



Research article

Error bounds for linear complementarity problems of  $SDD_1$  matrices and  $SDD_1$ - $B$  matrices

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Abstract: An upper bound of the infinity norm for the inverse of  $SDD_1$  matrix is presented. We apply the new bound to linear complementarity problems (LCPs) and obtain an alternative error bound for LCPs of  $SDD_1$  matrices and  $SDD_1$ - $B$  matrices. In addition, a new lower bound for the smallest singular value is also given. Numerical examples show the validity of the results.

Keywords: error bounds; linear complementarity problems; infinity norm;  $SDD_1$  matrices;  $SDD_1$ - $B$  matrices; singular values

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1. Introduction

The linear complementarity problem, denoted by  $LCP(A, q)$ , is to find a vector  $x \in R^n$  such that

$$Ax + q \geq 0, \quad (Ax + q)^T x = 0, \quad x \geq 0, \tag{1.1}$$

where  $A = (a_{ij}) \in R^{n \times n}$  and  $q \in R^n$ . The  $LCP(A, q)$  often arises from the various scientific areas of computing, economics and engineering, such as quadratic programs, optimal stopping, Nash equilibrium point of a bimatrix games, the contact problems and the free boundary problem for journal bearing, etc. for details, see [1–3].

The  $LCP(A, q)$  has a unique solution for any  $q \in R^n$  if and only if  $A$  is a  $P$ -matrix, see [2]. Some basic definitions for the special matrix are given below: a matrix  $A = (a_{ij}) \in R^{n \times n}$  is called a  $Z$ -matrix, if  $a_{ij} \leq 0$  for any  $i \neq j$ ; a  $P$ -matrix, if all its principal minors are positive; an  $M$ -matrix, if  $A^{-1} \geq 0$  and  $A$  is a  $Z$ -matrix; an  $H$ -matrix, if its comparison matrix  $\tilde{A} = (\tilde{a}_{ij})$  is an  $M$ -matrix, where

$$\tilde{a}_{ij} = \begin{cases} |a_{ij}|, & \text{if } i = j, \\ -|a_{ij}|, & \text{if } i \neq j. \end{cases}$$

In [4], Chen et al. gave the following error bound of the  $LCP(A, q)$  when  $A$  is a  $P$ -matrix:

$$\|x - x^*\|_\infty \leq \max_{d \in [0,1]^n} \|(I - D + DA)^{-1}\|_\infty \|r(x)\|_\infty, \quad \forall x \in \mathbb{R}^n, \quad (1.2)$$

where  $x^*$  is the solution of the  $LCP(A, q)$ ,  $r(x) = \min\{x, Ax + q\}$ ,  $D = \text{diag}(d_i)$  with  $0 \leq d_i \leq 1$ , and the min operator  $r(x)$  denotes the componentwise minimum of two vectors. When real  $H$ -matrix  $A$  with positive diagonal entries is a subclass of  $P$ -matrices, computation of error bounds for the  $LCP(A, q)$  becomes simpler (see formula (2.4) in [4]). In recent years, numerous scholars are keen to study special  $H$ -matrices, such as  $S$ - $QN$  matrices [5],  $S$ -Nekrasov matrices [6],  $S$ - $SDDS$  matrices [7] and  $DZT$  matrices [8]. The corresponding error bound of  $LCP(A, q)$  for  $S$ - $QN$  matrices is presented by Li et al. in [5]. And new error bounds of the  $LCP(A, q)$  for  $S$ -Nekrasov matrices are achieved by Gao et al. in [9] and Dai et al. in [10], which depends only on the entries of the matrix  $A$ .

When the class of involved matrices is subclass of  $P$ -matrices that are not  $H$ -matrices, error bounds of  $LCP(A, q)$  also need to be studied, such as,  $B_\pi^R$ -matrices [11],  $B$ -Nekrasov matrices [12] and [13], weakly chained diagonally dominant  $B$ -matrices [14],  $CKV$ -type  $B$ -matrices [15]. Specially, some error bounds for linear complementarity problems with  $B$ -type matrices have been derived by the papers [16–19].

In this paper, we provide an upper bound for infinity norm of the inverse of  $SDD_1$  matrix and apply the new bound to estimate the error for linear complementarity problems of  $SDD_1$  matrices and  $SDD_1$ - $B$  matrices. Meanwhile, a lower bound for the smallest singular value is also provided. Numerical examples are given to show the effectiveness of the obtained results.

## 2. Preliminaries

In this section, some definitions and lemmas are given. For any positive integer  $n$ ,  $N$  denotes the set  $\{1, 2, \dots, n\}$ . Let  $A = (a_{ij})$  be an  $n \times n$  matrix.

**Definition 1.** [20] Given a matrix  $A = (a_{ij}) \in \mathbb{R}^{n \times n}$ , let the  $i$ th deleted absolute row sum be

$$r_i(A) = \sum_{j=1, j \neq i}^n |a_{ij}|, \quad \forall i \in N.$$

**Definition 2.** [21]  $A = (a_{ij}) \in \mathbb{R}^{n \times n}$  is called a strictly diagonally dominant ( $SDD$ ) matrix if, for all  $i \in N$ ,

$$|a_{ii}| > r_i(A).$$

**Definition 3.** [22] Matrix  $A = (a_{ij}) \in \mathbb{R}^{n \times n}$  is called an  $SDD_1$  matrix if, for each  $i \in N_1(A)$ ,

$$|a_{ii}| > P_i(A),$$

where

$$P_i(A) = \sum_{j \in N_1(A) \setminus \{i\}} |a_{ij}| + \sum_{j \notin N_1(A) \cup \{i\}} \frac{r_j(A)}{|a_{jj}|} |a_{ij}|,$$

$$N_1(A) = \{i \in N \mid |a_{ii}| \leq r_i(A)\}.$$

**Definition 4.** [22] A real square matrix  $A$  is called an  $SDD_1$ - $B$  matrix (or  $B_1$ -matrix) if it can be written as

$$A = B^+ + C, \quad C = \begin{pmatrix} r_1^+ & \cdots & r_1^+ \\ \vdots & \ddots & \vdots \\ r_n^+ & \cdots & r_n^+ \end{pmatrix}, \quad (2.1)$$

with  $r_i^+ = \max\{0, a_{ij} | j \neq i\}$  and  $B^+$  is an  $SDD_1$  matrix with positive diagonal entries.

**Definition 5.** [11] Matrix  $A = (a_{ij}) \in R^{n \times n}$  with positive row sums is a  $B$ -matrix if all of its off-diagonal elements are bounded above by the corresponding row means, i.e., for all  $i \in N$ ,

$$\sum_{k=1}^n a_{ik} > 0 \quad \text{and} \quad \frac{1}{n} \left( \sum_{k=1}^n a_{ik} \right) > a_{ij}, \quad \forall j \neq i.$$

Now, we will introduce some useful lemmas.

**Lemma 1.** [22] If matrix  $A = (a_{ij}) \in R^{n \times n}$  is an  $SDD_1$  matrix by rows, then is also an  $H$ -matrix, if, in addition,  $A$  has positive diagonal entries, then  $\det A > 0$ .

**Lemma 2.** [22] If  $A = (a_{ij}) \in R^{n \times n}$  is an  $SDD_1$ - $B$  matrix, then  $A$  is also a  $P$ -matrix.

**Remark 1.** From Definitions 2–5, Lemmas 1 and 2, we have the following relationships:

$$SDD \text{ matrices} \subseteq SDD_1 \text{ matrices} \subseteq H\text{-matrices},$$

$$B\text{-matrices} \subseteq SDD_1\text{-}B \text{ matrices} \subseteq P\text{-matrices}.$$

**Remark 2.** In Example 1 we shall see that an  $SDD_1$  matrix is not necessarily an  $SDD$  matrix. In Example 7 we shall also see that an  $SDD_1$ - $B$  matrix is not necessarily a  $B$ -matrix.

**Lemma 3.** [23] Let  $\gamma > 0$  and  $\eta \geq 0$ , then for any  $x \in [0, 1]$ ,

$$\frac{1}{1-x+xy} \leq \frac{1}{\min\{\gamma, 1\}}, \quad \frac{\eta x}{1-x+xy} \leq \frac{\eta}{\gamma}.$$

**Lemma 4.** Matrix  $A = (a_{ij}) \in R^{n \times n}$  is an  $SDD_1$  matrix with positive diagonal entries, then  $A_D = (\bar{a}_{ij}) = I - D + DA$  is also an  $SDD_1$  matrix, where  $D = \text{diag}(d_i)$  with  $0 \leq d_i \leq 1$ .

*Proof.* Since  $A_D = (\bar{a}_{ij}) = I - D + DA$ , then

$$\bar{a}_{ij} = \begin{cases} 1 - d_i + d_i a_{ii}, & \text{if } i = j, \\ d_i a_{ij}, & \text{if } i \neq j. \end{cases}$$

By Lemma 3, for any  $i \in N_1(A)$ , we have

$$\begin{aligned} P_i(A_D) &= \sum_{j \in N_1(A)/\{i\}} |d_i a_{ij}| + \sum_{j \notin N_1(A) \cup \{i\}} \frac{d_j r_j(A)}{1 - d_j + a_{jj} d_j} |d_i a_{ij}| \\ &= d_i \left( \sum_{j \in N_1(A)/\{i\}} |a_{ij}| + \sum_{j \notin N_1(A) \cup \{i\}} \frac{d_j r_j(A)}{1 - d_j + a_{jj} d_j} |a_{ij}| \right) \end{aligned}$$

$$\begin{aligned} &\leq d_i \left( \sum_{j \in N_1(A)/\{i\}} |a_{ij}| + \sum_{j \notin N_1(A) \cup \{i\}} \frac{r_j(A)}{|a_{jj}|} |a_{ij}| \right) \\ &= d_i P_i(A), \end{aligned}$$

and

$$d_i P_i(A) < 1 - d_i + d_i |a_{ii}| = |\bar{a}_{ii}|.$$

That is, for each  $i \in N_1(A_D) \subseteq N_1(A)$ ,  $|\bar{a}_{ii}| > P_i(A_D)$ , so  $A_D = I - D + DA$  is an  $SDD_1$  matrix.  $\square$

**Lemma 5.** [7] Let  $A = (a_{ij}) \in R^{n \times n}$ ,  $n \geq 2$ , be a nonsingular matrix. Then there is a vector  $x \in C^n$  such that  $\|x\|_\infty = 1$  and

$$\|A^{-1}\|_\infty^{-1} = \|Ax\|_\infty = \max_{1 \leq i \leq n} |(Ax)_i|.$$

### 3. An upper bound for $\|A^{-1}\|_\infty$ of $SDD_1$ matrices

In this section, a new error bound of  $\|A^{-1}\|_\infty$  is presented when  $A$  is an  $SDD_1$  matrix.

**Theorem 1.** Let  $A = (a_{ij}) \in R^{n \times n}$ ,  $n \geq 2$ , be an  $SDD_1$  matrix with poistive diagonal entries. Then

$$\|A^{-1}\|_\infty \leq \max \left\{ \frac{1}{\min_{i \in N_1(A)} \{ |a_{ii}| - P_i(A) \}}, \frac{1}{\min_{i \in N_2(A)} \left\{ |a_{ii}| - \sum_{j \neq i} |a_{ij}| \right\}} \right\},$$

where  $N_1(A) = \{i \in N \mid |a_{ii}| \leq r_i(A)\}$ ,  $N_2(A) = \{i \in N \mid |a_{ii}| > r_i(A)\}$ .

*Proof.* In accordance with Lemma 5, choose a vector  $x = (x_i) \in C^n$  such that

$$\|x\|_\infty = 1 \quad \text{and} \quad |(Ax)_i| \leq \|A^{-1}\|_\infty^{-1} \quad \text{for all } i = 1, \dots, n. \quad (3.1)$$

Assume that

$$|x_k| = 1 = \|x\|_\infty \quad \text{and} \quad |x_j| = \frac{r_j(A)}{|a_{jj}|} \quad \text{for all } j \in N_2(A). \quad (3.2)$$

When  $k \in N_1(A)$ , we have

$$(Ax)_k = \sum_{j \in N} a_{kj} x_j = a_{kk} x_k + \sum_{j \in N_1(A)/\{k\}} a_{kj} x_j + \sum_{j \notin N_1(A) \cup \{k\}} a_{kj} x_j.$$

implying, in view of (3.1) and (3.2), that

$$\begin{aligned} |a_{kk}| &= |a_{kk}| |x_k| = \left| (Ax)_k - \sum_{j \in N_1(A)/\{k\}} a_{kj} x_j - \sum_{j \notin N_1(A) \cup \{k\}} a_{kj} x_j \right| \\ &\leq |(Ax)_k| + \sum_{j \in N_1(A)/\{k\}} |a_{kj}| |x_j| + \sum_{j \notin N_1(A) \cup \{k\}} |a_{kj}| |x_j| \\ &\leq |(Ax)_k| + \sum_{j \in N_1(A)/\{k\}} |a_{kj}| + \sum_{j \notin N_1(A) \cup \{k\}} \frac{r_j(A)}{|a_{jj}|} |a_{kj}| = |(Ax)_k| + P_k(A). \end{aligned}$$

Then, by (3.1), we have

$$\|A^{-1}\|_{\infty} \leq \frac{1}{|a_{kk}| - P_k(A)} \leq \frac{1}{\min_{i \in N_1(A)} \{|a_{ii}| - P_i(A)\}}. \quad (3.3)$$

When  $k \in N_2(A)$ , by (3.1), we get

$$0 < \min_{i \in N_2(A)} \left( |a_{ii}| - \sum_{j \neq i} |a_{ij}| \right) \leq |a_{kk}| - \sum_{j \neq k} |a_{kj}|,$$

and

$$\begin{aligned} 0 < \min_{i \in N_2(A)} \left( |a_{ii}| - \sum_{j \neq i} |a_{ij}| \right) |x_k| &\leq |a_{kk}x_k| - \sum_{j \neq k} |a_{kj}| |x_j| \\ &\leq |a_{kk}x_k| - \left| \sum_{j \neq k} a_{kj}x_j \right| \leq \left| \sum_{k \in N_2(A), j \in N} a_{kj}x_j \right| \\ &\leq \max_{i \in N_2(A)} \left| \sum_{j \in N} a_{ij}x_j \right| \leq \|A^{-1}\|_{\infty}^{-1}. \end{aligned}$$

Then

$$\|A^{-1}\|_{\infty} \leq \frac{1}{\min_{i \in N_2(A)} \left\{ |a_{ii}| - \sum_{j \neq i} |a_{ij}| \right\}}. \quad (3.4)$$

From (3.3) and (3.4), the conclusion follows.  $\square$

We illustrate our results by the following two examples.

**Example 1.** Consider the bound of  $\|A_1^{-1}\|_{\infty}$  for an  $SDD_1$  matrix  $A_1$ , where

$$A_1 = \begin{pmatrix} 9.5 & 5 & 5 & -6 \\ -1 & 20 & 1 & 2 \\ -2 & 1 & 5 & 0 \\ -2 & 1 & 7 & 12 \end{pmatrix}.$$

Then  $N_1(A_1) = \{1\}$ ,  $N_2(A_1) = \{2, 3, 4\}$ . By Theorem 1, we have

$$|a_{11}| - P_1(A_1) = |a_{11}| - \sum_{j \in N_2, j \neq 1} \frac{r_j(A_1)}{|a_{jj}|} |a_{1j}| = 0.5,$$

and

$$|a_{22}| - r_2(A_1) = 16, \quad |a_{33}| - r_3(A_1) = 2, \quad |a_{44}| - r_4(A_1) = 3.$$

Thus  $\|A_1^{-1}\|_{\infty} \leq \max \left\{ \frac{1}{0.5}, \frac{1}{2} \right\} = 2$ .



Then  $N_1(A_3) = \{2, n-1\}$ ,  $N_2(A_3) = \{1, 3, \dots, n-2, n\}$ . By Theorem 1, for  $i \in N_1(A_3)$ , we have

$$|a_{22}| - P_2(A_3) = |a_{22}| - \left( \sum_{j \in N_1(A_3), j \neq 2} |a_{2j}| + \sum_{j \in N_2(A_3), j \neq 2} \frac{r_j(A_3)}{|a_{jj}|} |a_{2j}| \right) = 0.4,$$

$$|a_{n-1, n-1}| - P_{n-1}(A_3) = |a_{n-1, n-1}| - \left( \sum_{j \in N_1(A_3), j \neq n-1} |a_{n-1, j}| + \sum_{j \in N_2(A_3), j \neq n-1} \frac{r_j(A_3)}{|a_{jj}|} |a_{n-1, j}| \right) = 0.4.$$

For  $i \in N_2(A_3)$ , we get

$$|a_{ii}| - \sum_{j \neq i} |a_{ij}| = 1, \quad i = 1, n,$$

$$|a_{ii}| - \sum_{j \neq i} |a_{ij}| = 0.8, \quad i = 3, 4, \dots, n-2.$$

Thus  $\|A_3^{-1}\|_{\infty} \leq \max\left\{\frac{1}{0.4}, \frac{1}{0.8}\right\} = 2.5000$ .

#### 4. Error bound for the linear complementarity problems involving $SDD_1$ matrices

In this section, new error bound of  $LCP(A, q)$  is presented when  $A$  is an  $SDD_1$  matrix.

**Corollary 1.** Let  $A = (a_{ij}) \in R^{n \times n}$ ,  $n \geq 2$ , be an  $SDD_1$  matrix with positive diagonal entries, and  $A_D = I - D + DA$ , where  $D = \text{diag}(d_i)$  with  $0 \leq d_i \leq 1$ . Then

$$\max_{d \in [0, 1]^n} \|A_D^{-1}\|_{\infty} \leq \max \left\{ \frac{1}{\min_{i \in N_1(A)} \{|a_{ii}| - P_i(A), 1\}}, \frac{1}{\min_{i \in N_2(A)} \left\{ |a_{ii}| - \sum_{j \neq i} |a_{ij}|, 1 \right\}} \right\},$$

where  $N_1(A) = \{i \in N \mid |a_{ii}| \leq r_i(A)\}$ ,  $N_2(A) = \{i \in N \mid |a_{ii}| > r_i(A)\}$ .

*Proof.* Since  $A$  is an  $SDD_1$  matrix, by Lemma 4,  $A_D = I - D + DA$  is also an  $SDD_1$  matrix, and

$$A_D = \begin{cases} 1 - d_i + d_i a_{ii}, & i = j, \\ d_i a_{ij}, & i \neq j. \end{cases}$$

By Theorem 1, we obtain

$$\|A_D^{-1}\|_{\infty} \leq \max \left\{ \frac{1}{\min_{i \in N_1(A_D)} (|\bar{a}_{ii}| - P_i(A_D))}, \frac{1}{\min_{i \in N_2(A_D)} \left( |\bar{a}_{ii}| - \sum_{i \neq j} |\bar{a}_{ij}| \right)} \right\}.$$

By Lemma 3, for  $i \in N_1(A)$ , it holds that

$$\frac{1}{|\bar{a}_{ii}| - P_i(A_D)} = \frac{1}{1 - d_i + d_i a_{ii} - \left( d_i \sum_{j \in N_1(A) \setminus \{i\}} |a_{ij}| + \sum_{j \notin N_1(A) \cup \{i\}} \frac{d_j r_j(A)}{1 - d_j + d_j a_{jj}} |d_i a_{ij}| \right)}$$

$$\begin{aligned} &\leq \frac{1}{1 - d_i + d_i a_{ii} - d_i \left( \sum_{j \in N_1(A) \setminus \{i\}} |a_{ij}| + \sum_{j \notin N_1(A) \cup \{i\}} \frac{r_j(A)}{a_{jj}} |a_{ij}| \right)} \\ &= \frac{1}{1 - d_i + d_i (|a_{ii}| - P_i(A))} \leq \frac{1}{\min \{|a_{ii}| - P_i(A), 1\}}. \end{aligned}$$

For  $i \in N_2(A)$ , by Lemma 3, we get

$$\frac{1}{|\bar{a}_{ii}| - \sum_{i \neq j} |\bar{a}_{ij}|} = \frac{1}{1 - d_i + d_i a_{ii} - \sum_{j \neq i} |d_i a_{ij}|} \leq \frac{1}{\min \left\{ |a_{ii}| - \sum_{j \neq i} |a_{ij}|, 1 \right\}}.$$

The proof is complete.  $\square$

**Example 4.** Consider the bound for  $\max_{d \in [0,1]^n} \|(I - D + DA_4)^{-1}\|_\infty$  of an  $SDD_1$  matrix  $A_4$ , where

$$A_4 = \begin{pmatrix} 16 & -8 & 4.1 & 8 \\ 0 & 8 & 3.1 & 1 \\ -8 & 8 & 20 & 8 \\ 1 & 1.2 & 5 & 8 \end{pmatrix}.$$

Then  $N_1(A_4) = \{1, 3\}$ ,  $N_2(A_4) = \{2, 4\}$ . By Corollary 1, we have

$$a_{11} - P_1(A_4) = a_{11} - \left( |a_{13}| + \frac{r_2(A_4)}{|a_{22}|} |a_{12}| + \frac{r_4(A_4)}{|a_{44}|} |a_{14}| \right) = 0.6,$$

$$a_{33} - P_3(A_4) = a_{33} - \left( |a_{31}| + \frac{r_2(A_4)}{|a_{22}|} |a_{32}| + \frac{r_4(A_4)}{|a_{44}|} |a_{34}| \right) = 0.7,$$

and

$$|a_{22}| - \sum_{j \neq 2} |a_{2j}| = 3.9, \quad |a_{44}| - \sum_{j \neq 4} |a_{4j}| = 0.8.$$

So  $\max_{d \in [0,1]^4} \|(I - D + DA_4)^{-1}\|_\infty \leq \max \left\{ \frac{1}{0.6}, \frac{1}{0.8} \right\} = 1.25$ .

**Example 5.** Consider the bound for  $\max_{d \in [0,1]^n} \|(I - D + DA_5)^{-1}\|_\infty$  of an  $SDD_1$  matrix  $A_5$ , where

$$A_5 = \begin{pmatrix} 15 & 2 & 1 & 0.5 & 1 & 1 & 1.2 & 2 & 0 & 1 \\ 3 & 25 & -2.1 & 2 & 6 & 1 & 3 & 2 & -5 & 0.5 \\ 2.5 & 1 & 12 & 1.5 & 0.5 & 2 & 1 & -1.2 & 1.3 & 1.6 \\ 2 & 3 & 1 & 20 & 4 & 1 & 1.5 & 0 & 0 & 1 \\ -2 & 1 & 3 & 4 & 20 & 1 & 2.1 & 1 & 1.2 & 5 \\ 1.5 & 0 & 1 & 4 & 3 & 23 & 1 & 5 & 7 & 1 \\ 5 & 1 & 6 & 2 & 3 & 2 & 25 & 1 & 5 & 1 \\ 1 & 1.2 & 3 & 2 & 5 & 6 & 4 & 26 & 1 & 2 \\ 3 & 5 & 7 & 6 & 4 & 3 & 1 & 2 & 35 & 1 \\ 2 & 1 & 1 & 2 & 3 & 1 & 1 & 2 & 2 & 17 \end{pmatrix}.$$





## 5. Error bound for linear complementarity problems involving $SDD_1$ - $B$ matrices

In this section, a new error bound of  $LCP(A, q)$  is presented when  $A$  is an  $SDD_1$ - $B$  matrix.

**Corollary 2.** Let  $A = (a_{ij}) \in \mathbb{R}^{n \times n}$ ,  $n \geq 2$ , be an  $SDD_1$ - $B$  matrix with positive diagonal entries, and let  $B^+ = (b_{ij})$  be the matrix as (2.1). Denote  $A_D = I - D + DA$ , where  $D = \text{diag}(d_i)$  with  $0 \leq d_i \leq 1$ . Then

$$\max_{d \in [0,1]^n} \|(I - D + DA)^{-1}\|_{\infty} \leq (n-1) \max \left\{ \frac{1}{\min_{i \in N_1(B^+)} \{|b_{ii}| - P_i(B^+), 1\}}, \frac{1}{\min_{i \in N_2(B^+)} \left\{ |b_{ii}| - \sum_{i \neq j} |b_{ij}|, 1 \right\}} \right\},$$

where

$$\begin{aligned} N_1(B^+) &= \{i \in N \mid |b_{ii}| \leq r_i(B^+)\}, \\ N_2(B^+) &= \{i \in N \mid |b_{ii}| > r_i(B^+)\}, \\ P_i(B^+) &= \sum_{j \in N_1(B^+)/\{i\}} |b_{ij}| + \sum_{j \notin N_1(B^+) \cup \{i\}} \frac{r_j(B^+)}{|b_{jj}|} |b_{ij}|. \end{aligned}$$

*Proof.* By the assumption,  $A$  is an  $SDD_1$ - $B$  matrix and  $A = B^+ + C$  is given as (2.1),  $B^+$  is an  $SDD_1$  matrix with positive diagonal entries. Thus for each diagonal matrix  $D = \text{diag}(d_i)$  with  $0 \leq d_i \leq 1$  ( $i \in N$ ), we have

$$A_D = I - D + DA = (I - D + DB^+) + DC = B_D^+ + C_D,$$

where  $B_D^+ = (\bar{b}_{ij}) = I - D + DB^+$  and  $C_D = DC$ . Similar to the proof of Theorem 2.2 in [24], we obtain

$$\|A_D^{-1}\|_{\infty} \leq \left\| [I + (B_D^+)^{-1}C_D]^{-1} \right\|_{\infty} \cdot \|(B_D^+)^{-1}\|_{\infty} \leq (n-1) \|(B_D^+)^{-1}\|_{\infty}.$$

We now bound  $\|(B_D^+)^{-1}\|_{\infty}$ , notice that  $B^+$  is an  $SDD_1$  matrix, by Lemma 4,  $B_D^+ = I - D + DB^+$  is also an  $SDD_1$  matrix. Hence, by Theorem 1, it holds that

$$\|(B_D^+)^{-1}\|_{\infty} \leq \left\{ \frac{1}{\min_{i \in N_1(B_D^+)} (|\bar{b}_{ii}| - P_i(B_D^+))}, \frac{1}{\min_{i \in N_2(B_D^+)} (|\bar{b}_{ii}| - \sum_{j \neq i} |\bar{b}_{ij}|)} \right\}.$$

By Lemma 3, for  $i \in N_1(B^+)$ , it holds that

$$\begin{aligned} \frac{1}{|\bar{b}_{ii}| - P_i(B_D^+)} &= \frac{1}{1 - d_i + d_i|b_{ii}| - \left( d_i \sum_{j \in N_1(B^+)/\{i\}} |b_{ij}| + \sum_{j \notin N_1(B^+) \cup \{i\}} \frac{d_j r_j(B^+)}{1 - d_j + d_j b_{jj}} |d_i b_{ij}| \right)} \\ &\leq \frac{1}{1 - d_i + d_i|b_{ii}| - d_i \left( \sum_{j \in N_1(B^+)/\{i\}} |b_{ij}| + \sum_{j \notin N_1(B^+) \cup \{i\}} \frac{r_j(B^+)}{|b_{jj}|} |b_{ij}| \right)} \end{aligned}$$

$$\begin{aligned}
&= \frac{1}{1 - d_i + d_i(|b_{ii}| - P_i(B^+))} \\
&\leq \frac{1}{\min\{|b_{ii}| - P_i(B^+), 1\}}.
\end{aligned}$$

For  $i \in N_2(B^+)$ , we have

$$\frac{1}{|\bar{b}_{ii}| - \sum_{j \neq i} |\bar{b}_{ij}|} = \frac{1}{1 - d_i + d_i|b_{ii}| - \sum_{j \neq i} |d_i b_{ij}|} \leq \frac{1}{\min\left\{|b_{ii}| - \sum_{j \neq i} |b_{ij}|, 1\right\}}.$$

Therefore, we derive

$$\|(B_D^+)^{-1}\|_\infty \leq \left\{ \frac{1}{\min_{i \in N_1(B^+)} \{|b_{ii}| - P_i(B^+), 1\}}, \frac{1}{\min_{i \in N_2(B^+)} \left\{|b_{ii}| - \sum_{j \neq i} |b_{ij}|, 1\right\}} \right\}.$$

The proof is complete. □

**Example 7.** Consider the bound for  $\max_{d \in [0,1]^n} \|(I - D + DA_7)^{-1}\|_\infty$  of an  $SDD_1$ - $B$  matrix  $A_7$ , where

$$A_7 = \begin{pmatrix} 16 & -8 & -4 & -8 \\ 5 & 13 & 2 & 4 \\ -8.5 & -8 & 20 & -8 \\ -1 & -1 & -5.4 & 8 \end{pmatrix}, \quad B^+ = \begin{pmatrix} 16 & -8 & -4 & -8 \\ 0 & 8 & -3 & -1 \\ -8.5 & -8 & 20 & -8 \\ -1 & -1 & -5.4 & 8 \end{pmatrix}.$$

Then  $N_1(A_7) = \{1, 3\}$ ,  $N_2(A_7) = \{2, 4\}$ . By Corollary 2, we have

$$b_{11} - P_1(B^+) = b_{11} - \left( |b_{13}| + \frac{r_2(B^+)}{|b_{22}|} |b_{12}| + \frac{r_4(B^+)}{|b_{44}|} |b_{14}| \right) = 1,$$

$$b_{33} - P_3(B^+) = b_{33} - \left( |b_{31}| + \frac{r_2(B^+)}{|b_{22}|} |b_{32}| + \frac{r_4(B^+)}{|b_{44}|} |b_{43}| \right) = 0.5,$$

and

$$|b_{22}| - \sum_{j \neq 2} |b_{2j}| = 4, \quad |b_{44}| - \sum_{j \neq 4} |b_{4j}| = 0.6.$$

So

$$\max_{d \in [0,1]^4} \|(I - D + DA_7)^{-1}\|_\infty \leq (4 - 1) \max\left\{\frac{1}{0.5}, \frac{1}{0.6}\right\} = 6.$$

**Example 8.** Consider the bound for  $\max_{d \in [0,1]^n} \|(I - D + DA_8)^{-1}\|_\infty$  of an  $SDD_1$ - $B$  matrix  $A_8$ , where

$$A_8 = \begin{pmatrix} 8 & -2 & -1 & -1 \\ 4 & 13 & 4 & 5 \\ -8 & -8 & 15 & -8 \\ -4 & -4 & -2 & 6 \end{pmatrix}, \quad B^+ = \begin{pmatrix} 8 & -2 & -1 & -1 \\ -1 & 8 & -1 & 0 \\ -8 & -8 & 15 & -8 \\ -4 & -4 & -2 & 6 \end{pmatrix}.$$



For  $i \in N_2(B^+)$ , we have

$$|b_{11}| - \sum_{j \neq 1} |b_{1j}| = 0.8, \quad |b_{nn}| - \sum_{j \neq n} |b_{nj}| = 0.9,$$

$$|a_{ii}| - \sum_{j \neq i} |a_{ij}| = 0.7, \quad i = 3, \dots, n-2.$$

So

$$\max_{d \in [0,1]^n} \|(I - D + DA_9)^{-1}\|_{\infty} \leq (n-1) \max \left\{ \frac{1}{0.42}, \frac{1}{0.7} \right\} = 2.381(n-1).$$

## 6. A lower bound of the smallest singular value for $SDD_1$ matrices

The singular values of a matrix  $A$  are the eigenvalues of  $(AA^*)^{\frac{1}{2}}$ , and are denoted as

$$\sigma_1(A) \geq \sigma_2(A) \geq \dots \geq \sigma_n(A) \geq 0.$$

The smallest singular value plays a special role in expressing properties of matrices, see [25,26] for details. The singular value  $\sigma_n(A)$  indicates not only whether  $A$  is nonsingular, but also how far (in the spectral norm) from the singular matrices  $A$  is. In addition, it is also a key ingredient in the spectral condition number  $\sigma_1(A)/\sigma_n(A)$ , as  $\sigma_1(A) \leq \sqrt{\|A\|_1 \|A\|_{\infty}}$  is a simple upper of  $\sigma_1(A)$  and hence if a lower bound of  $\sigma_n(A)$  is obtained, then one can give an upper bound for  $\sigma_1(A)/\sigma_n(A)$ . The spectral condition number  $\sigma_1(A)/\sigma_n(A)$  is commonly used in studying numerical calculations involving  $A$  [27]. Hence, lower bounds for the smallest singular value of matrices are of interest. Based on Theorem 1, by a similar proof to that of Theorem 7 in [28], a new lower bound for  $\sigma_n(A)$  can be obtained.

**Theorem 2.** Let  $A = (a_{ij}) \in C^{n \times n}$ ,  $n \geq 2$ . If  $A$  and its transpose  $A^T$  are all  $SDD_1$  matrices with poistive diagonal entries, then

$$\sigma_n(A) \geq \sqrt{\frac{1}{Bnd_{S,C}(A) \cdot Bnd_{S,C}(A^T)}},$$

where

$$Bnd_{S,C}(A) = \max \left\{ \frac{1}{\min_{i \in N_1(A)} (|a_{ii}| - P_i(A))}, \frac{1}{\min_{i \in N_2(A)} \left( |a_{ii}| - \sum_{j \neq i} |a_{ij}| \right)} \right\},$$

and

$$N_1(A) = \{i \in N \mid |a_{ii}| \leq r_i(A)\},$$

$$N_2(A) = \{i \in N \mid |a_{ii}| > r_i(A)\},$$

$$P_i(A) = \sum_{j \in N_1(A) \setminus \{i\}} |a_{ij}| + \sum_{j \notin N_1(A) \cup \{i\}} \frac{r_j(A)}{a_{jj}} |a_{ij}|.$$

*Proof.* Since  $A^T$  is an  $SDD_1$  matrix, it follows that

$$\|A^{-1}\|_1 = \|(A^{-1})^T\|_\infty = \|(A^T)^{-1}\|_\infty \leq Bnd_{S.C}(A^T).$$

By the well-known inequality (see[15]):

$$\|A^{-1}\|_2^2 \leq \|A^{-1}\|_1 \|(A^{-1})\|_\infty,$$

it holds that

$$\|A^{-1}\|_2^{-1} \geq \sqrt{\frac{1}{Bnd_{S.C}(A) \cdot Bnd_{S.C}(A^T)}}.$$

From  $\|A^{-1}\|_2^{-1} = \sigma_n(A)$ , the conclusion follows.  $\square$

**Example 10.** Consider the  $SDD_1$  matrix,

$$A_{10} = (a_{ij}) = \begin{pmatrix} 8 & 2 & 1 & 1 \\ 1 & 8 & 1 & 0 \\ 8 & 8 & 15 & 4 \\ 4 & 4 & 2 & 6 \end{pmatrix}, \quad A_{10}^T = (\alpha_{ij}) = \begin{pmatrix} 8 & 1 & 8 & 4 \\ 2 & 8 & 8 & 4 \\ 1 & 1 & 15 & 2 \\ 1 & 0 & 4 & 6 \end{pmatrix}.$$

Obviously,  $N_1(A_{10}) = \{3, 4\}$ ,  $N_2(A_{10}) = \{1, 2\}$ ,  $N_1(A_{10}^T) = \{1, 2\}$  and  $N_2(A_{10}^T) = \{3, 4\}$ . By Theorem 2, we have

$$|a_{11}| - \sum_{j \neq 1} |a_{1j}| = 4, \quad |a_{22}| - \sum_{j \neq 2} |a_{2j}| = 6,$$

$$a_{33} - P_3(A_{10}) = a_{33} - \left( |a_{34}| + \frac{r_1(A_{10})}{|a_{11}|} |a_{31}| + \frac{r_2(A_{10})}{|b_{22}|} |a_{32}| \right) = 5,$$

$$a_{44} - P_4(A_{10}) = a_{44} - \left( |a_{43}| + \frac{r_1(A_{10})}{|a_{11}|} |a_{41}| + \frac{r_2(A_{10})}{|b_{22}|} |a_{42}| \right) = 1,$$

and

$$\alpha_{11} - P_1(A_{10}^T) = \alpha_{11} - \left( |\alpha_{12}| + \frac{r_3(A_{10}^T)}{|\alpha_{33}|} |\alpha_{13}| + \frac{r_4(A_{10}^T)}{|\alpha_{44}|} |\alpha_{14}| \right) = 1.5334,$$

$$\alpha_{22} - P_2(A_{10}^T) = \alpha_{22} - \left( |\alpha_{21}| + \frac{r_3(A_{10}^T)}{|\alpha_{33}|} |\alpha_{23}| + \frac{r_4(A_{10}^T)}{|\alpha_{44}|} |\alpha_{24}| \right) = 0.5334,$$

$$|\alpha_{33}| - \sum_{j \neq 3} |\alpha_{3j}| = 11, \quad |\alpha_{44}| - \sum_{j \neq 4} |\alpha_{4j}| = 1.$$

So

$$Bnd_{S.C}(A_{10}) = 1, \quad Bnd_{S.C}(A_{10}^T) = 1.8748,$$

$$\sigma_n(A_{10}) \geq \sqrt{\frac{1}{Bnd_{S.C}(A_{10}) \cdot Bnd_{S.C}(A_{10}^T)}} = 0.7303.$$

## 7. Conclusions

Based on the fact that  $I - D + DA$  is an  $SDD_1$  matrix when  $A$  is an  $SDD_1$  matrix, we present an upper bound for the inverse of  $SDD_1$  matrix and give alternative bounds for  $\max_{d \in [0,1]^n} \|(I - D + DA)^{-1}\|_\infty$  when  $A$  is an  $SDD_1$  matrix or an  $SDD_1$ - $B$  matrix. A lower bound for smallest singular value is also provided. Also we illustrate the results with numerical examples. Finding computable global error bounds of the extended vertical linear complementarity problems for  $SDD_1$  matrices and  $SDD_1$ - $B$  matrices is an interesting problem. It is worth studying in the future.

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## Conflict of interest

The authors declare that they have no competing interests.

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