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INVERSE INTERVAL MATRIX: A SURVEY*

JIRI ROHN[†] AND RAENA FARHADSEFAT[‡]

Abstract. Results on the inverse interval matrix, both theoretical and computational, are surveyed. Described are, among others, formulae for the inverse interval matrix, NP-hardness of its computation, various classes of interval matrices for which the inverse can be given explicitly, and closed-form formulae for an enclosure of the inverse.

Key words. Interval matrix, Inverse interval matrix, NP-hardness, Enclosure, Unit midpoint, Inverse sign stability, Nonnegative invertibility, Absolute value equation, Algorithm.

AMS subject classifications. 15A09, 65G20, 65G40.

1. Introduction. In our recent paper [20], we presented a survey of forty necessary and sufficient conditions for regularity of interval matrices. It is now followed by a survey of properties of the inverse interval matrix which is closely related to the previous topic because the inverse interval matrix is only defined for regular interval matrices.

After some preliminaries in Sections 2 and 3, the inverse interval matrix is defined in Section 4. Next we introduce matrices B_y defined for each ± 1 -vector y and demonstrate their use for inverse matrix representation (Theorem 7.1) and for establishing finite formulae for the inverse interval matrix (Theorem 8.1). Then we present Coxson's result [3] showing that computing the inverse interval matrix is NP-hard. In Section 10 we show that for an interval matrix with unit midpoint the inverse interval matrix can be given explicitly by simple formulae (Theorem 10.2). Explicit formulae for an enclosure of the inverse of a strongly regular interval matrix are given in Section 11. In the next four sections, we give explicit formulae for the interval inverse of interval matrices that are either inverse sign stable (Section 12), or are of inverse sign pattern (Section 13), or are nonnegative invertible (Section 14), or have uniform width (Section 15). In the last Section 16, we describe available software

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for computing the inverse interval matrix or its enclosure. The Appendix contains a MATLAB-like description of an algorithm for solving an absolute value equation which is used in Section 6 for computation of the matrices B_y .

2. Notation. We use the following notation. A_{ij} denotes the ij th entry, $A_{i\bullet}$ the i th row and $A_{\bullet j}$ the j th column of a matrix A . Matrix inequalities, as $A \leq B$ or $A < B$, are understood componentwise. $A \circ B$ denotes the Hadamard (entrywise) product of $A, B \in \mathbb{R}^{m \times n}$, i.e., $(A \circ B)_{ij} = A_{ij}B_{ij}$ for each i, j . The minimum (or maximum) matrix of a compact (in particular, finite) set of matrices X is defined componentwise, i.e.,

$$(\min\{A \mid A \in X\})_{ij} = \min\{A_{ij} \mid A \in X\},$$

$$(\max\{A \mid A \in X\})_{ij} = \max\{A_{ij} \mid A \in X\}$$

for each i, j . The absolute value of a matrix $A = (a_{ij})$ is defined by $|A| = (|a_{ij}|)$. For each matrix A we define its sign matrix $\text{sgn}(A)$ by

$$(\text{sgn}(A))_{ij} = \begin{cases} 1 & \text{if } A_{ij} \geq 0, \\ -1 & \text{if } A_{ij} < 0 \end{cases}$$

for each i, j . The same notation also applies to vectors that are considered one-column matrices. I is the unit matrix, e_j is the j th column of I , $e = (1, \dots, 1)^T$ is the vector of all ones, and $E = ee^T$ is the matrix of all ones. $Y_n = \{y \mid |y| = e\}$ is the set of all ± 1 -vectors in \mathbb{R}^n , so that its cardinality is 2^n . For each $y \in \mathbb{R}^n$ we denote

$$T_y = \text{diag}(y_1, \dots, y_n) = \begin{pmatrix} y_1 & 0 & \dots & 0 \\ 0 & y_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & y_n \end{pmatrix},$$

and $\rho(A)$ is the spectral radius of A .

3. Interval matrices. Given two $n \times n$ matrices A_c and Δ , $\Delta \geq 0$, the set of matrices

$$\mathbf{A} = \{A \mid |A - A_c| \leq \Delta\}$$

is called a (square) interval matrix with midpoint matrix A_c and radius matrix Δ . Since the inequality $|A - A_c| \leq \Delta$ is equivalent to $A_c - \Delta \leq A \leq A_c + \Delta$, we can also write

$$\mathbf{A} = \{A \mid \underline{A} \leq A \leq \overline{A}\} = [\underline{A}, \overline{A}],$$

where $\underline{A} = A_c - \Delta$ and $\overline{A} = A_c + \Delta$ are called the bounds of \mathbf{A} .

Given an $n \times n$ interval matrix \mathbf{A} , we define matrices

$$A_{yz} = A_c - T_y \Delta T_z \quad (3.1)$$

for each $y, z \in Y_n$. The definition implies that

$$(A_{yz})_{ij} = (A_c)_{ij} - y_i \Delta_{ij} z_j = \begin{cases} \overline{A}_{ij} & \text{if } y_i z_j = -1, \\ \underline{A}_{ij} & \text{if } y_i z_j = 1 \end{cases} \quad (i, j = 1, \dots, n),$$

so that $A_{yz} \in \mathbf{A}$ for each $y, z \in Y_n$. Since the cardinality of Y_n is 2^n , the cardinality of the set of matrices $\{A_{yz} \mid y, z \in Y_n\}$ is at most 2^{2n} .

4. Definition of the inverse interval matrix. A square interval matrix \mathbf{A} is called *regular* if each $A \in \mathbf{A}$ is nonsingular, and it is said to be *singular* otherwise (i.e., if it contains a singular matrix). In particular, an interval matrix $\mathbf{A} = [A_c - \Delta, A_c + \Delta]$ with

$$\rho(|A_c^{-1}| \Delta) < 1 \quad (4.1)$$

is regular (Beeck [1]); interval matrices satisfying (4.1) are called *strongly regular*. The inverse interval matrix is defined only for regular interval matrices.

DEFINITION 4.1. For a regular interval matrix \mathbf{A} we define its inverse interval matrix $\mathbf{A}^{-1} = [\underline{B}, \overline{B}]$ by

$$\begin{aligned} \underline{B} &= \min \{ A^{-1} \mid A \in \mathbf{A} \}, \\ \overline{B} &= \max \{ A^{-1} \mid A \in \mathbf{A} \} \end{aligned}$$

(componentwise).

Comment. Thus, \mathbf{A}^{-1} is the narrowest interval matrix enclosing the set of matrices $\{A^{-1} \mid A \in \mathbf{A}\}$. Instead of “inverse interval matrix”, we sometimes simply say “interval inverse”.

5. The matrices B_y . First we show that regularity of an $n \times n$ interval matrix implies the existence of 2^n uniquely determined matrices.

THEOREM 5.1. [15, Thm. 5.1, (A3)] For a square interval matrix $\mathbf{A} = [A_c - \Delta, A_c + \Delta]$, the following assertions are equivalent:

- (i) \mathbf{A} is regular,
- (ii) for each $y \in Y_n$ the matrix equation

$$A_c B - T_y \Delta |B| = I \quad (5.1)$$

has a unique matrix solution B_y ,

(iii) for each $y \in Y_n$ the matrix equation (5.1) has a solution.

The main message here is the implication “(i) \Rightarrow (ii)”; (iii) is added for completeness.

It is useful to formulate the equation (5.1) columnwise.

THEOREM 5.2. *Let \mathbf{A} be regular. Then for each $y \in Y_n$ and for each $j \in \{1, \dots, n\}$ we have*

$$(B_y)_{\bullet j} = x_{yj},$$

where x_{yj} is the unique solution of the equation

$$A_c x - T_y \Delta |x| = e_j. \quad (5.2)$$

This theorem forms the basis of an algorithm for computing the B_y 's presented in the next section. We have still another expression for the j th column of B_y by means of the matrices A_{yz} introduced in (3.1).

THEOREM 5.3. *Let \mathbf{A} be regular. Then for each $y \in Y_n$ and for each $j \in \{1, \dots, n\}$ we have*

$$(B_y)_{\bullet j} = (A_{yz(j)}^{-1})_{\bullet j}, \quad (5.3)$$

where

$$z(j) = \text{sgn}((B_y)_{\bullet j}).$$

Since $z(j)$ depends on j , we cannot generally state that $B_y = A_{yz}^{-1}$ for some z . It may even be that $B_y^{-1} \notin \mathbf{A}$. As a consequence of (5.3), we obtain that

$$(B_y)_{ij} = (A_{yz(j)}^{-1})_{ij} \quad (5.4)$$

for each y, i, j . Of course, (5.3) and (5.4) cannot be directly used for computation of $(B_y)_{\bullet j}$ since they contain $z(j)$, the sign vector of the result.

6. Computation of the B_y 's. Theorem 5.2 shows us a way how to compute the matrix B_y column-by-column provided we are able to solve an equation of the type

$$Ax + B|x| = b, \quad (6.1)$$

called an *absolute value equation*. This can be done by a finite algorithm **signaccord** from [21] whose detailed MATLAB-like description is given in the Appendix. Its syntax is

$$[x, S, flag] = \text{signaccord}(A, B, b),$$

where A, B, b is the data of (6.1), x is a solution of (6.1) (if found), S is a singular matrix in the interval matrix $[A - |B|, A + |B|]$ (if found), and $flag$ is a verbal description of the output ('solution' or 'singular'). The behavior of the algorithm is described in Theorem 17.1. Its important feature is that for a regular interval matrix $[A - |B|, A + |B|]$ it always finds a solution to (6.1) (in infinite precision arithmetic), which in this case is unique [21]. As reported in [21], the algorithm takes on average about $0.11 \cdot n$ steps (passes through the **while** loop), where n is the matrix size.

Solving the equations (5.2) for $j = 1, \dots, n$, we obtain an algorithm (Fig. 6.1) for computing the matrix B_y for a given y .

```

function  $B_y = \text{bymatrix}(\mathbf{A}, y)$ 
for  $j = 1 : n$ 
     $[x, S, flag] = \text{signaccord}(A_c, -T_y \Delta, e_j);$ 
    if  $flag = 'singular'$ ,  $B_y = []$ ; return
    end
     $(B_y)_{\bullet j} = x;$ 
end
    
```

FIG. 6.1. An algorithm for computing B_y .

The following theorem (unpublished) follows directly from Theorems 5.2 and 17.1.

THEOREM 6.1. *For each square interval matrix \mathbf{A} and for each $y \in Y_n$ the algorithm (Fig. 6.1) either finds a matrix B_y satisfying (5.1), or issues an empty matrix B_y in which case \mathbf{A} is singular.*

It should be noted that success in computation of a *single* matrix B_y does not guarantee regularity; it is the existence of solutions of *all* the equations (5.1), $y \in Y_n$ that implies regularity of \mathbf{A} (Theorem 5.1, (iii)).

7. Inverse matrix representation theorem. The following theorem, which is of independent interest, brings us closer to the formulae for the inverse interval matrix to be given in the next section.

THEOREM 7.1. [15, Thm. 6.1] *Let \mathbf{A} be regular. Then for each $A \in \mathbf{A}$ there exist nonnegative diagonal matrices L_y , $y \in Y_n$, satisfying $\sum_{y \in Y_n} L_y = I$ such that*

$$A^{-1} = \sum_{y \in Y_n} B_y L_y \tag{7.1}$$

holds.

The formula (7.1) implies that for each i, j we have

$$(A^{-1})_{ij} = \sum_{y \in Y_n} (B_y)_{ij} (L_y)_{jj}, \quad (7.2)$$

where all the $(L_y)_{jj}$'s are nonnegative and $\sum_{y \in Y_n} (L_y)_{jj} = I_{jj} = 1$. Hence, $(A^{-1})_{ij}$ is a convex combination of the values $(B_y)_{ij}$ over all $y \in Y_n$.

Using the formula (5.3), we can reformulate the representation theorem in terms of the matrices A_{yz} defined in (3.1).

THEOREM 7.2. [18, Thm. 1.1] *Let \mathbf{A} be regular. Then for each $A \in \mathbf{A}$ there exist nonnegative diagonal matrices L_{yz} , $y, z \in Y_n$, satisfying $\sum_{y, z \in Y_n} L_{yz} = I$ such that*

$$A^{-1} = \sum_{y, z \in Y_n} A_{yz}^{-1} L_{yz} \quad (7.3)$$

holds.

Hence, $(A^{-1})_{ij}$ is a convex combination of the values $(A_{yz}^{-1})_{ij}$ over all $y, z \in Y_n$. The expansion (7.3) is perhaps more clear than (7.1) because it employs explicitly expressed matrices A_{yz}^{-1} instead of rather obscure matrices B_y , but the number of matrices A_{yz}^{-1} is 2^{2n} compared to "only" 2^n matrices B_y .

8. Formulae for the inverse interval matrix. Finally, using (7.2) and (5.4), we obtain the following simply formulated, but important result.

THEOREM 8.1. [15, Thm. 6.2] *Let \mathbf{A} be regular. Then its inverse $\mathbf{A}^{-1} = [\underline{B}, \overline{B}]$ is given by*

$$\underline{B} = \min_{y \in Y_n} B_y,$$

$$\overline{B} = \max_{y \in Y_n} B_y.$$

Similarly, from Theorem 7.2 we can derive an analogous result.

THEOREM 8.2. [18, (1.3), (1.4)] *Let \mathbf{A} be regular. Then its inverse $\mathbf{A}^{-1} = [\underline{B}, \overline{B}]$ is given by*

$$\underline{B} = \min_{y, z \in Y_n} A_{yz}^{-1},$$

$$\overline{B} = \max_{y, z \in Y_n} A_{yz}^{-1}.$$

The formulation of Theorem 8.2 is advantageous in that it leads us to some clues about matrices at which bounds of the inverse interval matrix are attained.

THEOREM 8.3. [18, Thm. 1.2] *Let \mathbf{A} be regular and let $i, j \in \{1, \dots, n\}$. Then we have:*

$$(i) \underline{B}_{ij} = (A_{yz}^{-1})_{ij} \text{ for some } y, z \in Y_n \text{ satisfying}$$

$$y^T \circ (A_{yz}^{-1})_{i\bullet} \leq 0^T, \tag{8.1}$$

$$z \circ (A_{yz}^{-1})_{\bullet j} \geq 0, \tag{8.2}$$

$$(ii) \overline{B}_{ij} = (A_{yz}^{-1})_{ij} \text{ for some } y, z \in Y_n \text{ satisfying}$$

$$y^T \circ (A_{yz}^{-1})_{i\bullet} \geq 0^T,$$

$$z \circ (A_{yz}^{-1})_{\bullet j} \geq 0.$$

For instance, the Hadamard product inequalities (8.1), (8.2) are equivalent to

$$y_k (A_{yz}^{-1})_{ik} \leq 0 \quad (k = 1, \dots, n), \tag{8.3}$$

$$z_h (A_{yz}^{-1})_{hj} \geq 0 \quad (h = 1, \dots, n).$$

Thus, if we know in advance that, e.g., $\underline{B}_{ik} > 0$, then $(A_{yz}^{-1})_{ik} > 0$ for each $y, z \in Y_n$ and (8.3) implies that $y_k = -1$; similarly, if $\overline{B}_{ik} < 0$, then (8.3) gives $y_k = 1$. Hence, preliminary knowledge of the signs of the bounds may lead us to a reduction, sometimes significant, of the number of matrices A_{yz} to be inverted. We shall explore these ideas further in Section 12.

9. NP-hardness. The formulae given for the inverse interval matrix in Theorems 8.1 and 8.2 are inherently exponential. The question whether essentially simpler formulae may be found was answered in the negative by Coxson [3] who proved that computation of the inverse interval matrix is NP-hard.

THEOREM 9.1. [3] *The following problem is NP-hard:*

Instance. *A strongly regular interval matrix $\mathbf{A} = [A_c - \Delta, A_c + \Delta]$ with symmetric rational A_c and Δ .*

Question. *Is $\overline{B}_{11} \geq 1$, where $[\underline{B}, \overline{B}] = \mathbf{A}^{-1}$?*

Hence, if the famous conjecture “P \neq NP” is true, then there does not exist a polynomial-time algorithm for computing the interval inverse. In view of this fact, in what follows we shall concentrate on special classes of interval matrices for which the inverse can be computed by simpler means.

10. Inverse of an interval matrix with unit midpoint. The first such a class is formed by interval matrices with unit midpoint, i.e., of the form $\mathbf{A} = [I - \Delta, I + \Delta]$. Such matrices are regular if and only if $\varrho(\Delta) < 1$ holds [22, Prop. 4.1], which is equivalent to

$$M := (I - \Delta)^{-1} \geq 0. \quad (10.1)$$

Hence, we assume that $\varrho(\Delta) < 1$ throughout this section. The main point here consists in the fact all the matrices B_y , $y \in Y_n$ can be described explicitly. The following theorem gives a general matrix formula (10.2) as well as three different componentwise formulae (10.3), (10.4), and (10.5). We use $M = (m_{ij})$ given by (10.1) and $\mu = (\mu_j)$ defined by

$$\mu_j = \frac{m_{jj}}{2m_{jj} - 1} \quad (j = 1, \dots, n).$$

THEOREM 10.1. [22, Thm. 4.2] *Let $\varrho(\Delta) < 1$. Then for each $y \in Y_n$ the unique solution of the matrix equation¹*

$$B - T_y \Delta |B| = I$$

is given by

$$B_y = T_y M T_y + T_y (M - I) T_\mu (I - T_y), \quad (10.2)$$

i.e., componentwise

$$(B_y)_{ij} = y_i y_j m_{ij} + y_i (1 - y_j) (m_{ij} - I_{ij}) \mu_j, \quad (10.3)$$

or

$$(B_y)_{ij} = \begin{cases} y_i m_{ij} & \text{if } y_j = 1, \\ y_i (2\mu_j - 1) m_{ij} & \text{if } y_j = -1 \text{ and } i \neq j, \\ \mu_j & \text{if } y_j = -1 \text{ and } i = j, \end{cases} \quad (10.4)$$

or

$$(B_y)_{ij} = \frac{(y_i + (1 - y_i) I_{ij}) m_{ij}}{y_j + (1 - y_j) m_{jj}} \quad (10.5)$$

$(i, j = 1, \dots, n)$.

Using Theorem 8.1, we obtain simple formulae for the interval inverse in this case.

¹This is the equation (5.1) with $A_c = I$.

THEOREM 10.2. [22, Thm. 4.3] *Let $\mathbf{A} = [I - \Delta, I + \Delta]$ with $\rho(\Delta) < 1$. Then the inverse interval matrix $\mathbf{A}^{-1} = [\underline{B}, \overline{B}]$ is given by*

$$\begin{aligned} \underline{B} &= -M + T_{\kappa}, \\ \overline{B} &= M, \end{aligned} \tag{10.6}$$

where

$$\kappa_j = \frac{2m_{jj}^2}{2m_{jj} - 1} \quad (j = 1, \dots, n),$$

or componentwise

$$\begin{aligned} \underline{B}_{ij} &= \begin{cases} -m_{ij} & \text{if } i \neq j, \\ \mu_j & \text{if } i = j, \end{cases} \\ \overline{B}_{ij} &= m_{ij} \end{aligned}$$

$(i, j = 1, \dots, n)$.

In particular, we have this consequence.

THEOREM 10.3. [22, Cor. 4.4] *If $\rho(\Delta) < 1$, then the inverse interval matrix $[I - \Delta, I + \Delta]^{-1} = [\underline{B}, \overline{B}]$ satisfies*

$$\frac{1}{2} \leq \underline{B}_{jj} \leq 1 \leq \overline{B}_{jj}$$

for each j .

According to (10.6), $\overline{B} = (I - \Delta)^{-1}$. The last theorem of this section reveals at what matrices the entries of \underline{B} are attained.

THEOREM 10.4. [22, Thm. 5.1] *For each i, j we have:*

(i) *if $i \neq j$, then*

$$\underline{B}_{ij} = (I - T_y \Delta T_y)_{ij}^{-1}$$

for each $y \in Y$ satisfying $y_i y_j = -1$,

(ii) *if $i = j$, then*

$$\underline{B}_{jj} = (I - T_y \Delta T_z)_{jj}^{-1}$$

for each $y \in Y$ satisfying $y_j = -1$ and $z = y + 2e_j$.

11. Enclosure of the inverse interval matrix. An interval matrix \mathbf{C} is called an *enclosure* of \mathbf{A}^{-1} if $\mathbf{A}^{-1} \subseteq \mathbf{C}$. Computation of an enclosure of the inverse of a strongly regular interval matrix can be performed in polynomial time, as shown in

the following theorem which is a follow-up of previous results by Hansen [8], Bliok [2] and Rohn [17] on interval linear equations.

THEOREM 11.1. [4, Thm. 2.40] *Let $\mathbf{A} = [A_c - \Delta, A_c + \Delta]$ be strongly regular. Then we have*

$$\mathbf{A}^{-1} \subseteq [\underline{\underline{B}}, \overline{\overline{B}}],$$

where

$$\begin{aligned} M &= (I - |A_c^{-1}|\Delta)^{-1}, \\ \mu &= (M_{11}, \dots, M_{nn})^T, \\ T_\nu &= (2T_\mu - I)^{-1}, \\ \underline{\underline{B}} &= -M|A_c^{-1}| + T_\mu(A_c^{-1} + |A_c^{-1}|), \\ \tilde{B} &= M|A_c^{-1}| + T_\mu(A_c^{-1} - |A_c^{-1}|), \\ \underline{B} &= \min\{\underline{\underline{B}}, T_\nu \underline{\underline{B}}\}, \\ \overline{\overline{B}} &= \max\{\tilde{B}, T_\nu \tilde{B}\}. \end{aligned}$$

Other types of enclosures were studied by Hansen [6], Hansen and Smith [7], Herzberger and Bethke [11], and Herzberger [9], [10].

The preliminary knowledge of an enclosure may make computation of the interval inverse easier, see Theorem 12.2 below.

12. Inverse sign stability. Let Z be a matrix satisfying $|Z| = E$, i.e., a ± 1 -matrix. We say that a regular interval matrix \mathbf{A} is *inverse Z -stable* if

$$Z \circ A^{-1} > 0$$

holds for each $A \in \mathbf{A}$. This means that for each i, j , either $(A^{-1})_{ij} < 0$ for each $A \in \mathbf{A}$ (if $Z_{ij} = -1$), or $(A^{-1})_{ij} > 0$ for each $A \in \mathbf{A}$ (if $Z_{ij} = 1$). We say simply that \mathbf{A} is inverse sign stable if it is inverse Z -stable for some Z .

We have the following finite characterization.

THEOREM 12.1. [18, Thm. 2.1] *\mathbf{A} is inverse Z -stable if and only if each A_{yz} is nonsingular and*

$$Z \circ A_{yz}^{-1} > 0 \tag{12.1}$$

holds for each $y, z \in Y_n$.

Notice that regularity of \mathbf{A} is not assumed; it follows from (12.1). The next theorem gives a sufficient inverse Z -stability condition verifiable in polynomial time.

THEOREM 12.2. [Improved version of [18], Thm. 2.2] *If \mathbf{A} is strongly regular and if*

$$\underline{\underline{B}} \circ \overline{\overline{B}} > 0$$

holds, where $\underline{\underline{B}}, \overline{\overline{B}}$ are as in Theorem 11.1, then \mathbf{A} is inverse Z -stable, where $Z = \text{sgn}(\underline{\underline{B}})$.

The main reason for introducing inverse Z -stable matrices is the following theorem which gives explicit componentwise formulae for entries of the bounds of the inverse interval matrix. It is an easy consequence of Theorem 8.3.

THEOREM 12.3. [18, Thm. 2.3] *Let \mathbf{A} be inverse Z -stable. Then the bounds of its inverse $\mathbf{A}^{-1} = [\underline{\underline{B}}, \overline{\overline{B}}]$ are given by the explicit formulae*

$$\begin{aligned} \underline{\underline{B}}_{ij} &= (A_{-y(i), z(j)}^{-1})_{ij} \\ \overline{\overline{B}}_{ij} &= (A_{y(i), z(j)}^{-1})_{ij} \quad (i, j = 1, \dots, n), \end{aligned}$$

where $y(i) = \text{sgn}((Z_{i\bullet})^T)$ and $z(j) = \text{sgn}(Z_{\bullet j})$ for each i, j .

13. Inverse sign pattern. Let \mathbf{A} be regular. If there exist (fixed) $z, y \in Y_n$ such that

$$(zy^T) \circ A^{-1} \geq 0 \tag{13.1}$$

for each $A \in \mathbf{A}$, then \mathbf{A} is said to be of the *inverse sign pattern* (z, y) . In other words, for each i, j we have $z_i y_j (A^{-1})_{ij} \geq 0$ for each $A \in \mathbf{A}$, so that $z_i y_j$ prescribes the sign of $(A^{-1})_{ij}$. If strict inequality holds in (13.1), then \mathbf{A} is inverse zy^T -stable. The property (13.1) can be succinctly reformulated as

$$T_z A^{-1} T_y \geq 0$$

for each $A \in \mathbf{A}$. It is a rather surprising fact that for both the characterization and the explicit form of interval inverse we need only two matrices in this case, namely A_{yz}^{-1} and $A_{-y,z}^{-1}$.

THEOREM 13.1. [15, Thm. 4.6] *\mathbf{A} is of the inverse sign pattern (z, y) if and only if A_{yz} and $A_{-y,z}$ are nonsingular and*

$$T_z A_{yz}^{-1} T_y \geq 0, \tag{13.2}$$

$$T_z A_{-y,z}^{-1} T_y \geq 0 \tag{13.3}$$

hold².

²Which implicitly asserts that the two conditions (13.2) and (13.3) imply regularity of \mathbf{A} .

The following theorem has not been published so far.

THEOREM 13.2. *If \mathbf{A} is of the inverse sign pattern (z, y) , then its inverse interval matrix is given by*

$$\mathbf{A}^{-1} = [\min\{A_{yz}^{-1}, A_{-y,z}^{-1}\}, \max\{A_{yz}^{-1}, A_{-y,z}^{-1}\}]. \quad (13.4)$$

See Garloff [5] for the special case of $y = z = (1, -1, 1, -1, \dots, (-1)^{n-1})^T$.

14. Nonnegative invertibility. An interval matrix \mathbf{A} is said to be *nonnegative invertible* if it is of the inverse sign pattern (e, e) , i.e., if

$$A^{-1} \geq 0$$

holds for each $A \in \mathbf{A}$. As immediate consequences of Theorems 13.1 and 13.2, we obtain the following two results.

THEOREM 14.1. [12] *\mathbf{A} is nonnegative invertible if and only if $\underline{A}^{-1} \geq 0$ and $\overline{A}^{-1} \geq 0$.*

THEOREM 14.2. [13] *If $\mathbf{A} = [\underline{A}, \overline{A}]$ is nonnegative invertible, then*

$$\mathbf{A}^{-1} = [\overline{A}^{-1}, \underline{A}^{-1}]. \quad (14.1)$$

The last formula follows from the fact that $\underline{A}^{-1} - \overline{A}^{-1} = \underline{A}^{-1}(\overline{A} - \underline{A})\overline{A}^{-1} \geq 0$ which gives $\underline{A}^{-1} \geq \overline{A}^{-1}$, hence (13.4) implies (14.1). Finally, we have the following inverse expansion theorem.

THEOREM 14.3. [14, Thm. 2] *If \mathbf{A} is inverse nonnegative, then for each $A \in \mathbf{A}$ there holds*

$$A^{-1} = \left(\sum_{j=0}^{\infty} (\overline{A}^{-1}(\overline{A} - A))^j \right) \overline{A}^{-1}.$$

15. Uniform width. An interval matrix \mathbf{A} is said to be of *uniform width* if it is of the form

$$\mathbf{A} = [A_c - \alpha E, A_c + \alpha E] \quad (15.1)$$

for some $\alpha \geq 0$. For sufficiently small α , its inverse can be again expressed explicitly. Let us denote

$$\begin{aligned} c &= |A_c^{-1}|e, \\ d &= |A_c^{-1}|^T e. \end{aligned}$$

THEOREM 15.1. [16, Thm. 2] *Let A_c be nonsingular and let $\alpha \geq 0$ satisfy*

$$\alpha(cd^T + \|c\|_1|A_c^{-1}|) < |A_c^{-1}|. \quad (15.2)$$

Then for the interval inverse $[\underline{B}, \overline{B}]$ of (15.1) we have

$$\begin{aligned} \underline{B}_{ij} &= (A_c^{-1})_{ij} - \frac{\alpha c_i d_j}{1 + \alpha z(j)^T A_c^{-1} y(i)}, \\ \overline{B}_{ij} &= (A_c^{-1})_{ij} + \frac{\alpha c_i d_j}{1 - \alpha z(j)^T A_c^{-1} y(i)} \quad (i, j = 1, \dots, n), \end{aligned}$$

where

$$\begin{aligned} y(i) &= \text{sgn}(((A_c^{-1})_{i\bullet})^T), \\ z(j) &= \text{sgn}((A_c^{-1})_{\bullet j}). \end{aligned}$$

The condition (15.2) provides for both strong regularity and inverse sign stability of \mathbf{A} .

16. Software. The freely available verification software package VERSOFT [26] written in INTLAB [23], [24], a toolbox of MATLAB, contains a file VERINVERSE.M [25] for computing a verified inverse of a square interval matrix. Its syntax is

$$[\mathbf{B}, \mathbf{S}] = \text{verinverse}(\mathbf{A}),$$

where A is an interval matrix, B is its verified interval inverse (if found), and S is a very tight interval matrix which is a part of A and is verified to contain a singular matrix in A (if found). B and S are never assigned numerical values simultaneously; at least one of them is a matrix of NaN's as the two options - regularity and singularity - exclude each other. The interval matrix B , if computed, is verified to contain the interval inverse of A and the overestimation is solely due to the outward rounding committed; in infinite precision arithmetic it would compute the exact interval inverse. It is based on a not-a-priori-exponential algorithm **hull** for solving interval linear equations described in [19]; its theoretical basis and implementation details have not been published. Nevertheless, the computation may occasionally last long as the problem is NP-hard (Theorem 9.1). In such cases we recommend computation of a polynomial-time enclosure described in Theorem 11.1. This enclosure has not been included into VERSOFT. INTLAB users may employ the function INV.M adapted for an interval argument by S.M. Rump [24].

17. Appendix: An algorithm for solving the absolute value equation.

This appendix contains a MATLAB-like description of an algorithm for solving the absolute value equation accompanied by a finite termination theorem. Both these results were referred to in Section 6.

THEOREM 17.1. [21, Thm. 3.1] *For each $A, B \in \mathbb{R}^{n \times n}$ and each $b \in \mathbb{R}^n$, the sign accord algorithm (Fig. 17.1) in a finite number of steps either finds a solution of the equation*

$$Ax + B|x| = b,$$

or states singularity of the interval matrix $[A - |B|, A + |B|]$ (and, in most cases, also finds a singular matrix $S \in [A - |B|, A + |B|]$).

```

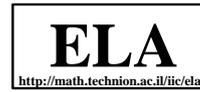
function [x, S, flag] = signaccord (A, B, b)
% Finds a solution to  $Ax + B|x| = b$  or states
% singularity of  $[A - |B|, A + |B|]$ .
x = []; S = []; flag = 'singular';
if A is singular, S = A; return, end
p = 0  $\in \mathbb{R}^n$ ;
z = sgn( $A^{-1}b$ );
if  $A + BT_z$  is singular, S =  $A + BT_z$ ; return, end
x =  $(A + BT_z)^{-1}b$ ;
C =  $-(A + BT_z)^{-1}B$ ;
while  $z_j x_j < 0$  for some j
    k = min{j |  $z_j x_j < 0$ };
    if  $1 + 2z_k C_{kk} \leq 0$ 
        S =  $A + B(T_z + (1/C_{kk})e_k e_k^T)$ ;
        x = [];
        return
    end
    p_k = p_k + 1;
    if  $\log_2 p_k > n - k$ , x = []; return, end
    z_k = -z_k;
     $\alpha = 2z_k / (1 - 2z_k C_{kk})$ ;
    x = x +  $\alpha x_k C_{\bullet k}$ ;
    C = C +  $\alpha C_{\bullet k} C_{k \bullet}$ ;
end
flag = 'solution';
    
```

FIG. 17.1. The sign accord algorithm [21].

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