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FRNT05 Nonlinear Control Systems and Servo Systems

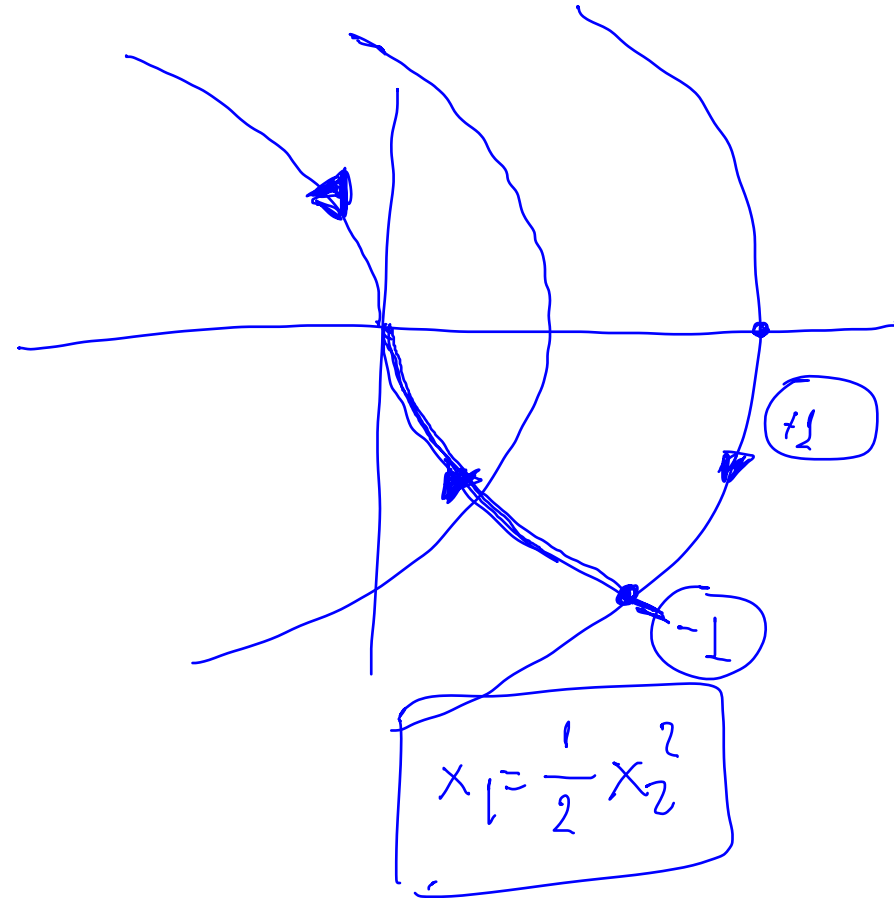
Lecture 13: Control Systems with discontinuities

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Outline

- ~~Time optimal control~~
- Sliding mode control
- Friction



Sliding set

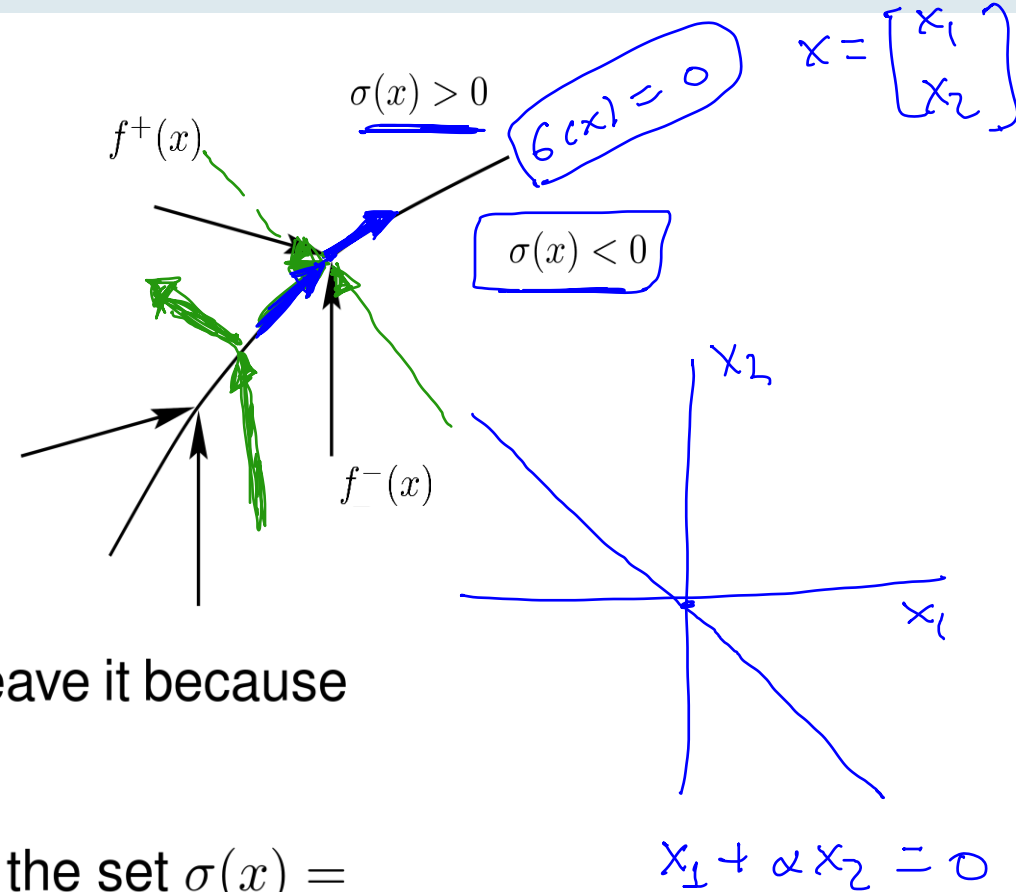
- Sliding set:**

→ $\sigma(x) = 0$

$$\left. \begin{aligned} \frac{\partial \sigma}{\partial x} f^+ &< 0 \\ \frac{\partial \sigma}{\partial x} f^- &> 0 \end{aligned} \right\}$$

f^+ and f^- point towards $\sigma(x) = 0$

$$\dot{x} = \begin{cases} f^+(x) & \sigma(x) > 0 \\ f^-(x) & \sigma(x) < 0 \end{cases}$$



- Once the trajectory hits the switching surface S , it cannot leave it because the vector fields on both sides point towards the surface
- If f^+ and f^- point “in the same direction” on both sides of the set $\sigma(x) = 0$, then the solution curves will just pass through and this region will not belong to the sliding set.
- If $f^+(x) = f^-(x)$ on $\sigma(x)$ then the sliding set is actually a set of equilibrium points

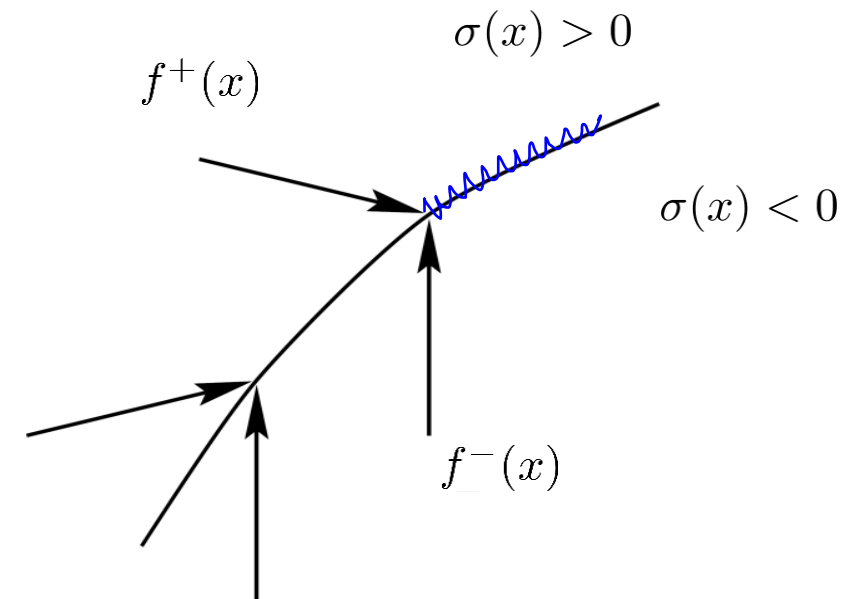
Sliding set

- The general behavior for the solution is to slide on $\sigma(x)$ (sliding mode.)
- The sliding motion can be described by noting that there is a unique convex combination of $f^+(x)$ and $f^-(x)$ that is tangent to $\sigma(x)$ at the point x .
- **Sliding dynamics:**

$$\dot{x} = \alpha(x)f^+(x) + (1 - \alpha(x))f^-(x),$$

where $\alpha(x)$ is obtained from

$$\begin{aligned} 0 &= \frac{d\sigma}{dt} = \frac{\partial \sigma}{\partial x} \cdot \dot{x} \\ &= \frac{\partial \sigma}{\partial x} f^-(x) + \alpha(x) \frac{\partial \sigma}{\partial x} [f^+(x) - f^-(x)]. \end{aligned}$$



- The fast switches will give rise to average dynamics that slide along the set where $\sigma(x) = 0$.

Example

$$\dot{x} = \begin{bmatrix} 0 & -1 \\ 1 & -1 \end{bmatrix} x + \begin{bmatrix} 1 \\ 1 \end{bmatrix} u = \underline{Ax} + \underline{Bu}$$

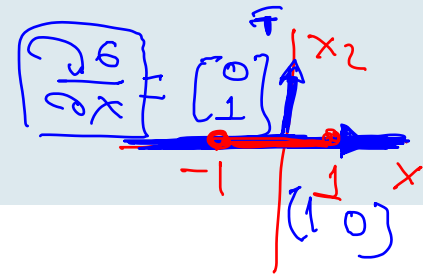
$$u = -\operatorname{sgn}\sigma(x) = -\operatorname{sgn}x_2 = -\operatorname{sgn}(Cx)$$

which means that

$$\dot{x} = \begin{cases} Ax - B, & x_2 > 0 \\ Ax + B, & x_2 < 0 \end{cases}$$

Determine the *sliding set* and the *sliding dynamics*.

Example: Sliding set



$$\begin{aligned}\dot{x}_1 &= -x_2 + u = -x_2 - \text{sgn}(x_2) \\ \dot{x}_2 &= x_1 - x_2 + u = x_1 - x_2 - \text{sgn}(x_2)\end{aligned}$$

$$\sigma(x) = 0 \quad f^+ = \begin{bmatrix} -x_2 - 1 \\ x_1 - x_2 - 1 \end{bmatrix} \quad f^- = \begin{bmatrix} -x_2 + 1 \\ x_1 - x_2 + 1 \end{bmatrix}$$

$$\sigma(x) = x_2$$

$$\begin{aligned}\sigma(x) &= 0 \\ \frac{\partial \sigma}{\partial x} f^+ &< 0 \\ \frac{\partial \sigma}{\partial x} f^- &> 0\end{aligned}$$

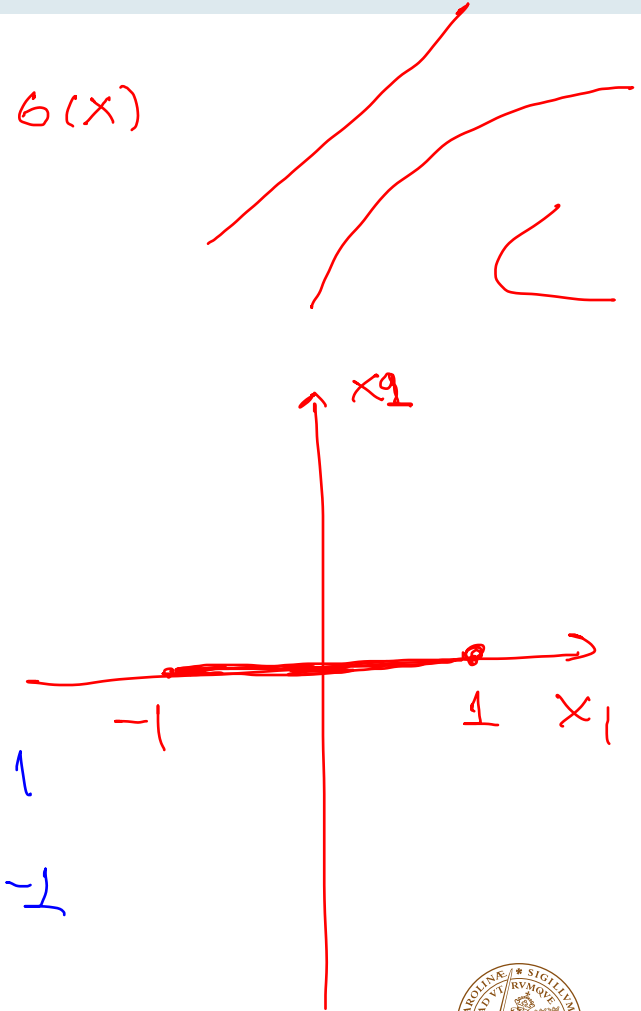
$$x_2 = 0$$

$$x_1 - x_2 - 1 < 0$$

$$x_1 - x_2 + 1 > 0$$

$$x_1 < 1$$

$$x_1 > -1$$



We thus have the sliding set $\{-1 < x_1 < 1, x_2 = 0\}$

Example: Sliding dynamics

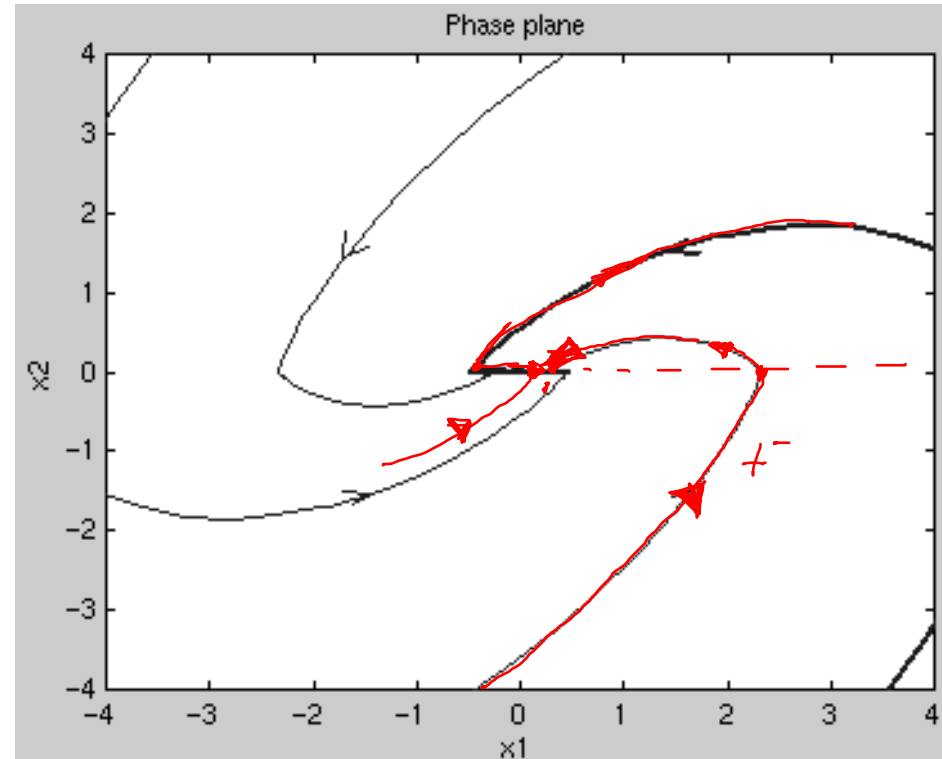
The *sliding dynamics* are the given by

$$\dot{x} = \alpha f^+ + (1 - \alpha) f^- \quad \text{sliding dynamics}$$
$$0 = \frac{\partial \sigma}{\partial x} f^-(x) + \alpha(x) \frac{\partial \sigma}{\partial x} [f^+(x) - f^-(x)] \quad \leftarrow$$

On the sliding set $\{-1 < x < 1, x_2 = 0\}$, this gives

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \alpha \begin{bmatrix} -x_2 - 1 \\ x_1 - x_2 - 1 \end{bmatrix} + (1 - \alpha) \begin{bmatrix} -x_2 + 1 \\ x_1 - x_2 + 1 \end{bmatrix}$$
$$0 = x_1 - x_2 + 1 - 2\alpha$$

Eliminating α gives $\begin{cases} \dot{x}_1 = -x_1 \\ \dot{x}_2 = 0 \end{cases}$ Hence, any initial condition the sliding set will give exponential convergence to $x_1 = x_2 = 0$.



The dynamics along the sliding set in $\sigma(x) = 0$ can also be obtained by finding $u = u_{\text{eq}} \in [-1, 1]$ such that $\dot{\sigma}(x) = 0$.
 u_{eq} is called the **equivalent control**.

Sliding mode control

- Define an output $y = \sigma(x)$ such that the relative degree of the input-output relationship is 1 and the origin for the the system $\sigma(x) = 0$ is stable.

Pendulum:
$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -\theta \cos(x_1) + bu \end{cases}$$

angle of pendulum
 $x_1 = \phi - \pi/2$
 $x_2 = \dot{\phi}$

$$\sigma = [a \ 1] \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \underline{a}x_1 + \underbrace{x_2}_{\dot{x}_1} \quad \longrightarrow \quad \boxed{\dot{x}_1 = -ax_1}$$

$$\sigma = x_2 + ax_1 \Rightarrow \boxed{\dot{\sigma}} = -\theta \cos(x_1) + \underbrace{ax_2}_{\dot{x}_1} + \underline{bu}$$

Sliding mode control

- Design a control input that makes the state starting in $\sigma(x) = 0$ to stay there for all t . ($\sigma(x) = 0$ invariant)

Pendulum:
$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -\theta \cos(x_1) + bu \end{cases}$$

$$\sigma = x_2 + ax_1 \Rightarrow \dot{\sigma} = -\theta \cos(x_1) + ax_2 + bu$$

$$\begin{aligned} \dot{\sigma} &= 0 & \sigma(x(0)) &= 0 \\ \sigma(x(t)) &= 0 \end{aligned}$$

$$u = \frac{1}{b} (-ax_2 + \theta \cos(x_1))$$

u_{eq.}

Sliding mode control

- Use feedback linearization and design a control input that makes $\sigma(x) = 0$.

Pendulum:
$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -\theta \cos(x_1) + bu \end{cases}$$

$$u_{\text{equiv}} = \frac{1}{b} (-ax_2 + \theta \cos(x_1))$$

Equivalent control

$$u = u_{\text{equiv}} + v$$

$$V = \frac{1}{2} \sigma^2$$

$$\dot{V} = \cancel{\frac{1}{2} \dot{\sigma}^2} = \dot{\sigma} [-\theta \cos(x_1) + ax_2 + bu_{\text{equiv}} + bv]$$

$$\boxed{\text{Set } v = -K\sigma} = -K\sigma^2$$

$$\sigma = x_2 + ax_1 \Rightarrow \dot{\sigma} = -\theta \cos(x_1) + ax_2 + bu$$

Sliding mode control

- Design a discontinuous control input v as part of the control $u = u_{\text{equiv}} + v$ that drives $\sigma(x)$ to zero in finite time.

$$\boxed{v = -K \text{sign}(\sigma)} \quad \text{sign}(\sigma) = \begin{cases} 1, & \sigma > 0 \\ -1, & \sigma < 0 \end{cases}$$

- Examine how $|\sigma(t)|$ changes over time

discontinuous
feedback
 $v = -K \text{sgn}(\sigma)$

$$\frac{d}{dt} |\sigma(t)| = \frac{d}{dt} \sqrt{\sigma^2} = \text{sign}(\sigma) \dot{\sigma} = -K_b$$

continuous
feedback
 $v = -k\sigma$

$$\frac{d}{dt} \sigma = -k_b \sigma \Rightarrow$$

$$\sigma(t) = e^{-k_b t} \sigma(0)$$

exponential convergence

$$|\sigma(t)| = |\sigma(0)| - k_b t \quad \text{for } t \leq \frac{|\sigma(0)|}{k_b}$$

$$\boxed{t_s = \frac{|\sigma(0)|}{k_b}}$$

The solution is not extended after this time instant.

Sliding mode control

- Choose K such that $u = \hat{u}_{eq} + v$ drives $\sigma(x)$ to zero in finite time.
 \hat{u}_{eq} is designed by using some estimate

$$\hat{\theta} \rightarrow u_{eq} = -\frac{a}{b}x_2 + \frac{\hat{\theta}}{b}\cos(x_1)$$

$$v = -K \text{sign}(\sigma)$$

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

By choosing K sufficiently large robustness to uncertainty

$$\dot{\sigma} = -\theta \cos x_1 + ax_2 + b(\hat{u}_{eq} + v)$$

$$\dot{\sigma} = -(\theta - \hat{\theta}) \cos x_1 - bK \text{sign}(\sigma)$$

$$V = \frac{1}{2} \sigma^2 \Rightarrow$$

$$\dot{V} = \sigma \dot{\sigma} = -bK|\sigma| - \sigma(\theta - \hat{\theta}) \cos(x_1)$$

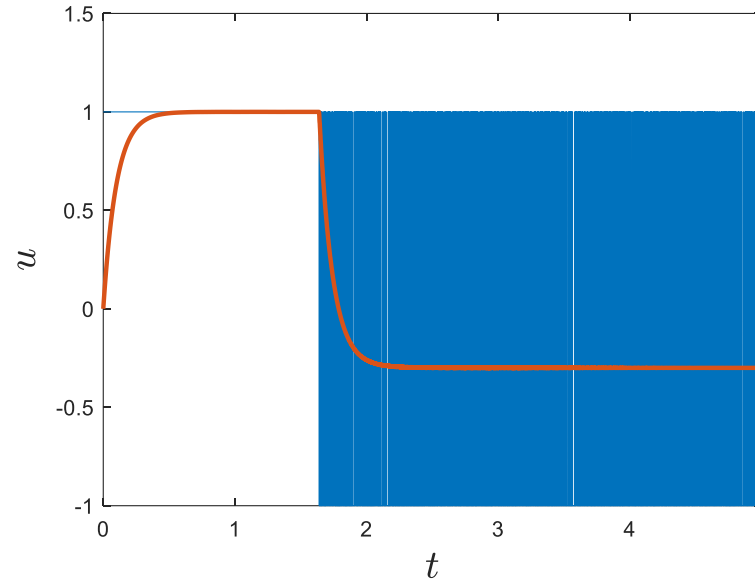
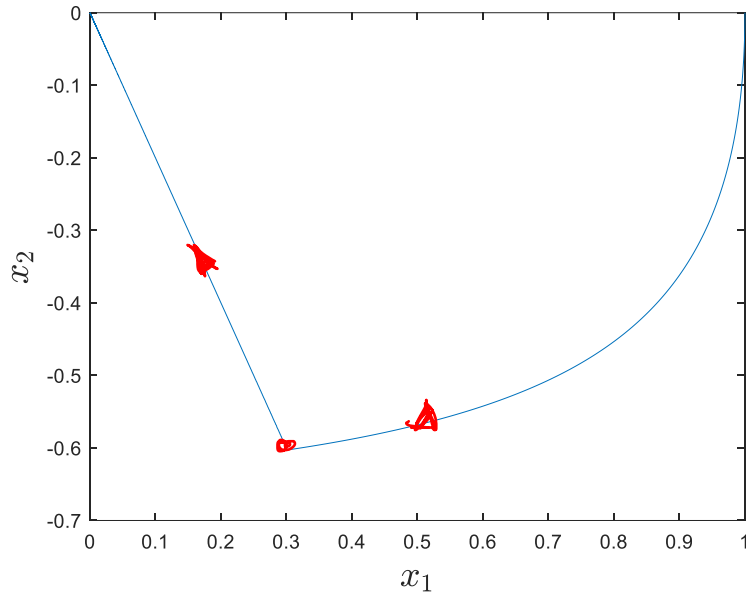
$$\leq -bK|\sigma| + \bar{\theta}|\sigma|$$

$$\leq -(bK - \bar{\theta})|\sigma| < 0 \text{ for } K > \frac{\bar{\theta}}{b}$$

$|\cos(x_1)| \leq 1$
 $|\theta - \hat{\theta}| \leq \bar{\theta}$
 "worst" case
 most positive



Example



- Chattering \rightarrow wear on mechanical devices

Robustness to uncertainty comes with the expense of high-frequency input signal

If it is smoothed there are steady state errors. See next slide

Pendulum:
$$\begin{cases} \dot{x}_1 = x_2 \\ \dot{x}_2 = -0.3 \cos(x_1) - u \end{cases}$$

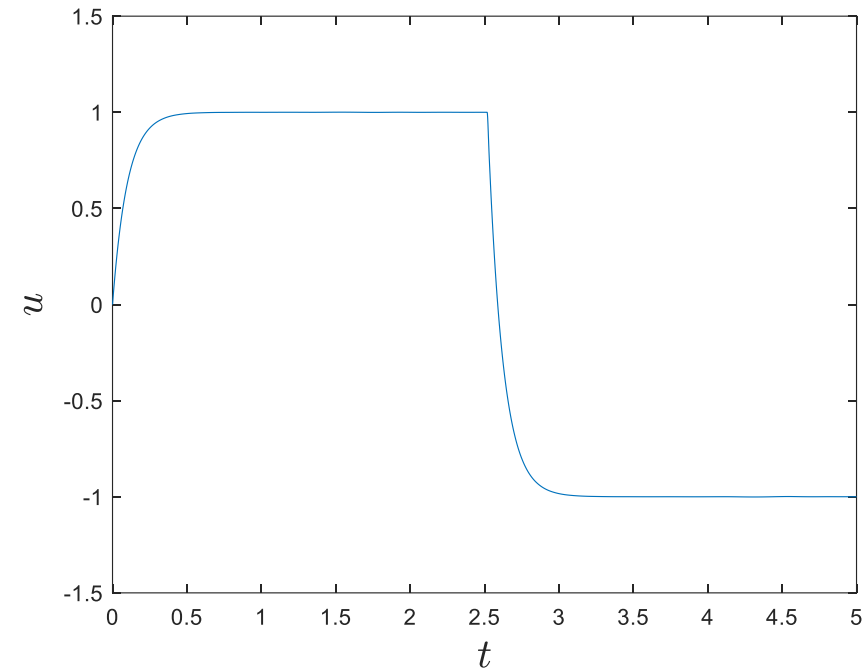
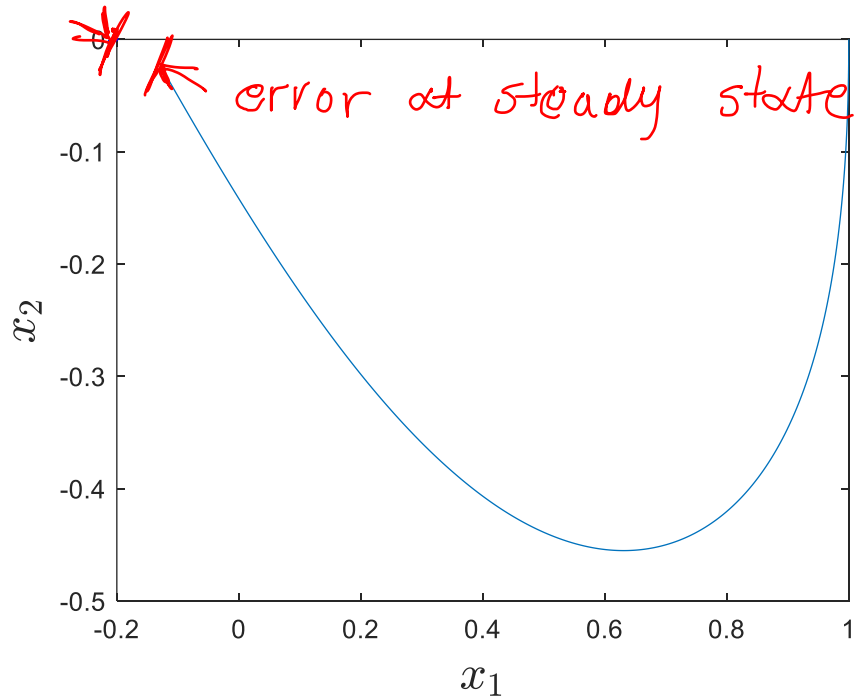
Sliding mode controller:
$$u = \text{sign}(x_2 + 2x_1)$$

Even higher uncertainty (no equivalent controller added)

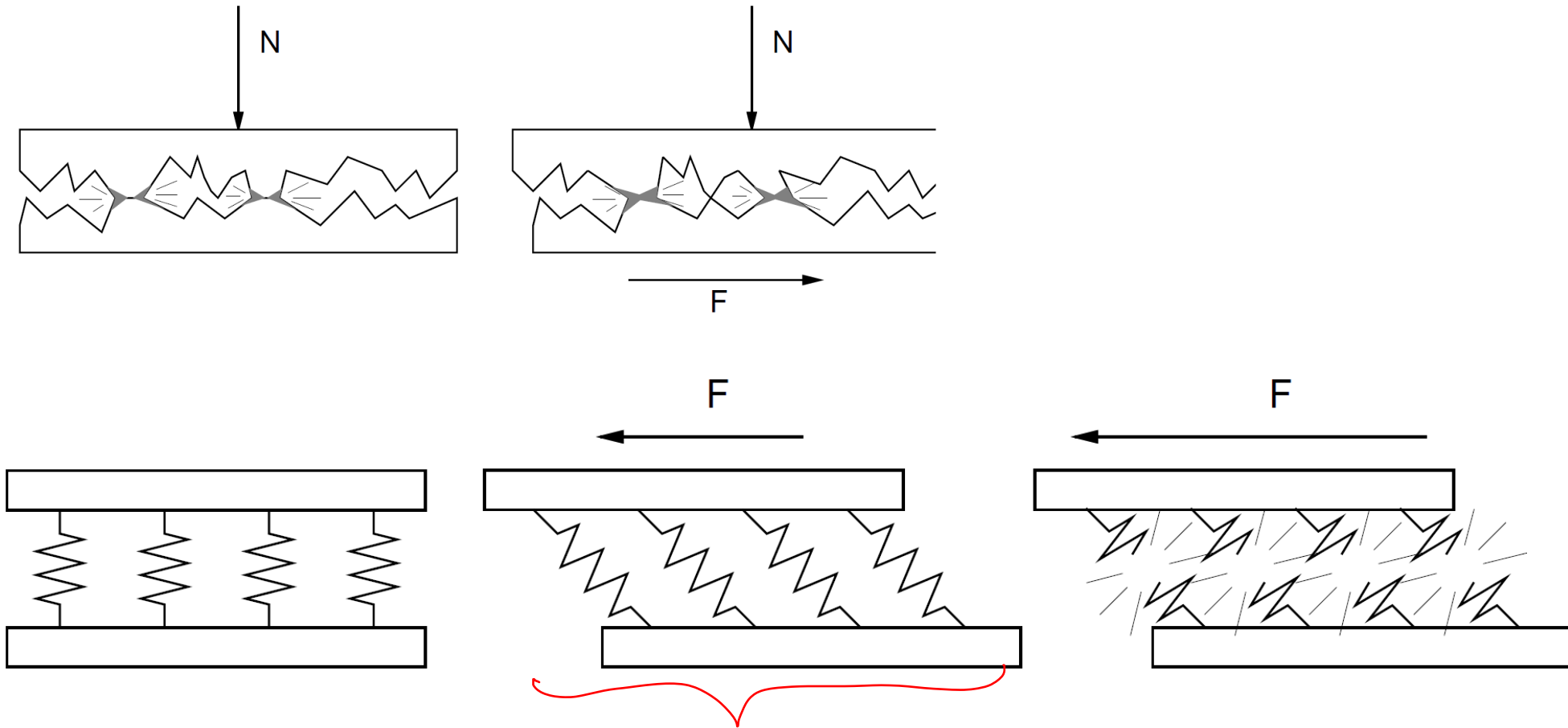
Example

- Continuous control

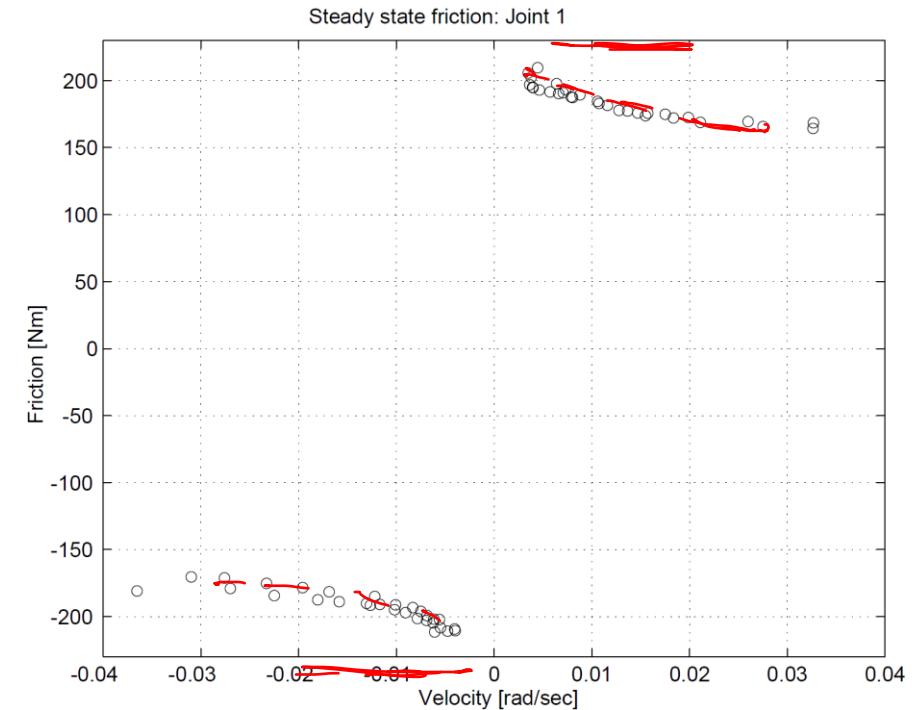
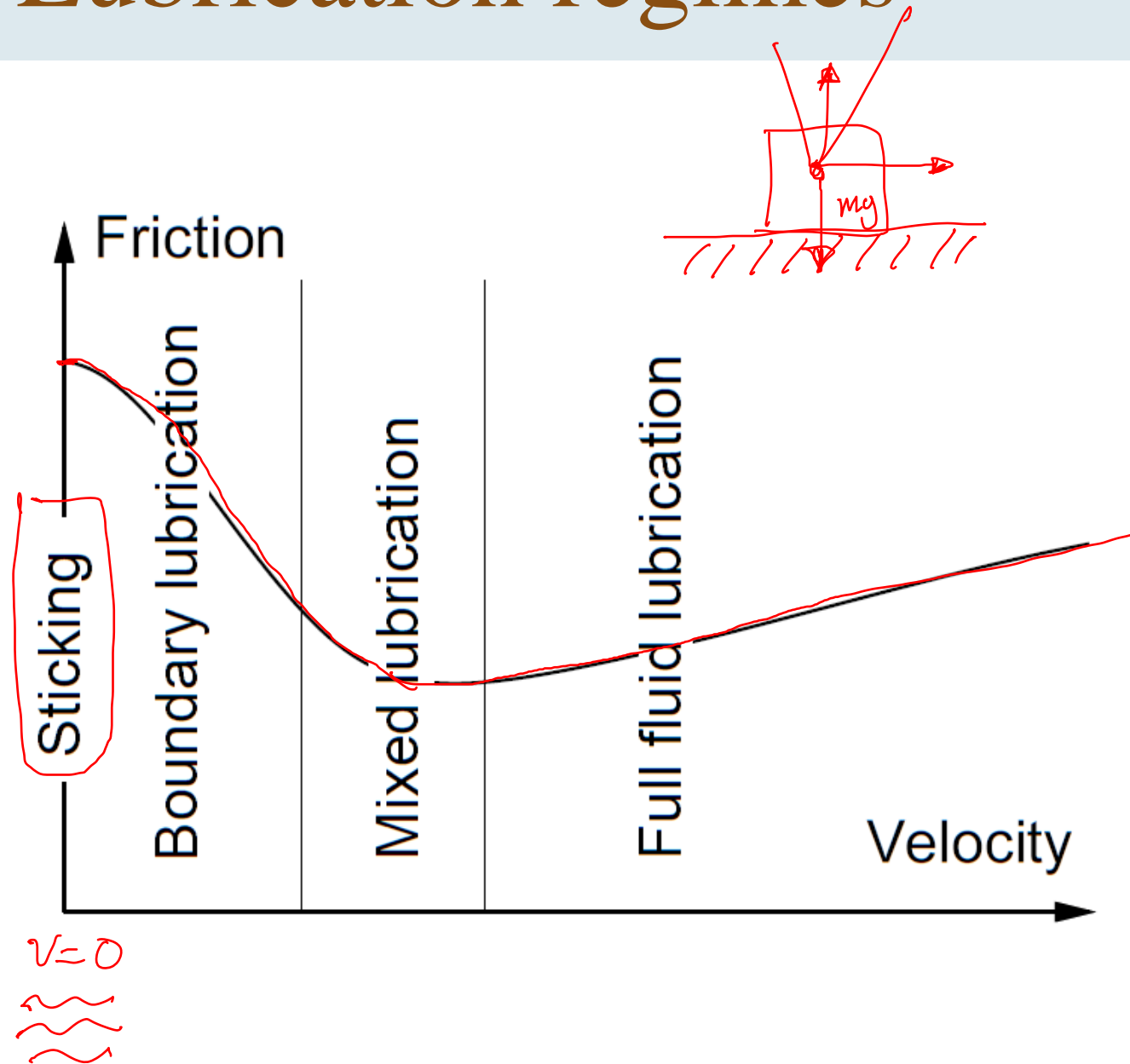
$$u = \tanh(x_2 + 2x_1)$$



Friction



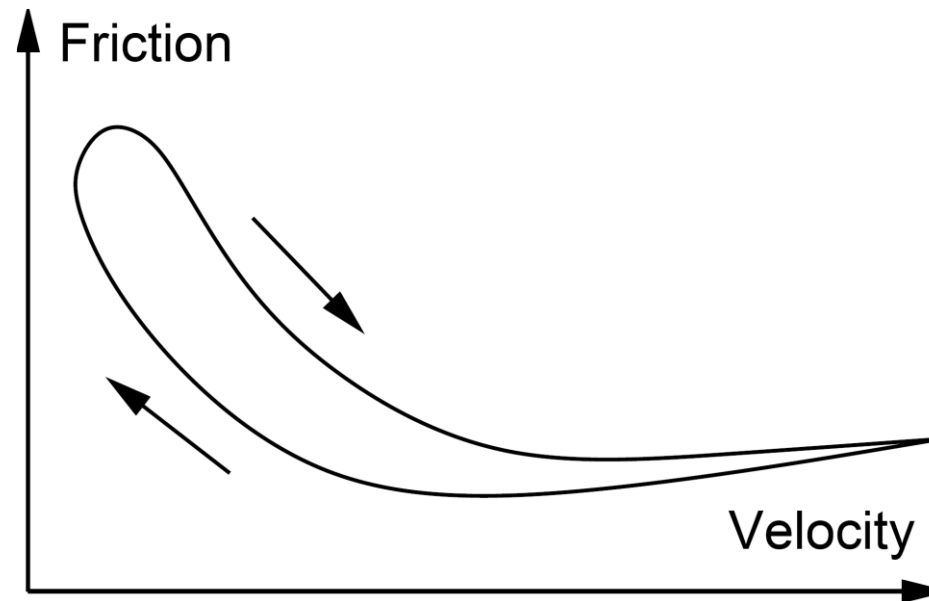
Lubrication regimes



For low velocity: friction increases with decreasing velocity
Stribeck (1902)

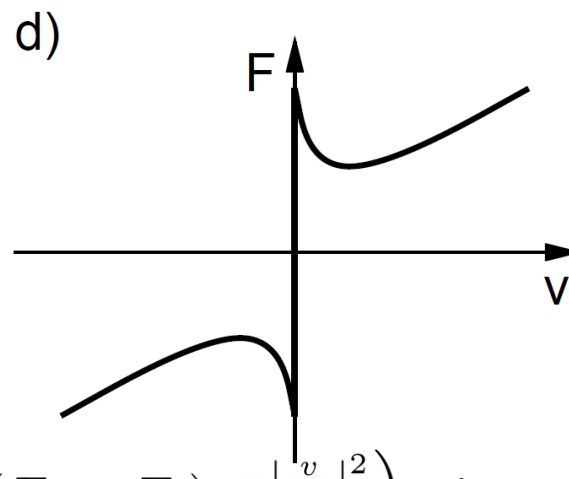
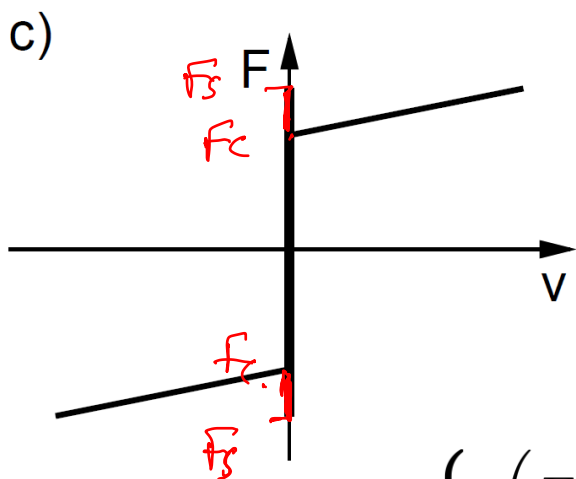
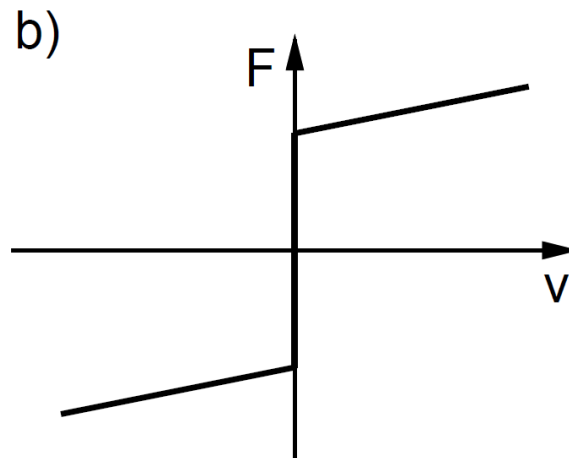
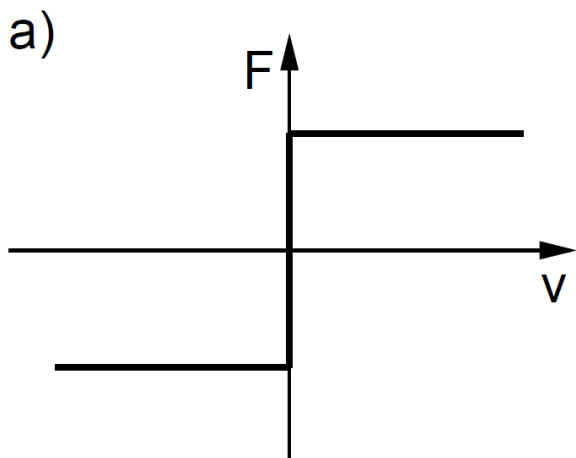
Hysteresis and friction

- Dynamics are important also outside sticking regime
- Hess and Soom (1990)
- Experiment with unidirectional motion $v(t) = v_0 + a \sin(\omega t)$
- Hysteresis effect!



Classical friction models

$$m\ddot{x} + \underline{F_c(t)} = F_e(t)$$



$F_e(t)$ = external applied force , F_c, F_v, F_s constants

$$F(t) = F_c \operatorname{sign} v(t) \quad (\alpha)$$

$$F(t) = \underbrace{F_v v(t)}_{\text{viscous}} + \underbrace{F_c \operatorname{sign} v(t)}_{\text{Coulomb}} \quad (b)$$

$$F(t) = \begin{cases} F_c \operatorname{sign} v(t) + F_v v(t) & v(t) \neq 0 \\ \max(\min(F_e(t), \underline{F_s}), -\underline{F_s}) & v(t) = 0 \end{cases} \quad (c)$$

equivalent expression

$$F(t) = \begin{cases} \left(F_c + (F_s - F_c) e^{-|\frac{v}{v_s}|^2} \right) \operatorname{sign} v(t) + F_v v(t) & v(t) \neq 0 \\ F_e(t) & v(t) = 0 \text{ and } |F_e(t)| < F_s \\ \operatorname{sign}(F_e) F_s & v(t) = 0 \text{ and } |F_e(t)| \geq F_s \end{cases} \quad (d)$$



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Advanced Friction Models

- Karnopp model
- Armstrong's seven parameter model
- Dahl model
- Bristle model
- Reset integrator model
- Bliman and Sorine
- LuGre model (Lund-Grenoble)

See PhD-thesis by Henrik Olsson

<https://lucris.lub.lu.se/ws/portalfiles/portal/4768278/8840259.pdf>

Friction models with extended state

Dahl's model

$$\dot{F}(t) = \sigma_0(v(t) - F_c|v(t)|)$$

LuGre model

$$\dot{z}(t) = v - \frac{|v|}{g(v)}z(t)$$

$$g(v) = \left(F_c + (F_s - F_c)e^{-|\frac{v}{v_s}|^2}\right)$$

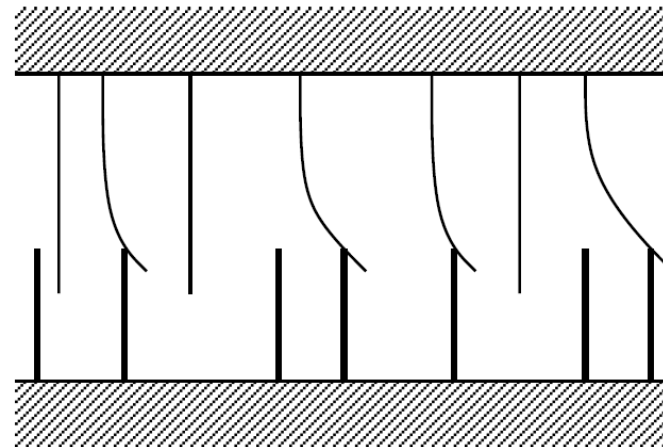
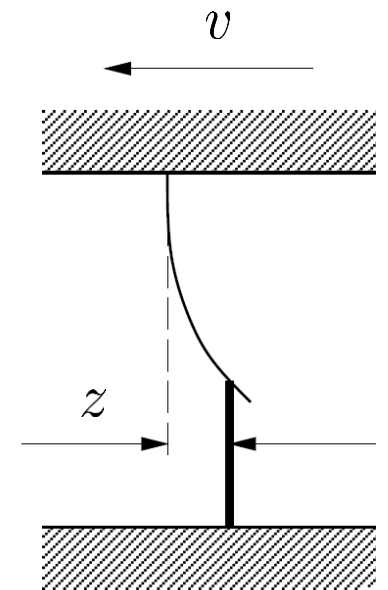
$$F(t) = \sigma_0 z(t) + \sigma_1 \dot{z} + F_v v(t)$$

↓
stiffness
of bristles

↓
damping

↓
viscous
friction

Contact between bristles

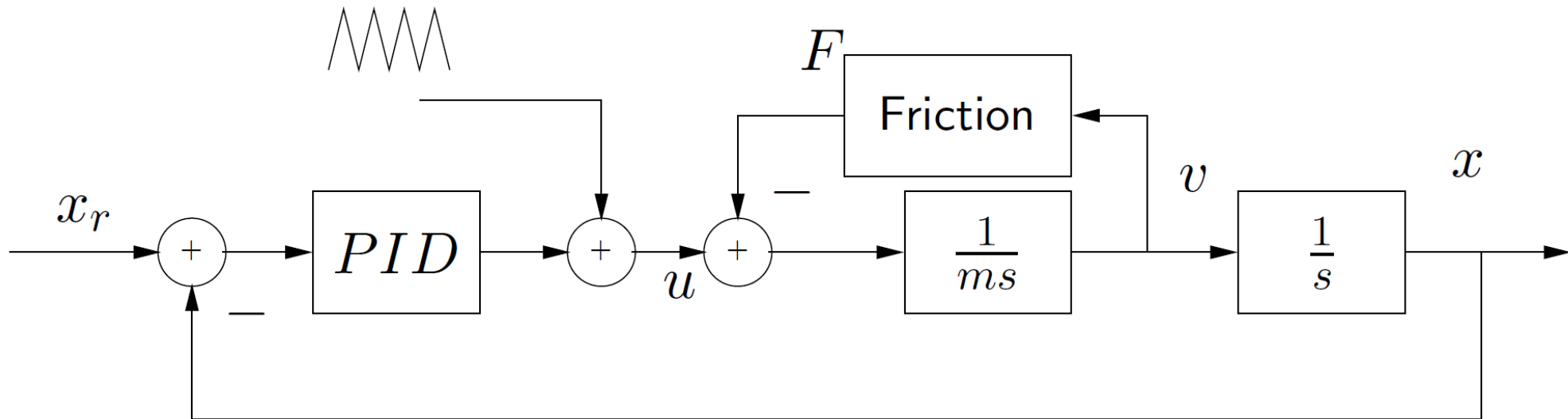


Friction and control

- Friction compensation
 - Lubrication
 - Dither
 - Integral action
 - Non-model based control
 - **Model-based friction compensation**
 - Adaptive friction compensation

High-frequency mechanical vibration: used to avoid sticking

Integral term compensate for slowly varying disturbances



Friction and control

- Friction compensation
 - Lubrication
 - Dither
 - Integral action
 - Non-model based control
 - **Model-based friction compensation**
 - Adaptive friction compensation
- To be useful for control the model should be:
 - sufficiently accurate,
 - suitable for simulation,
 - simple, few parameters to determine.
 - physical interpretations, insight
- Simple models should be preferred.
- If no stiction occurs the $v=0$ -models are not needed.

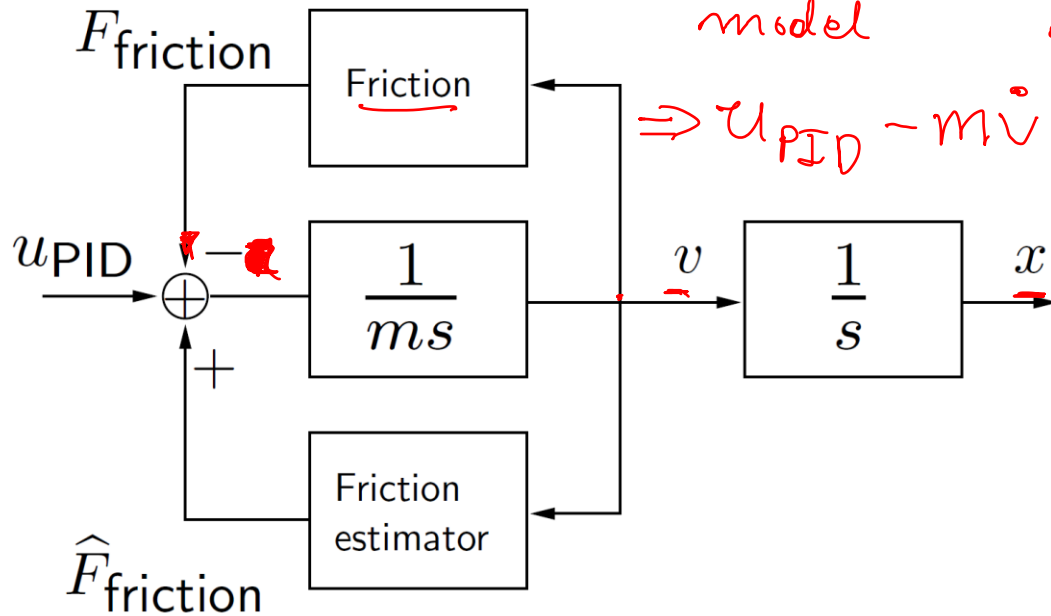
Adaptive friction compensation

$$F_{\text{friction}} = a \text{sgn}(v)$$

$$m\ddot{x} + F_{\text{friction}} = u_{\text{PID}} + F_{\text{friction}}$$

model

Result: $e = \underline{a} - \underline{\hat{a}} \rightarrow 0$ as $t \rightarrow \infty$:



$$\Rightarrow u_{\text{PID}} - m\dot{v} = F - \hat{F}$$

$$\frac{d}{dt} \sqrt{v^2} = \text{sgn}(v) \dot{v}$$

$$\frac{de}{dt} = \frac{-d\hat{a}}{dt} = \frac{dz}{dt} - km \frac{d}{dt} |v|$$

$$= k u_{\text{PID}} \text{sgn}(v) + km \dot{v} \text{sgn}(v)$$

$$= k \text{sgn}(v) (u_{\text{PID}} - m\dot{v})$$

$$= -k \text{sgn}(v) (F - \hat{F}) = -k \text{sgn}(v) (a \cdot \text{sgn}(v) - \hat{a} \cdot \text{sgn}(v))$$

$$= -k(a - \hat{a})$$

$$= -ke$$

\Rightarrow

$$e(t) = e(0) \exp(-kt)$$

$$e(t) \rightarrow 0$$

Assumption: v measurable.

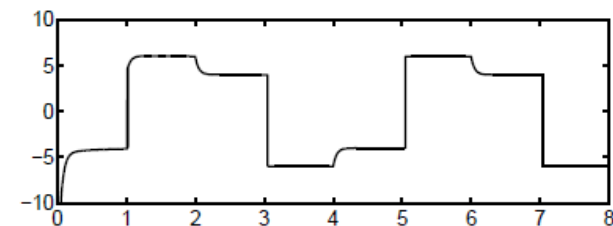
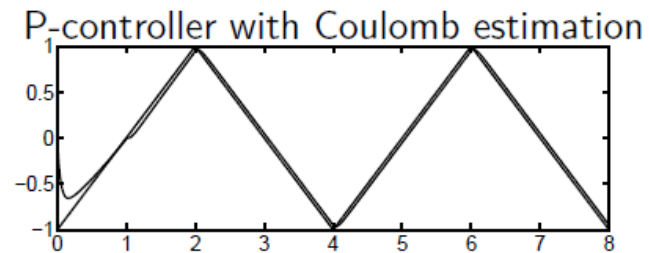
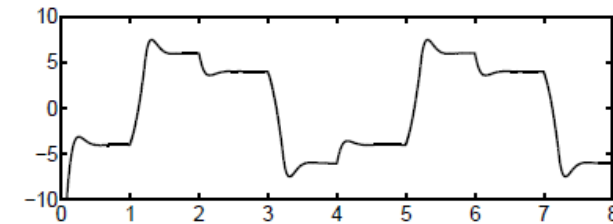
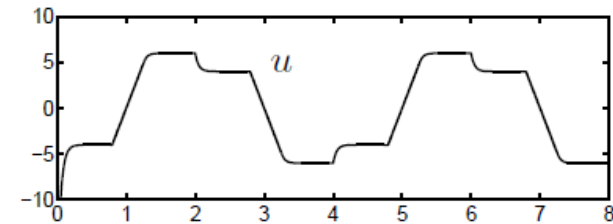
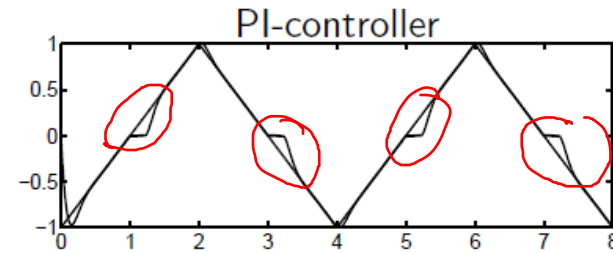
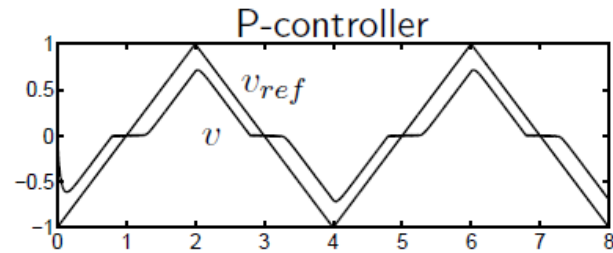
Friction estimator:

$$\begin{cases} \dot{z} = k u_{\text{PID}} \text{sgn}(v) \\ \hat{a} = z - km |v| \\ \hat{F}_{\text{friction}} = \hat{a} \text{sgn}(v) \end{cases}$$

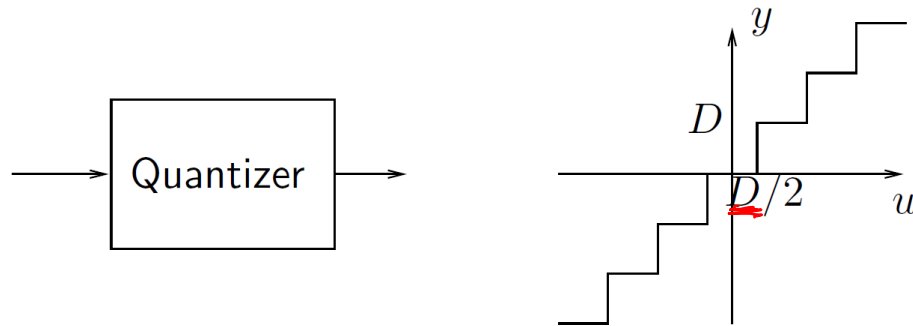


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Velocity Control – Results



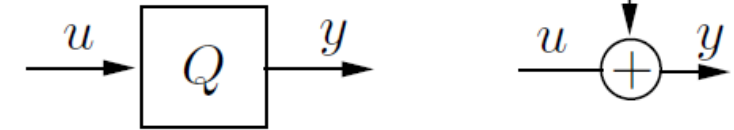
Quantization



- Digital signals have specific number of bits (accuracy and range of signals) (e.g. 8 bits, 64 bits)
- Quantization in A/D and D/A converters
- Quantization of parameters
- Roundoff, overflow, underflow in operations



Linear Model of Quantization e



- e : Noise independent of u with rectangular distribution over the quantization size with rectangular distribution with variance

$$Var(e) = \int_{-\infty}^{+\infty} e^2 f_e de = \int_{-D/2}^{D/2} e^2 \frac{1}{D} de = \frac{D^2}{12}$$

- It works if quantization level is small compared to the variations in u



For prediction of limit cycles

- Use describing function analysis

Sliding mode control

- Choose K such that $u = v$ drives $\sigma(x)$ to zero in finite time.

$$u = -K \operatorname{sign}(\sigma)$$

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

Sliding mode control

- Choose K such that $u = v$ drives $\sigma(x)$ to zero in finite time.

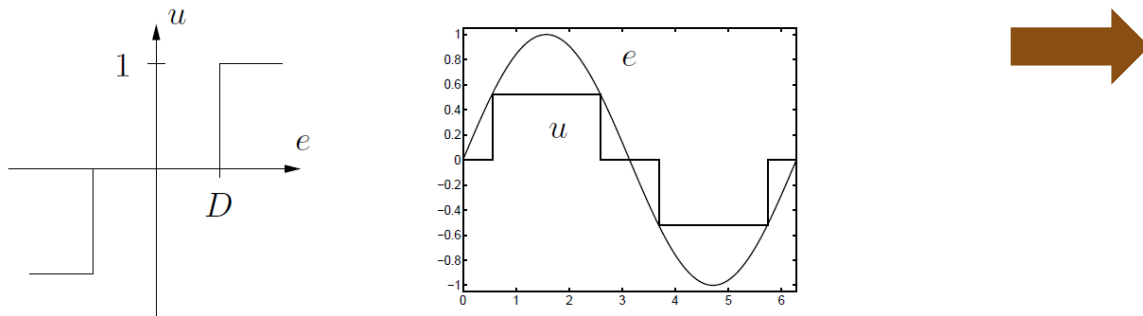
$$u = -K \operatorname{sign}(\sigma)$$

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

Quantization: Describing function



- Recall the deadzone nonlinearity



$$\text{Var}(e) = \int_{-\infty}^{+\infty} e^2 f_e de = \int_{-D/2}^{D/2} e^2 \frac{1}{D} de = \frac{D^2}{12}$$