

About observer synthesis for a system whose solutions converge to a set of indistinguishable states

ALAIN RAPAPORT

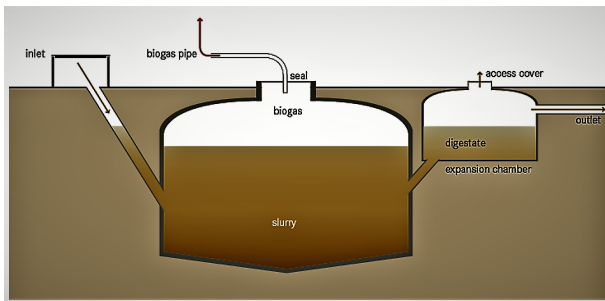
UMR MISTEA, Montpellier, France

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joint work with Denis Dochain, U. Louvain-la-Neuve (Belgium)

Motivation and context

Biogas reactor:



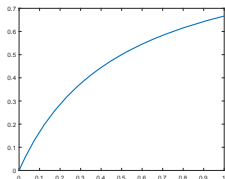
Principle: "substrate $\xrightarrow{\text{biomass}}$ biomass + biogas"

research conducted at LBE (Laboratory of Environmental Biotechnology), INRAE, Narbonne

Batch process with single biogas measurement

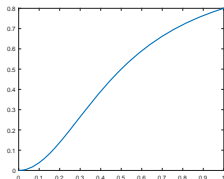
$$(S) : \begin{cases} \dot{b} = \mu(s)b \\ \dot{s} = -\mu(s)b \end{cases} \quad y = \mu(s)b$$

Assumption. μ is C^∞ with $\mu(0) = 0$ and $\mu(s) > 0$ for $s > 0$.



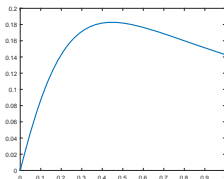
Monod:

$$\mu(s) = \frac{\bar{\mu}s}{K+s}$$



Hill:

$$\mu(s) = \frac{\bar{\mu}s^\alpha}{K^\alpha + s^\alpha}$$



Haldane:

$$\mu(s) = \frac{\bar{\mu}s}{K+s+\frac{s^2}{K_i}}$$

Batch process with single biogas measurement

In coordinates $\begin{bmatrix} z \\ s \end{bmatrix} = \begin{bmatrix} b + s \\ s \end{bmatrix}$,

$$(S) : \begin{cases} \dot{z} = 0 \\ \dot{s} = -\mu(s)(z - s) \end{cases} \quad y = \mu(s)(z - s)$$

defined on $\mathcal{C} := \{(z, s) \in \mathbb{R}^2 \mid z > s > 0\}$

Property. Any solution of (S) on \mathcal{C} is such that $\lim_{t \rightarrow +\infty} s(t) = 0$ and $\mathbb{R}_+ \times \{0\}$ is a set of indistinguishable states...

Observability analysis

$$\frac{d}{dt} \begin{bmatrix} z \\ s \end{bmatrix} = f(z, s) := \begin{bmatrix} 0 \\ -\mu(s)(z-s) \end{bmatrix}, \quad y := h(z, s) = \mu(s)(z-s)$$

Definition. (Gauthier-Kupka 98) (f, h) is **differentially observable** on \mathcal{D} (of order n) when it exists $n \in \mathbb{N}_*$ such that

$$x \mapsto (h(x), L_f(x), \dots, f_f^{n-1}h(x)) \text{ is injective on } \mathcal{D}$$

where $L_f h(x) := \partial_x h(x) \cdot f(x)$, $L_f^i h(x) := L_f(L_f^{i-1}h)(x)$.

Proposition. If μ is

- **concave increasing**, or
- **increasing with log μ decreasing**,

then (f, g) is differentially observable of order 2 on \mathcal{C} but not on $\partial\mathcal{C}$.

Observability analysis

$$\begin{aligned}L_f h(z, s) &= [\mu'(s)(z - s) - \mu(s)]\mu(s)(z - s) \\ &= -\left(\frac{\mu'(s)}{\mu(s)}h(z, s) - \mu(s)\right)h(z, s)\end{aligned}$$

For a given $y > 0$, define $\varphi_y(s) = \mu(s) - \frac{\mu'(s)}{\mu(s)}y$

$$\varphi'_y(s) = \mu'(s) - \left(\frac{\mu''(s)}{\mu(s)} - \frac{\mu'(s)^2}{\mu(s)^2}\right)y > 0 \Rightarrow \varphi_y \text{ invertible}$$

$$\Rightarrow \begin{cases} s = \varphi_{h(z,s)}^{-1} \left(\frac{L_f h(z, s)}{h(z, s)} \right) \\ z = \frac{h(z, s)}{\mu(s)} + s \end{cases} \quad \text{if } h(z, s) \neq 0$$

Consideration of Luenberger observers

In (s, y) coordinates:

Observer:

$$\begin{cases} \dot{\hat{s}} &= -y + G_1(\hat{y} - y) \\ \dot{\hat{y}} &= -\frac{\mu'(\hat{s})}{\mu(\hat{s})}y^2 + \mu(\hat{s})y + G_2(\hat{y} - y) \end{cases}$$

has a singularity at $\hat{s} = 0...$

Consideration of Luenberger observers

In (y, \dot{y}) coordinates:

Assume that μ is increasing and μ' is bounded.

$$\begin{cases} \dot{x} = y \\ \dot{y} = \phi(x, y)y \end{cases} \quad \text{with } \phi(x, y) = -\mu' \circ \mu^{-1} \left(\frac{y}{x} \right) x + \frac{y}{x}$$

Observer:

$$\begin{cases} \dot{\hat{x}} = y(t) + G_1(\hat{y} - y(t)) \\ \dot{\hat{y}} = \tilde{\phi}(\hat{x}, y(t))y(t) + G_2(\hat{y} - y(t)) \end{cases}$$

with $\tilde{\phi}(x, y) = \phi(\max(x, \epsilon, y/\bar{\mu}), y)$, $\bar{\mu} = \sup_{s>0} \mu(s)$ and $\epsilon > 0$ such that $x_0 > \epsilon$.

Convergence of the Luenberger observer

$$\frac{d}{dt} \begin{bmatrix} e_x \\ e_y \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & G_1 \\ \delta(t)y(t) & G_2 \end{bmatrix}}_{A(t)} \begin{bmatrix} e_x \\ e_y \end{bmatrix}$$

$$\text{with } \delta(t) = \begin{cases} \frac{\tilde{\phi}(\hat{x}(t), y(t)) - \phi(x(t), y(t))}{\hat{x}(t) - x(t)} & \text{if } \hat{x}(t) \neq x(t), \\ \partial_x \phi(x(t), y(t)) & \text{if } \hat{x}(t) = x(t). \end{cases}$$

$$\implies \lim_{t \rightarrow +\infty} A(t) = \begin{bmatrix} 0 & G_1 \\ 0 & G_2 \end{bmatrix}$$

Luenberger observer with log measurement

$$(x, y) \mapsto (x, w) := (x, \log y)$$

Observer:

$$\begin{cases} \dot{\hat{x}} = y(t) + G_1(\hat{w} - \log y(t)) \\ \dot{\hat{w}} = \tilde{\phi}(\hat{x}, y(t)) + G_2(\hat{w} - \log y(t)) \end{cases}$$

$$\Rightarrow \frac{d}{dt} \begin{bmatrix} e_x \\ e_w \end{bmatrix} = \underbrace{\begin{bmatrix} 0 & G_1 \\ \delta(t) & G_2 \end{bmatrix}}_{A(t)} \begin{bmatrix} e_x \\ e_w \end{bmatrix}$$

$$\text{with } \lim_{t \rightarrow +\infty} A(t) = \begin{bmatrix} 0 & G_1 \\ -\mu'(0) & G_2 \end{bmatrix} \text{ (for } \epsilon \text{ small enough)}$$

Note that $w(t) \rightarrow +\infty$ when $t \rightarrow +\infty$.

Construction of an asymptotic estimator

A forward integral invariant:

$$\begin{aligned}\lim_{t \rightarrow +\infty} s(t) = 0 &\Rightarrow s(t) = \int_t^{+\infty} y(\tau) d\tau \\ \Rightarrow x(t) &= \frac{y(t)}{\mu \left(\int_t^{+\infty} y(\tau) d\tau \right)}\end{aligned}$$

$$\Rightarrow z = x(t) + s(t) = \boxed{\frac{y(t)}{\mu \left(\int_t^{+\infty} y(\tau) d\tau \right)} + \int_t^{+\infty} y(\tau) d\tau = Cst}$$

Contraction of an asymptotic estimator

$$z = \frac{y(t)}{\mu \left(\int_t^{+\infty} y(\tau) d\tau \right)} + \int_t^{+\infty} y(\tau) d\tau, \quad \forall t > 0$$

$$\Rightarrow \hat{z}(t) = \frac{y(t_0)}{\mu \left(\int_{t_0}^t y(\tau) d\tau \right)} + \int_{t_0}^t y(\tau) d\tau \text{ is an estimator of } z$$

Estimator:

$$\hat{x}(t) = \frac{y(t_0)}{\mu \left(\int_{t_0}^t y(\tau) d\tau \right)} + \int_{t_0}^t y(\tau) d\tau, \quad \hat{s}(t) = \mu^{-1} \left(\frac{y(t)}{\hat{x}(t)} \right)$$

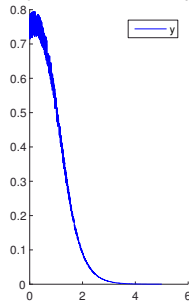
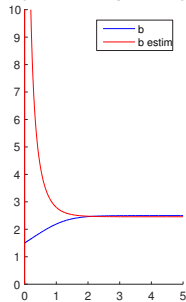
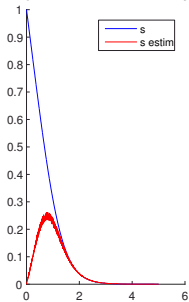
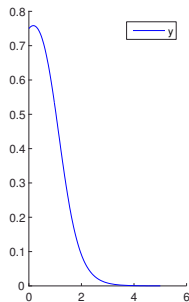
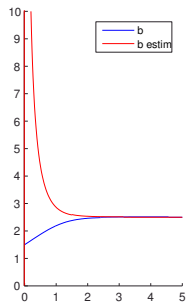
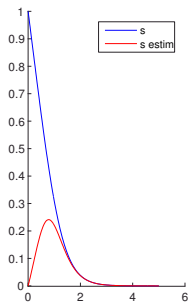
Asymptotic observer

$$\left\{ \begin{array}{l} \dot{v}(t) = y(t), \quad v(0) = 0 \\ \hat{x}(t) = \frac{y(0)}{\mu(v(t))} + v(t) \quad (t > 0) \\ \hat{s}(t) = \mu^{-1} \left(\frac{y(t)}{\hat{x}(t)} \right) \quad (t > 0) \end{array} \right.$$

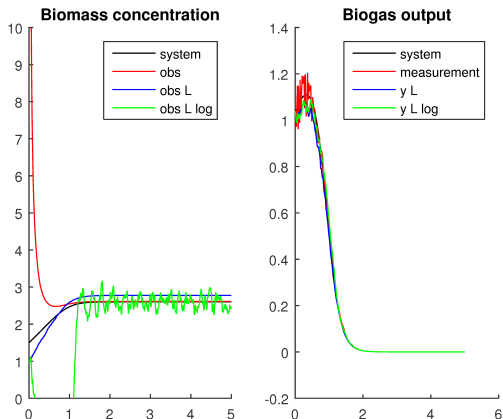
Property. When μ is increasing, $\hat{x} - x$ is decreasing with time:

$$\frac{d}{dt}(\hat{x} - x)(t) = -y(0) \frac{\mu' \left(\int_0^t y(\tau) d\tau \right)}{\left[\mu \left(\int_0^t y(\tau) d\tau \right) \right]^2} y(t)$$

Simulations of the asymptotic estimator



Numerical comparisons - the Monod case

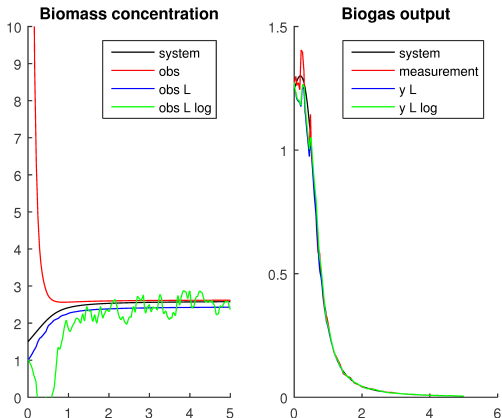


*small
noise
on y*

Luenberger: $G_1 = G_2 = -20$;

Luenberger with log: $G_1 = \lambda_1 \lambda_2 / \mu'(0)$, $G_2 = \lambda_1 + \lambda_2$

Numerical comparisons - the Hill case

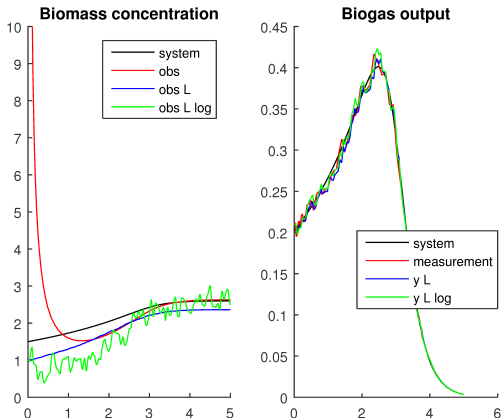


*small
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Luenberger: $G_1 = G_2 = -20$;

Luenberger with log: $G_1 = \lambda_1 \lambda_2 / \mu'(0)$, $G_2 = \lambda_1 + \lambda_2$

Numerical comparisons - the Haldane case



*small
noise
on y*

Luenberger: $G_1 = G_2 = -20$;

Luenberger with log: $G_1 = \lambda_1 \lambda_2 / \mu'(0)$, $G_2 = \lambda_1 + \lambda_2$

Robustness issue

Consider corrupted measurements: $y_{obs}(t) = y(t) + p(t)$,
(assumed to be unbiased i.e. with $\int_0^{+\infty} p(t)dt = 0$)

$$\Rightarrow \lim_{t \rightarrow +\infty} \hat{x}(t) - x(t) = \frac{p(0)}{\mu \left(\int_0^{+\infty} y(\tau) d\tau \right)}.$$

Assumption. There exists $T < +\infty$ such that

$$\frac{1}{T} \int_t^{t+T} p(\tau) d\tau = 0, \quad \forall t > 0.$$

Robust asymptotic estimator

Proposition. The estimator

$$\hat{\chi}(t) = \frac{\int_{t_0=0}^{t_0=T} \left\{ y_{obs}(t_0) + \int_{t_0}^t y_{obs}(\tau) d\tau \cdot \mu \left(\int_{t_0}^t y_{obs}(\tau) d\tau \right) \right\} dt_0}{\int_{t_0=0}^{t_0=T} \mu \left(\int_{t_0}^t y_{obs}(\tau) d\tau \right) dt_0}, \quad t > T$$

is unbiased.

Remark. $\hat{\chi}(\cdot)$ IS NOT the averaging of the asymptotic observer.

Robust asymptotic observer

Define $I(t_0, t) = \int_{t_0}^t y_{obs}(\tau) d\tau$, $t > t_0$

Let $\dot{v}_i = y_{obs(t)} \mathbb{1}_{\{t > (i-1)h\}}(t)$, $v_i(0) = 0$ for $i = 1 \dots N$

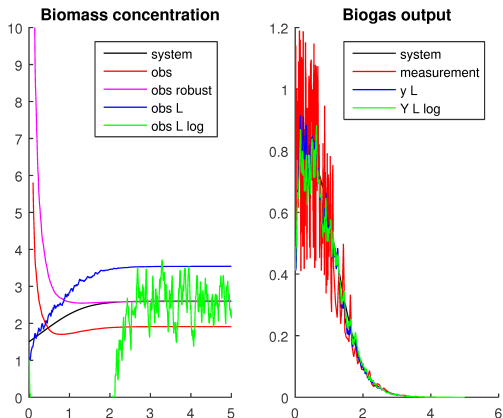
Then for any integrable function γ , one has the approximation

$$\int_{t_0=0}^{t_0=T} \gamma(I(t_0, t)) \simeq h \sum_{i=1}^N \gamma(v_i(t)), \quad t > T$$

Finite dimensional observer.

$$\hat{x}(t) = \frac{\frac{N}{T} v_1(T) + \sum_{i=1}^N v_i(t) \mu(v_i(t))}{\sum_{i=1}^N \mu(v_i(t))} \quad t > T,$$

Numerical comparisons - the Monod case



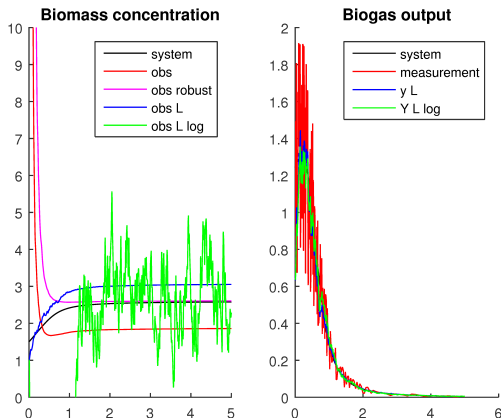
$$T = 0.1$$

$$N = 10$$

Luenberger: $G_1 = G_2 = -20$;

Luenberger with log: $G_1 = \lambda_1 \lambda_2 / \mu'(0)$, $G_2 = \lambda_1 + \lambda_2$

Numerical comparisons - the Hill case



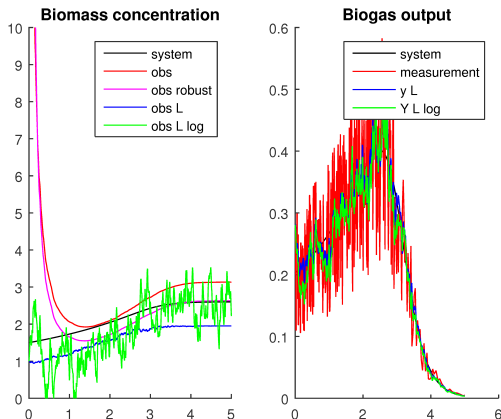
$$T = 0.1$$

$$N = 10$$

Luenberger: $G_1 = G_2 = -20$;

Luenberger with log: $G_1 = \lambda_1 \lambda_2 / \mu'(0)$, $G_2 = \lambda_1 + \lambda_2$

Numerical comparisons - the Haldane case



$$T = 0.1$$

$$N = 10$$

Luenberger: $G_1 = G_2 = -20$;

Luenberger with log: $G_1 = \lambda_1 \lambda_2 / \mu'(0)$, $G_2 = \lambda_1 + \lambda_2$

Other attempts...

The Kazantis-Kravaris approach

$$(S) : \begin{cases} \dot{x} = f(x) \\ \dot{y} = h(x) \end{cases}$$

Proposition (Kazantis-Kravaris 98): If there exists a diffeomorphism $T : \mathbb{R}^n \mapsto \mathbb{R}^n$ such that

$$\frac{d}{dt} T(x) = AT(x) + Bh(x) \quad \text{with } A \text{ Hurwitz}$$

then $\dot{z} = Az + By(t)$, $\hat{x} = T^{-1}(z)$ is an observer for (S) .

Remark. Such a transformation T is solution of the p.d.e.

$$\partial_x T(x)f(x) = AT(x) + Bh(x)$$

(and one can try $A = \text{diag}(-\lambda_1, \dots, -\lambda_n)$, $B = [1, \dots, 1]^t$).

Observer construction

Let $T_\lambda : (z, s) \mapsto T_\lambda(z, s)$ be a solution on of the o.d.e.

$$\mu(s)(z - s) \frac{\partial T_\lambda}{\partial s} = \lambda T_\lambda - \mu(s)(z - s) \quad \text{where } z \text{ is a parameter}$$

for $(z, s) \in \mathcal{T} = \{(z, s) \in \mathbb{R}^2 \mid s \in (0, z - z_{\min}), z \in [z_{\min}, z_{\max}]\}$.

$$\text{Define } \phi_{\lambda_1, \lambda_2} : \begin{bmatrix} z \\ s \end{bmatrix} \mapsto \begin{bmatrix} T_{\lambda_1}(z, s) \\ T_{\lambda_2}(z, s) \end{bmatrix} \quad (\lambda_1 \neq \lambda_2)$$

► Can one extend $\phi_{\lambda_1, \lambda_2}$ outside \mathcal{T} to be globally Lipschitz on \mathbb{R}^2 with a Lipschitz inverse?

$$\text{Then } \begin{cases} \dot{\hat{T}}_{\lambda_1} = -\lambda_1 \hat{T}_{\lambda_1} + y(t) \\ \dot{\hat{T}}_{\lambda_2} = -\lambda_2 \hat{T}_{\lambda_2} + y(t) \end{cases}, \quad \begin{bmatrix} \hat{z} \\ \hat{s} \end{bmatrix} = \phi_{\lambda_1, \lambda_2}^{-1} \left(\begin{bmatrix} \hat{T}_{\lambda_1} \\ \hat{T}_{\lambda_2} \end{bmatrix} \right)$$

is an observer of (\mathcal{S}) .

Transformation construction

Define $\gamma_z(s) = \mu(s)(z - s)$, $s \in [0, z]$ for $z > 0$

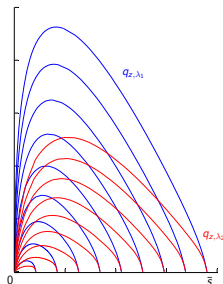
Lemma. Fix $\lambda > 0$, $z > 0$ and $\bar{s} \in]0, z[$

1. There exists a unique $q_{z,\lambda} \in C^1([0, \bar{z}], \mathbb{R}^+)$ s.t.

$$\gamma_z(s)q'_{z,\lambda}(s) = \lambda q_{z,\lambda}(s) - \gamma_z(s), \quad \forall s \in [0, \bar{s}]$$

with $q_{z,\lambda}(0) = q_{z,\lambda}(\bar{s}) = 0$.

2. $\lambda_2 > \lambda_1 > 0 \Rightarrow q_{z,\lambda_1}(s) > q_{z,\lambda_2}(s) > 0$, $\forall s \in]0, \bar{s}[$
3. γ_z concave $\Rightarrow q_{z,\lambda}$ (strictly) concave function on $[0, \bar{s}]$.

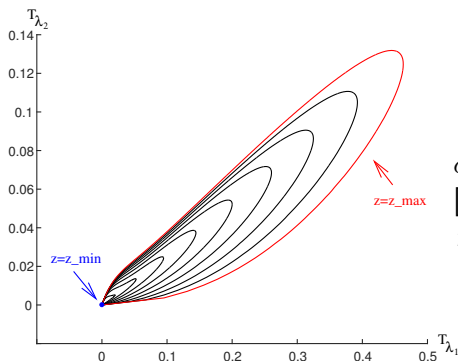


*Functions $q_{\lambda,z}$ with $\lambda_1 < \lambda_2$
for various z*

Transformation construction

$$\phi_{\lambda_1, \lambda_2}(z, s) = \begin{bmatrix} q_{z, \lambda_1}(s) \\ q_{z, \lambda_2}(s) \end{bmatrix} \text{ with } \phi_{\lambda_1, \lambda_2}(z, z - z_{\min}) = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

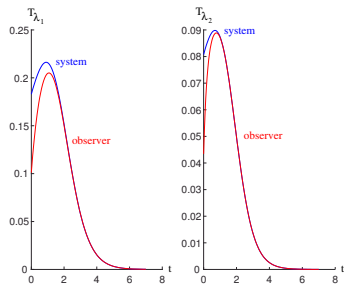
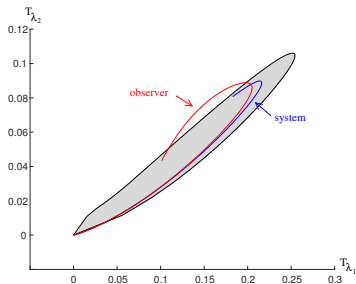
Proposition For $\lambda_2 > \lambda_1 > 0$, the map $\phi_{\lambda_1, \lambda_2}$ is injective on $\phi_{\lambda_1, \lambda_2}(\bar{\mathcal{T}}) \setminus \{(0, 0)\}$.



Sets
 $\phi_{\lambda_1, \lambda_2}(\{z\} \times [0, z - z_{\min}])$
for various z

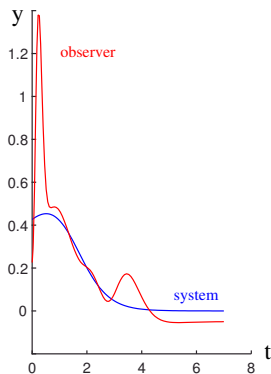
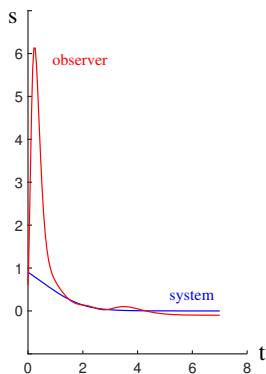
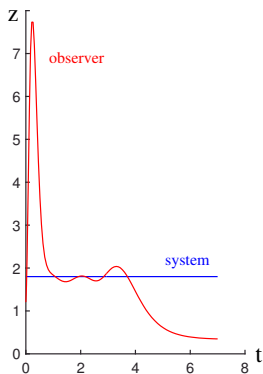
Simulations in $(T_{\lambda_1}, T_{\lambda_2})$ coordinates

$$\mu(s) = \frac{s}{1+s}, \quad \lambda_1 = 2, \quad \lambda_2 = 5$$



Come back in (z, s) coordinates

- ▶ Open problem: how to write **explicitly** $\phi_{\lambda_1, \lambda_2}^{-1}$?
- ▶ numerical approximation with spline functions:



A particular case: $\mu(s) = s$

$$Z = \begin{bmatrix} \log(xs) \\ s - x \end{bmatrix} \Rightarrow \dot{Z} = \begin{bmatrix} Z_2 \\ -2y(t) \end{bmatrix}$$

We consider then the observer

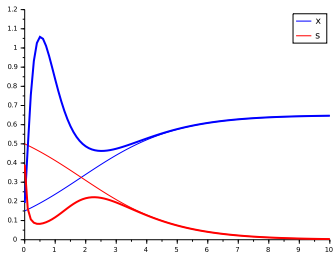
$$\begin{cases} \dot{\hat{Z}}_1 &= \hat{Z}_2 + G_1(\log(y(t)) - \hat{Z}_1) \\ \dot{\hat{Z}}_2 &= -2y(t) + G_2(\log(y(t)) - \hat{Z}_1) \end{cases} \quad \text{defined for } y(t) > 0$$

$$\text{with } \begin{cases} \hat{x} &= \phi_x(\hat{Z}_2, y(t)) = \frac{-\hat{Z}_2 + \sqrt{\hat{Z}_2^2 + 4y(t)}}{2} \\ \hat{s} &= \phi_s(\hat{Z}_2, y(t)) = \frac{\hat{Z}_2 + \sqrt{\hat{Z}_2^2 + 4y(t)}}{2} \end{cases}$$

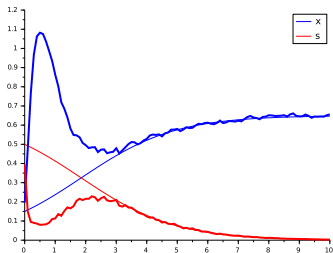
$$\Rightarrow \dot{E} = \begin{pmatrix} G_1 & 1 \\ G_2 & 0 \end{pmatrix} E \text{ in } Z \text{ coordinates,}$$

$|\partial_{Z_2} \phi_x| \leq 1, |\partial_{Z_2} \phi_s| \leq 1$ preserve the convergence speed in (x, s)

A particular case: $\mu(s) = s$



without noise



with noise

Conclusion and perspectives

- ▶ The asymptotic observer converges to the **exact** unobservable state in $\partial\mathcal{C}$
- ▶ Can we improve the convergence speed without asymptotic bias?
- ▶ Extension of the method to a larger class of problems?

Reference: Rapaport, A. and Dochain, D. (2020). A robust asymptotic observer for systems that converge to unobservable states. A batch reactor case study. *IEEE Transactions on Automatic Control*, 65 (6), 2693–2699.