

# Use of Grothendieck's Inequality in Interval Computations: Quadratic Terms are Estimated Accurately (Modulo a Constant Factor)

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## 1. Interval Computations (IC): Brief Reminder

- One of the main problem of interval computations:

- *Given:* a function  $f(x_1, \dots, x_n)$  and intervals

$$\mathbf{x}_i = [\underline{x}_i, \bar{x}_i] = [\tilde{x}_i - \Delta_i, \tilde{x}_i + \Delta_i],$$

- *Compute:* the range

$$\mathbf{y} = f(\mathbf{x}_1, \dots, \mathbf{x}_n) = \{f(x_1, \dots, x_n) : x_i \in \mathbf{x}_i\}.$$

- Computing the exact range is known to be NP-hard, even for quadratic  $f(x_1, \dots, x_n)$ .
- So, instead, we compute an enclosure  $\mathbf{Y} \supseteq \mathbf{y}$ , with excess width  $\text{wid}(\mathbf{Y}) - \text{wid}(\mathbf{y}) > 0$ .
- One of the most widely used methods of efficiently computing  $\mathbf{Y}$  is the Mean Value (MV) method:

$$\mathbf{Y} = f(\tilde{x}_1, \dots, \tilde{x}_n) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{x}_1 \times \dots \times \mathbf{x}_n) \cdot [-\Delta_i, \Delta_i].$$

## 2. Interval Computations: Reminder (cont-d)

- Mean Value (MV) method (reminder):

$$\mathbf{Y} = f(\tilde{x}_1, \dots, \tilde{x}_n) + \sum_{i=1}^n \frac{\partial f}{\partial x_i}(\mathbf{x}_1 \times \dots \times \mathbf{x}_n) \cdot [-\Delta_i, \Delta_i].$$

- The ranges of the derivatives  $f_{,i} \stackrel{\text{def}}{=} \frac{\partial f}{\partial x_i}$  can be estimated, e.g., by using straightforward IC:
  - *parse* the expression  $f_{,i}$ , i.e., represent it as a sequence of elementary arithmetic operations, and
  - replace each operation with numbers by the corresponding operation of interval arithmetic.
- The Mean Value method has excess width  $O(\Delta^2)$ , where

$$\Delta \stackrel{\text{def}}{=} \max \Delta_i.$$

### 3. Can We Get Better Enclosures?

- The Mean Value method has excess width  $O(\Delta^2)$
- Can we come up with more accurate enclosures?
- We cannot get too drastic an improvement:
  - even for quadratic functions  $f(x_1, \dots, x_n)$ , computing the interval range is NP-hard
  - and therefore (unless  $P=NP$ ), a feasible algorithm with excess width  $O(\Delta^{2+\varepsilon})$  is impossible.
- What we can do is try to decrease the overestimation of the quadratic term.
- It turns out that such a possibility follows from an inequality proven by A. Grothendieck in 1953.

## 4. Main Idea

- The MV method is based on the 1st order Mean Value Theorem (MVT):

$$f(\tilde{x} + \Delta x) = f(\tilde{x}) + \sum f_{,i}(\tilde{x} + \eta) \cdot \Delta x_i \text{ for some } \eta_i \in [-\Delta_i, \Delta_i].$$

- Instead, we propose to use 3rd order MVT:

$$f(\tilde{x} + \Delta x) = f(\tilde{x}) + \sum f_{,i}(\tilde{x}) \cdot \Delta x_i + \frac{1}{2} \cdot \sum f_{,ij}(\tilde{x}) \cdot \Delta x_i \cdot \Delta x_j + \frac{1}{6} \cdot \sum f_{,ijk}(\tilde{x} + \eta) \cdot \Delta x_i \cdot \Delta x_j \cdot \Delta x_k.$$

- Specifically, we propose to add estimates for ranges of linear, quadratic, and cubic terms.
- The range of the cubic term is estimated via straightforward interval comp.; the estimate is  $O(\Delta^3)$ .
- The range of the linear term  $f(\tilde{x}) + \sum f_{,i}(\tilde{x}) \cdot \Delta x_i$  can be explicitly described as  $[\tilde{y} - \Delta, \tilde{y} + \Delta]$ , where

$$\tilde{y} \stackrel{\text{def}}{=} f(\tilde{x}) \text{ and } \Delta = \sum |f_{,i}(\tilde{x})| \cdot \Delta_i.$$

## 5. Main Idea (cont-d)

- Reminder: we use the 3rd order MVT:

$$f(\tilde{x} + \Delta x) = f(\tilde{x}) + \sum f_{,i}(\tilde{x}) \cdot \Delta x_i + \frac{1}{2} \cdot \sum f_{,ij}(\tilde{x}) \cdot \Delta x_i \cdot \Delta x_j + \frac{1}{6} \cdot \sum f_{,ijk}(\tilde{x} + \eta) \cdot \Delta x_i \cdot \Delta x_j \cdot \Delta x_k.$$

- Specifically, we propose to add estimates for ranges of linear, quadratic, and cubic terms.
- The range of the linear term can be computed exactly.
- The range of the cubic term is  $O(\Delta^3) \ll O(\Delta^2)$ .
- What remains is to estimate the range  $[-Q, Q]$  of the quadr. term  $\sum_{i,j=1}^n a_{ij} \cdot \Delta x_i \cdot \Delta x_j$   $\left( a_{ij} \stackrel{\text{def}}{=} \frac{1}{2} \cdot f_{,ij}(\tilde{x}) \right)$  on

$$[-\Delta_1, \Delta_1] \times \dots \times [-\Delta_n, \Delta_n].$$

## 6. Relation to Grothendieck Inequality

- *Problem*: estimating the range  $[-Q, Q]$  of

$$\sum_{i,j=1}^n a_{ij} \cdot \Delta x_i \cdot \Delta x_j \text{ on } [-\Delta_1, \Delta_1] \times \dots \times [-\Delta_n, \Delta_n].$$

- *Re-scaling*: for  $z_i \stackrel{\text{def}}{=} \Delta x_i / \Delta_i$ , we have  $z_i \in [-1, 1]$ ,  $\Delta x_i = \Delta_i \cdot z_i$ , and the quadratic form becomes:

$$\sum_{i,j=1}^n b_{ij} \cdot z_i \cdot z_j, \text{ with } b_{ij} \stackrel{\text{def}}{=} a_{ij} \cdot \Delta_i \cdot \Delta_j.$$

- *Thus*:  $Q = \max \left\{ \sum_{i,j=1}^n b_{ij} \cdot z_i \cdot z_j : z_i \in [-1, 1] \right\}$ .
- *Grothendieck's inequality* enables us to estimate the maximum  $Q'$  of a related bilinear function

$$b(z, t) \stackrel{\text{def}}{=} \sum_{i,j=1}^n b_{ij} \cdot z_i \cdot t_j, \quad z_i, t_j \in \{-1, 1\}.$$

## 7. Grothendieck Inequality (cont-d)

- *Auxiliary problem*: estimating

$$Q' = \max \left\{ \sum_{i,j=1}^n b_{ij} \cdot z_i \cdot t_j : z_i, t_j \in \{-1, 1\} \right\}.$$

- This problem is known to be NP-hard.
- *General fact*: discrete optimization problems are more complex than continuous ones.
- *Observation*: the discrete set  $\{-1, 1\}$  is a unit sphere in 1-D Euclidean space.
- *Interesting*: for larger dimensions, a unit sphere is connected (hence not discrete).
- *Grothendieck's idea*: consider  $z_i$  and  $t_j$  from the unit sphere in a Hilbert space (=  $\infty$ -dim. Euclidean space).



## 8. Grothendieck's Result and Related Algorithm

- We want to compute:

$$Q' = \max \left\{ \sum_{i,j=1}^n b_{ij} \cdot z_i \cdot t_j : z_i, t_j \in \{-1, 1\} \right\}.$$

- We estimate instead:

$$Q'' \stackrel{\text{def}}{=} \max \left\{ \sum_{i,j=1}^n b_{ij} \cdot \langle z_i, t_j \rangle : z_i, t_j \in S \right\}.$$

- *Grothendieck's inequality*: for some universal constant  $K_G \in [1, 1.782]$ , we have  $\frac{1}{K_G} \cdot Q'' \leq Q' \leq Q''$ .
- *Comment*: the part  $Q' \leq Q''$  is trivial, since we can have all  $z_i$  and  $t_j$  equal to  $\pm e$  for some unit vector  $e$ .
- *Computational result*: an ellipsoid method – similar to linear programming one – can feasibly compute  $Q''$ .

## 9. How to Use This Algorithm to Estimate the Range $[-Q, Q]$ of the Quadratic Part

- We want to estimate:  $Q = \max\{B(z) : z_i \in [-1, 1]\}$ , where  $B(z) \stackrel{\text{def}}{=} b(z, z)$  and  $b(z, t) = \sum_{i,j=1}^n b_{ij} \cdot z_i \cdot t_j$ .
- We know:  $Q' = \max\{b(z, t) : z_i \in \{-1, 1\}, t_j \in \{-1, 1\}\}$ .
- Fact: a bilinear f-n  $b(z, t)$  attains its max at endpoints.
- Hence:  $Q' = \max\{b(z, t) : z_i \in [-1, 1], t_j \in [-1, 1]\}$ .
- Since  $b(z, t) = B((z+t)/2) - B((z-t)/2)$ , we have  $Q' \leq 2Q$ . Clearly,  $Q \leq Q'$ , hence  $Q'/2 \leq Q \leq Q'$ .
- From  $K_G^{-1} \cdot Q'' \leq Q' \leq Q''$ , we can now conclude that
$$\frac{Q''}{2K_G} \leq Q \leq Q''.$$
- Hence: by computing  $Q''$ , we can feasibly estimate  $Q$  accurately modulo a small constant factor  $2K_G \leq 3.6$ .

## 10. Resulting Algorithm

- According to the 3rd order Mean Value Theorem, for  $\Delta x_i \in [-\Delta_i, \Delta_i]$ , we have:

$$f(\tilde{x} + \Delta x) = T_1 + T_2 + T_3, \text{ where:}$$

$$T_1 \stackrel{\text{def}}{=} f(\tilde{x}) + \sum f_{,i}(\tilde{x}) \cdot \Delta x_i;$$

$$T_2 \stackrel{\text{def}}{=} \sum a_{ij} \cdot \Delta x_i \cdot \Delta x_j, \text{ where } a_{ij} = \frac{1}{2} \cdot f_{,ij}(\tilde{x}); \text{ and}$$

$$T_3 \stackrel{\text{def}}{=} \frac{1}{6} \cdot \sum f_{,ijk}(\tilde{x} + \eta) \cdot \Delta x_i \cdot \Delta x_j \cdot \Delta x_k.$$

- As an enclosure for the range of  $f$ , we take the sum of enclosures for  $T_1$ ,  $T_2$ , and  $T_3$ .
- For  $T_1$ , we compute the exact range in linear time  $O(n)$ .
- For  $T_3$ , we use straightforward interval computations and get an enclosure of width

$$O(\Delta^3) \ll O(\Delta^2).$$

## 11. Resulting Algorithm (cont-d)

- To estimate the range  $[-Q, Q]$  of the quadratic term  $T_2 = \sum a_{ij} \cdot \Delta x_i \cdot \Delta x_j$ , we do the following:
  - compute an auxiliary matrix  $b_{ij} = a_{ij} \cdot \Delta_i \cdot \Delta_j$ , and
  - use the ellipsoid method to compute

$$Q'' \stackrel{\text{def}}{=} \max \left\{ \sum_{i,j=1}^n b_{ij} \cdot \langle z_i, t_j \rangle : z_i, t_j \in S \right\}.$$

- Then,  $\frac{Q''}{2K_G} \leq Q \leq Q''$ , with  $2 \leq 2K_G \leq 3.6$ .
- *Why this is better* that the Mean Value method:
  - we still get excess width  $O(\Delta^2)$ , but
  - this time, we overestimate the quadratic terms by no more than a known constant factor.

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