# On the asymptotic optimality of error bounds for some linear complementarity problems

M. García-Esnaola · J. M. Peña

Received: date / Accepted: date

**Abstract** We introduce strong *B*-matrices and strong *B*-Nekrasov matrices, for which some error bounds for linear complementarity problems are analyzed. In particular, it is proved that the bounds of [5] and of [8] are asymptotically optimal for strong B-matrices and strong B-Nekrasov matrices, respectively. Other comparisons with a bound of [11] are performed.

**Keywords** Error bounds  $\cdot$  Linear complementarity problems  $\cdot$  *B*-matrices  $\cdot$  *B*-Nekrasov matrices  $\cdot$  *P*-matrices

Mathematics Subject Classification (2000) 90C33 · 90C31 · 65G50 · 15A48

### 1 Introduction

The linear complementarity problem (LCP(M,q)) looks for a vector  $x \in \mathbf{R}^n$  such that

$$Mx + q \ge 0, \quad x \ge 0, \quad x^T (Mx + q) = 0,$$
 (1)

where *M* is an  $n \times n$  real matrix and  $q \in \mathbf{R}^n$ .

It is well known that this problem has a unique solution if and only if M has positive principal minors (i.e., M is a P-matrix). Important applications of this problem can be seen in [2].

Research partially supported by the Spanish Research Grant MTM2015-65433-P (MINECO/FEDER) and by Gobierno de Aragón and Fondo Social Europeo.

M. García-Esnaola Departamento de Matemática Aplicada Universidad de Zaragoza Tel.: (+34) 976 761985 E-mail: gesnaola@unizar.es

Juan Manuel Peña Departamento de Matemática Aplicada Universidad de Zaragoza Tel.: (+34) 976 761129

E-mail: jmpena@unizar.es

Error bounds for LCP of *P*-matrices have been studied (cf., [1], [12]). For particular subclasses of *P*-matrices, the bounds can be refined: see [1] and [6] for the subclass of *H*-matrices with positive diagonal entries or [5] and [11] for the subclass of *B*-matrices. For classes of matrices containing *B*-matrices, error bounds for the LCP have also been obtained ([4], [7], [8], [9]). Among these classes of matrices we can mention the *B*-Nekrasov matrices, which will be also considered in this paper.

In some examples, the bound of [5] for *B*-matrices was improved by the bounds of [11]. We present and characterize in Section 2 a subclass of *B*-matrices called *strong B-matrices*, for which the bound of [5] is linear and asymptotically optimal (see Theorem 1) and for which the bound of [11] is worse than or equal to quadratic (see Theorem 3). At the end of Section 2, we also include a family of matrices that are simultaneously strong *B*-matrices and *H*-matrices and for which our bound is 1 and the bound of formula (2.4) of [1] (valid for *H*-matrices with positive diagonal entries) is arbitrarily large. A final example in Section 3 shows that our bound of [5] can improve that of [11] even for *B*-matrices that are not strong *B*-matrices. Finally, Section 4 introduces the class of strong *B*-Nekrasov matrices, which contains strong *B*-matrices and is contained in the class of *B*-Nekrasov matrices. We provide an error bound for the LCP of a strong *B*-Nekrasov matrix that is asymptotically optimal.

## 2 A class of B-matrices with an asymptotically optimal bound

The class of *B*-matrices is a subclass of *P*-matrices presented in [13] and has been applied to eigenvalues localization problems ([13], [14]) and to linear complementarity problems ([5], [11]). We recall the definition of a *B*-matrix.

**Definition 1** A square real matrix  $M = (m_{ij})_{1 \le i,j \le n}$  is a B- matrix if it has positive row sums and all its off-diagonal entries are bounded above by the corresponding row means, that is, for each i = 1, ..., n,

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

Given a matrix  $M = (m_{ij})_{1 \le i,j \le n}$ , we define for each  $i = 1, ..., n, r_i^+ := \max\{0, m_{ij} | j \ne i\}$  and we can decompose M into the form  $M = B^+ + C$ , where

$$B^{+} = \begin{pmatrix} m_{11} - r_{1}^{+} & \cdots & m_{1n} - r_{1}^{+} \\ \vdots & & \vdots \\ \vdots & & \vdots \\ m_{n1} - r_{n}^{+} & \cdots & m_{nn} - r_{n}^{+} \end{pmatrix} \text{ and } C = \begin{pmatrix} r_{1}^{+} & \cdots & r_{1}^{+} \\ \vdots & & \vdots \\ \vdots & & \vdots \\ r_{n}^{+} & \cdots & r_{n}^{+} \end{pmatrix}.$$
 (2)

Then by Proposition 2.1 of [5], M is a B-matrix if and only if  $B^+ = (b_{ij})_{1 \le i,j \le n}$  is a strictly diagonally dominant matrix  $(|b_{ii}| > \sum_{j \ne i} |b_{ij}|, i = 1, ..., n)$  with positive diagonal entries. In this paper we introduce the following subclass of B-matrices by requiring a stronger diagonal dominant property to  $B^+$ .

**Definition 2** Let M be a B-matrix and let us consider  $M = B^+ + C$  as in (2). Given  $B^+ =: (b_{ij})_{1 \le i,j \le n}$ , we define, for each  $i = 1, \ldots, n$ ,  $\beta_i := b_{ii} - \sum_{j \ne i} |b_{ij}|$ . Then we say that M is a strong B-matrix if  $\beta_i > 1$  for all  $i = 1, \ldots, n$ .

Given a complex matrix  $M = (m_{ij})_{1 \le i,j \le n}$ , its *comparison* matrix  $\tilde{M} = (\tilde{m}_{ij})_{1 \le i,j \le n}$  is given by  $\tilde{m}_{ij} := -|m_{ij}|$  if  $i \ne j$  and  $\tilde{m}_{ii} := |m_{ii}|$  for  $i = 1, \ldots, n$ . Let us recall that M is an H-matrix if  $\tilde{M}$  is a nonsingular M-matrix, that is, if  $\tilde{M}^{-1}$  is nonnegative. Error bounds for LCP with H-matrices cannot be applied to LCP with strong B-matrices because a strong B-matrix is not necessarily an H-matrix, as the following example shows.

Example 1 Let us consider the matrix

$$M = \begin{pmatrix} 61.1 & 30 & 20 & 10 \\ -20 & 37.5 & 0 & -16 \\ 0 & -40 & 51.5 & -10 \\ 50 & 50 & 10 & 91.5 \end{pmatrix}.$$

The decomposition  $M = B^+ + C$  with  $B^+$  and C given in (2) leads to

Observe that M is a B-matrix and even a strong B-matrix. However, M is not an H-matrix because its comparison matrix  $\tilde{M}$  has an inverse with nonpositive entries.

The following result characterizes strong *B*-matrices.

**Proposition 1**  $M := (m_{ij})_{1 \le i,j \le n}$  is a strong B-matrix if and only if for each i = 1, ..., n,

$$\sum_{j=1}^{n} m_{ij} > 1 \quad and \quad \frac{1}{n} \left( \sum_{j=1}^{n} m_{ij} - 1 \right) > m_{ik} \quad \forall k \neq i.$$
 (3)

*Proof* Let us assume first that M is a strong B-matrix. Then  $\beta_i = b_{ii} - \sum_{j \neq i} |b_{ij}| > 1$ , for all  $i = 1, \ldots, n$ , and  $B^+ = (b_{ij})_{1 \leq i,j \leq n}$  is given in (2). Taking into account that  $r_i = \max\{0, m_{ij} | j \neq i\}$ , we have that  $r_i^+ \geq m_{ij}$  for all  $j \neq i$ , and we derive from the previous formula

$$1 < m_{ii} - r_i^+ - \sum_{j \neq i} (r_i^+ - m_{ij}) = \sum_{j=1}^n m_{ij} - nr_i^+$$
 (4)

and so we conclude that  $\sum_{j=1}^{n} m_{ij} > 1 + nr_i^+ \ge 1$ . From (4) we also have  $\sum_{j=1}^{n} m_{ij} - 1 > nr_i^+ \ge nm_{ik}$  for all  $k \ne i$  and (3) holds.

Let us now assume that (3) holds. Clearly, (3) implies that M is a B-matrix. Since  $r_i^+ \ge m_{ij}$  for all  $j \ne i$ ,

$$\beta_i = m_{ii} - r_i^+ - \sum_{i \neq i} (r_i^+ - m_{ij}) = \sum_{i=1}^n m_{ij} - nr_i^+.$$
 (5)

If  $r_i^+=0$ , then by (5) and (3),  $\beta_i=\sum_{j=1}^n m_{ij}>1$ . Finally, if  $r_i^+\neq 0$ , then there exists  $k\neq i$  such that  $r_i^+=m_{ik}>0$ . Then, by (5) and (3) we obtain  $\beta_i=\sum_{j=1}^n m_{ij}-nm_{ik}=n(\frac{1}{n}\sum_{i=1}^n m_{ij}-m_{ik})>n(\frac{1}{n})=1$  and M is a strong B-matrix.

Example 2 Now we present some examples arising in practical applications. Given the tridiagonal  $n \times n$  matrix

$$M = \begin{pmatrix} b + \alpha \sin\left(\frac{1}{n}\right) & c \\ & a & b + \alpha \sin\left(\frac{2}{n}\right) & c \\ & \ddots & \ddots & \ddots \\ & & \ddots & \ddots & c \\ & & a & b + \alpha \sin(1) \end{pmatrix}, \tag{6}$$

the LCP(M,q) with various q in an interval vector arises from the finite difference method for free boundary problems (cf. [15], [1]). Observe that, for b > 3, a = c = -1,  $\alpha = 0$ , the corresponding family is formed by strong B-matrices and also for b = 3, a = c = -1,  $\alpha > 0$ . In fact, the decomposition (2) for these matrices is of the form  $M = B^+ + 0$  and  $B^+$  satisfies the properties of Definition 2. Besides, for any  $k_1, \ldots, k_n \ge 0$ , we can form matrices

$$C = \begin{pmatrix} k_1 & \cdots & k_1 \\ \vdots & & \vdots \\ \vdots & & \vdots \\ k_n & \cdots & k_n \end{pmatrix}$$

and M + C is again a strong B-matrix by Definition 2.

In this section we shall prove that the bound of Theorem 2.2 of [5] is asymptotically optimal for the class of strong *B*-matrices. For this purpose, we consider the following family of  $n \times n$  strong *B*-matrices ( $n \ge 2$ ),

$$M_{m} = \begin{pmatrix} m+k & m & \cdots & m \\ 1 & 1+k & \cdots & 1 \\ \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots \\ 1 & 1 & \cdots & 1+k \end{pmatrix}, \tag{7}$$

where m is a positive integer and k > 1. Observe that, if  $k \in (0,1]$ , then  $M_m$  is a B-matrix but it is not a strong B-matrix.

First, we recall some notations for the linear complementarity problem (1). Its solution  $x^*$  is unique if and only if M is a P-matrix. In this case, by Theorem 2.3 of [1],

$$||x-x^*||_{\infty} \le \max_{d \in [0,1]^n} ||(I-D+DM)^{-1}||_{\infty} ||r(x)||_{\infty},$$

where I is the identity matrix, D is the diagonal matrix  $\operatorname{diag}(d_i)$  with  $0 \le d_i \le 1$  for all  $i = 1, \ldots, n$ , and  $r(x) := \min(x, Mx + q)$ , where the min operator denotes the componentwise minimum of two vectors. If M is a B-matrix and  $\beta_i$ ,  $i = 1, \ldots, n$ , are defined as in Definition 2, let us denote by

$$\beta := \min_{i \in \{1, \dots, n\}} \{ \beta_i \}. \tag{8}$$

Then, by Theorem 2.2 of [5],

$$\max_{d \in [0,1]^n} \| (I - D + DM)^{-1} \|_{\infty} \le \frac{n-1}{\min\{\beta,1\}}. \tag{9}$$

**Theorem 1** For strong B-matrices, the bound (9) is asymptotically optimal and it is equal to n-1.

*Proof* First observe that for strong *B*-matrices the bound (9) is equivalent to

$$\max_{d \in [0,1]^n} ||(I - D + DM)^{-1}||_{\infty} \le n - 1, \tag{10}$$

because of (8) and  $\beta > 1$ . Let us consider the family of matrices  $M_m$  given by (7), with m a positive integer and k > 1, and the particular choice in the left side of (9) given by  $d = (1, ..., 1)^T$ , which corresponds to the diagonal matrix D = I. So, with this choice, we have the following inequality

$$\max_{d \in [0,1]^n} \| (I - D + DM_m)^{-1} \|_{\infty} \ge \| M_m^{-1} \|_{\infty}. \tag{11}$$

Observe that  $M_m$  can be written in the form (2) as  $M_m = K + ue^T$  where

$$K := \begin{pmatrix} k & 0 & \cdots & 0 \\ 0 & k & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & k \end{pmatrix}, \quad u := \begin{pmatrix} m \\ 1 \\ \vdots \\ \vdots \\ 1 \end{pmatrix}, \quad e := \begin{pmatrix} 1 \\ 1 \\ \vdots \\ \vdots \\ 1 \end{pmatrix}.$$

Then we have

$$M_m = K(I + K^{-1}ue^T) = K(I + u_k e^T),$$
 (12)

where  $u_k := (m/k, 1/k, ..., 1/k)^T$ . Then, since  $e^T u_k \neq -1$ , by the Sherman-Morrison formula (see formula (2.1.5) of page 65 of [10]), we obtain from (12)

$$M_m^{-1} = (I + u_k e^T)^{-1} K^{-1} = (I - \frac{u_k e^T}{1 + e^T u_k}) K^{-1}$$

and so

$$M_m^{-1} = \begin{pmatrix} 1 - \frac{m}{\frac{m+n+k-1}{-1}} & \frac{-m}{\frac{m+n+k-1}{-1}} & \cdots & \frac{-m}{\frac{m+n+k-1}{-1}} \\ \frac{-1}{m+n+k-1} & 1 - \frac{1}{\frac{m+n+k-1}{-1}} & \cdots & \frac{-1}{\frac{m+n+k-1}{-1}} \\ \vdots & \vdots & \vdots & \vdots \\ \frac{-1}{m+n+k-1} & \frac{-1}{m+n+k-1} & \cdots & 1 - \frac{1}{m+n+k-1} \end{pmatrix}.$$

Then  $\|M_m^{-1}\|_{\infty} = 1 - \frac{m}{m+n+k-1} + \frac{(n-1)m}{m+n+k-1} = \frac{n+(k-1)+(n-1)m}{m+n+k-1}$  and we derive

$$\lim_{m \to \infty} ||M_m^{-1}|| = n - 1. \tag{13}$$

By (11) and (13),  $\lim_{m\to\infty} \max_{d\in[0,1]^n} \|(I-D+DM_m)^{-1}\|_{\infty} = n-1$  and (10) is asymptotically optimal.

Let us illustrate the asymptotic optimality of the previous result with some particular values of n,k,m for the  $n\times n$  matrix  $M_m$  of (7). A lower bound for  $\max_{d\in[0,1]^n}\|(I-D+DM_m)^{-1}\|_{\infty}$  is  $\|M_m^{-1}\|_{\infty}$  by (11) and in the line previous to (13) we have seen that  $\|M_m^{-1}\|_{\infty}=1-\frac{m}{m+n+k-1}+\frac{(n-1)m}{m+n+k-1}=\frac{n+(k-1)+(n-1)m}{m+n+k-1}$ . For n=10 and k=2, we obtain that this lower bound is  $\frac{20}{12}$  for m=1,  $\frac{101}{21}$  for m=10,  $\frac{911}{111}$  for m=100 and  $\frac{9011}{1011}$  for m=1000, which shows the approximation to our upper bound n-1=9 of (10).

We now present a family of  $2 \times 2$  matrices of Example 2 that are simultaneously strong *B*-matrices and *H*-matrices. For these matrices our bound (10) is 1 and the bound of formula (2.4) of [1] (valid for *H*-matrices with positive diagonal entries) is arbitrarily large.

Example 3 Let us consider  $2 \times 2$  matrices of Example 2 with b = 4, a = c = -1,  $\alpha = 0$  and  $k_1 = k_2 = k > 1$ . So our matrices are of the form

$$M = \begin{pmatrix} 4+k & k-1 \\ k-1 & 4+k \end{pmatrix}.$$

These strong B-matrices are also H-matrices because the comparison matrix

$$\tilde{M} = \begin{pmatrix} 4+k & -k+1 \\ -k+1 & 4+k \end{pmatrix}$$

has nonnegative inverse

$$\tilde{M}^{-1} = \begin{pmatrix} \frac{4+k}{5(3+2k)} & \frac{k-1}{5(3+2k)} \\ \frac{k-1}{5(3+2k)} & \frac{4+k}{5(3+2k)} \end{pmatrix}.$$

Our bound (10) is n-1=1. Let us now consider the bound of (2.4) of [1]:

$$\max_{d \in [0,1]^n} \| (I - D + DM)^{-1} \|_{\infty} \le \| \tilde{M}^{-1} \max(\Lambda, I) \|_{\infty},$$

where  $\Lambda$  is the diagonal part of M ( $\Lambda := \operatorname{diag}(m_{ii})$ ) and  $\max(\Lambda, I) := \operatorname{diag}(\max\{m_{11}, 1\}, \dots, \max\{m_{nn}, 1\})$ . For our matrices M,  $\max_{d \in [0,1]^n} \|(I - D + DM)^{-1}\|_{\infty} = \frac{4}{5} + \frac{k}{5}$ , which is greater than 1 and can be arbitrarily large.

# 3 Comparisons with another recent bound

In [11] the authors provided for a B-matrix M an upper bound for  $\max_{d \in [0,1]^n} \| (I - D + DM)^{-1} \|_{\infty}$  different from (9) and they showed some examples where this bound (presented in Theorem 4 of [11]) improves (9). We are now going to prove that for strong B-matrices the bound (9) (or its equivalent form (10) as shown in Theorem 1) is better than the bound of Theorem 4 of [11]. We shall prove in Theorem 2 that the bound of Theorem 4 of [11] is worse than or equal to quadratic n(n-1), in contrast with our linear bound n-1 of (10). Previously we recall Theorem 4 of [11].

**Theorem 2** Let  $M = (m_{ij})_{1 \le i,j \le n}$  be a B- matrix with the form  $M = B^+ + C$ , where  $B^+ := (b_{ij})_{1 \le i,j \le n}$  and C are given in (2). Then

$$\max_{d \in [0,1]^n} \| (I - D + DM)^{-1} \|_{\infty} \le \frac{n-1}{\min{\{\bar{\beta}_1, 1\}}} + \sum_{i=2}^n \frac{n-1}{\min{\{\bar{\beta}_i, 1\}}} \prod_{j=1}^{i-1} (1 + \frac{1}{\bar{\beta}_j} \sum_{k=j+1}^n |b_{jk}|) =: b_n,$$
(14)

where for each i = 1, ..., n,

$$\bar{\beta}_i := b_{ii} - l_i(B^+) \sum_{j=i+1}^n |b_{ij}| \quad and \quad l_i(B^+) := \max_{i \le k \le n} \{ \frac{1}{|b_{kk}|} \sum_{j=i, j \ne k}^n |b_{kj}| \}.$$

Let us now provide a lower bound for the bound of Theorem 2 in the case of strong *B*-matrices.

**Theorem 3** Let  $M = (m_{ij})_{1 \le i,j \le n}$  be a strong B-matrix. Then the bound  $b_n$  of (14) satisfies

$$b_n \ge n(n-1) \tag{15}$$

7

*Proof* With the notations of Theorem 2, let us observe that for each  $i=1,\ldots,n,$   $l_i(B^+)<1$  (see (11) of [11]). Therefore  $\bar{\beta}_i\geq \beta_i$ , for each  $i=1,\ldots,n$ . Since M is a strong B-matrix, for  $i=1,\ldots,n,$   $\beta_i>1$ , which implies that  $\bar{\beta}_i>1$  and so,  $\min\{\bar{\beta}_i,1\}=1$ . Then, since  $\prod_{j=1}^{i-1}\left(1+\frac{1}{\bar{\beta}_j}\sum_{k=j+1}^n|b_{jk}|\right)\geq 1$ , we can deduce that  $b_n\geq n-1+\sum_{i=2}^n(n-1).1=n(n-1)$ .

Although theorems 1 and 3 only hold for strong B-matrices, the bound (9) can be sharp and even better than (14), also for B-matrices that are not strong B-matrices. In the following example we consider a B-matrix that is not strong and we shall compare the bound (9) with (14). Let us consider the matrix

$$M = \begin{pmatrix} 60.5 & 30 & 20 & 10 \\ -20 & 40 & 0 & -16 \\ 0 & -40 & 51 & -10 \\ 50 & 50 & 10 & 91 \end{pmatrix}.$$

The decomposition  $M = B^+ + C$  with  $B^+$  and C given in (2) leads to

Observe that M is a B-matrix (because  $B^+$  is strictly diagonally dominant with positive diagonal entries) and that it is not a strong B-matrix (because  $\beta = \beta_1 = 0.5$ ). One can check that the bound of Theorem 4 of [11] (that is, the bound  $b_n$  of (14)) is in this case

$$\max_{d \in [0,1]^n} ||(I - D + DM)^{-1}||_{\infty} \le b_n = 445.321.$$

Let us now compute our bound (9). Taking into account that n = 4 and  $\beta_1 = 0.5$ ,  $\beta_2 = 4$ ,  $\beta_3 = \beta_4 = 1$ , (and so,  $\beta = 0.5$  by (8)), we deduce that (9) is now 3/0.5 = 6.

# 4 Strong B-Nekrasov matrices and asymptotically optimal bounds

Let us recall the definition of a Nekrasov matrix. For this purpose, we need some notations. If  $M = (m_{ij})_{1 \le i,j \le n}$  is a complex matrix with  $m_{ii} \ne 0$  for all i = 1, ..., n, let us define

$$h_1(M) := \sum_{j \neq 1} |m_{1j}|, \ h_i(M) := \sum_{j=1}^{i-1} |m_{ij}| \frac{h_j(M)}{|m_{jj}|} + \sum_{j=i+1}^{n} |m_{ij}|, \quad i = 2, \dots, n.$$
 (16)

Then M is a *Nekrasov matrix* if  $|m_{ii}| > h_i(M)$  for all i = 1, ..., n (cf. [16]). It is well-known that Nekrasov matrices are nonsingular matrices.

The following definition recalls the concept of *B*-Nekrasov matrix, which was introduced in [8], and introduces the new definition of strong *B*-Nekrasov matrices.

**Definition 3** A real matrix M is a B-Nekrasov matrix if  $M = B^+ + C$ , where  $B^+$  and C are given in (2) and  $B^+$  is a Nekrasov Z-matrix with all diagonal entries positive. If  $B^+ - I$  is a Nekrasov Z-matrix with all diagonal entries positive, then we say that M is a strong B-Nekrasov matrix.

By Remark 1 of [8], *B*-matrices are *B*-Nekrasov matrices and, by Proposition 1 of [8], *B*-Nekrasov matrices are *P*-matrices. Since a strong *B*-matrix *M* can be written  $M = B^+ + C$  as in (2) and satisfies that  $B^+ - I$  is a strictly diagonally dominant matrix with positive diagonal entries, we can deduce that a strong *B*-matrix is a strong *B*-Nekrasov matrix.

The following result shows that a strong *B*-Nekrasov matrix is also a *B*-Nekrasov matrix.

## **Proposition 2** If M is a strong B-Nekrasov matrix, then M is also B-Nekrasov.

*Proof* It is sufficient to prove that if  $B^+ - I$  is a Nekrasov Z-matrix with positive diagonal entries, then  $B^+ = (b_{ij})_{1 \le i,j \le n}$  is a Nekrasov Z-matrix with positive diagonal entries. So, we assume that

$$b_{ii} - 1 > h_i(B^+ - I), \quad i = 1, \dots, n,$$
 (17)

and let us prove that

$$b_{ii} > h_i(B^+), \quad i = 1, \dots, n.$$
 (18)

Since  $b_{ii} > b_{ii} - 1$  for all i = 1, ..., n, in order to prove (18) from (17), we shall prove that

$$h_i(B^+ - I) \ge h_i(B^+), \quad i = 1, \dots, n.$$
 (19)

Let us prove (19) by induction on *i*. Since  $h_1(B^+ - I) = \sum_{j \neq 1} |b_{1j}| = h_1(B^+)$ , (19) holds for i = 1. Let us now assume that (19) holds for all j < i and let us prove it for *i*. Then

$$h_i(B^+ - I) = \sum_{j=1}^{i-1} |b_{ij}| \frac{h_j(B^+ - I)}{b_{jj} - 1} + \sum_{j=i+1}^{n} |b_{ij}|.$$

Taking into account the induction hypothesis and that  $b_{jj} - 1 < b_{jj}$ , we obtain

$$h_i(B^+ - I) \ge \sum_{i=1}^{i-1} |b_{ij}| \frac{h_j(B^+)}{b_{jj}} + \sum_{i=i+1}^{n} |b_{ij}| = h_i(B^+)$$

and the induction holds.

The converse of the previous proposition does not hold as the following example shows. Any matrix

$$M_k = \begin{pmatrix} k & -k+1 \\ -k+1 & k \end{pmatrix}, \quad k \ge 2,$$

is *B*-Nekrasov because  $M_k = B_k^+ + C_k$  with  $B_k^+ = M_k$  and  $C_k = 0$  and it can be checked that  $B_k = M_k$  is Nekrasov. However,  $M_k$  is not strong *B*-Nekrasov because

$$B_k^+ - I = \begin{pmatrix} k-1 & -k+1 \\ -k+1 & k-1 \end{pmatrix}$$

is singular and so,  $B_k^+ - I$  can not be Nekrasov.

It is well–known that a complex matrix A is an B-matrix if there exists a diagonal matrix  $W = \operatorname{diag}(w_i)$  such that AW is strictly diagonal dominant. Observe that the diagonal matrix W can be taken with positive diagonal entries (it is sufficient using  $\operatorname{diag}(|w_i|)$ ). It is known that a Nekrasov matrix is an B-matrix (see [16] and p. 5021 of [3]). So, given a B-Nekrasov matrix M, there exists a diagonal matrix W with positive diagonal entries such that  $B^+W$  is a strictly diagonally dominant Z-matrix (where  $B^+$  is given by (2)). Then this holds, in particular, for strong B-Nekrasov matrices. Since a strong B-matrix satisfies that  $B^+ - I$  is

a strictly diagonally dominant matrix with positive diagonal entries, we can deduce that a strong *B*-matrix is a strong *B*-Nekrasov matrix.

In Theorem 2 of [8], we obtained an error bound for the LCP of a B-Nekrasov matrix  $A = B^+ + C$  satisfying certain hypotheses that allowed us to construct a particular diagonal matrix W such that  $B^+W$  is a strictly diagonally dominant Z-matrix. As we have recalled in the previous paragraph, for any B-Nekrasov matrix  $A = B^+ + C$ , there exists a diagonal matrix W with positive diagonal entries such that  $B^+W$  is a strictly diagonally dominant Z-matrix. The same proof of Theorem 2 of [8] can be used to prove the following result, which does not require any additional hypothesis on the B-Nekrasov matrices.

**Theorem 4** Let  $M = (m_{ij})_{1 \le i,j \le n}$  be a B-Nekrasov matrix, let  $B^+$  be the matrix of (2) and let  $W = diag(w_i)$  be the diagonal matrix with positive diagonal entries such that  $\bar{B} := B^+W = (\bar{b}_{ij})_{1 \le i,j \le n}$  is a strictly diagonally dominant Z-matrix. Let  $\beta_i := \bar{b}_{ii} - \sum_{j \ne i} |\bar{b}_{ij}|$  and  $\delta_i := \frac{\beta_i}{w_i}$ , for  $i = 1, \ldots, n$  and  $\delta = min_i \{\delta_i\}$ . Then

$$\max_{d \in [0,1]^n} \| (I - D + DM)^{-1} \|_{\infty} \le \frac{(n-1) \max_i \{ w_i \}}{\min\{1, \delta\} \min_i \{ w_i \}}.$$
 (20)

Now, we shall provide an error bound for the LCP of a strong B-Nekrasov matrix M.

**Theorem 5** Let  $M = (m_{ij})_{1 \le i,j \le n}$  be a strong B-Nekrasov matrix, let  $B^+$  be the matrix of (2) and let  $W = diag(w_i)$  be the diagonal matrix with positive diagonal entries such that  $(B^+ - I)W$  is a strictly diagonally dominant Z-matrix. Then

$$\max_{d \in [0,1]^n} \| (I - D + DM)^{-1} \|_{\infty} \le \frac{(n-1) \max_i \{w_i\}}{\min_i \{w_i\}}.$$
 (21)

Moreover, the bound (21) is asymptotically optimal.

*Proof* Since M is a strong B-matrix,  $B^+ - I$  is a Nekrasov matrix and so it is an H-matrix. Therefore there exists a diagonal matrix W with positive diagonal entries such that  $(B^+ - I)W$  is a strictly diagonally dominant Z-matrix. Then  $B^+W$  is also a strictly diagonally dominant Z-matrix and, since a strong B-Nekrasov matrix is a B-Nekrasov matrix, we can apply Theorem 4 with  $\bar{B} := B^+W$  and (20) holds. Since  $(B^+ - I)W = \bar{B} - W$  is also a strictly diagonally dominant Z-matrix,  $\beta_i - w_i > 0$  for all  $i = 1, \ldots, n$ . This implies that  $\delta_i := \frac{\beta_i}{w_i} > 1$  for all  $i = 1, \ldots, n$ . If  $\delta = \min_i \{ \delta_i \}$ , then  $\min\{1, \delta\} = 1$  and so (20) becomes (21).

Now, let us consider the family of strong *B*-Nekrasov matrices  $M_m$  given by (7) with m a positive integer and k > 1, and the particular choice in the left side of (21) given by  $d = (1, ..., 1)^T$ , which corresponds to the diagonal matrix D = I. So, with this choice we have (11). Observe that if we write  $M = B^+ + C$  as in (2), then  $B^+$  is the diagonal matrix whose diagonal entries are equal to k. So,  $B^+ - I$  is a strictly diagonally dominant Z-matrix and we can choose the matrix W = I to obtain the formula (21). In this case the right side of (21) is n - 1. Finally, since (11) and (13) hold, we see that (21) is asymptotically optimal.

**Acknowledgements** The authors wish to thank the anonymous referees for their valuable suggestions to improve the paper. This work was partially supported by the Spanish Research Grant MTM2015-65433-P (MINECO/FEDER), Gobierno de Aragón and Fondo Social Europeo.

### References

- 1. X. Chen and S. Xiang, Computation of error bounds for P-matrix linear complementarity problems. Math. Program., Ser. A, 106 (2006), 513–525.
- 2. R. W. Cottle, J.-S. Pang and R. E. Stone, *The Linear Complementarity Problems*. Academic Press, Boston MA, 1992.
- L. Cvetković, P-F. Dai, K. Doroslovaški, Y-T. Li, Infinity norm bounds for the inverse of Nekrasov matrices. Appl. Math. Comput. 219 (2013), 5020–5024.
- P.-F. Dai, Error bounds for linear complementarity problems of DB-matrices. Linear Algebra Appl. 434 (2011), 830–840.
- 5. M. García-Esnaola and J. M. Peña, Error bounds for linear complementarity problems for B-matrices. Appl. Math. Lett. 22 (2009), 1071–1075.
- M. García-Esnaola and J. M. Peña, A comparison of error bounds for linear complementarity problems of H-matrices. Linear Algebra Appl. 433 (2010), 956–964.
- M. García-Esnaola and J. M. Peña, Error bounds for linear complementarity problems of B<sup>S</sup>-matrices. Appl. Math. Lett. 25 (2012), 1379–1383.
- 8. M. García-Esnaola and J. M. Peña, *B*-Nekrasov matrices and error bounds for linear complementarity problems. Numer. Algor. 72 (2016), 435–445.
- 9. M. García-Esnaola and J. M. Peña,  $B_{\pi}^{R}$ -matrices and error bounds for linear complementarity problems. Calcolo 54 (2017), 813–822.
- 10. G. H. Golub and C. F. Van Loan, *Matrix Computations* (Fourth edition). The Johns Hopkins University Press, Baltimore, 2013.
- 11. C. Li and Y. Li, Note on error bounds for linear complementarity problems for *B*-matrices. Appl. Math. Lett. 57 (2016), 108–113.
- 12. R. Mathias and J.-S. Pang, Error bounds for the linear complementarity problem with a *P*-matrix. Linear Algebra Appl. 132 (1990), 123–136.
- 13. J. M. Peña, A class of *P*-matrices with applications to the localization of the eigenvalues of a real matrix. SIAM J. Matrix Anal. Appl. 22 (2001), 1027–1037.
- 14. J. M. Peña, On an alternative to Gerschgorin circles and ovals of Cassini. Numer. Math. 95 (2003), 337–345.
- 15. U. Schäfer, An enclosure method for free boundary problems based on a linear complementarity problem with interval data. Numer. Funct. Anal. Optim. 22 (2001), 991–1011.
- 16. T. Szulc, Some remarks on a theorem of Gudkov, Linear Algebra Appl. 225 (1995), 221–235.