

Meanwhile ...

I don't like it, Thomson ...
... If we don't get
somewhere soon ...



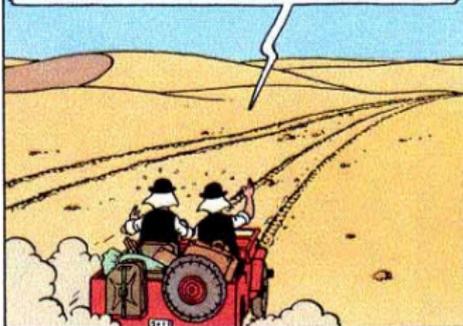
It's all right! ... Look! ... There!
... Tracks of a car!

Quite correct! And they
aren't a mirage, either!



An hour later ...

Hooray! ... More tracks! ... A
second car joined the first one ...



A real stroke of luck
hitting this road.

To be precise:
we've really had
a stroke!



Verifying loops in robot trajectories under uncertainties

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SWIM

26th July 2019, Palaiseau



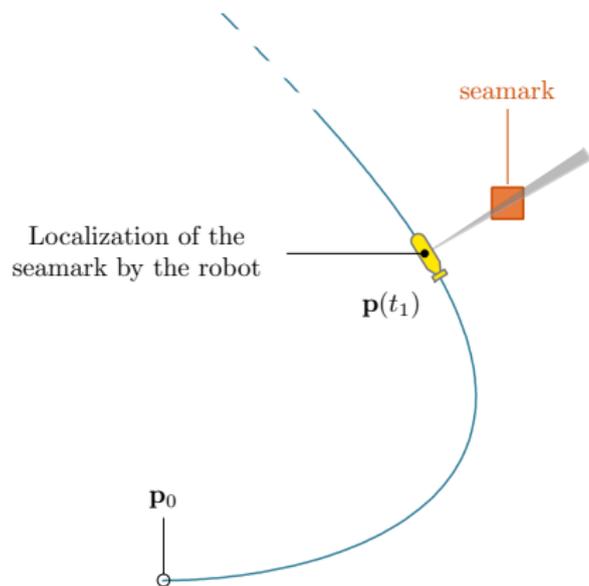
Section 2

Motivations

Motivations

Simultaneous Localization and Mapping

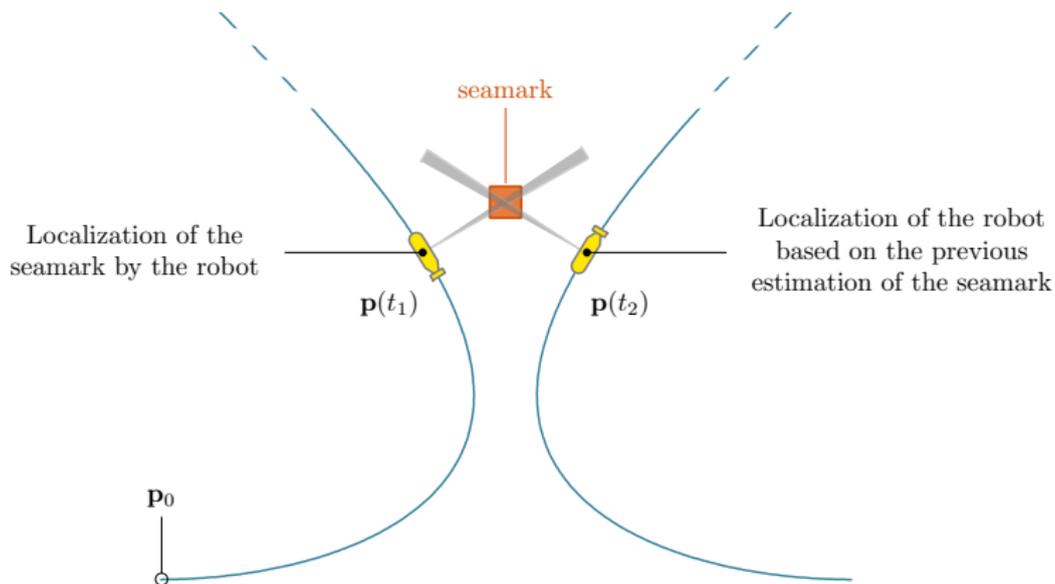
- ▶ **come back** to a previous pose and **recognize** the environment
- ▶ problem: loop closure detection



Motivations

Simultaneous Localization and Mapping

- ▶ **come back** to a previous pose and **recognize** the environment
- ▶ problem: loop closure detection



Motivations

Simultaneous Localization and Mapping

The problem of **false loop detections** in similar environments.

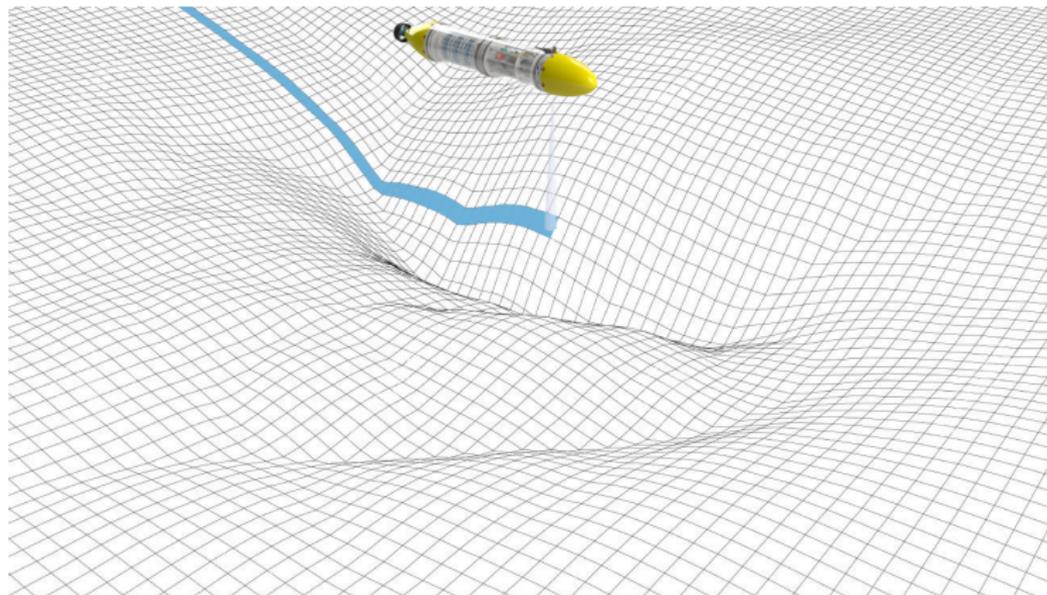


In Versailles' gardens: similar places. *Did we really come back to a previous place?*

Motivations

Underwater robot localization

An underwater robot performing a loop during an exploration:

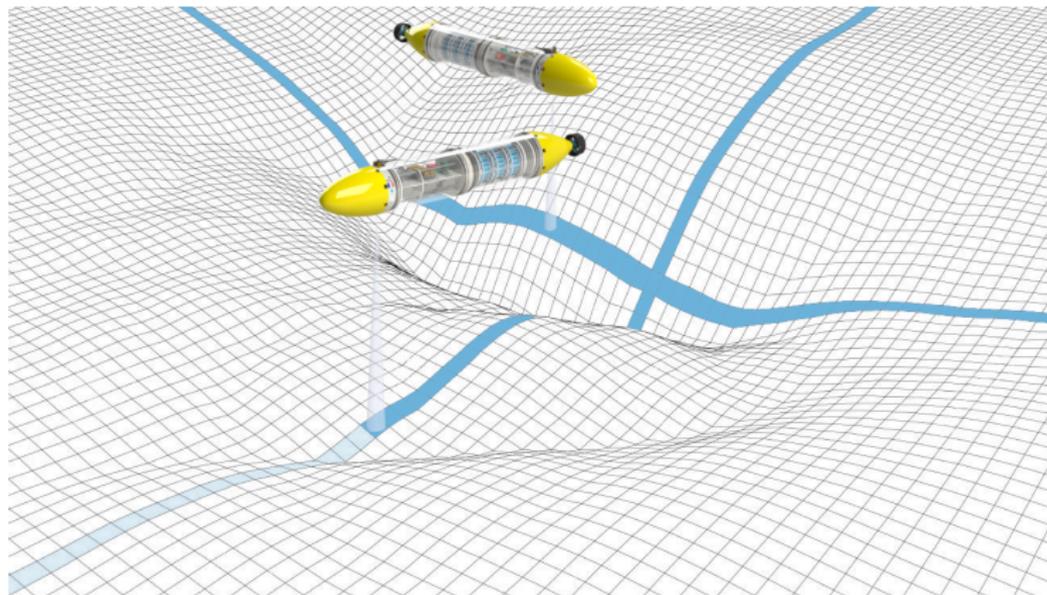


Robot's trajectory is projected in blue on the seabed.

Motivations

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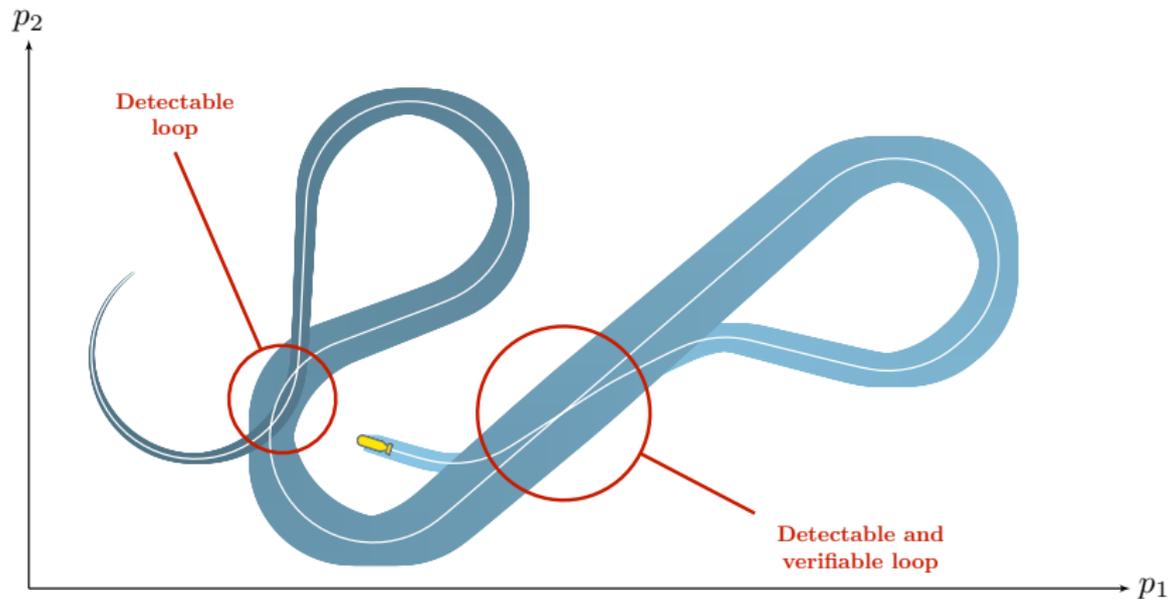


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Motivations

Uncertainties: detection vs verification

Uncertain trajectories enclosed by **tubes**.



Only one loop can be verified – at least two feasible loops are detected

Section 3

Looped trajectories

Looped trajectories

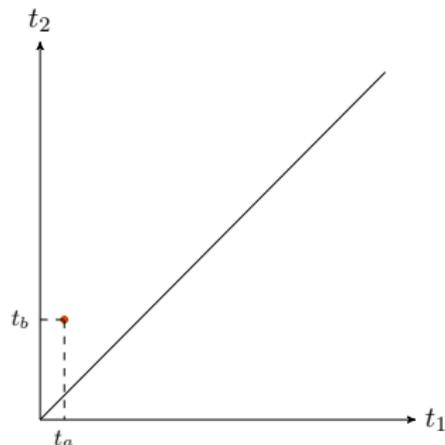
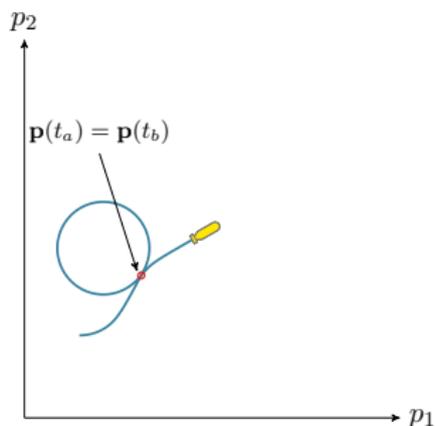
Definitions (Aubry, 2013)

- ▶ robot position: $\mathbf{p} = (x, y)^T \in \mathbb{R}^2$
- ▶ 2D robot trajectory: $\mathbf{p}(t) : \mathbb{R} \rightarrow \mathbb{R}^2, t \in [t_0, t_f]$
- ▶ looped trajectory \Leftrightarrow trajectory that crosses itself
 - ▶ $\mathbf{p}(t_1) = \mathbf{p}(t_2), t_1 \neq t_2$
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Looped trajectories

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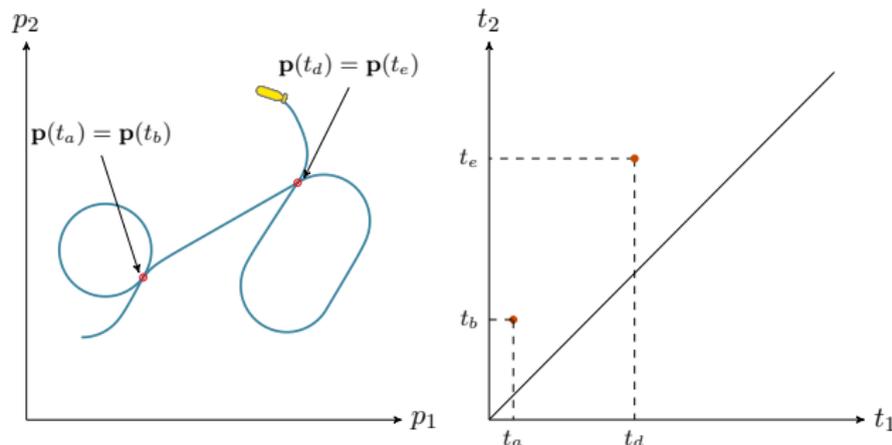
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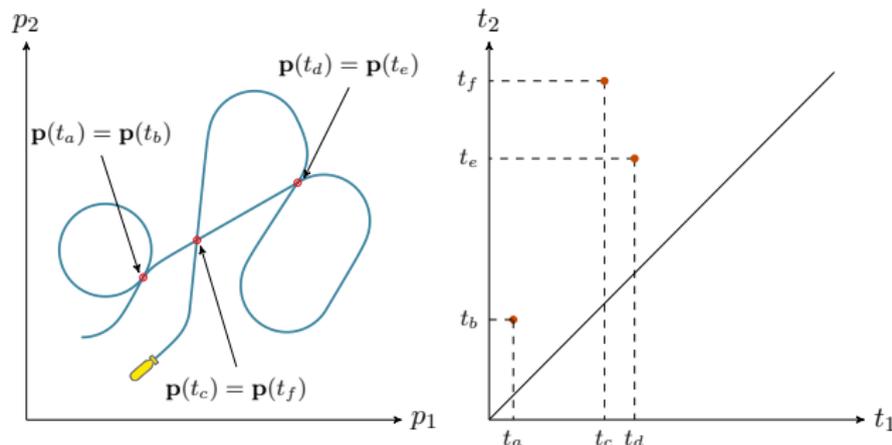
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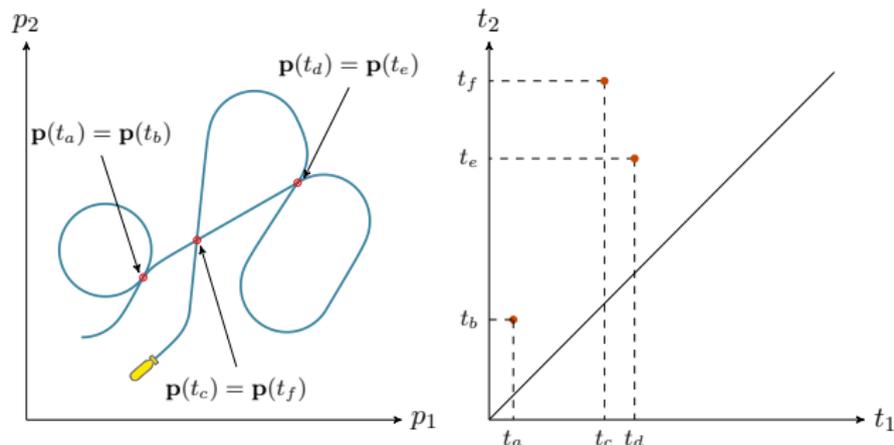
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Looped trajectories

Definitions (Aubry, 2013)

- ▶ t -plane \Leftrightarrow all feasible t -pairs = $[t_0, t_f]^2$
- ▶ loop set \mathbb{T}^* :
 - ▶ $\mathbb{T}^* = \{(t_1, t_2) \in [t_0, t_f]^2 \mid \mathbf{p}(t_1) = \mathbf{p}(t_2), t_1 < t_2\}$
- ▶ loop set of below example:
 - ▶ $\mathbb{T}^* = \{(t_a, t_b), (t_c, t_f), (t_d, t_e)\}$



Looped trajectories

Computing loops from robot sensors

Context: robot trajectory $\mathbf{p}(t)$ cannot be directly sensed.

Computation from speed measurements:

$$\mathbf{p}(t) = \int_{t_0}^t \mathbf{v}(\tau) d\tau + \mathbf{p}_0, \quad (1)$$

with $\mathbf{v}(t) \in \mathbb{R}^2$: robot velocity vector at time $t \in [t_0, t_f]$.

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Loop-set from velocity:

$$\mathbb{T}^* = \{(t_1, t_2) \in [t_0, t_f]^2 \mid \mathbf{p}(t_1) = \mathbf{p}(t_2), t_1 < t_2\} \quad (2)$$

$$= \left\{ (t_1, t_2) \in [t_0, t_f]^2 \mid \int_{t_1}^{t_2} \mathbf{v}(\tau) d\tau = \mathbf{0}, t_1 < t_2 \right\} \quad (3)$$

Section 4

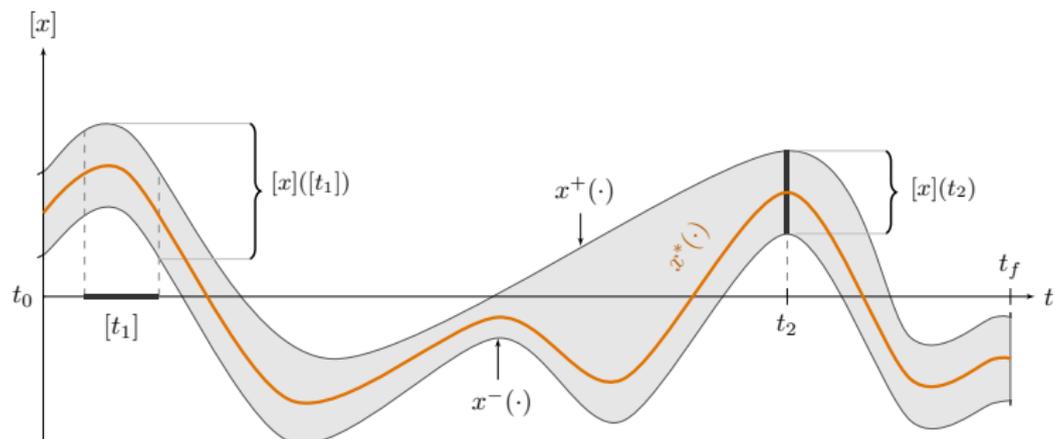
Loop detection

Loop detection

Bounded-error context

Tubes: sets of trajectories

$[x](\cdot)$, interval of trajectories $[x^-(\cdot), x^+(\cdot)]$
 such that $\forall t \in \mathbb{R}, x^-(t) \leq x^+(t)$



Tube $[x](\cdot)$ enclosing an uncertain trajectory $x^*(\cdot)$

Loop detection

Bounded-error context

Actual loop-set \mathbb{T}^* (error free):

$$\mathbb{T}^* = \left\{ (t_1, t_2) \mid \int_{t_1}^{t_2} \mathbf{v}^*(\tau) d\tau = \mathbf{0} \right\} \quad (4)$$

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Bounded-error context, assuming $\mathbf{v}^*(\cdot) \in [\mathbf{v}](\cdot)$:

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Set-membership approach:

$$\mathbb{T}^* \subset \mathbb{T} \subset [t_0, t_f]^2 \quad (6)$$

Loop detection

Inclusion function

Simplification:defining the actual but unknown function $\mathbf{f}^* : \mathbb{R}^2 \rightarrow \mathbb{R}^2$

$$\mathbf{f}^*(t_1, t_2) = \int_{t_1}^{t_2} \mathbf{v}^*(\tau) d\tau \quad (7)$$

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Assessed knowledge: $[\mathbf{f}] : \mathbb{R}^2 \rightarrow \mathbb{I}\mathbb{R}^2$ is an *interval function* of \mathbf{f}^* :

$$\mathbf{f}^*(t_1, t_2) \in [\mathbf{f}](t_1, t_2) = \int_{t_1}^{t_2} [\mathbf{v}](\tau) d\tau \quad (8)$$

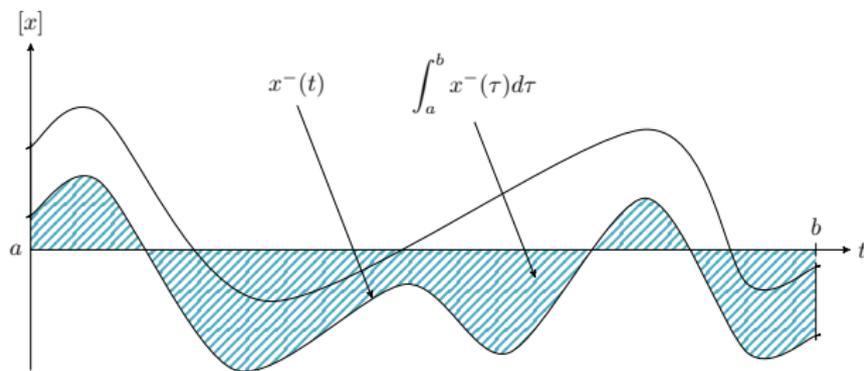
Loop detection

Integral of tubes

Definition: the integral of a tube $[x](\cdot) = [x^-, x^+]$ is an interval:

$$\int_a^b [x](\tau) d\tau = \left\{ \int_a^b x(\tau) d\tau \mid x(\cdot) \in [x](\cdot) \right\} = \left[\int_a^b x^-(\tau) d\tau, \int_a^b x^+(\tau) d\tau \right]$$

[Aubry2013]



blue area: lower bound of the tube's integral

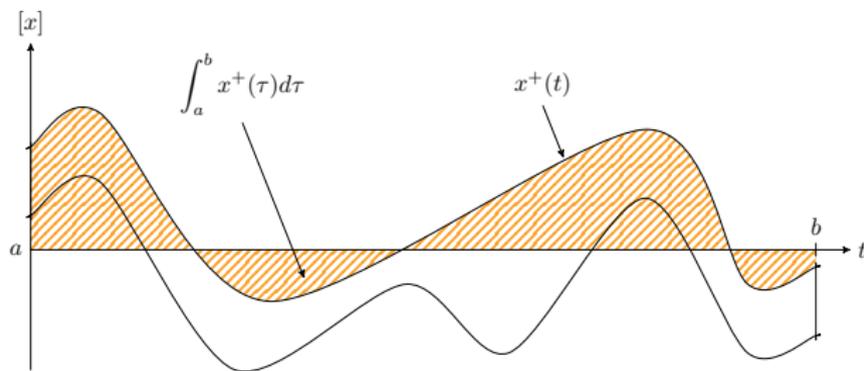
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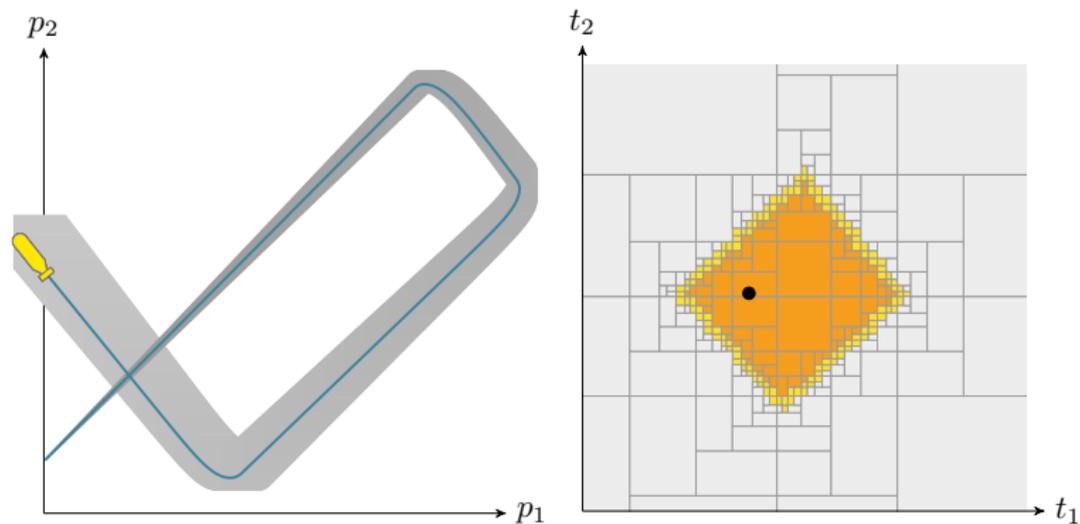
[Aubry2013]



orange area: upper bound of the tube's integral

Loop detection

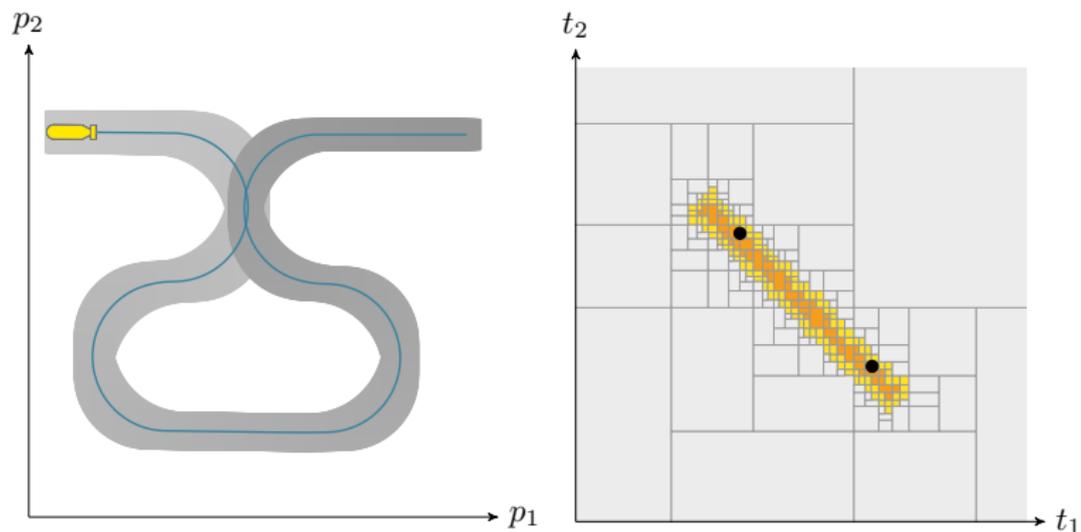
Reliable approximation of a loop set



Undeniable looped trajectory

Loop detection

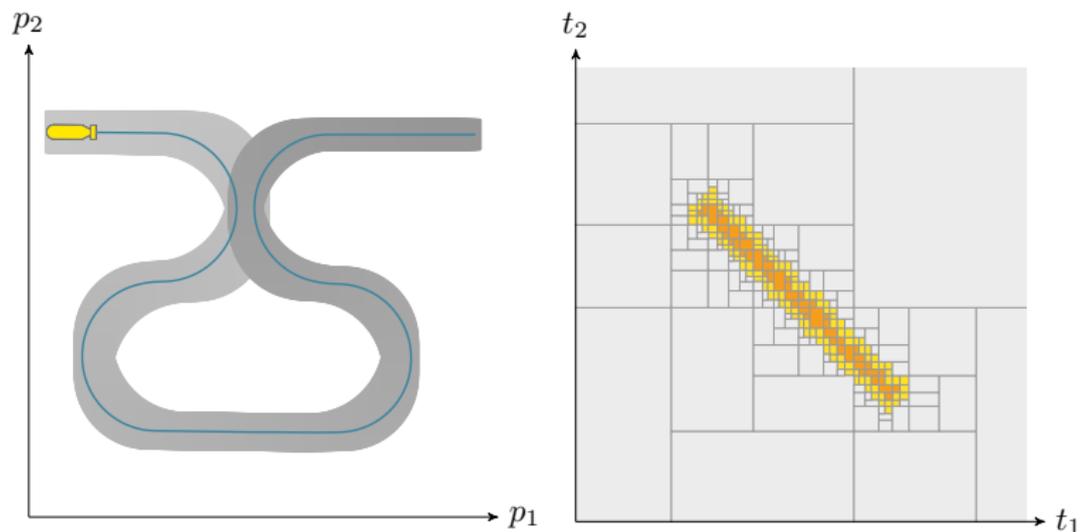
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Doubtful looped trajectory

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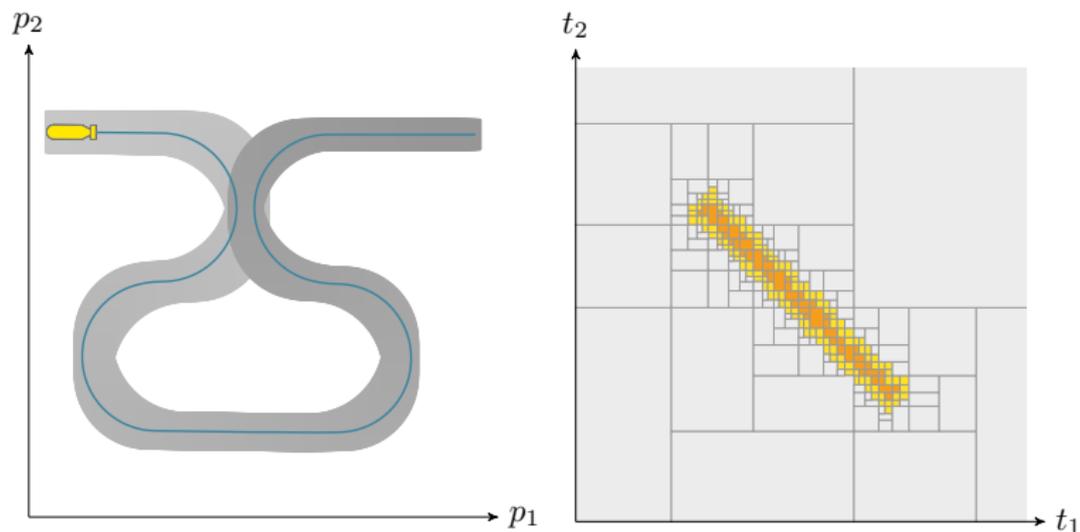
Reliable approximation of a loop set



Doubtful looped trajectory

Loop detection

Reliable approximation of a loop set



Doubtful looped trajectory

$$\forall \mathbf{f} \in [\mathbf{f}], \exists \mathbf{t} \in \mathbb{T}_i \mid \mathbf{f}(\mathbf{t}) = \mathbf{0} \implies \underbrace{\exists \mathbf{t} \in \mathbb{T}_i \mid \mathbf{f}^*(\mathbf{t}) = \mathbf{0}}_{\text{loop existence proof}} \quad (9)$$

Section 5

Topological degree for zero verification

Topological degree for zero verification

Problem statement

Usually we cannot find exactly a zero of an uncertain function, but we can prove it exists within some domain Ω .

Assumptions:

-
- ▶ known inclusion function $[\mathbf{f}] : \mathbb{IR}^n \rightarrow \mathbb{IR}^m$ of the unknown function $\mathbf{f}^* : \mathbb{R}^n \rightarrow \mathbb{R}^m$
 - ▶ possibly in the form of an algorithm for computing $[\mathbf{f}](\mathbf{t})$
 - ▶ $n = m = 2$
 - ▶ Ω possibly made of a union of finitely many boxes in \mathbb{IR}^2

→ need to isolate and verify zeros of \mathbf{f}^*

Topological degree for zero verification

Powerful topological degree

Topological degree $\deg(\mathbf{f}^*, \Omega)$:

- ▶ unique integer assigned to \mathbf{f}^* and a compact set $\Omega \subset \mathbb{R}^n$ such that $\mathbf{f}^*(\mathbf{t}) \neq \mathbf{0}$ for all $\mathbf{t} \in \partial\Omega$

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Most important property of it:

$$\deg(\mathbf{f}^*, \Omega) \neq 0 \quad \implies \quad \exists \mathbf{t} \in \Omega \mid \mathbf{f}^*(\mathbf{t}) = \mathbf{0} \quad (10)$$

■ Topological degree theory and applications

Y. J. Cho, Y. Q. Chen *Mathematical Analysis and Applications*, 2006

■ Degree theory in analysis and applications

I. Fonseca, W. Gangbo *Oxford lecture series*, 1995

■ A set of axioms for the degree of a tangent vector field on differentiable manifolds

M. Furi, M. P. Pera, M. Spadini *Fixed Point Theory and Applications*, 2010

Topological degree for zero verification

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Assets of topological degree:

- ▶ can be computed in case where only an inclusion function $[f]$ of f^* is given
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- ▶ useful to count the number of 0

Topological degree for zero verification

Powerful topological degree

Our application for loop detection:

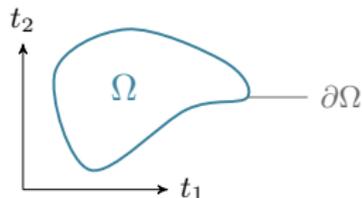
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Topological degree for zero verification

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 - ▶ *winding number* of the curve $\partial\Omega \xrightarrow{\mathbf{f}^*} \mathbb{R}^2 \setminus \{\mathbf{0}\}$ around $\mathbf{0}$

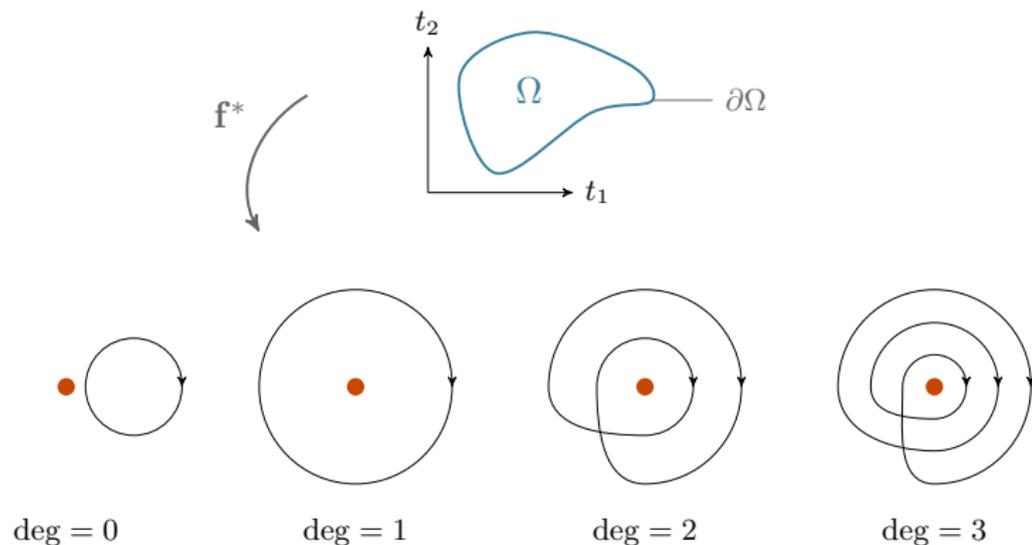


Topological degree for zero verification

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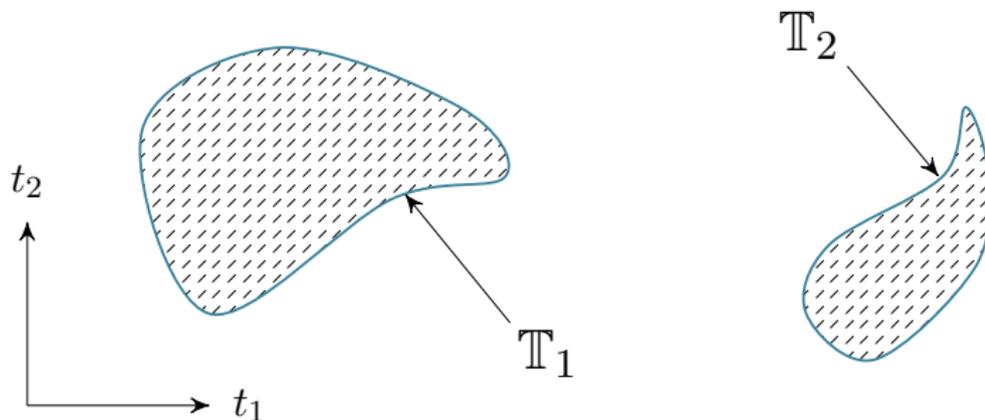
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Topological degree for zero verification

Outer approximation of a set \mathbb{T} with SIVIA

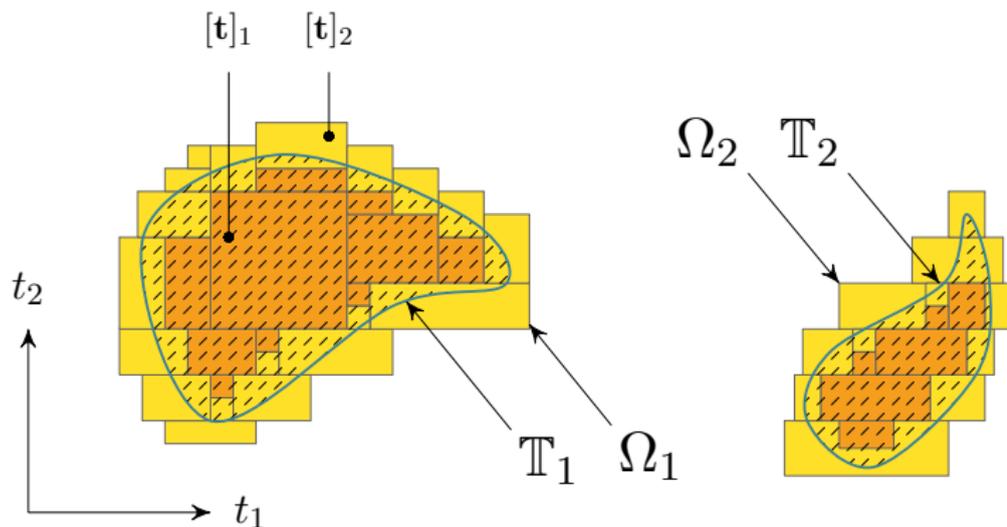
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Topological degree for zero verification

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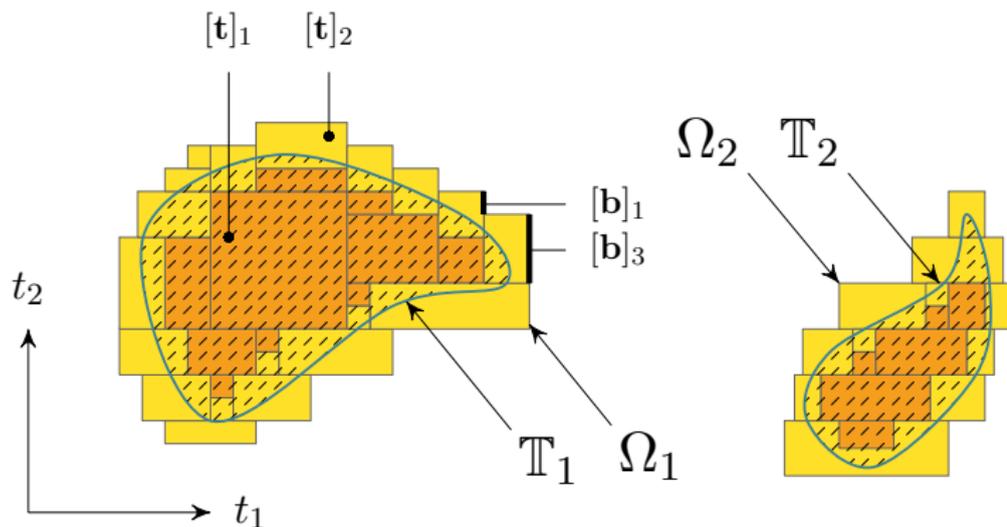
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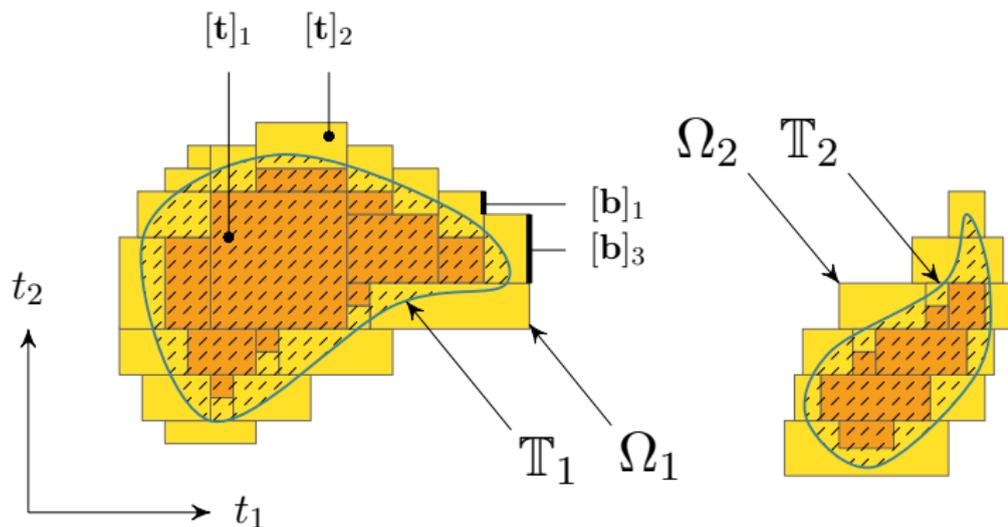
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Topological degree for zero verification

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Consider $\mathbb{T} \subset \mathbb{R}^n$ in which we want to find zeros of \mathbf{f}^* .



Outer set has the properties required for Ω : $\mathbf{f}^*(\mathbf{t}) \neq \mathbf{0}, \forall \mathbf{t} \in \partial\Omega$

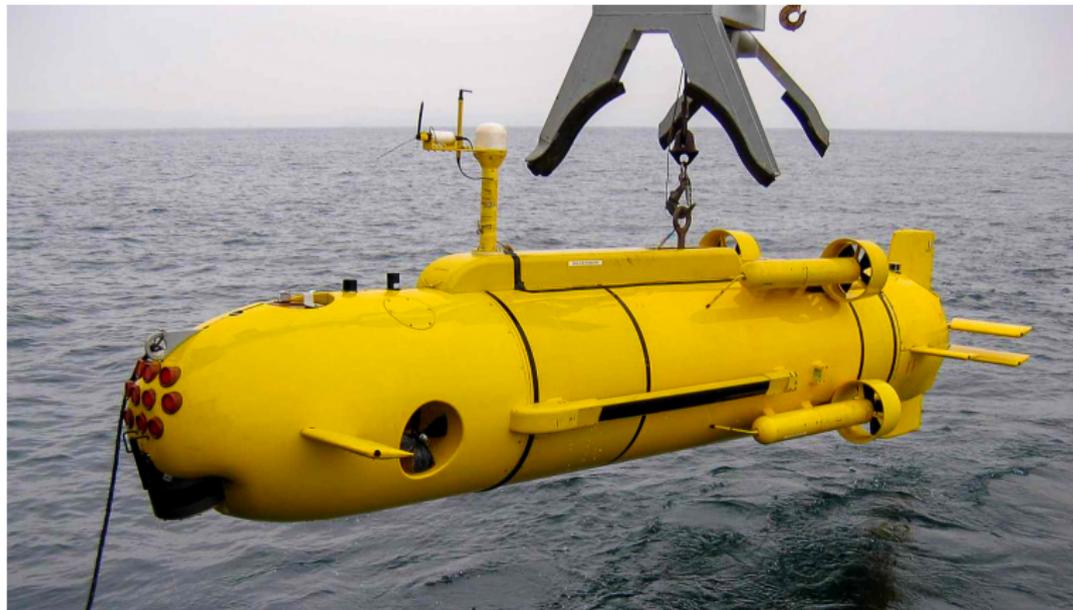
Section 6

Application

Application

Redermor mission

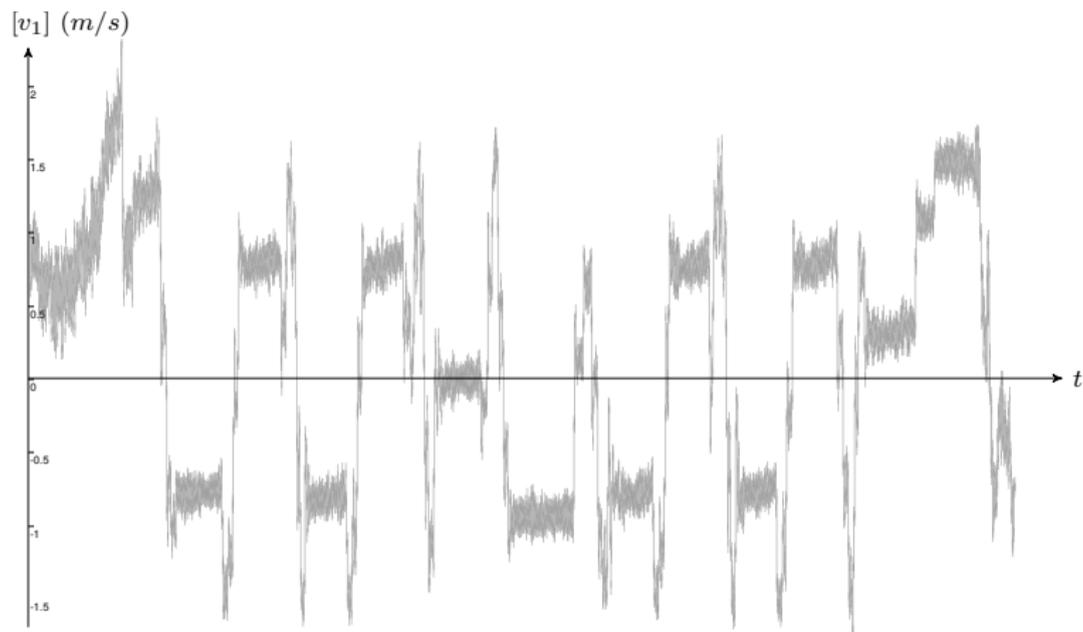
2 hours experimental mission in Brittany (France)



The *Redermor* Autonomous Underwater Vehicle (AUV)

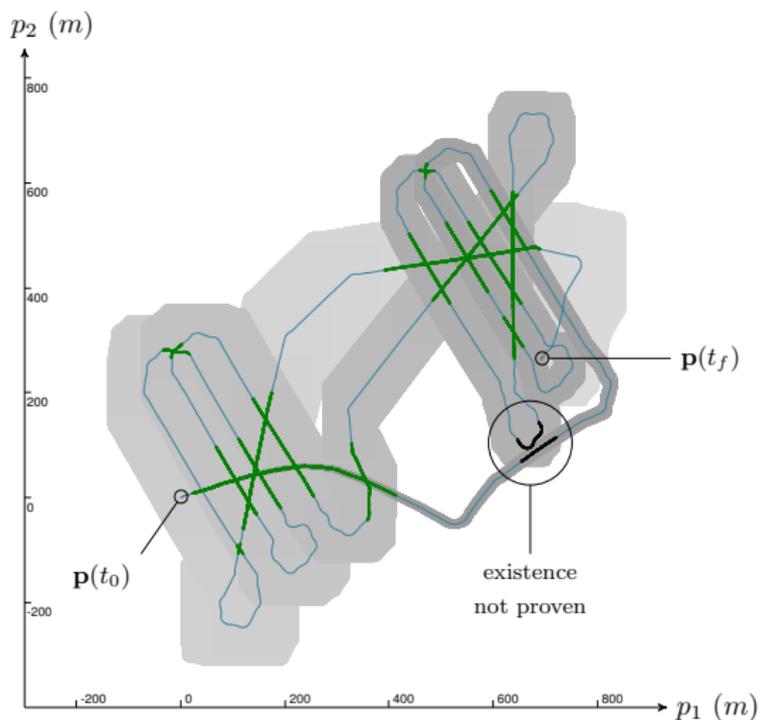
Application

Redermor mission

Tube of proprioceptive measurements $[\mathbf{v}](\cdot)$:East speed velocity tube $[v_1](\cdot)$

Application

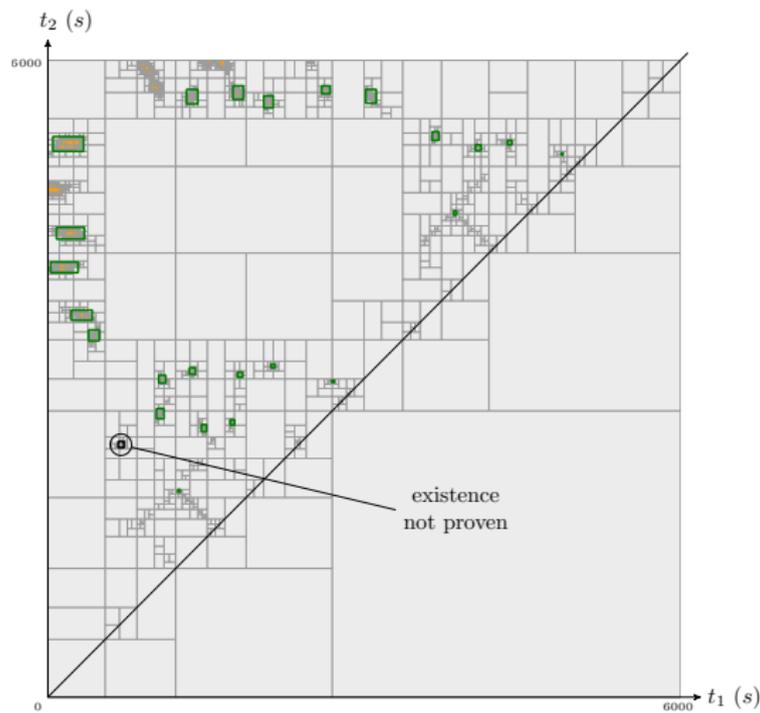
Guaranteed computation of the trajectory



2d trace of Redermor AUV

Application

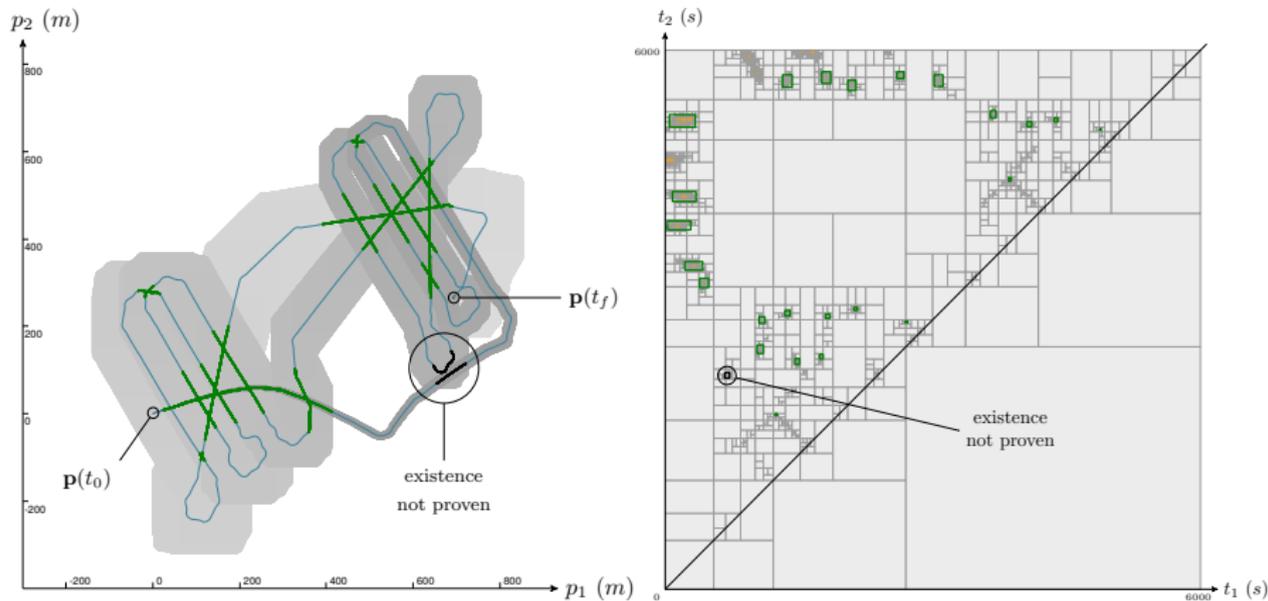
t -plane of the mission: $\mathbb{T} = \{(t_1, t_2) \mid \mathbf{0} \in [\mathbf{f}](t_1, t_2), t_1 < t_2\}$



t -plane corresponding to Redermor's mission

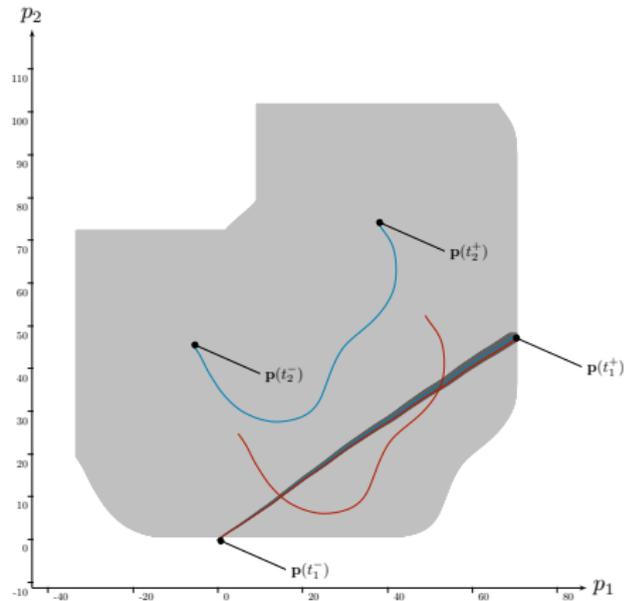
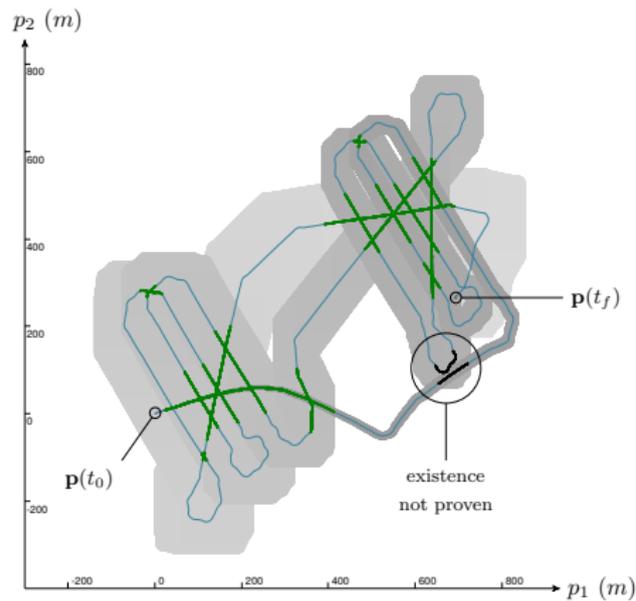
Application

Overview and results



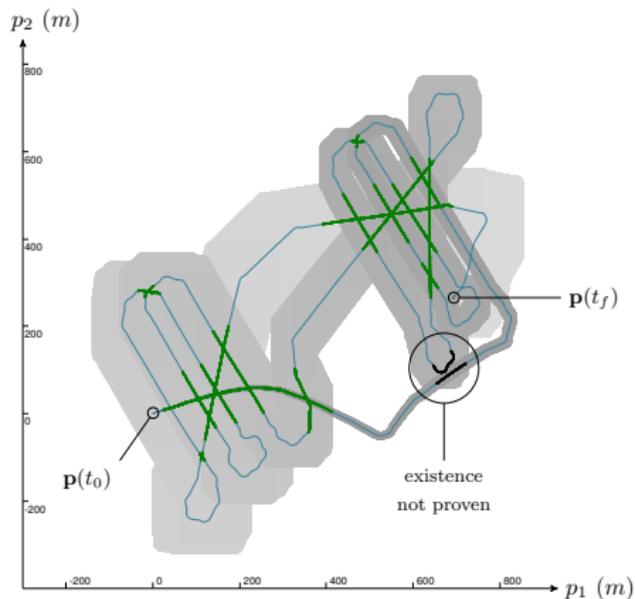
Application

Overview and results



Application

Overview and results

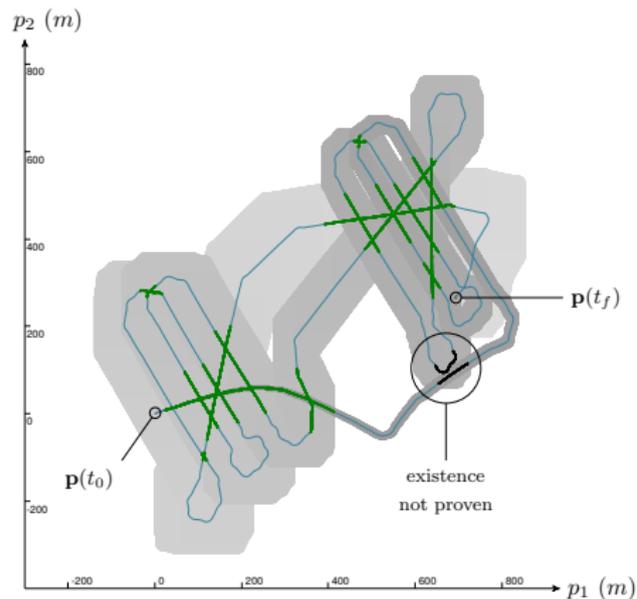
**Loop proof number**

Without uncertainties:

$$\lambda^* = \#\{\mathbf{t} \mid \mathbf{f}^*(\mathbf{t}) = \mathbf{0}, t_1 < t_2\}$$

Application

Overview and results

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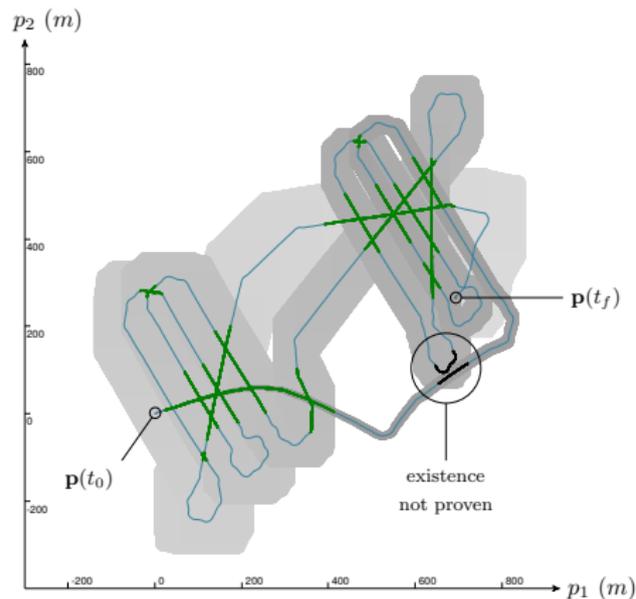
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Results:Newton operator test: $\lambda_{\mathcal{N}} = 14$

Application

Overview and results

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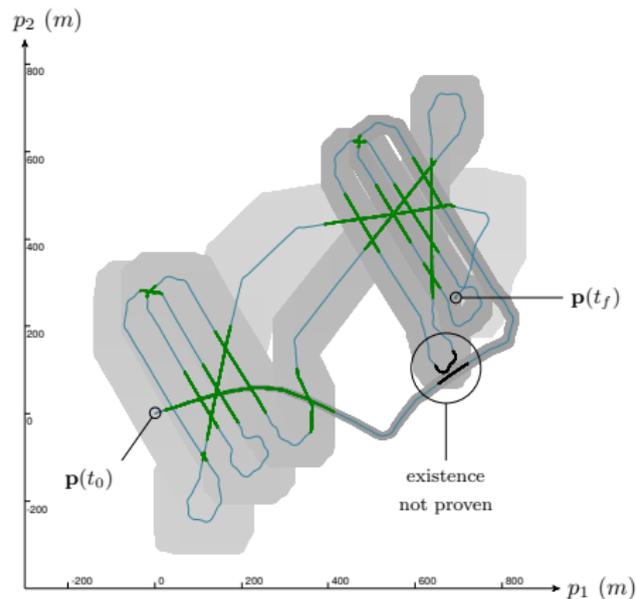
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Results:Newton operator test: $\lambda_N = 14$ Topological degree test: $\lambda_T = 24$ Truth: $\lambda^* = 24$

Conclusion

Loop proof \Leftrightarrow **verified existence of a 0** of an uncertain function:

- ▶ situation where the exact values of the function are not known
- ▶ have to deal with a reliable approximation of it

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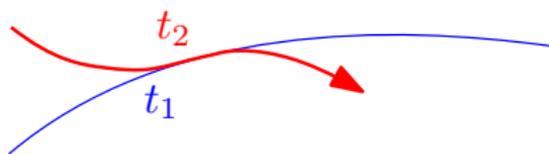
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- ▶ applied in a 2d context
- ▶ optimal results



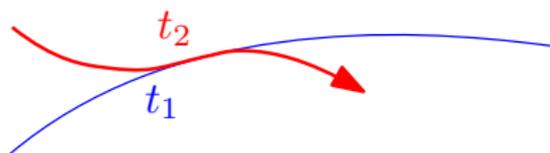
Conclusion

Loop proof \Leftrightarrow **verified existence of a 0** of an uncertain function:

- ▶ situation where the exact values of the function are not known
- ▶ have to deal with a reliable approximation of it

Topological degree theory:

- ▶ well suited in this case
- ▶ applied in a 2d context
- ▶ optimal results



■ Proving the existence of loops in robot trajectories

S. Rohou, P. Franek, C. Aubry, L. Jaulin *The International Journal of Robotics Research*, 2018

Future work

Now that loops are proved with proprioceptive measurements...

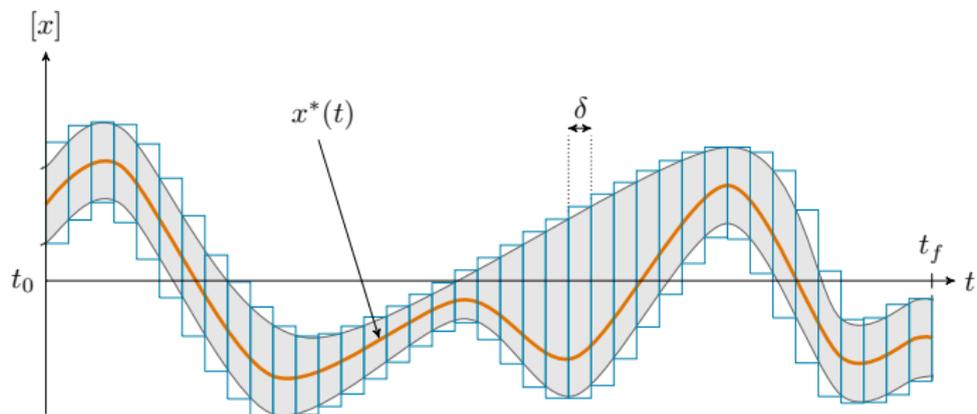


...it remains to perform localization by adding environment perceptions.

Tubex library

An open-source C++ (Python) library based on IBEX and providing tools for constraint programming over dynamical systems.

- ▶ Tube, TubeVector, ...
- ▶ contractors (delays, differential eq., time uncertainties...)
- ▶ robotic tools and applications

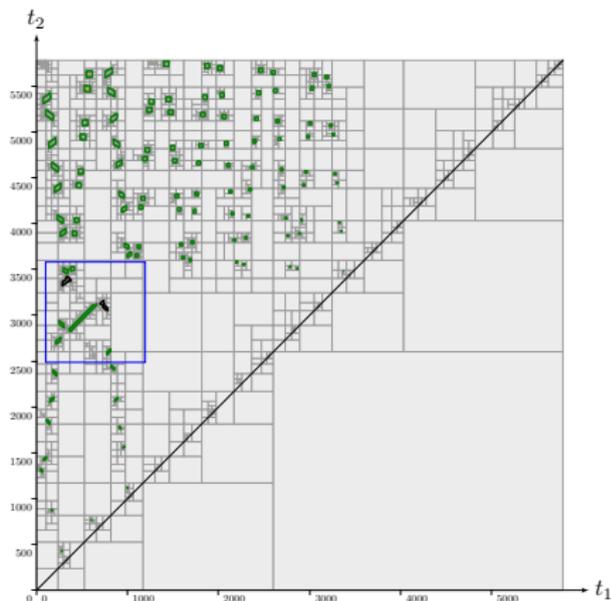
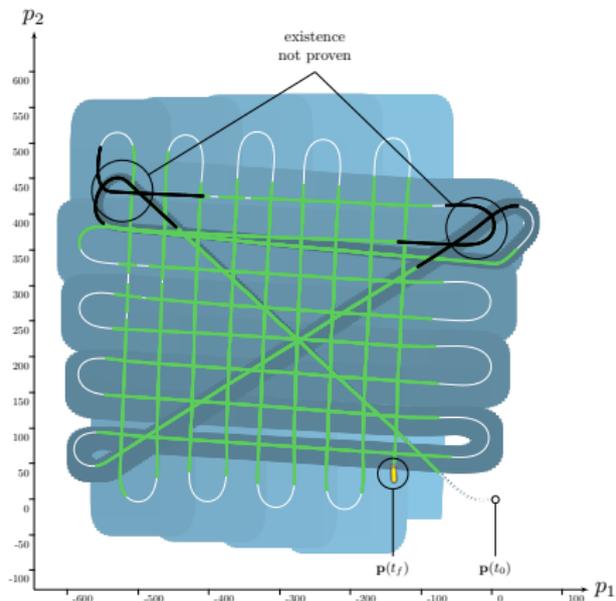


<http://www.simon-rohou.fr/research/tubex-lib/>

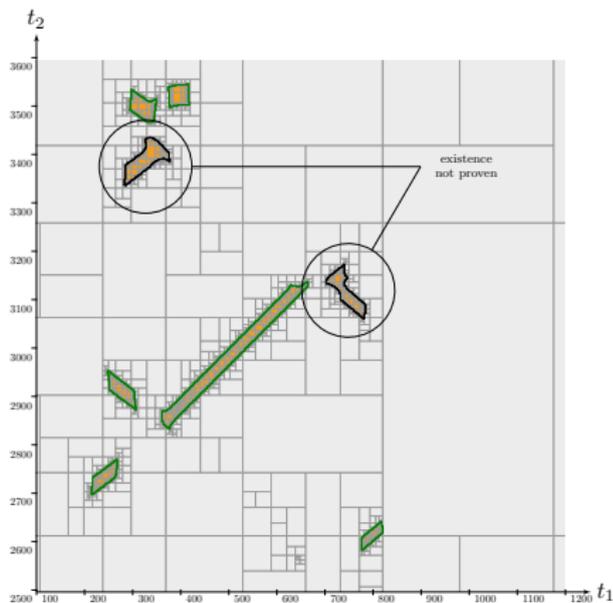
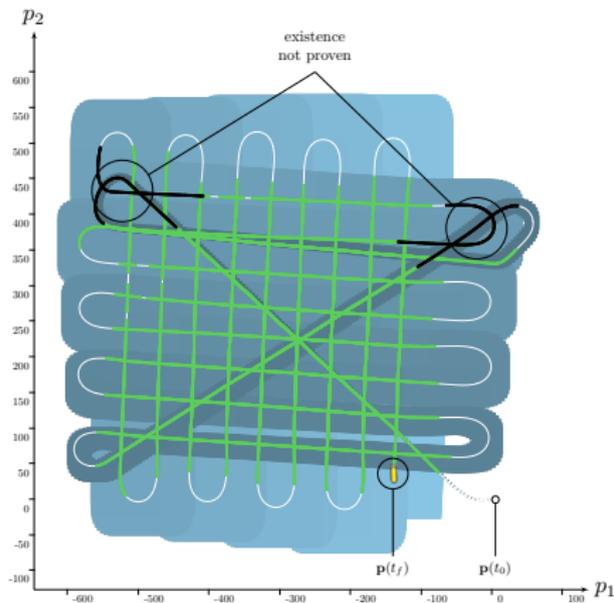
Section 10

Appendix

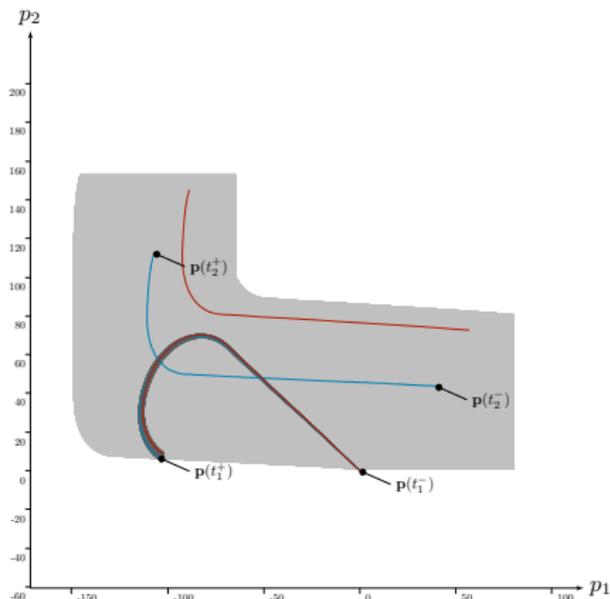
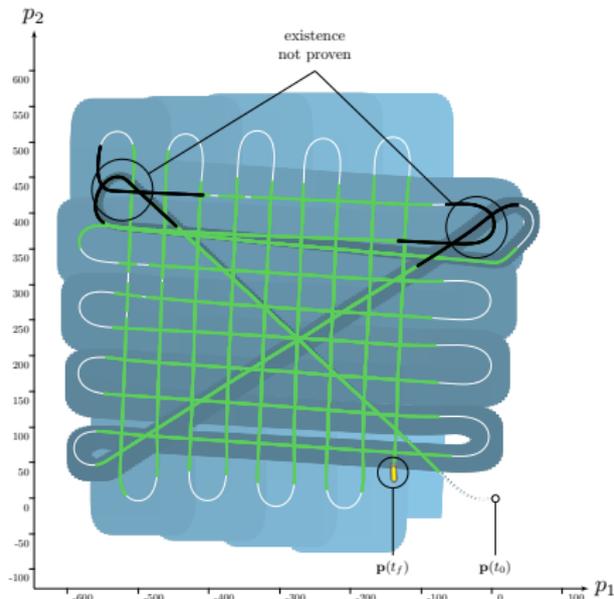
Appendix

Another experiment (*Daurade* AUV)

Appendix

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Appendix

Reliable approximation of absolute speed $\mathbf{v}^*(\cdot)$

Robot sensors for **absolute speed computation**:

- ▶ velocity sensor (DVL)
- ▶ inertial measurement unit

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Reliable approximation of absolute speed $\mathbf{v}^*(\cdot)$

Robot sensors for **absolute speed computation**:

- ▶ velocity sensor (DVL)
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Uncertainties:

- ▶ datasheets \implies standard deviation σ for each sensor

Appendix

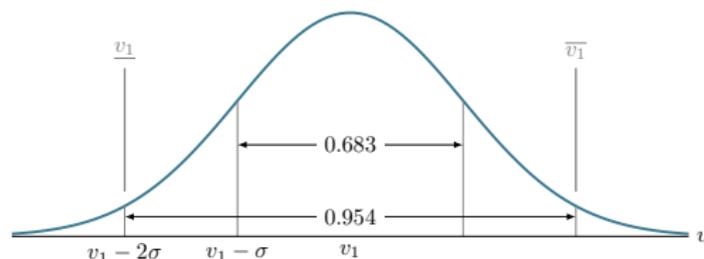
Reliable approximation of absolute speed $\mathbf{v}^*(\cdot)$

Robot sensors for **absolute speed computation**:

- ▶ velocity sensor (DVL)
- ▶ inertial measurement unit

Uncertainties:

- ▶ datasheets \implies standard deviation σ for each sensor
- ▶ 95% confidence rate: $v_1^* \in [v_1] = [v_1 - 2\sigma, v_1 + 2\sigma]$



Appendix

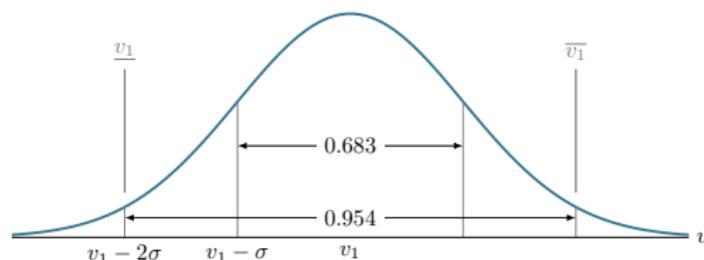
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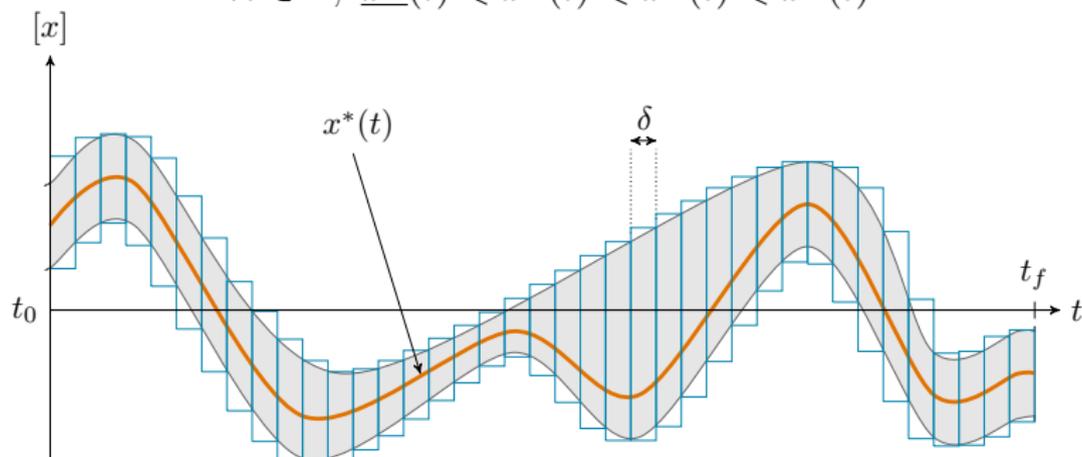
- ▶ uncertainties propagated thanks to interval arithmetic

Appendix

Tubes: computer representation

Implementation **enclosing** $[x^-(\cdot), x^+(\cdot)]$ inside an interval of step functions $[\underline{x}(\cdot), \overline{x}(\cdot)]$ such that:

$$\forall t \in \mathbb{R}, \underline{x}(t) \leq x^-(t) \leq x^+(t) \leq \overline{x}(t)$$



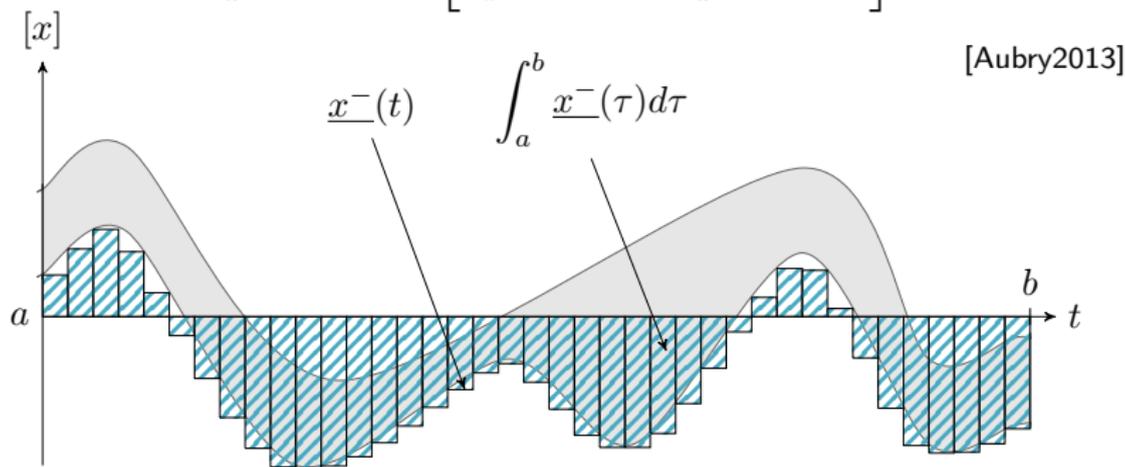
tube implementation with a set of boxes – this outer representation adds pessimism but enables guaranteed and simple computations

Appendix

Tubes integral: implementation

Outer approximation of the integral computed by:

$$\int_a^b [x](\tau) d\tau \subset \left[\int_a^b \underline{x}^-(\tau) d\tau, \int_a^b \overline{x}^+(\tau) d\tau \right]$$



blue area: outer approximation of the lower bound of the tube's integral