 2.3. Consider $\mathcal{A} = (a_{ijkl}) \in \mathbb{R}^{[4,2]}$ defined by

$$a_{ijkl} = \begin{cases} a_{1111} = 10; a_{1122} = 9; a_{1121} = a_{1211} = -1; \\ a_{2222} = 5; a_{2211} = 6; a_{2122} = a_{2212} = -1; \\ a_{ijkl} = 0, \text{ otherwise.} \end{cases}$$

All *Z*-eigenvalues of \mathcal{A} are 5.0000 and 10.0000. We choose different parameters $\alpha_1 = [3, 8]^{\top}$, $\alpha_2 = [10, 7]^{\top}$, $\alpha_3 = [9, 5]^{\top}$ and $\alpha_4 = [9, 5.5]^{\top}$, respectively. Set $\alpha_1 = [3, 8]^{\top}$ and $I_Z = (I_{ijkl})$ as Case I of Definition 1.1

$$I_{ijkl} = \begin{cases} I_{1111} = I_{1122} = I_{2211} = I_{2222} = 1; \\ 0, \text{ otherwise.} \end{cases}$$

Accordingly to Theorem 2.1, we obtain

$$\mho(\mathcal{A}, \alpha_1 = (3, 8)) = [-7.5917, 16.5498] \cup [-3.8102, 15.7178] = [-7.5917, 16.5498];$$

Similarly, we can obtain the following table:

α	[3,8]⊤	[10,7] ^T	[9 <i>,</i> 5] [⊤]	[9 <i>,</i> 5.5] [⊤]
$\mho(\mathcal{A}, \alpha)$	[-7.5917, 16.5498]	[3.5949, 12.6533]	[3.6277,11]	[3.6088, 10.6225]
$\mathcal{G}(\mathcal{A}, \alpha)$	[-12, 18]	[2, 13]	[2,12]	[2.5, 12]

Numerical results show that the bound of Theorem 2.1 is tighter than that of Theorem 2 of [10] and the suitable parameter α has a great influence on the numerical effect.

3. Positive definiteness of multivariate homogeneous forms

In this section, based on the inclusion set $\mathcal{O}(\mathcal{A}, \alpha)$ in Theorem 2.1, we propose a sufficient condition for the positive definiteness of even-order tensors. Before proceeding further, we introduce the results of [1, 10].

Definition 3.1. (*i*) We say that \mathcal{A} is symmetric if

$$a_{i_1\ldots i_m} = a_{i_{\pi(1)}\ldots i_{\pi(m)}}, \forall \pi \in \Gamma_m,$$

where Γ_m is the permutation group of *m* indices.

(ii) We say that \mathcal{A} is weakly symmetric if the associated homogeneous polynomial $f_{\mathcal{A}}(x)$ satisfies

$$\nabla f_{\mathcal{A}}(x) = m \mathcal{A} x^{m-1}.$$

Obviously, if tensor \mathcal{A} is symmetric, then \mathcal{A} weakly symmetric. However, the converse result may not hold.

Lemma 3.2. (Theorem 3 of [10]) Let λ be a Z-eigenvalue of $\mathcal{A} = (a_{i_1i_2...i_m}) \in \mathbb{R}^{[m,n]}$ and $I_Z \in \mathbb{R}^{[m,n]}$ be a Z-identity tensor with m being even. If there exists a positive real vector $\alpha = (\alpha_1, ..., \alpha_n)^T$ such that

$$\alpha_i > R_i(\mathcal{A}, \alpha_i), \forall i \in N,$$

then $\lambda > 0$. Further, if \mathcal{A} is symmetric, then \mathcal{A} is positive definite and $f_{\mathcal{A}}(x)$ defined in (1) is positive definite.

Theorem 3.3. Let λ be a Z-eigenvalue of $\mathcal{A} = (a_{i_1i_2...i_m}) \in \mathbb{R}^{[m,n]}$ and $I_Z \in \mathbb{R}^{[m,n]}$ be a Z-identity tensor with m being even. For $i \in N$, if there exist a positive real vector $\alpha = (\alpha_1, ..., \alpha_n)^{\top}$ and $j \neq i$ such that

$$(\alpha_i - r_i^{\overline{\Theta}_j}(\mathcal{A}, \alpha_i))\alpha_j > r_i^{\Theta_j}(\mathcal{A}, \alpha_i)R_j(\mathcal{A}, \alpha_j),$$
(14)

then $\lambda > 0$. Further, if \mathcal{A} is symmetric, then \mathcal{A} is positive definite and $f_{\mathcal{A}}(x)$ defined in (1) is positive definite.

Proof. Suppose on the contrary that $\lambda \leq 0$. From Theorem 2.1, there exists $t \in N$ with $\lambda \in O_{t,i}(\mathcal{A}, \alpha_t)$, i.e.,

$$\left(|\lambda - \alpha_t| - r_t^{\Theta_j}(\mathcal{A}, \alpha_t))|\lambda - \alpha_j| \le r_t^{\Theta_j}(\mathcal{A}, \alpha_t)R_j(\mathcal{A}, \alpha_j), \forall j \ne t.$$

Further, it follows from $\alpha_i > 0$ and $\lambda \le 0$ that

$$(\alpha_t - r_t^{\Theta_j}(\mathcal{A}, \alpha_t))\alpha_j \leq r_t^{\Theta_j}(\mathcal{A}, \alpha_t)R_j(\mathcal{A}, \alpha_j), \forall j \neq t,$$

which contradicts (14). Thus, $\lambda > 0$. When \mathcal{A} is a symmetric tensor and all *Z*-eigenvalues are positive, \mathcal{A} is positive definite and $f_{\mathcal{A}}(x)$ defined in (1) is positive definite. \Box

The following example shows the validity of Theorem 3.3.

Example 3.4. Consider $f_{\mathcal{A}}(x) = \mathcal{A}x^m$ deduced by symmetric tensor $\mathcal{A} = (a_{ijkl}) \in \mathbb{R}^{[4,3]}$ as follows

 $\begin{aligned} a_{1111} &= 1.4; a_{2222} = 3.2; a_{3333} = 2.6; a_{1112} = a_{1121} = a_{1211} = a_{2111} = -0.1; \\ a_{1122} &= a_{1212} = a_{1221} = a_{2112} = a_{2121} = a_{2211} = 0.8; \\ a_{1133} &= a_{1313} = a_{1331} = a_{3113} = a_{3131} = a_{3311} = 1.1; \\ a_{1233} &= a_{1323} = a_{1332} = a_{2133} = a_{2313} = a_{2331} = -0.1; \\ a_{3123} &= a_{3132} = a_{3213} = a_{3221} = a_{3312} = a_{3321} = 0.1; \\ a_{2223} &= a_{2232} = a_{2232} = a_{3222} = 0.1; \\ a_{2233} &= a_{2332} = a_{2332} = a_{3223} = a_{3232} = a_{3322} = 1.0; \\ a_{1112} &= a_{1112} = a_{1112} = a_{1111} = a_{1111} = a_{1111} = -0.1; \\ a_{1113} &= a_{1113} = a_{1113} = a_{1113} = a_{1113} = a_{1113} = a_{1113} = a_{1111} = 0.1; \\ a_{2223} &= a_{2232} = a_{2232} = a_{2232} = a_{2322} = 0.1; \\ a_{2233} &= a_{2332} = a_{2332} = a_{3223} = a_{3232} = a_{3322} = a_{3323} = a_{3323} = a_{3332} = a_{3333} = a_{3332} = a_{3333} = a_{3333$

Taking I_Z as Case II (Case I) of Definition 1.1, by simple computations, we cannot find positive real number α_1 such that

$$\alpha_1 > R_1(\mathcal{A}, \alpha_1),$$

which shows that Theorem 3 of [10] cannot check the positive definiteness of \mathcal{A} and $f_{\mathcal{A}}(x)$. Set $\alpha = (2.85, 3.0, 2.7)$ and let $I_Z = (I_{ijkl})$ be Case II of Definition 1.1

$$I_{ijkl} = \left\{ \begin{array}{l} I_{1111} = I_{2222} = I_{3333} = 1; \\ I_{1122} = I_{1212} = I_{1221} = I_{1133} = I_{1313} = I_{1331} = \frac{1}{3}; \\ I_{2112} = I_{2121} = I_{2221} = I_{2233} = I_{2323} = I_{2332} = \frac{1}{3}; \\ I_{3113} = I_{3131} = I_{3311} = I_{3223} = I_{3232} = I_{3322} = \frac{1}{3}; \\ 0, \ otherwise. \end{array} \right.$$

From Theorem 3.3, we can calculate the following corresponding values

	$(\alpha_i - r_i^{\Theta_j}(\mathcal{A}, \alpha_i))\alpha_j$	$r_i^{\Theta_j}(\mathcal{A},\alpha_i)R_j(\mathcal{A},\alpha_j)$
i = 1, j = 2	2.85	1.575
i = 1, j = 3	1.755	1.275
i = 2, j = 1	4.56	2.065
i = 2, j = 3	6.21	2.55
i = 3, j = 1	6.27	3.54
i = 3, j = 2	6	1.5

From the above table, we verify

$$(\alpha_i - r_i^{\Theta_j}(\mathcal{A}, \alpha_i)))\alpha_j > r_i^{\Theta_j}(\mathcal{A}, \alpha_i)R_j(\mathcal{A}, \alpha_j), \forall i \neq j \in N,$$

which implies that \mathcal{A} is positive definite and $f_{\mathcal{A}}(x)$ is positive definite.

4. Estimations of Z-spectral radius and convergence rate on the greedy rank-one algorithms

As we know, the best rank-one approximation which has numerous applications in wireless communication systems, image processing, data analysis [7, 15–17, 21]. The best rank-one approximation of $\mathcal{A} = (a_{i_1i_2...i_m})$ is to find a rank-one tensor $\kappa x^m = (\kappa x_{i_1} x_{i_2} \dots x_{i_m})$ such that

$$\min_{\kappa \in \mathbb{R}, x} \{ \|\mathcal{A} - \kappa x^m\|_F : x^T x = 1 \},\$$

where $\|\mathcal{A}\|_F := \sqrt{\sum_{i_1,i_2,...,i_m \in \mathbb{N}} a_{i_1i_2...i_m}^2}$. When \mathcal{A} is nonnegative and weakly symmetric, $\rho(\mathcal{A})x_0^m$ is a best rank-one approximation of \mathcal{A} , i.e.,

$$\min_{\kappa \in \mathbb{R}, x^T x = 1} \|\mathcal{A} - \kappa x^m\|_F = \|\mathcal{A} - \rho(\mathcal{A}) x_0^m\|_F = \sqrt{\|\mathcal{A}\|_F^2 - \rho(\mathcal{A})^2}.$$

Further, Qi [17] defined the quotient on the residual of the best rank-one approximation of tensor \mathcal{A} as follows:

$$\omega = \frac{\|\mathcal{A} - \rho(\mathcal{A})x_0^m\|_F}{\|\mathcal{A}\|_F} = \sqrt{1 - \frac{\rho(\mathcal{A})^2}{\|\mathcal{A}\|_F^2}},$$

which can estimate the convergence rate of the greedy rank-one algorithm [2, 17, 18, 25]. Hence, we shall devote to finding sharp upper bounds of the *Z*-spectral radius of weakly symmetric nonnegative tensors to estimate the convergence rate of the greedy rank-one algorithms. We recall some fundamental results of nonnegative tensors [1].

Lemma 4.1. (*Theorem 3.11 of [1]*) Assume \mathcal{A} is a weakly symmetric nonnegative tensor. Then, $\rho(\mathcal{A}) = \lambda^*$, where λ^* denotes the largest Z-eigenvalue.

Lemma 4.2. (Corollary 4.10 of [1]) Assume A is a weakly symmetric nonnegative tensor. Then,

$$\rho(\mathcal{A}) \geq \max_{i \in \mathbb{N}} a_{i\ldots i}.$$

Theorem 4.3. Let $\mathcal{A} = (a_{i_1i_2...i_m}) \in \mathbb{R}^{[m,n]}$ be a weakly symmetric nonnegative tensor and $I_Z \in \mathbb{R}^{[m,n]}$ be a Z-identity tensor (Case I or Case II) with m being even. For real vector $\alpha = (\alpha_1, ..., \alpha_n)^{\top} \in \mathbb{R}^n$ with $\alpha_i \leq \max_{i \in \mathbb{N}} a_{i...i}$, then

$$\rho(\mathcal{A}) \leq \max_{i \in \mathbb{N}} \min_{j \in \mathbb{N}, i \neq j, \alpha \in \mathbb{R}^n} \frac{1}{2} (\alpha_i + \alpha_j + r_i^{\overline{\Theta}_j}(\mathcal{A}, \alpha_i) + \Lambda_{i,j}^{\frac{1}{2}}(\mathcal{A}, \alpha_i)),$$

where $\Lambda_{i,j}(\mathcal{A}) = (\alpha_i - \alpha_j + r_i^{\overline{\Theta}_j}(\mathcal{A}, \alpha_j))^2 + 4r_i^{\Theta_j}(\mathcal{A}, \alpha_i)R_j(\mathcal{A}, \alpha_j).$

Proof. From Lemma 4.1, we assume that $\rho(\mathcal{A}) = \lambda^*$ is the largest *Z*-eigenvalue. It follows from Theorem 2.1 that there exists $t \in N$ such that

$$(|\rho(\mathcal{A}) - \alpha_t| - r_t^{\Theta_j}(\mathcal{A}, \alpha_t))|\rho(\mathcal{A}) - \alpha_j| \le r_t^{\Theta_j}(\mathcal{A}, \alpha_t)R_j(\mathcal{A}, \alpha_j), \forall j \ne t.$$
(15)

Since \mathcal{A} is nonnegative and Lemma 4.2 holds, for $\alpha_i \leq \max_{i \in \mathcal{N}} a_{i...i}$, we have

$$\rho(\mathcal{A}) \geq \alpha_t \text{ and } \rho(\mathcal{A}) \geq \alpha_j.$$

Thus, (15) is equivalent to

$$(\rho(\mathcal{A}) - \alpha_t - r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t))(\rho(\mathcal{A}) - \alpha_j) \le r_t^{\Theta_j}(\mathcal{A}, \alpha_t)R_j(\mathcal{A}, \alpha_j), \forall j \ne t.$$
(16)

Solving for (16), we obtain

$$\rho(\mathcal{A}) \leq \frac{1}{2}(\alpha_j + \alpha_t + r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t) + \Lambda_{t,j}^{\frac{1}{2}}(\mathcal{A}, \alpha_t)),$$

where $\Lambda_{t,s}(\mathcal{A}) = (\alpha_t - \alpha_s + r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t))^2 + 4r_t^{\Theta_j}(\mathcal{A}, \alpha_t)R_j(\mathcal{A}, \alpha_j)$. Since $j \in N$ and α are chosen arbitrarily, it holds

$$\rho(\mathcal{A}) \leq \min_{j \in N, t \neq j, \alpha \in \mathbb{R}^n} \frac{1}{2} (\alpha_j + \alpha_t + r_t^{\overline{\Theta}_j}(\mathcal{A}, \alpha_t) + \Lambda_{t,j}^{\frac{1}{2}}(\mathcal{A}, \alpha_t)).$$

Consequently,

$$\rho(\mathcal{A}) \leq \max_{i \in N} \min_{j \in N, i \neq j, \alpha \in \mathbb{R}^n} \frac{1}{2} (\alpha_i + \alpha_j + r_i^{\overline{\Theta}_j}(\mathcal{A}, \alpha_i) + \Lambda_{i,j}^{\frac{1}{2}}(\mathcal{A}, \alpha_i)).$$

Thus, the conclusion holds. \Box

The following numerical experiment shows validity of Theorem 4.3 and gives an estimation for the convergence rate of the greedy rank-one algorithms.

Example 4.4. Consider tensor $\mathcal{A} = (a_{ijkl}) \in \mathbb{R}^{[4,2]}$ defined by

$$a_{ijkl} = \begin{cases} a_{1111} = 1; a_{2222} = 3; a_{1122} = a_{1212} = a_{1221} = a_{2112} = a_{2121} = a_{2211} = \frac{1}{3}; \\ a_{1112} = a_{1121} = a_{1211} = a_{2111} = \frac{1}{3}; a_{ijkl} = 0, \text{ otherwise.} \end{cases}$$

By simple computation, we obtain ($\rho(\mathcal{A}), x$) = (3, (0, 1)) and $||\mathcal{A}||_F$ = 3.3166. For this tensor, set α = (1, 1) and let I_Z = (I_{ijkl}) be Case II of Definition 1.1. The bounds via different estimations given in the literature are shown in the following table:

<i>G</i> . '	Wang	et al./	'Filomat 3	4:9 (2020)), 3131–3139
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References	upper bound	parameter α
Theorem 3.11 of [1]	$\rho(\mathcal{A}) \le 6.1283$	No
Corollary 4.5 of [19]	$\rho(\mathcal{A}) \leq 4.3333$	No
Theorems 4.5-4.7 of [20]	$\rho(\mathcal{A}) \le 4.1985$	No
Theorem 7 of [18]	$\rho(\mathcal{A}) \leq 4.0000$	No
Theorem 1 of [10]	$\rho(\mathcal{A}) \leq 3.3333$	$\alpha = (1,1)$
Theorems 4.1	$\rho(\mathcal{A}) \leq 3.1055$	$\alpha = (1, 1)$

From the table above, it is easy to see that only the upper bound obtained by Theorem 4.1 is smaller than $\|\mathcal{A}\|_{F}$. Consequently, we have

$$\min_{\kappa \in \mathbb{R}, \kappa \in \mathbb{R}^n, \kappa^T x = 1} \|\mathcal{A} - \kappa x^m\|_F = \sqrt{\|\mathcal{A}\|_F^2 - \rho(\mathcal{A})^2} \ge 1.3559.$$

Further, we obtain that the quotient on the residual of the best rank-one approximation of \mathcal{A} is

$$\omega = \frac{\|\mathcal{A} - \rho(\mathcal{A})x_0^m\|_F}{\|\mathcal{A}\|_F} = \sqrt{1 - \frac{\rho(\mathcal{A})^2}{\|\mathcal{A}\|_F^2}} \ge 0.3511,$$

which implies the convergence rate of the greedy rank-one algorithm [2, 17, 18, 24, 25].

5. Conclusions

In this paper, we established a Brauer-type *Z*-eigenvalue inclusion set for even-order real tensors by *Z*-identity tensor and proposed some sufficient conditions for the positive definiteness of multivariate homogeneous forms. Note that the suitable parameter α has a great influence on the numerical effects and positive definiteness of $f_{\mathcal{A}}(x)$. Therefore, how to select the suitable parameter α is our further research.

Competing Interests

The authors declare that they have no competing interests.

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