

Introduction to Nonlinear Control

Stability, control design, and estimation

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Part II:

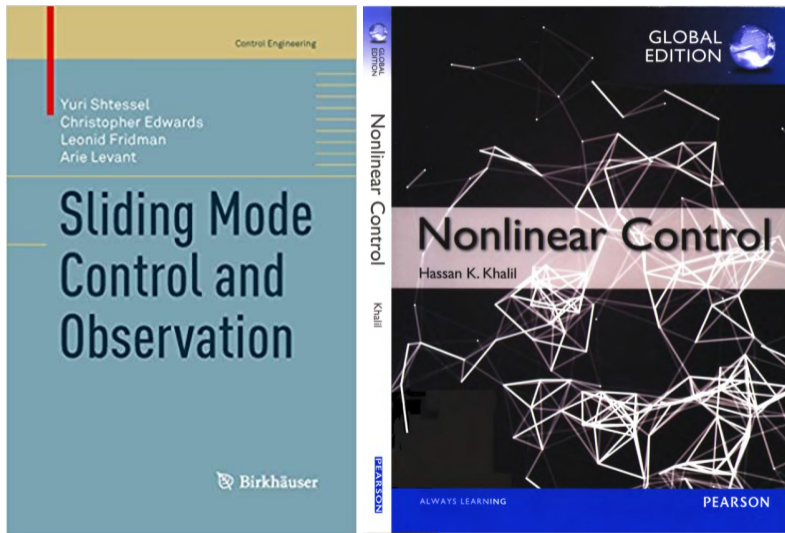
Chapter 10: Sliding Mode Control

- 10.1 Finite-Time Stability
- 10.2 Basic Sliding Mode Control
- 10.3 A More General Setting
- 10.4 Estimating the Disturbance
- 10.5 Output Tracking



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Sliding Mode Control



Nonlinear Systems - Fundamentals

- 1 Finite-Time Stability
- 2 Basic Sliding Mode Control
 - Terminology
 - Chattering & Chattering Avoidance
- 3 A More General Structure
- 4 Estimating the Disturbance
- 5 Output Tracking

Sliding Mode Control

We consider systems of the form

$$\dot{x} = f(x, u, \delta(t, x))$$

$$y = h(x)$$

with

- state $x \in \mathbb{R}^n$
- input $u \in \mathbb{R}^m$
- output $y \in \mathbb{R}$
- potentially time and state dependent unknown disturbance $\delta : \mathbb{R}_{\geq 0} \times \mathbb{R}^n \rightarrow \mathbb{R}^n$

We will be interested in

- stabilizing the origin
- output tracking

despite the presence of the disturbance.

↪ First we have to discuss *finite-time stability*.

Section 1

Finite-Time Stability

Finite-Time Stability

Consider $(f : \mathbb{R}^n \rightarrow \mathbb{R}^n)$

$$\dot{x} = f(x), \quad x(0) = x_0 \in \mathbb{R}^n$$

We assume $f(0) = 0$, and thus $x^e = 0$ is an equilibrium.

Definition (Finite-time stability)

The origin is said to be **finite-time stable** if there exist an open neighborhood $\mathcal{D} \subset \mathbb{R}^n$ of the origin and a function $T : \mathcal{D} \setminus \{0\} \rightarrow (0, \infty)$, called the **settling-time function**, such that the following statements hold:

- **(Stability)** For every $\varepsilon > 0$ there exists a $\delta > 0$ such that, for every $x(0) = x_0 \in \mathcal{B}_\delta \cap \mathcal{D} \setminus \{0\}$, $x(t) \in \mathcal{B}_\varepsilon$ for all $t \in [0, T(x_0))$.
- **(Finite-time convergence)** For every $x(0) = x_0 \in \mathcal{D} \setminus \{0\}$, $x(\cdot)$ is defined on $[0, T(x_0))$, $x(t) \in \mathcal{D} \setminus \{0\}$ for all $t \in [0, T(x_0))$, and $x(t) \rightarrow 0$ for $t \rightarrow T(x_0)$.

The origin is said to be a **globally finite-time stable** if it is finite-time stable with $\mathcal{D} = \mathbb{R}^n$.

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Example

Consider

$$\dot{x} = f(x) = -\sqrt[3]{x^2}, \quad (\text{with } f(0) = 0)$$

Note that

- f is not Lipschitz at the origin
- uniqueness of solutions can only be guaranteed if $x(t) \neq 0$

We can verify that

$$x(t) = -\frac{1}{27}(t - 3 \operatorname{sign}(x(0)) \sqrt[3]{|x(0)|})^3$$

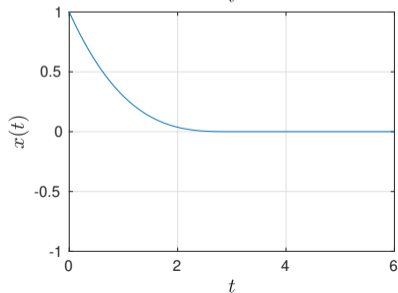
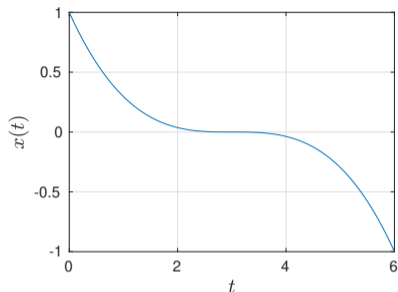
is a solution for all $x \in \mathbb{R}$.

However, for $x(0) > 0$

$$x(t) = \begin{cases} -\frac{1}{27}(t - 3 \sqrt[3]{|x(0)|})^3 & \text{if } t \leq 3 \sqrt[3]{|x(0)|} \\ 0 & \text{if } t \geq 3 \sqrt[3]{|x(0)|} \end{cases}$$

is also a solution.

Finite-Time Stability (2)



Example

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Finite-Time Stability (3)

Example

Consider

$$\dot{x} = f(x) = -\text{sign}(x) \sqrt[3]{x^2}.$$

We can verify

$$x(t) = \begin{cases} -\frac{1}{27} \text{sign}(x(0))(t - 3\sqrt[3]{|x(0)|})^3 & \text{if } t \leq 3\sqrt[3]{|x(0)|} \\ 0 & \text{if } t \geq 3\sqrt[3]{|x(0)|} \end{cases}$$

~> The ODE admits unique solutions

Once the equilibrium is reached, the inequalities

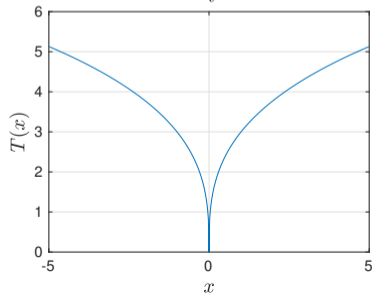
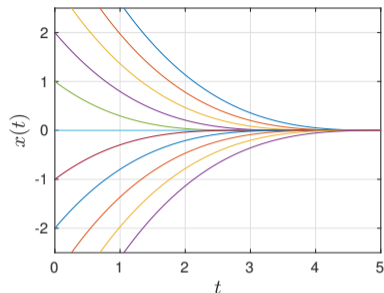
$$-\text{sign}(x) \sqrt[3]{x^2} < 0 \text{ for all } x > 0, \quad \text{and}$$

$$-\text{sign}(x) \sqrt[3]{x^2} > 0 \text{ for all } x < 0$$

ensure that the origin is attractive.

It follows from the explicit solution that

- The origin is finite-time stable
- Settling time $T(x) = 3\sqrt[3]{|x|}$



Finite-Time Stability (4)

Theorem (Lyapunov fcn for finite-time stability)

Consider $\dot{x} = f(x)$ with $f(0) = 0$. Assume there exist a continuous function $V : \mathbb{R}^n \rightarrow \mathbb{R}_{\geq 0}$, which is continuously differentiable on $\mathbb{R}^n \setminus \{0\}$, $\alpha_1, \alpha_2 \in \mathcal{K}_\infty$ and a constant $\kappa > 0$ such that

$$\alpha_1(|x|) \leq V(x) \leq \alpha_2(|x|),$$

$$\dot{V}(x) = \langle \nabla V(x), f(x) \rangle \leq -\kappa \sqrt{V(x)} \quad \forall x \neq 0.$$

Then the origin is globally finite-time stable.

Moreover, the *settling-time* $T(x) : \mathbb{R}^n \rightarrow \mathbb{R}_{\geq 0}$ is upper bounded by

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Proof.

Comparison principle:

$$\int_0^t \frac{dV(x(t))}{\sqrt{V(x(t))}} \leq - \int_0^t \kappa dt$$

leads to

$$\sqrt{V(x(t))} \leq \sqrt{V(x(0))} - \frac{\kappa t}{2}.$$

Using the lower and upper bound

$$|x(t)| \leq \alpha_1^{-1} \left(\left(\sqrt{\alpha_2(|x(0)|)} - \frac{\kappa t}{2} \right)^2 \right)$$

\rightsquigarrow Finite-time convergence

Moreover,

$$\sqrt{\alpha_2(|x(0)|)} - \frac{\kappa t}{2} = 0,$$

implies $|x(T)| \leq 0$ for all $T \geq \frac{2}{\kappa} \sqrt{\alpha_2(|x(0)|)}$ □

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For quadratic Lyapunov functions: ($a_1, a_2 > 0$)

$$a_1|x|^2 \leq V(x) \leq a_2|x|^2$$

$$|x(t)| \leq \frac{1}{\sqrt{a_1}} \left(\sqrt{a_2}|x(0)| - \frac{\kappa t}{2} \right)$$

$$T(|x|) \leq |x| \frac{2\sqrt{a_2}}{\kappa}$$

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Finite-Time Stability (5)

Example

Consider again

$$\dot{x} = f(x) = -\operatorname{sign}(x) \sqrt[3]{x^2}$$

Candidate Lyapunov function (continuously differentiable for all $x \neq 0$)

$$V(x) = \sqrt[3]{x^2}$$

Define $\alpha_1, \alpha_2 \in \mathcal{K}_\infty$,

$$\alpha_1(s) = \alpha_2(s) = \sqrt[3]{s^2}$$

Then, for all $x \neq 0$, it holds that

$$\begin{aligned}\dot{V}(x) &= \langle \nabla V(x), -\operatorname{sign}(x) \sqrt[3]{x^2} \rangle = \frac{2}{3} \operatorname{sign}(x) |x|^{-\frac{1}{3}} (-\operatorname{sign}(x) |x|^{\frac{2}{3}}) \\ &= -\frac{2}{3} |x|^{\frac{1}{3}} = -\frac{2}{3} \sqrt{V(x)}\end{aligned}$$

$\rightsquigarrow V$ is a Lyapunov function & the origin is finite-time stable

Bound on the settling time

$$T(x) \leq \frac{2}{\kappa} \sqrt{\alpha_2(|x|)} = \frac{2}{\frac{2}{3}} \sqrt{|x|^{\frac{2}{3}}} = 3 \sqrt[3]{|x|}$$

Section 2

Basic Sliding Mode Control

Basic Sliding Mode Control

As an example, consider:

$$\begin{aligned}\dot{x} &= x^3 + z, \\ \dot{z} &= u + \delta(t, x, z).\end{aligned}$$

- **Unknown disturbance** $\delta : \mathbb{R}_{\geq 0} \times \mathbb{R}^2 \rightarrow \mathbb{R}$
- **Assumption:** there exists $L_\delta \in \mathbb{R}_{>0}$ such that

$$|\delta(t, x, z)| \leq L_\delta \quad (t, x, z) \in \mathbb{R}_{\geq 0} \times \mathbb{R}^2$$

- Thus, δ is bounded but not necessarily continuous

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Goal: Exponential stability of the x -subsystem

- I.e., we want x to behave as $\dot{x} = -x$ (for all bounded disturbances)
- The desired behavior implies $\dot{x} + x = 0$
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Approach: Define a new state

$$\sigma \doteq x^3 + z + x \quad \text{and} \quad V(\sigma) = \frac{1}{2}\sigma^2$$

- Then

$$\begin{aligned}\dot{V}(\sigma) &= \sigma \dot{\sigma} = \sigma (3x^2 \dot{x} + \dot{z} + \dot{x}) \\ &= \sigma (3x^5 + 3x^2 z + u + \delta(t, x, z) + x^3 + z).\end{aligned}$$

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$$u = v - 3x^5 - 3x^2 z - x^3 - z$$

so that $\dot{V}(\sigma) = \sigma (v + \delta(t, x, z))$ (with new input v)

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so that $\dot{V}(\sigma) = \sigma (v + \delta(t, x, z))$ (with new input v)

- Selecting $v = -\rho \operatorname{sign}(\sigma)$, $\rho > 0$, provides the estimate

$$\begin{aligned}\dot{V}(\sigma) &= \sigma (-\rho \operatorname{sign}(\sigma) + \delta(t, x, z)) = -\rho|\sigma| + \sigma\delta(t, x, z) \\ &\leq -\rho|\sigma| + L_\delta|\sigma| = -(\rho - L_\delta)|\sigma|.\end{aligned}$$

- Finally, with $\rho = L_\delta + \frac{\kappa}{\sqrt{2}}$, $\kappa > 0$, we have

$$\dot{V}(\sigma) \leq -\frac{\kappa|\sigma|}{\sqrt{2}} = -\alpha\sqrt{V(\sigma)} \rightsquigarrow \text{finite-time stab. of } \sigma = 0$$

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- Note that the control

$$u = -\left(L_\delta + \frac{\kappa}{\sqrt{2}}\right) \operatorname{sign}(x^3 + z + x) - 3x^5 - 3x^2 z - x^3 - z$$

is independent of the term $\delta(t, x, z)$.

Basic Sliding Mode Control (2)

Consider:

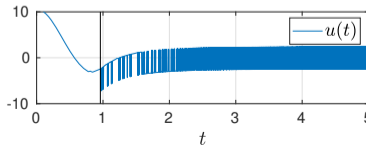
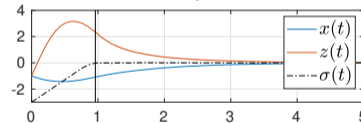
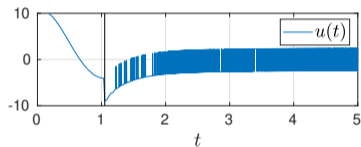
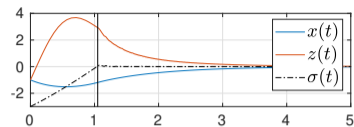
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Control law:

$$u = -\left(L_\delta + \frac{\kappa}{\sqrt{2}}\right) \text{sign}(x^3 + z + x) - 3x^5 - 3x^2z - x^3 - z$$

Parameter selection for the simulations:

- $L_\delta = 1$ and $\kappa = 2$
- $\delta(t, x, z) = \sin(t)$ (top)
- $\delta(t, x, z) = \text{sign}(\cos(2t) \sin(2t))$ (bottom)



Basic Sliding Mode Control (2)

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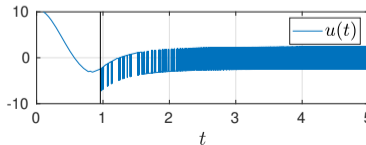
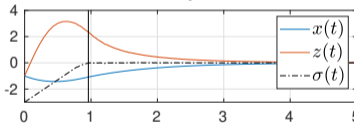
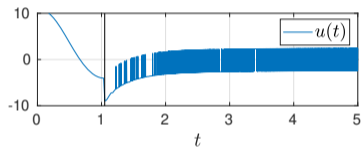
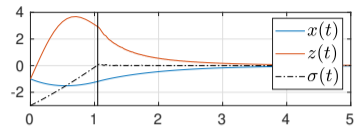
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We observe that

- σ converges to zero in finite-time
- Afterwards (x, z) asymptotically approach the origin
- Since the ordinary differential equation is solved numerically, σ is not exactly zero!



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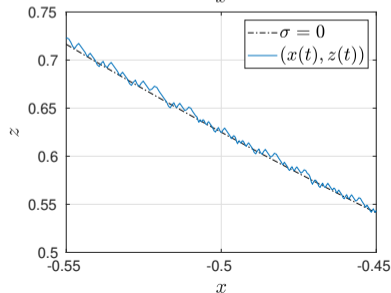
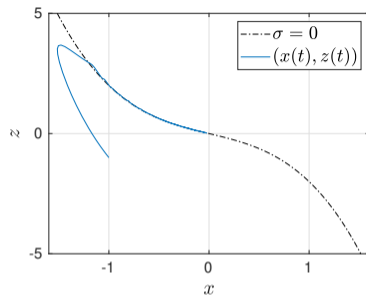
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We observe that

- σ converges to zero in finite-time
- Afterwards (x, z) asymptotically approach the origin
- Since the ordinary differential equation is solved numerically, σ is not exactly zero!

Convergence structure:

↪ Similar to backstepping/forwarding

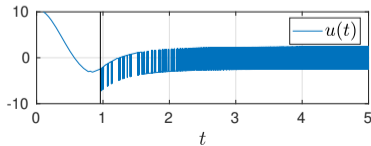
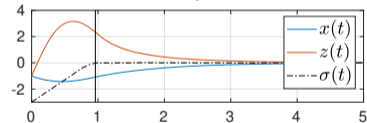
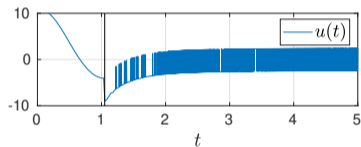
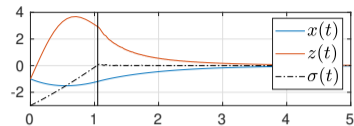


Terminology

- Sliding variable: σ
- Sliding surface

$$\{(x, z) \in \mathbb{R}^2 : \sigma(x, z) = 0, (x, z) \in \mathbb{R}^2\},$$

↪ The sliding variable, and thus implicitly the sliding surface, is defined such that the origin of the x -subsystem is exponentially stable if $\sigma(t) = 0$, $t \in \mathbb{R}_{\geq 0}$, is satisfied.



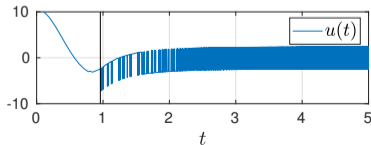
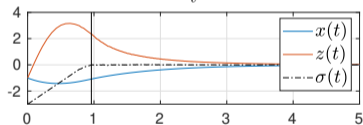
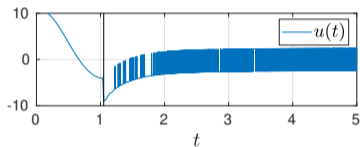
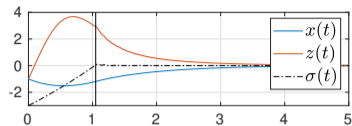
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- We have defined σ based on the condition $\dot{x} = -x$. We could have also used $\dot{x} = -2x$ or $\dot{x} = -x^3$ (asymptotic stability), for example.
- The control law u is derived such that states converge to the sliding surface in finite-time.

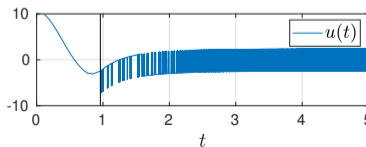
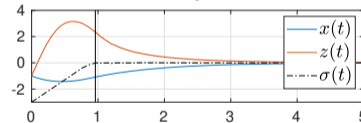
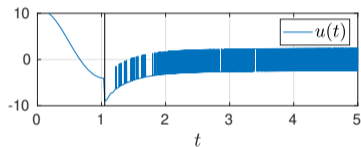
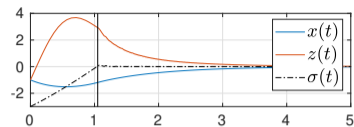


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 - The control law u is derived such that states converge to the sliding surface in finite-time.
- ↪ Convergence of $\sigma(t) \rightarrow 0$ is called the *reaching phase*.
- On the sliding surface the selection of u ensures that the dynamics behave like $\dot{x} = -x$.
- ↪ This is called *sliding phase* and guarantees asymptotic stability of the origin for the overall closed-loop system.



Chattering & Chattering Avoidance

Note that:

- The control law u is discontinuous due to $v = \rho \operatorname{sign}(\sigma)$
 - v switches between ρ and $-\rho$ depending on the sign of the sliding variable σ
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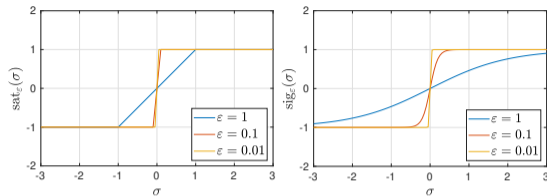
Approximation of the sign-function:

- (Continuous) saturation: ($\operatorname{sat}_\varepsilon : \mathbb{R} \rightarrow [-1, 1]$)

$$\operatorname{sat}_\varepsilon(\sigma) = \operatorname{sat}\left(\frac{\sigma}{\varepsilon}\right) = \begin{cases} 1, & \frac{\sigma}{\varepsilon} \geq 1 \\ \frac{\sigma}{\varepsilon}, & -1 \leq \frac{\sigma}{\varepsilon} \leq 1 \\ -1, & \frac{\sigma}{\varepsilon} \leq -1 \end{cases}$$

- (Smooth) Sigmoid function: ($\operatorname{sig}_\varepsilon : \mathbb{R} \rightarrow [-1, 1]$)

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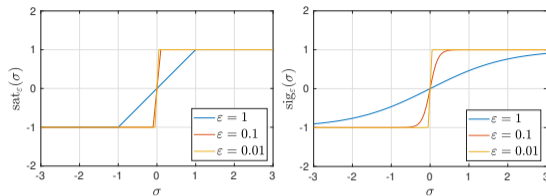
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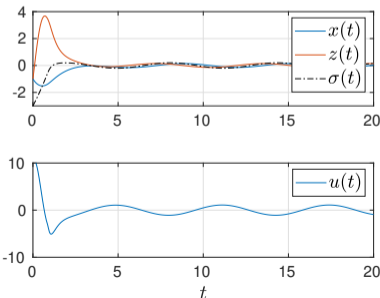
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Back to the example: ($\varepsilon = 0.5$)

$$u_\varepsilon = -\left(L_\delta + \frac{\kappa}{\sqrt{2}}\right) \operatorname{sat}_\varepsilon(x^3 + z + x) - 3x^5 - 3x^2z - x^3 - z$$



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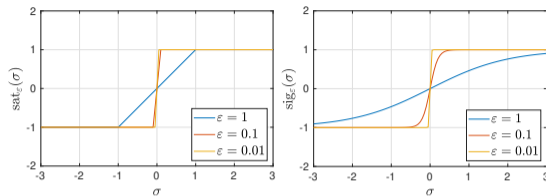
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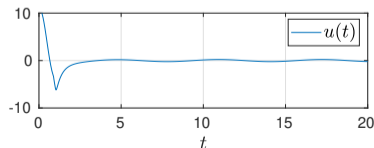
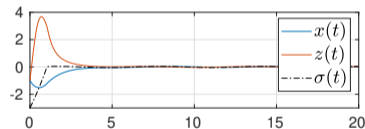
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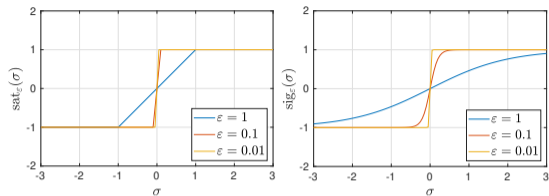
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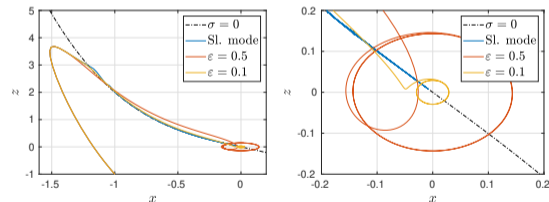
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Section 3

A More General Structure

A More General Structure

Consider:

$$\begin{aligned}\dot{x} &= f_1(x, z) \\ \dot{z} &= f_2(x, z) + g(x, z)(u + \delta(t, x, z))\end{aligned}$$

where

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Assumptions:

- $|\delta(t, x, z)| \leq L_\delta$ for all $(t, x, z) \in \mathbb{R}_{\geq 0} \times \mathbb{R}^{n+1}$
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- Select the new input:

$$v = - \left(\frac{\kappa}{\sqrt{2}} + L_g L_\delta \right) \text{sign}(\sigma)$$

- Thus, with the bounds it holds that:

$$\dot{V}(\sigma) \leq \sigma v + |\sigma|L_g L_\delta = -\frac{\kappa}{\sqrt{2}}|\sigma| = -\kappa\sqrt{V(\sigma)}$$

$\rightsquigarrow \sigma(t) = 0$ in finite time.

A More General Structure (2)

Theorem

Consider the dynamics

$$\begin{aligned}\dot{x} &= f_1(x, z) \\ \dot{z} &= f_2(x, z) + g(x, z)(u + \delta(t, x, z))\end{aligned}$$

with $f_1(0, 0) = 0$. Assume that $g(x, z) \neq 0$ and there exists a constant $L_g > 0$ such that $|g(x, z)| \leq L_g$ for all $(x, z) \in \mathbb{R}^{n+1}$. Additionally assume that $k(x)$ is defined such that the origin of $\dot{x} = f(x, k(x))$ is asymptotically stable with $k(0) = 0$.

Then for all disturbances δ satisfying the condition

$$|\delta(t, x, z)| \leq L_\delta \quad \forall (t, x, z) \in \mathbb{R}^{n+1} \times \mathbb{R}_{\geq 0}$$

for some $L_\delta > 0$, **the feedback law**

$$u = \frac{1}{g(x, z)} \left(-f_2(x, z) + \frac{\partial k}{\partial x}(x) f_1(x, z) \right) - \left(\frac{\kappa}{\sqrt{2}} + L_g L_\delta \right) \frac{\text{sign}(z - k(x))}{g(x, z)}$$

asymptotically stabilizes the origin of the system for all $\kappa > 0$.

Additionally, the **sliding surface $\sigma = z - k(x) = 0$ is reached no later than**

$$T(\sigma(0)) = T(z(0) - k(x(0))) = \frac{1}{\sqrt{2}\kappa} |z(0) - k(x(0))|.$$

Section 4

Estimating the Disturbance

Estimating the Disturbance

So far:

- We have introduced the sliding variable σ
- Through Lyapunov arguments we ensure that (in theory) $\sigma(t) = 0$ in finite time
- In numerical simulations (or in practice) σ will not be exactly zero.
- To dominate the disturbance, we defined the control law

$$u = \frac{\left(-f_2(x, z) + \frac{\partial k}{\partial x}(x)f_1(x, z)\right) - \left(\frac{\kappa}{\sqrt{2}} + L_g L_\delta\right) \text{sign}(z - k(x))}{g(x, z)}$$

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Now:

- Consider the **unimplementable control**

$$u_{eq} = \frac{1}{g(x, z)} \left(-f_2(x, z) + \frac{\partial k}{\partial x}(x)f_1(x, z)\right) - \delta(t, x, z)$$

(which is called the *equivalent control*)

- Note that: (Assuming that δ is sufficiently smooth,) on the sliding surface where $\dot{\sigma} = 0$, it follows that u_{eq} guarantees $\sigma(t) = 0$ for all $t \geq T$ if $\sigma(T) = 0$ without the chattering effects
- Moreover, if δ is a smooth function, u_{eq} is a smooth average of the chattering control u

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- Moreover, if δ is a smooth function, u_{eq} is a smooth average of the chattering control u

Estimation of δ through a low-pass filter:

- **Idea:** Apply low-pass filter to the chattering input

$$v = - \left(\frac{\kappa}{\sqrt{2}} + L_g L_\delta \right) \text{sign}(\sigma).$$

- In particular, consider **augmented dynamics** ($\tau > 0$ small)

$$\dot{x} = f_1(x, z)$$

$$\dot{z} = f_2(x, z) + g(x, z)(\hat{u}_{eq} + \delta(t, x, z))$$

$$\dot{\xi} = -\frac{1}{\tau}\xi + \frac{1}{\tau} \text{sign}(z - k(x))$$

- **Approximated equivalent control:**

$$\hat{u}_{eq} = \frac{\left(-f_2(x, z) + \frac{\partial k}{\partial x}(x)f_1(x, z)\right)}{g(x, z)} - \left(\frac{\kappa}{\sqrt{2}} + L_g L_\delta\right) \frac{\xi}{g(x, z)}.$$

(where we have replaced $\text{sign}(\sigma)$ by ξ)

- ↔ The approximated equivalent control is an alternative to u

Estimating the Disturbance (Detour: Low-pass filter)

Remark (Low-pass filter)

The dynamics $\dot{\xi} = -\frac{1}{\tau}\xi + \frac{1}{\tau}\text{sign}(z - k(x))$ represent a *low-pass filter*. To see this, consider the one-dimensional system

$$\dot{x} = -\frac{1}{\tau}x + \frac{1}{\tau}u$$

$$y = x$$

and its representation in the frequency domain

$$\hat{y}(s) = (s + \frac{1}{\tau})^{-1} \frac{1}{\tau} \hat{u}(s) = \frac{\frac{1}{\tau}}{s + \frac{1}{\tau}} \hat{u}(s).$$

For $\tau > 0$ small we observe from the transfer function

$$G(s) = \frac{\frac{1}{\tau}}{s + \frac{1}{\tau}}$$

that for low frequencies the system approximately satisfies $\hat{y}(s) \approx \hat{u}(s)$ and for high frequencies it holds that $\hat{y}(s) \approx 0$.

Estimating the Disturbance (Example)

Example

Original example with augmented state: ($\tau > 0$, small)

$$\dot{x} = x^3 + z,$$

$$\dot{z} = u + \delta(t, x, z)$$

$$\dot{\xi} = -\frac{1}{\tau}\xi + \frac{1}{\tau} \text{sign}(z - k(x))$$

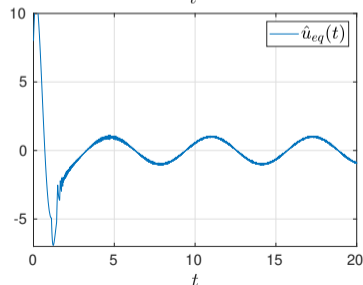
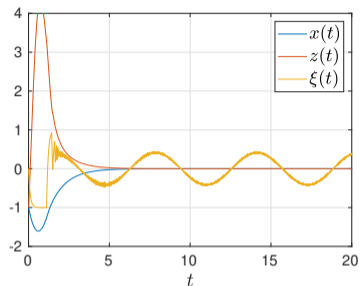
- We follow the steps so far, define $z = k(x) = -x^3 - x$, i.e., $\dot{x} = x^3 - x^3 - x = -x$ (i.e. $x = 0$ is exponentially stable).
- The sliding mode control law (from the theorem)

$$\begin{aligned} u &= (-3x^2 - 1)(x^3 + z) - \left(\frac{\kappa}{\sqrt{2}} + L_\delta\right) \text{sign}(z + (x^3 + x)) \\ &= -3x^5 - x^3 - 3x^2z - z - \left(\frac{\kappa}{\sqrt{2}} + L_\delta\right) \text{sign}(z + x^3 + x) \end{aligned}$$

- The approximated equivalent control

$$\hat{u}_{eq} = -3x^5 - x^3 - 3x^2z - z - \left(\frac{\kappa}{\sqrt{2}} + L_\delta\right) \xi$$

Here: $\delta(t, x, z) = \sin(t)$, $L_\delta = 1$, $\kappa = 2$ and $\tau = 0.1$.



Estimating the Disturbance

Compare equivalent & approximated equivalent control:

$$u_{eq} = \frac{1}{g(x, z)} \left(-f_2(x, z) + \frac{\partial k}{\partial x}(x) f_1(x, z) \right) - \delta(t, x, z)$$

$$\hat{u}_{eq} = \frac{\left(-f_2(x, z) + \frac{\partial k}{\partial x}(x) f_1(x, z) \right)}{g(x, z)} - \left(\frac{\kappa}{\sqrt{2}} + L_g L_\delta \right) \frac{\xi}{g(x, z)}$$

An estimation of the disturbance:

$$\hat{\delta}(t, x, z) = \left(\frac{\kappa}{\sqrt{2}} + L_g L_\delta \right) \frac{\xi}{g(x, z)}$$

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Example

Consider (u sliding mode control law)

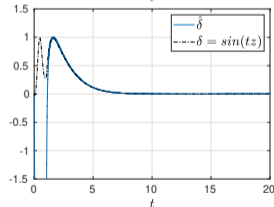
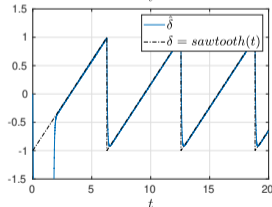
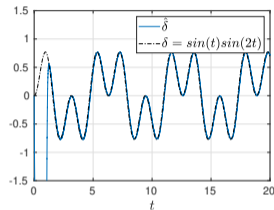
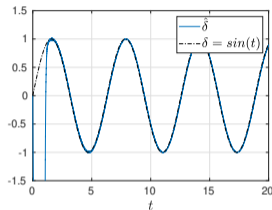
$$\dot{x} = x^3 + z,$$

$$\dot{z} = u + \delta(t, x, z)$$

$$\dot{\xi} = -\frac{1}{0.05} \xi + \frac{1}{0.05} \text{sign}(z - k(x))$$

Estimated disturbance:

$$\hat{\delta}(t, x, z) = (\sqrt{2} + 1)\xi$$



Section 5

Output Tracking

Output Tracking

So far:

- (Asymptotic) Stabilization of the origin

Now:

- Tracking of a reference signal

Output Tracking

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- (Asymptotic) Stabilization of the origin

Now:

- Tracking of a reference signal

In particular, consider

- $y = x, \quad \dot{x} = x^3 + z, \quad \dot{z} = u + \delta(t, x, z)$
- reference signal $y_r : \mathbb{R}_{\geq 0} \rightarrow \mathbb{R}$ twice cont. diff.

Goal:

- $y(t) \rightarrow y_r(t)$ for $t \rightarrow \infty$

Output Tracking

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Controller design:

- Define error dynamics

$$e(t) = y_r(t) - y(t) \quad \text{and demand} \quad e(t) \xrightarrow{t \rightarrow \infty} 0$$

- and the requirement that $e(t) \rightarrow 0$ for $t \rightarrow \infty$.

- Define the sliding variable (based on $\dot{e} = -e$)

$$\sigma = \dot{e} + e = \dot{y}_r - \dot{y} + y_r - y = \dot{y}_r + y_r - x^3 - z - x$$

- Calculating the time derivative:

$$\dot{\sigma} = \ddot{y}_r + \dot{y}_r - 3x^2\dot{x} - \dot{z} - \dot{x}$$

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- Candidate Lyapunov function $V(\sigma) = \frac{1}{2}\sigma^2$:

$$\dot{V}(\sigma) = \dot{\sigma}\sigma$$

$$= \sigma(\ddot{y}_r + \dot{y}_r - 3x^2\dot{x} - \dot{z} - \dot{x} - u - \delta(t, x, z) - x^3 - z)$$

- Define the input (with new degree of freedom v):

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$$\psi(t, x, z) = \ddot{y}_r + \dot{y}_r - \delta(t, x, z)$$

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$$\psi(t, x, z) = \ddot{y}_r + \dot{y}_r - \delta(t, x, z)$$

- Then, the candidate Lyapunov functions satisfies

$$\dot{V}(\sigma) = \sigma(\psi(t, x, z) + v)$$

- Assume that $|\psi(t, x, z)| \leq L_\psi$, for $L_\psi > 0$ & define

$$v = -(L_\psi + \frac{\kappa}{\sqrt{2}}) \text{sign}(\sigma), \quad (\kappa > 0)$$

- Then $\dot{V}(\sigma)$ satisfies

$$\begin{aligned} \dot{V}(\sigma) &= \sigma(\psi(t, x, z) - (L_\psi + \frac{\kappa}{\sqrt{2}}) \text{sign}(\sigma)) \\ &\leq |\sigma|L_\psi - |\sigma| \left(L_\psi + \frac{\kappa}{\sqrt{2}} \right) = -\kappa\sqrt{V(x)} \end{aligned}$$

Output Tracking

Example

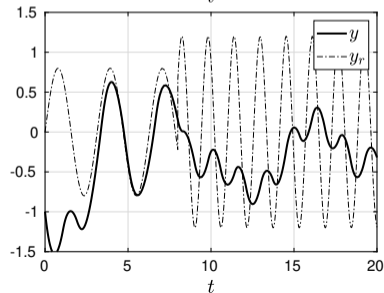
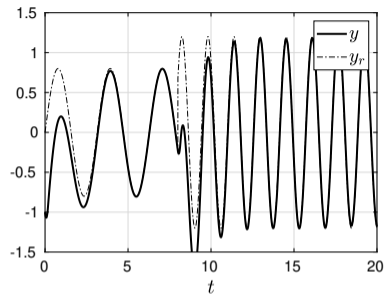
Consider the system

$$y = x, \quad \dot{x} = x^3 + z, \quad \dot{z} = u + \delta(t, x, z)$$

with output together with the reference signal

$$y_r(t) = \begin{cases} 0.8 \sin(2t) & \text{for } t < 8, \\ 1.2 \sin(4t) & \text{for } t \geq 8. \end{cases}$$

- y_r is twice continuously differentiable for all $t \neq 8$
- For the simulation, the disturbance $\delta(t, x, z) = \sin(t)$ is used
- The new disturbance $\psi = \ddot{y}_r + \dot{y}_r - \delta$ satisfies $|\psi(t, x, z)| \leq 25$ for all $t \neq 8$.
- Top figure: $L_\psi = 25$
- Bottom figure: $L_\psi = 1$
- Additionally: $\kappa = 2$



Introduction to Nonlinear Control

Stability, control design, and estimation

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Part III:

Chapter 10: Sliding Mode Control

- 10.1 Finite-Time Stability
- 10.2 Basic Sliding Mode Control
- 10.3 A More General Setting
- 10.4 Estimating the Disturbance
- 10.5 Output Tracking



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