

# On Mixed-Monotonicity of Chen-Fliess series\*

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**Ivan Perez Avellaneda and Luis A. Duffaut Espinosa**

Electrical and Biomedical Engineering Department

University of Vermont

\*Supported by



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# Outline

1. Motivation and Problem Statement
2. Preliminaries
  - 2.1. Chen-Fließ Series
  - 2.2. Mixed-Monotonicity Reachability
3. Data-Driven Mixed-Monotonicity
4. Illustrative Simulations
5. Conclusion

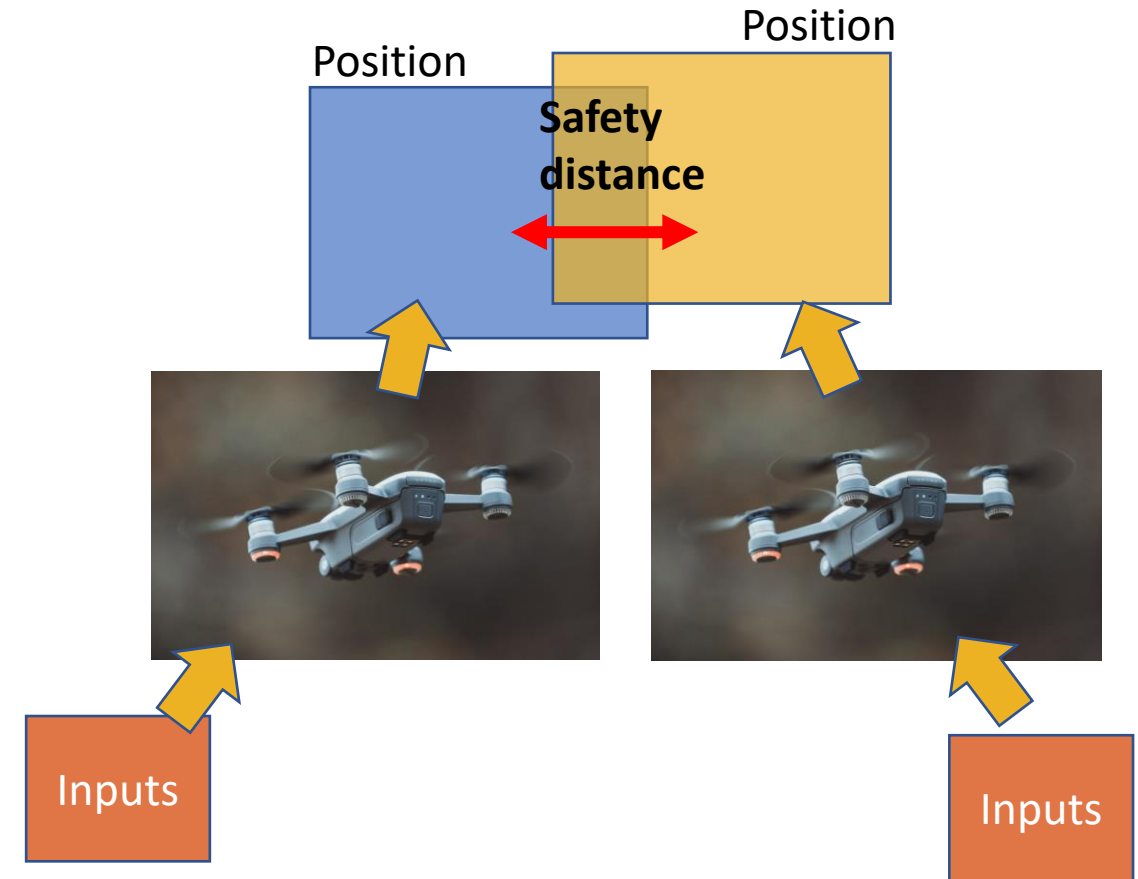
# 1. Motivation and Problem Statement

## Definition 1 (Reachable set):

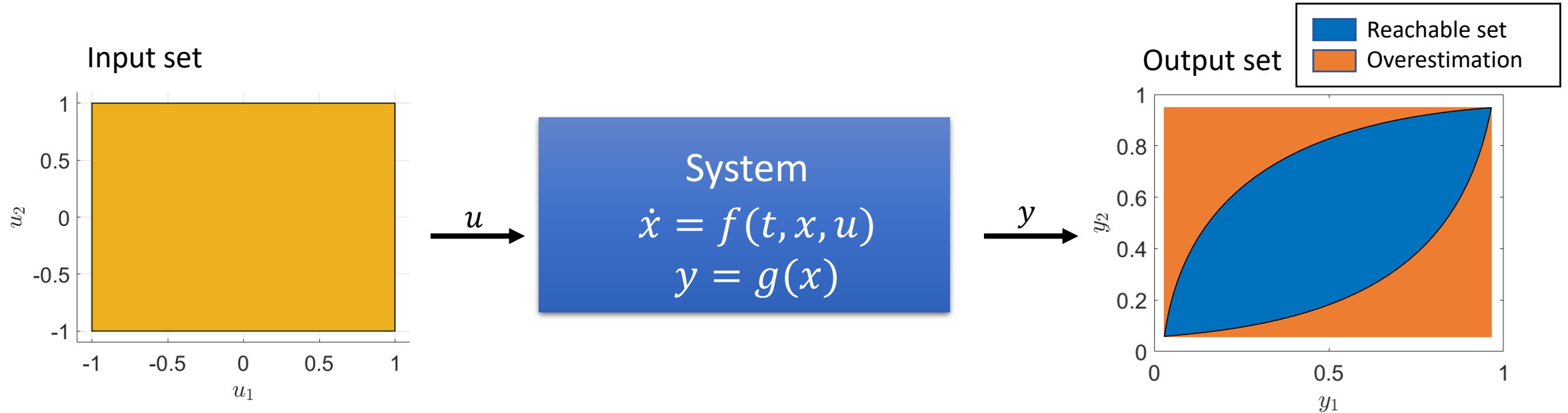
The set of outputs of the system as a response to a set of inputs and a set of initial conditions.

## Applications:

1. Collision avoidance of quadcopters
2. Power systems safety
3. Robotics safety

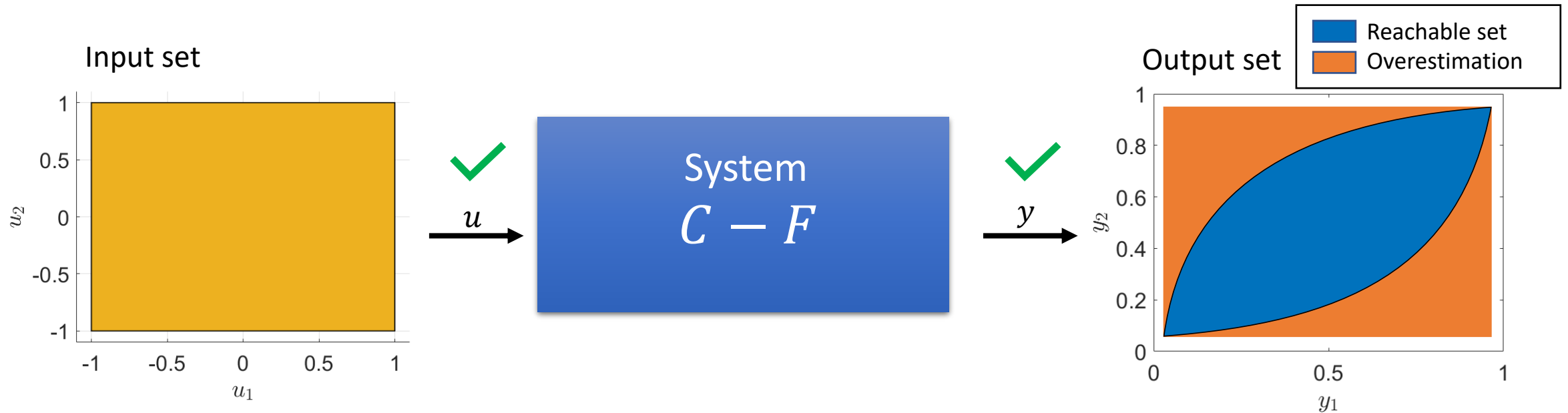


# 1. Motivation and Problem Statement



- Common methodologies: Set-based methods, Hamilton-Jacobi reachability, neural networks.
- Why a square? It only needs two points to compute.
- Approach: mixed-monotonicity.

# 1. Motivation and Problem Statement



- Why? Most real systems are impossible to know with certainty.
- What to do? Use a data-driven approach.
- Data-driven approaches: neural networks, **Chen-Fliess series**.
- Objective: extend MM to Chen-Fliess series and provide an overestimation.

## 2. Preliminaries

### 2.1 Chen-Fliess Series

Definition 2 (Chen-Fliess series):

$$y = F_c[u](t) := \sum_{\eta \in X^*} \underbrace{(c, \eta)}_{\substack{\text{Vector} \\ \text{in } \mathbb{R}^l}} E_\eta[u](t)$$

words  $\nearrow$   $\eta \in X^*$   $\nwarrow$  words  
 $\nwarrow$  Space of all words  $\nearrow$  Iterative integral

Word:  $x_{i_1} \cdot x_{i_2} \cdots x_{i_n}$   
Language:  $X = \{x_1, \dots, x_m\}$

Definition 3 (Iterative integral):

$$\underbrace{E_{x_i \nu}[u](t)}_{E_\phi[u](t)} := \int_0^t \underbrace{u_i(\tau)}_{\text{words}} \underbrace{E_\nu[u](\tau)}_{\text{Iterative integral}} d\tau,$$
$$E_\phi[u](t) := 1$$

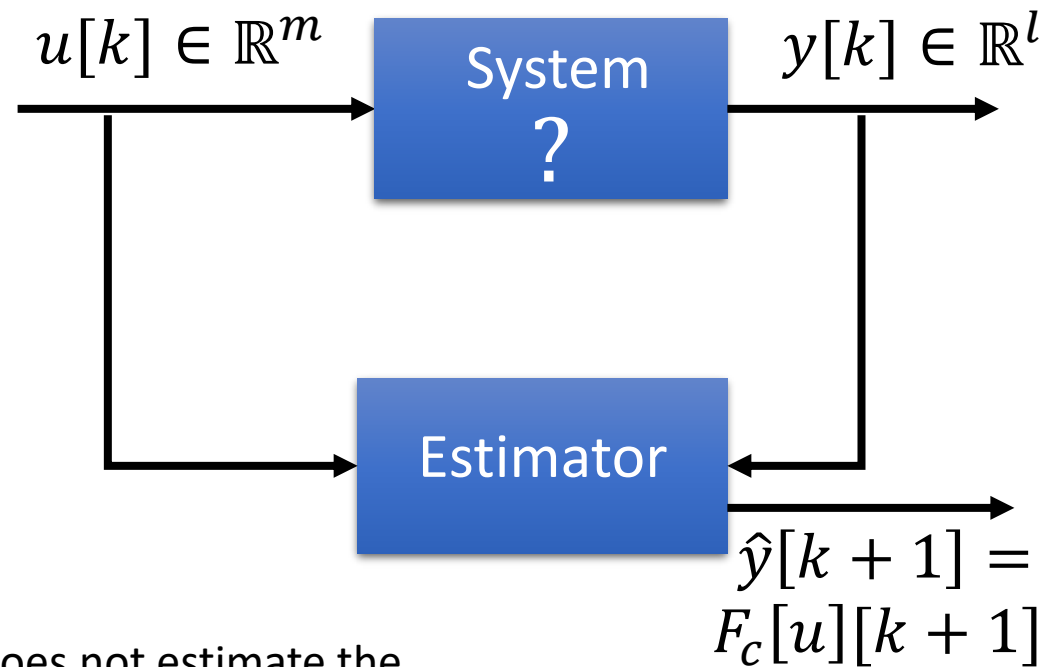
**Remark:** we need to truncate to a word length to use them

[Fliess, 1981]

## 2. Preliminaries

### 2.1 Chen-Fliess Series

Application:

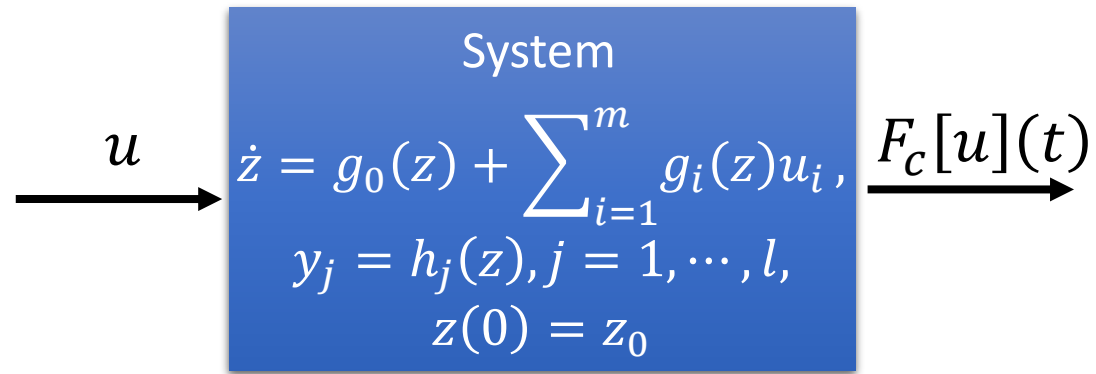


**Remark:** Chen-Fliess series does not estimate the system, but it can be used as a framework to do so.

[Gray, Venkatesh, Duffaut Espinosa, 2019]

## 2. Preliminaries

### 2.1 Chen-Fliess Series



**Remark:** the coefficients are computed deterministically in terms of Lie derivatives.

[Fliess, 1981]

**Example:** Lotka-Volterra system

$$\dot{x}_1 = -x_1x_2 + x_1u_1,$$

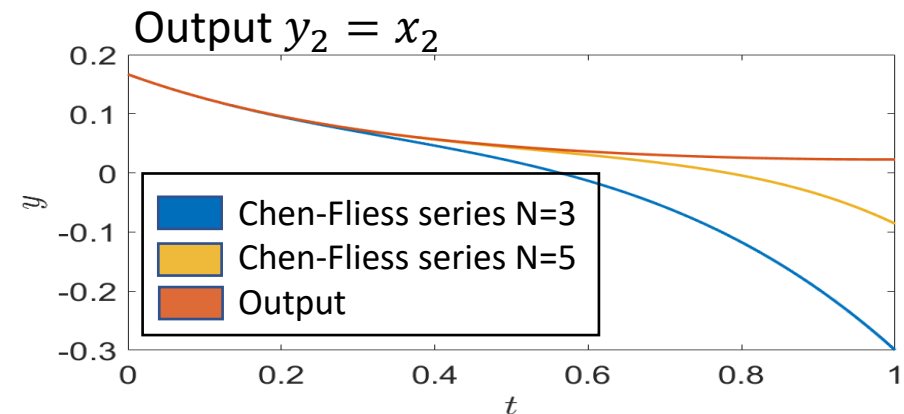
$$\dot{x}_2 = x_1x_2 - x_2u_2,$$

$$y = x,$$

$$x(0) = (1/6, 1/6).$$

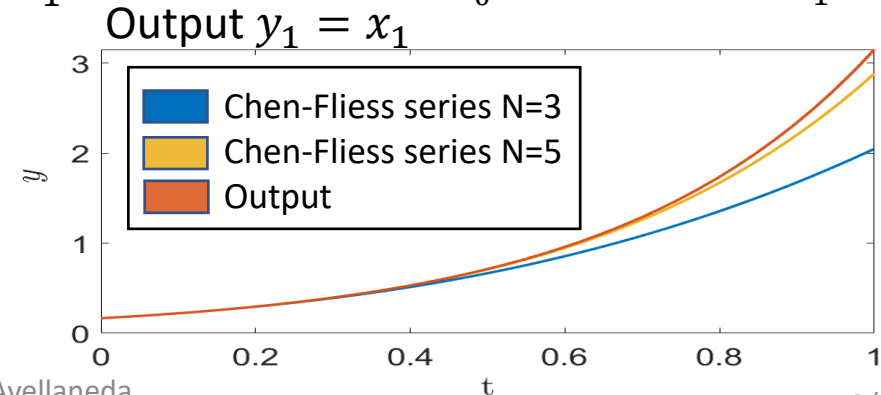
Chen-Fliess series:

$$F_{c_2}[u] = 0.17 + 0.28E_{x_0}[u](t) - 0.17E_{x_2}[u](t) + \dots$$



Chen-Fliess series:

$$F_{c_1}[u] = 0.17 - 0.28E_{x_0}[u](t) + 0.17E_{x_1}[u](t) + \dots$$





## 2. Preliminaries

### 2.2 Mixed-Monotonicity (MM)

#### Definition 4 (Mixed-Monotone system):

A system  $\dot{x} = f(x, u)$ ,  $x \in \mathcal{X} \subset \mathbb{R}^n$ ,  $u \in \mathcal{U} \subset \mathbb{R}^m$  is mixed-monotone with respect to the decomposition function  $d: \mathcal{X} \times \mathcal{U} \times \mathcal{X} \times \mathcal{U} \rightarrow \mathbb{R}^n$  if

- i.  $d(x, u, x, u) = f(x, u)$ ,
- ii.  $\frac{\partial d_i}{\partial x_j}(x, u, \hat{x}, \hat{u}) \geq 0$  for all  $i \neq j$ ,
- iii.  $\frac{\partial d_i}{\partial \hat{x}_j}(x, u, \hat{x}, \hat{u}) \leq 0$  for all  $i, j$ ,
- iv.  $\frac{\partial d_i}{\partial u_k}(x, u, \hat{x}, \hat{u}) \geq 0$  and  $\frac{\partial d_i}{\partial \hat{u}_k}(x, u, \hat{x}, \hat{u}) \leq 0$  for all  $i, k$ .

#### Definition 5 (Embedding system):

The embedding system of a mixed-monotone system with respect to the decomposition function  $d: \mathcal{X} \times \mathcal{U} \times \mathcal{X} \times \mathcal{U} \rightarrow \mathbb{R}^n$  is

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

[Abate, Dutreix, Coogan, 2020]

## 2. Preliminaries

### 2.2 Mixed-Monotonicity (MM)

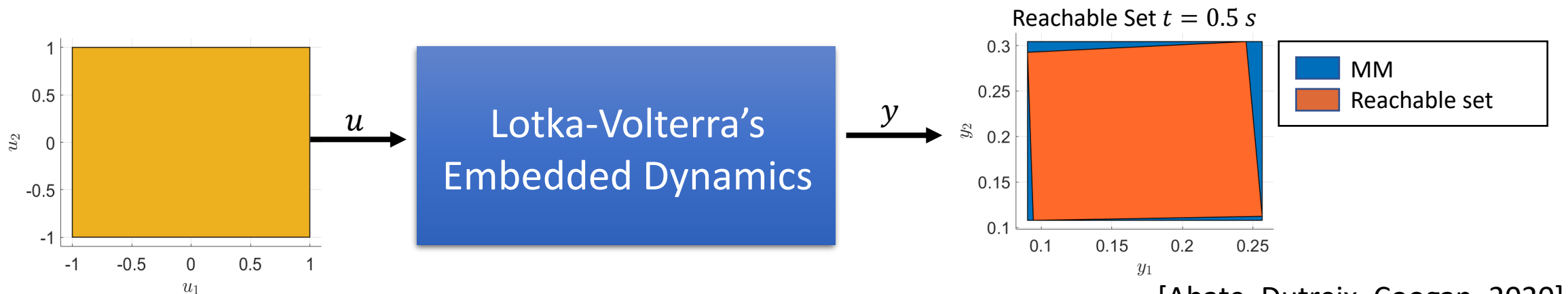
**Example:** Lotka-Volterra system

$$\dot{x}_1 = -x_1x_2 + x_1u_1,$$

$$\dot{x}_2 = x_1x_2 - x_2u_2,$$

$$y = x,$$

$$x(0)=(1/6,1/6), u_{1,2} \in [-1, 1].$$



[Abate, Dutreix, Coogan, 2020]

### 3. Input-Output Mixed-Monotonicity (IOMM)

Revisiting the objective:

- Extend mixed-monotonicity to Chen-Fliess series.
- Provide an overestimation of the reachable set of a Chen-Fliess series.

1

Define a partial order

$$u \preceq \hat{u}$$

2

Express the Chen-Fliess series as a difference of two positive series

$$F_c[u](t) = \mathcal{F}_{c^+}[u](t) - \mathcal{F}_{c^-}[u](t)$$

3

Obtain a decomposition function by lifting to a higher domain

$$d[u, \hat{u}] := \mathcal{F}_{c^+}[u](t) - \mathcal{F}_{c^-}[\hat{u}](t)$$

4

Use mixed-monotonicity

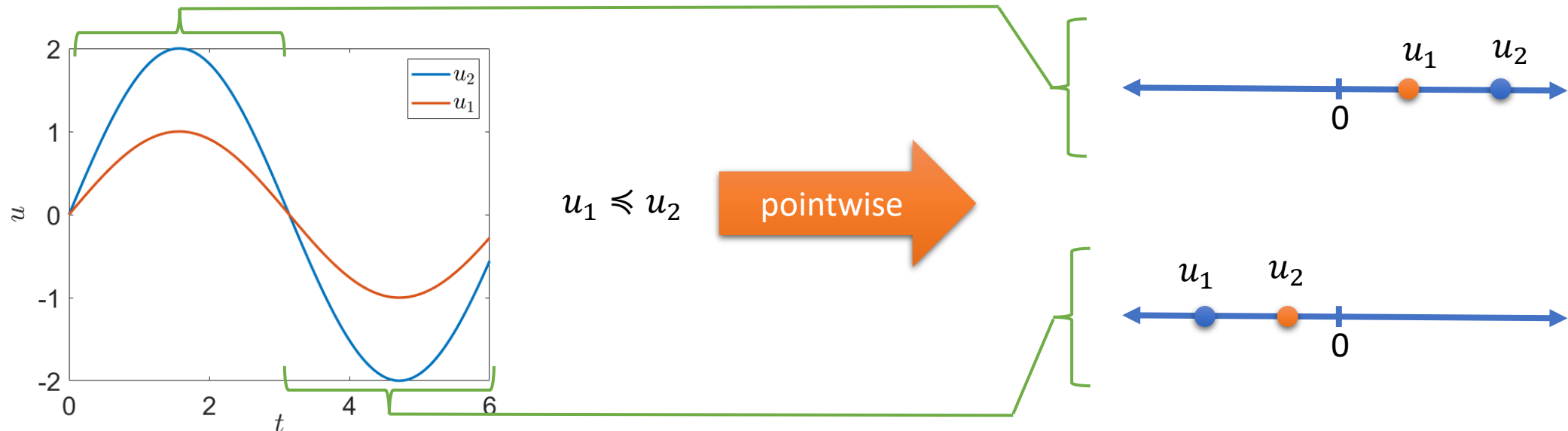
$$F_c[u](t) \in [d[\hat{u}, u](t), d[u, \hat{u}](t)]$$

### 3. Input-Output Mixed-Monotonicity (IOMM)

#### Definition 6 (Partial Order):

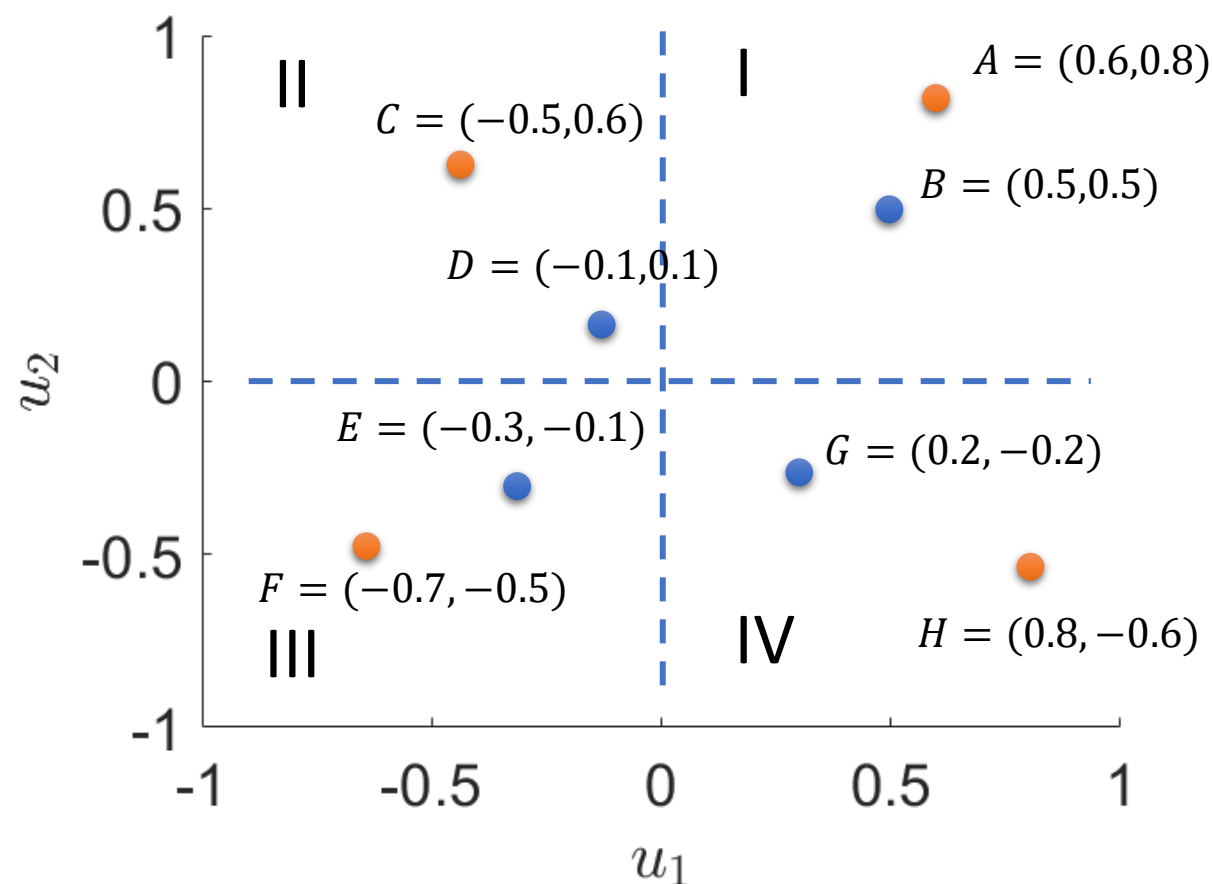
Consider the input functions  $u_1: \mathbb{R} \rightarrow \mathbb{R}^m$  and  $u_2: \mathbb{R} \rightarrow \mathbb{R}^m$ . These functions are ordered  $u_1(t) \preceq u_2(t)$  if and only if  $u_1^+(t) \leq u_2^+(t)$  and  $u_1^-(t) \leq u_2^-(t)$ , where  $\leq$  is the standard order of real numbers (componentwise) and  $u^+(t) = \max\{u(t), 0\}$ ,  $u^-(t) = -\min\{u(t), 0\}$ .

Example 1:



### 3. Input-Output Mixed-Monotonicity (IOMM)

Example 2:



I

$$B \preceq A$$

$$(0.5, 0.5) = B^+ \leq A^+ = (0.6, 0.8)$$

$$(0, 0) = B^- \leq A^- = (0, 0)$$

II

$$D \preceq C$$

$$(0, 0.1) = D^+ \leq C^+ = (0, 0.6)$$

$$(0.1, 0) = D^- \leq C^- = (0.5, 0)$$

III

$$E \preceq F$$

$$(0, 0) = E^+ \leq F^+ = (0, 0)$$

$$(0.3, 0.1) = E^- \leq F^- = (0.7, 0.5)$$

IV

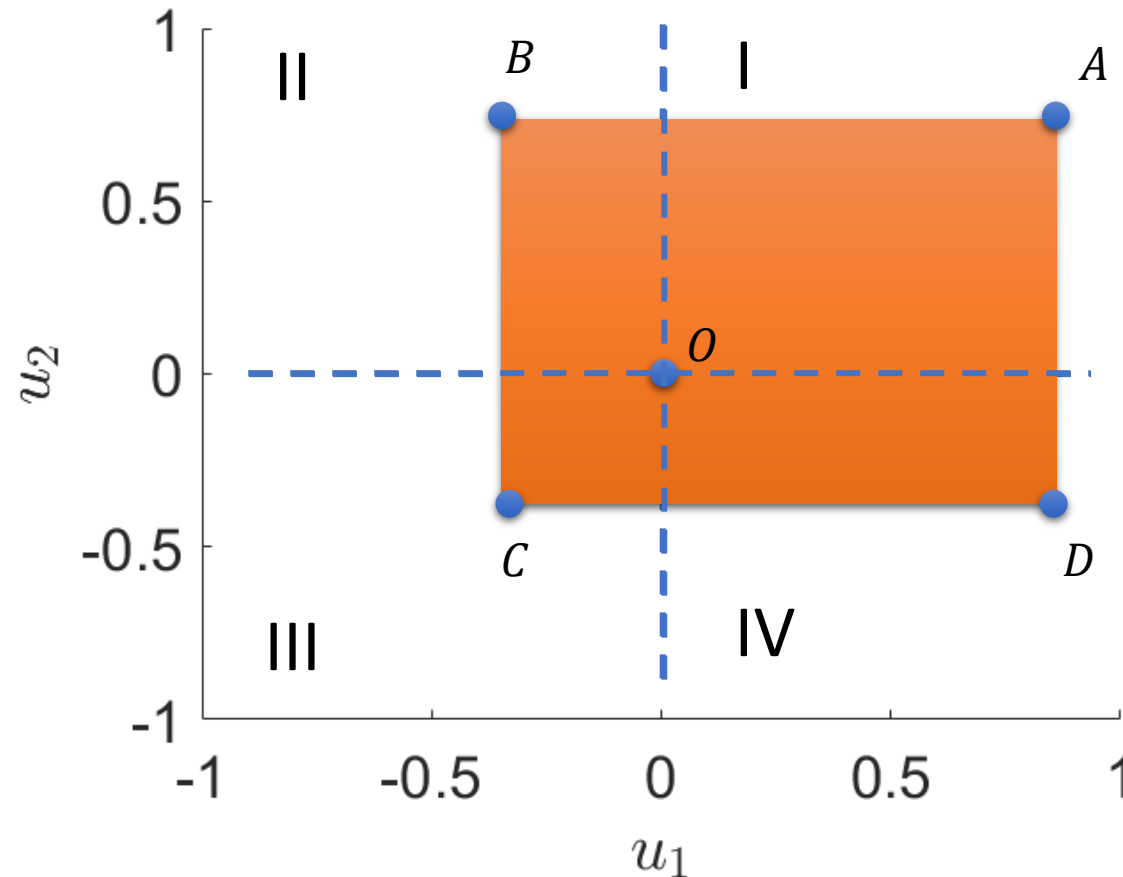
$$G \preceq H$$

$$(0, 0.2) = G^+ \leq H^+ = (0.8, 0)$$

$$(0, 0.2) = G^- \leq H^- = (0, 0.6)$$

### 3. Input-Output Mixed-Monotonicity (IOMM)

Example 3:



$$\{u \in \mathbb{R}^2 : C_x \leq u_1 \leq A_x, D_y \leq u_2 \leq B_y\}$$

=

$$[O, A] \cup [O, B] \cup [O, C] \cup [O, D]$$

where

$$[O, A] := \{u \in \mathbb{R}^2 : O \preceq u \preceq A\}$$

### 3. Input-Output Mixed-Monotonicity (IOMM)

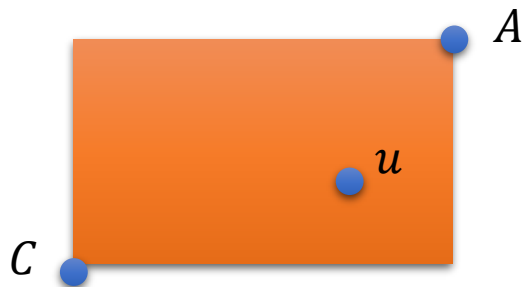
#### Definition 7 (IOMM):

A Chen-Fliess series  $F_c[u](t)$  is input-output mixed-monotone if there exists a decomposition function  $d_{F_c}[u, \hat{u}](t)$  such that

- i.  $d_{F_c}[u, u](t) = F_c[u](t)$ ,
- ii.  $d_{F_c}[u, \hat{u}](t)$  is non-decreasing in  $u$ ,
- iii.  $d_{F_c}[u, \hat{u}](t)$  is non-increasing in  $\hat{u}$ .

Overestimation:

Consider the input function  $u \in L_p^m[0, T]$  living in a square in the first orthant, this is  $C \leq u \leq A$ .



$$d_{F_c}[C, A](t) \leq F_c[u](t) = d_{F_c}[u, u](t) \leq d_{F_c}[A, C](t)$$

### 3. Input-Output Mixed-Monotonicity (IOMM)

#### Definition 8 (Extended iterative integral):

Consider the alphabets  $X$  and  $Y$  associated to the functions  $u, v \in L_p^m[0, T]$ , respectively

$$\mathcal{E}_{z_i \eta}[u, v](t) := \begin{cases} \int_0^t u_i(\tau) \mathcal{E}[u, v](\tau) d\tau, & z_i \in X \\ \int_0^t v_i(\tau) \mathcal{E}[u, v](\tau) d\tau, & z_i \in Y \end{cases}, \quad \mathcal{E}_\phi[u, v](t) := 1$$

#### Lemma 1:

Consider the alphabets  $X$  and  $Y$  associated to the functions  $u, v \in L_p^m[0, T]$ , respectively

$$F[u + v](t) = \sum_{k=0}^{\infty} \sum_{\xi \in \mathbb{S}_{X^*, Y^k}} (c, \sigma_x(\xi)) \mathcal{E}_\xi[u, v](t)$$

Substitution  
homomorphism from  
 $Z = X \cup Y$  to  $X$ :  
 $y_i \rightarrow x_i$



### 3. Input-Output Mixed-Monotonicity (IOMM)

#### Lemma 2:

Consider the input function  $u \in L_p^m[0, T]$  and its positive and negative parts  $u^+, u^-$  associated to the alphabets  $X$  and  $Y$ , respectively

$$\begin{aligned}
 F_c[u](t) &= \mathcal{F}_{c^+}[u](t) - \mathcal{F}_{c^-}[u](t) \\
 \mathcal{F}_{c^+}[u](t) &= \sum_{k=0}^{\infty} \sum_{\eta \in X^*} \sum_{\xi \in \mathbb{S}_{\eta, Y^k}} \overbrace{(c^+, \xi) \mathcal{E}_{\xi}[u^+, u^-](t)}^{\boxed{\max\{(-1)^k(c, \sigma_X(\xi)), 0\}}} , \\
 \mathcal{F}_{c^-}[u](t) &= \sum_{k=0}^{\infty} \sum_{\eta \in X^*} \sum_{\xi \in \mathbb{S}_{\eta, Y^k}} \overbrace{(c^-, \xi) \mathcal{E}_{\xi}[u^+, u^-](t)}^{\boxed{-\min\{(-1)^k(c, \sigma_X(\xi)), 0\}}} ,
 \end{aligned}$$

#### Theorem 1:

The following is a decomposition function of the Chen-Fliess series  $F_c[u](t)$

$$d_{F_c}[u, \hat{u}](t) := \mathcal{F}_{c^+}[u](t) - \mathcal{F}_{c^-}[\hat{u}](t)$$

### 3. Input-Output Mixed-Monotonicity (IOMM)

#### Theorem 2:

Consider the Chen-Fliess series  $F_c[u](t)$  taking values in the hyper-rectangle  $U = [u, \hat{u}] \subset K$ . Then

$$\text{Reach}(F_c, U)(t) \subset [d_{F_c}[u, \hat{u}](t), d_{F_c}[\hat{u}, u](t)].$$

# 4. Illustrative Simulation

## Example 4: Lotka-Volterra system

$$\dot{x}_1 = -x_1x_2 + x_1u_1,$$

$$\dot{x}_2 = x_1x_2 - x_2u_2,$$

$$y = x,$$

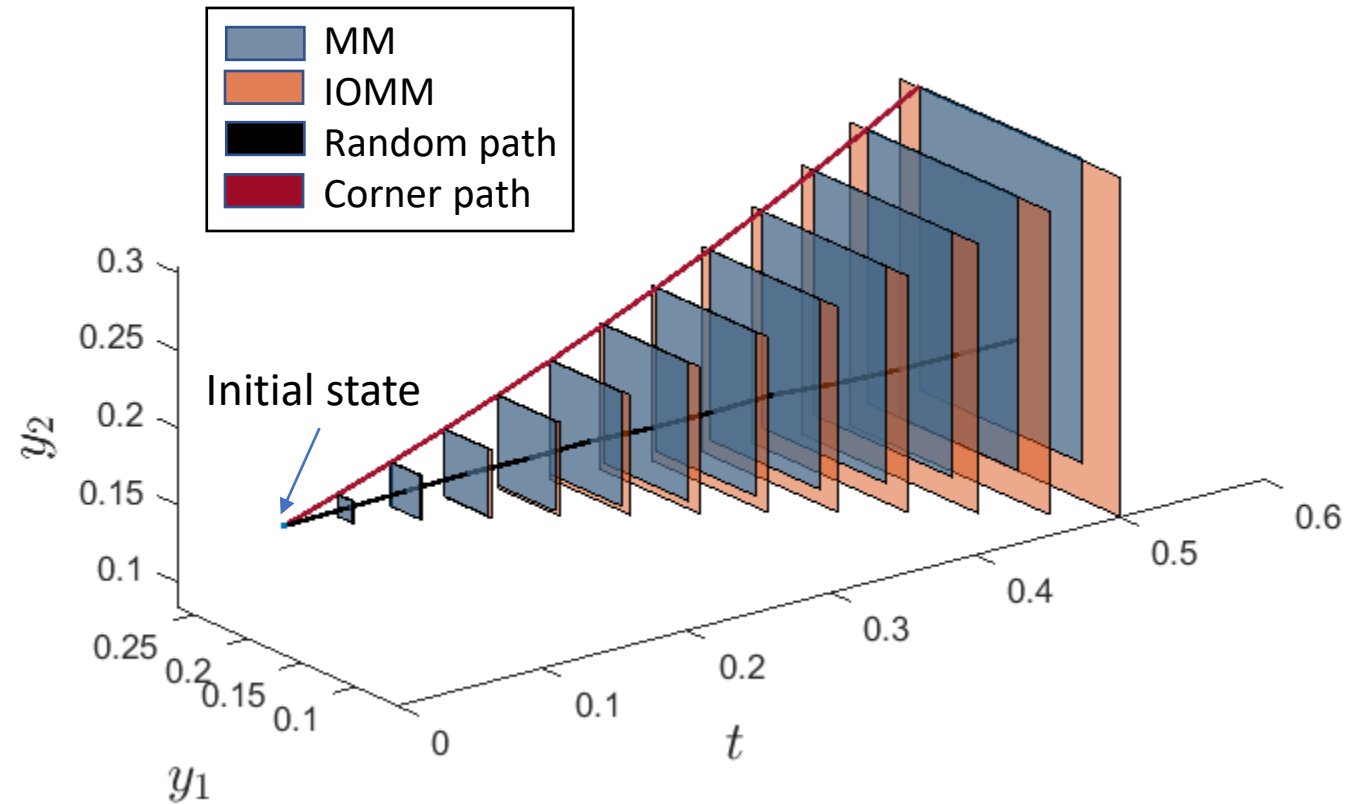
$$x(0) = (1/6, 1/6),$$

$$(u_1, u_2) \in [-1, 1] \times [-1, 1].$$

Positive and negative parts of the decomposition function:

$$\begin{aligned}\mathcal{F}_{c^+}[u](t) = & 0.167 + 0.028\mathcal{E}_{x_0}[u^+, u^-](t) \\ & + 0.167\mathcal{E}_{y_2}[u^+, u^-](t) + \dots\end{aligned}$$

$$\begin{aligned}\mathcal{F}_{c^-}[u](t) = & 0.167\mathcal{E}_{x_2}[u^+, u^-](t) \\ & + 0.028\mathcal{E}_{x_0x_2}[u^+, u^-](t) \\ & + 0.028\mathcal{E}_{x_0y_1}[u^+, u^-](t) + \dots\end{aligned}$$



**Remark:** IOMM always contains MM.

# 5. Conclusion

- We have provided a methodology to overestimate the reachable set of a system described as a Chen-Fliess series.
- The methodology provides a closed-form for the reachable set overestimation.
- The IOMM approach contains the sets that MM provides when the full Chen-Fliess series is considered.
- The accuracy of the method depends on the truncation of the Chen-Fliess series.
- Further work is needed to find the tightest overestimation.

# Questions?

Ivan.perez@uvm.edu