# An aymptotic minimal contractor for non-linear equations provided in the Codac library

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SWIM 2024, Maastricht  $6^{\mathrm{th}}$  June 2024









## Section 1

# Introduction

## Problem statement

We consider the problem of approximating the solutions of the system:

$$f(x) = 0$$

where  $\mathbf{f}: \mathbb{R}^n \to \mathbb{R}^p$  is a non-linear and differentiable function. Possibly non continuous.

In particular, we will consider systems where:

..for which the solution set  $\mathbb{X}=\{\mathbf{x}\in\mathbb{R}^n\ |\ \mathbf{f}(\mathbf{x})=\mathbf{0}\}$  has infinitely many solutions.

## Set-inversion problem

Set inversion can be done with

- **forward** evaluations of **f** 
  - (see *e.g.* SIVIA)
    - limits in high dimensions due to bisections
- forward/reverse evaluations of f (such as contractors + pavers)
  - lower complexity, bisections used as a last resort
  - efficient for constraint propagation

## Reverse evaluations

### Forward evaluations:

 $fwd_plus([x],[y])$ : returns  $[\{z \mid \exists x \in [x], \exists y \in [y], z = x + y\}]$ 

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### Reverse (backward) evaluations:

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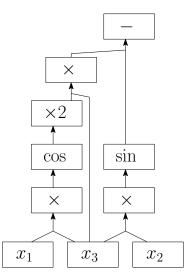
## Section 2

# The HC4Revise algorithm

"Forward-backward" algorithm

DAG associated with:

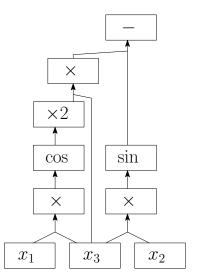
$$f(\mathbf{x}) = 2x_3 \cos(x_1 x_3) - \sin(x_2 x_3)$$



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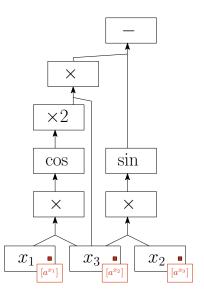


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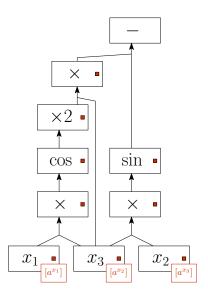


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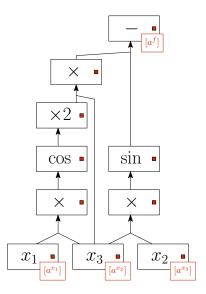


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- 3. intersecting  $\left[a^{f}\right]$  with  $\left[y\right]$

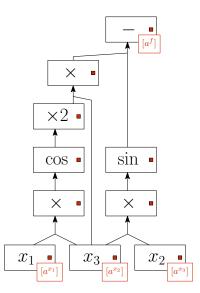


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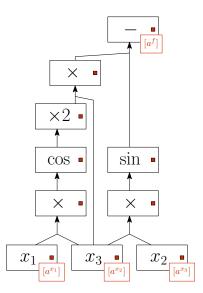


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- 3. intersecting  $[a^f]$  with [y]
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- 5. possible contraction of  $([a^{x_1}], [a^{x_2}], [a^{x_3}])$



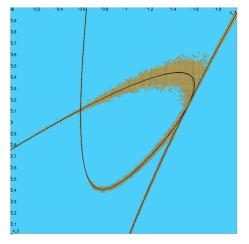
F. Benhamou, F. Goualard, L. Granvilliers, and J. F. Puget, "Revising hull and box consistency," in Logic Programming: The 1999 Inter- national Conference, Las Cruces, New Mexico, USA, November 29 - December 4, 1999, D. D. Schreye, Ed. MIT Press, 1999, pp. 230-244

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## The HC4Revise algorithm Limits of the algorithm

$$\left(\begin{array}{c} -x_3^2 + 2x_3\sin(x_3x_1) + \cos(x_3x_2)\\ 2x_3\cos(x_3x_1) - \sin(x_3x_2) \end{array}\right) = \mathbf{0}$$



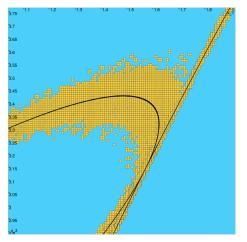
#### Example from:

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- delayed systems
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- projected approximations for  $\mathbf{x}_{1,2}$

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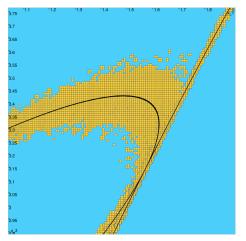
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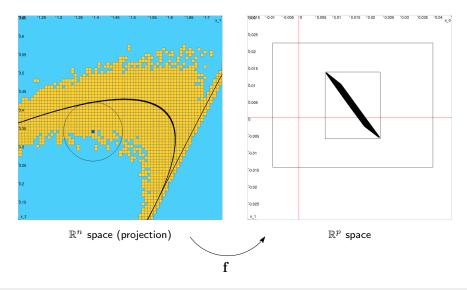
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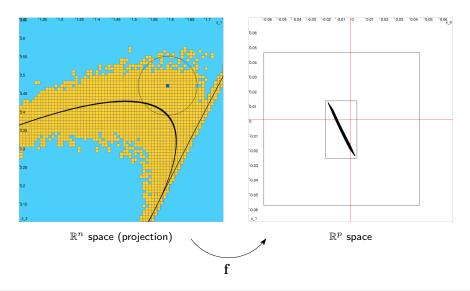
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### Two problems

Clustering effect due to: [1] dependency problem



Clustering effect due to: [2] wrapping effect



# Solutions for these clustering effects

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These solutions come from a recent contribution with proof:

L. Jaulin (2024). Asymptotically minimal interval contractors based on the centered form; Application to the stability analysis of linear time-delayed differential equations, Acta Cybernetica.

## Section 3

# A contractor involving the centered form

Forward evaluations with the "centered form" The classical formula is given as:

$$[\mathbf{f}_c]([\mathbf{x}]) = \mathbf{f}(\overline{\mathbf{x}}) + [\mathbf{J}_{\mathbf{f}}]([\mathbf{x}]) \cdot ([\mathbf{x}] - \overline{\mathbf{x}})$$

with  $\overline{\mathbf{x}}$  the center of the box  $[\mathbf{x}]$ , and  $[\mathbf{J_f}]([\mathbf{x}])$  the interval Jacobian matrix of f evaluated over  $[\mathbf{x}]$ .

R. Moore, Methods and Applications of Interval Analysis Society for Industrial and Applied Mathematics, jan 1979.

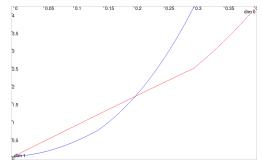
- traditionally used to enclose the range of a function over narrow intervals
- asymptotically small overestimation for sufficiently narrow boxes on scalar functions

The dependency problem: example

Consider for instance this expression with multiple occurrences of x:

 $2x^5 + x^3 - 3x^2$ 

Let us try the evaluation for several growing inputs:  $[x]=0.7+[-\epsilon,\epsilon],\ \epsilon=0\dots0.5$ 



Comparing pessimism: natural evaluation (red) vs centered form evaluation (blue).

Forward evaluations with the "centered form"

The constraint  $\{f(x)\in [y],\; x\in [x]\}$  is then expressed using a centered form expression:

 $\mathbf{f}(\mathbf{x}) \in [\mathbf{y}]$ 

Forward evaluations with the "centered form"

$$\mathbf{f}(\mathbf{x}) \in [\mathbf{y}] \cap \left(\mathbf{f}(\overline{\mathbf{x}}) + [\mathbf{J}_{\mathbf{f}}]([\mathbf{x}])([\mathbf{x}] - \overline{\mathbf{x}})\right)$$

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Simultaneous evaluation  $\Rightarrow$  operator overloading

Interval Automatic Differentiation (IAD)

Contribution of this work:

We use IAD to automatically compute the term  $[\mathbf{J}_{\mathbf{f}}]([\mathbf{x}])$ .

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About Automatic Differentiation:

- techniques to automatically evaluate the partial derivatives of a function
- HC4Revise  $\Rightarrow$  execution of a sequence of elementary algorithms:

fwd\_plus, fwd\_cos, etc (and their reverse counterparts)

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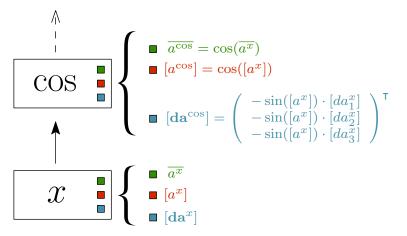
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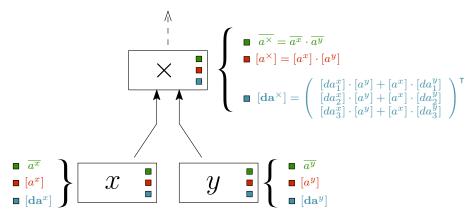
- using **chain rule** and interval analysis: compute AD for approximating  $[{\bf J_f}]([{\bf x}])$
- $\Rightarrow$  operator overloading in programming languages

Examples of operator overloading



Overloading the cos operator with automatic differentiation

Examples of operator overloading

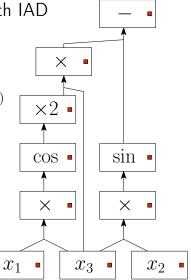


Overloading the product operator with automatic differentiation

Directed acyclic graph, now with IAD

Directed acyclic graph (DAG) involving automatic differentiation:

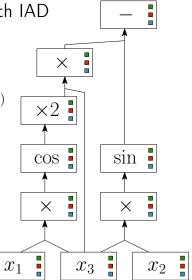
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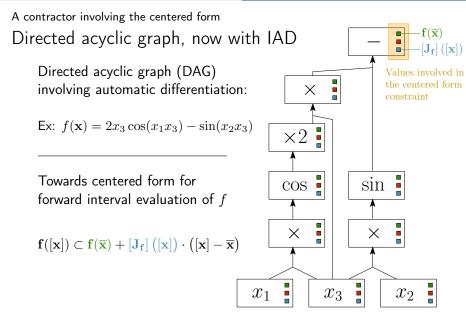
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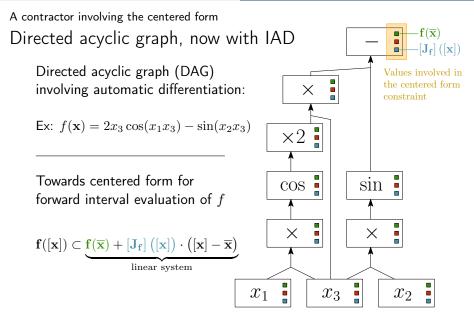
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A contractor involving the centered form Directed acyclic graph, now with IAD Directed acyclic graph (DAG) the centered form involving automatic differentiation: Х Ex:  $f(\mathbf{x}) = 2x_3 \cos(x_1 x_3) - \sin(x_2 x_3)$  $\times 2$ sin COS Х Х  $x_3$  $x_2$  $x_1$ 





## Section 4

# Top level algorithm for the CtcInverse contractor

Recall the constraint  $\{f(\mathbf{x})\in[\mathbf{y}],\ \mathbf{x}\in[\mathbf{x}]\}$ , expressed using a centered form expression:

$$\mathbf{f}(\mathbf{x}) \in [\mathbf{y}] \cap \left(\mathbf{f}(\overline{\mathbf{x}}) + [\mathbf{J}_{\mathbf{f}}]\big([\mathbf{x}]\big)\big([\mathbf{x}] - \overline{\mathbf{x}}\big)\right)$$

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This can be expressed under the form of a linear constraint:

 $\mathbf{b} = \mathbf{A} \cdot \mathbf{p}$ 

$$- \mathbf{b} \in ([\mathbf{y}] \cap [\mathbf{f}]([\mathbf{x}])) - \mathbf{f}(\overline{\mathbf{x}}), \in \mathbb{IR}^{p}$$
$$- \mathbf{A} \in [\mathbf{J}_{\mathbf{f}}]([\mathbf{x}]), \in \mathbb{IR}^{p \times n}$$
$$- \mathbf{p} \in [\mathbf{x}] - \overline{\mathbf{x}}, \in \mathbb{IR}^{n}$$

Resulting algorithm for the contractor involving the centered form, for dealing with the constraint  $f(\mathbf{x})\in[\mathbf{y}]$ 

```
\begin{aligned} \mathtt{CtcInverse}(in: \mathbf{f}, [\mathbf{y}], \ in/out: \ [\mathbf{x}]) \\ & \overline{\mathbf{x}} \leftarrow \mathrm{mid}([\mathbf{x}]) \\ & \left(\mathbf{a}^{\mathbf{f}}, [\mathbf{a}^{\mathbf{f}}], [\mathbf{d}\mathbf{a}^{\mathbf{f}}]\right) \leftarrow \mathtt{fwd\_dag\_eval}(\mathbf{f}, [\mathbf{x}], \overline{\mathbf{x}}) \\ & [\mathbf{p}] \leftarrow [\mathbf{x}] - \overline{\mathbf{x}} \\ & [\mathbf{b}] \leftarrow ([\mathbf{y}] \cap [\mathbf{a}^{\mathbf{f}}]) - \mathbf{a}^{\mathbf{f}} \\ & [\mathbf{A}] \leftarrow [\mathbf{d}\mathbf{a}^{\mathbf{f}}] \\ & ([\mathbf{A}], [\mathbf{p}]) \leftarrow \mathtt{reverse\_mul}([\mathbf{b}], [\mathbf{A}], [\mathbf{p}]) \\ & [\mathbf{x}] \leftarrow [\mathbf{p}] + \overline{\mathbf{x}} \end{aligned}
```

with the terms  $\mathbf{a}^{f}$ ,  $[\mathbf{a}^{f}]$  and  $[\mathbf{d}\mathbf{a}^{f}]$  computed during the forward evaluation of the DAG.

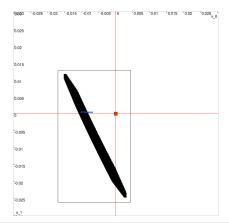
Top level algorithm for the CtcInverse contractor Dealing efficiently with the linear constraint

About the reverse matrix-vector product:  $([\mathbf{A}], [\mathbf{p}]) \leftarrow \texttt{reverse\_mul}([\mathbf{b}], [\mathbf{A}], [\mathbf{p}])$ 

Some preconditioning allows to significantly reduce wrapping effect.

This can be achieved using a Gauss Jordan band diagonalization method.



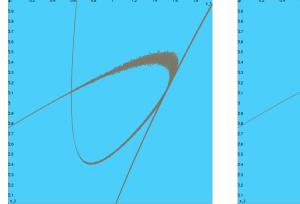


## Section 5

# Results of CtcInverse provided in the Codac library

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Previous example using the centered form

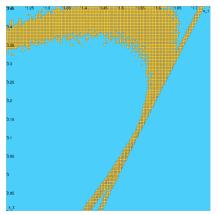


 $\mathbb X$  computed with HC4Revise. Computation time: 4.51s. 27430 boxes.

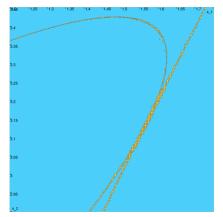
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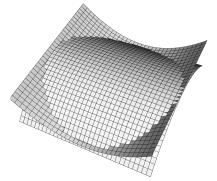


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Results of CtcInverse provided in the Codac library Comparison with affine arithmetic



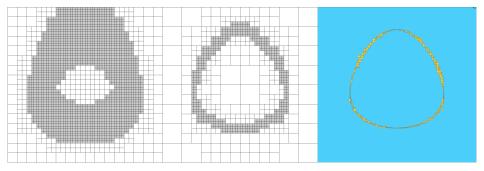
#### Intersection of two lofted parabolas:

L.H. De Figueiredo. Surface Intersection using Affine Arithmetic, Proceedings of Graphics Interface'96, 168-175. 1996

$$\begin{aligned} \alpha(u) &= \mathbf{a}_0 (1-u)^2 + 2\mathbf{a}_1 u (1-u) + \mathbf{a}_2 u^2 \\ \beta(u) &= \mathbf{b}_0 (1-u)^2 + 2\mathbf{b}_1 u (1-u) + \mathbf{b}_2 u^2 \\ \mathbf{f}(u,v) &= (1-v)\alpha(u) + v\beta(u) \end{aligned}$$

Results of CtcInverse provided in the Codac library Comparison with affine arithmetic

 $\mathbf{f}:\mathbb{R}^4\to\mathbb{R}^3$ 



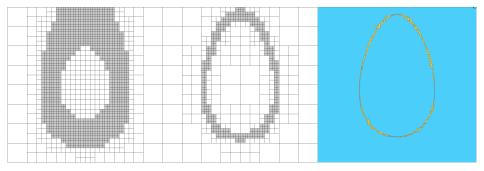
Results projected on the  $(u_1, v_1)$  space. Left: natural evaluation – center: affine evaluation – right: CtcInverse

Using the same  $\epsilon$  in the branch-and-band algorithm.

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Results of CtcInverse provided in the Codac library Comparison with affine arithmetic

 $\mathbf{f}:\mathbb{R}^4\to\mathbb{R}^3$ 



Results projected on the  $(u_2, v_2)$  space. Left: natural evaluation – center: affine evaluation – right: CtcInverse

Using the same  $\epsilon$  in the branch-and-band algorithm.

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## Section 6

## Conclusion

Contribution of this work

## - proposed algorithm involving IAD

- operator overloading:
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```
Conclusion
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```
Python example (Codac library)
```

from codac import \*



```
x = VectorVar(3)
f = AnalyticFunction([x], vec(
    -sqr(x[2])+2*x[2]*sin(x[2]*x[0])+cos(x[2]*x[1]),
    2*x[2]*cos(x[2]*x[0])-sin(x[2]*x[1])
))
ctc = CtcInverse(f, [[0],[0]])
pave([[0,2],[2,4],[0,10]], ctc, 0.004)
```

Using pre-release Codac 2.0.0:

```
pip install codac --pre
```

http://codac.io

```
Matlab example (Codac library)
```



```
import py.codac4matlab.*
```

```
x = VectorVar(3);
f = AnalyticFunction({x}, vec( ...
-sqr(x(3))+2*x(3)*sin(x(3)*x(1))+cos(x(3)*x(2)), ...
2*x(3)*cos(x(3)*x(1))-sin(x(3)*x(2)) ...
));
ctc = CtcInverse(f, {0,0});
pave(IntervalVector({{0,2},{2,4},{0,10}}), ctc, 0.004);
```

Using pre-release Codac 2.0.0 dedicated to Matlab:

```
pip install codac4matlab --pre
```

```
http://codac.io
```

```
Conclusion
```

```
C++ example (Codac library)
```

```
#include <codac-core.h>
```

```
using namespace std;
using namespace codac2;
int main()
{
    VectorVar x(3);
    AnalyticFunction f({x}, vec(
        -sqr(x[2])+2*x[2]*sin(x[2]*x[0])+cos(x[2]*x[1]),
        2*x[2]*cos(x[2]*x[0])-sin(x[2]*x[1])
    ));
```

```
CtcInverse_<IntervalVector> ctc(f, {0.,0.});
pave(IntervalVector({{0,2},{2,4},{0,10}}), ctc, 0.004);
}
```

### http://codac.io

++