

Reachability Analysis for Systems Affine in the Time Varying Bounded Uncertainties

François BIDET, Éric GOUBAULT and Sylvie PUTOT

Laboratoire d'Informatique de l'École polytechnique

October 15, 2020



Table of contents

Problem formulation

Enclosure of the solution

Application

Examples



Problem formulation



The problem

Initial Value Problem

$$\begin{cases} \dot{x} = f(t, x(t), u(t)) \\ x(t_0) \in \mathcal{X} \end{cases} \quad (1)$$

Affine in the uncertainties

$$f(t, x, u) = h_0(x, t) + \sum_{i=1}^m u_i h_i(x, t) \quad (2)$$

Immune Virus System

$$\begin{cases} \dot{x}(t) = \Lambda(t) - \gamma_1 x(t) - \beta x(t)v(t) \\ \dot{y}(t) = \beta x(t)v(t) - \gamma_2 y(t) \\ \dot{v}(t) = \kappa y(t) - \alpha(t)v(t) \end{cases} \quad (3)$$



The problem


Initial Value Problem

$$\begin{cases} \dot{x} = f(t, x(t), u(t)) \\ x(t_0) \in \mathcal{X} \end{cases} \quad (1)$$

Affine in the uncertainties

$$f(t, x, u) = h_0(x, t) + \sum_{i=1}^m u_i h_i(x, t) \quad (2)$$

Immune Virus System

$$\begin{pmatrix} \dot{x}(t) \\ \dot{y}(t) \\ \dot{v}(t) \end{pmatrix} = \begin{pmatrix} -\gamma_1 x(t) - \beta x(t)v(t) \\ \beta x(t)v(t) - \gamma_2 y(t) \\ \kappa y(t) \end{pmatrix} + \Lambda(t) \begin{pmatrix} 1 \\ 0 \\ 0 \end{pmatrix} + \alpha(t) \begin{pmatrix} 0 \\ 0 \\ -v(t) \end{pmatrix} \quad (3)$$


Problem hypotheses

Hypotheses

$$\begin{cases} \dot{x} = f(t, x(t), u(t)) = h_0(x(t), t) + \sum_{i=1}^m u_i(t) h_i(x(t), t) \\ x(0) \in \mathcal{X} \\ u(t) \in \mathcal{U} \end{cases} \quad (4)$$

with

- ▶ $\mathcal{X} \subset \mathbb{R}^n$ bounded
- ▶ $\mathcal{U} \subset \mathbb{R}^m$ close convex bounded
- ▶ $t \mapsto u(t)$ measurable

Carathéodory, for all $u : [0, T] \rightarrow \mathcal{U}$:

- ▶ There exists $m : [0, T] \rightarrow \mathbb{R}_+$ such that $\forall (t, x), \|f(t, x, u(t))\| \leq m(t)$
- ▶ There exists $k : [0, T] \rightarrow \mathbb{R}_+$ such that $\forall (t, x_1, x_2), \|f(t, x_1, u(t)) - f(t, x_2, u(t))\| \leq k(t) \|x_1 - x_2\|$



Difference with other tools

Other tools use at least Riemann-integrable uncertainties:

Flow*: uncertainties are assumed continuous

CORA: uncertainties are assumed Riemann-integrable



Enclosure of the solution



Inclusion using Lebesgue-integration

Lemma

Let a measurable function $u : [0, T] \rightarrow \mathcal{U}$ with a closed convex bounded set $\mathcal{U} \subset \mathbb{R}$. Let a Lebesgue-integrable function g with a decomposition in positive functions: $g = g^+ - g^-$. Then

$$\int_0^T g(s)u(s) ds \in \left\{ u_1 \int_0^T g^+(s) ds - u_2 \int_0^T g^-(s) ds \mid u_1 \in \mathcal{U}, u_2 \in \mathcal{U} \right\} \quad (5)$$



New operator

Parametric operator

$$\begin{aligned}\mathbb{P}_{(u_i)}(\rho) := t \mapsto & x_0 + \int_0^t h_0(\rho(s), s) ds \\ & + \sum_{i=1}^m u_{2i-1} \int_0^t h_i^+(\rho(s), s) ds \\ & - \sum_{i=1}^m u_{2i} \int_0^t h_i^-(\rho(s), s) ds \quad (6)\end{aligned}$$

Global operator

$$\mathbb{P}(\rho) := t \mapsto \left\{ \mathbb{P}_{(u_i)}(\rho)(t) \mid \forall i \in \llbracket 1, 2m \rrbracket, u_i \in \mathcal{U} \right\} \quad (7)$$



Fixed-point theorem

Theorem

Let x_0 an initial state and let \mathcal{U} a closed bounded convex set of possible values of the inputs. If there exists a set of functions φ such that for all $t \in [0, T]$:

$$\mathbb{P}(\varphi)(t) \subset \varphi(t) \quad (8)$$

then for all $t \in [0, T]$, $\varphi(t)$ is an over-approximation of the reachable set at time t with the initial state x_0 .



Application



Algorithm to compute an over-approximation

Algorithm

We use Taylor Models as sets representations and we replace all uncertainties by a centered Taylor Model: $u(t) \in [a, b]$ becomes $\text{TM}\left(\frac{a+b}{2}, \left[\frac{a-b}{2}, \frac{b-a}{2}\right]\right)$.

1. Compute a raw enclosure of the solution
2. Decompose the functions $h_{i \geq 1}(x(t), t)$ as difference of positive functions using the raw enclosure
3. Compute the polynomial expansion up to the expected order
4. Find a valid remainder



Decomposition

Affine decomposition

Assume for all $t \in [0, T]$, $h_i(x(t), t) \in [a, b]$ and $a < 0 < b$. We define

$$\begin{cases} h_i^+(x(t), t) = \frac{b}{b-a} h_i(x(t), t) - \frac{ab}{b-a} \\ h_i^-(x(t), t) = \frac{a}{b-a} h_i(x(t), t) - \frac{ab}{b-a} \end{cases} \quad (9)$$

and we have $h_i(x(t), t) = h_i^+(x(t), t) - h_i^-(x(t), t)$.

Optimality

This decomposition minimizes $\|h_i^+\|_1 + \|h_i^-\|_1$.



Examples



Example 1

Simple dynamics

$$\dot{x}(t) = (0.1 - t)u(t) \quad (10)$$

with $x(0) = 0$ and $\forall t \in [0, 0.2]$, $u(t) \in [-1, 1]$

Case: u constant

If u is constant ($u(t) = u(0) \in [-1, 1]$), then $x(0.2) = 0$ and $x(t) \in [-0.005, 0.005]$.

Exact reachable set

The exact reachable set at time $t = 0.2$ is $x(0.2) \in [-0.01, 0.01]$.



Example 1: Decomposition

Decomposition

For all $t \in [0, 0.2]$, $h_1(x(t), t) = (0.1 - t) \in [-0.1, 0.1]$.

We deduce $h_1(x, t) = h_1^+(x, t) - h_1^-(x, t)$ with

$$\begin{cases} h_1^+(x, t) = 0.1 - 0.5t \\ h_1^-(x, t) = 0.5t \end{cases} \quad (11)$$

Equivalent dynamics

The dynamics becomes

$$\dot{x}(t) = (0.1 - 0.5t) u(t) - (0.5t) u(t) \quad (12)$$



Example 1: Over-approximation

We replace all occurrences of $u(t)$ by $\text{TM}(0, [-1, 1])$.

We start with an expansion to the order 0 in time:

$$\varphi_0(x_0, t) = \text{TM}(x_0, [0]).$$

Then, to expected an higher order expansion, we iterate

$$\varphi_{n+1} = \mathbb{P}(\varphi_n):$$

$$\begin{aligned}\varphi_1(x_0, t) &= \text{TM}(x_0, [0]) + \text{TM}(0, [-1, 1]) \int_0^t h_1^+(\varphi_0(x_0, s), s) ds \\ &\quad - \text{TM}(0, [-1, 1]) \int_0^t h_1^-(\varphi_0(x_0, s), s) ds \\ &= \text{TM}(x_0, [0]) + \text{TM}(0, [-1, 1]) \cdot \text{TM}(0.1t - 0.25t^2, [0]) \\ &\quad - \text{TM}(0, [-1, 1]) \cdot \text{TM}(0.25t^2, [0]) \\ &= \text{TM}(x_0, [0]) + \text{TM}(0, [-0.01, 0.01]) \\ &\quad - \text{TM}(0, [-0.01, 0.01]) \\ &= \text{TM}(x_0, [-0.02, 0.02])\end{aligned}$$



Example 1: Over-approximation

Result

For all $t \in [0, 0.2]$, we have $x(t) \in [-0.02, 0.02]$.

(Remind) Exact reachable set

The exact reachable set at time $t = 0.2$ is $x(0.2) \in [-0.01, 0.01]$.



Decreasing exponential

Dynamics

$$\dot{x}(t) = -u(t)x(t) \quad (13)$$

with $x(0) \in [1, 1.1]$ and $\forall t, u(t) \in [1, 2]$.



Decreasing exponential

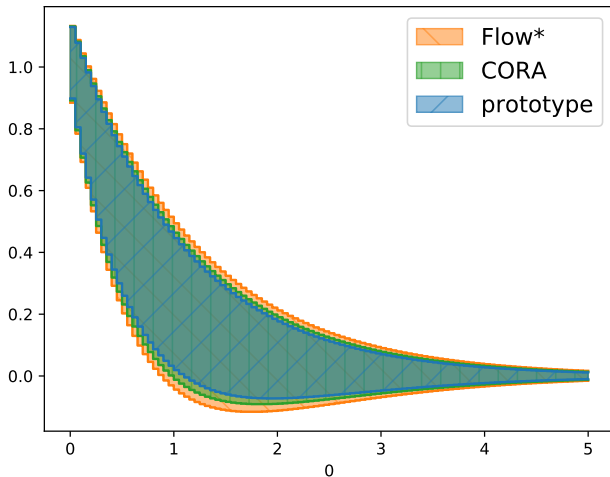


Figure: Over-approximations with fixed time-step equals to 0.05



Nonlinear perturbation

Dynamics

$$\begin{cases} \dot{x}(t) = -x(t) + x(t)y(t)u(t) \\ \dot{y}(t) = -y(t) \end{cases} \quad (14)$$

with $x(0) = 1$, $y(0) = 2$ and $u(t) \in [-1, 1]$.



Nonlinear perturbation

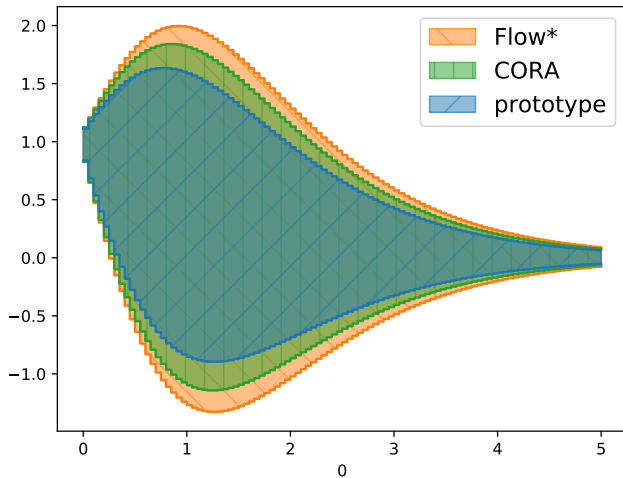


Figure: Over-approximations with fixed time-step equals to 0.05



Thank you for your attention

