

Switch observability: A novel approach towards fault detection

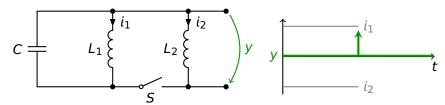
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Motivational example



Switch		obsv.
open	$y \equiv 0$ for arbitrary internal state	X
closed	equilibrium $i_1 = -i_2 = \text{const} \rightarrow y \equiv 0$	X
closing	$y = 0$ jumps to $\neq 0$	1
opening	non-equilibrium: $y \neq 0$ jumps to zero (+ lmp.)	1
	equilibrium: $y(t) = 0 \ \forall t$, but with impulse in y	✓

Transition "open \rightarrow close" ($y \not\equiv 0$ on (t_S , $t_S + \varepsilon$)) distinguishable from transition "close \rightarrow open" ($y \equiv 0$ on (t_S , $t_S + \varepsilon$))

Discussion of example

Circuit is modelled by a switched differential-algebraic equation (DAE):

$$E_{\sigma}\dot{x} = A_{\sigma}x(+B_{\sigma}u)$$
$$y = C_{\sigma}x$$

$$\sigma:\mathbb{R} \to \{1,\dots,P\}$$
 is the switching signal

Nonobservability on switch-free intervals

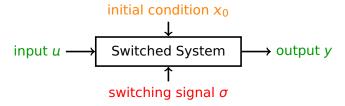
Using measurements only from switch-free intervals:

- Mode (i.e. switch position) cannot be recovered for some $x_0 \neq 0$
- > Each individual mode is not state-observable

Observability around switch

- > Modes before and after the switch can be recovered
- > Internal states can completely be recovered
- > Dirac impulses in output needed for observability

The observability problem

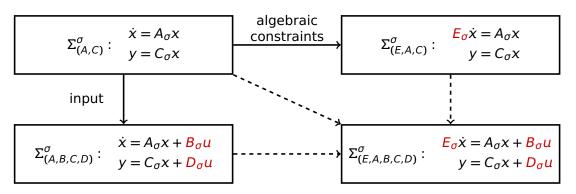


Observability questions

- > Is there a unique x_0 for any given σ , u, y? \rightarrow (t.v.) observability \checkmark
- \rightarrow Is there a unique (x_0, σ) for any given u and y?
 - \rightarrow (x, σ)-observability
- > Is there a unique σ for any given u, y and unknown x_0 ?
 - $\rightarrow \sigma$ -observability = fault detectability (+isolation)
- \rightarrow Is there a unique set $\{t_5\}$ of switching times for any u, y?
 - → ts-observability = fault detectability

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System classes



Future work: Nonlinear versions thereof ...

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Introduction

$$\dot{x} = A_{\sigma}x$$

$$\dot{x} = A_{\sigma}x + B_{\sigma}u$$

$$E_{\sigma}\dot{x} = A_{\sigma}x + B_{\sigma}u$$

Observer design

Summary

The simplest system class $\Sigma_{(A.C)}^{\sigma}$: $\begin{vmatrix} \dot{x} = A_{\sigma}x \\ y = C_{\sigma}x \end{vmatrix}$

$$\dot{x} = A_{\sigma} x
y = C_{\sigma} x$$

Formal Definition: (x, σ) -/ σ -Observability

$$\Sigma_{(A,C)}^{\sigma}(\mathbf{x}, \boldsymbol{\sigma})$$
-observable : $\Leftrightarrow \forall \sigma, \widehat{\sigma} \quad \forall \text{ sol. } x, \widehat{x} \text{ with } (x, \widehat{x}) \neq (0,0)$:

$$(x, \sigma) \neq (\widehat{x}, \widehat{\sigma}) \implies y \neq \widehat{y}$$

$$\Sigma_{(A,C)}^{\sigma}$$
 σ -observable : $\Leftrightarrow \forall \sigma, \widehat{\sigma} \quad \forall \text{ sol. } x, \widehat{x} \text{ with } (x, \widehat{x}) \neq (0, 0)$:

$$\sigma \neq \widehat{\sigma} \implies y \neq \widehat{y}$$

First (surprising?) result for $\Sigma_{(A,C)}^{\sigma}$

$$(x, \sigma)$$
-observability

 $\iff \sigma$ -observability

State-observability of each mode

In the context auf fault detection/isolation we have:

```
fault detection and isolation
          mode detectability
            \sigma-observability
          (x, \sigma)-observability
(x, \sigma)-observability with constant \sigma
(A_i, C_i) is observable for each mode i
```

Assuming (state-)observability for all faulty modes is not realistic.

Summary

Weaker observability notion

$$\dot{x} = \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} x$$

$$y = C_{\sigma} x$$
with
$$C_1 = \begin{bmatrix} 1, 0 \end{bmatrix} \\
C_2 = \begin{bmatrix} 0, 1 \end{bmatrix} \rightarrow \text{ not observable}$$

$$mode 1 \quad t_{\sigma} \quad mode 2$$

Switch observability ((x, σ_1) -/ σ_1 -observability)

Recover (x and) σ from u and y, if at least one switch occurs

Again: σ_1 -observability \iff (x, σ_1) -observability

Obs. characterizations for $\Sigma_{(A,C)}^{\sigma}$: $\begin{vmatrix} \dot{x} = A_{\sigma}x \\ y = C_{\sigma}x \end{vmatrix}$

$$\dot{x} = A_{\sigma} x \\ y = C_{\sigma} x$$

Kalman observability matrix of mode
$$k$$
: $\mathcal{O}_k := \begin{bmatrix} C_k \\ C_k A_k \\ C_k A_k^2 \\ \vdots \end{bmatrix}$

Theorem (cf. Küsters & Trenn, Automatica 2018)

$$\sigma$$
-observability $\iff \forall i \neq j : \operatorname{rank}[\mathcal{O}_i \mathcal{O}_j] = 2n$

$$\sigma_1$$
-observability $\iff \forall i \neq j, p \neq q, (i, j) \neq (p, q) : rank $\begin{bmatrix} \mathcal{O}_i & \mathcal{O}_p \\ \mathcal{O}_j & \mathcal{O}_q \end{bmatrix} = 2n$$

$$t_{S}$$
-observability $\iff \forall i \neq j : rank[\mathcal{O}_{i} - \mathcal{O}_{i}] = n$

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Summary

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Observer design

Summary

Adding inputs

$$\dot{x} = A_{\sigma}x + B_{\sigma}u$$
$$y = C_{\sigma}x + D_{\sigma}u$$

Input-depending observability

 $\Sigma(A_{\sigma}, C_{\sigma}) \sigma$ -observable $\Leftrightarrow \Sigma(A_{\sigma}, B_{\sigma}, C_{\sigma}, D_{\sigma}) \sigma$ -observable

Strong vs. weak observability

observable for all $u \Leftrightarrow$ observable for some/almost all u

Further technicalities

Analytic vs. smooth inputs and equivalent switching signals

Strong obs. for $\Sigma^{\sigma}_{(A,B,C,D)}$: $\begin{vmatrix} \dot{x} = A_{\sigma}x + B_{\sigma}u \\ y = C_{\sigma}x + D_{\sigma}u \end{vmatrix}$

$$\dot{x} = A_{\sigma}x + B_{\sigma}u$$
$$y = C_{\sigma}x + D_{\sigma}u$$

Definition

$$\Sigma^{\sigma}_{(A,B,C,D)}$$
 is strongly (x,σ) -/ σ -/ (x,σ_1) - $/\sigma_1$ -/ t_S -observable : \Leftrightarrow $\forall u$: $\Sigma^{\sigma}_{(A,B,C,D)}$ is (x,σ) -/ σ -/ (x,σ_1) - $/\sigma_1$ -/ t_S -observable

Again it holds:

strong
$$(x, \sigma)$$
-observability \iff strong σ -observability strong (x, σ_1) -observability \iff strong σ_1 -observability

Zero-state problem

Property

$$x \equiv 0 \iff \exists t_0 \in \mathbb{R} : x(t_0) = 0$$

not valid anymore

Avoiding zero-state-problem, variant 1

Additional assumptions

(A2) ker
$$\begin{bmatrix} B_i \\ B_j \\ D_i - D_j \end{bmatrix} = \{0\} \ \forall i \neq j$$

Notation:

(A1)
$$u$$
 is real analytic

(A2) $\ker \begin{bmatrix} B_i \\ B_j \\ D_i - D_j \end{bmatrix} = \{0\} \ \forall i \neq j$

$$\Gamma_k = \begin{bmatrix} D_k \\ C_k B_k & D_k \\ C_k A_k B_k & C_k B_k & D_k \\ C_k A_k^2 B_k & C_k A_k B_k & C_k B_k & D_k \\ \vdots & \ddots & \ddots \end{bmatrix}$$

Theorem (cf. Lou and Si 2009)

$$\Sigma^{\sigma}_{(A,B,C,D)}$$
 with (A1), (A2) is σ -observable \Leftrightarrow

$$rank[O_i \ O_i \ \Gamma_i - \Gamma_i] = 2n + rank(\Gamma_i - \Gamma_i) \ \forall i \neq j$$

Relationship to ui-observability

Theorem (see e.g. Kratz (1995) or Hautus (1983))

$$\operatorname{rank}[\mathcal{O}_{i} \mathcal{O}_{j} \Gamma_{i} - \Gamma_{j}] = 2n + \operatorname{rank}(\Gamma_{i} - \Gamma_{j})$$

$$\Leftrightarrow$$

$$\Sigma_{ij} : \begin{bmatrix} \dot{\xi} = \begin{bmatrix} A_{i} & 0 \\ 0 & A_{j} \end{bmatrix} + \begin{bmatrix} B_{i} \\ B_{j} \end{bmatrix} u \\ y_{\Delta_{i,j}} = [C_{i} - C_{j}]\xi + (D_{i} - D_{j})u \\ \text{is unknown-input (ui-) observable} \end{bmatrix}$$

Strong t_S -/ σ_1 -observability (under (A1), (A2))

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Theorem (Küsters and T. 2018)
\Sigma_{(A,B,C,D)}^{\sigma} is t_s-observable \iff \forall i \neq j:
                                           rank[\mathcal{O}_i - \mathcal{O}_i \quad \Gamma_i - \Gamma_i] = n + rk(\Gamma_i - \Gamma_i)
                                                                                       and
                                                                           \mathcal{R}(\Sigma_{ii}) = \{0\}
\Sigma_{(A,B,C,D)}^{\sigma} is \sigma_1-observable \iff \forall i \neq j, p \neq q, (i,j) \neq (p,q):
                                   \operatorname{rank}\begin{bmatrix} \mathcal{O}_{i} & \mathcal{O}_{p} & \Gamma_{i} - \Gamma_{p} \\ \mathcal{O}_{i} & \mathcal{O}_{q} & \Gamma_{i} - \Gamma_{q} \end{bmatrix} = 2n + \operatorname{rank}\begin{bmatrix} \Gamma_{i} - \Gamma_{p} \\ \Gamma_{i} - \Gamma_{q} \end{bmatrix}
                                                                                       and
                                                                           \mathcal{R}(\Sigma_{ii}) = \{0\}
```

Avoiding (A1) and (A2)

Definition (Equivalent switching signal, c.f. Kaba (2014))

For $\Sigma_{(A,B,C,D)}^{\sigma}$, initial value $x_0 \in \mathbb{R}^0$, input u

$$\sigma \overset{x_0,u}{\sim} \widetilde{\sigma}$$
 : \Leftrightarrow $x \equiv \widetilde{x}, \ y \equiv \widetilde{y} \ \text{and} \ \sigma(t) = \widetilde{\sigma}(t) \ \text{expect on}$ intervals where the state is identically zero

Corresponding equivalence class: $[\sigma_{x_0,u}] := \{ \tilde{\sigma} \mid \sigma \overset{x_0,u}{\sim} \tilde{\sigma} \}$

Definition

 $\Sigma^{\sigma}_{(A,B,C,D)}$ is called $(x,[\sigma])$ -, $[\sigma]$ -, $(x,[\sigma_1])$, $[\sigma_1]$ -, and $[t_S]$ -observable $:\Leftrightarrow$ replace in previos definitions $\sigma \neq \widehat{\sigma}$ by $[\sigma_{x_0,u}] \neq [\widehat{\sigma}_{x_0,u}]$

Exactly the same rank-conditions as before!

Introduction

 $\dot{x} = A_{\sigma}x$

 $\dot{x} = A_{\sigma}x + B_{\sigma}u$

 $E_{\sigma}\dot{x} = A_{\sigma}x + B_{\sigma}u$

Observer design

Overview for $\Sigma^{\sigma}_{(A,B,C,D)}$: $\begin{vmatrix} \dot{x} = A_{\sigma}x + B_{\sigma}u \\ y = C_{\sigma}x + D_{\sigma}u \end{vmatrix}$

$$\dot{x} = A_{\sigma}x + B_{\sigma}u$$
$$y = C_{\sigma}x + D_{\sigma}u$$

equivalence classes for σ , u analytical \wedge (A2) u = 0u smooth (x, σ) -observability strong (x, σ) -observability strong $(x, [\sigma])$ -observability $= \sigma$ -observability = strong σ -observability = strong $[\sigma]$ -observability $rk[\mathcal{O}_i \ \mathcal{O}_i] = 2n$ $rk[\mathcal{O}_i \quad \mathcal{O}_i \quad \Gamma_i - \Gamma_i] = 2n + rk(\Gamma_i - \Gamma_i)$ (x, σ_1) -observability strong (x, σ_1) -observability strong $(x, [\sigma_1])$ -observability $= \sigma_1$ -observability = strong σ_1 -observability = strong $[\sigma_1]$ -observability $\operatorname{rk}\begin{bmatrix} \mathcal{O}_i & \mathcal{O}_p \\ \mathcal{O}_i & \mathcal{O}_q \end{bmatrix} = 2n$ $\mathcal{R}(\Sigma_{i,j}) = \{0\} \land \mathsf{rk} \begin{bmatrix} \mathcal{O}_i & \mathcal{O}_p & \Gamma_i - \Gamma_p \\ \mathcal{O}_i & \mathcal{O}_q & \Gamma_i - \Gamma_q \end{bmatrix} = 2n + \mathsf{rk} \begin{bmatrix} \Gamma_i - \Gamma_p \\ \Gamma_i - \Gamma_q \end{bmatrix}$ ts-observability strong ts-obervability strong $[t_S]$ -observability $\mathcal{R}(\Sigma_{i,i}) = \{0\} \land \mathsf{rk}[\mathcal{O}_i - \mathcal{O}_i \quad \Gamma_i - \Gamma_i] = n + \mathsf{rk}(\Gamma_i - \Gamma_i)$ $rk(\mathcal{O}_i - \mathcal{O}_i) = n$

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Summary

Contents

$$\dot{x} = A_{\sigma}x$$

$$\dot{x} = A_{\sigma}x + B_{\sigma}u$$

$$E_{\sigma}\dot{x}=A_{\sigma}x+B_{\sigma}u$$

Summary

Switch-observability for switched DAEs

$$\Sigma_{(E,A,B,C,D)}^{\sigma}: \qquad E_{\sigma}\dot{x} = A_{\sigma}x + B_{\sigma}u \\ y = C_{\sigma}x + D_{\sigma}u$$

After quite a bit of new notations, theory and definitions ...

Theorem (Dissertation Küsters 2018)

 $\Sigma^{\sigma}_{(E,A,B,C,D)}$ is strongly $(x,[\sigma_1])$ -observable \Leftrightarrow $[t_S]$ -observability and

$$\begin{aligned} \operatorname{rk} \left[\begin{array}{l} \mathcal{O}_{i}^{\operatorname{diff}} & \mathcal{O}_{p}^{\operatorname{diff}} & \Gamma_{i} - \Gamma_{p} \\ \mathcal{O}_{j}^{\operatorname{diff}} \Pi_{i} & \mathcal{O}_{q}^{\operatorname{diff}} \Pi_{p} & (\Gamma_{j} - \mathcal{O}^{\operatorname{diff}} M_{i}^{\operatorname{imp}}) - (\Gamma_{q} - \mathcal{O}_{q}^{\operatorname{diff}} M_{p}^{\operatorname{imp}}) \\ \mathcal{O}_{j}^{\operatorname{imp}} \Pi_{i} & \mathcal{O}^{\operatorname{imp}} \Pi_{p} & \mathcal{O}_{j}^{\operatorname{imp}} (M_{j}^{\operatorname{imp}} - M_{i}^{\operatorname{imp}}) - \mathcal{O}_{q} (M_{q}^{\operatorname{imp}} - M_{p}^{\operatorname{imp}}) \end{array} \right] \\ &= \operatorname{dim} \overline{\mathcal{V}^{*}}_{i,p} - \operatorname{dim} \mathcal{M}_{i,j,p,q} + \operatorname{rk} \left(\begin{bmatrix} \Gamma_{i} - \Gamma_{p} \\ \Gamma_{j} - \Gamma_{q} \\ \Gamma_{j}^{\operatorname{imp}} - \Gamma_{q}^{\operatorname{imp}} \end{bmatrix} Z_{i,p}^{2} \right) & \forall i \neq j, p \neq q, \\ (i,j) \neq (p,q) \end{aligned}$$

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$$\dot{x} = A_{\sigma}x$$

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Observer design

Summary

"Trivial" observer design for (x, σ) -obs.

Instantenous observability

 (x, σ) -observability \implies local state and mode observability

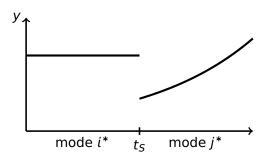
Observer design

- 1. For each mode run a classical state observer
- 2. Pick the one which converges \rightarrow mode and state estimation
- 3. Repeat

Nothing switch specific

Information at the switch (e.g. jumps) not utilized.

Overall observer design



- (0. Detect switching time t_{5} .)
- 1a. Run partial state observers on $(t_S \tau, t_S)$ for all modes.
- 1b. Run partial state observers on $(t_S, t_S + \tau)$ for all modes.
- 2. Combine partial information to find (i^*, j^*) and state estimation $\hat{x}(t_s)$

Summary

Partial state observer

$$\dot{x} = A_p \dot{x} + B_p u,
y = C_p x + D_p u,
\mathcal{O}_p := \begin{bmatrix} C_p \\ C_p A_p \\ \vdots \\ C_p A_p^{n-1} \end{bmatrix} \quad r_p := \operatorname{rank} \mathcal{O}_p$$

Choose orthogonal $Z_p \in \mathbb{R}^{n \times r_p}$ with $\operatorname{im} Z_p = \operatorname{im} \mathcal{O}_p^T$, then

$$\dot{z}_p = \mathbf{Z}_p^{\mathsf{T}} A_p \mathbf{Z}_p z_p + \mathbf{Z}_p^{\mathsf{T}} B_p u$$
$$y = C_p \mathbf{Z}_p z_p + D_p u$$

is observable

Definition (Partial state observer)

Any observer for $z_p = Z_p^T x$ is a partial state observer.

Mode dependence

 Z_p and size r_p are mode dependent.

Reasonable modes

Definition (Reasonable modes)

Mode *i* is reasonable on $(t_S - \tau, t_S)$: \Leftrightarrow

$$\exists x_i^{t_S} : y = C_i x_i + D_i u$$
 where $\dot{x}_i = A_i x_i + B_i u$, $x_i(t_S) = x_i^{t_S}$

In particular, i^* is reasonable on $(t_S - \tau, t_S)$.

Crucial property of reasonable modes

Partial state observers "converge" for all reasonable modes, i.e.

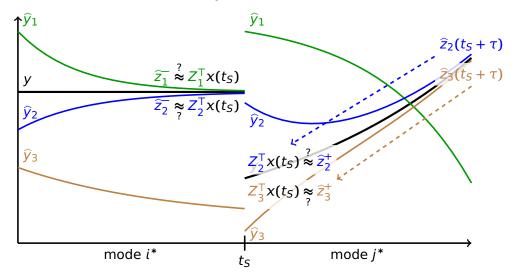
$$y \approx C_i Z_i \hat{z}_i + D_i u$$
 on $(t_S - \varepsilon, t_S) \forall$ reasonable i

Analog definition for reasonable modes j on $(t_S, t_S + \tau)$, with

$$y \approx C_i Z_i \hat{Z}_i + D_i u$$
 on $(t_S + \tau - \varepsilon, t_S + \tau)$ \forall reasonable i

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Illustration of Steps 1 and 2



Combining partial state estimations

Question

How to combine the obtained information before and after the switch?

Obvious fact

$$(x, \sigma_1)$$
-observability \implies observability for known σ with one switch \implies $\ker \mathcal{O}_i \cap \ker \mathcal{O}_j = \{0\} \quad \forall i \neq j$ \implies $\operatorname{rank}[Z_i, Z_j] = n \quad \forall i \neq j$

State estimation candidates

For
$$(i, j) = (i^*, j^*)$$
 we have
$$\begin{pmatrix} \widehat{z}_i^- \\ \widehat{z}_j^+ \end{pmatrix} \approx \begin{bmatrix} Z_i^\top \\ Z_j^\top \end{bmatrix} x(t_S) \implies x(t_S) \approx \begin{bmatrix} Z_i^\top \\ Z_j^\top \end{bmatrix}^{\dagger} \begin{pmatrix} \widehat{z}_i^- \\ \widehat{z}_j^+ \end{pmatrix} =: \widehat{x}_{ij}$$

Final step

Theorem (Küsters & T. 2017)

For sufficiently accurate partial observers and for all reasonable (i, j)

$$(i,j) = (i^*,j^*) \implies \begin{bmatrix} Z_i^\top \\ Z_j^\top \end{bmatrix} \widehat{x}_{ij} \approx \begin{bmatrix} \widehat{z}_i^\top \\ \widehat{z}_j^+ \end{bmatrix}$$
$$(i,j) \neq (i^*,j^*) \implies \begin{bmatrix} Z_i^\top \\ Z_j^\top \end{bmatrix} \widehat{x}_{ij} \not\approx \begin{bmatrix} \widehat{z}_i^\top \\ \widehat{z}_j^+ \end{bmatrix}$$

Summary

- Classical mode-detection property too restrictive
 - State-observability required for each individual mode
 - Information around switch not utilized
 - Novel concept: switch-observability (σ_1 -observability)
- Characterizations in the form of simple rank-tests
- Observer design based on partial state-observers

Future work and topics:

- Extension to nonlinear cases
- > Testing in "real" appplications
- > Distributed design for large networks
- > Using state- and mode-estimations for feedback-control