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Linear Programming with Inexact Data is NP-Hard

We prove that the problem of checking existence of optimal solutions to all linear programming problems whose data range in prescribed intervals is NP-hard.

1. The result

Consider a family of linear programming (LP) problems

$$\min\{c^T x; Ax = b, x \geq 0\} \quad (1)$$

for all data satisfying

$$A \in A^I, b \in b^I, c \in c^I, \quad (2)$$

where $A^I = \{A; \underline{A} \leq A \leq \overline{A}\}$ is an $m \times n$ interval matrix, $m \leq n$, and $b^I = \{b; \underline{b} \leq b \leq \overline{b}\}$, $c^I = \{c; \underline{c} \leq c \leq \overline{c}\}$ are interval vectors of dimensions m and n , respectively (the inequalities are understood componentwise). The family (1), (2) may be interpreted as a linear programming problem with inexact data, or as a fully parametrized parametric linear programming problem.

The problem of existence of optimal solutions to all linear programming problems in the family (1), (2) was addressed in [5]. There it was proved that each LP problem (1) with data satisfying (2) has an optimal solution if and only if the LP problem $\min\{c^T x; \underline{A}x \leq \underline{b}, \overline{A}x \geq \overline{b}, x \geq 0\}$ has an optimal solution, and each of the 2^m systems $Ax = b$ whose each row is either of the form $(\underline{A}x)_i = \underline{b}_i$ or of the form $(\overline{A}x)_i = \overline{b}_i$ ($i = 1, \dots, m$) has a nonnegative solution. Hence, we have a finitely verifiable necessary and sufficient condition, but the number of systems to be checked for nonnegative solvability is exponential in m .

In the main result of this paper we show that the problem in question is NP-hard. Hence, unless the famous conjecture “P≠NP” (see GAREY AND JOHNSON [1]) is false, there does not exist a polynomial-time algorithm for checking existence of optimal solutions to all LP problems (1), (2). The proof given below shows that even checking *feasibility* of all LP problems in the family (1), (2) is NP-hard.

Theorem 1. *The following decision problem is NP-hard:*

Instance. A^I, b^I, c^I (with rational bounds).

Question. *Does each LP problem (1) with data satisfying (2) have an optimal solution?*

Proof. 0) For the purpose of the proof, let us introduce $A_c = \frac{1}{2}(\underline{A} + \overline{A})$, $\Delta = \frac{1}{2}(\overline{A} - \underline{A})$, $b_c = \frac{1}{2}(\underline{b} + \overline{b})$ and $\delta = \frac{1}{2}(\overline{b} - \underline{b})$, so that $A^I = [A_c - \Delta, A_c + \Delta]$ and $b^I = [b_c - \delta, b_c + \delta]$. The proof goes through several steps.

1) First we prove that each system

$$Ax = b, x \geq 0 \quad (3)$$

with data satisfying

$$A \in A^I, b \in b^I \quad (4)$$

has a solution if and only if

$$(\forall y)(A_c^T y + \Delta^T |y| \geq 0 \Rightarrow b_c^T y - \delta^T |y| \geq 0) \quad (5)$$

holds. “Only if”: Let each system (3) with data (4) have a solution, and let $A_c^T y + \Delta^T |y| \geq 0$ for some $y \in \mathbb{R}^m$. Define a diagonal matrix T by $T_{ii} = 1$ if $y_i \geq 0$, $T_{ii} = -1$ if $y_i < 0$, and $T_{ij} = 0$ if $i \neq j$ ($i, j = 1, \dots, m$), then $|y| = Ty$. Consider now the system

$$(A_c + T\Delta)x = b_c - T\delta, x \geq 0. \quad (6)$$

Since $A_c + T\Delta \in A^I$ and $b_c - T\delta \in b^I$, the system (6) has a solution according to the assumption, and $(A_c + T\Delta)^T y = A_c^T y + \Delta^T |y| \geq 0$, hence FARKAS lemma applied to (6) gives that $b_c^T y - \delta^T |y| = (b_c - T\delta)^T y \geq 0$, which proves (5). “If”: Assuming that (5) holds, consider a system (3) with data satisfying (4). Let $A^T y \geq 0$ for some y ; then $A_c^T y + \Delta^T |y| \geq (A_c + A - A_c)^T y = A^T y \geq 0$, hence (5) gives that $b^T y = (b_c + b - b_c)^T y \geq b_c^T y - \delta^T |y| \geq 0$. Thus we have proved that for each y , $A^T y \geq 0$ implies $b^T y \geq 0$, and FARKAS lemma proves the existence of a solution to (3).

2) For a given square $m \times m$ interval matrix $A_0^I = [A_c^0 - \Delta^0, A_c^0 + \Delta^0]$, construct an $m \times 2m$ interval matrix

$$A^I = [A_c - \Delta, A_c + \Delta] \quad (7)$$

with

$$A_c = (A_c^{0T}, -A_c^{0T}), \quad \Delta = (\Delta^{0T}, \Delta^{0T}), \quad (8)$$

and interval vectors

$$b^I = [-e, e], \quad c^I = [e, e], \quad (9)$$

where $e = (1, \dots, 1)^T$. We shall prove that A_0^I is regular (i.e., each $A \in A_0^I$ is nonsingular) if and only if each LP problem (1) with data satisfying (2) (A^I , b^I , c^I given by (7)–(9)) has an optimal solution. In fact, since the objective $e^T x$ is bounded from below, a problem (1) has an optimal solution if and only if it is feasible. Hence, according to part 1), Eq. (5), some problem (1) with data (2) does *not* have an optimal solution if and only if there exists a vector y satisfying $\begin{pmatrix} A_c^0 \\ -A_c^0 \end{pmatrix} y + \begin{pmatrix} \Delta^0 \\ \Delta^0 \end{pmatrix} |y| \geq 0$ and $e^T |y| > 0$, which is equivalent to

$$|A_c^0 y| \leq \Delta^0 |y|, \quad y \neq 0. \quad (10)$$

Then the OETTLI–PRAGER theorem [3] gives that (10) is equivalent to existence of a singular matrix in $A_0^I = [A_c^0 - \Delta^0, A_c^0 + \Delta^0]$. This proves the assertion.

3) Given a square $m \times m$ interval matrix A_0^I , construct an $m \times 2m$ interval matrix A^I and interval vectors b^I , c^I by (7)–(9). According to part 2), checking regularity of A_0^I can be reduced in polynomial time to checking optimality of all problems (1), (2). But since the problem of checking regularity of interval matrices is NP-hard (POLJAK AND ROHN [4], Theorem 2.8), the problem of checking whether each LP problem (1) with data satisfying (2) has an optimal solution is NP-hard as well. ■

2. Concluding remarks

KHACHIYAN [2] proved that an LP problem (1) can be solved in polynomial time. The above result shows that this nice property is lost when inexact data are present. Nevertheless, the worst-case-type result of Theorem 1 does not preclude efficient solvability of many practical examples. The criterion from [5] quoted in the introduction requires solving one LP problem and checking nonnegative solvability of 2^p systems of linear equations, where p is the number of rows i having at least one inexact coefficient (i.e., either $\underline{b}_i < \bar{b}_i$, or $\underline{A}_{ij} < \bar{A}_{ij}$ for some j). Thus the criterion can be efficiently applied to practical examples with small values of p .

Acknowledgements

This work was supported by the Czech Republic Grant Agency under grant GAČR 201/95/1484.

3. References

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