

# Interval analysis with application to the safe navigation of autonomous vehicles

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# Interval analysis

**Problem.** Given  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  and a box  $[\mathbf{x}] \subset \mathbb{R}^n$ , prove that

$$\forall \mathbf{x} \in [\mathbf{x}], f(\mathbf{x}) \geq 0.$$

Interval arithmetic can solve efficiently this problem.

**Example.** Is the function

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

always positive for  $x_1, x_2 \in [-1, 1]$  ?

## Interval arithmetic

$$[-1,3] + [2,5] = ?,$$

$$[-1,3] \cdot [2,5] = ?,$$

$$\text{abs}([-7,1]) = ?$$

## Interval arithmetic

$$\begin{aligned} [-1,3] + [2,5] &= [1,8], \\ [-1,3] \cdot [2,5] &= [-5,15], \\ \text{abs}([-7,1]) &= [0,7] \end{aligned}$$

The interval extension of

$$f(x_1, x_2) = x_1 \cdot x_2 - (x_1 + x_2) \cdot \cos x_2 + \sin x_1 \cdot \sin x_2 + 2$$

is

$$\begin{aligned}[f]([x_1], [x_2]) &= [x_1] \cdot [x_2] - ([x_1] + [x_2]) \cdot \cos[x_2] \\ &\quad + \sin[x_1] \cdot \sin[x_2] + 2.\end{aligned}$$

## Theorem (Moore, 1970)

$$[f]([\mathbf{x}]) \subset \mathbb{R}^+ \Rightarrow \forall \mathbf{x} \in [\mathbf{x}], f(\mathbf{x}) \geq 0.$$

# Interval arithmetic

If  $\diamond \in \{+, -, \cdot, /, \max, \min\}$

$$[x] \diamond [y] = [\{x \diamond y \mid x \in [x], y \in [y]\}].$$

where  $[\mathbb{A}]$  is the smallest interval which encloses  $\mathbb{A} \subset \mathbb{R}$ .

## Exercise.

$$[-1, 3] + [2, 5] = [?, ?]$$

$$[-1, 3] \cdot [2, 5] = [?, ?]$$

$$[-2, 6] / [2, 5] = [?, ?]$$

## Solution.

$$[-1, 3] + [2, 5] = [1, 8]$$

$$[-1, 3] \cdot [2, 5] = [-5, 15]$$

$$[-2, 6] / [2, 5] = [-1, 3]$$

**Exercise.** Compute

$$[-2, 2] / [-1, 1] = [?, ?]$$

## Solution.

$$[-2, 2] / [-1, 1] = [-\infty, \infty]$$

$$\begin{aligned}[x^-, x^+] + [y^-, y^+] &= [x^- + y^-, x^+ + y^+], \\ [x^-, x^+] \cdot [y^-, y^+] &= [x^- y^- \wedge x^+ y^- \wedge x^- y^+ \wedge x^+ y^+, \\ &\quad x^- y^- \vee x^+ y^- \vee x^- y^+ \vee x^+ y^+],\end{aligned}$$

If  $f \in \{\cos, \sin, \text{sqr}, \sqrt{ }, \log, \exp, \dots\}$

$$f([x]) = [\{f(x) \mid x \in [x]\}].$$

## Exercise.

$$\sin([0, \pi]) = ?$$

$$\text{sqr}([-1, 3]) = [-1, 3]^2 = ?$$

$$\text{abs}([-7, 1]) = ?$$

$$\text{sqrt}([-10, 4]) = \sqrt{[-10, 4]} = ?$$

$$\log([-2, -1]) = ?.$$

## Solution.

$$\sin([0, \pi]) = [0, 1]$$

$$\text{sqr}([-1, 3]) = [-1, 3]^2 = [0, 9]$$

$$\text{abs}([-7, 1]) = [0, 7]$$

$$\text{sqrt}([-10, 4]) = \sqrt{[-10, 4]} = [0, 2]$$

$$\log([-2, -1]) = \emptyset.$$

# Inclusion functions

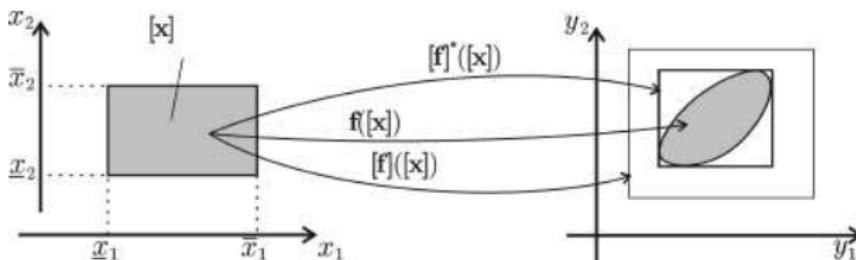
A *box*, or *interval vector*  $[\mathbf{x}]$  of  $\mathbb{R}^n$  is

$$[\mathbf{x}] = [x_1^-, x_1^+] \times \cdots \times [x_n^-, x_n^+] = [x_1] \times \cdots \times [x_n].$$

The set of all boxes of  $\mathbb{R}^n$  will be denoted by  $\mathbb{IR}^n$ .

$[\mathbf{f}] : \mathbb{IR}^n \rightarrow \mathbb{IR}^m$  is an *inclusion function* for  $\mathbf{f}$  if

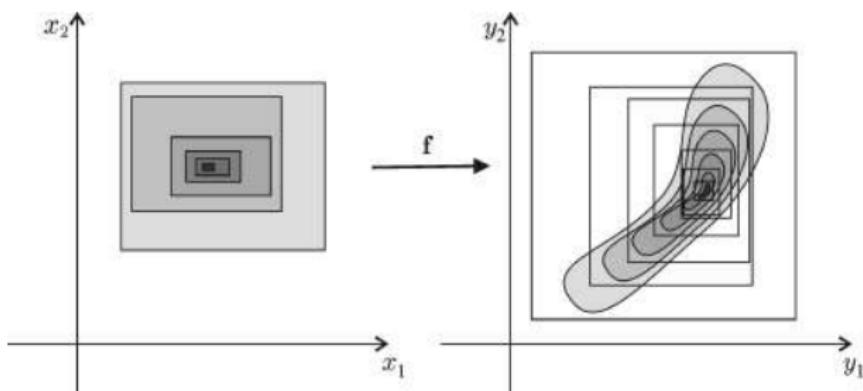
$$\forall [\mathbf{x}] \in \mathbb{IR}^n, \quad \mathbf{f}([\mathbf{x}]) \subset [\mathbf{f}]([\mathbf{x}]).$$



Inclusion functions  $[\mathbf{f}]$  and  $[\mathbf{f}]^*$ ; here,  $[\mathbf{f}]^*$  is minimal.

The inclusion function  $[\mathbf{f}]$  is

<i>monotonic</i>	if	$([\mathbf{x}] \subset [\mathbf{y}]) \Rightarrow ([\mathbf{f}](\mathbf{x}) \subset [\mathbf{f}](\mathbf{y}))$
<i>minimal</i>	if	$\forall \mathbf{x} \in \mathbb{IR}^n, [\mathbf{f}](\mathbf{x}) = [\mathbf{f}([\mathbf{x}])]$
<i>thin</i>	if	$w([\mathbf{x}]) = 0 \Rightarrow w([\mathbf{f}](\mathbf{x})) = 0$
<i>convergent</i>	if	$w([\mathbf{x}]) \rightarrow 0 \Rightarrow w([\mathbf{f}](\mathbf{x})) \rightarrow 0.$



**Exercise.** The natural inclusion function for  $f(x) = x^2 + 2x + 4$  is

$$[f]([x]) = [x]^2 + 2[x] + 4.$$

For  $[x] = [-3, 4]$ , compute  $[f]([x])$  and  $f([x])$ .

**Solution.** If  $[x] = [-3, 4]$ , we have

$$\begin{aligned}[f]([-3, 4]) &= [-3, 4]^2 + 2[-3, 4] + 4 \\&= [0, 16] + [-6, 8] + 4 \\&= [-2, 28].\end{aligned}$$

Note that  $f([-3, 4]) = [3, 28] \subset [f]([-3, 4]) = [-2, 28]$ .

A minimal inclusion function for

$$\begin{aligned}\mathbf{f}: \quad & \mathbb{R}^2 & \rightarrow & \mathbb{R}^3 \\ (x_1, x_2) & \mapsto & (x_1 x_2, x_1^2, x_1 - x_2).\end{aligned}$$

is

$$\begin{aligned}[\mathbf{f}]: \quad & \mathbb{IR}^2 & \rightarrow & \mathbb{IR}^3 \\ ([x_1], [x_2]) & \rightarrow & ([x_1] \cdot [x_2], [x_1]^2, [x_1] - [x_2]).\end{aligned}$$

If  $\mathbf{f}$  is given by

**Algorithm**  $\mathbf{f}(\text{in : } \mathbf{x} = (x_1, x_2, x_3), \text{ out : } \mathbf{y} = (y_1, y_2))$

```
z := x1
fork := 0 to 100
    z := x2(z + k · x3)
next
y1 := z
y2 := sin(zx1)
```

Its natural inclusion function is

**Algorithm**  $\mathbf{f}(\text{in} : [\mathbf{x}] = ([x_1], [x_2], [x_3]), \text{out} : [\mathbf{y}] = ([y_1], [y_2]))$

$[z] := [x_1]$

fork := 0 to 100

$[z] := [x_2] \cdot ([z] + k \cdot [x_3])$

next

$[y_1] := [z]$

$[y_2] := \sin([z] \cdot [x_1])$

Is  $\mathbf{f}$  convergent? thin? monotonic?

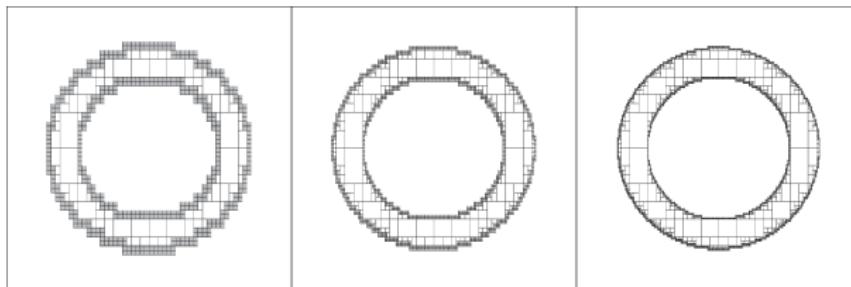
# Set inversion

A subpaving of  $\mathbb{R}^n$  is a set of non-overlapping boxes of  $\mathbb{R}^n$ .  
Compact sets  $X$  can be bracketed between inner and outer  
subpavings:

$$X^- \subset X \subset X^+.$$

## Example.

$$\mathbb{X} = \{(x_1, x_2) \mid x_1^2 + x_2^2 \in [1, 2]\}.$$



Let  $\mathbf{f}: \mathbb{R}^n \rightarrow \mathbb{R}^m$  and let  $\mathbb{Y}$  be a subset of  $\mathbb{R}^m$ . Set inversion is the characterization of

$$\mathbb{X} = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{f}(\mathbf{x}) \in \mathbb{Y}\} = \mathbf{f}^{-1}(\mathbb{Y}).$$

We shall use the following tests.

- (i)  $[\mathbf{f}](\mathbf{[x]}) \subset \mathbb{Y} \Rightarrow \mathbf{[x]} \subset \mathbb{X}$
- (ii)  $[\mathbf{f}](\mathbf{[x]}) \cap \mathbb{Y} = \emptyset \Rightarrow \mathbf{[x]} \cap \mathbb{X} = \emptyset.$

Boxes for which these tests failed, will be bisected, except if they are too small.

# Localization

A robot measures distances to three beacons.

$i$	$x_i$	$y_i$	$[d_i]$
1	1	3	[1,2]
2	3	1	[2,3]
3	-1	-1	[3,4]

The intervals  $[d_i]$  contain the true distance with a probability of  $\pi = 0.9$ .

Define

$$\mathbb{P}_i = \left\{ \mathbf{p} \in \mathbb{R}^2 \mid \sqrt{(p_1 - x_i)^2 + (p_2 - y_i)^2} \in [d_i] \right\}.$$

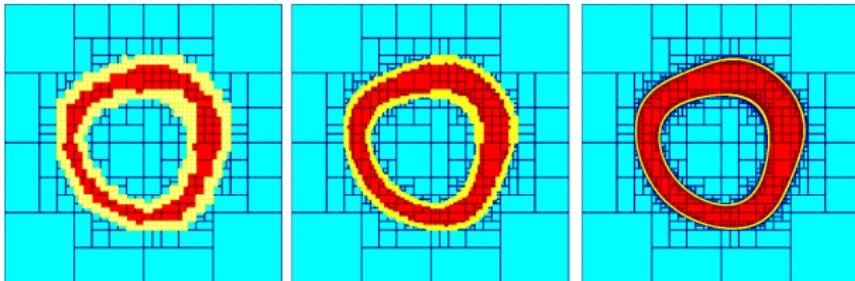
$$\text{prob}(\mathbf{p} \in \mathbb{P}^{\{0\}}) = 0.729$$

$$\text{prob}(\mathbf{p} \in \mathbb{P}^{\{1\}}) = 0.972$$

$$\text{prob}(\mathbf{p} \in \mathbb{P}^{\{2\}}) = 0.999$$



# Contractors

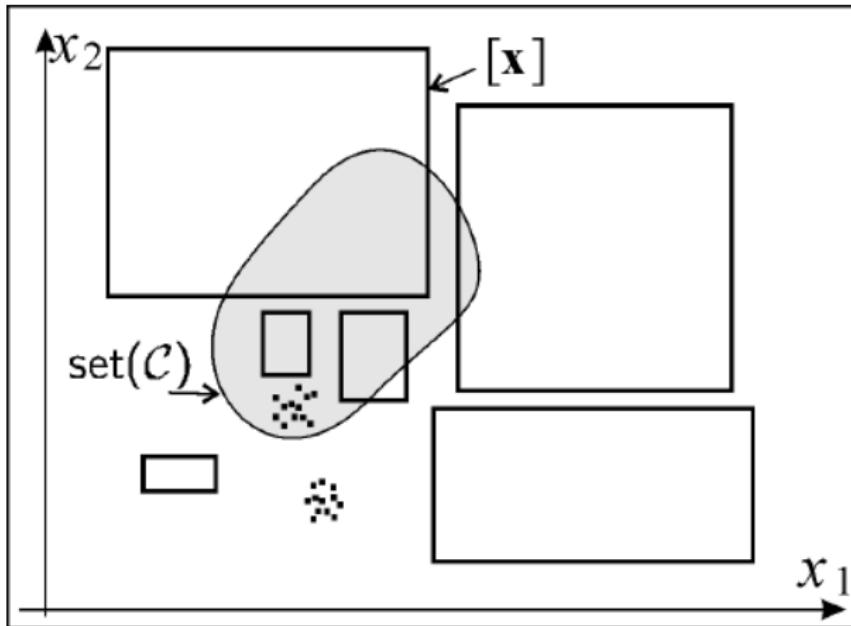


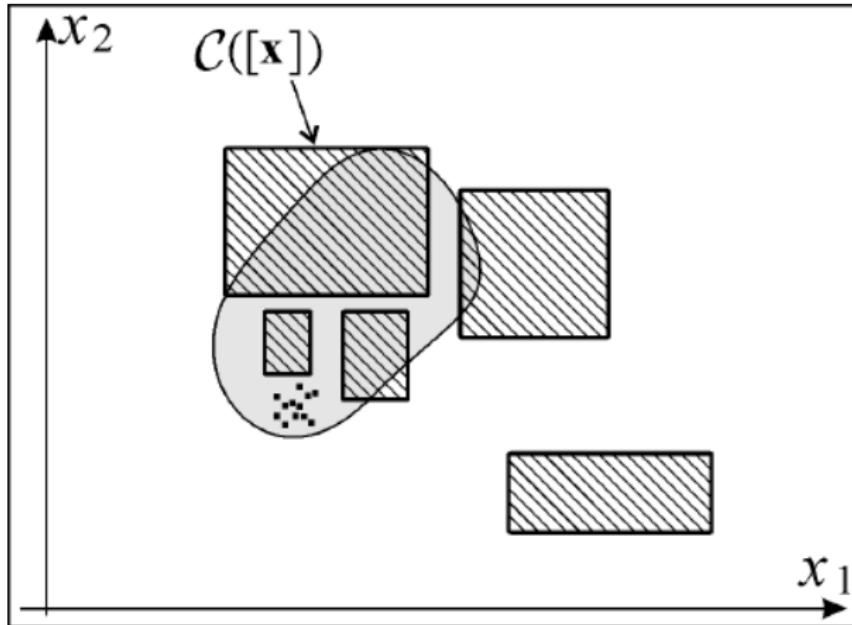
To characterize  $\mathbb{X} \subset \mathbb{R}^n$ , bisection algorithms bisect all boxes in all directions and become inefficient. Interval methods can still be useful if

- the solution set  $\mathbb{X}$  is small (optimization problem, solving equations),
- contraction procedures are used as much as possible,
- bisections are used only as a last resort.

The operator  $\mathcal{C} : \mathbb{IR}^n \rightarrow \mathbb{IR}^n$  is a *contractor* for  $\mathbb{X} \subset \mathbb{R}^n$  if

$$\forall [\mathbf{x}] \in \mathbb{IR}^n, \left\{ \begin{array}{ll} \mathcal{C}([\mathbf{x}]) \subset [\mathbf{x}] & \text{(contractance),} \\ \mathcal{C}([\mathbf{x}]) \cap \mathbb{X} = [\mathbf{x}] \cap \mathbb{X} & \text{(completeness).} \end{array} \right.$$





The operator  $\mathcal{C} : \mathbb{IR}^n \rightarrow \mathbb{IR}^n$  is a *contractor* for the equation  $f(\mathbf{x}) = 0$ , if

$$\forall [\mathbf{x}] \in \mathbb{IR}^n, \left\{ \begin{array}{l} \mathcal{C}([\mathbf{x}]) \subset [\mathbf{x}] \\ \mathbf{x} \in [\mathbf{x}] \text{ et } f(\mathbf{x}) = 0 \Rightarrow \mathbf{x} \in \mathcal{C}([\mathbf{x}]) \end{array} \right.$$

**Exercice.** Let  $x, y, z$  be 3 variables such that

$$x \in [-\infty, 5],$$

$$y \in [-\infty, 4],$$

$$z \in [6, \infty],$$

$$z = x + y.$$

Contract the intervals for  $x, y, z$ .

## Solution.

$$[x] = [2, 5]$$

$$[y] = [1, 4]$$

$$[z] = [6, 9]$$

Since  $x \in [-\infty, 5]$ ,  $y \in [-\infty, 4]$ ,  $z \in [6, \infty]$  and  $z = x + y$ , we have

$$\begin{aligned} z = x + y &\Rightarrow z \in [6, \infty] \cap ([-\infty, 5] + [-\infty, 4]) \\ &= [6, \infty] \cap [-\infty, 9] = [6, 9]. \end{aligned}$$

$$\begin{aligned} x = z - y &\Rightarrow x \in [-\infty, 5] \cap ([6, \infty] - [-\infty, 4]) \\ &= [-\infty, 5] \cap [2, \infty] = [2, 5]. \end{aligned}$$

$$\begin{aligned} y = z - x &\Rightarrow y \in [-\infty, 4] \cap ([6, \infty] - [-\infty, 5]) \\ &= [-\infty, 4] \cap [1, \infty] = [1, 4]. \end{aligned}$$

The contractor associated with  $z = x + y$  is:

**Algorithm pplus(inout:  $[z]$ ,  $[x]$ ,  $[y]$ )**

$[z] := [z] \cap ([x] + [y])$	// $z = x + y$
$[x] := [x] \cap ([z] - [y])$	// $x = z - y$
$[y] := [y] \cap ([z] - [x])$	// $y = z - x$

The contractor associated with  $z = x \cdot y$  is:

**Algorithm pmult (inout:  $[z], [x], [y]$ )**

$[z] := [z] \cap ([x] \cdot [y])$	// $z = x \cdot y$
$[x] := [x] \cap ([z] \cdot 1/[y])$	// $x = z/y$
$[y] := [y] \cap ([z] \cdot 1/[x])$	// $y = z/x$

The contractor associated with  $y = \exp x$  is:

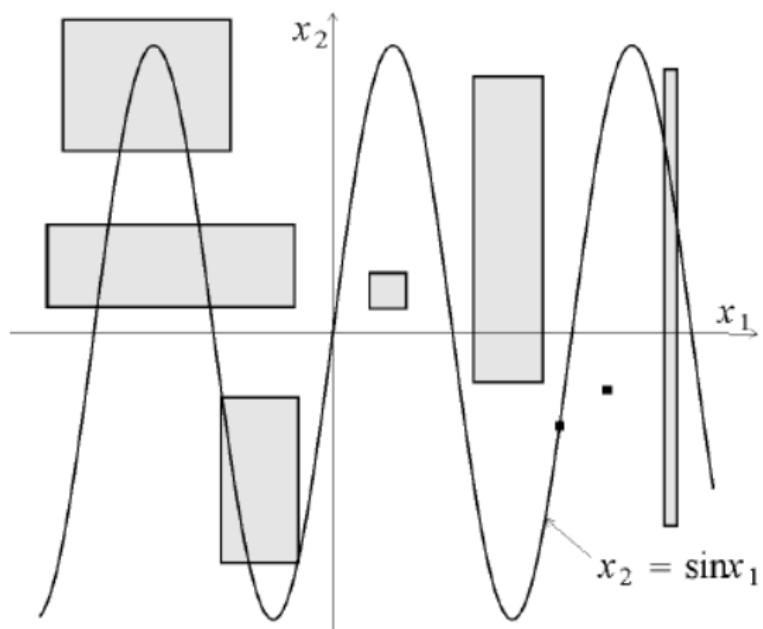
**Algorithm** pexp (inout:  $[y], [x]$ )

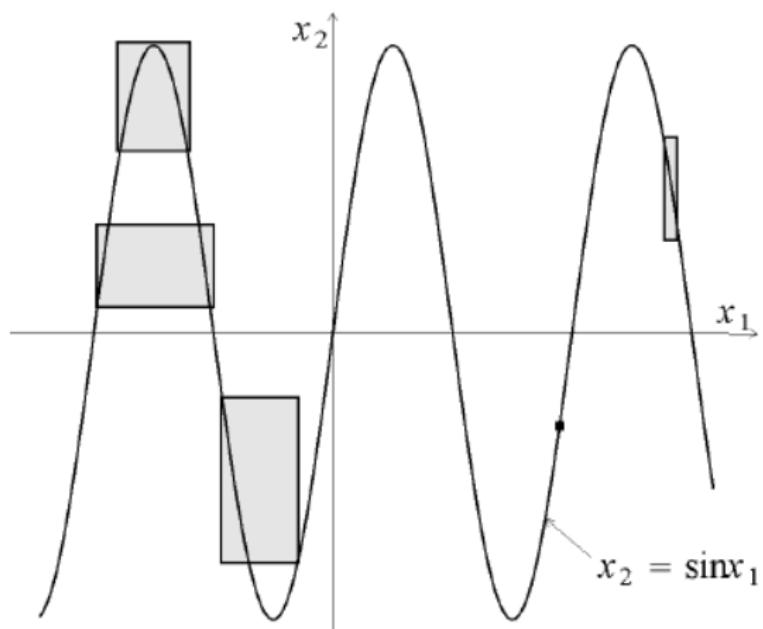
- 1  $[y] := [y] \cap \exp([x])$
- 2  $[x] := [x] \cap \log([y])$

Any constraint for which such a projection procedure is available will be called a *primitive constraint*.

**Example.** Consider the primitive equation:

$$x_2 = \sin x_1.$$





## Decomposition

$$x + \sin(xy) \leq 0, \\ x \in [-1, 1], y \in [-1, 1]$$

## Decomposition

$$\begin{aligned}x + \sin(xy) &\leq 0, \\x \in [-1, 1], y \in [-1, 1]\end{aligned}$$

can be decomposed into

$$\left\{ \begin{array}{lll} a = xy & x \in [-1, 1] & a \in [-\infty, \infty] \\ b = \sin(a) & , \quad y \in [-1, 1] & b \in [-\infty, \infty] \\ c = x + b & & c \in [-\infty, 0] \end{array} \right.$$

# Forward Backward contractor

For the equation

$$(x_1 + x_2) \cdot x_3 \in [1, 2],$$

we decompose into

$$\begin{aligned} a &= x_1 + x_2 \\ b &= a \cdot x_3 \\ b &\in [1, 2] \end{aligned}$$

For the equation

$$(x_1 + x_2) \cdot x_3 \in [1, 2],$$

we have the following contractor:

**Algorithm  $\mathcal{C}$  (inout $[x_1], [x_2], [x_3]$ )**

$$[a] = [x_1] + [x_2] \quad // a = x_1 + x_2$$

$$[b] = [a] \cdot [x_3] \quad // b = a \cdot x_3$$

$$[b] = [b] \cap [1, 2] \quad // b \in [1, 2]$$

$$[x_3] = [x_3] \cap \frac{[b]}{[a]} \quad // x_3 = \frac{b}{a}$$

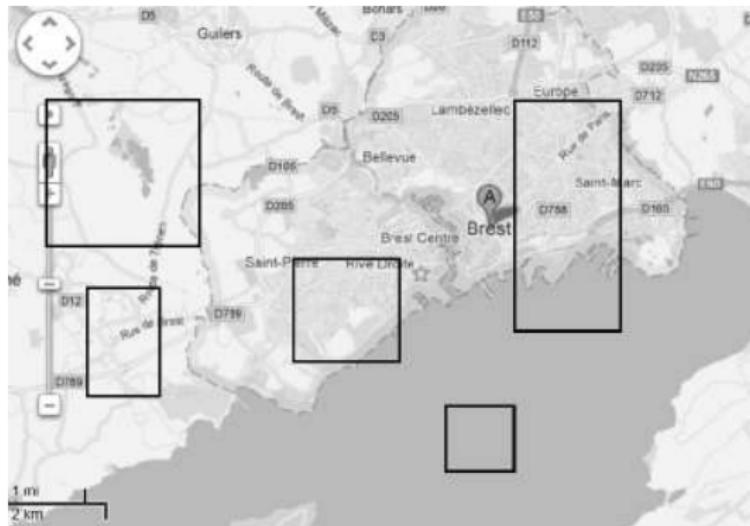
$$[a] = [a] \cap \frac{[b]}{[x_3]} \quad // a = \frac{b}{x_3}$$

$$[x_1] = [x_1] \cap [a] - [x_2] \quad // x_1 = a - x_2$$

$$[x_2] = [x_2] \cap [a] - [x_1] \quad // x_2 = a - x_1$$

## Contractor on images

The robot with coordinates  $(x_1, x_2)$  is in the water.





# Solving equations

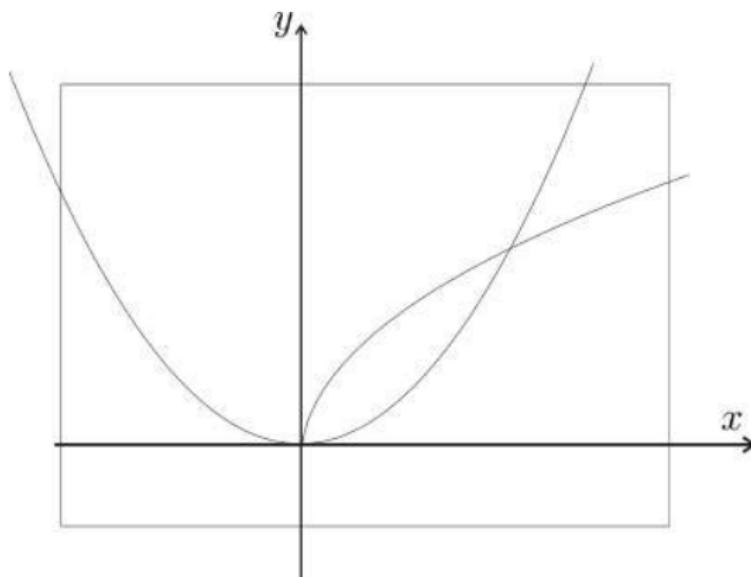
Consider the system of two equations.

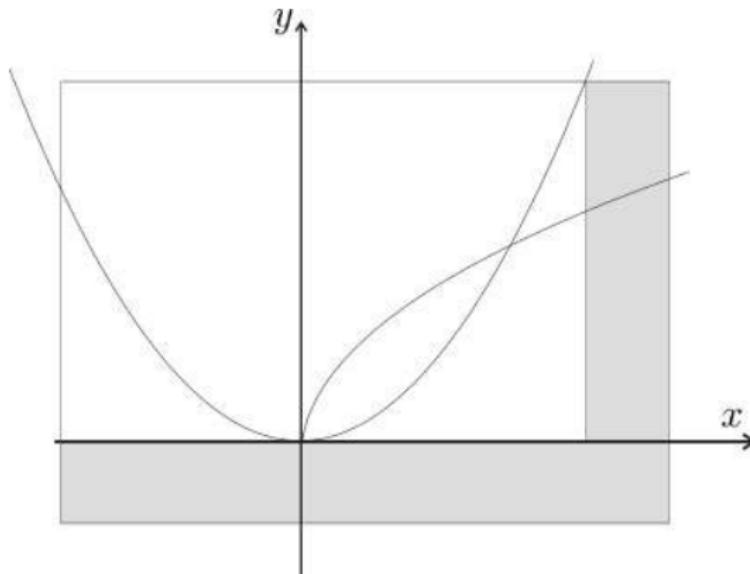
$$\begin{aligned}y &= x^2 \\y &= \sqrt{x}.\end{aligned}$$

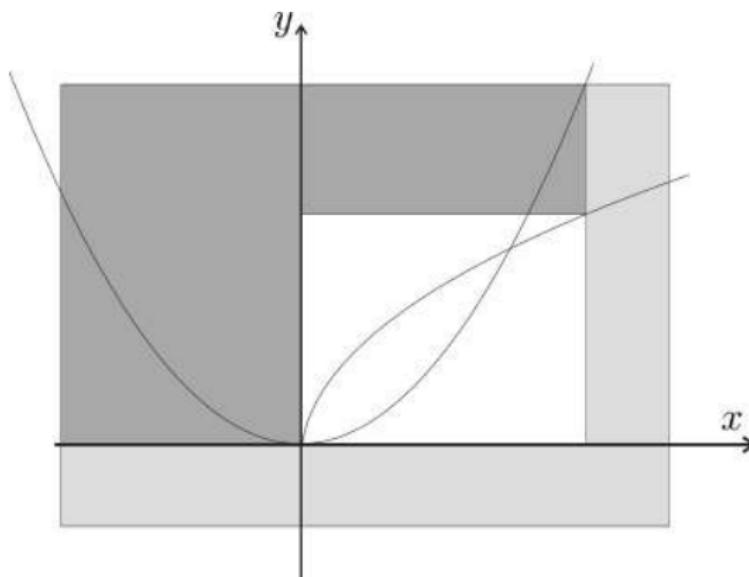
We can build two contractors

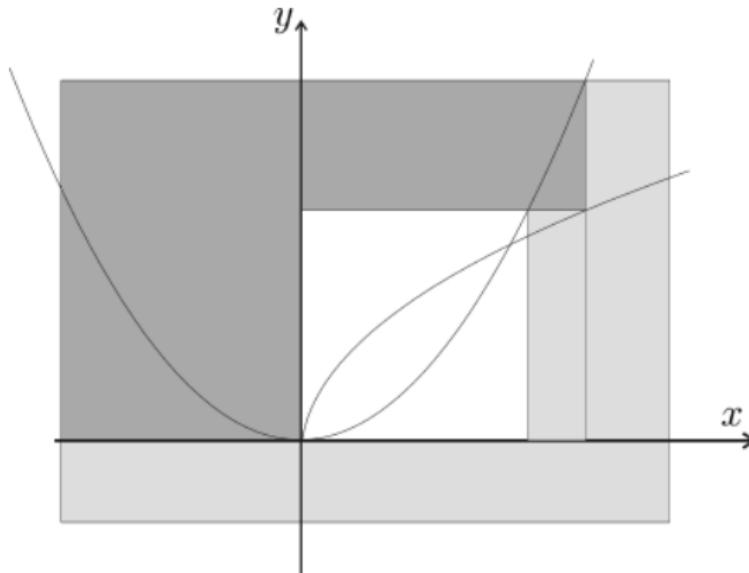
$$\mathcal{C}_1 : \begin{cases} [y] = [y] \cap [x]^2 \\ [x] = [x] \cap \sqrt{[y]} \end{cases} \quad \text{associated to } y = x^2$$

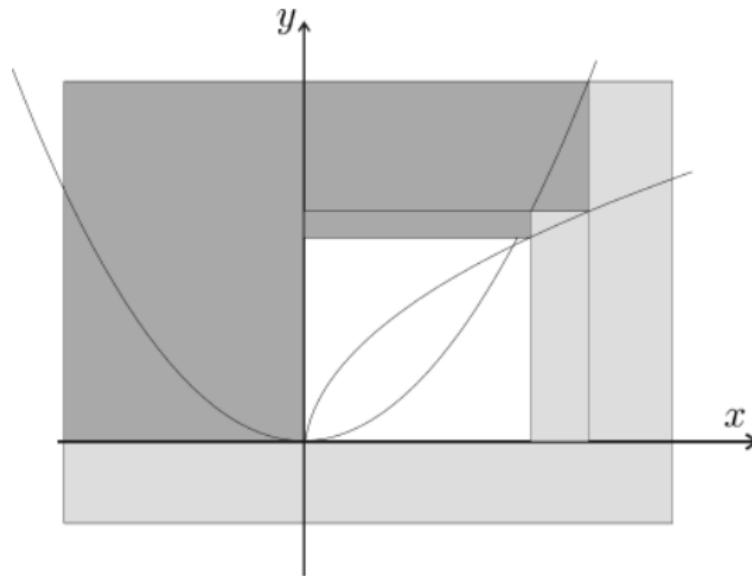
$$\mathcal{C}_2 : \begin{cases} [y] = [y] \cap \sqrt{[x]} \\ [x] = [x] \cap [y]^2 \end{cases} \quad \text{associated to } y = \sqrt{x}$$

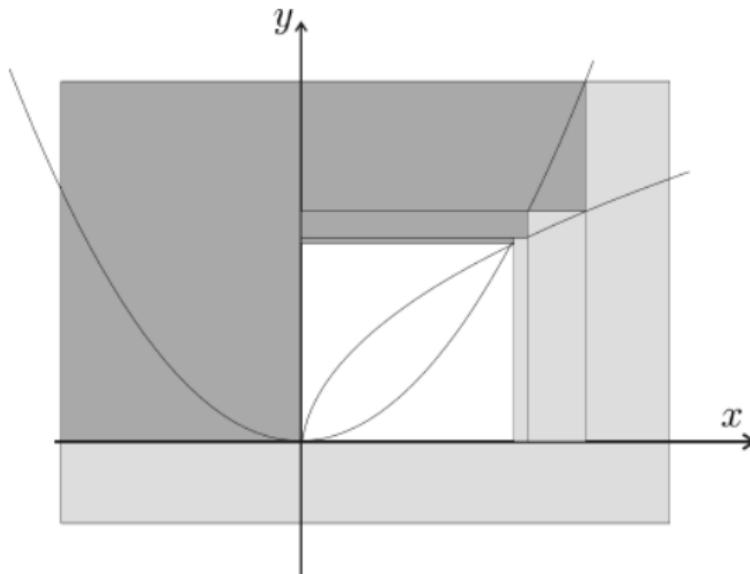


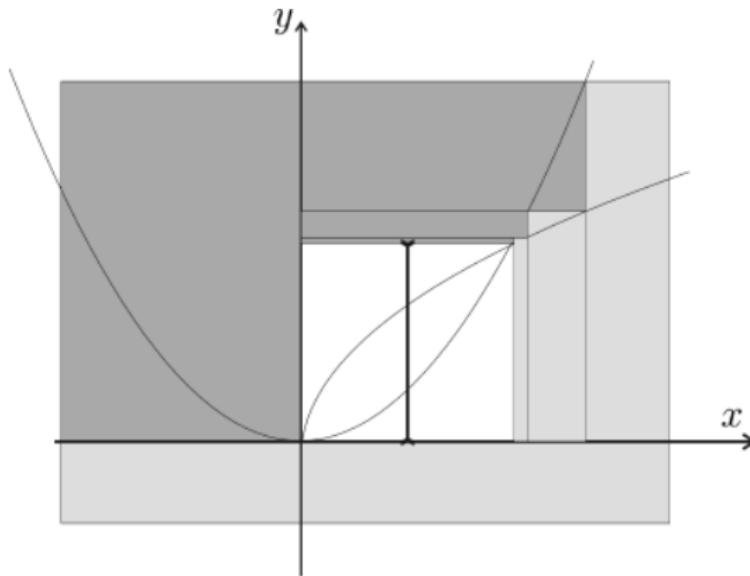


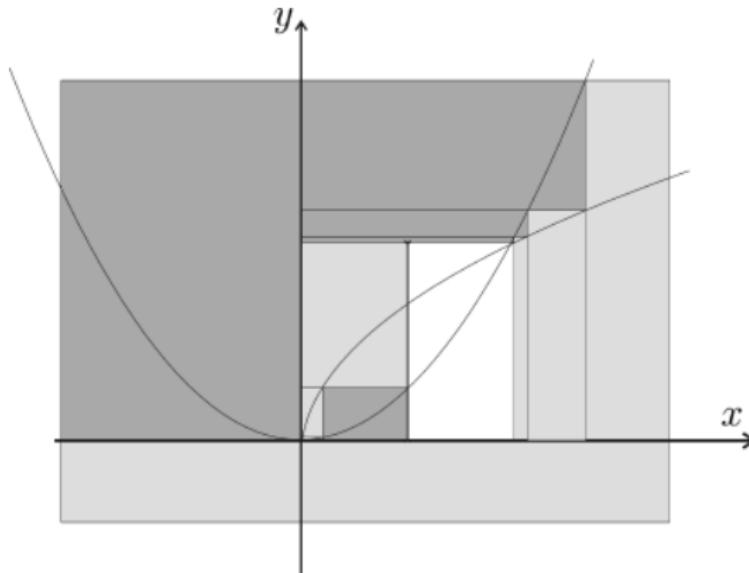


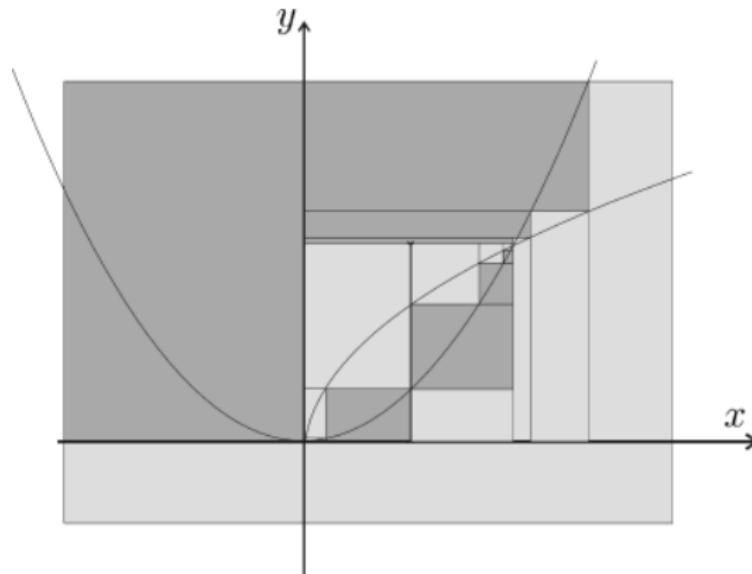








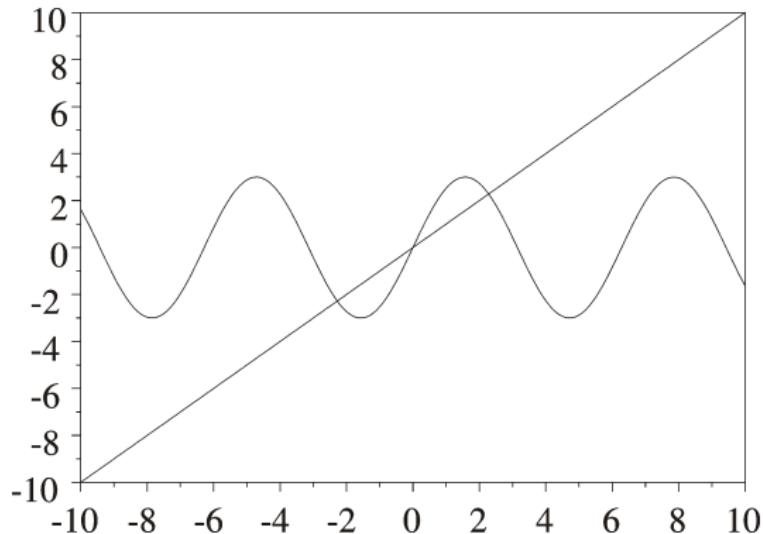


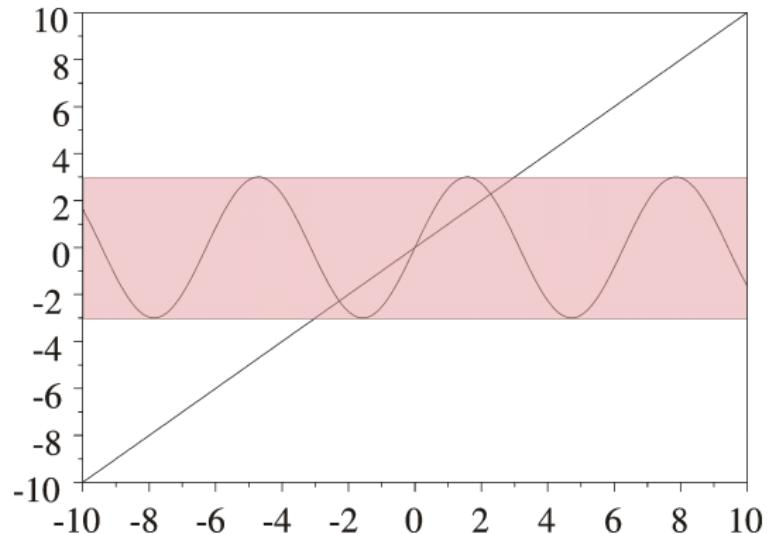


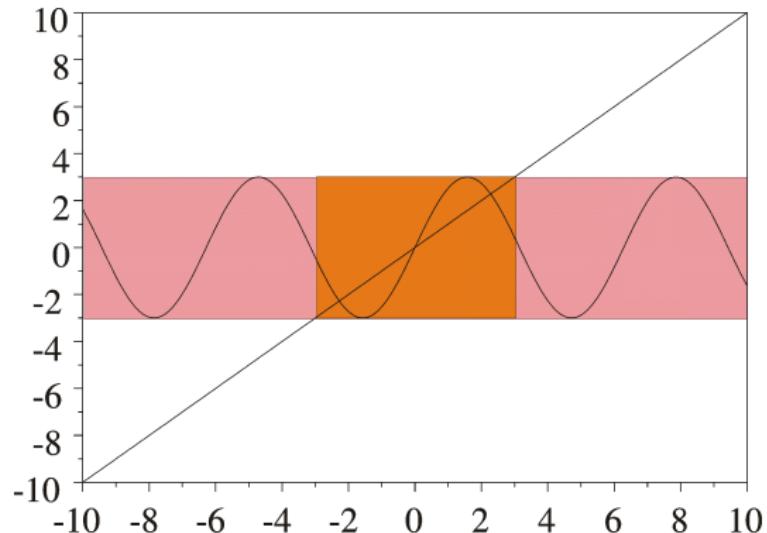
# Another example

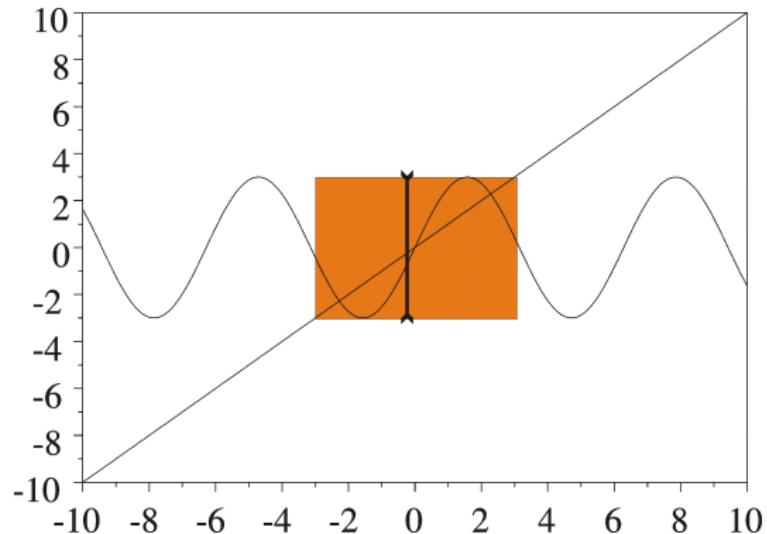
**Exemple.** Consider the system

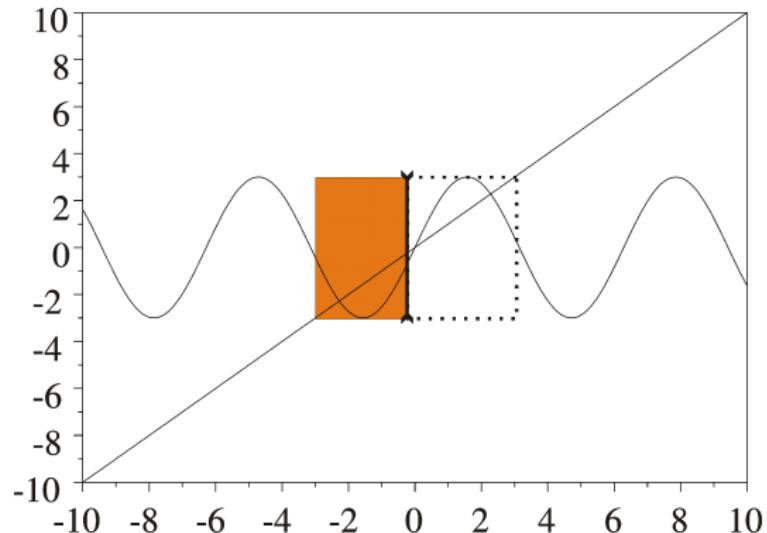
$$\begin{cases} y = 3 \sin(x) \\ y = x \end{cases} \quad x \in \mathbb{R}, y \in \mathbb{R}.$$

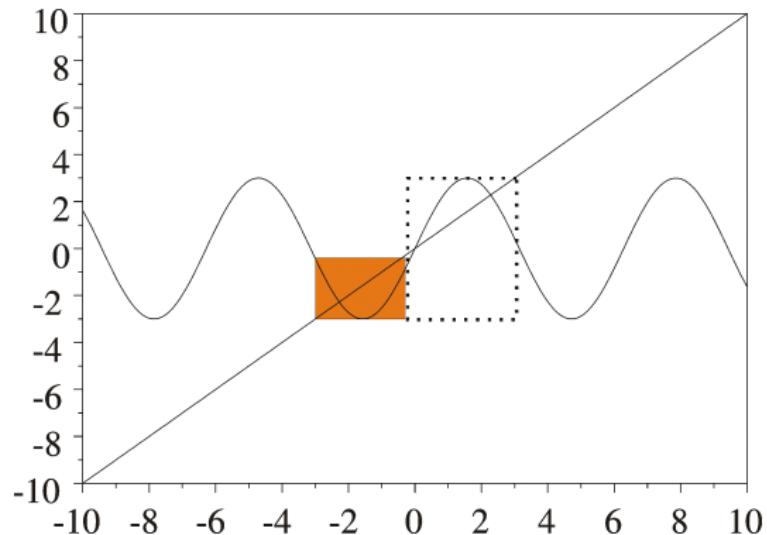


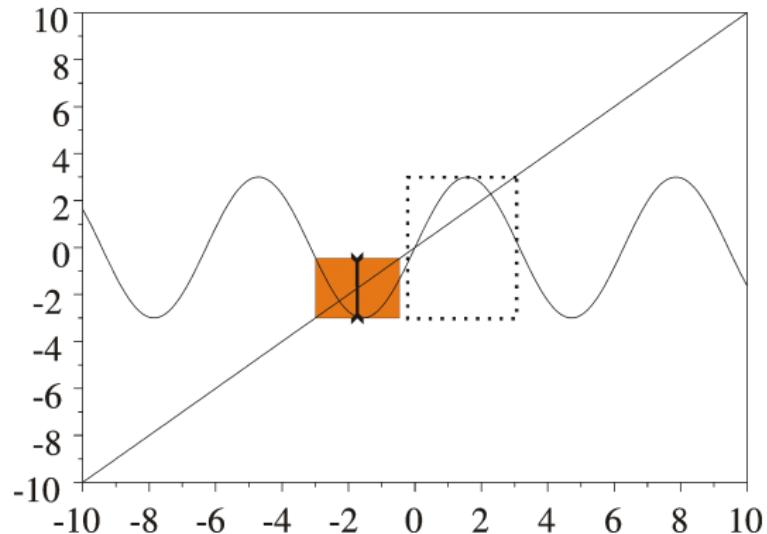


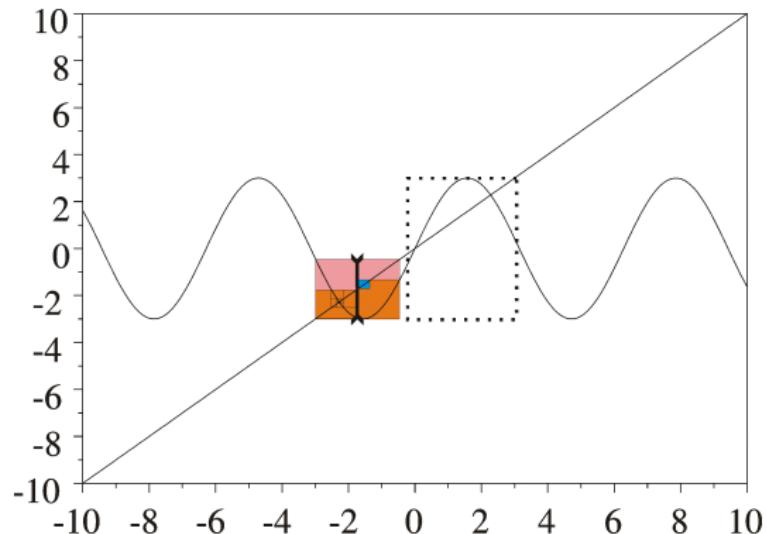


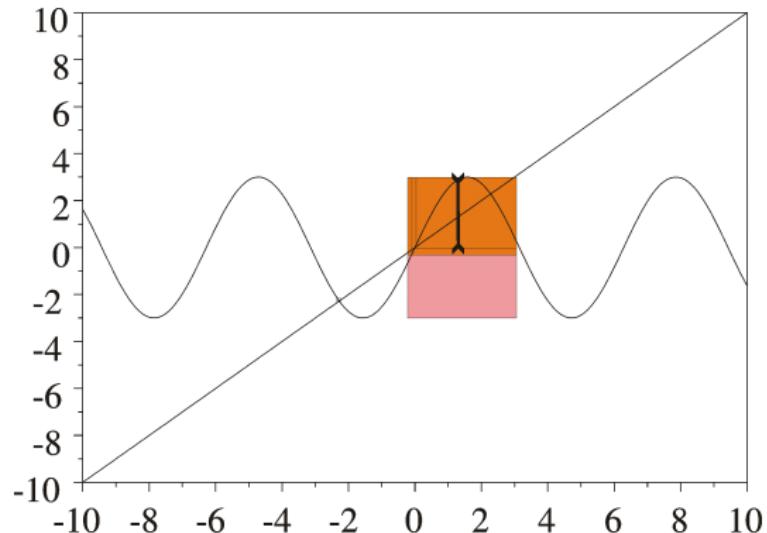


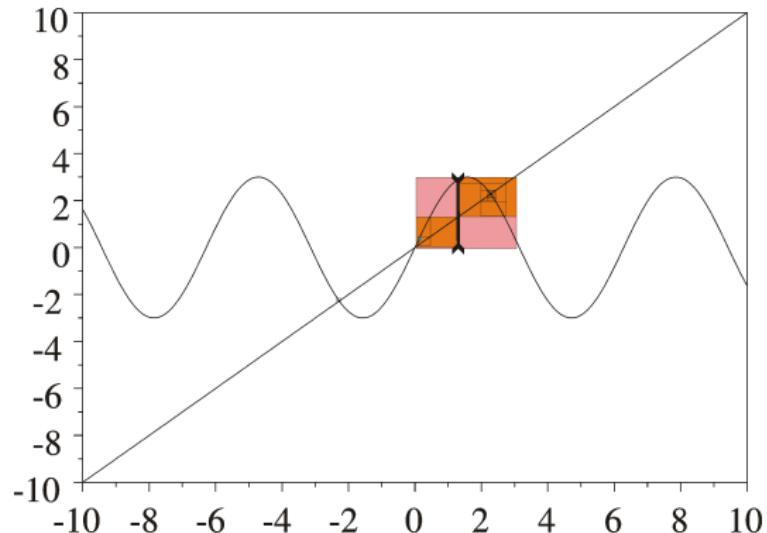












# Contractor algebra

intersection	$(\mathcal{C}_1 \cap \mathcal{C}_2)([\mathbf{x}]) = \mathcal{C}_1([\mathbf{x}]) \cap \mathcal{C}_2([\mathbf{x}])$
union	$(\mathcal{C}_1 \cup \mathcal{C}_2)([\mathbf{x}]) = [\mathcal{C}_1([\mathbf{x}]) \cup \mathcal{C}_2([\mathbf{x}])]$
composition	$(\mathcal{C}_1 \circ \mathcal{C}_2)([\mathbf{x}]) = \mathcal{C}_1(\mathcal{C}_2([\mathbf{x}])))$
repetition	$\mathcal{C}^\infty = \mathcal{C} \circ \mathcal{C} \circ \mathcal{C} \circ \dots$
repeat intersection	$\mathcal{C}_1 \sqcap \mathcal{C}_2 = (\mathcal{C}_1 \cap \mathcal{C}_2)^\infty$
repeat union	$\mathcal{C}_1 \sqcup \mathcal{C}_2 = (\mathcal{C}_1 \cup \mathcal{C}_2)^\infty$

# A link with matrices

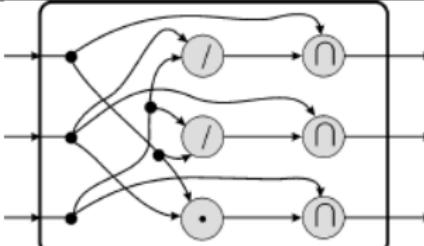
linear application       $\rightarrow$       matrices

$$\mathcal{L} : \begin{cases} \alpha = 2a + 3h \\ \gamma = h - 5a \end{cases} \rightarrow \mathbf{A} = \begin{pmatrix} 2 & 3 \\ 1 & -5 \end{pmatrix}$$

We have a matrix algebra and Matlab.

We have:  $\text{var}(\mathcal{L}) = \{a, h\}$ ,  $\text{covar}(\mathcal{L}) = \{\alpha, \gamma\}$ .

But we cannot write:  $\text{var}(\mathbf{A}) = \{a, h\}$ ,  $\text{covar}(\mathbf{A}) = \{\alpha, \gamma\}$ .

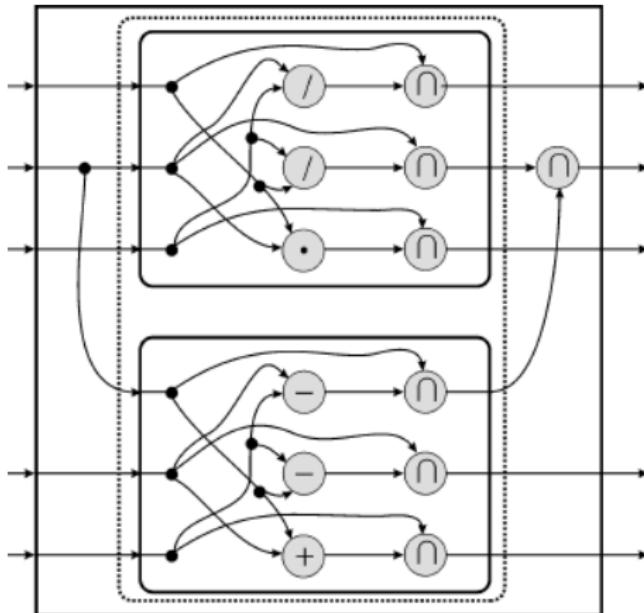
constraint	$\rightarrow$	contractor
$a \cdot b = z$	$\rightarrow$	

## Contractor fusion

$$\left\{ \begin{array}{l} a \cdot b = z \rightarrow \mathcal{C}_1 \\ b + c = d \rightarrow \mathcal{C}_2 \end{array} \right.$$

Since  $b$  occurs in both constraints, we fuse the two contractors as:

$$\begin{aligned}\mathcal{C} &= \mathcal{C}_1 \times \mathcal{C}_2|_{(2,1)} \\ &= \mathcal{C}_1 | \mathcal{C}_2 \text{ (for short)}\end{aligned}$$



# SLAM

$$\left\{ \begin{array}{l} \dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, \mathbf{u}) \quad (\text{evolution equation}) \\ \mathbf{y} = \mathbf{g}(\mathbf{x}, \mathbf{u}) \quad (\text{observation equation}) \\ \mathbf{z}_i = \mathbf{h}(\mathbf{x}, \mathbf{u}, \mathbf{m}_i) \quad (\text{mark equation}) \end{array} \right.$$



Redermor (GESMA, Brest)

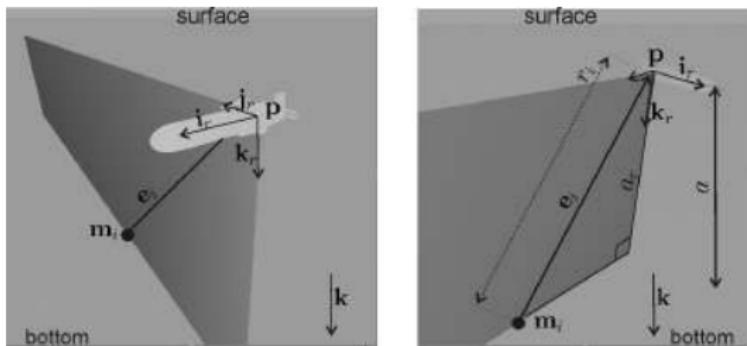
<https://youtu.be/X0lqZxb-tFs>

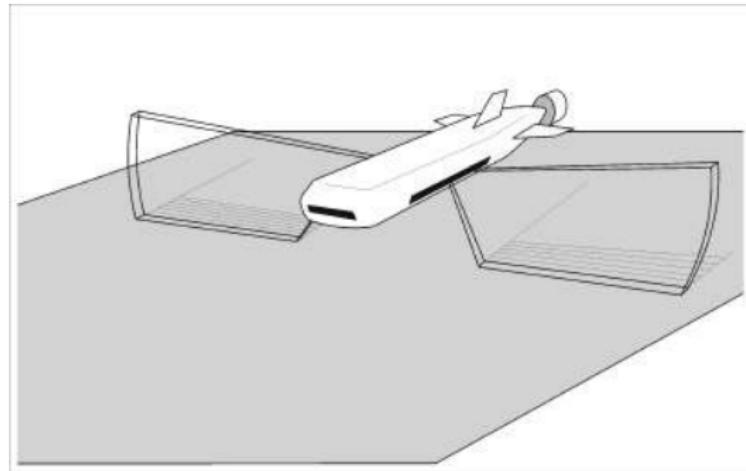


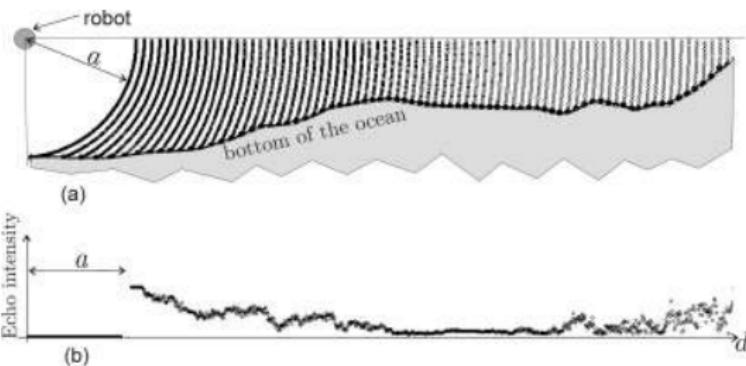
**GPS** (Global positioning system), only at the surface.

$$t_0 = 0000 \text{ s}, \quad \ell^0 = (-4.4582279^\circ, 48.2129206^\circ) \pm 2.5m$$
$$t_f = 6000 \text{ s}, \quad \ell^f = (-4.4546607^\circ, 48.2191297^\circ) \pm 2.5m$$

**Sonar (KLEIN 5400 side scan sonar).**

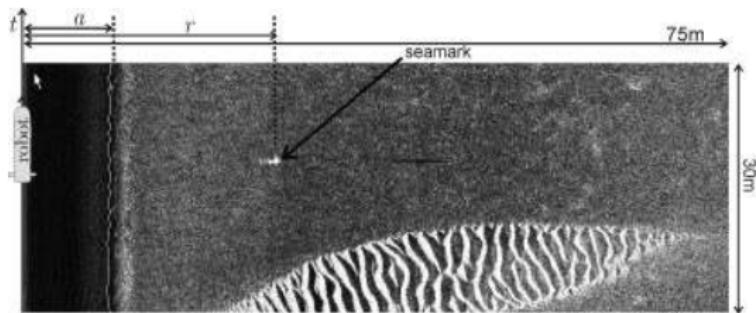








Screenshot of SonarPro



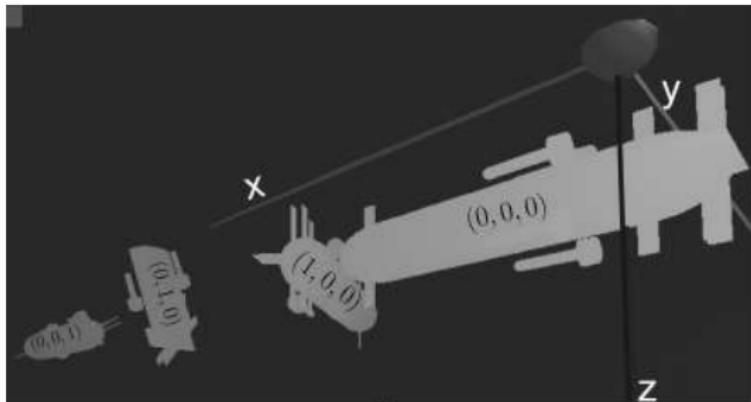
## Mine detection with SonarPro

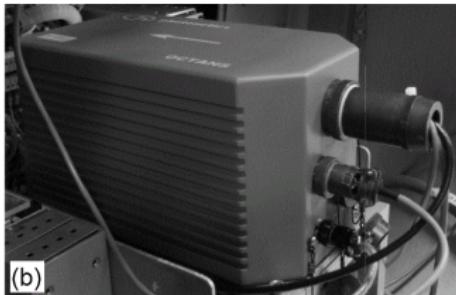
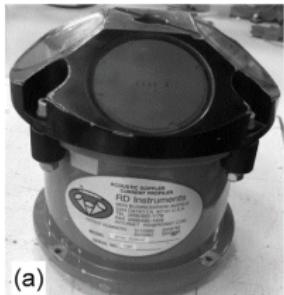
**Loch-Doppler** returns the speed robot  $\mathbf{v}_r$ .

$$\mathbf{v}_r \in \mathbf{x}_r + 0.004 * [-1, 1] . \mathbf{x}_r + 0.004 * [-1, 1]$$

## Inertial unit (Octans III).

$$\begin{pmatrix} \phi \\ \theta \\ \psi \end{pmatrix} \in \begin{pmatrix} \tilde{\phi} \\ \tilde{\theta} \\ \tilde{\psi} \end{pmatrix} + \begin{pmatrix} 1.75 \times 10^{-4} \cdot [-1, 1] \\ 1.75 \times 10^{-4} \cdot [-1, 1] \\ 5.27 \times 10^{-3} \cdot [-1, 1] \end{pmatrix}.$$





Six mines have been detected.

$i$	0	1	2	3	4	5
$\tau(i)$	1054	1092	1374	1748	3038	3688
$\sigma(i)$	1	2	1	0	1	5
$\tilde{r}(i)$	52.42	12.47	54.40	52.68	27.73	26.98

6	7	8	9	10	11
4024	4817	5172	5232	5279	5688
4	3	3	4	5	1
37.90	36.71	37.37	31.03	33.51	15.05

# Constraint network

$$t \in \{.0, 0.1, 0.2, \dots, 5999.9\},$$

$$i \in \{0, 1, \dots, 11\},$$

$$\begin{pmatrix} p_x(t) \\ p_y(t) \end{pmatrix} = 111120 \begin{pmatrix} 0 & 1 \\ \cos(\ell_y(t) * \frac{\pi}{180}) & 0 \end{pmatrix} \begin{pmatrix} \ell_x(t) - \ell_x^0 \\ \ell_y(t) - \ell_y^0 \end{pmatrix},$$

$$\mathbf{p}(t) = (p_x(t), p_y(t), p_z(t)),$$

$$\mathbf{R}_\psi(t) = \begin{pmatrix} \cos \psi(t) & -\sin \psi(t) & 0 \\ \sin \psi(t) & \cos \psi(t) & 0 \\ 0 & 0 & 1 \end{pmatrix},$$

$$\mathbf{R}_\theta(t) = \begin{pmatrix} \cos \theta(t) & 0 & \sin \theta(t) \\ 0 & 1 & 0 \\ -\sin \theta(t) & 0 & \cos \theta(t) \end{pmatrix},$$

$$\mathbf{R}_\varphi(t) = \begin{pmatrix} 1 & 0 & 0 \\ 0 & \cos \varphi(t) & -\sin \varphi(t) \\ 0 & \sin \varphi(t) & \cos \varphi(t) \end{pmatrix},$$

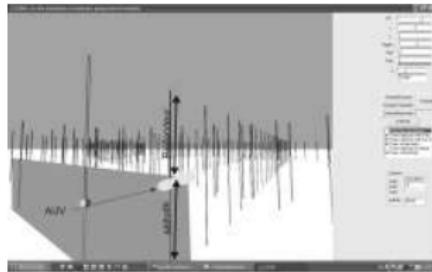
$$\mathbf{R}(t) = \mathbf{R}_\psi(t)\mathbf{R}_\theta(t)\mathbf{R}_\varphi(t),$$

$$\dot{\mathbf{p}}(t) = \mathbf{R}(t).\mathbf{v}_r(t),$$

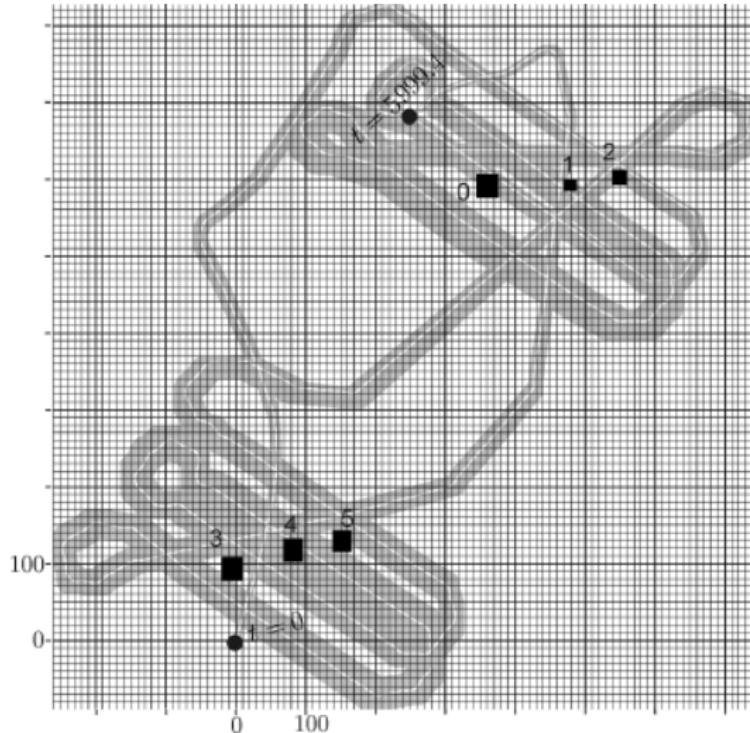
$$||\mathbf{m}(\sigma(i)) - \mathbf{p}(\tau(i))|| = r(i),$$

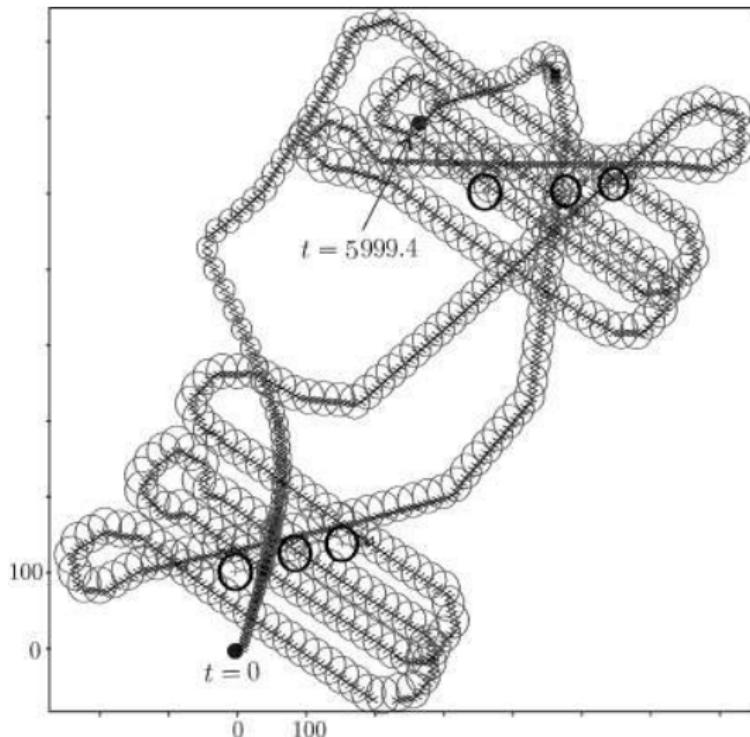
$$\mathbf{R}^\top(\tau(i))(\mathbf{m}(\sigma(i)) - \mathbf{p}(\tau(i))) \in [0] \times [0, \infty]^{\times 2},$$

$$m_z(\sigma(i)) - p_z(\tau(i)) - a(\tau(i)) \in [-0.5, 0.5]$$



[youtu.be/lzJtAfAT7h4](https://youtu.be/lzJtAfAT7h4)





# Tubes

A trajectory is a function  $\mathbf{f}: \mathbb{R} \rightarrow \mathbb{R}^n$ . For instance

$$\mathbf{f}(t) = \begin{pmatrix} \cos t \\ \sin t \end{pmatrix}$$

is a trajectory.

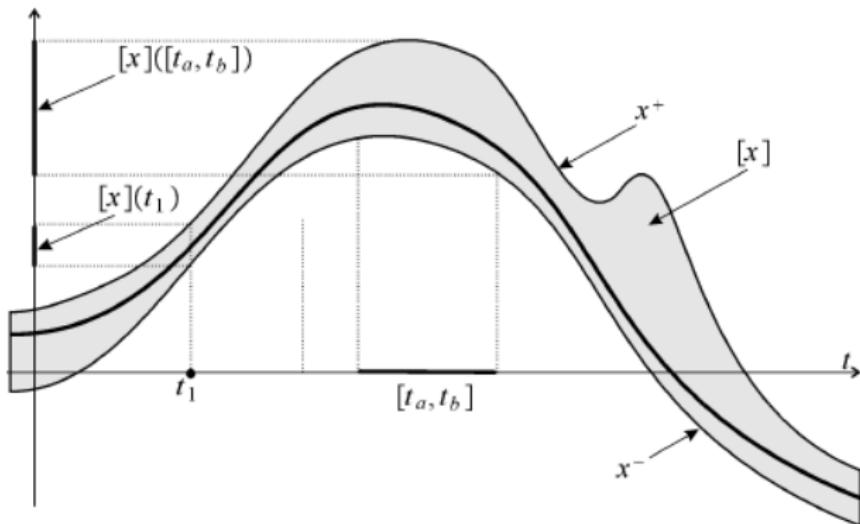
## Order relation

$$\mathbf{f} \leq \mathbf{g} \Leftrightarrow \forall t, \forall i, f_i(t) \leq g_i(t).$$

We have

$$\mathbf{h} = \mathbf{f} \wedge \mathbf{g} \Leftrightarrow \forall t, \forall i, h_i(t) = \min(f_i(t), g_i(t)),$$

$$\mathbf{h} = \mathbf{f} \vee \mathbf{g} \Leftrightarrow \forall t, \forall i, h_i(t) = \max(f_i(t), g_i(t)).$$



The set of trajectories is a lattice. Interval of trajectories (tubes) can be defined.

## Example.

$$[\mathbf{f}](t) = \begin{pmatrix} \cos t + [0, t^2] \\ \sin t + [-1, 1] \end{pmatrix}$$

is an interval trajectory (or tube).

# Tube arithmetic

If  $[x]$  and  $[y]$  are two scalar tubes [1], we have

$$[z] = [x] + [y] \Rightarrow [z](t) = [x](t) + [y](t) \quad (\text{sum})$$

$$[z] = \text{shift}_a([x]) \Rightarrow [z](t) = [x](t+a) \quad (\text{shift})$$

$$[z] = [x] \circ [y] \Rightarrow [z](t) = [x]([y](t)) \quad (\text{composition})$$

$$[z] = \int [x] \Rightarrow [z](t) = \left[ \int_0^t x^-(\tau) d\tau, \int_0^t x^+(\tau) d\tau \right] \quad (\text{integral})$$

# Tube contractor

Tube arithmetic allows us to build contractors.

Consider for instance the differential constraint

$$\begin{aligned}\dot{x}(t) &= x(t+1) \cdot u(t), \\ x(t) &\in [x](t), \dot{x}(t) \in [\dot{x}](t), u(t) \in [u](t)\end{aligned}$$

We decompose as follows

$$\begin{cases} x(t) = x(0) + \int_0^t y(\tau) d\tau \\ y(t) = a(t) \cdot u(t). \\ a(t) = x(t+1) \end{cases}$$

Possible contractors are

$$\left\{ \begin{array}{lcl} [x](t) & = & [x](t) \cap ([x](0) + \int_0^t [y](\tau) d\tau) \\ [y](t) & = & [y](t) \cap [a](t) \cdot [u](t) \\ [u](t) & = & [u](t) \cap \frac{[y](t)}{[a](t)} \\ [a](t) & = & [a](t) \cap \frac{[y](t)}{[u](t)} \\ [a](t) & = & [a](t) \cap [x](t+1) \\ [x](t) & = & [x](t) \cap [a](t-1) \end{array} \right.$$

**Example.** Consider  $x(t) \in [x](t)$  with the constraint

$$\forall t, x(t) = x(t+1)$$

Contract the tube  $[x](t)$ .

We first decompose into primitive trajectory constraints

$$\begin{aligned}x(t) &= a(t+1) \\x(t) &= a(t).\end{aligned}$$

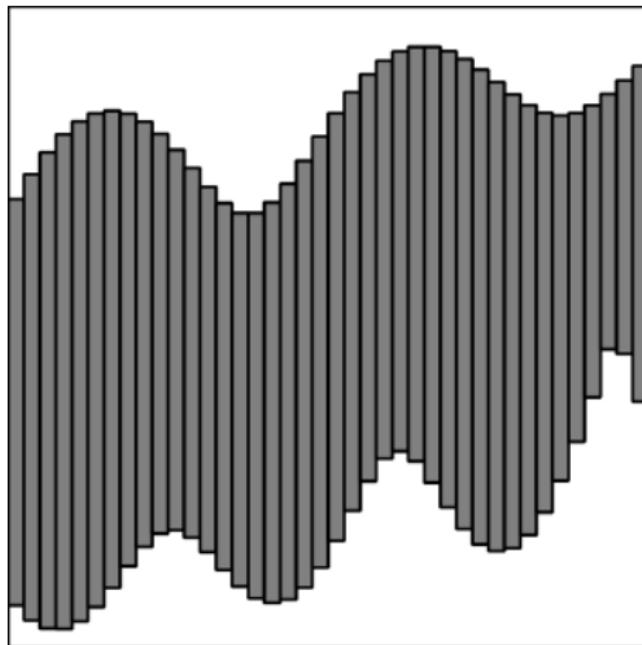
## Contractors

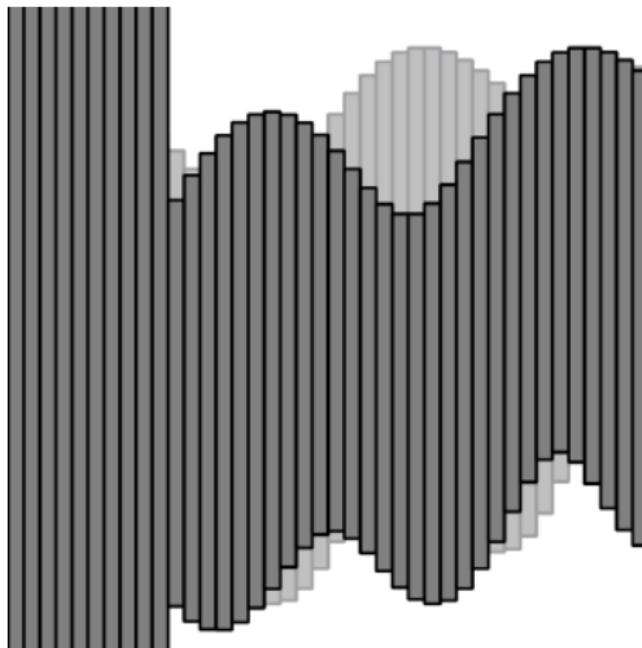
$$[x](t) : = [x](t) \cap [a](t+1)$$

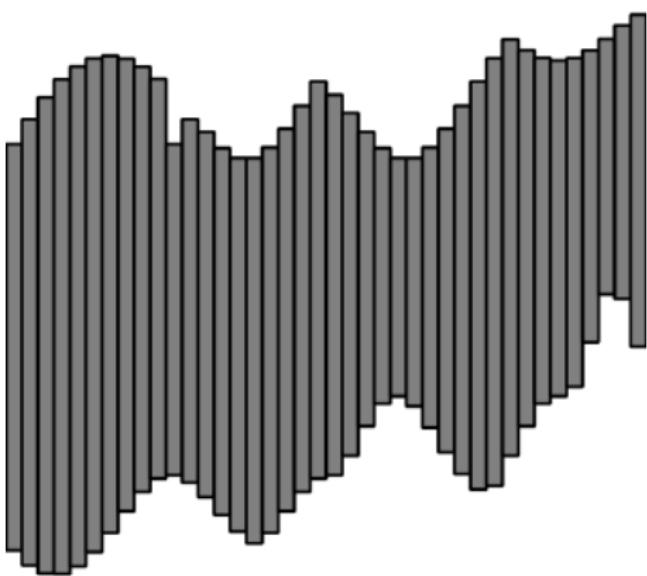
$$[a](t) : = [a](t) \cap [x](t-1)$$

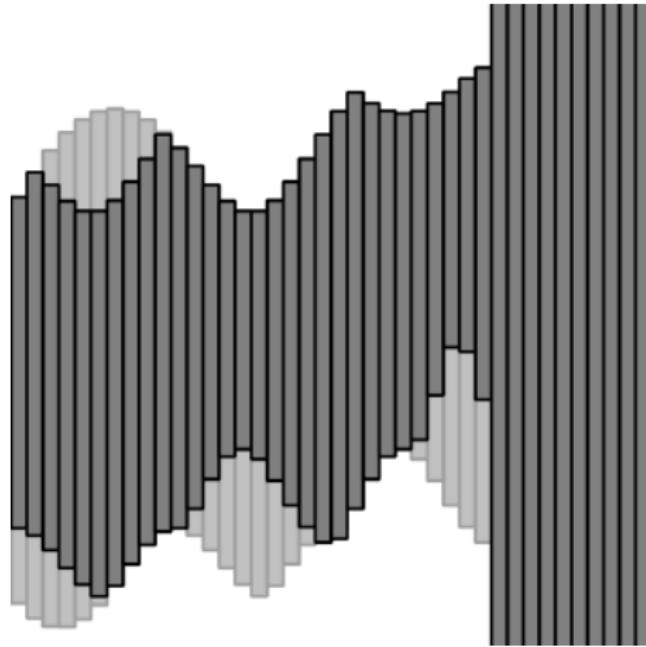
$$[x](t) : = [x](t) \cap [a](t)$$

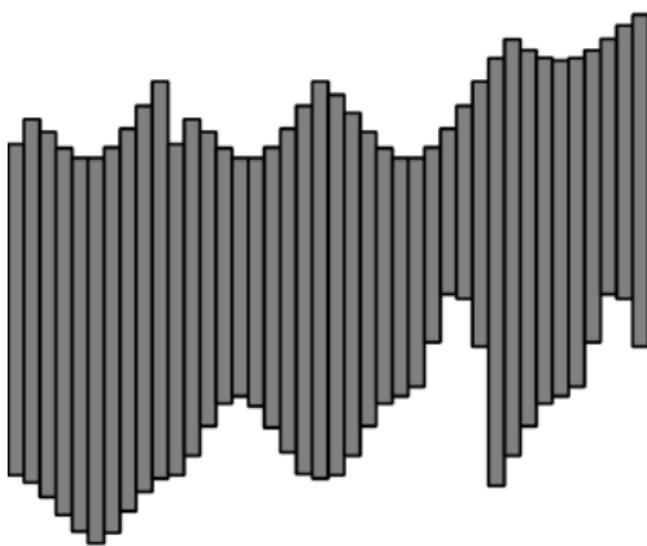
$$[a](t) : = [a](t) \cap [x](t)$$

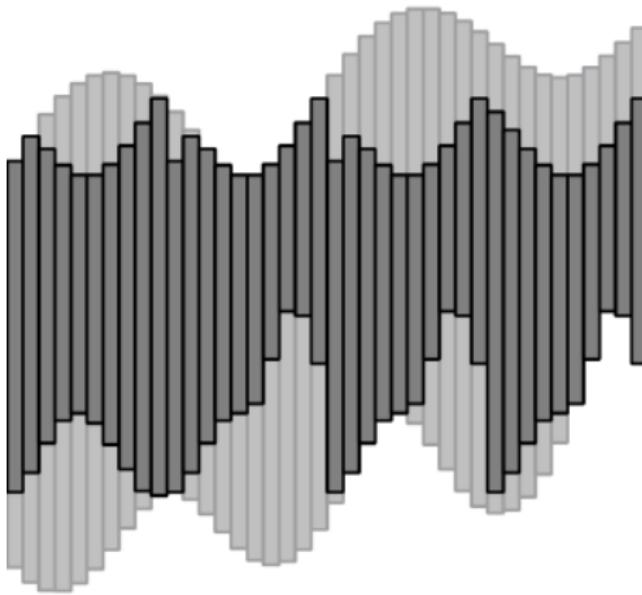


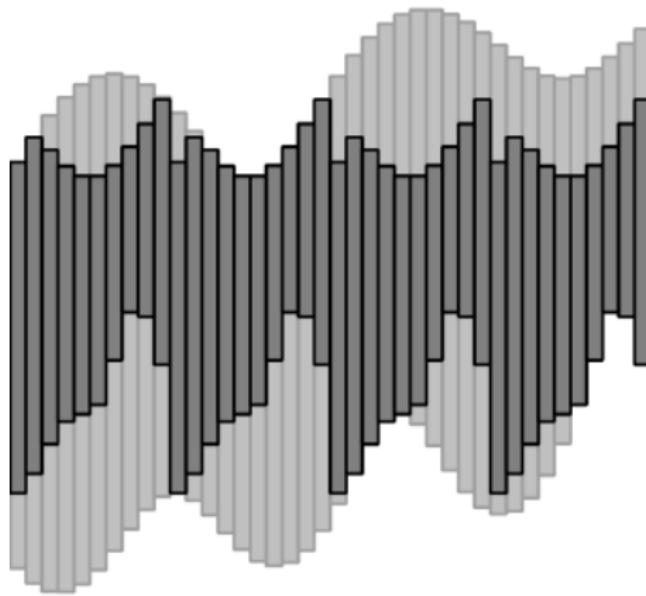


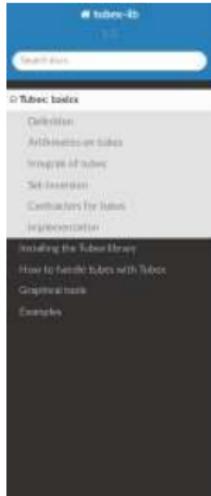












## Definition

A tube  $[x](\cdot)$  is defined as an envelope enclosing an uncertain trajectory  $x(\cdot) : \mathbb{R} \rightarrow \mathbb{R}^n$ . It is built as an interval of two functions  $[x^-(t), x^+(t)]$  such that  $\forall t, x^-(t) \leq x^+(t)$ . A trajectory  $x(\cdot)$  belongs to the tube  $[x](\cdot)$  if  $\forall t, x(t) \in [x](t)$ . Fig. 1 illustrates a tube implemented with a set of boxes. This sliced implementation is detailed hereinafter.

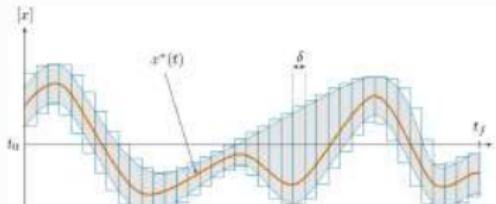


Fig. 1 A tube  $[x](\cdot)$  represented by a set of boxes. This representation can be used to enclose signals such as  $x^*(\cdot)$ .

## Code example

```
float tolerance = 0.3;
Interval domain[0,10];
Tube estimate, simulating, Function("x", "(x-5)^2 + [-0.3,0.3]");
```

<http://www.codac.io/>

# Space-time estimation

## Classical state estimation

$$\begin{cases} \dot{\mathbf{x}}(t) &= \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t)) & t \in \mathbb{R} \\ \mathbf{0} &= \mathbf{g}(\mathbf{x}(t), t) & t \in \mathbb{T} \subset \mathbb{R}. \end{cases}$$

Space constraint  $\mathbf{g}(\mathbf{x}(t), t) = 0$ .

## Example.

$$\begin{cases} \dot{x}_1 = x_3 \cos x_4 \\ \dot{x}_2 = x_3 \cos x_4 \\ \dot{x}_3 = u_1 \\ \dot{x}_4 = u_2 \\ (x_1(5) - 1)^2 + (x_2(5) - 2)^2 - 4 = 0 \\ (x_1(7) - 1)^2 + (x_2(7) - 2)^2 - 9 = 0 \end{cases}$$

With time-space constraints

$$\begin{cases} \dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t), \mathbf{u}(t)) & t \in \mathbb{R} \\ \mathbf{0} = \mathbf{g}(\mathbf{x}(t), \mathbf{x}(t'), t, t') & (t, t') \in \mathbb{T} \subset \mathbb{R} \times \mathbb{R}. \end{cases}$$

**Example.** An ultrasonic underwater robot with state

$$\mathbf{x} = (x_1, x_2, \dots) = (x, y, \theta, v, \dots)$$

At time  $t$  the robot emits an onmidirectional sound. At time  $t'$  it receives it

$$(x_1(t) - x_1(t'))^2 + (x_2(t) - x_2(t'))^2 - c(t - t')^2 = 0.$$

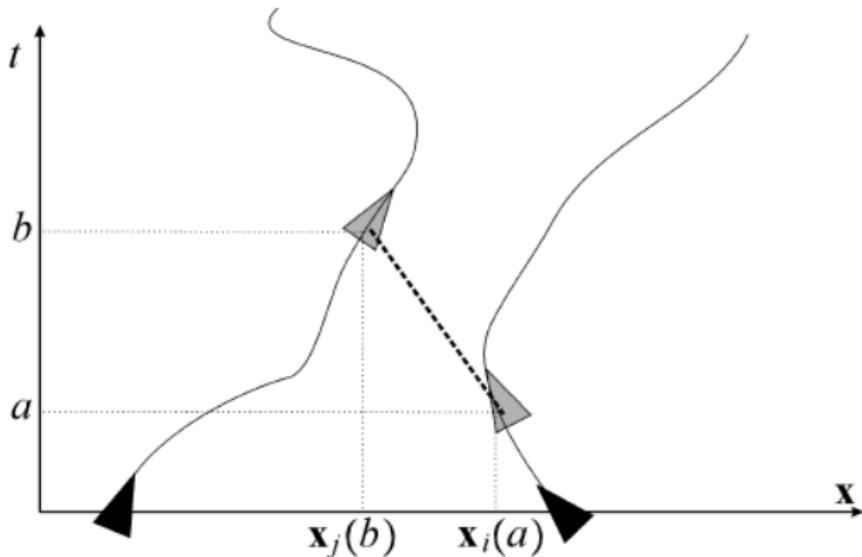
# Swarm localization

Consider  $n$  robots  $\mathcal{R}_1, \dots, \mathcal{R}_n$  described by

$$\dot{\mathbf{x}}_i = \mathbf{f}(\mathbf{x}_i, \mathbf{u}_i), \mathbf{u}_i \in [\mathbf{u}_i].$$

Omnidirectional sounds are emitted and received.

A *ping* is a 4-uple  $(a, b, i, j)$  where  $a$  is the emission time,  $b$  is the reception time,  $i$  is the emitting robot and  $j$  the receiver.



With the time space constraint

$$\dot{\mathbf{x}}_i = \mathbf{f}(\mathbf{x}_i, \mathbf{u}_i), \mathbf{u}_i \in [\mathbf{u}_i].$$

$$g(\mathbf{x}_{i(k)}(a(k)), \mathbf{x}_{j(k)}(b(k)), a(k), b(k)) = 0$$

where

$$g(\mathbf{x}_i, \mathbf{x}_j, a, b) = \|x_1 - x_2\| - c(b - a).$$

Clocks are uncertain. We only have measurements  $\tilde{a}(k), \tilde{b}(k)$  of  $a(k), b(k)$  thanks to clocks  $h_i$ . Thus

$$\begin{aligned}\dot{\mathbf{x}}_i &= \mathbf{f}(\mathbf{x}_i, \mathbf{u}_i), \mathbf{u}_i \in [\mathbf{u}_i]. \\ g\left(\mathbf{x}_{i(k)}(a(k)), \mathbf{x}_{j(k)}(b(k)), a(k), b(k)\right) &= 0 \\ \tilde{a}(k) &= h_{i(k)}(a(k)) \\ \tilde{b}(k) &= h_{j(k)}(b(k))\end{aligned}$$

The drift of the clocks is bounded

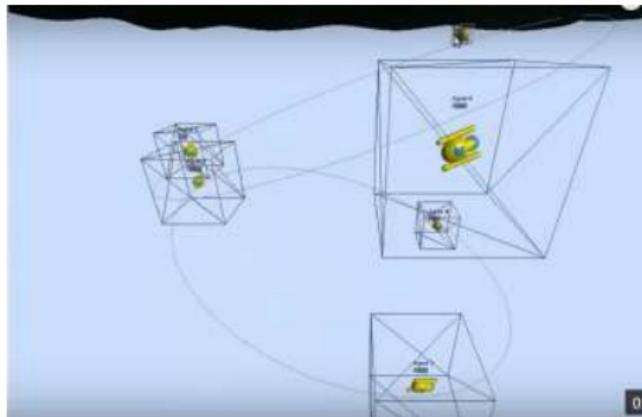
$$\dot{\mathbf{x}}_i = \mathbf{f}(\mathbf{x}_i, \mathbf{u}_i), \mathbf{u}_i \in [\mathbf{u}_i].$$

$$g(\mathbf{x}_{i(k)}(a(k)), \mathbf{x}_{j(k)}(b(k)), a(k), b(k)) = 0$$

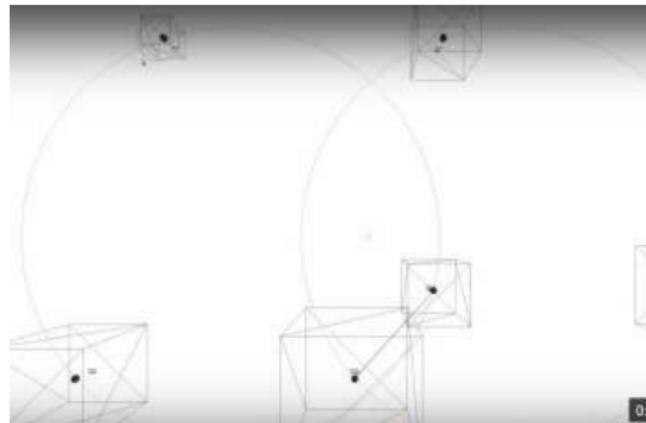
$$\tilde{a}(k) = h_{i(k)}(a(k))$$

$$\tilde{b}(k) = h_{j(k)}(b(k))$$

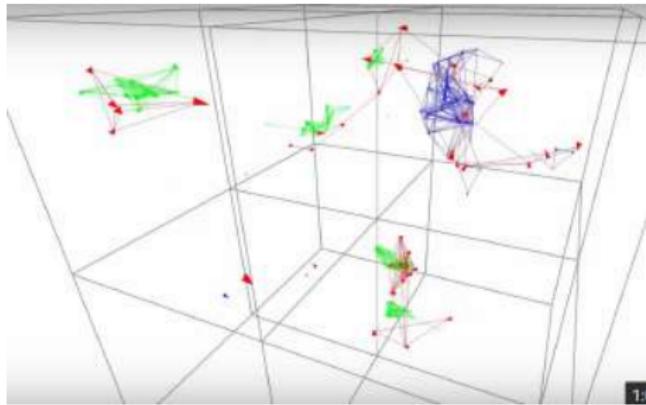
$$\dot{h}_i = 1 + n_h, n_h \in [n_h]$$



<https://youtu.be/j-ERcoXF1Ks>



<https://youtu.be/jr8xKle0Nds>



<https://youtu.be/GycJxGFvYE8>

# References

- ① Interval analysis [8, 5, 6]
- ② Localization with intervals : [7]
- ③ SLAM with intervals : [4]
- ④ Interval tubes [9], [3], [1][10]
- ⑤ Swarm localization [2]

-  F. L. Bars, J. Sliwka, O. Reynet, and L. Jaulin.  
State estimation with fleeting data.  
*Automatica*, 48(2):381–387, 2012.
-  A. Bethencourt and L. Jaulin.  
Cooperative localization of underwater robots with unsynchronized clocks.  
*Journal of Behavioral Robotics*, 4(4):233–244, 2013.
-  A. Bethencourt and L. Jaulin.  
Solving non-linear constraint satisfaction problems involving time-dependant functions.  
*Mathematics in Computer Science*, 8(3), 2014.
-  L. Jaulin.  
Range-only SLAM with occupancy maps; A set-membership approach.  
*IEEE Transaction on Robotics*, 27(5):1004–1010, 2011.

-  L. Jaulin, M. Kieffer, O. Didrit, and E. Walter.  
*Applied Interval Analysis, with Examples in Parameter and State Estimation, Robust Control and Robotics.*  
Springer-Verlag, London, 2001.
-  L. Jaulin, O. Reynet, B. Desrochers, S. Rohou, and J. Ninin.  
*IaMOOC, Interval analysis with applications to parameter estimation and robot localization ,*  
[www.ensta-bretagne.fr/iamooc/](http://www.ensta-bretagne.fr/iamooc/).  
ENSTA-Bretagne, 2019.
-  M. Kieffer, L. Jaulin, E. Walter, and D. Meizel.  
Robust autonomous robot localization using interval analysis.  
*Reliable Computing*, 6(3):337–362, 2000.
-  R. Moore.  
*Methods and Applications of Interval Analysis.*  
Society for Industrial and Applied Mathematics, jan 1979.

-  S. Rohou, L. Jaulin, M. Mihaylova, F. L. Bars, and S. Veres.  
Guaranteed Computation of Robots Trajectories.  
*Robotics and Autonomous Systems*, 93:76–84, 2017.
-  J. A. D. Sandretto and A. Chapoutot.  
Dynibex: a differential constraint library for studying dynamical systems.  
In *Conference on Hybrid Systems: Computation and Control*,  
Vienne, Austria, 2016.