

Lecture #5

(p1)

Robust Control of Nonlinear Systems

Models are usually imprecise

- actual plant uncertainty (unknown parameters)
- purposeful simplified representation

Robust controller deals w/ such imprecision

- through a nominal controller + additional terms to deal w/ uncertainty

For example,

(lots of systems come in this form)

~~Consider~~ Consider the scalar n^{th} order d.e.

$$\ddot{x}^{(n)} = f(x, \dot{x}, \dots, x^{(n-1)}, t) + b(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t) u$$

If f and b are known, and we had a trajectory tracking problem, we can define

$$\ddot{\tilde{x}}(t) = x(t) - x_{\text{des}}(t)$$

$$\text{and send } \ddot{\tilde{x}}(t) \rightarrow 0$$

We do this by setting $v = f + bu$ and then solve for u .

But of course, we do not know f & b

e.g. Robot in the presence of underwater currents

- robots w/ timevarying loads.
or
unknown

So you define

$$\| \underset{\substack{\uparrow \\ \text{unknown}}}{f(x,t)} - \hat{f}(x,t) \| \leq \underbrace{F(x,t)}_{\substack{\text{Some estimate of drift} \\ \downarrow \\ \text{worst-case}}}$$

and similarly for $b(x,t)$

Typical problem is to track trajectory and keep error small.

The key idea is to replace n^{th} order system w/ first order sys.

- and take advantage of simplicity of first order problem

(\rightarrow if output negative, apply +ve input
if " + , " -ve ")

Of course, this new variable when controlled represents original system

So create intermediate variable z :

such that (1) z contains u
and

$$(2) \ z \rightarrow 0 \Rightarrow \ddot{x}(t) \rightarrow 0$$

Example

$$n=2 \quad \ddot{x} = f + bu$$

$$\text{Pick } z = \dot{x}$$

or

$$z = \dot{x}_d - \dot{x}$$

But this does not satisfy requirement (2)

If $z \rightarrow 0$ only $\dot{x} \rightarrow 0$, not $\ddot{x}(t)$

Suppose we choose to satisfy the 2nd condition

(p3)

$$s = \frac{u}{x}$$

then as $s \rightarrow 0$, $\frac{u}{x} \rightarrow 0$. So that is good.

But that does not satisfy the 1st condition

So let's mix the two.

Choose $s = \frac{0}{x} + \lambda \frac{u}{x}$

\swarrow
this would satisfy condition #1

\downarrow
scalar > 0

Regarding condition #2, this equation is

like $s \rightarrow \boxed{\frac{1}{\frac{d}{dt} + \lambda}} \rightarrow \frac{u}{x}$

filter.

therefore as $s \rightarrow 0$, $\frac{u}{x} \rightarrow 0$.

For the n^{th} order case,

choose

$$s = \left(\frac{d}{dt} + \lambda \right)^{n-1} \frac{u}{x}$$

$n=3 \Rightarrow s = \frac{0}{x} + 2\lambda \frac{0}{x} + \lambda^2 \frac{u}{x}$

The reasoning is the same.

(1) s contains $\left(\frac{d}{dt} \right)^n \frac{u}{x}$ and therefore contains u .

(2) Now you have n^{th} order filter

$$s \rightarrow \boxed{\frac{1}{\left(\frac{d}{dt} + \lambda \right)^{n-1}}} \rightarrow \frac{u}{x}$$

Stable linear filter

So you replace n^{th} order problem w/ first order problem.

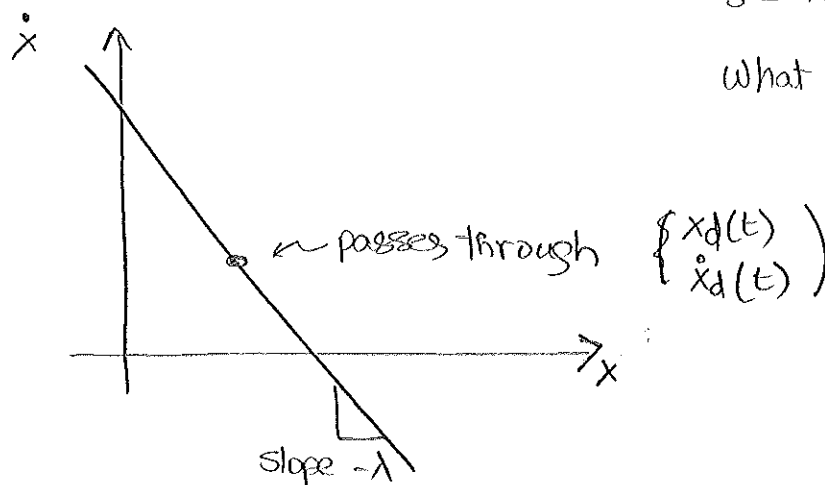
If you send $s \rightarrow 0$, $\frac{u}{x} \rightarrow 0$

Let's look @ this ^{second-order system} geometrically:

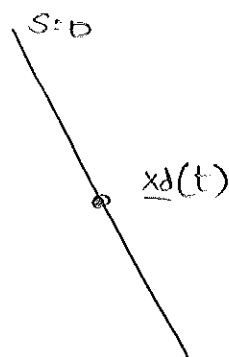
(p4)

$$\ddot{x} = \ddot{x} + \lambda \dot{x}$$

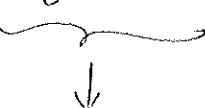
What does $s=0$ represent?



Let's go to n^{th} order system. The line becomes a ^{hyper}plane ($s=0$)

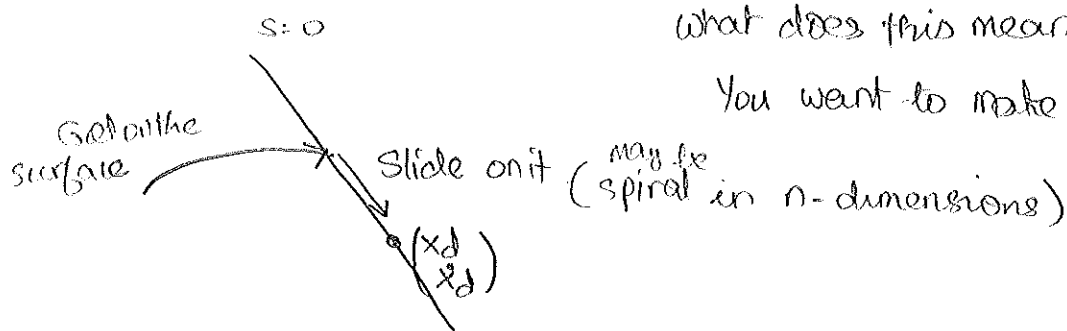


You want the system to tend to \underline{x}_d and stay there.



What does this mean?

You want to make it ^{an} invariant set.



So $s=0$ is both a place & a dynamics

Once you get onto hyperplane, motion is completely defined.

what if I can only guarantee

(p5)

$$|s| \leq \phi \text{ constant?}$$

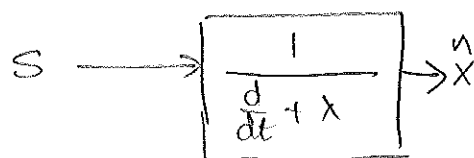
How does the residual error in z translate to x ?

let's examine $n=2$ case

$$z = \overset{0}{\underset{\sim}{x}} + \lambda \overset{y}{\underset{\sim}{x}}$$

$$s: (p + \lambda) \overset{y}{\underset{\sim}{x}}$$

$$\frac{\overset{y}{\underset{\sim}{x}}}{s} = \frac{1}{p + \lambda}$$



we know that $|s| \leq \phi$ after transient. (effect of initial conditions goes away)

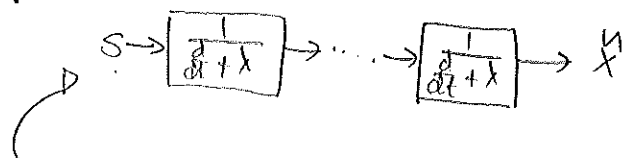
So the form for

$$\overset{y}{\underset{\sim}{x}}(t) = \underbrace{(\text{effect of initial cond'n})}_{\rightarrow 0} + \int_0^t e^{-\lambda(t-r)} z(r) dr$$

since $\lambda > 0$

$$\begin{aligned} \text{Now } \left| \int_0^t e^{-\lambda(t-r)} z(r) dr \right| &\leq \int_0^t \left| e^{-\lambda(t-r)} z(r) \right| dr \\ &\leq \phi \int_0^t |e^{-\lambda(t-r)}| dr \\ &\leq \frac{\phi}{\lambda} \end{aligned}$$

$\Rightarrow \overset{y}{\underset{\sim}{x}}$ bounded by $\frac{\phi}{\lambda}$



what about n^{th} order sys? similar calculation w/ $(n-1)$ consecutive filters.

$$\text{If } |z| \leq \phi, \text{ then } \underset{(\text{error})}{|\overset{y}{\underset{\sim}{x}}|} \leq \frac{\phi}{\lambda^{n-1}}$$

So let's solve the first order problem (we have delayed it all this while!)

consider second order sys.

7.1.3 $\ddot{x} = g(x,t) + b(x,t)u$

$$\ddot{x} \neq a(t) \dot{x}^2 \cos 3x \pm u \quad w/ \quad 1 \leq a(t) \leq 2$$

similar to a drag term

choose $\hat{f} = \underbrace{-1.5 \dot{x}^2}_{\downarrow \text{middle of range.}} \cos 3x$

So $F = 0.5 \dot{x}^2 |\cos 3x|$

$$s = \ddot{x} + \lambda \dot{x}$$

$$\dot{s} = \ddot{\ddot{x}} + \lambda \ddot{x} = f + bu - \ddot{x}d + \lambda \dot{x}$$

choose u such that $\frac{d}{dt} s^2 \leq 0$ (we want $|s|$ to always decrease)

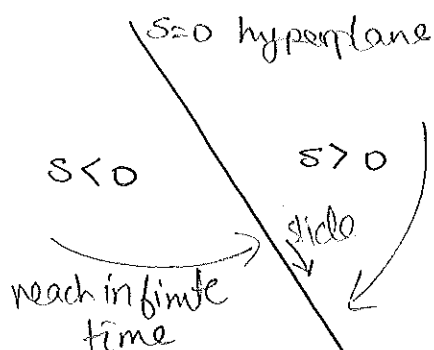
or more precisely

$$\frac{1}{2} \frac{d}{dt} s^2 \leq \underbrace{-\eta |s|}_{\text{constant} > 0}$$

called sliding condition

what does this mean?

what if we start w/ $s > 0$



$$\Rightarrow s \dot{s} \leq -\eta s$$

If $s > 0$

$$\dot{s} \leq -\eta$$

$\Rightarrow s=0$ is reached in finite time

$$\left(\leq \frac{s(t=0)}{\eta} \right)$$

trapped on this surface (invariant set)
then first order system takes over.

choose $b = 1$ for simplicity

(p7)

Define \hat{u} such that $\dot{s} = 0$ if you know $f + b$ exactly.

$$\text{i.e. } \hat{u} = -\hat{f} + \ddot{x}_d \approx \lambda \ddot{y}$$

(i.e. $b = \hat{b}$)

with $u = \hat{u}$

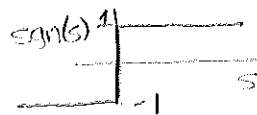
$$\dot{s} = \dot{b} - \dot{\hat{b}}$$

Also $\dot{b} - \dot{\hat{b}}$ may be of arbitrary sign

$$\Rightarrow \frac{1}{2} \frac{d}{dt} s^2 = s \dot{s} = s(\dot{b} - \dot{\hat{b}})$$

But this does not necessarily mean that $|\dot{b} - \dot{\hat{b}}| \leq \eta$

Add one more term.



$$u = \hat{u} - k \text{sign}(s)$$

push in the right direction

$$\text{Then } s\dot{s} = s(\dot{b} - \dot{\hat{b}}) - k|s| \stackrel{\text{needs to be}}{\leq} -\eta|s|$$

Choose $k = F + \eta$ to satisfy this condition

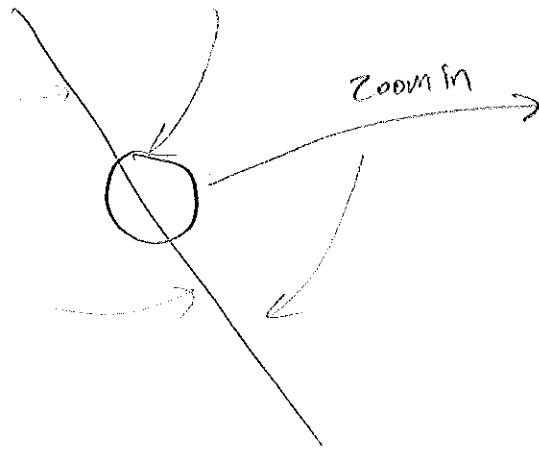
So for the example. \rightarrow uncertainty in b

$$u = 1.5 \ddot{x}^2 \cos 3x + \ddot{x}_d - \lambda \ddot{y} - (0.5 \ddot{x}^2 |\cos 3x| + \eta) \text{sign}(\ddot{y} + \lambda \ddot{x})$$

Why does this work?

You have chosen control law so that $s \dot{s}$ is pushed on to the surface $s = 0$ always (the effect of the term $k|s|$)

Let's look closely @ performance though.



may not seem like a big deal
in the s space but may
be a big deal in the
actuator space.

- large oscillations

Maybe you can ignore small tracking error and change the
control law to ~~not act~~ ^{not act} for such small errors.

(ignore $|e| \leq \phi$)