An Interval Arithmetic Approach to Input-Output Reachability*

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Overview

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1. Motivation and Problem Statement

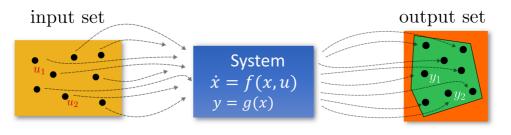


Figure 1: Reachable Set and Overestimation.

Definition 1 (Reachable Set)

Given the set of inputs ${\cal U}$ and a set of initial conditions ${\cal X}_0$, the reachable set at time t is

$$\mathsf{Reach}(\mathcal{X}_0,\mathcal{U})(t) = \left\{ \overbrace{\phi(t,u,x_0)}^{\mathsf{trajectory}} \in \mathbb{R}^n : \mathsf{for some} \ u : [0,t] \to \mathcal{U}, x_0 \in \mathcal{X}_0 \right\} \tag{1}$$

- ▶ **Goal:** compute the reachable set not by simulating each possible trajectory one by one.
- ► **Methodologies:** set-based methods, mixed-monotonicity, Hamilton-Jacobi reachability, Koopman operators, neural networks.

1. Motivation and Problem Statement

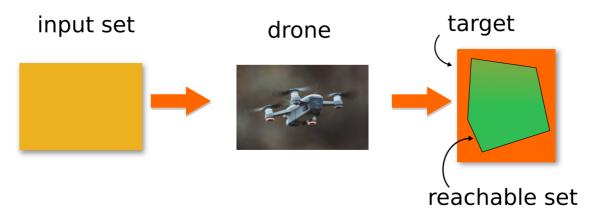


Figure 2: Target reaching.

2.1 Preliminaries: Chen-Fliess series

- ▶ An alphabet is a set of symbols (letters): $X = \{x_0, x_1, \dots, x_m\}$
- ightharpoonup A word is defined as the catenation $\eta = x_{i_1} \cdots x_{i_k}$ of letters from X.
- ightharpoonup The set of words of any length: X^*

Example 1

 $X = \{x_0, x_1\}$. We can form the words

$$\eta_1 = x_0 x_1$$
 $\eta_2 = x_1 x_0$
 $\eta_3 = x_1^2 = x_1 x_1$

▶ A power series c is a function $c: X^* \to \mathbb{R}^{\ell}$ and represented as $c = \sum_{\eta \in X^*} (c, \eta) \eta$.

Example 2

$$c_1 = \sum_{k=1}^{\infty} x_1^k, \qquad c_2 = \sum_{k=0}^{\infty} (k+1) \cdot x_0 x_1 x_2^k x_1$$

2.1 Preliminaries: Chen-Fliess series

Definition 2 (Iterated integrals)

Let $X=\{x_0,\cdots,x_m\}$, $\eta=x_i\xi\in X^*$, the input vector $u(t)=(u_1(t),\cdots,u_m(t))\in L_1^m[0,T]$

$$E_{\phi}[u](t) := 1, \quad E_{x_i \xi}[u](t) := \int_0^t \frac{u_i(\tau) E_{\xi}[u](\tau) d\tau}{u_i(\tau) E_{\xi}[u](\tau) d\tau}$$

Definition 3 (Chen-Fliess series)

Consider the power series $c = \sum_{\eta \in X^*} (c, \eta) \eta$, the Chen-Fliess series associated to c is

$$F_{c}[u](t) = \sum_{\eta \in X^{*}} (c, \eta) E_{\eta}[u](t)$$
word
$$\text{word} \qquad \text{word}$$

$$\text{vector}$$

$$\text{in } \mathbb{R}^{\ell} \qquad \text{iterated integral}$$
space of all words

2.1 Preliminaries: Chen-Fliess series

Theorem 1 (State-Space realization (Fliess 1983))

The Chen-Fliess series
$$F_c \in \mathbb{R}_{LC}\langle\langle X \rangle
angle$$
 represents the system $\underline{\underline{m}}$

$$\dot{z} = g_0(z) + \sum_{i=0}^{m} g_i(z)u_i, \ z(t_0) = z_0$$

$$y = h(z)$$

if and only if
$$(c, \eta) = L_{g_{\eta}} h(z)|_{z_0} \text{ where } L_{g_{\eta}} h_j(z_0) := L_{g_{i_1}} \cdots L_{g_{i_k}} h_j(z)|_{z_0} \text{ for } \eta = x_{i_k} \cdots x_{i_1} \in X^*$$

$$L_{g_{i_1}} \cdots L_{g_{i_k}} h(z_0) = \frac{\partial}{\partial z} \left(\cdots \left(\frac{\partial}{\partial z} \left(\frac{\partial h(z)}{\partial z} \cdot g_{i_k}(z) \right) \cdot g_{i_{k-1}}(z) \right) \cdots \right) \cdot g_{i_1}(z) \right] . \tag{4}$$

•
$$c = \sum_{n \in X^*} (c, \eta) \eta$$
 has finite Lie rank.

and $L_a f(z) = (\frac{\partial}{\partial z} f(z)) \cdot g(z)$. Equivalently,

► There exist $K, M \ge 0$ such that $|(c, \eta)| \le KM^{|\eta|} |\eta|!$

$$i{=}1$$

(3)

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(2)

2.2 Preliminaries: Mixed-Monotonicity

Consider the Lipschitz continuous function in each argument $f: \mathcal{X} \times \mathcal{U} \to \mathbb{R}^n$. The continuous-time dynamical system is defined

$$\dot{x}(t) = f(x(t), u(t))
x(0) = x_0$$
(5)

Definition 4 (Decomposition function)

Let $d: \mathcal{X} \times \mathcal{U} \times \mathcal{X} \times \mathcal{U} \to \mathbb{R}^n$ locally Lipschitz continuous satisfying

- $i. \ d(x, u, x, u) = f(x, u) \text{ for all } (x, u) \in \mathcal{X} \times \mathcal{U}.$
- ii. $\frac{\partial d_i}{\partial x_i}(x,u,\hat{x},\hat{u})\geq 0$ for any $i\neq j$, and for any $(x,u,\hat{x},\hat{u})\in\mathcal{X}\times\mathcal{U}\times\mathcal{X}\times\mathcal{U}$.
- $iii. \ \ \frac{\partial d_i}{\partial \hat{x}_\cdot}(x,u,\hat{x},\hat{u}) \leq 0 \ \text{for any} \ i,j, \ \text{and for any} \ (x,u,\hat{x},\hat{u}) \in \mathcal{X} \times \mathcal{U} \times \mathcal{X} \times \mathcal{U}.$
- $\begin{array}{ll} iv. & \frac{\partial d_i}{\partial u_k}(x,u,\hat{x},\hat{u}) \geq 0, & \frac{\partial d_i}{\partial \hat{u}_k}(x,u,\hat{x},\hat{u}) \leq 0 \text{ for any } i,k\text{, and for any } \\ & (x,u,\hat{x},\hat{u}) \in \mathcal{X} \times \mathcal{U} \times \mathcal{X} \times \mathcal{U}. \end{array}$

d is a decomposition function of (5).

2.2 Preliminaries: Mixed-Monotonicity

Theorem 2 (Coogan 2020)

The following is a **decomposition function** of the system:

$$d_{i}(x, u, \hat{x}, \hat{u}) = \begin{cases} \min_{\substack{y \in [x, \hat{x}] \\ y_{i} = x_{i} \\ z \in [u, \hat{u}]}} f_{i}(y, z), & x \leq \hat{x}, u \leq \hat{u}, \\ \max_{\substack{y \in [\hat{x}, x] \\ y_{i} = x_{i} \\ z \in [\hat{u}, u]}} f_{i}(y, z), & \hat{x} \leq x, \hat{u} \leq u. \end{cases}$$

$$(6)$$

Definition 5 (Coogan 2020)

Given the decomposition function d, the **embedding system** is

$$\begin{bmatrix} \dot{x} \\ \dot{\hat{x}} \end{bmatrix} = e(x, u, \hat{x}, \hat{u}) := \begin{bmatrix} d(x, u, \hat{x}, \hat{u}) \\ d(\hat{x}, \hat{u}, x, u) \end{bmatrix}, \tag{7}$$

where the input $(u, \hat{u}) \in \mathcal{U} \times \mathcal{U} \subset \mathbb{R}^{2m}$ and the state $(x, \hat{x}) \in \mathcal{X} \times \mathcal{X} \subset \mathbb{R}^{2n}$.

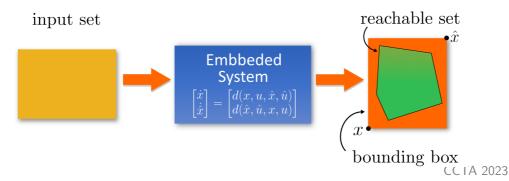
2.2 Preliminaries: Mixed-Monotonicity

- $lack \Phi^e(t,a,b)$: trajectory of the embedding system where $a,b\in\mathbb{R}^{2n}$.
- $[x, y] \subset \mathbb{R}^n$: hyperrectangle defined by the coordinatewise inequality $x \leq y$.
- $ightharpoonup \|a\| \subset \mathbb{R}^n$: hyperrectangle defined by $a \in \mathbb{R}^{2n}$ where a = (x, y) and $x \in \mathbb{R}^n$, $y \in \mathbb{R}^n$.

Theorem 3 (Coogan 2020)

Consider the decomposition d associated with (5), $\mathcal{X}_0 = [x, \overline{x}] \subset \mathcal{X}$ and $\mathcal{U} = [u, \overline{u}]$, then

$$\textit{Reach}([\underline{x},\overline{x}],[\underline{u},\overline{u}])(t) \subset [\![\Phi^e(t,(\underline{x},\overline{x}),(\underline{u},\overline{u}))]\!], \textit{ for all } t \geq 0$$



2.3 Preliminaries: Other Approaches Using Chen-Fliess Series

- 1. Mixed-Monotonicity for Chen-Fliess series (**IPA** and LDE 2022a)
- 2. Gradient Descent using Chen-Fliess series (**IPA** and LDE 2022b)
- 3. Newton Optimization using Chen-Flies series (IPA and LDE 2023a)
- 4. Trust regions Optimization using Chen-Flies series (IPA and LDE 2023b)
- 5. Critical Points of Chen-Flies series (LDE and SG and IPA 2023c)
- 6. Backward and Inner Approximation using Chen-Flies series (IPA and LDE 2023d)
- ► IPA = Ivan Perez Avellaneda
- ► LDE = Luis A. Duffaut Espinosa
- ► SG = W. Steven Grav

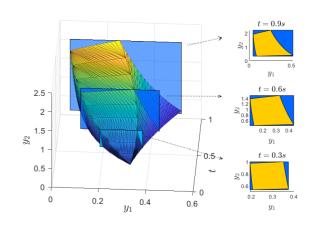


Figure 3: Output reachable set of a Lotka-Volterra system and its MBB in terms of its maximum and minimum outputs.

3.1 Main results: Interval Arithmetics

Goal: compute overestimation of the reachable set using IA and Chen-Fliess series.

Definition 6 (Interval Product)

Given $I_1 = [a_1, b_1] \subset \mathbb{R}$ and $I_2 = [a_2, b_2] \subset \mathbb{R}$, the product $I_1 \cdot I_2 := [\underline{I}, \overline{I}]$ where

$$\underline{I} = \min_{\substack{y_1 \in [a_1,b_1] \\ y_2 \in [a_2,b_2]}} y_1 y_2, \quad \overline{I} = \max_{\substack{y_1 \in [a_1,b_1] \\ y_2 \in [a_2,b_2]}} y_1 y_2.$$

Equivalently,

$$[a_1,b_1]\cdot [a_2,b_2] = [\min\{a_1a_2,a_1b_2,b_1a_2,b_1b_2\}, \ \max\{a_1a_2,a_1b_2,b_1a_2,b_1b_2\}].$$

and

$$\underbrace{[a,b]\cdots[a,b]}_{n \text{ times}} = [a,b]^n.$$

Definition 7

Given a set $\mathcal{X} \subset \mathbb{R}^n$ and $\lambda \in \mathbb{R}$, $\lambda \mathcal{X} := \{\lambda x : x \in \mathcal{X}\}.$

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3.1 Main results: Interval Arithmetics

Example 3

▶
$$[-2,1] \cdot [1,3] = [\min\{-2,-6,1,3\}, \max\{-2,-6,1,3\}] = [-6,3]$$
▶ $[-2,1]^2 = [-2,4]$

Example 4

3[-2,1] = [-6,3].

Lemma 1

Consider $u: \mathbb{R} \to \mathbb{R}^m \in [\mathbf{a}, \mathbf{b}]$ (i.e., $u_i(t) \in [a, b]$). For $\eta = x_{i_1} \cdots x_{i_n} \in X^*$, then

Consider
$$u: \mathbb{R} \to \mathbb{R} \in [\mathbf{d}, \mathbf{D}]$$
 (i.e., $u_i(t) \in [u, v]$). For $\eta = x_{i_1} \cdots x_{i_n} \in A$, then

 $Reach_{\eta}([\mathbf{a},\mathbf{b}])(t) \subset [a,b]^{|\eta|-|\eta|_{x_0}} \frac{t^{|\eta|}}{|\eta|!}, \forall \ t \in [0,T].$

Example 5

Consider
$$F_c[u](t)=E_{x_1^2}[u](t)$$
 and $u\in[-2,1]$. We want to overaproximate $\mathrm{Reach}_c(\mathcal{U})(t)$
$$E_{x_1^2}[u](t)=\int_0^t u(\tau_1)\int_0^{\tau_1} u(\tau)d\tau d\tau_1$$

Since $u \in [-2, 1]$ $-2\tau_1 \leq \int_0^{\tau_1} u(\tau)d\tau \leq \tau_1,$

3.2 Main results: Interval Arithmetics and Iterated Integrals

$$u(\tau_1) \int_0^{\tau_1} u(\tau) d\tau \in [-2, 1] \cdot [-2, 1] \tau_1$$

 $\min_{u \in \mathcal{U}} E_{\eta_1}[u](t) + \min_{u \in \mathcal{U}} E_{\eta_2}[u](t) \le \min_{u \in \mathcal{U}} (E_{\eta_1}[u](t) + E_{\eta_2}[u](t))$

 $\max_{u \in \mathcal{U}} (E_{\eta_1}[u](t) + E_{\eta_2}[u](t)) \le \max_{u \in \mathcal{U}} E_{\eta_1}[u](t) + \max_{u \in \mathcal{U}} E_{\eta_2}[u](t).$

Remark:

and

$$E_{x_1^2}[u](t) = \int_0^t u(\tau_1) \int_0^{\tau_1} u(\tau) d\tau d\tau_1 \in [-2, 1]^2 \frac{t^2}{2!}.$$

Therefore,

it follows that

$$-x_1^{-1}$$

$$\mathsf{Reach}_c(\mathcal{U})(t) \subset [-2,1]^2 \frac{t^2}{2!}$$

$$u(\tau)d\tau d\tau_1$$

$$d\tau d au_1 \in$$

$$d\tau_1 \in [-2,1]^2 \frac{1}{2}$$

$$|^2\frac{t^2}{2!}.$$

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3.3 Main results: Reachable Set Overestimation

Consider

$$x(0) = x_0$$
$$y = h(x)$$

 $\operatorname{\mathsf{Reach}}_c(\mathcal{U})(T) := \Big\{ y = F_c[u](T) \in \mathbb{R}^\ell \ : \text{for some } u : [0,t] \to \mathcal{U} \Big\}.$

 $\dot{x}(t) = f(x(t), u(t))$

and $\phi(t, u, x_0)$ is the trajectory function. Definition 8 (State-Space Reachable Set)

The RS of (8) with inputs in U:

$$\mathsf{Reach}(\mathcal{X}_0,\mathcal{U})(T) := \Big\{ \phi(T,u,x_0) \in \mathbb{R}^n : \mathsf{for \ some} \ x_0 \in \mathcal{X}_0, u : [0,t] o \mathcal{U} \Big\}.$$

Definition 9 (Input-Output Reachable Set)

The DC of the CES E [1](t) representing (0) and (0) with insure in 1/2

The RS of the CFS
$$F_c[u](t)$$
 representing (8) and (9) with inputs in \mathcal{U} :

3 15/21

(8)

(9)

3.3 Main results: Reachable Set Overestimation

Lemma 2

Given the interval $I = [a, b] \subset \mathbb{R}$, its n-th power I^n is given by

$$I^n = \begin{cases} [a^n, b^n], & a, b > 0, n \text{ even} \\ [b^n, a^n], & a, b < 0, n \text{ even} \\ [\min\{a^{n-1}b, ab^{n-1}\}, & \max\{|a|, |b|\}^n], a < 0, b > 0, n \text{ even} \\ [a^n, b^n], & ab > 0, n \text{ odd} \\ [\min\{a^n, ab^{n-1}\}, & \max\{a^{n-1}b, b^n\}], & a < 0, b > 0, & n \text{ odd} \end{cases}$$

 $1 = (1, \dots, 1)^{\top}$ for all $t \in [0, T]$.

Theorem 4

Given
$$c \in \mathbb{R}_{LC}\langle\langle X \rangle\rangle$$
 with $u(t) \in [\mathbf{a},\mathbf{b}]$ for all $t>0$. The RS of the CFS $F_c[u](t)$

$$\mathit{Reach}_c([\mathbf{a},\mathbf{b}])(t) \subset \left[F_{\underline{c}}[\mathbb{1}](t),F_{\overline{c}}[\mathbb{1}](t)\right], \ \forall t \in \mathbb{R},$$

$$\mathit{where}\ (\underline{c},\eta) = \min\left\{(c,\eta)[a,b]^{|\eta|-|\eta|_{x_0}}\right\}, \ (\overline{c},\eta) = \max\left\{(c,\eta)[a,b]^{|\eta|-|\eta|_{x_0}}\right\} \ \mathit{and}$$

(10)

4. Simulations

Example 6 Consider the system

$$\dot{x} = xu, \ y = x, \ x_0 = 1$$

with $u \in [1, 2.8]$. The Chen-Fliess series is given by

$$F_c[u](t) = 1 + \sum_{k=1}^{\infty} E_{x_1^k}[u](t).$$

The embedding system (Mixed-Monotonicity):

$$\dot{x} = xu, \ \dot{\hat{x}} = \hat{x}\hat{u}, (x_0, \hat{x}_0) = (1, 1)$$

Using interval arithmetics:

$$(\overline{c},\eta)=2.8^k$$
 and $(c,\eta)=1, |\eta|=k$

(11)

4. Simulations

1.5

0.05

0.15

0.2

$$\operatorname{Reach}_{c}([1,2.8])(t) \subset [F_{\underline{c}}[1](t), F_{\overline{c}}[1](t)] = \left[1 + \sum_{k=1}^{\infty} \frac{t^{k}}{k!}, 1 + \sum_{k=1}^{\infty} 2.8^{k} \frac{t^{k}}{k!}\right].$$

$$\begin{array}{c} 4.5 \\ 4 \\ 3.5 \end{array}$$

$$\begin{array}{c} CFS-IA \text{ upper bound} \\ --- MM \text{ lower bound} \\ --- MM \text{ upper bound} \end{array}$$

$$\begin{array}{c} 3 \\ 2.5 \\ 2 \end{array}$$

Figure 4: Reachable set overestimation using the mixed-monotonicity vs CFS with interval arithmetics.

0.25

0.3

0.35

0.4

0.45

4. Simulations

Example 7

Consider the Lotka-Volterra system

$$\dot{x}_1 = -x_1x_2 + x_1u_1, \quad \dot{x}_2 = x_1x_2 - x_2u_2, \quad y = x_1, \quad x_0 = (1/6, 1/6)^{\top}$$

The embedding system (Mixed-Monotonicity):

$$\dot{x}_1 = -x_1 \hat{x}_2 + x_1 u_1
\dot{x}_2 = x_2 x_1 - x_2 \hat{u}_2
\dot{\hat{x}}_1 = -\hat{x}_1 x_2 + \hat{x}_1 \hat{u}_1
\dot{\hat{x}}_2 = \hat{x}_2 \hat{x}_1 - \hat{x}_2 u_2$$

with $(x_{1,0}, x_{2,0}, \hat{x}_{1,0}, \hat{x}_{2,0}) = (1/6, 1/6, 1/6, 1/6)$

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4. Simulation

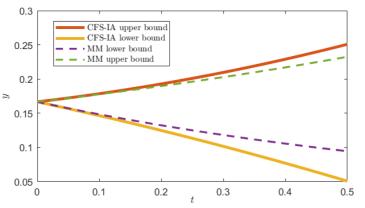


Figure 5: Reachable set overestimation using the mixed-monotonicity vs CFS with interval arithmetics.

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5. Conclusions

- 1. **Closed form:** for overestimating reachable sets.
- 2. **Faster computation:** compared with non-convex optimization approaches.
- 3. **Good accuracy:** short periods of time.

Questions? https://iperezav.github.io