Interval trajectory tracking with flatness

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Context: CIPEGE¹

CIPEGE goal

Improve the employability of university students in Earth Sciences.

The CIPEGE tool

A decision making tool to satisfy qualitatively (level of study) and quantitavely (number of graduated) the labor market.

¹International prospective employment center in Earth sciences and environment (Centre international de prospective pour l'emploi en géosciences et en environnement)

Context

Discrete time nonlinear system:

$$\begin{cases} x(k+1) = f(x(k), u(k)), \ x(0) = x_0 \\ y(k) = h(x(k)) \end{cases}$$

avec

- $x(k) \in \mathbb{R}^n$ the state;
- $y(k) \in \mathbb{R}^p$ the output;
- $u(k) \in \mathbb{R}^m$ the input (control);
- x₀ the initial condition ;
- $f : \mathbb{R}^n \times \mathbb{R}^m \to \mathbb{R}^n$ and $h : \mathbb{R}^n \to \mathbb{R}^p$ are two analytic functions.

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Objective

Determine the input u(k) to track the *reference trajectory* $y_{ref}(k)$ of the output y(k) over a given prediction horizon.

Uncertain trajectory tracking

We consider uncertain trajectory

 $y_{\text{ref}}(k) \in [y_{\text{ref}}(k)]$



Issue

Goal

Characterize the set of admissible controls at each time k such that the ouput value remains in reference trajectory intervals.

$$\mathcal{U}^{*}(k) = \{u(k) | h(f(x(k), u(k))) \in [y_{ref}(k+1)]\}$$

Computing such set is generally intractable.

 \Rightarrow Computation of an inner approximation of the solution set.

Existing Approaches

2 Contribution





Existing Approaches

2 Contribution

3 SISO example



"Global numerical approach to nonlinear discrete-time control", IEEE Trans. on Automatic Control [Jaulin, Walter (1997)]

The result is a subpaving where all boxes belong to the characterized image set.

$$\mathcal{U}_{1..\ell} = \{(u(1), ..., u(\ell)) | \forall i \in [1, \ell], f(x(i), u(i)) \in [y_{\mathsf{ref}}(i+1)]\}$$

with y(k) = x(k).

Drawbacks

- Computation time : time complexity of the branch & Prune algorithm is in $O(e^{m \times \ell})$ with *m* the dimension of *u* and ℓ the prediction horizon.
- **Precision** : The precision is a user defined parameter according the stopping criterion:
 - minimum size of the considered intervals;
 - allocated time for computation.
- **Restriction** : the reference trajectory is defined over the entire (whole) state vector (y(k) = x(k)).

"Set computation for nonlinear control", Reliable Computing [Jaulin, Ratschan et Hardouin (2004)]

The result is a validated sequence of controls on a prediction horizon ℓ :

 $(\tilde{u}(1),...,\tilde{u}(\ell)) \in \mathcal{U}_{1..\ell} = \{(u(1),...,u(\ell)) | \forall i \in [1,\ell], f(x(i),u(i)) \in [y_{\mathsf{ref}}(i+1)]\}$

with y(k) = x(k).

Drawbacks

- **Computation time** : complexity is reduced since at each time *i* the control is fixed;
- Precision : we only get one sequence of controls;
- **Restriction** : the reference trajectory is defined over the entire (whole) state vector (y(k) = x(k)).

Existing Approaches

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3 SISO example



Objectives

- Restriction : the trajectory can be on a part of the components of the state vector;
- Computation time and precision : enhance time computation according to the precision.

Proposed method

Use of the flatness concept for dynamic systems.

Flatness (Fliess et al.)

Definition of x and u in function of a set of fundamental variables of the system: the *flat output*.

$$\begin{cases} x(k+1) = f(x(k), u(k)) \\ y(k) = h(x(k)). \end{cases}$$

This system is flat if there exist for a certain integer M, a function μ of x(k) and u(k) such that

$$F_k = \mu(x(k), u(k), \ldots, u(k+M)).$$

 F_k is called the flat output and verify the following relations for all k and for all J = n - 1:

$$\begin{aligned} x(k) &= \psi(F_k, F_{k+1}, \dots, F_{k+J}); \\ y(k) &= h \circ \psi(F_k, F_{k+1}, \dots, F_{k+J}); \\ u(k) &= \varphi(F_k, F_{k+1}, \dots, F_{k+J+1}). \end{aligned}$$

Computation of the set of admissible control for a flat system

Flatness gives

$$u(k) = \varphi(F_k, F_{k+1}, \ldots, F_{k+J+1})$$

If we consider that the reference trajectory gives the appropriate values for the flat ouput. The set to characterize at each time k is

$$\mathcal{U}^*(k) = \{\varphi(F_k, F_{k+1}, \dots, F_{k+J+1}) | F_k \in [y_{\mathsf{ref}}(k)] \, \forall k\}$$

This corresponds to the computation of the set

 $\{\mathcal{F}(x)|x\in[x]\}.$

Two distinctive cases

Computation of an inner approximation of the set

- $u \in \mathbb{R}$ (SISO) : generalized affine forms ;
- $u \in \mathbb{R}^m$, $m \ge 1$ (MIMO) : Branch & Prune algorithm.

SISO case

To define a generalized affine form, one need to:

- compute $[\Delta_i]$, an outer approximation of the Jacobian of f over a box;
- compute f^ε(t₁,..., t_n) for a given (t₁,..., t_n) ⇒ corresponding to the center by taking t = 0.

Generalized affine set[1]

A generalized affine set of a function is a triplet (Z, c, J) with

- $Z \in \mathcal{M}(n+m+1,p)$ a vector of affine forms;
- $c \in \mathbb{R}^p$ a vector corresponding to the centre;
- $J \in (\mathcal{M}(n,p))^n$ a matrix of affine forms.

Combining affine arithmetics and Kaucher arithmetics. The generalized affine forms allow to produce an interval $[y] \subseteq \{f(x)|x \in [x]\}$.

 E. Goubault, O. Mullier, S. Putot, and M. Kieffer. Inner approximated reachability analysis. In *Hybrid Systems: Computation and Control*, pages 163–172, New York, NY, USA, 2014. ACM.
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MIMO case

Employed method

- Extension of the results from Jaulin-Goldsztejn to the case $f : \mathbb{R}^n \to \mathbb{R}^m$;
- Computation of a subpaving of the set $\{f(x)|x \in [x]\}$;
- Branch & Prune algorithm + sufficient condition for a box [y] to belong to $\{f(x)|x \in [x]\}$.

Existing Approaches

2 Contribution





SISO Example

$$\begin{cases} x_1(k+1) = x_1(k)x_2(k) \\ x_2(k+1) = x_2(k) + u(k) \end{cases}$$

with $x_0 = (1 \ 1)^T$.

We have computed the flat output F_k and the expression of u(k) according to the flat output:

$$u(k) = \frac{F_{k+2}}{F_{k+1}} - \frac{F_{k+1}}{F_k}$$

and we know the reference trajectory

$$(y_{ref}(k))_{k=1..3} = (F_1, F_2, F_3)$$

= $(1, [1, 2], [5, 7]).$

We apply the both presented methods to compute u(0) and u(1).

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- Method 1 : generalized affine forms.
- Method 2 : intervals.

Method 1 (Generalized affine forms)

$$(F_1, F_2, F_3) = (1, [1, 2], [5, 7])$$

 $u(k) = \frac{F_{k+2}}{F_{k+1}} - \frac{F_{k+1}}{F_k}$

We compute the affine forms corresponding to F_i :

$$\hat{F}_{0} = 1$$

 $\hat{F}_{1} = 1$
 $\hat{F}_{2} = \frac{3}{2} + \frac{1}{2}\varepsilon_{2}$
 $\hat{F}_{3} = 6 + \varepsilon_{3}$

Method 1 (Generalized affine forms)

The computation gives:

• for *u*(0)

$$\check{u}_{0} = \begin{pmatrix} \hat{u}_{0} &= \frac{1}{2} + \frac{1}{2}\varepsilon_{2} \\ c &= \frac{1}{2} \\ [J] &= \begin{pmatrix} 0 & \frac{1}{2} & 0 \\ \hat{J} &= \begin{pmatrix} 0 & \frac{1}{2} & 0 \end{pmatrix} \\ \hat{J} &= \begin{pmatrix} 0 & \frac{1}{2} & 0 \end{pmatrix} \end{pmatrix}$$

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• for *u*(1)

$$\check{u}_{1} = \begin{pmatrix} \hat{u}_{1} &= 3 - \frac{5}{4}\varepsilon_{2} + \frac{3}{4}\varepsilon_{3} + \eta \\ c &= \frac{5}{2} \\ [J] &= \left(0 \ \left[-4, -\frac{9}{8}\right] \ \left[\frac{1}{2}, 1\right]\right) \\ \hat{J} &= \left(\left(0\right) \ \left(-\frac{23}{8} + 0.28125\varepsilon_{2} - 0.395833\varepsilon_{3} + 1.61458\eta\right) \\ & \left(0.75 - 0.125\varepsilon_{2} + 0.125\eta\right)\right) \end{cases}$$

Method 2 (bisection and intervals)

The bisection method requires to know the expression of the function and its derivatives:

$$u(k) = \frac{F_{k+2}}{F_{k+1}} - \frac{F_{k+1}}{F_k}$$
$$J_u = \left(\begin{array}{cc} \frac{F_{k+1}}{F_k^2} & -\frac{F_{k+2}}{F_{k+1}^2} - \frac{1}{F_k} & \frac{1}{F_{k+1}} \end{array}\right)$$

Results : discussion

Stopping criterion for method 2 and (Jaulin et al.) : minimum size for an interval to be treated is 0.1.

Method	Method 1	Method 2	(Jaulin et al.)
Results	[0, 1]	[0, 1]	[0, 0.9765625]
Computation time	$1.37672 imes 10^{-6}$ s	0.0002s	0.37935s ²

Table: Results for u(0).

Method	Method 1	Method 2	(Jaulin et al.)
Results	[0.875, 4.125]	[0.59375, 5.6875]	[0.8789, 5.2734]
Computation time	$2.86311\times10^{-6}\mathrm{s}$	0.013656s	$0.37935s^{1}$

Table: Results for u(1).

²time to compute u(1) and u(2) together

Existing Approaches

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Contribution

New method to characterize the set of admissible controls

- enhances the result according to the existing approach in term of precision and computation time;
- combination of set-membership methods and flatness of dynamic systems;
- handle a larger class of problems $(y(k) \neq x(k))$.

Contribution

New method to characterize the set of admissible controls

- enhances the result according to the existing approach in term of precision and computation time;
- combination of set-membership methods and flatness of dynamic systems;
- handle a larger class of problems $(y(k) \neq x(k))$.

Perspectives

Extend the results to the class of continuous-time systems:

- control *u* is a function of the flat output and its successive time derivatives;
- use of validated integration methods.