

Approximate Version of Interval Computation Is Still NP-Hard

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Abstract

It is known that, in general, the problem of computing the range of a given polynomial on given intervals is NP-hard. For some NP-hard optimization problems, the approximate version – e.g., if we want to find the value differing from the maximum by no more than a factor of 2 – becomes feasible. Thus, a natural question is: what if instead of computing the exact range, we want to compute the enclosure which is, e.g., no more than twice wider than the actual range? In this paper, we show that this approximate version is still NP-hard, whether we want it to be twice wider or k times wider, for any k .

1 Formulation of the Problem

Need for interval computations. In practice, we often need to estimate the value of a difficult-to-measure quantity y by using its known relation $y = f(x_1, \dots, x_n)$ with easier-to-measure quantities $x_1, \dots, x_n$. This relation is usually described by a continuous function $f(x_1, \dots, x_n)$.

Measurements are never 100% accurate. The measurement result $\tilde{x}_i$ is, in general, different from the actual (unknown) value x_i of the corresponding quantity. Often, the only information that we have about the measurement error $\Delta x_i \stackrel{\text{def}}{=} \tilde{x}_i - x_i$ is the upper bound Δ_i on its absolute value: $|\Delta x_i| \leq \Delta_i$; see, e.g., [9]. In this case, once we know the measurement result $\tilde{x}_i$, the only information that we have about the actual value x_i is that this value belongs to the interval $[\underline{x}_i, \bar{x}_i]$, where $\underline{x}_i = \tilde{x}_i - \Delta_i$ and $\bar{x}_i = \tilde{x}_i + \Delta_i$.

For different values x_i from these intervals, we get, in general, different values of $y = f(x_1, \dots, x_n)$. It is therefore important to find the range of possible values of y , i.e., find the interval

$$[\underline{y}, \bar{y}] = f([\underline{x}_1, \bar{x}_1], \dots, [\underline{x}_n, \bar{x}_n]) = \{f(x_1, \dots, x_n) : x_i \in [\underline{x}_i, \bar{x}_i]\}. \quad (1)$$

The problem of computing this range based on the function $f(x_1, \dots, x_n)$ and intervals $[\underline{x}_i, \bar{x}_i]$ is known as the main problem of *interval computations*; see, e.g., [4, 6, 7].

The main problem of interval computations is known to be NP-hard. It is known that already for polynomials $f(x_1, \dots, x_n)$ – i.e., for algorithms that start with variables and constants and involve only addition, subtraction, and multiplication – the main problem of interval computations is NP-hard. This was first proven in [2, 3]; see also [5].

The fact that this problem is NP-hard means that – unless $P = NP$, which most computer scientists believe to be impossible – no feasible algorithm can solve all the instances of the interval computation problem. Since no feasible algorithm can always compute the exact range, the currently used feasible algorithms compute *enclosures*, i.e., intervals $[\underline{Y}, \bar{Y}]$ that contain (enclose) the desired range: $[\underline{y}, \bar{y}] \subseteq [\underline{Y}, \bar{Y}]$.

What about approximate version of this problem? The main problem of interval computation is, in effect, an optimization problem: $\underline{y}$ is the minimum of the functions $f(x_1, \dots, x_n)$ under the constraints $x_i \in [\underline{x}_i, \bar{x}_i]$, while $\bar{y}$ is the corresponding maximum.

It is known that for many NP-hard optimization problems, their approximate versions can be solved by feasible algorithms; see, e.g., [1]. For example, it is known that the following *knapsack* optimization problem is NP-hard. We are given the prices $p_1, \dots, p_n$ of n items, their weights $w_1, \dots, w_n$, and the knapsack's capacity W . We need to find, among all selections $S \subseteq \{1, \dots, n\}$ that can fit into the knapsack (i.e., for which $\sum_{i \in S} w_i \leq W$), the selection with the largest possible overall price $\sum_{i \in S} p_i$. Interestingly, for every $k < 1$, there are feasible algorithms for finding a selection for which the overall price is larger than k times the maximum.

Similar algorithms are known for approximate versions of many other NP-hard optimization problems. A natural question is: will an approximate version of interval computations be feasible? In this paper, we show that such a version is still NP-hard.

2 Main Result

Proposition. *For any $k > 1$, the following problem is NP-hard:*

- *given: a polynomial $f(x_1, \dots, x_n)$ and intervals $[\underline{x}_i, \bar{x}_i]$, $i = 1, \dots, n$,*
- *compute an enclosure $[\underline{Y}, \bar{Y}]$ for the range (1) whose width is no more than k times larger than the width of the actual range: $\bar{Y} - \underline{Y} \leq k \cdot (\bar{y} - \underline{y})$.*

Proof.

1°. By definition, a problem is NP-hard if every problem from the class NP can be reduced to this problem; see, e.g., [5, 8]. Thus, a usual way to prove

that a problem is NP-hard is to show that a known NP-hard problem P_0 can be reduced to this problem. Indeed, in this case, every problem from the class NP can be reduced to P_0 , and since P_0 can be reduced to our problem, we can thus conclude that every problem from the class NP can be reduced to our problem as well.

As a known NP-hard problem P_0 , we will consider the following problem: given positive integers $s_1, \dots, s_n$ find values $\varepsilon_i \in \{-1, 1\}$ for which $\sum_{i=1}^n \varepsilon_i \cdot s_i = 0$. Equivalently, we want to divide the given set of positive integers into two parts whose sums are equal: the first part is formed by integers s_i with $\varepsilon_i = -1$, the second part is formed by the remaining integers.

2°. Let us start with a problem of estimating the range of the variance

$$v(x_1, \dots, x_n) = \frac{1}{n} \cdot \sum_{i=1}^n x_i^2 - \left(\frac{1}{n} \cdot \sum_{i=1}^n x_i \right)^2 \quad (2)$$

when $x_i \in [-s_i, s_i]$. The value of the variance is always non-negative, and 0 is attained when all x_i are 0s. Thus, the lower endpoint $\underline{v}$ of the range $[\underline{v}, \bar{v}]$ of the range of the function $v(x_1, \dots, x_n)$ is equal to 0.

Here, $x_i^2 \leq s_i^2$, so $\frac{1}{n} \cdot \sum_{i=1}^n x_i^2 \leq S \stackrel{\text{def}}{=} \frac{1}{n} \cdot \sum_{i=1}^n s_i^2$, thus $v(x_1, \dots, x_n) \leq S$. When the corresponding instance of problem P_0 has a solution ε_i , then for $x_i = \varepsilon_i \cdot s_i$, the value S is attained: $v(x_1, \dots, x_n) = S$. Thus, in this case, $\bar{v} = S$.

Let us show that if the corresponding instance of the problem P_0 does not have a solution, then we have $\bar{x} < S - \frac{1}{4n^2}$. We will prove this by reduction to a contradiction. Let us prove that if $\bar{x} \geq S - \frac{1}{4n^2}$, then the corresponding instance of the problem P_0 has a solution. Indeed, every continuous function attains its maximum at some point in a compact set – in particular, in a box $[-s_1, s_1] \times \dots \times [-s_n, s_n]$. Thus, there exists a tuple $(x_1, \dots, x_n)$ for which $v(x_1, \dots, x_n) \geq S - \frac{1}{4n^2}$, thus

$$S \leq v(x_1, \dots, x_n) + \frac{1}{4n^2}. \quad (3)$$

Then, we have

$$\frac{1}{n} \cdot s_i^2 + \frac{1}{n} \cdot \sum_{j \neq i} s_j^2 = S \leq \frac{1}{n} \cdot \sum_{i=1}^n x_i^2 - \left(\frac{1}{n} \cdot \sum_{i=1}^n x_i \right)^2 + \frac{1}{4n^2}.$$

Since $x_j^2 \leq s_j^2$, we thus get

$$\frac{1}{n} \cdot s_i^2 + \frac{1}{n} \cdot \sum_{j \neq i} s_j^2 \leq \frac{1}{n} \cdot x_i^2 + \frac{1}{n} \cdot \sum_{j \neq i} s_j^2 + \frac{1}{4n^2},$$

hence

$$\frac{1}{n} \cdot s_i^2 \leq \frac{1}{n} \cdot x_i^2 + \frac{1}{4n^2}$$

and

$$x_i^2 \leq s_i^2 \leq x_i^2 + \frac{1}{4n}$$

and so

$$0 \leq s_i^2 - x_i^2 = s_i^2 - |x_i|^2 = (s_i - |x_i|) \cdot (s_i + |x_i|) \leq \frac{1}{4n}.$$

Here, s_i is a positive integer, so $s_i \geq 1$ hence $s_i + |x_i| \geq 1$ and thus,

$$0 \leq s_i - |x_i| \leq \frac{1}{4n}.$$

The right-hand side is smaller than 1, so, for $\varepsilon_i = \text{sign}(x_i)$ (which is 1 if $x_i > 0$, -1 if $x_i < 0$, and 0 if $x_i = 0$), we get

$$|s_i \cdot \varepsilon_i - x_i| \leq \frac{1}{4n \cdot (s_i + |x_i|)} \leq \frac{1}{4n}.$$

Thus,

$$\left| \sum_{i=1}^n s_i \cdot \varepsilon_i - \sum_{i=1}^n x_i \right| \leq \sum_{i=1}^n |s_i \cdot \varepsilon_i - x_i| \leq n \cdot \frac{1}{4n} = \frac{1}{4}. \quad (4)$$

Also, from (3), taking into account that $x_i^2 \leq s_i^2$, we conclude that

$$\frac{1}{n} \cdot \sum_{i=1}^n s_i^2 \leq \sum_{i=1}^n x_i^2 - \left(\frac{1}{n} \cdot \sum_{i=1}^n x_i \right)^2 + \frac{1}{4n^2} \leq \frac{1}{n} \cdot \sum_{i=1}^n s_i^2 - \left(\frac{1}{n} \cdot \sum_{i=1}^n x_i \right)^2 + \frac{1}{4n^2},$$

thus

$$\left(\frac{1}{n} \cdot \sum_{i=1}^n x_i \right)^2 \leq \frac{1}{4n^2}$$

hence

$$\left| \frac{1}{n} \cdot \sum_{i=1}^n x_i \right| \leq \frac{1}{2n},$$

thus

$$\left| \sum_{i=1}^n x_i \right| \leq \frac{1}{2}. \quad (5)$$

From (4) and (5), we conclude that

$$\left| \sum_{i=1}^n s_i \cdot \varepsilon_i \right| \leq \left| \sum_{i=1}^n s_i \cdot \varepsilon_i - \sum_{i=1}^n x_i \right| + \left| \sum_{i=1}^n x_i \right| \leq \frac{1}{4} + \frac{1}{2} < 1.$$

Since the sum $\sum_{i=1}^n s_i \cdot \varepsilon_i$ is an integer, this means that this integer is equal to 0, i.e., that the values ε_i indeed solve the given instance of the problem P_0 .

3°. Now, let us consider a polynomial $f(x_1, \dots, x_n) = (v(x_1, \dots, x_n))^N$, where N is such that

$$\left(1 - \frac{1}{4n^2 \cdot S}\right)^N < \frac{1}{k},$$

e.g.,

$$N = \left\lceil \frac{\ln(k)}{-\ln\left(1 - \frac{1}{4n^2 \cdot S}\right)} \right\rceil + 1.$$

Asymptotically, $n \sim \text{const} \cdot n^2 \cdot S$, so this polynomial is still feasible.

If the original instance of the problem P_0 has a solution, then the range $[\underline{y}, \bar{y}]$ of the function $f(x_1, \dots, x_n)$ is equal to $[0, S^N]$, and thus, the width of any enclosure $[\underline{Y}, \bar{Y}]$ for this range is at least S^N .

If the original instance has no solutions, then the range $[\underline{y}, \bar{y}]$ is contained in $\left[0, \left(S - \frac{1}{4n^2}\right)^N\right]$, so its width is smaller than or equal to

$$\left(S - \frac{1}{4n^2}\right)^N = S^N \cdot \left(1 - \frac{1}{4n^2 \cdot S}\right)^N < \frac{1}{k} \cdot S^N.$$

Since the width of the enclosure $[\underline{Y}, \bar{Y}]$ is no more than k times the width of the actual range, this width is thus smaller than S^N .

So, if we had an algorithm computing such a no-more-than- k -times wider enclosure $[\underline{Y}, \bar{Y}]$, we would be able to tell whether the original instance of the problem P_0 has a solution or not: if $\bar{Y} - \underline{Y} \geq S^N$, the instance has a solution, otherwise it does not have a solution. Thus, we reduced the NP-hard problem P_0 to our problem, hence our problem is also NP-hard. The proposition is proven.

Remaining open problems. Interval computations problem is NP-hard even if we limit ourselves to quadratic polynomials. In our proof that an approximate version is NP-hard we used polynomials of arbitrary degrees. What if we limit ourselves to quadratic polynomials only? Will the problem still be NP-hard for all k ?

Similar questions can be asked about other situations when computing the exact range is NP-hard. For example, it is known that if we only know intervals of possible values of all components a_{ij} of a matrix, then computing the range of possible eigenvalues is NP-hard. What if we consider enclosures for this range which are no more than k times wider than the actual range? Will the problem still be NP-hard?

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