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## Complexity of Solving Linear Interval Equations

*It is proved that computing enclosures of solutions of linear interval equations with overestimation bounded by a polynomial in the system size is NP-hard.*

### 1. Introduction

Solving linear interval equations usually means computing enclosures. For a system of linear interval equations

$$A^I x = b^I \tag{1}$$

( $A^I$  square), *enclosure* is defined as an interval vector  $[y, \bar{y}]$  satisfying

$$X \subseteq [y, \bar{y}],$$

where  $X$  is the solution set:

$$X = \{x; Ax = b \text{ for some } A \in A^I, b \in b^I\}.$$

Various enclosure methods can be found in Alefeld and Herzberger [1]. If  $A^I$  is regular (i.e., each  $A \in A^I$  is nonsingular), then there exists the narrowest enclosure  $[\underline{x}, \bar{x}]$  given by

$$\begin{aligned} \underline{x}_i &= \min_X x_i, \\ \bar{x}_i &= \max_X x_i \end{aligned}$$

for each  $i$ . Computing  $[\underline{x}, \bar{x}]$  was proved to be NP-hard (Rohn and Kreinovich [7]; also, Kreinovich, Lakeyev and Noskov [4] for the rectangular case). In the main result of this paper we show that computing enclosures with overestimation bounded by a polynomial in the system size is NP-hard. The result holds true even for a very restricted class of systems (1) with  $A^I = [A_c - \Delta, A_c + \Delta]$  having nondegenerate interval coefficients in one row only and satisfying  $\varrho(|A_c^{-1}| \Delta) = 0$ . Hence, the problem of computing sufficiently narrow enclosures turns out to be more difficult than previously believed.

### 2. Preliminaries

A real symmetric  $n \times n$  matrix  $A = (a_{ij})$  is called an *MC-matrix* [5] if it is of the form

$$a_{ij} \begin{cases} = n & \text{if } i = j \\ \in \{0, -1\} & \text{if } i \neq j \end{cases}$$

( $i, j = 1, \dots, n$ ). In the proof of the main theorem we shall essentially utilize the following result ([6], Corollary 7) concerning the norm

$$\|A\|_{\infty,1} = \max\{\|Ax\|_1; \|x\|_{\infty} = 1\}$$

(where  $\|x\|_1 = \sum_i |x_i|$  and  $\|x\|_{\infty} = \max_i |x_i|$ ; see Golub and van Loan [3], p. 15):

**Proposition 1.** *Computing  $\|A\|_{\infty,1}$  is NP-hard for MC-matrices.*

Next we introduce a class of systems (1) of a special form. For each pair of rational numbers  $\varepsilon > 0$ ,  $\delta > 0$  we shall denote by  $H_{\varepsilon\delta}$  the family of systems of linear interval equations

$$A^I x = b^I$$

with  $A^I$  of the form

$$A^I = \begin{pmatrix} a & [-\varepsilon e^T, \varepsilon e^T] \\ 0 & A^{-1} \end{pmatrix}, \quad (2)$$

where  $a$  is a positive rational number,  $A$  is an  $n \times n$  MC-matrix ( $n$  arbitrary,  $n \geq 1$ ),  $e = (1, 1, \dots, 1)^T \in \mathbb{R}^n$  (i.e.,  $A^I$  is  $(n+1) \times (n+1)$ ), and

$$b^I = \begin{pmatrix} 0 \\ [-\delta e, \delta e] \end{pmatrix} \quad (3)$$

is an  $(n+1)$ -dimensional interval vector. If we write (2) as

$$A^I = [A_c - \Delta, A_c + \Delta],$$

then

$$A_c = \begin{pmatrix} a & 0^T \\ 0 & A^{-1} \end{pmatrix}$$

is nonnegative symmetric positive definite [5], the radius matrix

$$\Delta = \begin{pmatrix} 0 & \varepsilon e^T \\ 0 & 0 \end{pmatrix}$$

has nonzero coefficients in the first row only, and

$$|A_c^{-1}| \Delta = \begin{pmatrix} 0 & \frac{\varepsilon}{a} e^T \\ 0 & 0 \end{pmatrix},$$

hence

$$\varrho(|A_c^{-1}| \Delta) = 0.$$

Thus the interval matrix (2) is strongly regular (i.e.  $\varrho(|A_c^{-1}| \Delta) < 1$ ); problems with strongly regular interval matrices have been usually considered "tractable".

In order to be able to formulate a unifying complexity result, we introduce the following concept: *enclosure algorithm* is an algorithm which for each system  $A^I x = b^I$  with rational data (and square  $A^I$ ) in a finite number of steps either computes a rational enclosure, or fails (i.e., issues an error message). Failure of an enclosure algorithm may be caused by various reasons: 1) no enclosure exists since the solution set is unbounded (in case of a singular  $A^I$ ), 2) the algorithm cannot be continued (e.g. in case of the interval Gaussian algorithm), 3) the algorithm works under some condition only (e.g., strong regularity), 4) a prescribed number of steps has been reached, etc.

### 3. Main result

**Theorem 1.** *If  $P \neq NP$ , then for each polynomial-time enclosure algorithm and each rational  $\varepsilon > 0$ ,  $\delta > 0$  either (i), or (ii) holds:*

(i) *the algorithm fails for some system in  $H_{\varepsilon\delta}$ ,*

(ii) *for each rational  $\alpha > 0$  and each integer  $k \geq 0$  there exists a system of size  $n \geq 2$  in  $H_{\varepsilon\delta}$  for which the enclosure  $[\underline{y}, \bar{y}]$  computed by the algorithm satisfies*

$$\underline{y}_1 \leq \underline{x}_1 - \alpha n^k < \bar{x}_1 + \alpha n^k \leq \bar{y}_1. \quad (4)$$

**Remark 1.** 1) P and NP are the well-known complexity classes. The conjecture that  $P \neq NP$ , although unproved, is widely believed to be true (cf. Garey and Johnson [2]). 2) If the conjecture holds true, then each polynomial-time enclosure algorithm which works for at least one family  $H_{\varepsilon\delta}$  may produce arbitrarily large overestimations (4); hence, no (even arbitrarily bad) accuracy can be guaranteed to be achievable by a polynomial-time enclosure algorithm.

Proof. Assume to the contrary that there exists a polynomial-time enclosure algorithm, rational numbers  $\varepsilon > 0$ ,  $\delta > 0$ ,  $\alpha > 0$  and an integer  $k \geq 0$  such that for each system in  $H_{\varepsilon\delta}$  the algorithm computes an enclosure  $[\underline{y}, \bar{y}]$  satisfying either

$$\underline{x}_1 - \alpha n^k < \underline{y}_1$$

or

$$\bar{y}_1 < \bar{x}_1 + \alpha n^k,$$

where  $n$  is the system size. Let  $A$  be an arbitrary  $MC$ -matrix of size  $m$ . Let us construct an  $(m+1) \times (m+1)$  interval matrix

$$A^I = \begin{pmatrix} \frac{\varepsilon\delta}{\gamma} & [-\varepsilon e^T, \varepsilon e^T] \\ \mathbf{0} & A^{-1} \end{pmatrix},$$

where

$$\gamma = \alpha(m+1)^k,$$

and an  $(m+1)$ -dimensional interval vector

$$b^I = \begin{pmatrix} \mathbf{0} \\ [-\delta e, \delta e] \end{pmatrix},$$

and apply the algorithm to the system

$$A^I x = b^I \tag{5}$$

(which obviously belongs to  $H_{\varepsilon\delta}$ ) to compute an enclosure  $[\underline{y}, \bar{y}]$  which, according to the assumption, satisfies either

$$\underline{x}_1 - \gamma < \underline{y}_1 \tag{6}$$

or

$$\bar{y}_1 < \bar{x}_1 + \gamma. \tag{7}$$

This can be done in polynomial time. We shall prove that

$$\|A\|_{\infty,1} = \left\lceil \frac{1}{\gamma} \min\{-\underline{y}_1, \bar{y}_1\} \right\rceil \tag{8}$$

holds, where  $\lceil \dots \rceil$  denotes the integer part. Hence,  $\|A\|_{\infty,1}$  can be computed in polynomial time; but since this is an NP-hard problem (Proposition 1), P=NP will follow. To prove (8), first observe that the system (5) can be written as

$$\frac{\varepsilon\delta}{\gamma} x_1 + [-\varepsilon e^T, \varepsilon e^T] x' = \mathbf{0},$$

$$-\delta e \leq A^{-1} x' \leq \delta e,$$

where  $x' = (x_2, \dots, x_m)^T$ . Hence

$$\begin{aligned} \bar{x}_1 &= \frac{\gamma}{\varepsilon\delta} \max\{\varepsilon e^T |x'|; -\delta e \leq A^{-1} x' \leq \delta e\} \\ &= \gamma \max\{\|x''\|_1; -e \leq A^{-1} x'' \leq e\} \\ &= \gamma \max\{\|Ax''' \|_1; -e \leq x''' \leq e\} \\ &= \gamma \max\{\|Ax''' \|_1; \|x''' \|_\infty = 1\} \\ &= \gamma \|A\|_{\infty,1} \end{aligned}$$

and in a quite similar way,

$$\underline{x}_1 = -\gamma \|A\|_{\infty,1}.$$

Hence from (6) and (7) we obtain that either

$$-\frac{1}{\gamma} \underline{y}_1 < \|A\|_{\infty,1} + 1$$

or

$$\frac{1}{\gamma} \bar{y}_1 < \|A\|_{\infty,1} + 1$$

holds, in both the cases

$$\frac{1}{\gamma} \min\{-\underline{y}_1, \bar{y}_1\} < \|A\|_{\infty,1} + 1. \quad (9)$$

But since  $[\underline{y}_1, \bar{y}_1]$  encloses  $[\underline{x}_1, \bar{x}_1]$ , from  $\underline{y}_1 \leq \underline{x}_1$ ,  $\bar{x}_1 \leq \bar{y}_1$  we have

$$\|A\|_{\infty,1} \leq \frac{1}{\gamma} \min\{-\underline{y}_1, \bar{y}_1\}$$

which together with (9) gives

$$\|A\|_{\infty,1} \leq \frac{1}{\gamma} \min\{-\underline{y}_1, \bar{y}_1\} < \|A\|_{\infty,1} + 1. \quad (10)$$

However, the number

$$\|A\|_{\infty,1} = \max\{\|Ax\|_1; \|x\|_{\infty} = 1\} = \max\{\|Ax\|_1; x_j \in \{-1, 1\} \text{ for each } j\}$$

is integer for an *MC*-matrix  $A$  (which is integer by definition), hence from (10) we finally obtain that

$$\|A\|_{\infty,1} = \left\lceil \frac{1}{\gamma} \min\{-\underline{y}_1, \bar{y}_1\} \right\rceil,$$

which is (8). Hence,  $\|A\|_{\infty,1}$  can be computed in polynomial time for an *MC*-matrix  $A$ , which in view of Proposition 1 implies that  $P=NP$ . This concludes the proof by contradiction.

#### 4. Application: interval Gaussian algorithm

For each rational  $\varepsilon > 0, \delta > 0$ , the interval Gaussian algorithm with partial pivoting [1] (which is polynomial-time) is performable for each system in  $H_{\varepsilon\delta}$  since all the pivots are real and nonzero due to the special form of the system matrix (2). Hence, if  $P \neq NP$ , then arbitrarily large overestimations (4) may occur for arbitrarily narrow system matrices (2) and arbitrarily narrow right-hand sides (3).

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