

Observability of Nonlinear Systems Using Interval Analysis

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Motivation

Why deal with observability?

- To assure the functionality of an observer first the observability has to be proven
- Many applications have a finite state space, therefore only observability for this finite state space is of interest
- In the case of a system which is not observable, it can be of interest which states can not be distinguished of each other

Nonlinear autonomous Systems

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

$$y(t) = h(x(t))$$

with $f : \mathbb{R}^n \rightarrow \mathbb{R}^n$ and $h : \mathbb{R}^n \rightarrow \mathbb{R}^m$ as real-analytic functions

Problem formulation

Lie derivatives

$$y(t) = \sum_{k \geq 0} \frac{t^k}{k!} (L_f^k h)(x_0)$$

$$L_f^k h = \frac{\partial}{\partial x} (L_f^{k-1} h) f \text{ and } L_f^0 h = h$$

Distinguishability

$$z^1 \text{ and } z^2 \text{ are indistinguishable} \Leftrightarrow y^1(t) \equiv y^2(t)$$

$$\Leftrightarrow (L_f^k h)(z^1) = (L_f^k h)(z^2), k \geq 0$$

$$\mathcal{M} = \left\{ \begin{pmatrix} z^1 \\ z^2 \end{pmatrix} \mid (L_f^k h)(z^1) = (L_f^k h)(z^2) \right\} / \left\{ \begin{pmatrix} z^1 \\ z^2 \end{pmatrix} \mid z^1 = z^2 \right\} = \emptyset$$

Problem formulation

Observability function

$$q(x) = \begin{bmatrix} L_f^0 h(x) \\ \vdots \\ L_f^\kappa h(x) \end{bmatrix}$$

with κ as the necessary amount of Lie derivatives.

z^1 and z^2 are indistinguishable $\Leftrightarrow q(z^1) = q(z^2)$

Local observability

$$\text{rank} \left(\frac{\partial q(x)}{\partial x} \right) = n$$

Lie derivatives symbolically

Example nonlinear system

$$\begin{aligned} f(x) &= \begin{pmatrix} -x_2 + x_1(1 - x_1^2 - x_2^2) \\ x_1 + x_2(1 - x_1^2 - x_2^2) \end{pmatrix} \\ h(x) &= x_1 + x_2 \end{aligned}$$

Lie derivative $L_f^4 h(x)$

$$\begin{aligned}
L_f^4 h(x) = & (x_2 + x_1(x_1^2 + x_2^2 - 1))((x_2 + x_1(x_1^2 + x_2^2 - 1))(6x_2 - 2x_1 + 2(2x_1 + 2x_2)(2x_1x_2 - 1) + 2x_2(2x_1x_2 + x_1^2 + 3x_2^2) + 6x_1(2x_1x_2 + 3x_1^2 + x_2^2 - 2) \\
& + 2(6x_1 + 2x_2)(3x_1^2 + x_2^2 - 1) + 6x_1(x_1^2 + x_2^2 - 1) + 2x_2(x_1^2 + x_2^2 - 1)) - (x_1 - x_2(x_1^2 + x_2^2 - 1))(2x_2 - 2x_1 + (2x_1 + 6x_2)(2x_1x_2 - 1) + (6x_1 + 2x_2)(2x_1x_2 + 1) \\
& + 2x_1(2x_1x_2 + x_1^2 + 3x_2^2) + 2x_2(2x_1x_2 + 3x_1^2 + x_2^2 - 2) + (2x_1 + 2x_2)(3x_1^2 + x_2^2 - 1) + (2x_1 + 2x_2)(x_1^2 + 3x_2^2 - 1) + 2x_1(x_1^2 + x_2^2 - 1) + 2x_2(x_1^2 + x_2^2 - 1)) \\
& + (2x_1x_2 - 1)((2x_1 + 2x_2)(x_2 + x_1(x_1^2 + x_2^2 - 1)) - (2x_1 + 6x_2)(x_1 - x_2(x_1^2 + x_2^2 - 1)) + (x_1^2 + 3x_2^2 - 1)(2x_1x_2 + x_1^2 + 3x_2^2) + (2x_1x_2 + 1)(2x_1x_2 + 3x_1^2 + x_2^2 - 2)) \\
& + (3x_1^2 + x_2^2 - 1)((6x_1 + 2x_2)(x_2 + x_1(x_1^2 + x_2^2 - 1)) - (2x_1 + 2x_2)(x_1 - x_2(x_1^2 + x_2^2 - 1)) + (3x_1^2 + x_2^2 - 1)(2x_1x_2 + 3x_1^2 + x_2^2 - 2) + (2x_1x_2 - 1)(2x_1x_2 + x_1^2 + 3x_2^2)) \\
& - (x_1 - x_2(x_1^2 + x_2^2 - 1))((x_2 + x_1(x_1^2 + x_2^2 - 1))(2x_2 - 2x_1 + (2x_1 + 6x_2)(2x_1x_2 - 1) + (6x_1 + 2x_2)(2x_1x_2 + 1) + 2x_1(2x_1x_2 + x_1^2 + 3x_2^2) + 2x_2(2x_1x_2 + 3x_1^2 + x_2^2 - 2) \\
& + (2x_1 + 2x_2)(3x_1^2 + x_2^2 - 1) + (2x_1 + 2x_2)(x_1^2 + 3x_2^2 - 1) + 2x_1(x_1^2 + x_2^2 - 1) + 2x_2(x_1^2 + x_2^2 - 1)) - (x_1 - x_2(x_1^2 + x_2^2 - 1))(2x_2 - 6x_1 + 2(2x_1 + 2x_2)(2x_1x_2 + 1) \\
& + 6x_2(2x_1x_2 + x_1^2 + 3x_2^2) + 2x_1(2x_1x_2 + 3x_1^2 + x_2^2 - 2) + 2(2x_1 + 6x_2)(x_1^2 + 3x_2^2 - 1) + 2x_1(x_1^2 + x_2^2 - 1) + 6x_2(x_1^2 + x_2^2 - 1)) + (2x_1x_2 + 1)((6x_1 + 2x_2)(x_2 \\
& + x_1(x_1^2 + x_2^2 - 1)) - (2x_1 + 2x_2)(x_1 - x_2(x_1^2 + x_2^2 - 1)) + (3x_1^2 + x_2^2 - 1)(2x_1x_2 + 3x_1^2 + x_2^2 - 2) + (2x_1x_2 - 1)(2x_1x_2 + x_1^2 + 3x_2^2)) + (x_1^2 + 3x_2^2 - 1)((2x_1 + 2x_2) \\
& (x_2 + x_1(x_1^2 + x_2^2 - 1)) - (2x_1 + 6x_2)(x_1 - x_2(x_1^2 + x_2^2 - 1)) + (x_1^2 + 3x_2^2 - 1)(2x_1x_2 + x_1^2 + 3x_2^2) + (2x_1x_2 + 1)(2x_1x_2 + 3x_1^2 + x_2^2 - 2)))
\end{aligned}$$

Using power series for Lie derivatives

Power series

$$a(t) := \sum_{k=0}^{\infty} a_k t^k$$

| | |
|------------------|--|
| $(a \pm b)(t)$ | $\left \sum_{k=0}^{\infty} (a_k \pm b_k) t^k \right.$ |
| $(a \cdot b)(t)$ | $\left \sum_{k=0}^{\infty} \left(\sum_{l=0}^k a_l \cdot b_{k-l} \right) t^k \right.$ |
| $(a : b)(t)$ | $\left \sum_{k=0}^{\infty} \left(\frac{a_k}{b_0} - \frac{1}{b_0} \sum_{l=0}^k c_l \cdot b_{k-l} \right) t^k \right.$ |

Using power series for Lie derivatives

Computation of $L_f^0 h(x_0)$

$$x_0 \text{ given} \quad L_f^0 h(x_0) = 0! h_0(x_0)$$

Computation of $L_f^1 h(x_0)$

$$x_1 = \frac{1}{1+1} f_0(x_0) \quad L_f^1 h(x_0) = 1! h_1(x_0, x_1)$$

Computation of $L_f^2 h(x_0)$

$$x_2 = \frac{1}{2+1} f_1(x_0, x_1) \quad L_f^2 h(x_0) = 2! h_2(x_0, x_1, x_2)$$

Computation of k Lie derivatives

$$x_{k+1} = \frac{1}{k+1} f_k \text{ with } f_k = f_k(x_0, \dots, x_k) \text{ and } f(x(t)) = \sum_{k \geq 0} f_k t^k$$

$$L_f^k h(x_0) = k! h_k(x_0, \dots, x_k)$$

Notation

Notations in this talk

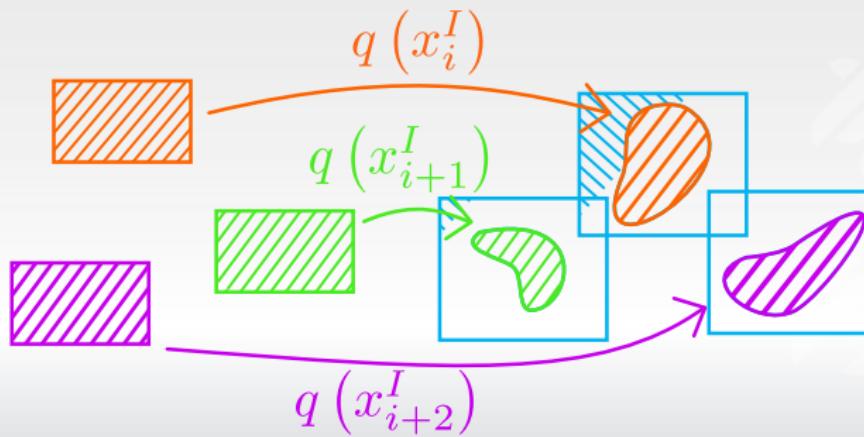
Interval $x^I = [\underline{x}, \bar{x}]$

Interval midpoint $\check{x} = \frac{\underline{x} + \bar{x}}{2}$

Interval radius $\hat{x} = \frac{\bar{x} - \underline{x}}{2}$

Hull of intervals $\tilde{x}^I = \bigcup x_i^I$ with $i = 1, \dots, n$

List structure



Double layered list structure

$$\mathcal{L}_i = \{x_i^I \mid \Theta^I(x_i^I)\}, \quad \Theta^I(x_i^I) = x_{i+1}^I, x_{i+2}^I$$

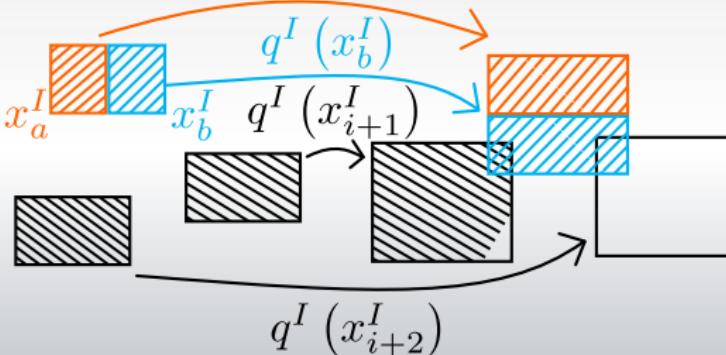
$$\mathcal{L}_{i+1} = \{x_{i+1}^I \mid \Theta^I(x_{i+1}^I)\}, \quad \Theta^I(x_{i+1}^I) = x_i^I$$

$$\mathcal{R} = \{\mathcal{L}_i, \mathcal{L}_{i+1}, \mathcal{L}_{i+2}, \dots\}$$

Distinguish the mapping $q^I(x^I)$

$$\mathcal{L}_i = \{x_i^I \mid x_{i+1}^I, x_{i+2}^I\}$$

$$q^I(x_a^I)$$



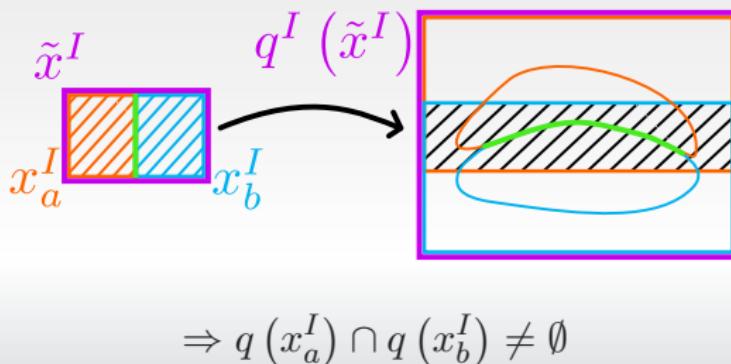
Distinguishable if $q(x_1^I) \cap q(x_2^I) = \emptyset$

$$\mathcal{L}_a = \{x_a^I \mid x_b^I\}$$

$$\mathcal{L}_b = \{x_b^I \mid x_a^I, x_{i+1}^I, x_{i+2}^I\}$$

- 1: take first \mathcal{L}_i out of \mathcal{R}
- 2: bisect $\mathcal{L}_i \ni x_i^I = x_a^I \cup x_b^I$ and remove x_i^I entirely from \mathcal{R}
- 3: calculate $q(x_a^I)$ and $q(x_b^I)$
- 4: test Distinguishability and create \mathcal{L}_a and \mathcal{L}_b
- 5: **for** $j = a, b$ **do**
- 6: get hull \tilde{x}_j^I for \mathcal{L}_j
- 7: **if** \tilde{x}_j^I local observable **then**
- 8: remove x_j^I from \mathcal{R}
- 9: **else**
- 10: put \mathcal{L}_j at the end of \mathcal{R}
- 11: **end if**
- 12: **end for**
- 13: change κ if necessary

Problem of the Distinguishability



Therefore we need the local observability criterion

$$\text{rank} \left(\frac{\partial q^I(\tilde{x}^I)}{\partial x} \right) = n$$

- 1: take first \mathcal{L}_i out of \mathcal{R}
- 2: bisect $\mathcal{L}_i \ni x_i^I = x_a^I \cup x_b^I$ and remove x_i^I entirely from \mathcal{R}
- 3: calculate $q(x_a^I)$ and $q(x_b^I)$
- 4: test Distinguishability and create \mathcal{L}_a and \mathcal{L}_b
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- 7: **if** \tilde{x}_j^I local observable **then**
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- 11: **end if**
- 12: **end for**
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Local observability criterion

With J^I been the Jacobian of $q^I(\tilde{x}^I)$

$$V^I := \begin{pmatrix} 0 & J^I \\ J^{I\top} & 0 \end{pmatrix}$$

Theorem of Rohn [98]:

$$\underline{\lambda}_i^I(V^I) = \lambda_i(\check{V}) - \rho(\hat{V})$$

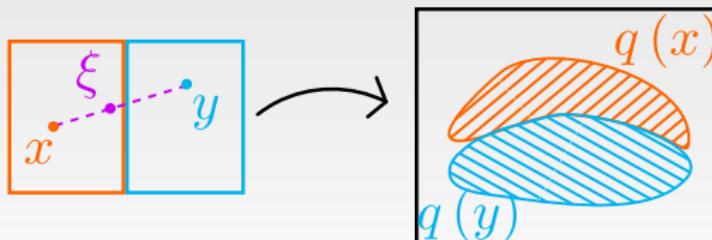
$$\bar{\lambda}_i^I(V^I) = \lambda_i(\check{V}) + \rho(\hat{V})$$

$\rho(\hat{V})$: spectral radius of \hat{V}

If $2n \lambda_i^I \not\geq 0 \Rightarrow V^I$ is of full rank and therefore J^I is of full rank

- 1: take first \mathcal{L}_i out of \mathcal{R}
- 2: bisect $\mathcal{L}_i \ni x_i^I = x_a^I \cup x_b^I$ and remove x_i^I entirely from \mathcal{R}
- 3: calculate $q(x_a^I)$ and $q(x_b^I)$
- 4: test Distinguishability and create \mathcal{L}_a and \mathcal{L}_b
- 5: **for** $j = a, b$ **do**
- 6: get hull \tilde{x}_j^I for \mathcal{L}_j
- 7: **if** \tilde{x}_j^I local observable **then**
- 8: remove x_j^I from \mathcal{R}
- 9: **else**
- 10: put \mathcal{L}_j at the end of \mathcal{R}
- 11: **end if**
- 12: **end for**
- 13: change κ if necessary

Local observability criterion



$$q_i(x) - q_i(y) = \left(\frac{\partial q_i(\xi_i)}{\partial x} \right)^T (x - y)$$

$$q(x) - q(y) = \underbrace{\begin{bmatrix} \left(\frac{\partial q_1(\xi_1)}{\partial x} \right)^T \\ \left(\frac{\partial q_2(\xi_2)}{\partial x} \right)^T \\ \vdots \end{bmatrix}}_M (x - y)$$

$$M \in M^I = \frac{\partial q^I(x^I)}{\partial x}$$

- 1: take first \mathcal{L}_i out of \mathcal{R}
- 2: bisect $\mathcal{L}_i \ni x_i^I = x_a^I \cup x_b^I$ and remove x_i^I entirely from \mathcal{R}
- 3: calculate $q(x_a^I)$ and $q(x_b^I)$
- 4: test Distinguishability and create \mathcal{L}_a and \mathcal{L}_b
- 5: **for** $j = a, b$ **do**
- 6: get hull \tilde{x}_j^I for \mathcal{L}_j
- 7: **if** \tilde{x}_j^I local observable **then**
- 8: remove x_j^I from \mathcal{R}
- 9: **else**
- 10: put \mathcal{L}_j at the end of \mathcal{R}
- 11: **end if**
- 12: **end for**
- 13: change κ if necessary

Determination the amount of Lie derivatives

- Since $h : \mathbb{R}^n \rightarrow \mathbb{R}^m$ every further Lie derivative adds m equations to $q(x)$
- Necessary amount of Lie derivatives vary due to the observability singularities
- At least n equations are mandatory
- Always computing $\kappa + 1$ Lie derivatives
- Test Distinguishability and local observability for $\kappa - 1$, κ and $\kappa + 1$

```
1: take first  $\mathcal{L}_i$  out of  $\mathcal{R}$ 
2: bisect  $\mathcal{L}_i \ni x_i^I = x_a^I \cup x_b^I$  and
   remove  $x_i^I$  entirely from  $\mathcal{R}$ 
3: calculate  $q(x_a^I)$  and  $q(x_b^I)$ 
4: test Distinguishability
   and create  $\mathcal{L}_a$  and  $\mathcal{L}_b$ 
5: for  $j = a, b$  do
6:   get hull  $\tilde{x}_j^I$  for  $\mathcal{L}_j$ 
7:   if  $\tilde{x}_j^I$  local observable then
8:     remove  $x_j^I$  from  $\mathcal{R}$ 
9:   else
10:    put  $\mathcal{L}_j$  at the end of  $\mathcal{R}$ 
11:   end if
12: end for
13: change  $\kappa$  if necessary
```

Nonlinear example

Nonlinear state space representation

$$f(x) = \begin{pmatrix} x_2 \\ -x_1 \end{pmatrix}$$

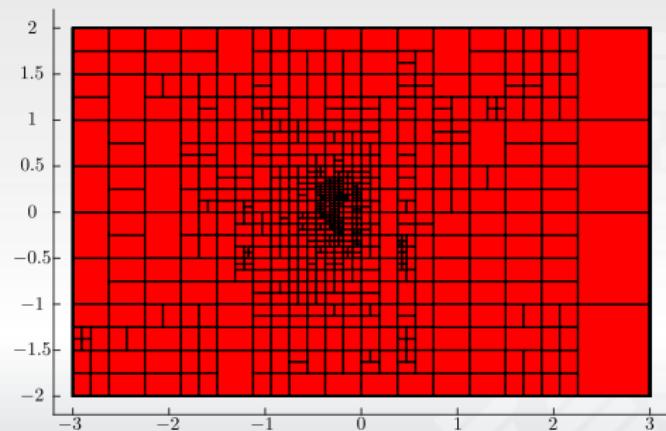
$$h(x) = x_1^2 + x_1$$

$$L_f^0 h(x) = x_1^2 + x_1$$

$$L_f^1 h(x) = 2x_1 x_2 + x_2$$

Solutions for $L_f^0 h(x) = L_f^0 h(z)$ and $L_f^1 h(x) = L_f^1 h(z)$ are

$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$ is indistinguishable from $\begin{pmatrix} z_1 \\ z_2 \end{pmatrix}$ and $\begin{pmatrix} -z_1 - 1 \\ -z_2 \end{pmatrix}$



Further Lie derivatives are necessary to exclude the second solution

$$L_f^2 h(x) = 2x_2^2 - 2x_1^2 - x_1$$

$$L_f^3 h(x) = -8x_1 x_2 - x_2$$

What was shown

- Numeric approach to determine whether a nonlinear system is observable
- The presented algorithm can handle nonlinear systems with real-analytic functions
- The presented algorithm can adjust the necessary amount of Lie derivatives dynamically
- In case of a not observable system the algorithm is capable of finding the states that can not be distinguished of each other

What we currently work on

- Reduction of wrapping effect
- Improving the runtime further by parallelizing the algorithm

Thank you for your attention
Any questions?