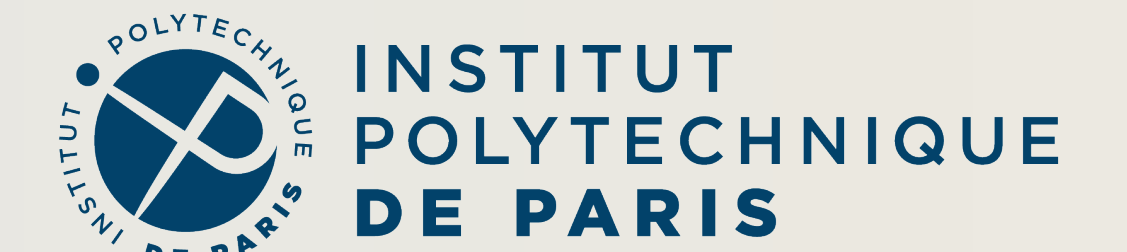


# Reachability Analysis with Measurable Time-Varying Uncertainties\*



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## Introduction

- Safety properties can be proved knowing the set of reachable states
- Computing exact sets of reachable states is often impossible
- We compute an over-approximation: a set guaranteed to contain all reachable states
- Our method is able to handle a broader class of uncertainties than most of state-of-the-art tools

## Problem formulation

- We want to compute over-approximations of the set of reachable states of the initial value problem

$$\begin{cases} \dot{x}(t) = g(u(t)) \cdot h(t, x(t)) \\ x(0) = x_0 \in \mathcal{X}_0 \subset \mathbb{R}^n \end{cases} \quad (1)$$

with  $u : [0, T] \rightarrow \mathcal{U}$  a Lebesgue-measurable bounded function, where  $\mathcal{U}$  is a compact set,  $g : \mathcal{U} \rightarrow \mathbb{R}^{n \times m}$  and  $h : \mathbb{R} \times \mathbb{R}^n \rightarrow \mathbb{R}^m$  two continuous functions.

- Right-hand side not Riemann-integrable  $\implies$  most tools produce non guaranteed approximations

## Theory

Let  $h^+$  and  $h^-$  be continuous functions such that  $h = h^+ - h^-$ . An over-approximation of the initial value problem

$$\begin{cases} \dot{x}(t) = Ah^+(t, x(t)) - Bh^-(t, x(t)) \\ x(0) = x_0 \in \mathcal{X}_0 \\ A, B \text{ such that } \exists u_A, u_B \in \mathcal{U} : g(u_A) = A \text{ and } g(u_B) = B \end{cases}$$

for all  $x_0$ ,  $A$ , and  $B$ , is an over-approximation of the problem 1.

The right-hand side is continuous, so we can apply the Picard iteration to guarantee the over-approximation of the set of reachable states, parametrized by  $t$ ,  $x_0$ ,  $A$  and  $B$ .

## Picard iteration

Let  $\dot{x}(t) = f(t, x(t))$  be an ordinary differential equation with  $f$  a continuous function and  $x(0) = x_0$  be an initial state.

Given a set-valued function  $\bar{\varphi}(t)$ , whose all images are closed, convex and bounded, such that

$$x_0 + \int_0^t f(s, \bar{\varphi}(s)) ds \subset \bar{\varphi}(t)$$

then  $\bar{\varphi}(t)$  is an over-approximation of the set of reachable states of the system at time  $t$ : for all solution  $x$ ,  $x(t) \in \bar{\varphi}(t)$ .

## Sets representation: Taylor models (TMs)

- triple  $(\mathcal{D}, p, I)$  with  $\mathcal{D} \subset \mathbb{R}^k$  be a domain,  $p : \mathcal{D} \rightarrow \mathbb{R}$  be a polynomial, and  $I \subset \mathbb{R}$  be an interval called remainder
- a Taylor model  $(\mathcal{D}, p, I)$  is an over-approximation of  $f$  on  $\mathcal{D}$  if  $\forall x \in \mathcal{D}$ ,  $f(x) \in \{p(x) + v \mid v \in I\}$
- Easy to over-approximate usual operators and easy to implement

## Positive decomposition

- Best decompositions of  $h$  using TMs minimize  $\|h_i^+\|_1 + \|h_i^-\|_1$
- If  $h(t, x) \in [a, b]$  with  $a < 0 < b$ , the best affine decomposition is

$$h_i^+ = \frac{b}{b-a}h - \frac{ab}{b-a} \quad \text{and} \quad h_i^- = \frac{a}{b-a}h - \frac{ab}{b-a}$$

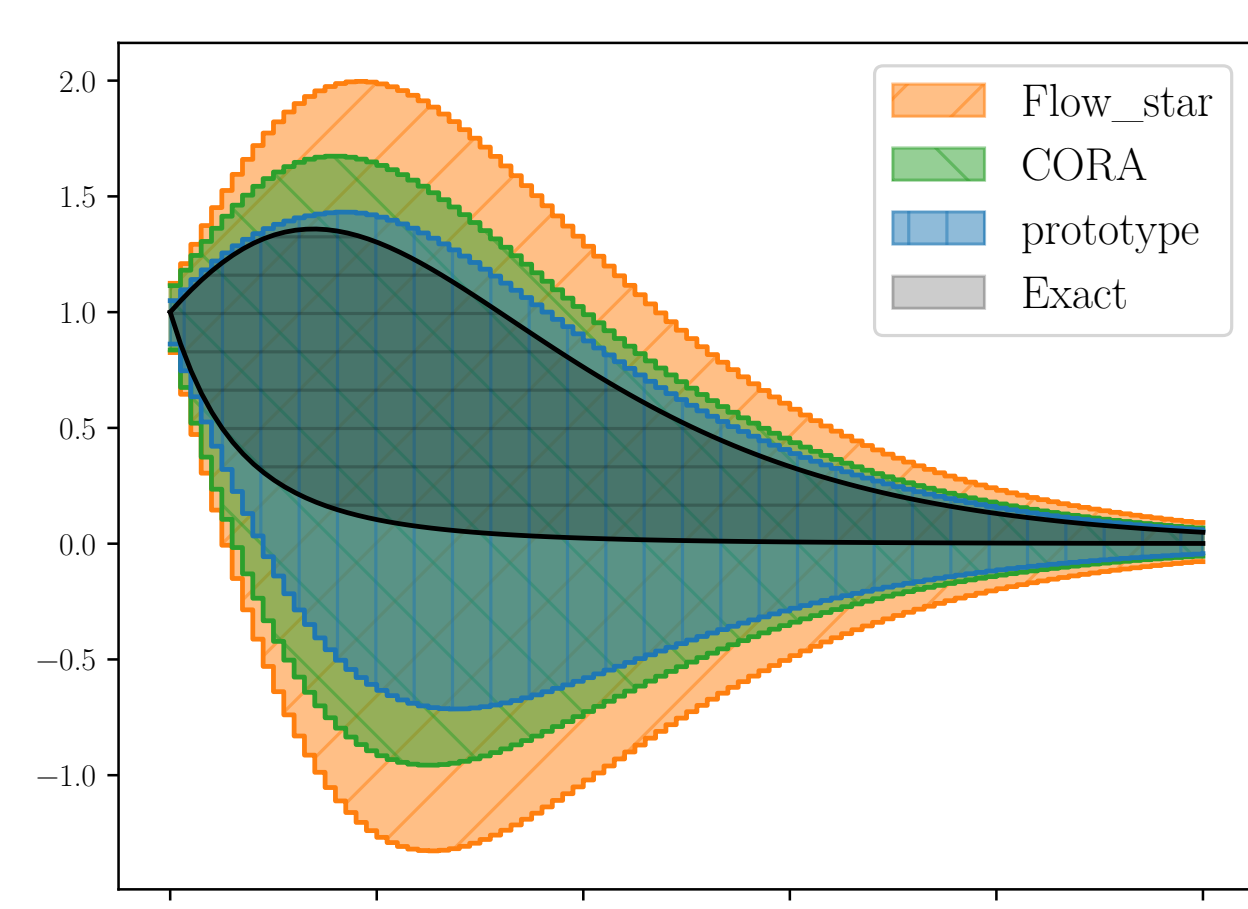
- We tried multiple decompositions on the problem  $\dot{x}(t) = (0.1-t)u(t)$  with  $x(0) = 0$  and  $u(t) \in [-1, 1]$  and displayed the results below

Decomposition	Over-Approximation
$(h + 0.1) - 0.1$	$[-0.040, 0.040]$
$(0.5h + 0.05) - (0.05 - 0.5h)$	$[-0.020, 0.020]$
$(h + 0.5)^2 - (h^2 + 0.25)$	$[-0.102, 0.102]$
$(h + 0.25)^2 - (h - 0.25)^2$	$[-0.027, 0.027]$

## Example: Nonlinear

Simple example with nonlinear dynamics:

$$\begin{cases} \dot{x}(t) = -x(t) - x(t)y(t)u(t) \\ \dot{y}(t) = -y(t) \\ t \in [0, 20] \\ u(t) \in [0, 1] \\ x(0) = 3 \end{cases}$$

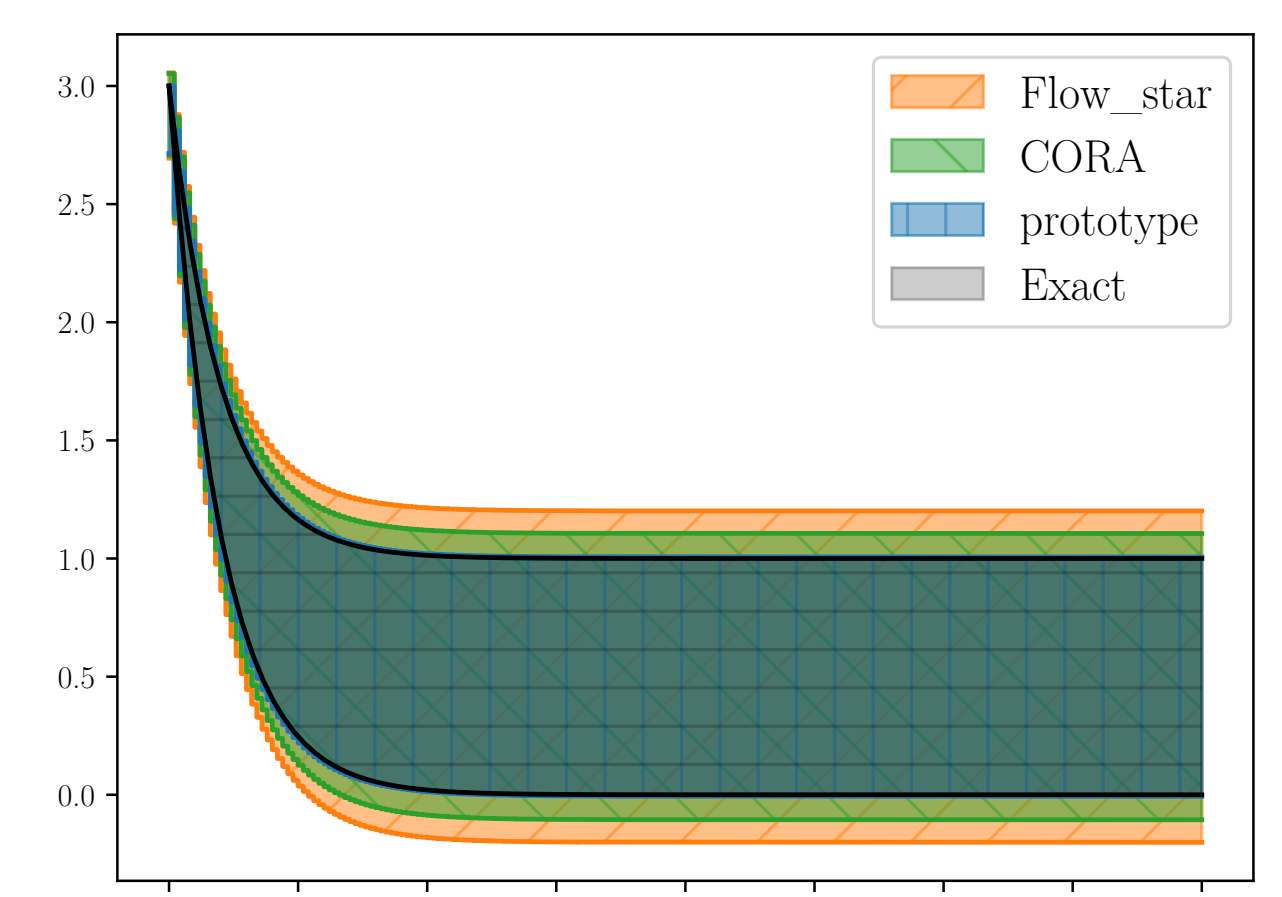


Over-approximations of  $x(t)$

## Example: Switching

Simple example with multiple possible fixed-points:

$$\begin{cases} \dot{x}(t) = u(t) - x(t) \\ t \in [0, 20] \\ u(t) \in [0, 1] \\ x(0) = 3 \end{cases}$$

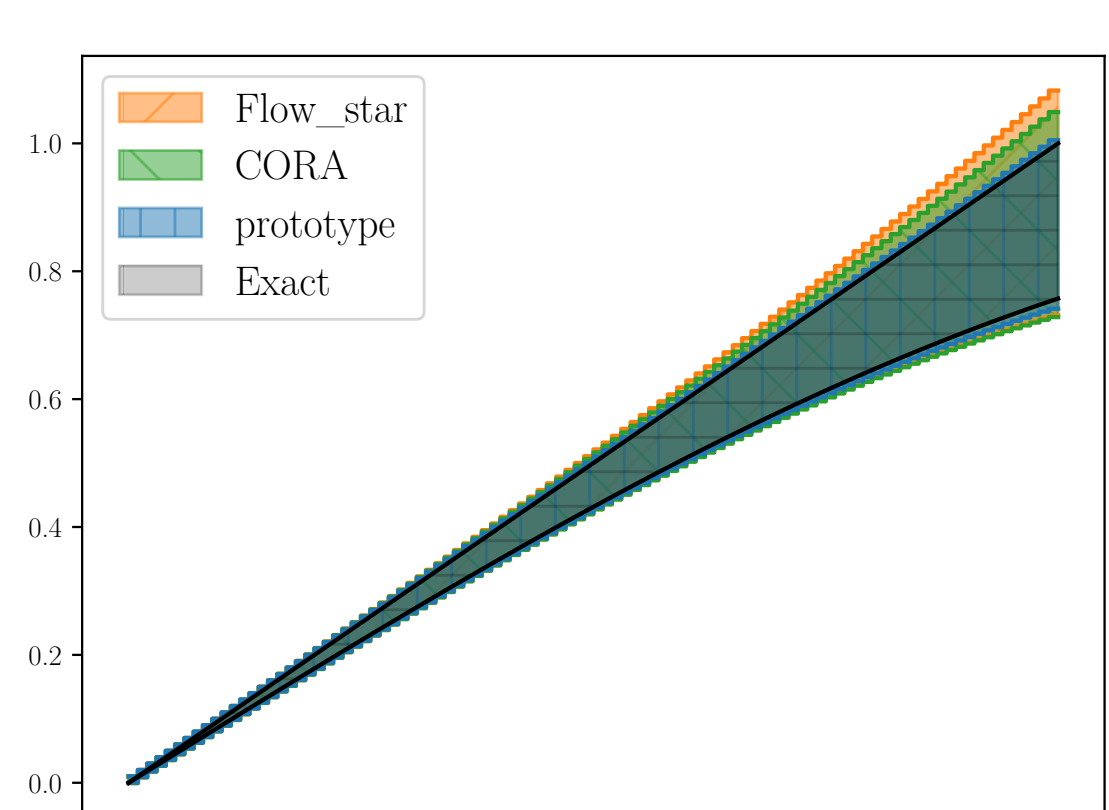


Over-approximations of  $x(t)$

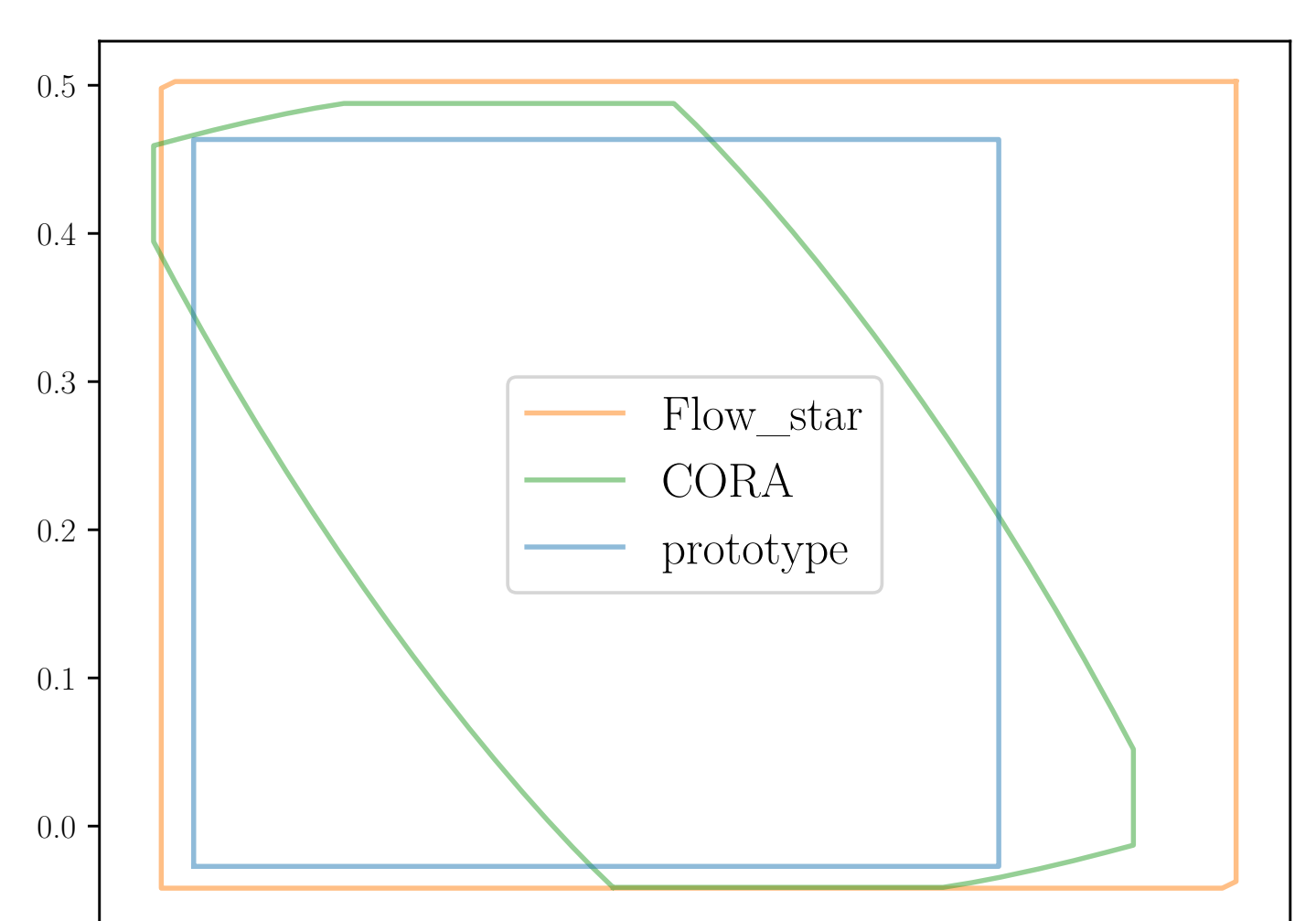
## Example: Dubins car

Variation of the Dubins car model:

$$\begin{cases} \dot{x}(t) = u_1(t) \cos(z(t)) \\ \dot{y}(t) = u_1(t) \sin(z(t)) \\ \dot{z}(t) = u_2(t) \end{cases} \quad \text{with} \quad \begin{cases} t \in [0, 1] \\ u_1(t) \in [0.9, 1]; u_2(t) \in [0, 1] \\ x(0) = y(0) = z(0) = 0 \end{cases}$$



Over-approximations of  $x(t)$



Over-approximations of  $y(1)$  wrt.  $x(1)$

## Discussion about examples

Areas	CORA	Flow*	proto	exact
<b>Nonlinear</b>	5.956	7.735	4.866	3.576
<b>Switching</b>	23.33	26.81	19.48	19.00
<b>Dubins car</b>	0.114	0.114	0.099	0.086

- Tighter over-approximations of  $x(t)$  (cf. table above)
- Loss of dependencies between variables (cf. Dubins car example)
- Handle broader class of uncertainties (Lebesgue-measurable)

## Conclusion and future work

- New method to handle Lebesgue-measurable uncertainties
- Could be generalized to stochastic uncertainties (Itô calculus)
- Tight over-approximation on simple examples  $\implies$  useful for safety properties proofs
- Could be used with transitions' abstractions of hybrid automata

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