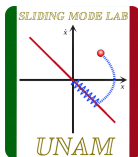


Main Ideas of Second Order Sliding Mode Control

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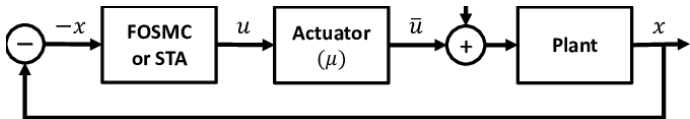
Idea of SOSMC.

Idea

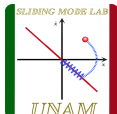
Substitute a discontinuous control with a continuous one.

Presence of actuators.

Presence of actuator is growing the relative degree of the system.

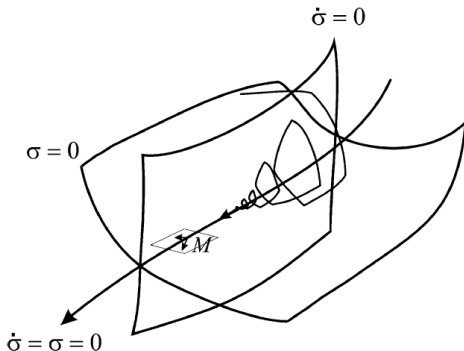


Idea of SOSM.



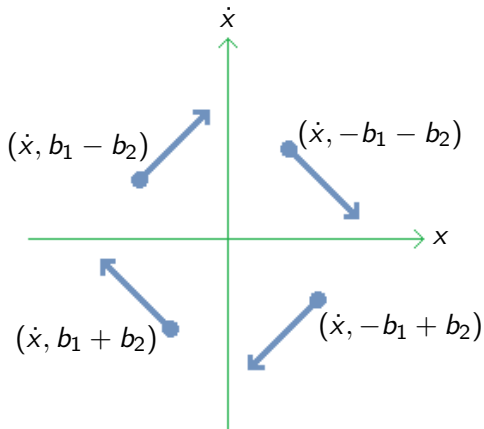
Main Idea

The idea is to reach $\sigma = 0$ and $\dot{\sigma} = 0$.

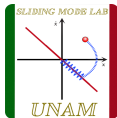


Twisting phase plane

$$\ddot{x} = -b_1 \text{sign}(x) - b_2 \text{sign}(\dot{x}), \quad b_1 > b_2 > 0$$



Twisting Algorithm Convergence



To describe the system behavior in the phase plane is convenient

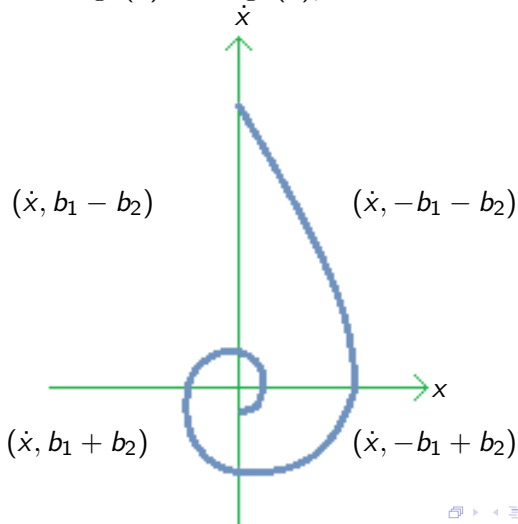
$$\ddot{x} = \frac{d\dot{x}}{dt} = \frac{d\dot{x}}{dx} \frac{dx}{dt} = \frac{d\dot{x}}{dx} \dot{x} = -b_1 \text{sign}(x) - b_2 \text{sign}(\dot{x})$$

or

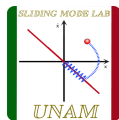
$$\frac{d\dot{x}}{dx} = \frac{-b_1 \text{sign}(x) - b_2 \text{sign}(\dot{x})}{\dot{x}}$$

Trajectories of Twisting Algorithm

$$\ddot{x} = -b_1 \text{sign}(x) - b_2 \text{sign}(\dot{x}), \quad b_1 > b_2 > 0$$



Twisting Algorithm Convergence



$$\dot{x}d\dot{x} = -(b_1 + b_2)dx$$

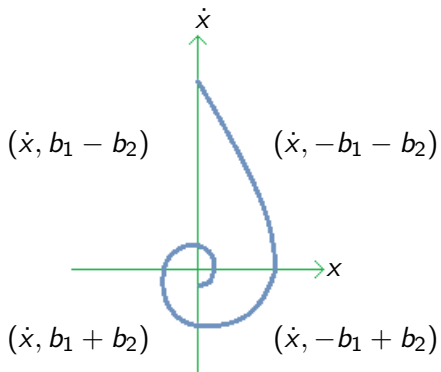
$$\frac{1}{2}\dot{x}^2 = -(b_1 + b_2)x - x_1$$

for $\dot{x} > 0$

$$x = x_1 - \frac{\dot{x}^2}{2(b_1 + b_2)}$$

for $\dot{x} \leq 0$

$$x = x_1 - \frac{\dot{x}^2}{2(b_1 - b_2)}$$



Twisting Algorithm Convergence

For $x = 0, \dot{x} = \dot{x}_0$

$$0 = x_1 - \frac{\dot{x}_0^2}{2(b_1 + b_2)}$$

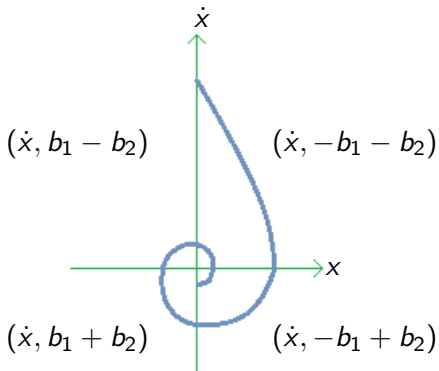
$$\Rightarrow \dot{x}_0^2 = 2(b_1 + b_2)x_1$$

For $x = 0, \dot{x} = \dot{x}_1$

$$\dot{x}_1^2 = 2(b_1 - b_2)x_1$$

Convergence rate

$$\frac{|\dot{x}_1|}{|\dot{x}_0|} = \sqrt{\frac{b_1 - b_2}{b_1 + b_2}} := q < 1$$

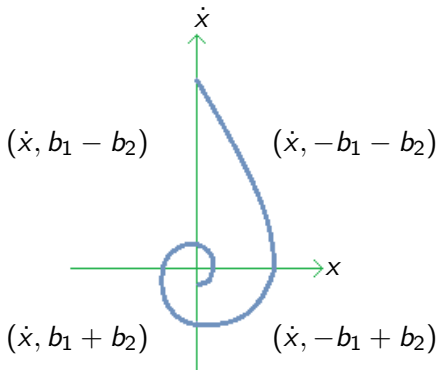


Twisting Algorithm Convergence

Extending the trajectory to $x < 0$ and using the same reasoning in successive crossing of $x = 0$ axis we obtained

$$\frac{|\dot{x}_{i+1}|}{|\dot{x}_i|} = q < 1$$

thus, the algorithm converge to origin. The real trajectory consist of infinite number of segments belonging to $x \geq 0$ and $x \leq 0$, the convergence time can be estimated.



Twisting Algorithm Convergence

$$\dot{x} > 0, x > 0,$$

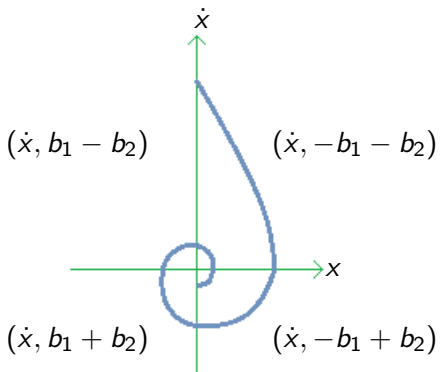
$$\ddot{x} = -b_1 \text{sign}(x) - b_2 \text{sign}(\dot{x})$$

$$\dot{x}(t) = -(b_1 + b_2)t + \dot{x}_0$$

$$\dot{x}(t_1^+) = 0 \implies t_1^+ = \frac{\dot{x}_0}{b_1 + b_2}.$$

$$x(t) = -\frac{1}{2}(b_1 - b_2)t^2 + x_1$$

$$\begin{aligned} t_1^- &= \sqrt{\frac{2x_1}{b_1 - b_2}}, \\ &= \sqrt{\frac{1}{(b_1 - b_2)(b_1 + b_2)}} \dot{x}_0 \end{aligned}$$



Twisting Algorithm Convergence

Then the time

$$t_1 := t_1^+ + t_1^- = \eta \dot{x}_0, \text{ con } \eta = \frac{1}{b_1 + b_2} + \sqrt{\frac{1}{(b_1 - b_2)(b_1 + b_2)}}$$

corresponds to the trajectory $\dot{x}_0 x_1 \dot{x}_1$. In the same way, the time

$$t_i = \eta |\dot{x}_{i-1}| = \eta q^{i-1} \dot{x}_0$$

corresponds to the trajectory $\dot{x}_{i-1} x_i \dot{x}_i$. In this case, the total convergence time is

$$T = \sum_{i=1}^{\infty} t_i = \sum_{i=1}^{\infty} \eta |\dot{x}_{i-1}| = \sum_{i=1}^{\infty} \eta q^{i-1} \dot{x}_0 = \frac{\eta \dot{x}_0}{1 - q}$$

Zeno phenomenon

Twisting Algorithm for Perturbed Systems

$$\ddot{x} = a(t, x) + b(t, x)u, \quad |a(t, x)| \leq C, \quad 0 < K_m \leq b(t, x) \leq K_M,$$

The control

$$u = -b_1 \text{sign}(x) - b_2 \text{sign}(\dot{x}), \quad b_1 > b_2 > 0$$

Lemma

Let b_1 and b_2 satisfy the conditions

$$K_m(b_1 + b_2) - C > K_M(b_1 - b_2) + C, \quad K_m(b_1 - b_2) > C.$$

Then, the controller u provides for the appearance of a 2-sliding mode $x = \dot{x} = 0$ attracting the trajectories of the system in finite time.

Twisting Algorithm for Perturbed Systems

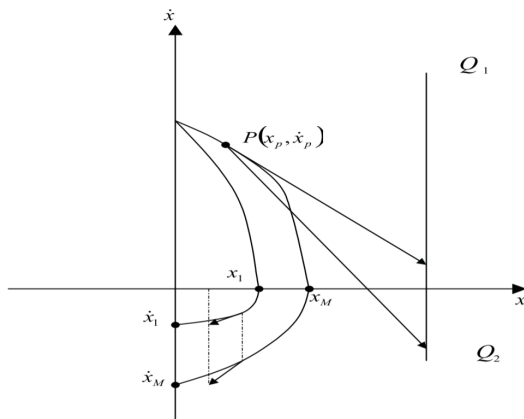
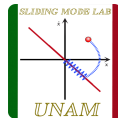
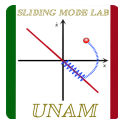


Figure: Majorant curves of the twisting controller.

Suboptimal Algorithm



where

$$r_1 - r_2 > \frac{C}{K_m},$$

$$r_1 + r_2 > \frac{4C + K_M(r_1 - r_2)}{3K_m},$$

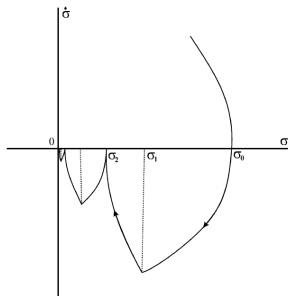
Consider

$$\ddot{\sigma} \in [-C, C] + [K_m, K_M] u,$$

The suboptimal controller is given by

$$u = r_1 \text{sign}(\sigma - \sigma^*/2) + r_2 \text{sign}(\sigma^*),$$

$$r_1 > r_2 > 0,$$



Suboptimal Algorithm

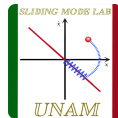
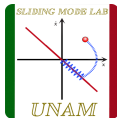


Figure: Bartolini's workgroup.

Terminal Sliding Mode Surface



$$\begin{aligned}\dot{x}_1 &= x_2, \quad \dot{x}_2 = u(x), \\ u(x) &= -\alpha \operatorname{sign}(s(x)), \\ s(x) &= x_2 + \beta \sqrt{|x_1|} \operatorname{sign}(x_1).\end{aligned}$$



Figure: Prof. Z. Man

Terminal Sliding Variable

Time derivative of the switching surface

$$\dot{s}(x) = \dot{x}_2 + \beta \frac{x_2}{2\sqrt{|x_1|}} = -\alpha \operatorname{sign}(s(x)) + \beta \frac{x_2}{2\sqrt{|x_1|}}.$$

- $s(x)$ is singular for $x_1 = 0$, and **the relative degree of the switching surface does not exist**
- On $x_2 = -\beta\sqrt{|x_1|} \operatorname{sign}(x_1)$

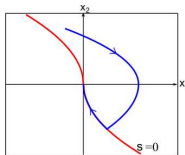
$$\dot{s} = -\alpha \operatorname{sign}(s(x)) - \frac{\beta^2}{2} \operatorname{sign}(x_1).$$

- Two types of behavior for the solution of the system are possible

Two types of behavior

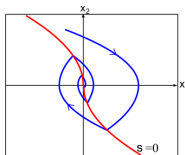
Terminal mode:

- $\beta^2 < 2\alpha$,
- Trajectories of the system reach the surface $s(x) = 0$ and remain there.

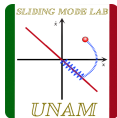


Twisting mode

- $\beta^2 > 2\alpha$
- Trajectories do not slide on the surface $s(x) = 0$



How to overcome the Singularity?



Singularity can be overcome by rewriting the function s

$$\bar{s}(x) = \beta^2 x_1 + x_2^2 \operatorname{sign}(x_2).$$

Quasi-Continuous Algorithm

This algorithm is given by

$$u = -\alpha \frac{\dot{\sigma} + \beta|\sigma|^{1/2}\text{sign}(\sigma)}{|\dot{\sigma}| + \beta|\sigma|^{1/2}}$$

where

$$\alpha, \beta > 0, \quad \alpha K_m - C > 0,$$

and the inequality

$$\alpha K_m - C - 2\alpha K_m \frac{\beta}{\rho + \beta} - \frac{1}{2}\rho^2 > 0,$$

must be satisfied for some positive $\rho > \beta$.

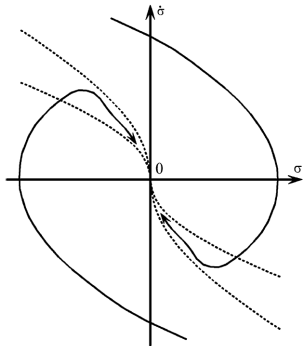
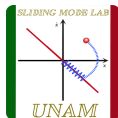


Figure: Trajectories of the quasi-continuous controller.

Anti-chattering Strategy



$$\dot{X} = F(t, X) + G(t, X)u, X \in R^n, u \in R, |F| < F^+,$$

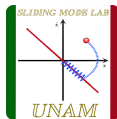
The switching variable $\sigma(X) : \dot{\sigma} = f(\sigma, t) + g(\sigma, t)u$.

Anti-chattering strategy:

Add an Integrator in control input:

If $\dot{u} = v = -a \operatorname{sign}(\dot{\sigma}(t)) - b \operatorname{sign}(\sigma(t))$, so u is a Lipschitz continuous control signal ensuring finite-time convergence to $\sigma = 0$

Anti-chattering Strategy

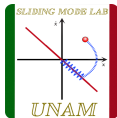


Criticism(1987) If it is possible to measure $\dot{\sigma} = f(t, \sigma) + g(t, \sigma)u$, then the uncertainty $f(t, \sigma) = \dot{\sigma} - g(t, \sigma)u$ is also known and can be compensated without any discontinuous control!

Counter-argument

If g is uncertain so $\ddot{\sigma}$ depends on u through uncertainty! The anti-chattering strategy is reasonable for the case of uncertain control gains.

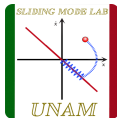
Discussion about SOSM



Advantages of SOSM

- 1 Allows to compensate **bounded** matched uncertainties for the systems with relative degree two **with discontinuous control signal**
- 2 Allows to compensate **Lipschitz** matched uncertainties with continuous control signal using the first derivative of sliding inputs
- 3 Ensures quadratic precision of convergence with respect to the sliding output
- 4 For one degree of freedom mechanical systems: the sliding surface design is no longer needed.
- 5 For systems with relative degree r : the order of the sliding dynamics is reduced up to $(r - 2)$. The design of the sliding surface of order $(r - 2)$ is still necessary!

OPEN PROBLEMS:EARLY 90th



- To reduce the chattering substituting **discontinuous control signal with continuous one** the derivative of the sliding input still needed!
- The problem of exact finite-time stabilization and exact disturbance compensation for SISO systems with arbitrary relative degree remains open. More deep decomposition is still needed
- Theoretically exact differentiators are needed to realize theoretically exact compensation of the Lipschitz matched uncertainties

The basic differential inclusion

Defining $x = (x_1, \dots, x_\rho)^T = (\sigma, \dot{\sigma}, \dots, \sigma^{(\rho-1)})^T$, $\sigma^{(i)} = \frac{d}{dt}h(z, t)$

- The regular form

$$\Sigma_T : \begin{cases} \dot{x}_i = x_{i+1}, & i = 1, \dots, \rho - 1, \\ \dot{x}_\rho = w(t, z) + b(t, z)u, & x_0 = x(0), \\ \dot{\zeta} = \phi(\zeta, x) & \zeta_0 = \zeta(0), \end{cases}$$

$$0 < K_m \leq b(t, z) \leq K_M, \quad |w(t, z)| \leq C.$$

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- Reduced Dynamics Asymptotically stable, i.e., $\dot{\zeta} = \phi(\zeta, 0) \zeta_0 = \zeta(0)$,

The basic differential inclusion

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- Reduced Dynamics Asymptotically stable, i.e., $\dot{\zeta} = \phi(\zeta, 0)$ $\zeta_0 = \zeta(0)$,
- The basic Differential Inclusion (DI)

$$\Sigma_{DI} : \begin{cases} \dot{x}_i = x_{i+1}, & i = 1, \dots, \rho - 1, \\ \dot{x}_\rho \in [-C, C] + [k_m, K_M]u, \end{cases}$$

The basic differential inclusion

If x_1 is a plain output then the zero dynamics does not exists.

Defining $x = (x_1, \dots, x_\rho)^T = (\sigma, \dot{\sigma}, \dots, \sigma^{(\rho-1)})^T, \sigma^{(i)} = \frac{d}{dt}h(z, t)$

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- The regular form

$$\Sigma_T : \begin{cases} \dot{x}_i = x_{i+1}, & i = 1, \dots, \rho - 1, \\ \dot{x}_\rho = w(t, z) + b(t, z)u, & x_0 = x(0), \end{cases}$$

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