



УНИВЕРСИТЕТ ИТМО

Факультет систем управления и робототехники

Nonlinear systems

Zimenko Konstantin



5. Feedback linearization

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✓ Motivation

Why and when do we need linearization ?

5. Feedback linearization

✓ Motivation

Example...

$$\begin{aligned}\dot{x}_1 &= x_2 \\ \dot{x}_2 &= -a[\sin(x_1 + \delta) - \sin \delta] - bx_2 + cu\end{aligned}$$

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$$u = \frac{a}{c} [\sin(x_1 + \delta) - \sin \delta] + \frac{1}{c} [-k_1x_1 - k_2x_2]$$

5. Feedback linearization

✓ Generalize

Consider the nonlinear system

$$\dot{x} = f(x) + G(x)u$$

$$f(0) = 0, x \in \mathbb{R}^n, u \in \mathbb{R}^m$$

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$$u = \alpha(x) + \gamma^{-1}(x)v \quad \longrightarrow \quad \dot{z} = Az + Bv$$

$$v = -Kz$$

Design K such that $(A - BK)$ is Hurwitz

5. Feedback linearization

$$u = \alpha(x) - \gamma^{-1}(x)KT(x)$$

Closed-loop system in the x -coordinates:

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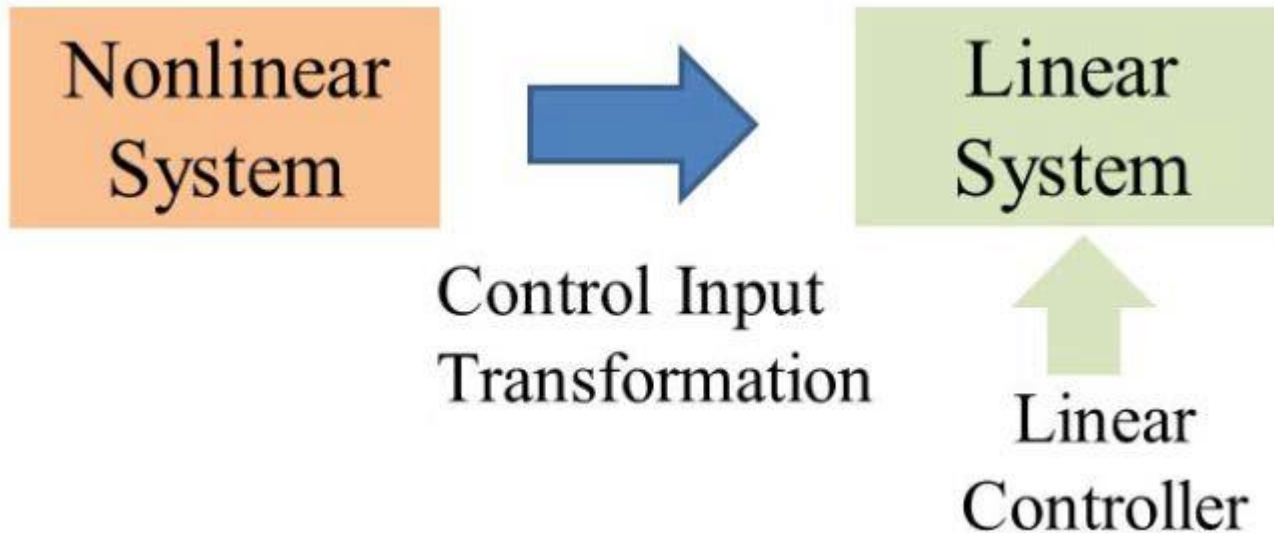
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In fact, we have

$$\hat{\alpha}, \hat{\gamma}, \hat{T}$$

$$u = \hat{\alpha}(x) - \hat{\gamma}^{-1}(x)K\hat{T}(x)$$

5. Feedback linearization

Closed-loop system:

$$u = \hat{\alpha}(x) - \hat{\gamma}^{-1}(x)K\hat{T}(x)$$

$$\dot{z} = (A - BK)z + B\delta(z)$$

$$\delta = \gamma[\hat{\alpha} - \alpha + \gamma^{-1}KT - \hat{\gamma}^{-1}K\hat{T}]$$

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where: $\hat{\alpha}$, $\hat{\gamma}$, \hat{T} are nominal models of α , γ and T .

$$V(z) = z^T P z,$$

$$P(A - BK) + (A - BK)^T P = -I$$

If $\|\delta(z)\| \leq k\|z\|$ for all z , where

$$0 \leq k < \frac{1}{2\|PB\|}$$

then the origin is globally exponentially stable

5. Feedback linearization

✓ Example: pendulum equation

$$\ddot{\theta} = -a \sin \theta - b \dot{\theta} + cT$$

$$x_1 = \theta - \delta, \quad x_2 = \dot{\theta}, \quad u = T - T_{ss} = T - \frac{a}{c} \sin \delta,$$

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$$\dot{x}_2 = -a[\sin(x_1 + \delta) - \sin \delta] - bx_2 + cu$$

$$u = \frac{1}{c} \{a[\sin(x_1 + \delta) - \sin \delta] - k_1 x_1 - k_2 x_2\}$$

$$A - BK = \begin{bmatrix} 0 & 1 \\ -k_1 & -(k_2 + b) \end{bmatrix} \quad \text{is Hurwith}$$

5. Feedback linearization

- ✓ Example: pendulum equation

$$T = u + \frac{a}{c} \sin \delta = \frac{1}{c} [a \sin(x_1 + \delta) - k_1 x_1 - k_2 x_2]$$

Let \hat{a} and \hat{c} be nominal models of a and c

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Let \hat{a} and \hat{c} be nominal models of a and c

$$T = \frac{1}{\hat{c}} [\hat{a} \sin(x_1 + \delta) - k_1 x_1 - k_2 x_2]$$

$$\dot{x} = (A - BK)x + B\delta(x)$$

$$\delta(x) = \left(\frac{\hat{a}c - a\hat{c}}{\hat{c}} \right) \sin(x_1 + \delta) - \left(\frac{c - \hat{c}}{\hat{c}} \right) (k_1 x_1 + k_2 x_2)$$

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✓ Example: pendulum equation

$$\delta(x) = \left(\frac{\hat{a}c - a\hat{c}}{\hat{c}} \right) \sin(x_1 + \delta_1) - \left(\frac{c - \hat{c}}{\hat{c}} \right) (k_1 x_1 + k_2 x_2)$$

$$|\delta(x)| \leq k\|x\| + \varepsilon$$

$$k = \left| \frac{\hat{a}c - a\hat{c}}{\hat{c}} \right| + \left| \frac{c - \hat{c}}{\hat{c}} \right| \sqrt{k_1^2 + k_2^2}, \quad \varepsilon = \left| \frac{\hat{a}c - a\hat{c}}{\hat{c}} \right| |\sin(x_1 + \delta_1)|$$

$$P = \begin{bmatrix} p_{11} & p_{12} \\ p_{12} & p_{22} \end{bmatrix}, \quad PB = \begin{bmatrix} p_{12} \\ p_{22} \end{bmatrix}$$

$$k < \frac{1}{2\sqrt{p_{12}^2 + p_{22}^2}}$$

$$\sin \delta_1 = 0 \Rightarrow \varepsilon = 0$$

6. Backstepping



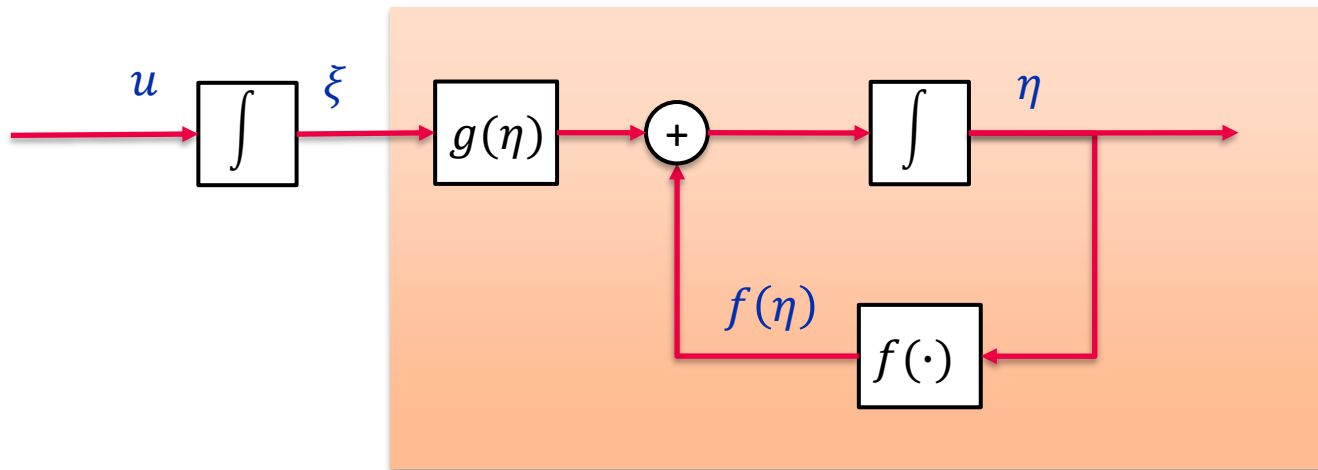
6. Backstepping

Consider the system

$$\dot{\eta} = f(\eta) + g(\eta)\xi$$

$$\dot{\xi} = u$$

$$\eta \in \mathbb{R}^n, \quad \xi, u \in \mathbb{R}$$



6. Backstepping

Consider the system

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Stabilize the origin using state feedback

View ξ as “virtual” control input to

$$\dot{\eta} = f(\eta) + g(\eta)\xi$$

Suppose there is $\xi = \phi(\eta)$ that stabilizes the origin of

$$\dot{\eta} = f(\eta) + g(\eta)\phi(\eta)$$

With the Lyapunov function in the form

$$\frac{\partial V}{\partial \eta} [f(\eta) + g(\eta)\phi(\eta)] \leq -W(\eta), \quad \forall \eta \in \mathbb{D}$$

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Suppose there is $\xi = \phi(\eta)$ that stabilizes the origin of

$$\dot{\eta} = f(\eta) + g(\eta)\phi(\eta)$$

Transform the original system by adding and subtracting $g(\eta)\phi(\eta)$

$$\begin{aligned}\dot{\eta} &= f(\eta) + g(\eta)\xi + g(\eta)\phi(\eta) - g(\eta)\phi(\eta) \\ \dot{\xi} &= u\end{aligned}$$

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$$\begin{aligned}\dot{\eta} &= f(\eta) + g(\eta)\xi + g(\eta)\phi(\eta) - g(\eta)\phi(\eta) \\ \dot{\xi} &= u \\ \dot{\eta} &= [f(\eta) + g(\eta)\phi(\eta)] + g(\eta)[\xi - \phi(\eta)] \\ \dot{\xi} &= u\end{aligned}$$

And we will introduce a replacement

$$z = \xi - \phi(\eta)$$

6. Backstepping

We get:

$$\dot{\eta} = [f(\eta) + g(\eta)\phi(\eta)] + g(\eta)z$$
$$\dot{z} = u - \dot{\phi}$$

$$\dot{\phi}(\eta) = \frac{\partial \phi(\eta)}{\partial \eta} \dot{\eta} \quad \longrightarrow \quad \dot{\phi}(\eta) = \frac{\partial \phi(\eta)}{\partial \eta} [f(\eta) + g(\eta)\xi]$$

Taking $v = u - \dot{\phi}$ reduces the system to the cascade connection

$$\dot{\eta} = [f(\eta) + g(\eta)\phi(\eta)] + g(\eta)z$$
$$\dot{z} = v$$

now the first component has an asymptotically stable origin when the input is zero.

6. Backstepping

Consider the Lyapunov function

$$V_c(\eta, \xi) = V(\eta) + \frac{1}{2}z^2$$

Its derivative is equal to

$$\begin{aligned}\dot{V}_c &= \frac{\partial V}{\partial \eta} [f(\eta) + g(\eta)\phi(\eta)] + \frac{\partial V}{\partial \eta} g(\eta)z + zv \\ &\leq -W(\eta) + \frac{\partial V}{\partial \eta} g(\eta)z + zv\end{aligned}$$

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Choosing

$$v = -\frac{\partial V}{\partial \eta} g(\eta) - kz \quad k > 0$$

Yields

$$\dot{V}_c \leq -W(\eta) - kz^2$$

6. Backstepping

$$\dot{\eta} = f(\eta) + g(\eta)\xi$$

$$\dot{\xi} = u$$

Substituting for u , v and $\dot{\phi}$ we obtain the state feedback control law:

$$v = u - \dot{\phi} \quad z = \xi - \phi(\eta)$$

$$\dot{\phi}(\eta) = \frac{\partial \phi(\eta)}{\partial \eta} [f(\eta) + g(\eta)\xi]$$

$$v = -\frac{\partial V}{\partial \eta} g(\eta) - kz$$

$$u = \frac{\partial \phi(\eta)}{\partial \eta} [f(\eta) + g(\eta)\xi] - \frac{\partial V}{\partial \eta} g(\eta) - k[\xi - \phi(\eta)]$$

$$\dot{\eta} = f(\eta) + g(\eta)\xi \quad (*)$$

$$\dot{\xi} = u$$

$$\frac{\partial V}{\partial \eta} [f(\eta) + g(\eta)\phi(\eta)] \leq -W(\eta), \quad (**)$$

$$u = \frac{\partial \phi(\eta)}{\partial \eta} [f(\eta) + g(\eta)\xi] - \frac{\partial V}{\partial \eta} g(\eta) - k[\xi - \phi(\eta)] \quad (***)$$

Lemma

Consider the system (*). Let $\phi(\eta)$ be a stabilizing state feedback control for (*) with $\phi(0) = 0$, and $V(\eta)$ be a Lyapunov function that satisfies (**) with some definite function $W(\eta)$. Then, the state feedback control (***) stabilizes the origin of (*), with $V(\eta) + \frac{1}{2}[\xi - \phi(\eta)]^2$ as a Lyapunov function. Moreover, if all the assumptions hold globally and $V(\eta)$ is radially unbounded, the origin will be globally asymptotically stable.

6. Backstepping

✓ Example 1

Consider the system:

$$\dot{x}_1 = x_1^2 - x_1^3 + \underline{x_2}$$

$$\dot{x}_2 = u$$

$$x_2 = \phi(x_1) = -x_1^2 - x_1 \quad \longrightarrow \quad \dot{x}_1 = -x_1 - x_1^3$$

$$V(x_1) = \frac{1}{2}x_1^2 \quad \longrightarrow \quad \dot{V} = -x_1^2 - x_1^4 \quad \forall x_1 \in \mathbb{R}$$

Next, we will introduce a replacement:

$$z_2 = x_2 - \phi(x_1) = x_2 + x_1^2 + x_1$$

$$\dot{x}_1 = -x_1 - x_1^3 + z_2$$

$$\dot{z}_2 = u + (1 + 2x_1)(-x_1 - x_1^3 + z_2)$$

6. Backstepping

✓ Example 1

$$\dot{x}_1 = -x_1 - x_1^3 + z_2$$

$$\dot{z}_2 = u + (1 + 2x_1)(-x_1 - x_1^3 + z_2)$$

$$V_c(x) = V(x) + \frac{1}{2}z^2$$
$$V(x) = \frac{1}{2}x_1^2$$

$$V_c(x) = \frac{1}{2}x_1^2 + \frac{1}{2}z_2^2$$

$$\dot{V}_c = x_1(-x_1 - x_1^3 + z_2) + z_2[u + (1 + 2x_1)(-x_1 - x_1^3 + z_2)]$$

$$\dot{V}_c = -x_1^2 - x_1^4 + z_2[x_1 + u + (1 + 2x_1)(-x_1 - x_1^3 + z_2)]$$

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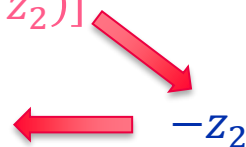
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$$\dot{V}_c = x_1(-x_1 - x_1^3 + z_2) + z_2[u + (1 + 2x_1)(-x_1 - x_1^3 + z_2)]$$

$$\dot{V}_c = -x_1^2 - x_1^4 + z_2[x_1 + u + (1 + 2x_1)(-x_1 - x_1^3 + z_2)]$$

$$u = -x_1 - (1 + 2x_1)(-x_1 - x_1^3 + z_2) - z_2$$



6. Backstepping

✓ Example 1

$$u = -x_1 - (1 + 2x_1)(-x_1 - x_1^3 + z_2) - z_2$$

$$z_2 = x_2 + x_1^2 + x_1$$

$$\dot{V}_c = -x_1^2 - x_1^4 - z_2^2$$

$$\dot{V}_c = -x_1^2 - x_1^4 - (x_2 + x_1 + x_1^2)^2$$

6. Backstepping

✓ Example 1

For system:

$$\dot{x}_1 = x_1^2 - x_1^3 + x_2$$

$$\dot{x}_2 = u$$

$$u = -x_1 - (1 + 2x_1)(-x_1^3 + x_2 + x_1^2) - (x_2 + x_1^2 + x_1)$$

the origin $x = 0$ is globally asymptotically stable

6. Backstepping

✓ Example 2

Consider the third-order system:

$$\begin{aligned}\dot{x}_1 &= x_1^2 - x_1^3 + x_2 \\ \dot{x}_2 &= x_3 \\ \dot{x}_3 &= u\end{aligned}$$

After one step of backstepping, we know that the second-order system:

$$\begin{aligned}\dot{x}_1 &= x_1^2 - x_1^3 + x_2 \\ \dot{x}_2 &= x_3\end{aligned}$$

with x_3 as input, can be globally stabilized by the control:

$$x_3 = -x_1 - (1 + 2x_1)(-x_1^3 + x_2 + x_1^2) - (x_2 + x_1^2 + x_1)$$

6. Backstepping

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$$x_3 = -x_1 - (1 + 2x_1)(-x_1^3 + x_2 + x_1^2) - (x_2 + x_1^2 + x_1) \stackrel{\text{def}}{=} \phi(x_1, x_2)$$

$$\text{with} \quad V(x_1, x_2) = \frac{1}{2}x_1^2 + \frac{1}{2}(x_2 + x_1^2 + x_1)^2$$

6. Backstepping

✓ Example 2

By analogy, we introduce a replacement:

$$z_3 = x_3 - \phi(x_1, x_2)$$

It was

$$\begin{aligned}\dot{x}_1 &= x_1^2 - x_1^3 + x_2 \\ \dot{x}_2 &= x_3 \\ \dot{x}_3 &= u\end{aligned}$$



$$\begin{aligned}\dot{x}_1 &= x_1^2 - x_1^3 + x_2 \\ \dot{x}_2 &= \cancel{x_3} - \cancel{\phi(x_1, x_2)} + \phi(x_1, x_2) \\ \dot{x}_3 &= u \\ \dot{x}_3 &= \dot{z}_3 + \dot{\phi}(x_1, x_2)\end{aligned}$$

We get:

$$\begin{aligned}\dot{x}_1 &= x_1^2 - x_1^3 + x_2 \\ \dot{x}_2 &= z_3 + \phi(x_1, x_2) \\ \dot{z}_3 &= u - \dot{\phi}(x_1, x_2)\end{aligned}$$

6. Backstepping

✓ Example 2

$$\dot{x}_1 = x_1^2 - x_1^3 + x_2$$

$$\dot{x}_2 = z_3 + \phi(x_1, x_2)$$

$$\dot{z}_3 = u - \dot{\phi}(x_1, x_2)$$



$$\dot{x}_1 = \boxed{x_1^2 - x_1^3 + x_2}$$

$$\dot{x}_2 = \boxed{z_3 + \phi}$$

$$\dot{z}_3 = u - \frac{\partial \phi}{\partial x_1} \boxed{(x_1^2 - x_1^3 + x_2)} - \frac{\partial \phi}{\partial x_2} \boxed{(z_3 + \phi)}$$

6. Backstepping

✓ Example 2

Consider the Lyapunov function:

$$V_c = V + \frac{1}{2}z_3^2$$

Its derivative is equal to:

$$\dot{V}_c = \frac{\partial V}{\partial x_1} \dot{x}_1 + \frac{\partial V}{\partial x_2} \dot{x}_2 + z_3 \dot{z}_3$$

$$\dot{V}_c = \frac{\partial V}{\partial x_1} (x_1^2 - x_1^3 + x_2) + \frac{\partial V}{\partial x_2} (z_3 + \phi) + z_3 \left(u - \dot{\phi} \frac{\partial \phi}{\partial x_1} (x_1^2 - x_1^3 + x_2) - \frac{\partial \phi}{\partial x_2} (z_3 + \phi) \right)$$

6. Backstepping

✓ Example 2

Consider the Lyapunov function:

$$V_c = V + \frac{1}{2}z_3^2$$

Its derivative is equal to:

$$\dot{V}_c = \frac{\partial V}{\partial x_1} \dot{x}_1 + \frac{\partial V}{\partial x_2} \dot{x}_2 + z_3 \dot{z}_3$$

$$\dot{V}_c = \frac{\partial V}{\partial x_1} (x_1^2 - x_1^3 + x_2) + \frac{\partial V}{\partial x_2} (z_3 + \phi) + z_3 \left(u - \phi \frac{\partial \phi}{\partial x_1} (x_1^2 - x_1^3 + x_2) - \frac{\partial \phi}{\partial x_2} (z_3 + \phi) \right)$$

$$\dot{V}_c = -x_1^2 - x_1^4 - (x_2 + x_1 + x_1^2)^2 + z_3 \left(\frac{\partial V}{\partial x_2} + u - \frac{\partial \phi}{\partial x_1} (x_1^2 - x_1^3 + x_2) - \frac{\partial \phi}{\partial x_2} (z_3 + \phi) \right)$$

6. Backstepping

✓ Example 2

We obtain:

$$u = -\frac{\partial V}{\partial x_2} + \frac{\partial \phi}{\partial x_1}(x_1^2 - x_1^3 + x_2) + \frac{\partial \phi}{\partial x_2}(z_3 + \phi) - \underbrace{z_3}_{\text{for a square}}$$

and

$$\dot{V}_c = -x_1^2 - x_1^4 - (x_2 + x_1 + x_1^2)^2 + z_3^2$$

6. Backstepping

✓ Generalize

$$\dot{x} = f_0(x) + g_0(x)z_1$$

$$\dot{z}_1 = f_1(x, z_1) + g_1(x, z_1)z_2$$

$$\dot{z}_2 = f_2(x, z_1, z_2) + g_2(x, z_1, z_2)z_3$$

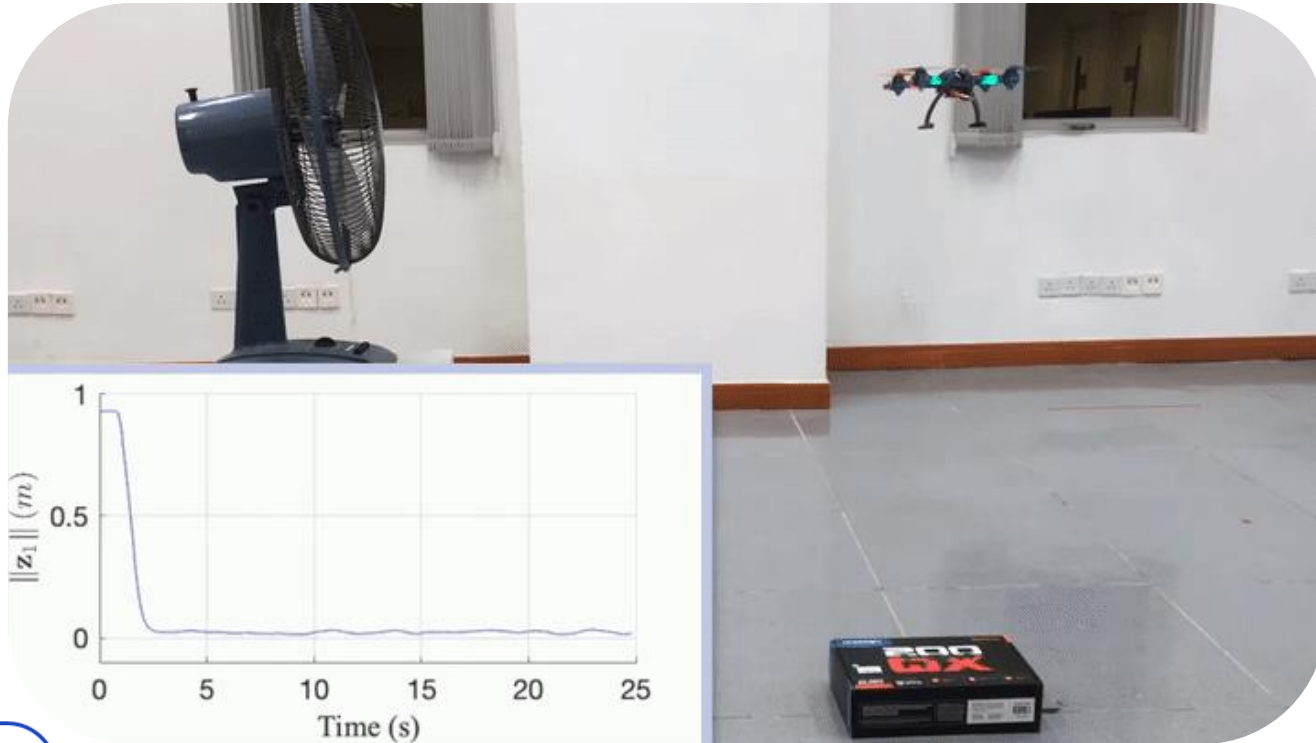
$$\vdots$$

$$\dot{z}_{k-1} = f_{k-1}(x, z_1, \dots, z_{k-1}) + g_{k-1}(x, z_1, \dots, z_{k-1})z_k$$

$$\dot{z}_k = f_k(x, z_1, \dots, z_k) + g_k(x, z_1, \dots, z_k)u$$

$$g_i(x, z_1, \dots, z_i) \neq 0 \quad \text{for } 1 \leq i \leq k$$

Adaptive Backstepping Control of a Quadcopter



7. Sliding-mode control

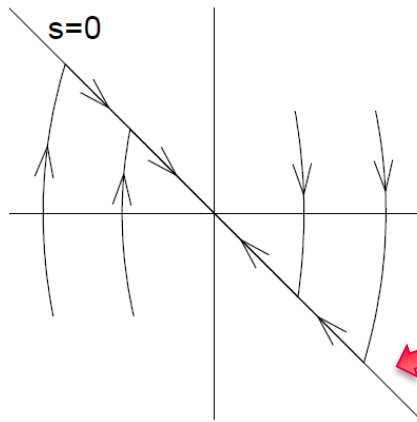


Consider the system :

$$\dot{x}_1 = x_2$$

$$\dot{x}_2 = h(x) + g(x)u \quad g(x) \geq g_0 > 0$$

Sliding Manifold (Surface):



$$s = a_1 x_1 + x_2 = 0$$

$$s(t) \equiv 0 \quad \longrightarrow \quad \dot{x}_1 = -a_1 x_1$$

$$a_1 > 0 \quad \longrightarrow \quad \lim_{t \rightarrow \infty} x_1(t) = 0$$


$$\dot{s} = a_1 \dot{x}_1 + \dot{x}_2 = a_1 x_2 + h(x) + g(x)u$$

Suppose: $\left| \frac{a_1 x_2 + h(x)}{g(x)} \right| \leq \varrho(x)$

With $V = \frac{1}{2}s^2$ as a Lyapunov function candidate we have:

$$\dot{V} = s\dot{s} = s[a_1 x_2 + h(x)] + g(x)su \leq g(x)|s|\varrho(x) + g(x)su$$

Taking $u = -\beta \operatorname{sgn}(s)$ where $\beta(x) \geq \varrho(x) + \beta_0, \beta_0 > 0$

 $s > 0, \quad u = -\beta(x)$



$$\dot{V} \leq g(x)|s|\varrho(x) - g(x)\beta(x)|s|$$

$$\dot{V} \leq g(x)|s|\varrho(x) - g(x)(\varrho(x) + \beta_0)|s| = -g(x)\beta_0|s|$$

$$s < 0, \quad u = \beta(x)$$

$$\dot{V} \leq g(x)|s|\varrho(x) + g(x)su \quad \longrightarrow \quad \dot{V} \leq g(x)|s|\varrho(x) - g(x)\beta(x)|s|$$

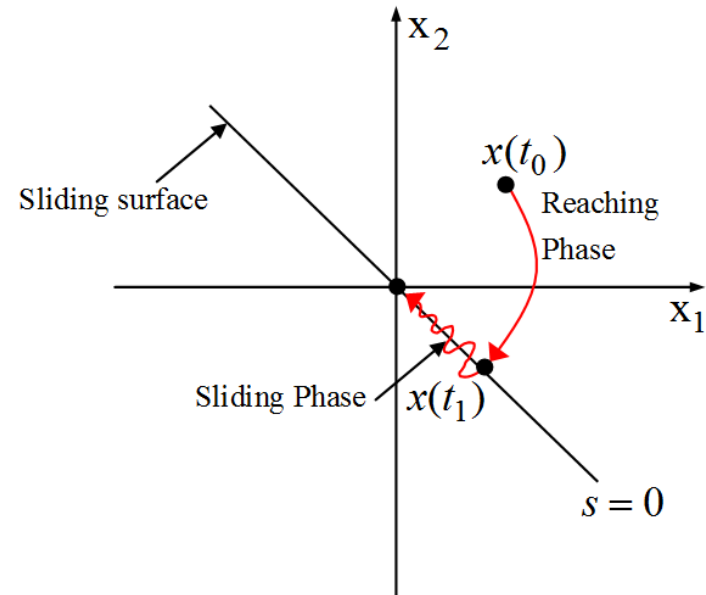
$$\dot{V} \leq g(x)|s|\varrho(x) - g(x)(\varrho(x) + \beta_0)|s| = -g(x)\beta_0|s|$$

$$\text{sign}(s) = \begin{cases} 1, & s > 0 \\ -1, & s < 0 \end{cases}$$

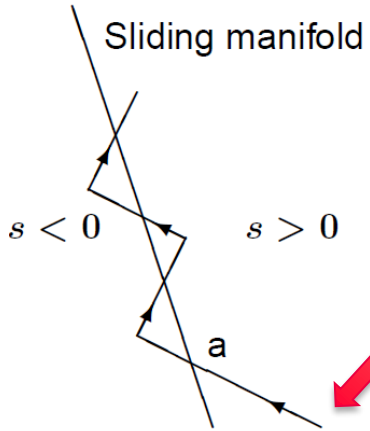
$$u = -\beta(x)\text{sign}(s)$$

$$\dot{V} \leq -g(x)\beta_0|s| \leq -g_0\beta_0|s|$$

$$\dot{V} \leq -g_0\beta_0\sqrt{2V}$$



7. Sliding-mode control



Chattering due to delay in control switching.

Solutions

1. divide the control into continuous and switching components

2. eliminate chattering is to replace the signum function by a high-slope saturation function

7. Sliding-mode control

1. Divide the control into continuous and switching components

$$\hat{h}(x) \xrightarrow{\text{Nominal models}} h(x)$$

$$\hat{g}(x) \xrightarrow{\text{Nominal models}} g(x)$$

$$\text{Taking: } u = \frac{[a_1 x_2 + \hat{h}(x)]}{\hat{g}(x)} + v$$

$$s' = a_1 \left[1 - \frac{g(x)}{\hat{g}(x)} \right] x_2 + h(x) - \frac{g(x)}{\hat{g}(x)} \hat{h}(x) + g(x)v \stackrel{\text{def}}{=} \delta(x) + g(x)$$

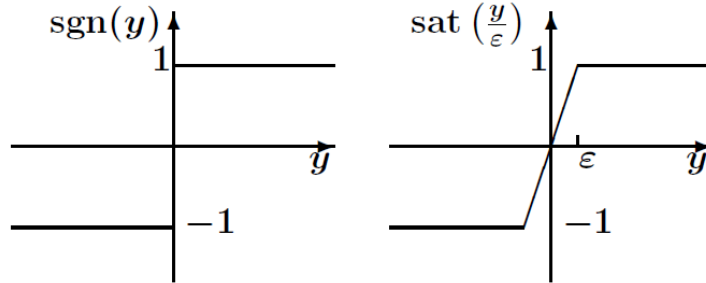
$$\text{If: } \left| \frac{\delta(x)}{g(x)} \right| \leq \varrho(x) \xrightarrow{\text{Nominal models}} v = -\beta(x) \text{sgn}(s)$$

7. Sliding-mode control

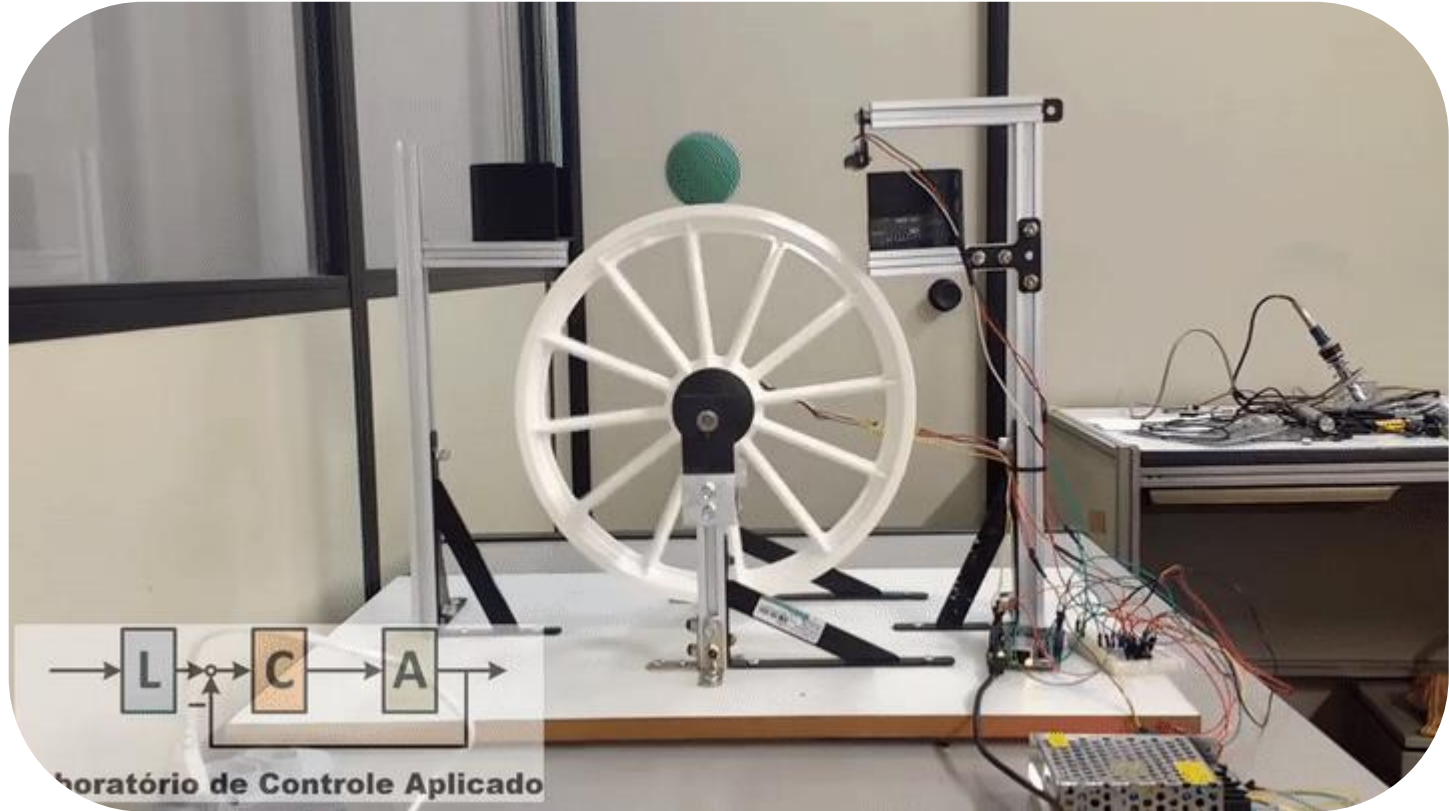
2. Eliminate chattering is to replace the signum function by a high-slope saturation function

$$u = -\beta(x)\text{sat}\left(\frac{s}{\varepsilon}\right)$$

$$\text{sat}(y) = \begin{cases} y, & \text{if } |y| \leq 1 \\ \text{sgn}(y), & \text{if } |y| > 1 \end{cases}$$



Sliding-Mode Control of a Ball on Wheel System



Stabilization of a Ball on Sphere System Using Sliding Mode Control



The background is a complex, low-poly geometric pattern in various shades of blue, ranging from dark navy to light sky blue. The shapes are irregular polygons that fit together like a mosaic. In the lower center, there is a white rounded rectangular border containing the text 'IT'sMOre than a UNIVERSITY'.

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