

Interval analysis for the validation of autonomous vehicles properties

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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

$$f(\mathbf{x}) = x_1x_2 - (x_1 + x_2)\cos x_2 + \sin x_1 \cdot \sin x_2 + 2$$

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 $\mathbb{A} \subset \mathbb{R}$.

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

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

Inclusion functions

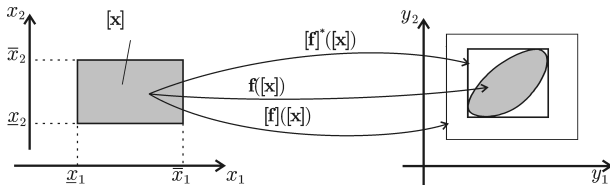
A *box*, or *interval vector* $[\mathbf{x}]$ of $\mathbb{R}^n$ is

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

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

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

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



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

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 $[f](in : [x] = ([x_1], [x_2], [x_3]), out : [y] = ([y_1], [y_2]))$

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

Is $[f]$ convergent? thin? monotonic?

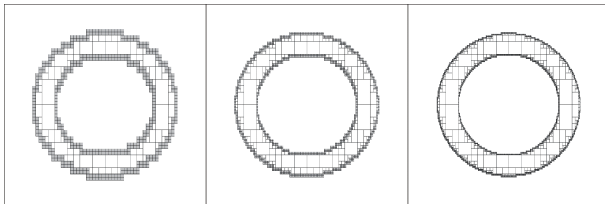
Set inversion

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

$$\mathbb{X}^- \subset \mathbb{X} \subset \mathbb{X}^+.$$

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 Y be a subset of $\mathbb{R}^m$. Set inversion is the characterization of

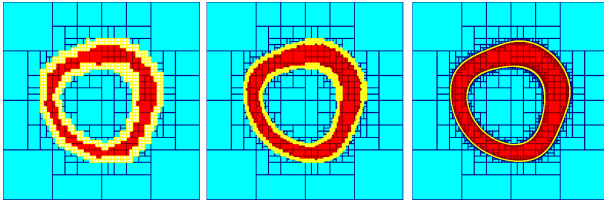
$$X = \{\mathbf{x} \in \mathbb{R}^n \mid \mathbf{f}(\mathbf{x}) \in Y\} = \mathbf{f}^{-1}(Y).$$

We shall use the following tests.

$$\begin{aligned} \text{(i)} \quad & [\mathbf{f}]([\mathbf{x}]) \subset \mathbb{Y} \quad \Rightarrow \quad [\mathbf{x}] \subset \mathbb{X} \\ \text{(ii)} \quad & [\mathbf{f}]([\mathbf{x}]) \cap \mathbb{Y} = \emptyset \quad \Rightarrow \quad [\mathbf{x}] \cap \mathbb{X} = \emptyset. \end{aligned}$$

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

Contractors

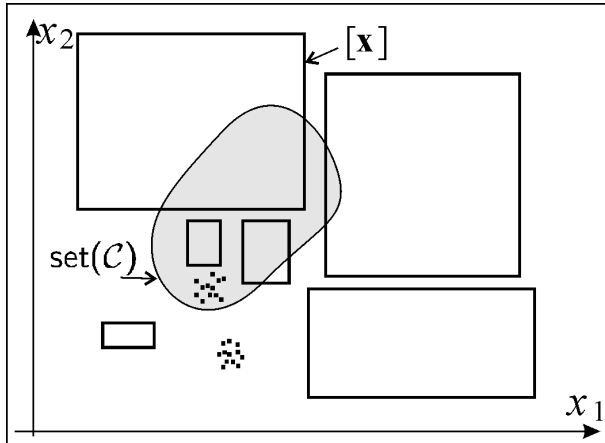


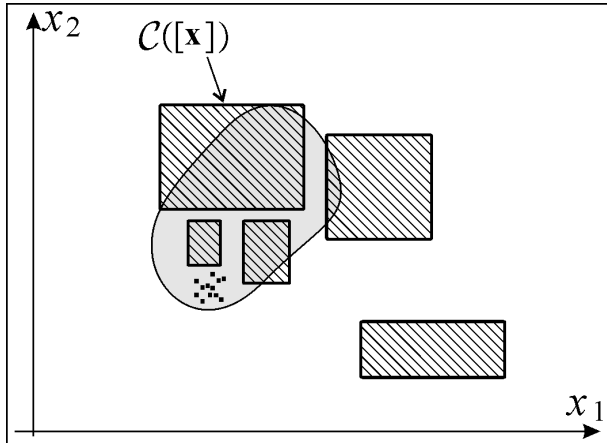
To characterize $\mathbb{X} \subset \mathbb{R}^n$, bisection algorithms bisect all boxes in all directions and become inefficient. Interval methods can still be useful if

- 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

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





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

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

Exercise. 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. We have

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

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

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

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: $[z], [x], [y]$)

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

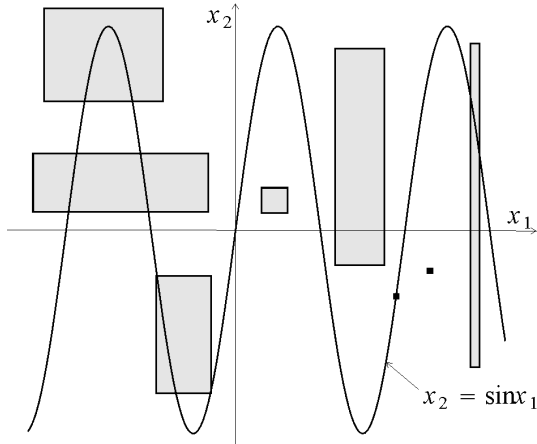
The contractor associated with $y = \exp x$ is:

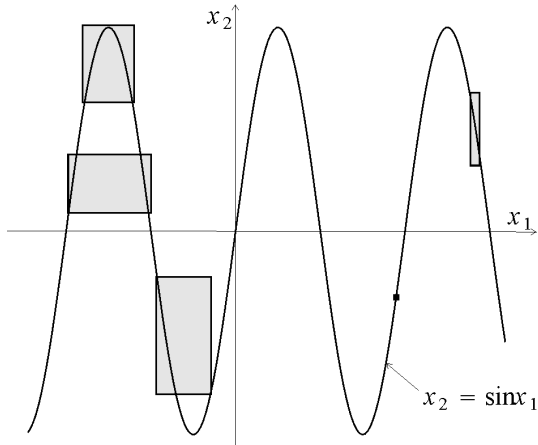
Algorithm $\text{pexp}(\text{inout: } [y], [x])$
1 $[y] := [y] \cap \exp([x])$
2 $[x] := [x] \cap \log([y])$

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

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

Decomposition

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

can be decomposed into

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

Forward Backward contractor

For the equation

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

we decompose into

$$a = x_1 + x_2$$

$$b = a \cdot x_3$$

$$b \in [1, 2]$$

For the equation

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

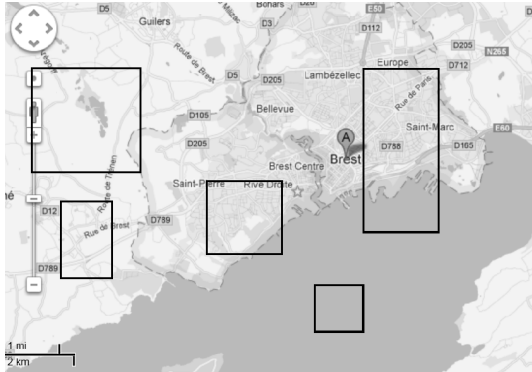
we have the following contractor:

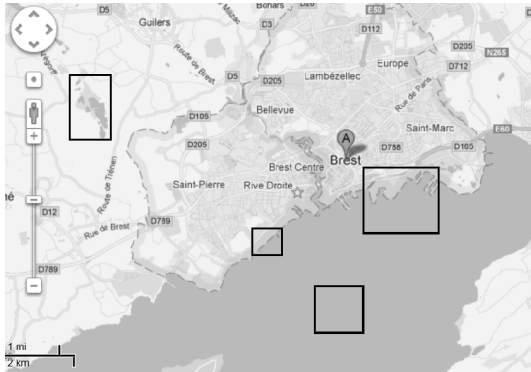
Algorithm $\mathcal{C}$ (inout $[x_1], [x_2], [x_3]$)

$[a] = [x_1] + [x_2]$	// $a = x_1 + x_2$
$[b] = [a] \cdot [x_3]$	// $b = a \cdot x_3$
$[b] = [b] \cap [1, 2]$	// $b \in [1, 2]$
$[x_3] = [x_3] \cap \frac{[b]}{[a]}$	// $x_3 = \frac{b}{a}$
$[a] = [a] \cap \frac{[b]}{[x_3]}$	// $a = \frac{b}{x_3}$
$[x_1] = [x_1] \cap [a] - [x_2]$	// $x_1 = a - x_2$
$[x_2] = [x_2] \cap [a] - [x_1]$	// $x_2 = a - x_1$

Contractor on images

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





Solving equations

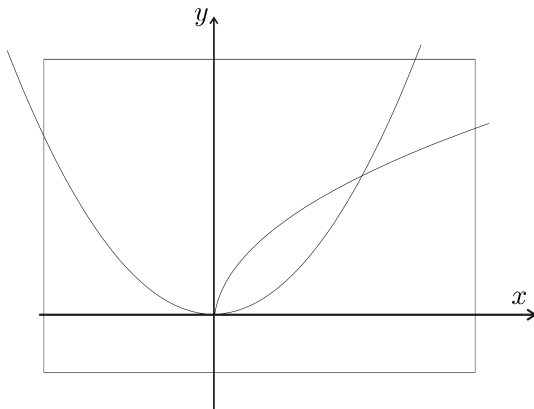
Consider the system of two equations.

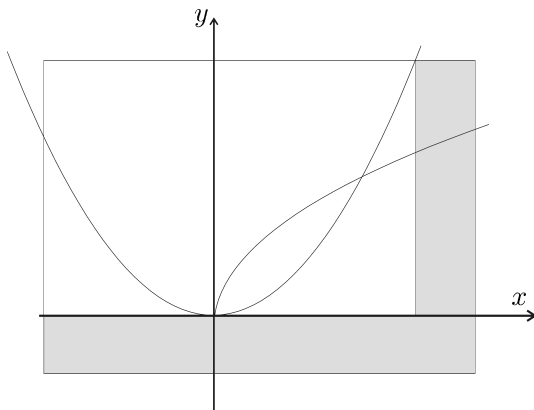
$$\begin{aligned}y &= x^2 \\ y &= \sqrt{x}.\end{aligned}$$

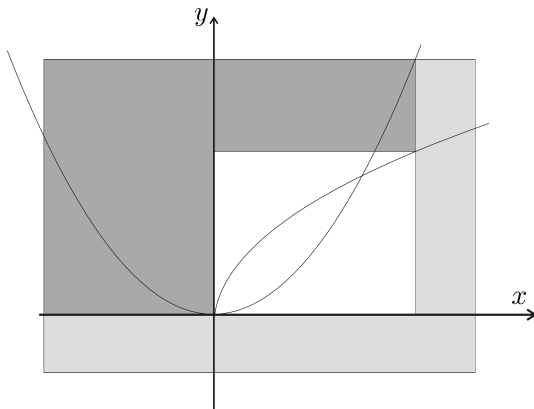
We can build two contractors

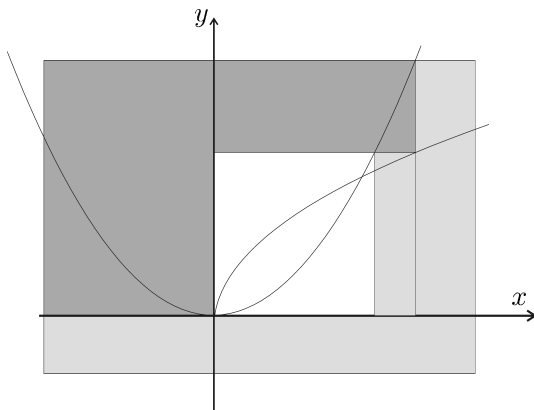
$$\mathcal{C}_1 : \begin{cases} [y] = [y] \cap [x]^2 \\ [x] = [x] \cap \sqrt{[y]} \end{cases} \quad \text{associated to } y = x^2$$

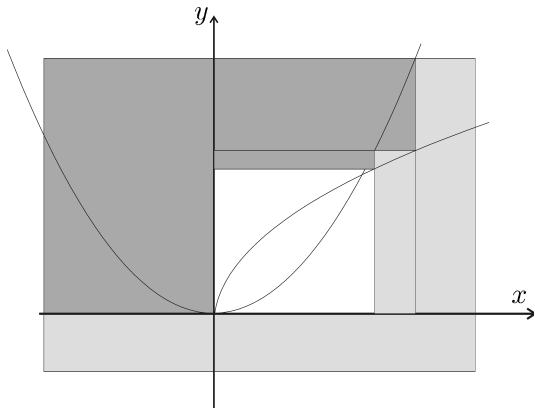
$$\mathcal{C}_2 : \begin{cases} [y] = [y] \cap \sqrt{[x]} \\ [x] = [x] \cap [y]^2 \end{cases} \quad \text{associated to } y = \sqrt{x}$$

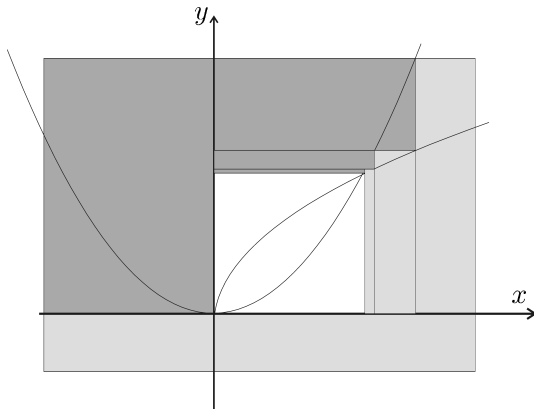


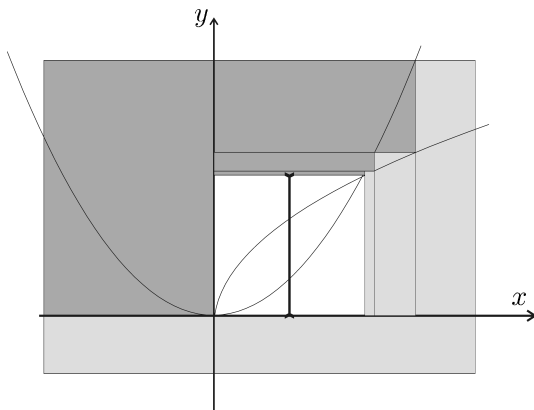


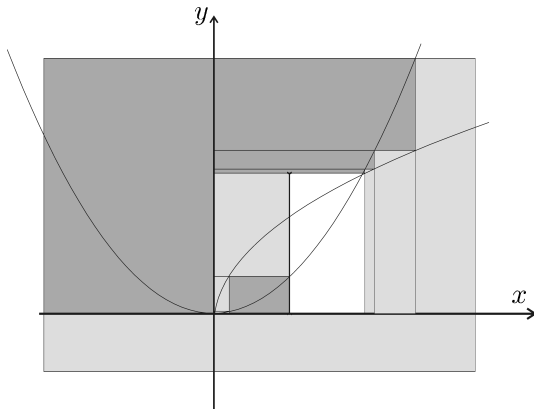


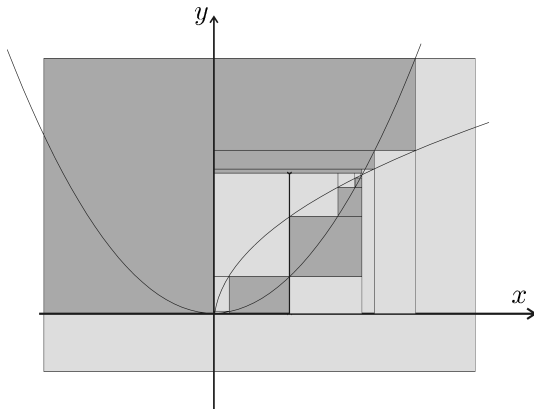






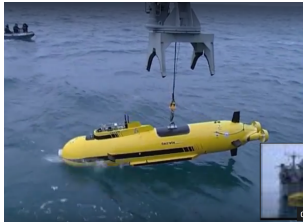






SLAM

$$\begin{cases} \dot{\mathbf{x}} &= \mathbf{f}(\mathbf{x}, \mathbf{u}) & \text{(evolution equation)} \\ \mathbf{y} &= \mathbf{g}(\mathbf{x}, \mathbf{u}) & \text{(observation equation)} \\ \mathbf{z}_i &= \mathbf{h}(\mathbf{x}, \mathbf{u}, \mathbf{m}_i) & \text{(mark equation)} \end{cases}$$



Redermor (GESMA, Brest)

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

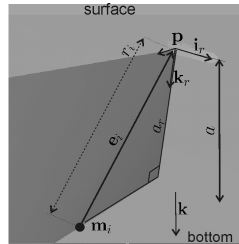
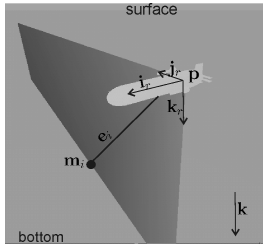


GPS (Global positioning system), only at the surface.

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

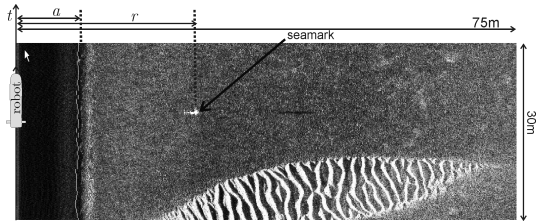
$$t_f = 6000 \text{ s}, \quad \ell^f = (-4.4546607^\circ, 48.2191297^\circ) \pm 2.5m$$

Sonar (KLEIN 5400 side scan sonar).





Screenshot of SonarPro



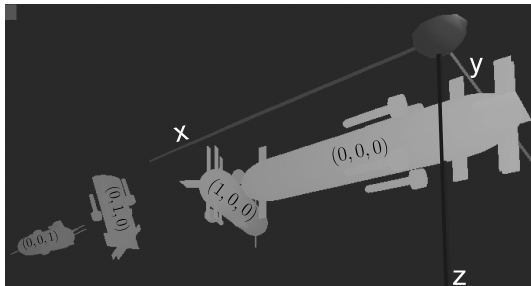
Mine detection with SonarPro

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 landmarks 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

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

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

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

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

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

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

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

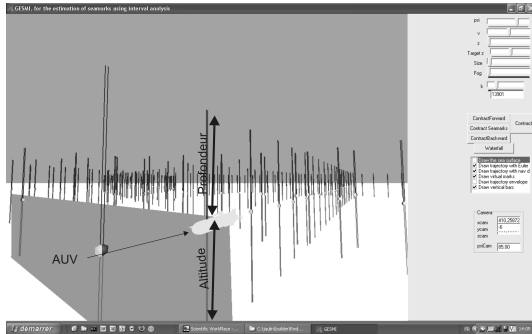
$$\mathbf{R}(t) = \mathbf{R}_\psi(t) \mathbf{R}_\theta(t) \mathbf{R}_\phi(t)$$

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

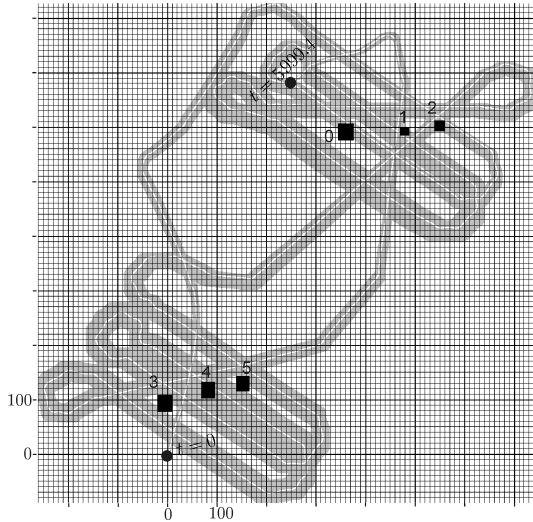
$$\|\mathbf{m}(\sigma(i)) - \mathbf{p}(\tau(i))\| = r(i)$$

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

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



youtu.be/lzJtAfAT7h4



Definition

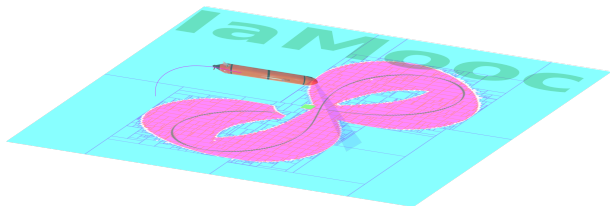
A tube $[X](\cdot)$ is defined as an enclosing an uncertain trajectory $\mathbf{x}(\cdot) : \mathbb{R} \rightarrow \mathbb{R}^n$. It is built as an interval of two functions $[X^-(\cdot), X^+(\cdot)]$ such that $\forall t, \mathbf{x}^*(t) \in X^*(t)$. A trajectory $\mathbf{x}(\cdot)$ belongs to the tube $[X](\cdot)$ if $\forall t, \mathbf{x}(t) \in [X](t)$. 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 slices. This representation can be used to enclose signals such as $x^*(\cdot)$.

Code example:

```
float timestep = 0.1;
Interval domain(0,10);
Tube x(domain, timestep, Function("t", "(t-5)*2 + [-0.5,0.5]*"));
```

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



<https://www.ensta-bretagne.fr/iamooc/>

References

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



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