

Sliding Mode

Motivating example:

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

$$\dot{x}_2 = h(x) + g(x)u$$

where $h(x)$ and $g(x)$ are unknown but

$$g(x) \geq g_0 > 0$$

Step 1: Choose a sliding manifold:

$$\text{Let } s = a_1 x_1 + x_2 = 0$$

Note that when $s = 0$, $x_2 = \dot{x}_1 = a_1 x_1$,

which implies that $x_1 \rightarrow 0$, if \dot{x}_1 also goes to zero, then $x_2 \rightarrow 0$.

Step 2 ~~is~~ pick u to drive all trajectories to $s = 0$

$$\begin{aligned} \text{Note that } \dot{s} &= a_1 \dot{x}_1 + \dot{x}_2 \\ &= a_1 x_2 + h + g u \end{aligned}$$

$$\text{Let } V = \frac{1}{2} s^2$$

then

$$\begin{aligned} \dot{V} &= s \dot{s} = s(a_1 x_2 + h + g u) \\ &= s g \left[\frac{a_1 x_2 + h}{g} + u \right] \end{aligned}$$

$$\text{Suppose } \left| \frac{a_1 x_2 + h(x)}{g(x)} \right| \leq \rho(x) \quad \forall x \in \mathbb{R}^2$$

Then

$$\dot{v} \leq g(x) |s| \left| \frac{a_1 x_2 + h}{g} \right| + g s u$$

$$\leq g(x) |s| p(x) + g(x) s u$$

Let $u = -\beta(x) \operatorname{sgn}(s)$

where $\beta(x) \geq p(x) + \beta_0$ then

$$\dot{v} \leq g(x) |s| p(x) - g(x) |s| p(x) - g(x) |s| \beta_0$$

$$= -g(x) \beta_0 |s|$$

$$\leq -g_0 \beta_0 |s|$$

\therefore Trajectories starting outside of s , converge to s .

In fact they converge to s in finite time as shown by the following:

Let $w = \sqrt{2v}$, then

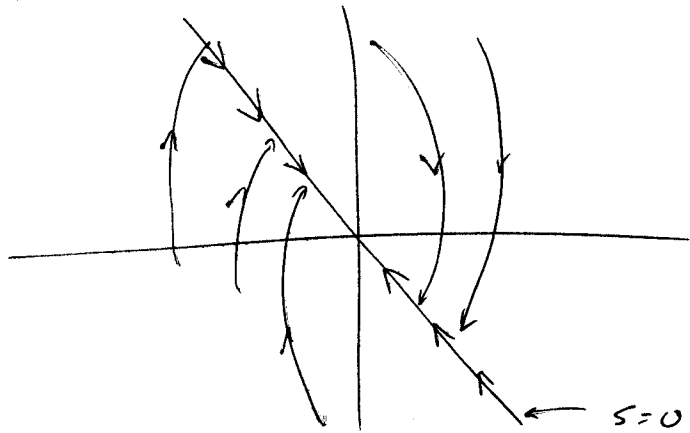
$$\dot{w} = \frac{\dot{v}}{\sqrt{2v}} \leq \frac{-g_0 \beta_0 |s|}{\sqrt{2v}} = \frac{-g_0 \beta_0 |s|}{|s|} = -g_0 \beta_0$$

$$\Rightarrow w(s(t)) \leq w(s(0)) - g_0 \beta_0 t$$

until $s(t) = 0$ in finite time, namely

$$t^* \leq \frac{w(s(0))}{g_0 \beta_0}$$

Phase portrait



The set $\{x: s=0\}$ is called the "sliding manifold"

The motion before reaching the sliding manifold is called the "reaching phase."

The motion on the sliding manifold is called the "sliding phase."

The controller $u = -\beta(x) \operatorname{sgn}(s)$ is the sliding mode control.

Advantages: does not require exact knowledge of $f(x)$ and $g(x)$, is robustness

Region of Attraction

If we know that $\left| \frac{a_1 x_2 + h(x)}{g(x)} \right| \leq k_1$

Then we can let

$$u = -k \operatorname{sgn}(s), \quad k > k_1$$

Consider the set $\tilde{\Omega} = \{s : |s| \leq c\}$

The condition $\dot{s} \leq 0$ implies $\tilde{\Omega}$ is positively invariant

$$\text{Since } s = a_1 x_1 + x_2 \Rightarrow x_2 = -a_1 x_1 + s$$

$$\therefore \dot{x}_1 = x_2 = -a_1 x_1 + s$$

Letting $V_1 = \frac{1}{2} x_1^2$ gives

$$\dot{V}_1 = x_1 \dot{x}_1 = -a_1 x_1^2 + x_1 s$$

$$\leq -a_1 x_1^2 + |x_1| |s|$$

$$\leq -a_1 x_1^2 + |x_1| c$$

$$\leq 0 \quad \text{if } |x_1| \geq \frac{c}{a_1}$$

$$\therefore |x_1(t_0)| \leq \frac{c}{a_1} \Rightarrow |x_1(t)| \leq \frac{c}{a_1}$$

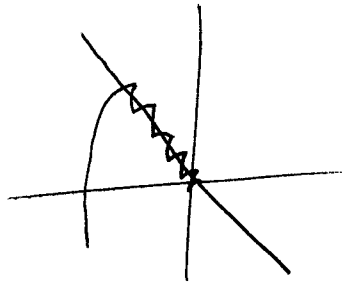
$\therefore \Omega = \{ |x_1| \leq \frac{c}{a_1}, |s| \leq c \}$ is positively invariant

$$\text{if } \left| \frac{a_1 x_2 + h(x)}{g(x)} \right| \leq k_1 \quad \forall x \in \Omega$$

$$\text{Since } x_1(t) \rightarrow 0 \quad \text{and } s(t) \rightarrow 0 \Rightarrow x_2(t) \rightarrow 0$$

we have that every trajectory starting in Ω goes to the origin.

Disadvantage: chattering - since can't switch instantaneously, the state will chatter around the sliding surface.



Example: inverted pendulum

$$\begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= \underbrace{-(g_0/l) \sin(x_1 + \delta_1)}_{h(x)} - \underbrace{(h_0/m)}_{g(x)} x_2 + \left(\frac{1}{ml^2}\right) u \end{aligned}$$

Note

$$\begin{aligned} \left| \frac{a_1 x_2 + h(x)}{g(x)} \right| &= \left| ml^2 a_1 x_2 - mg_0 l \sin(x_1 + \delta) - h_0 l^2 x_2 \right| \\ &= \left| l^2 (m a_1 - h_0) x_2 - mg_0 l \sin(x_1 + \delta) \right| \\ &\leq l^2 (m a_1 - h_0) |x_2| + mg_0 l \end{aligned}$$

pick $a_1 = 1$ and let $|x_1| \leq \pi \equiv c$

then on the set $\{ |x_1| \leq \pi, |s| = |x_1 + x_2| < \frac{\pi}{2} \}$

we have

$$\begin{aligned} \left| \frac{a_1 x_2 + h(x)}{g(x)} \right| &\leq l^2 |m a_1 - h_0| (2\pi) + mg_0 l \\ &\leq \bar{l}^2 |\bar{m} - \bar{h}_0| (2\pi) + \bar{m} \bar{g}_0 \bar{l} \end{aligned}$$

where $\underline{l} \leq l \leq \bar{l}$, $\underline{m} \leq m \leq \bar{m}$, $\underline{h}_0 - h \leq \bar{h}_0$

Example

$$\dot{y} = h(y, \dot{y}) + g(y)u$$

where we know $\hat{h}(y, \dot{y})$ and $\hat{g}(y)$ st.

$$|\hat{h} - h| < \bar{h}$$

$$|\hat{g} - g| < \bar{g}$$

Let $s = \dot{y} + a_1 y$ where $a_1 > 0$

Note that if $s = 0$ then $\dot{y} = -a_1 y \Rightarrow y(t) = e^{-a_1 t} y(0)$

Let $x_1 = y$

$x_2 = \dot{y}$

$$\Rightarrow \begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= h(x) + g(x)u \end{aligned}$$

where $s = x_2 + a_1 x_1$

Let $v = \frac{1}{2} s^2$

$$\begin{aligned} \Rightarrow \dot{v} &= s \dot{s} = s(\dot{x}_2 + a_1 \dot{x}_1) \\ &= s(h(x) + g(x)u + a_1 x_2) \\ &= s(\hat{h} + (h - \hat{h}) + \hat{g}u + (g - \hat{g})u + a_1 x_2) \end{aligned}$$

Let $u = -\frac{[a_1 x_2 + \hat{h}]}{\hat{g}} + v$

$$\begin{aligned} \Rightarrow \dot{v} &= s \left[(h - \hat{h}) + \hat{g}v + \frac{(g - \hat{g})}{\hat{g}} (a_1 x_2 + \hat{h}) + (g - \hat{g})v \right] \\ &= s \left[(h - \hat{h}) + \frac{(g - \hat{g})}{\hat{g}} (a_1 x_2 + \hat{h}) + gv \right] \end{aligned}$$

Example

$$\dot{y} = h(y, \dot{y}) + u$$

where we know $\hat{h}(y, \dot{y})$ and

$$\forall y, \dot{y} \quad |h - \hat{h}| \leq \bar{h}$$

Let $s = \dot{y} + a_1 y$ where $a_1 > 0$.

Note that if $s > 0$ then $\dot{y} = -a_1 y \Rightarrow y(t) = e^{-a_1 t} y(0)$

Let $x_1 = y$
 $x_2 = \dot{y}$

$$\Rightarrow \begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= h + u = \hat{h} + (h - \hat{h}) + u \end{aligned}$$

Let $v = \frac{1}{2} s^2 \Rightarrow \dot{v} = s \dot{s} = s(\hat{h} + (h - \hat{h}) + u + a_1 x_2)$

let $u = -\hat{h} - a_1 x_2 + v$

then $\dot{v} = s(\hat{h} - \hat{h} + v)$

$$\begin{aligned} & \leq |s| |\hat{h} - \hat{h}| + s v \\ & \leq |s| \bar{h} + s v \end{aligned}$$

Let $u = -k \operatorname{sgn}(s)$

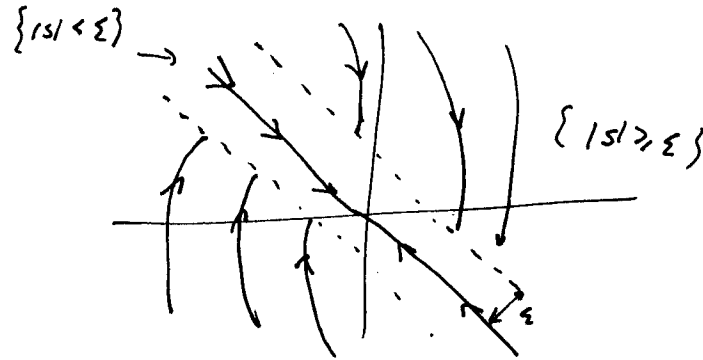
$$\begin{aligned} \Rightarrow \dot{v} &= |s| \bar{h} - k |s| \\ &= -(k - \bar{h}) |s| \end{aligned}$$

\therefore if $k > \bar{h}$ then $\dot{v} < 0$
& converge to sliding mode in finite time.

What about chattering?

Basic idea:

Instead of switching exactly on s ,
place a buffer around s



Instead of using

$$u = -\beta(x) \operatorname{sgn}(s)$$

use

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

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

In the set ~~$\{ |s| \geq \epsilon \}$~~ $\{ |s| \geq \epsilon \}$ we again have

$$\dot{V} \leq -\gamma_0 \beta_0 |s|$$

with finite time convergence to the set

$$\{ |s| < \epsilon \}$$

In side the set $\{ |s| < \epsilon \}$ we have that

$$\dot{x}_1 = \dot{x}_2 = -a x_1 + s$$

\therefore The derivative of $V_1 = \frac{1}{2} x_1^2$

$$\begin{aligned}
 \dot{V}_1 &= x_1 \dot{x}_1 \\
 &= -a_1 x_1^2 + x_1 s \\
 &\leq -a_1 x_1^2 + |x_1| |s| \\
 &\leq -a_1 x_1^2 + |x_1| \varepsilon \\
 &= -(1-\theta) a_1 x_1^2 + (-\theta a_1 x_1^2 + |x_1| \varepsilon) \quad 0 < \theta < 1 \\
 &\leq -(1-\theta) a_1 x_1^2 \quad \text{if} \quad |x_1| \geq \frac{\varepsilon}{a_1 \theta}
 \end{aligned}$$

$$\therefore x \rightarrow \Omega_\varepsilon = \left\{ |x_1| \leq \frac{\varepsilon}{a_1 \theta}, |s| \leq \varepsilon \right\}$$

In general, the origin may not be stabilized, but ε can be made arbitrarily small.

The origin may be stabilized but this must be investigated independently.



Generalization

Given a nonlinear system $\dot{x} = f(x, u)$, suppose that we can find a change of variables to put the system in the form

$$\dot{y} = f_a(y, \xi)$$

$$\dot{\xi} = f_b(y, \xi) + G(x) E(x) u + \delta(t, x, u)$$

where f_a, f_b, E - known, E - invertable

G, δ - unknown

G - diagonal + bounded away from zero

Also, assume that if we think of ξ as an input to the first equation, that we can find a stabilizing control $\phi(y)$ s.t.

$\dot{y} = f_a(y, \phi(y))$ is asymptotically stable.

Since we desire $\xi = \phi(y)$, let

$$s \equiv \xi - \phi(y)$$

Then
$$\dot{s} = \dot{\xi} - \frac{\partial \phi}{\partial y} \dot{y}$$

$$= f_b + GEu + \delta - \frac{\partial \phi}{\partial y} f_a$$

$$\text{Let } u = E^{-1} \left\{ -\hat{G}^{-1} \left[f_0 - \frac{\partial \phi}{\partial y} f_a \right] + v \right\}$$

where \hat{G} is the nominal model of G

$$\text{Since } G(s) = \begin{pmatrix} g_1(s) & & 0 \\ & \ddots & \\ 0 & & g_p(s) \end{pmatrix} \text{ is diagonal we have}$$

$$\dot{s}_i = g_i(x) v_i + \Delta_i(t, x, u)$$

where Δ_i is all of the left over junk.

Assume that we know bounds on Δ_i , i.e.

$$\left| \frac{\Delta_i(t, x, u)}{g_i(x)} \right| \leq \rho(x) + \alpha_0 \|v\|_\infty \quad \bullet \alpha_0 \in (0, 1)$$

$$\text{Let } v_i = \frac{1}{s_i} \dot{s}_i \quad \text{then}$$

$$\begin{aligned} \dot{v}_i &= s_i \dot{s}_i \\ &= s_i (g_i v + \Delta_i) \\ &\leq g_i \left\{ s_i v_i + |s_i| \left| \frac{\Delta_i}{g_i} \right| \right\} \\ &\leq g_i \left\{ s_i v_i + |s_i| [\rho + \alpha_0 \|v\|_\infty] \right\} \end{aligned}$$

$$\text{Take } v_i = -\beta(x) \text{sgn}(s_i) \quad \text{then}$$

$$\begin{aligned} \dot{v}_i &\leq g_i \left\{ -\beta(x) |s_i| + |s_i| [\rho + \alpha_0 \|v\|_\infty] \right\} \\ &\leq g_i \left\{ -\beta(x) + \rho + \alpha_0 \beta \right\} |s_i| \\ &= g_i \left[-(1 - \alpha_0) \beta + \rho \right] |s_i| \end{aligned}$$

Example 1

Given

1st order linear system

$$\dot{y} = ay + bu$$

a, b are not known precisely.

Assume we have a nominal model

\hat{a} $\hat{b} > 0$ and that we know that

~~$$|a - \hat{a}| \leq \bar{a}$$~~

$$a = \hat{a} + \delta a \quad \text{where } |\delta a| \leq \bar{a}$$

$$b = \hat{b} + \delta b \quad \text{where } 0 < \underline{b} \leq \delta b \leq \bar{b}$$

Similar to the adaptive control problem, assume that our objective is to follow a reference model

$$\dot{y}_m = a_m y_m + b_m r$$

Let $\tilde{y} = y - y_m$

Find a sliding mode controller that drives $\tilde{y} \rightarrow 0$.

The ~~slid~~ sliding surface is one dimension less than the ^{order} ~~degree~~ of the system. Since the original system has order 1, the sliding surface will have dimension = 0.

$$\text{Let } s = \tilde{y} = y - y_m.$$

Also let $v = \frac{1}{2} \dot{s}^2$, then

$$\dot{v} = s \dot{s} \quad \text{to where}$$

$$\dot{s} = \dot{\tilde{y}} = \dot{y} - \dot{y}_m = ay + bu - a_m y_m - b_m r$$

so

$$\dot{v} = s [ay + bu - a_m y_m - b_m r]$$

$$= s \left[\hat{a} y + s a y + bu - a_m y_m - b_m r \right]$$

$$= b s \left[\frac{\hat{a} y + \frac{sa}{b} - a_m y_m - b_m r}{b} + u + \frac{s a}{b} y \right]$$

$$\text{Let } u = - \left[\frac{\hat{a} y - a_m y_m - b_m r}{\hat{b}} \right] + v$$

Then

$$\begin{aligned} \dot{v} &= bs \left\{ (\hat{a}y - a_m y_m - b_m r) \left(\frac{1}{b} - \frac{1}{\bar{b}} \right) + \frac{\delta a}{b} y + v \right\} \\ &= s \left\{ (\hat{a}y - a_m y_m - b_m r) \left(\frac{\bar{b} - b}{b} \right) + \delta a y \right\} + bs v \\ &\leq |s| \left\{ |\hat{a}y - a_m y_m - b_m r| \frac{\bar{b}}{b} + \bar{a}|y| \right\} + bs v \end{aligned}$$

Let $v = -\beta \operatorname{sgn}(s)$

$$\begin{aligned} \Rightarrow \dot{v} &\leq |s| \left\{ |\hat{a}y - a_m y_m - b_m r| \frac{\bar{b}}{b} + \bar{a}|y| \right\} - b\beta |s| \\ &= -b|s| \left\{ \beta - \left[\frac{|\hat{a}y - a_m y_m - b_m r| \frac{\bar{b}}{b} + \bar{a}|y|}{b} \right] \right\} \\ &\leq -b|s| \left\{ \beta - \left[\frac{|\hat{a}y - a_m y_m - b_m r| \frac{\bar{b}}{b} + \bar{a}|y|}{\underline{b}} \right] \right\} \end{aligned}$$

Let $\beta = \left[\frac{|\hat{a}y - a_m y_m - b_m r| \frac{\bar{b}}{b} + \bar{a}|y|}{\underline{b}} \right] + \beta_0$

Then

$$\begin{aligned} \dot{v} &\leq -b\beta_0 |s| \\ &\leq -\underline{b} \beta_0 |s| \end{aligned}$$

\therefore we can conclude finite-time convergence to the set $\{s=0\}$

Summarizing, the control law is

$$u = - \underbrace{\left[\frac{\hat{a}y - a_m y_m - b_m r}{\hat{b}} \right]}_{\text{control based on nominal model}} - \underbrace{\left\{ \left[\frac{|\hat{a}y - a_m y_m - b_m r| \frac{\bar{b}}{\underline{b}} + \bar{a}|y|}{\underline{b}} \right] + \beta_0 \right\} \text{sgn}(y)}_{\text{makes control robust with respect to bounded uncertainty.}}$$

control based
on nominal
model

makes control robust
with respect to bounded
uncertainty.

Matlab example

$$a = -5, \quad b = 2$$

$$\hat{a} = -3, \quad \hat{b} = 1$$

$$a_m = -3, \quad b_m = -3$$

$$\bar{b} = 3, \quad \underline{b} = 0.5, \quad \beta_0 = 3$$

Sliding Mode Example 2

Simplified model of underwater vehicle heading dynamics.

$$\ddot{\psi} + a \dot{\psi} |\dot{\psi}| = \tau$$

$\psi \equiv$ heading direction

$\tau \equiv$ normalized torque

$a =$ positive unknown parameter

Let \hat{a} be nominal value and

$$|S_a| = |a - \hat{a}| < \bar{a}$$

The control objective is to pick τ to follow a reference heading ψ_r

Let the sliding surface be

$$S = \ddot{\tilde{\psi}} + \gamma \dot{\tilde{\psi}}$$

where $\tilde{\psi} = \psi - \psi_r$ is the heading error,

and let

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

Then $\dot{V} = S\dot{S}$ where

$$\dot{S} = \ddot{\tilde{\psi}} + \gamma \dot{\tilde{\psi}} = \ddot{\psi} - \ddot{\psi}_r + \gamma \dot{\tilde{\psi}} = -a \dot{\psi} |\dot{\psi}| + \tau - \ddot{\psi}_r + \gamma \dot{\tilde{\psi}}$$

Then

$$\begin{aligned}\dot{v} &= s \left(-a \dot{\psi} |\dot{\psi}| + \tau - \ddot{\psi}_r + \gamma \ddot{\psi} \right) \\ &= s \left(-\hat{a} \dot{\psi} |\dot{\psi}| - \delta a \dot{\psi} |\dot{\psi}| + \tau - \ddot{\psi}_r + \gamma \ddot{\psi} \right)\end{aligned}$$

Select

$$\tau = \hat{a} \dot{\psi} |\dot{\psi}| + \ddot{\psi}_r - \gamma \ddot{\psi} + v$$

Then

$$\begin{aligned}\dot{v} &= s \left(-\delta a \dot{\psi} |\dot{\psi}| + v \right) \\ &\leq |s| \left(|\delta a| |\dot{\psi}|^2 \right) + s v\end{aligned}$$

Select

$$v = -\beta \operatorname{sgn}(s)$$

Then

$$\dot{v} \leq -|s| \left[\beta - |\delta a| |\dot{\psi}|^2 \right]$$

~~Then~~ ~~use~~

$$\leq -|s| \left[\beta - \bar{a} |\dot{\psi}|^2 \right]$$

Select

$$\beta = \bar{a} |\dot{\psi}|^2 + \beta_0 \quad \text{then}$$

$$\dot{v} \leq -\beta_0 |s|$$

control is

$$\tau = \hat{a} \dot{\psi} |\dot{\psi}| + \ddot{\psi}_r - \gamma \ddot{\psi} - \left[\bar{a} |\dot{\psi}|^2 + \beta_0 \right] \operatorname{sgn}(\dot{\psi} + \gamma \ddot{\psi})$$

< Matlab simulation >

Let's try the following observer

$$\dot{\hat{x}} = A\hat{x} + Bu + L(y - C\hat{x}) + S(\hat{x}, y)$$

Define $\tilde{x} = x - \hat{x}$ and let

$$V = \tilde{x}^T P \tilde{x} \quad \text{where } P \text{ comes from}$$

Assumption A2

then

$$\dot{V} = \dot{\tilde{x}}^T P \tilde{x} + \tilde{x}^T P \dot{\tilde{x}}$$

where

$$\dot{\tilde{x}} = \dot{x} - \dot{\hat{x}}$$

$$= A\tilde{x} + f(t, x) + B\tilde{u} + B\tilde{v} - A\hat{x} - B\tilde{u} - L\tilde{C}\tilde{x} + L\hat{C}\hat{x} - S(\hat{x}, y)$$

$$= (A - LC)\tilde{x} + P^T C^T h(t, x) + P^T C^T w(t) - S(\hat{x}, y)$$

Select $S(\hat{x}, y) = P^T C^T R(\hat{x}, y)$ then

$$\dot{\tilde{x}} = (A - LC)\tilde{x} + P^T C^T [h(t, x) + w(t) - R(\hat{x}, y)]$$

$$\therefore \dot{V} = \tilde{x}^T (A - LC)^T P \tilde{x} + \tilde{x}^T P (A - LC) \tilde{x}$$

$$+ [h(t, x) + w - R(\hat{x}, y)]^T C P^T P \tilde{x}$$

$$+ \tilde{x}^T P P^T C^T [h(t, x) + w(t) - R(\hat{x}, y)]$$

$$= -\tilde{x}^T Q \tilde{x} + 2\tilde{x}^T C^T [h(t, x) + w(t) - R(\hat{x}, y)]$$

$$\Rightarrow \dot{v} \leq -\tilde{x}^T Q \tilde{x} + 2 \|c\tilde{x}\| \|h(x) + w(x)\| - 2 (c\tilde{x})^T R(\hat{x}, y) \\ \leq -\tilde{x}^T Q \tilde{x} + 2 \|c\tilde{x}\| \rho(y) - 2 (c\tilde{x})^T R(\hat{x}, y)$$

$$\text{Let } R(\hat{x}, y) = \beta \frac{c\tilde{x}}{\|c\tilde{x}\|} = \beta \frac{(y - c\hat{x})}{\|y - c\hat{x}\|}$$

$$\text{Then } \dot{v} \leq -\tilde{x}^T Q \tilde{x} - 2 \|c\tilde{x}\| [\beta - \rho(y)]$$

$$\text{Let } \beta = \rho(y) + \beta_0$$

$$\Rightarrow \dot{v} \leq -\tilde{x}^T Q \tilde{x} - 2\beta_0 \|c\tilde{x}\|$$

Therefore the sliding mode observer is

$$\dot{\hat{x}} = A\hat{x} + Bu + L(y - c\hat{x}) + \rho^{-1} c^T \left(\frac{y - c\hat{x}}{\|y - c\hat{x}\|} \right) [\rho(y) + \beta_0]$$

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Output Feedback Control of Underwater Vehicles Using Nonlinear State Observer

Myung-Hyun Kim and Daniel J. Inman

Fourth Semi-Annual Meeting-MURI
Nonlinear Active Control of Dynamical Systems
Oct.6 1998



Outline

- Summary of Previous Work
- Problem Formulation
- Output Feedback Control
- Dive Plane Equation of Motion
- Observer Design
- Controller Design
- Simulation Results
- Conclusion



Summary of previous work

■ Effects of actuator dynamics

- ◆ Actuators in UUVs such as thrusters, rudders have time lag and this tends to degrade system performance and make quick and precise control difficult.
- ◆ Thruster dynamics has significant influence on the dynamics of vehicle.
- ◆ Investigation of the influence of the rudder dynamics on the system response. However, since UUVs are operated in low speed, rudder dynamics does not influence the UUV dynamics.
- ◆ Rudder dynamics is substantial only at high speed such as in torpedo application.



Problem Formulation

■ Sliding Mode Controller

◆ Advantages

- ◆ Capable of handling nonlinear systems
- ◆ Robust to uncertainties/parameter variation

◆ Disadvantages

- ◆ All states measurement required
- ◆ Occurrence of undesirable chattering phenomena

◆ Chattering can be avoided by including boundary layer

◆ Design of nonlinear observer for state estimation

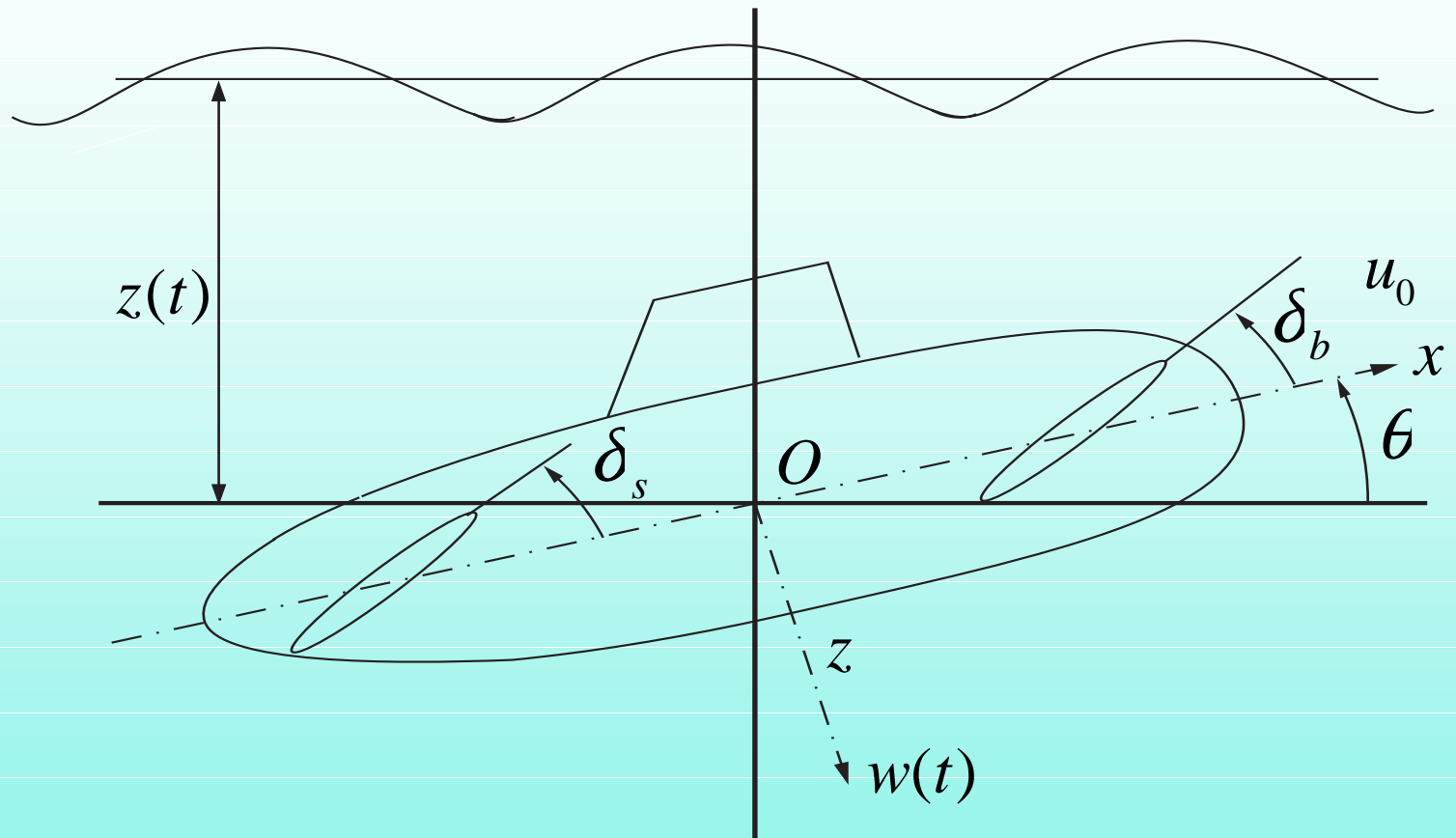


Output Feedback Control

- Healey(NPS) et.al.
 - ◆ Design of Robust Observer
 - ◆ Sliding Mode Control with estimated states
- Fossen(NIT) et.al.
 - ◆ Nonlinear Observer from the kinematic relationship with additional acceleration measurements
 - ◆ Adaptive Control with estimated states
- Nonlinear State observer using Sliding Mode



Dive Plane Equation of Motion



Dive Plane Equation of Motion(Cont'd)

$$\begin{aligned} & m[\dot{w} - uq - x_G \dot{q} - z_G q^2] \\ & = z_{\dot{q}} \dot{q} + z_{\dot{w}} \dot{w} + z_q uq + z_w w + u^2 (z_b \delta_b + z_s \delta_s) \end{aligned}$$

$$\begin{aligned} & I_y \dot{q} - m[x_G (\dot{w} - uq) - z_G (\dot{u} + wq)] \\ & = M_{\dot{q}} \dot{q} + M_{\dot{w}} \dot{w} + M_q uq + M_w uw + u^2 (M_b \delta_b + M_s \delta_s) \\ & - (x_G mg - x_B B) \cos \theta - (z_G mg - z_B B) \sin \theta \end{aligned}$$

$$\dot{\theta} = q$$

$$\dot{z} = w \cos \theta - u \sin \theta$$



Dive Plane Equation of Motion(Cont'd)

State Space Form

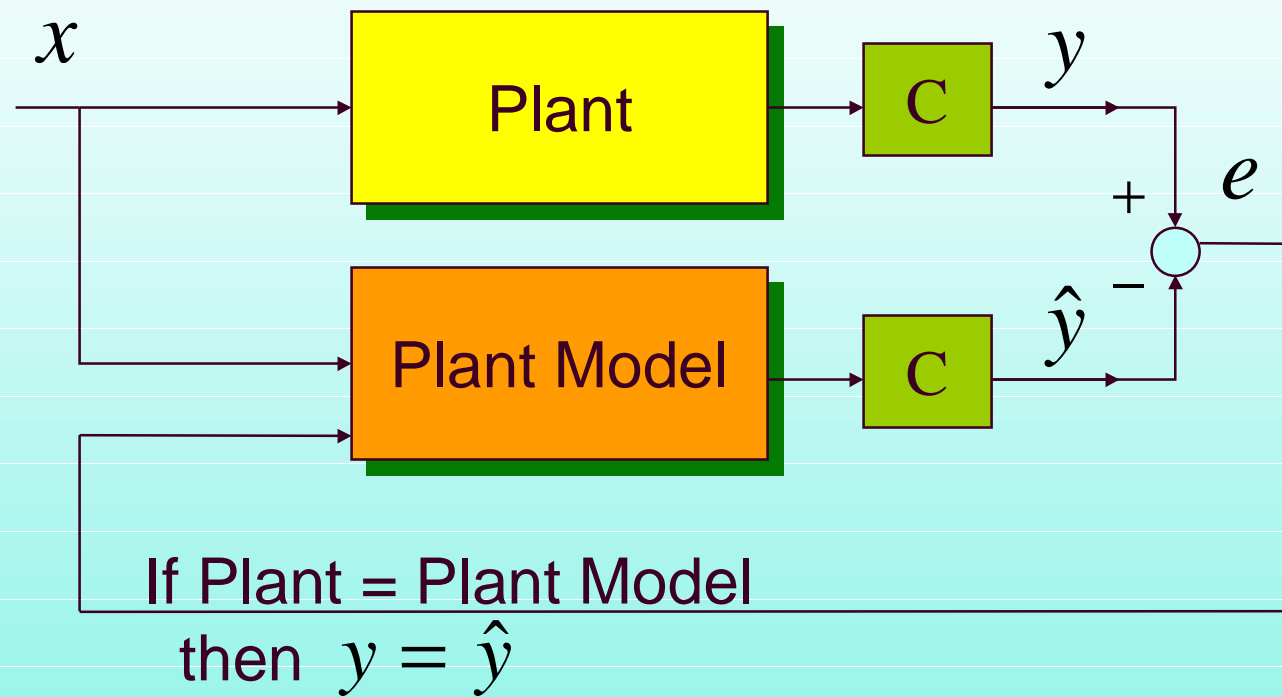
$$\begin{bmatrix} \dot{w} \\ \dot{q} \\ \dot{\theta} \\ \dot{z} \end{bmatrix} = \begin{bmatrix} a_{11} u & a_{12} u & a_{13} & 0 \\ a_{21} u & a_{22} u & a_{23} & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & -u & 0 \end{bmatrix} \begin{bmatrix} w \\ q \\ \theta \\ z \end{bmatrix} + \begin{bmatrix} b_{11} u^2 & b_{12} u^2 \\ b_{21} u^2 & b_{22} u^2 \\ 0 & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \delta_b \\ \delta_s \end{bmatrix} \\ + \begin{bmatrix} F_d \cos \theta \\ M_d \cos \theta \\ 0 \\ 0 \end{bmatrix} + \begin{bmatrix} e_{11} q^2 + e_{12} qw \\ e_{21} q^2 + e_{22} qw \\ 0 \\ 0 \end{bmatrix}$$

Output Equation

$$\begin{bmatrix} z \\ q + \theta \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 \end{bmatrix} \begin{bmatrix} w \\ q \\ \theta \\ z \end{bmatrix}$$



Observer Design



Observer Design(Cont'd)

$$\dot{x} = Ax + f(t, x) + B[u(t) + v(t)]$$

$$y = Cx$$

Assumptions

A1) The system is detectable so that we can find G .
The spectrum of $A - GC = 0$ is in the LHP

A2) There exists $Q \in R^{n \times m}$, symmetric and positive definite,
and function h and w such that:

$$f(t, x) = P^{-1}C^T h(t, x)$$

$$Bv(t) = P^{-1}C^T w(t)$$

$$\text{where } A_0^T P + PA_0 = -Q$$

A3) Let $\xi(t, x) = h(t, x) + w(t, x)$. Then, there exists a positive scalar

$$\text{Such that: } \|\xi(t, x)\| < \rho$$



Observer Design(Cont'd)

Sliding Mode Observer

$$\dot{\hat{x}} = A_o \hat{x} + Gy + S(\hat{x}, y) + Bu$$

$$\text{where } S(\hat{x}, y) = \begin{cases} -\frac{P^{-1}C^T C e}{\|Ce\|} \rho & \text{for } e \neq 0 \\ 0 & \text{for } e = 0 \end{cases}$$

- We measure the depth and the pitch rate.
- The pitch angle can be obtained by integrating pitch rate.



Controller Design

- PD Controller

$$\begin{bmatrix} \delta_b \\ \delta_s \end{bmatrix} = \begin{bmatrix} K_{11} & K_{12} & K_{13} & K_{14} \\ K_{21} & K_{22} & K_{23} & K_{24} \end{bmatrix} \begin{bmatrix} w - w_{com} \\ q - q_{com} \\ \theta - \theta_{com} \\ z - z_{com} \end{bmatrix}$$

- Sliding Mode Controller

Switching surface

$$\sigma(t, x) = [\sigma_1(t, x) \quad \dots \quad \sigma_m(t, x)]^T = 0$$

Reaching condition

$$\frac{d}{dt} \left(\frac{1}{2} \sigma^T \sigma \right) = \sigma^T \dot{\sigma} < 0$$



Controller Design(Cont'd)

Gradient of σ

$$\frac{\partial \sigma(t, x)}{\partial x} = S(t, x)$$

Control input

$$u = u_{eq_{nom}} + \bar{u}$$

$$\begin{cases} u_{eq_{nom}} = -[S(t, x)B]^{-1}(S(t, x)Ax + \frac{\partial \sigma}{\partial t}) \\ \bar{u} = -[S(t, x)B]^{-1} \eta \text{sign}(\sigma) \end{cases}$$



Simulation results

- Nominal forward speed 6 knot
- Bounds on nonlinear terms

$|\theta| < 10^\circ$: pitch angle

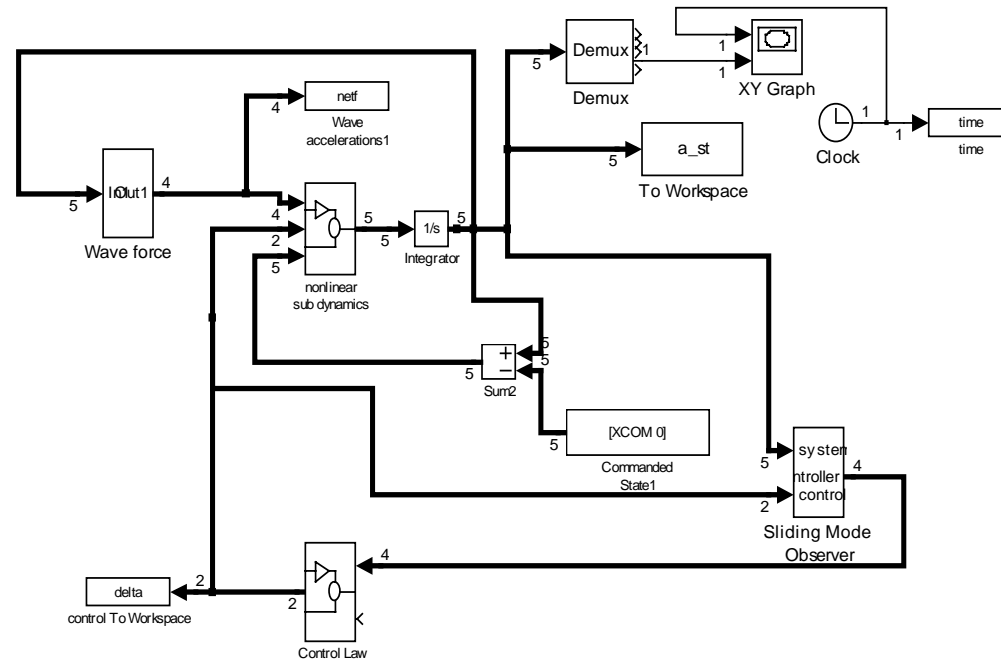
$|q| < 0.2 \text{ rad / sec}$: pitch rate

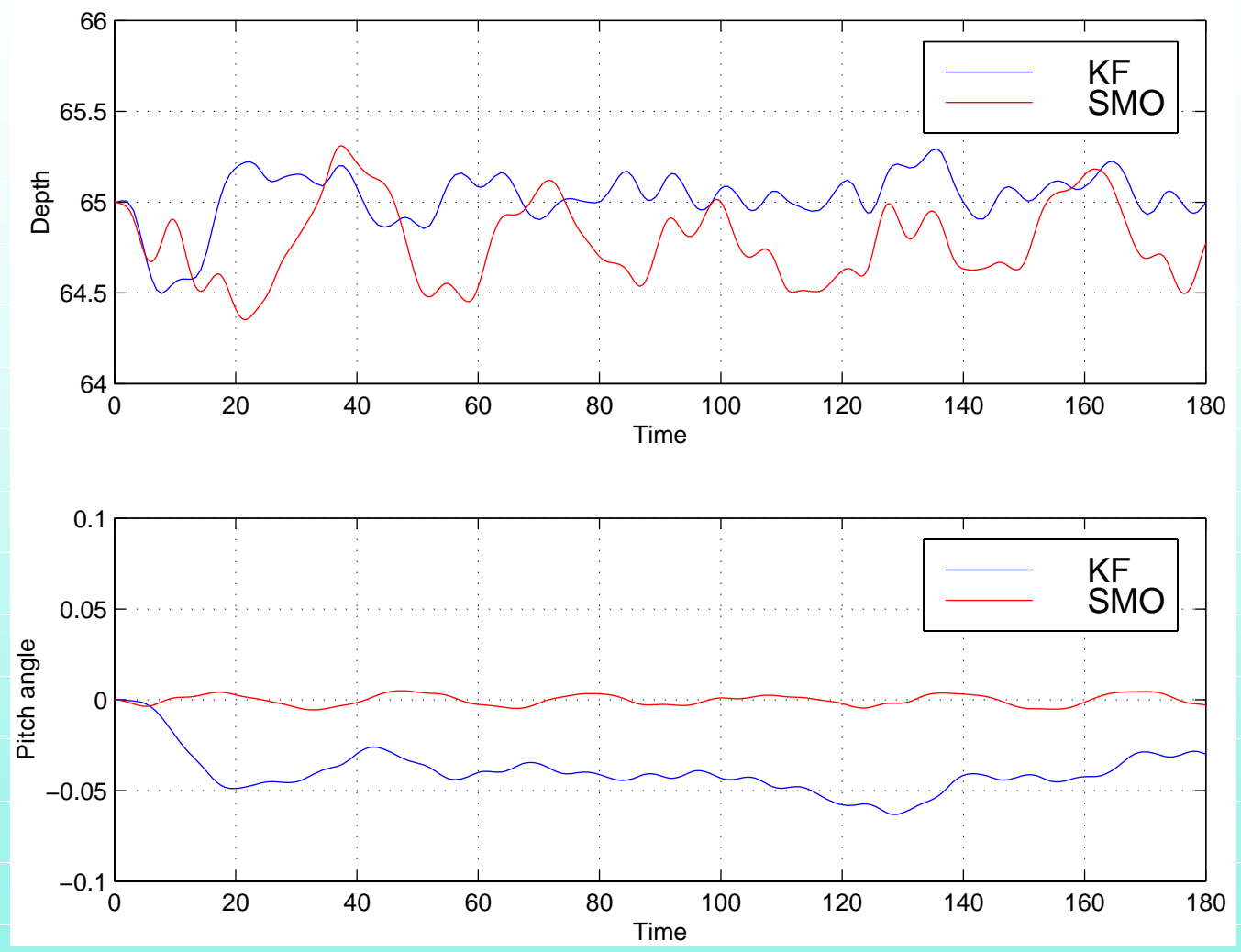
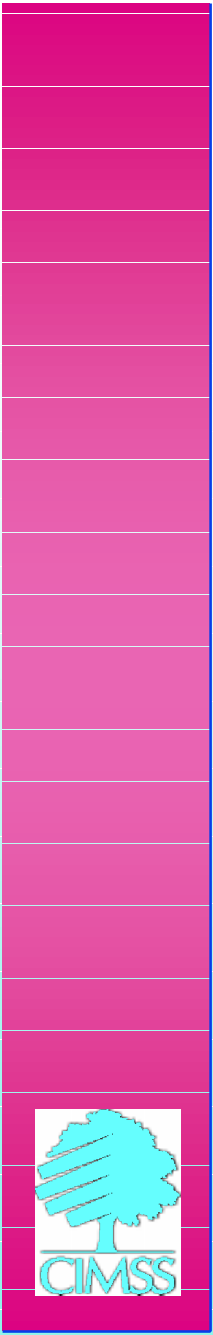
$|w| < 1.5 \text{ ft / s}$: heave speed

- Include bounds of wave force

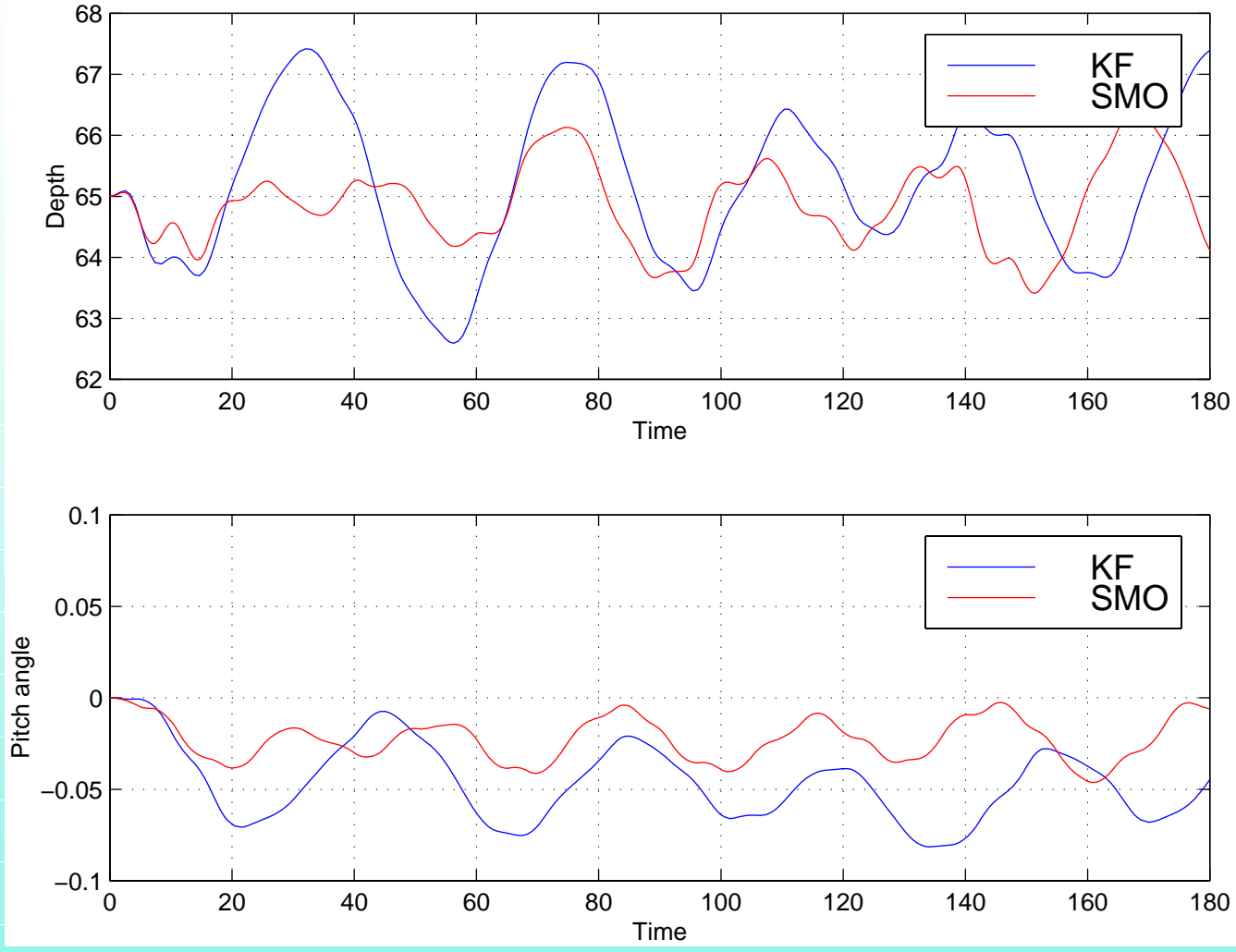
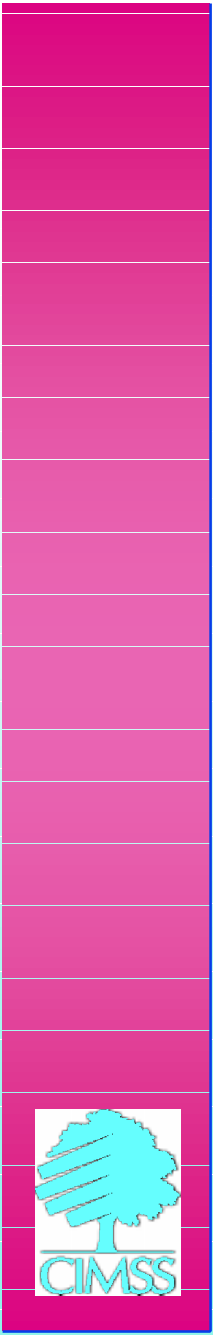


Simulink Model

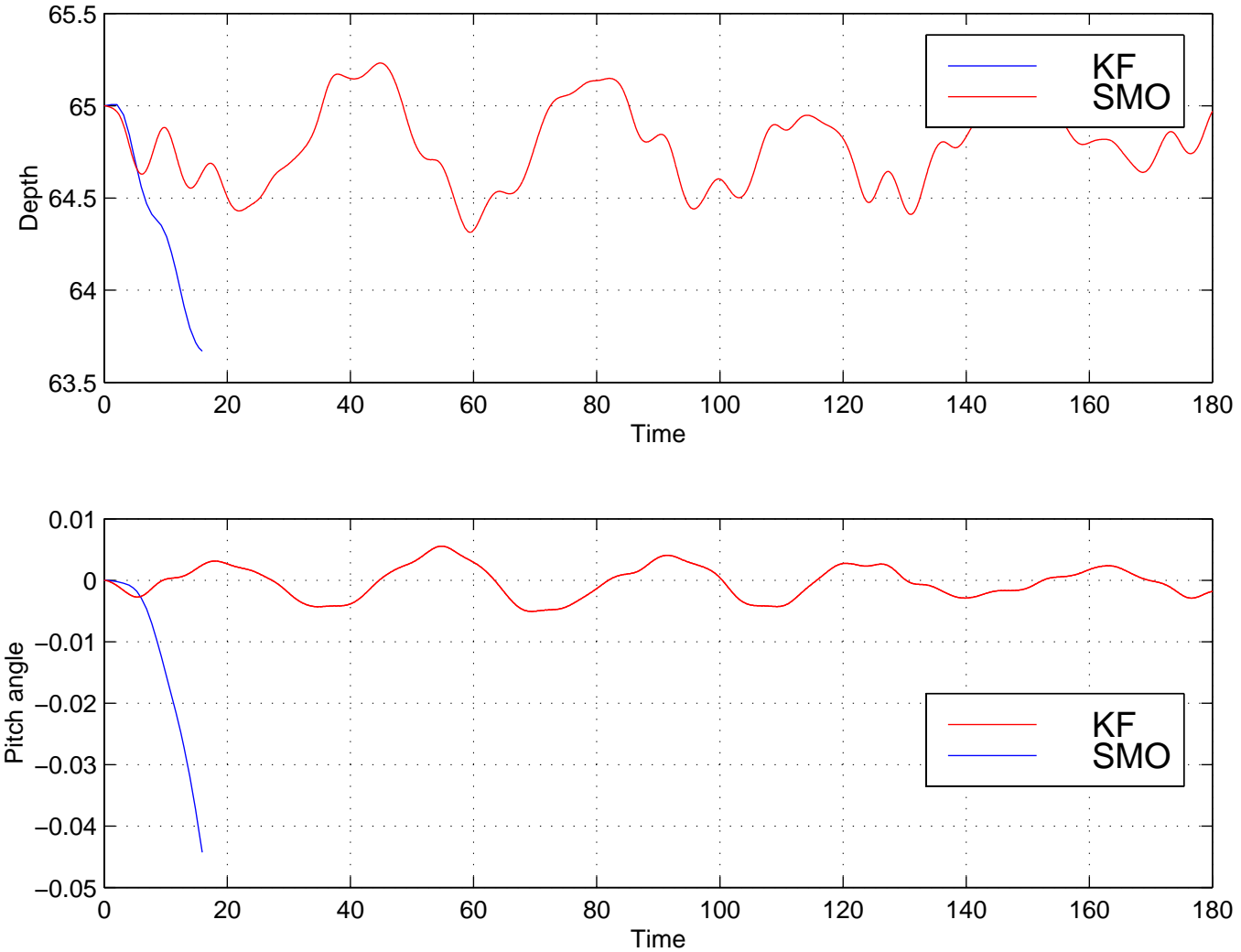




Depth and pitch response to 65ft dive command at sea state 3
-PD control-

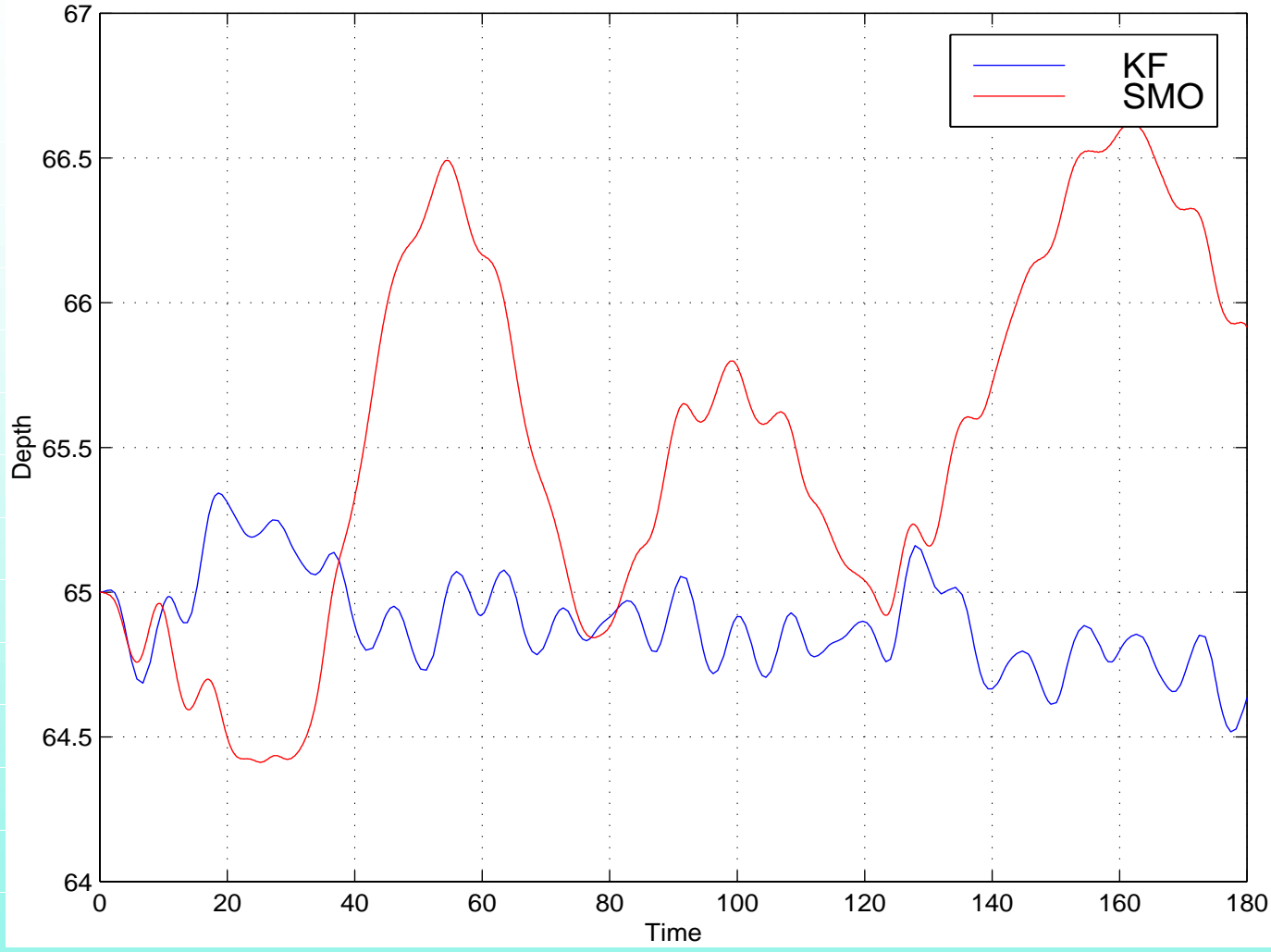
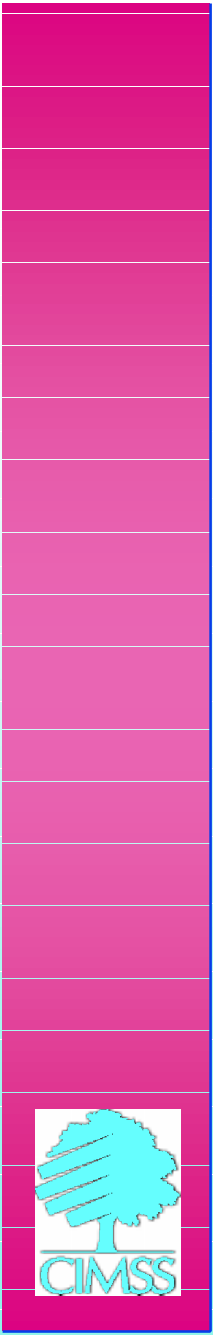


Depth and pitch response to 65ft dive command at sea state 4
-PD control-



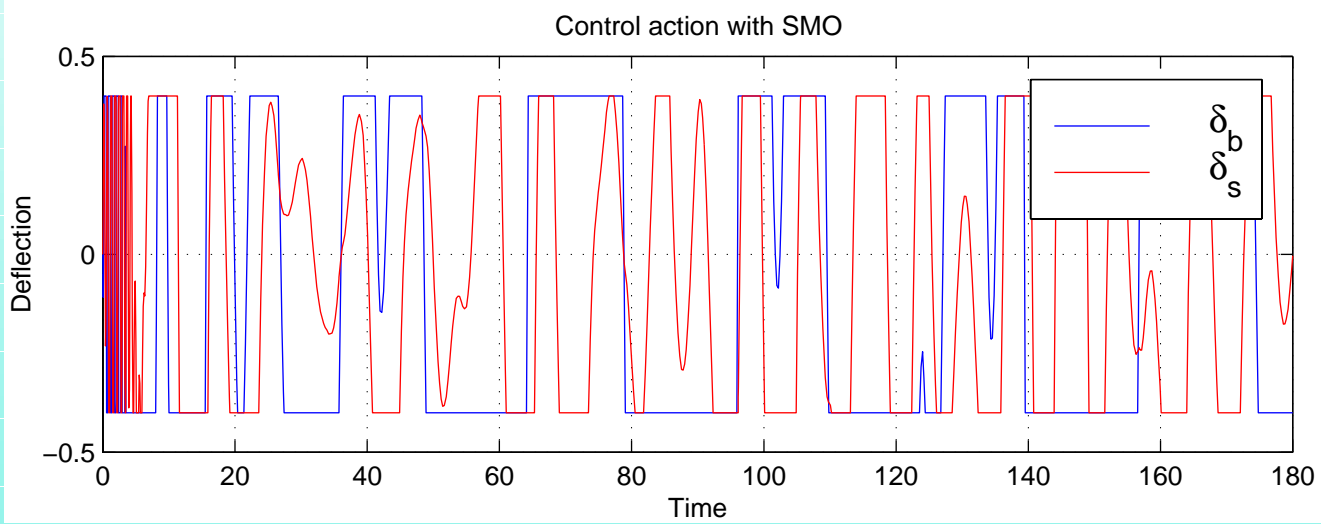
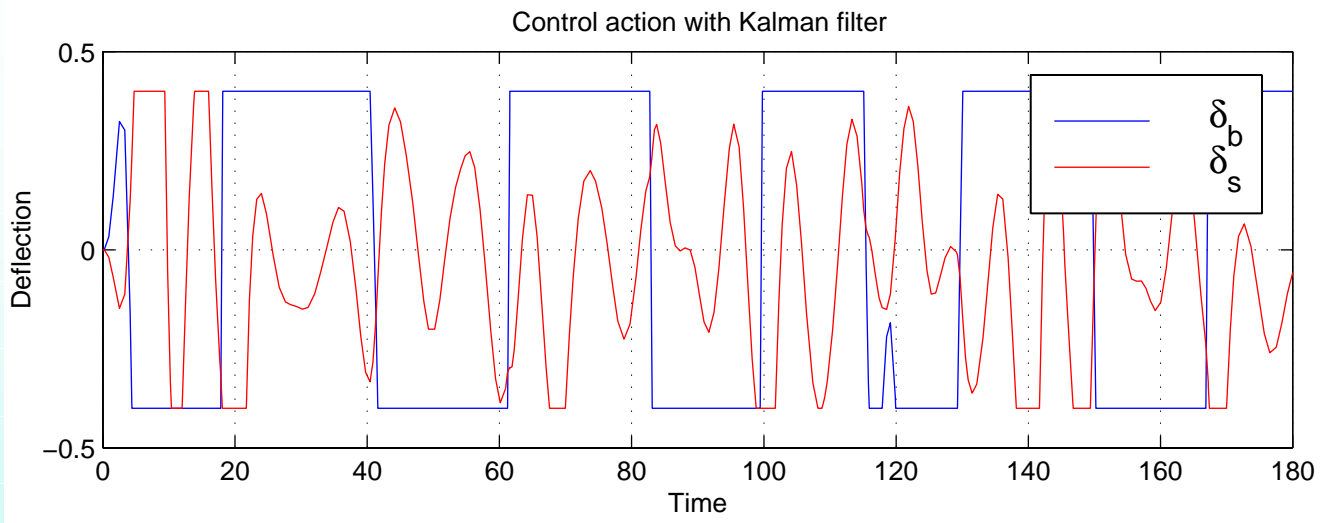
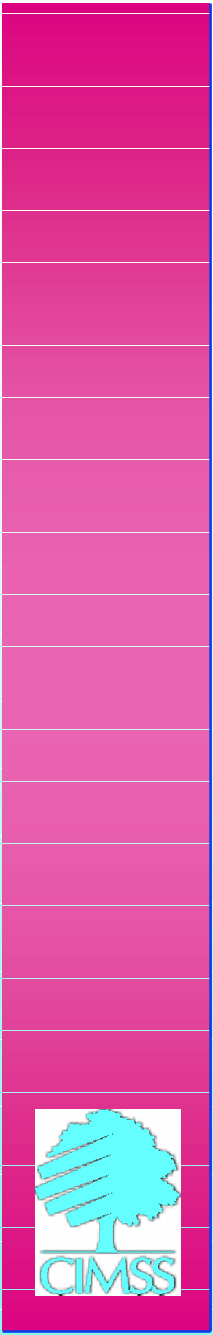
Depth and pitch response to 65ft dive command
with 30% parameter variation
-PD control-





Depth and pitch response to 65ft dive command at sea state 3
-Sliding mode control-



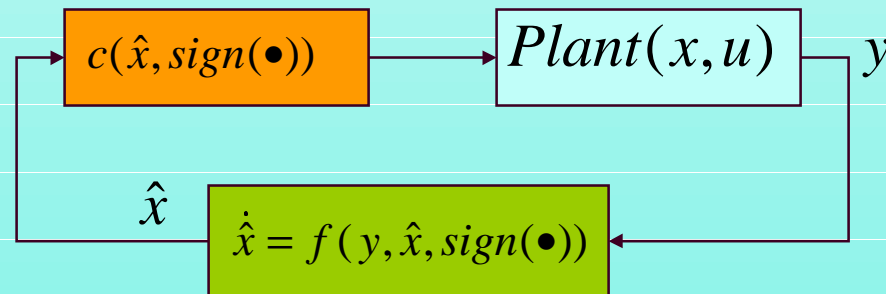


Comparison of control action

SMC and SMO Combination

- Unexpected results are observed when SMC(Sliding Mode Controller) and SMO(Sliding Mode Observer) are combined.
- Possible problem
 - ◆ nested discontinuities

$$f = \text{sign}(\text{sign}(u))$$



Conclusion

- Sliding mode observer is designed and applied in the dive plane of submarine in order to estimate linear velocity component (heave velocity).
- It is compared with the Kalman filter estimator which is based on linearized dynamics and showed better performance in terms of robustness to wave disturbance, model parameter uncertainties, and parameter variation.
- When, however, the SMC and SMO are combined, undesirable results are obtained. More investigation is required.
- This sliding mode observer can be applied to fault detection algorithm in replace of Kalman filter.
- It takes more computational time in estimating the states when sliding mode observer is used. More efficient computational algorithm needs to be investigated.

