

## Chapter #5

### EEE8086-EEE8115

## Robust and Adaptive Control Systems

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## Robust Control

### 1. Ideal Systems

Assume that we have a  $2^{nd}$  order system:

$$\ddot{x} + A\dot{x} + Bx = u \quad (1)$$

and that we want to follow a specific signal  $x_d$

then if we choose  $u = \ddot{x}_d + A\dot{x}_d + Bx_d$ :

$$\begin{aligned} \ddot{x} + A\dot{x} + Bx &= \ddot{x}_d + A\dot{x}_d + Bx_d \Leftrightarrow \\ (\ddot{x} - \ddot{x}_d) + A(\dot{x} - \dot{x}_d) + B(x - x_d) &= 0 \end{aligned}$$

We can define now as a tracking error:

$$\tilde{x} = x - x_d \quad (2)$$

and hence the error dynamics are given by:

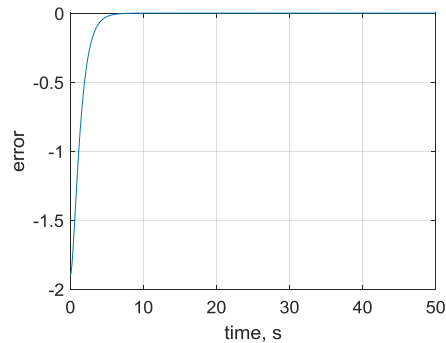
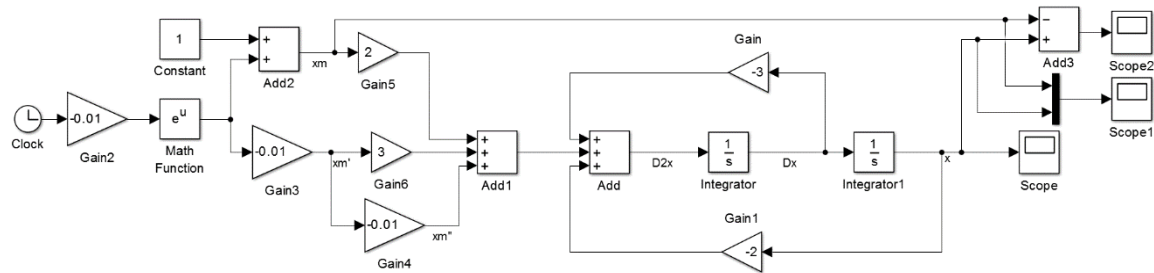
$$\ddot{\tilde{x}} + A\dot{\tilde{x}} + B\tilde{x} = 0 \quad (3)$$

So if the scalars  $A$  and  $B$  define a stable system the tracking error will converge to zero and hence the system will have the desired response.

**Example 1.1:** A system is given by the following  $2^{nd}$  order ODE:

$$\ddot{x} + 3\dot{x} + 2x = u \text{ and we want it to track the desired trajectory } x_d = 1 + e^{-0.01t}$$

Then we can set the control signal  $u = \ddot{x}_d + 3\dot{x}_d + 2x_d$  and the response is:



If on the other hand the matrices  $A$  and  $B$  are not stable (or fast enough) we can use:

$$u = A\dot{x} + Bx - C\ddot{x} - D\dot{x} + \ddot{x}_d \quad (4)$$

which will give me:

$$\ddot{x} + A\dot{x} + Bx = A\dot{x} + Bx - C\ddot{x} - D\dot{x} + \ddot{x}_d \Leftrightarrow$$

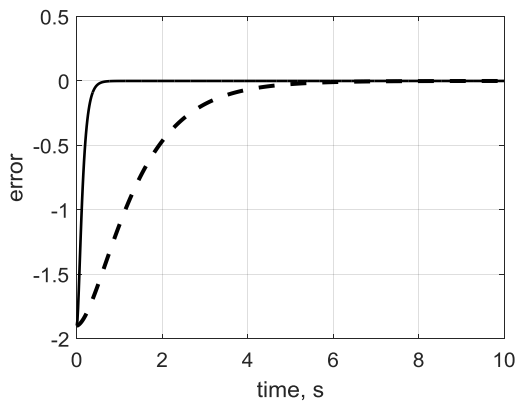
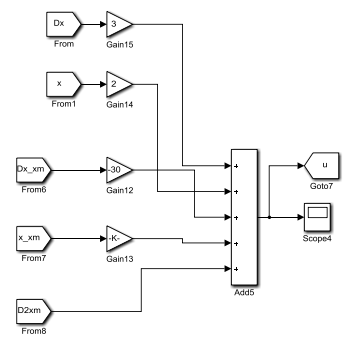
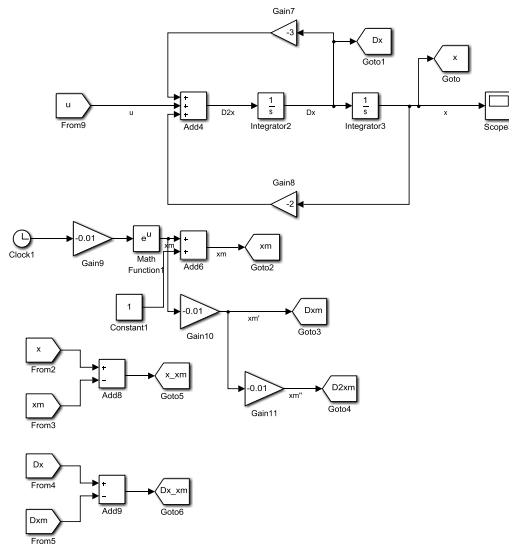
$$\ddot{x} - \ddot{x}_d + C\ddot{x} + D\dot{x} = 0 \Leftrightarrow$$

$$\ddot{x} + C\ddot{x} + D\dot{x} = 0$$

and hence if we properly choose  $C$  and  $D$  in order to have fast and stable dynamics for the error we can again ensure that the system will have the desired response.

**Example 1.2:** Assume that we want to make the speed of the error 10 times faster, i.e. to place the poles of the error dynamics at -10 and -20:

$$u = 3\dot{x} + 2x - 30\dot{\hat{x}} - 200\hat{x} + \ddot{x}_d$$



This method can be applied now to a nonlinear  $n^{\text{th}}$  order system as if

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

We can always choose:

$$u = \frac{-h\left(\tilde{x}, \dot{\tilde{x}}, \ddot{\tilde{x}}, \dots, \tilde{x}^{(n-1)}, t\right) + x_d^{(n)} - f\left(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t\right)}{g\left(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t\right)}$$

$$x^{(n)} = f\left(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t\right) + g\left(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t\right) \left( \frac{-h\left(\tilde{x}, \dot{\tilde{x}}, \ddot{\tilde{x}}, \dots, \tilde{x}^{(n-1)}, t\right) + x_d^{(n)} - f\left(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t\right)}{g\left(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t\right)} \right)$$

$$x^{(n)} = -h\left(\tilde{x}, \dot{\tilde{x}}, \ddot{\tilde{x}}, \dots, \tilde{x}^{(n-1)}, t\right) + x_d^{(n)} \Leftrightarrow$$

$$x^{(n)} - x_d^{(n)} = -h\left(\tilde{x}, \dot{\tilde{x}}, \ddot{\tilde{x}}, \dots, \tilde{x}^{(n-1)}, t\right) \Leftrightarrow$$

$$\tilde{x}^{(n)} + h\left(\tilde{x}, \dot{\tilde{x}}, \ddot{\tilde{x}}, \dots, \tilde{x}^{(n-1)}, t\right) = 0$$

Hence if we properly choose the function  $h$  we can make sure that regardless of the desired signal the system will behave satisfactory.

## 2. Sliding mode control

### 2.1 Ideal systems

Assume that we have a system:

$$x^{(n)} = f\left(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t\right) + g\left(x, \dot{x}, \ddot{x}, \dots, x^{(n-1)}, t\right)u \quad (5)$$

with a desired tracking trajectory  $x_d(t)$ , the error between the desired and real trajectory is defined as:

$$\tilde{x}(t) = x(t) - x_d(t) \quad (6)$$

We know that studying an  $n^{\text{th}}$  order nonlinear system is a cumbersome task, while linear systems are much easier to handle. So the first question that we have here is, can a linear system represent our system given in (5)? Let's denote the variable “ $s$ ”<sup>1</sup> whose ODE describes our system. We impose 2 properties on  $s$  here:

1. We need to differentiate  $s$  only once in order to have an expression of the control signal  $u$ .
2. When  $s \rightarrow 0 \Rightarrow \tilde{x}(t) \rightarrow 0$

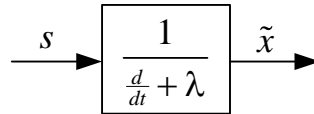
If we have a 2<sup>nd</sup> order system:  $\ddot{x} = f(x, \dot{x}, t) + g(x, \dot{x}, t)u$  then the conditions are verified if we choose:

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<sup>1</sup> Do not be confused, this  $s$  has nothing to do with the Laplace variable.

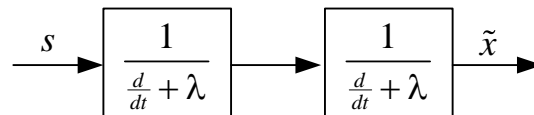
$$s = \dot{\tilde{x}} + \lambda\tilde{x} = \left( \frac{d}{dt} + \lambda \right) \tilde{x} \quad (7)$$

which can be seen as a stable linear filter:

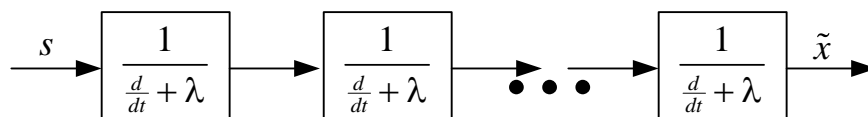


For 3<sup>rd</sup> order systems:

$$s = \left( \frac{d}{dt} + \lambda \right)^2 \tilde{x} = \left( \frac{d^2}{dt^2} + 2\lambda \frac{d}{dt} + \lambda^2 \right) \tilde{x} = \frac{d^2 \tilde{x}}{dt^2} + 2\lambda \frac{d\tilde{x}}{dt} + \lambda^2 \tilde{x}$$



For the general case  $s = \left( \frac{d}{dt} + \lambda \right)^{n-1} \tilde{x}$  (8)

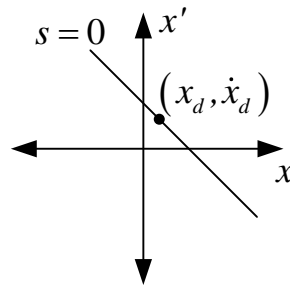


So by studying the error dynamics (given by (8)) we have replaced a nonlinear system with a linear one (and of a smaller order).

**Our task now is to find the control law that will make the ODE given by (8) a stable one, i.e. that the error tends to zero (in finite time).**

From this point we will focus on 2<sup>nd</sup> order systems, but the same analysis can be carried out in a general  $n^{\text{th}}$  order system.

Geometrically the condition  $s = \dot{\tilde{x}} + \lambda\tilde{x}$  with  $s$  is zero we have  $\dot{\tilde{x}} + \lambda\tilde{x} = 0$  or  $(\dot{x} - \dot{x}_d) + \lambda(x - x_d) = 0$ . Which is the equation of a straight line (or a surface in  $n^{\text{th}}$  dimensional systems) in the state space:



Now let's try to solve the ODE of the error dynamics (by assuming that  $s=0$ ):  $\dot{\tilde{x}} + \lambda\tilde{x} = 0 \Rightarrow \tilde{x}(t) = \tilde{x}(t_0)e^{-\lambda t}$  which implies that  $\dot{\tilde{x}} = 0, \tilde{x} = 0$ . Hence, if the trajectory at some point hits the surface defined by  $s=0$  at  $t=t_0$  we have that  $\tilde{x}(t) = 0, \forall t > t_0$ . Hence the surface defined by  $s$  is invariant and this implies that we will have the desired response  $\forall t > t_0$ .

Note: At this point we have **NOT** solved our control problem. We have just changed a nonlinear problem with a linear one and we have seen its geometric interpretation in the state space. The task now is to find the control law  $u$  that will make  $s=0$  in finite time.

Now assume that you have a 2<sup>nd</sup> order system:

$$\ddot{x} = f(x, \dot{x}, t) + g(x, \dot{x}, t)u$$

We want to converge to  $s=0$  by the appropriate choice of  $u$ . In order to guarantee that the above equation is stable we can look for a Lyapunov



function like:  $V(s) = \frac{1}{2}s^2$  with  $V(0) = 0$  and  $V(s) > 0, s \neq 0$ . So now we have

to find the appropriate  $u$  such as  $\frac{dV(s)}{dt} = s\dot{s} < 0$  and hence according to

Lyapunov to have a stable system. One obvious way to make sure that is to set  $\dot{s} = -s$ :

$$s = \dot{\tilde{x}} + \lambda\tilde{x} \Leftrightarrow \dot{s} = \ddot{\tilde{x}} + \lambda\dot{\tilde{x}} = \ddot{x} - \ddot{x}_d + \lambda\dot{\tilde{x}} = f(x, \dot{x}, t) + g(x, \dot{x}, t)u - \ddot{x}_d + \lambda\dot{\tilde{x}}$$

$$\text{Hence: } f(x, \dot{x}, t) + g(x, \dot{x}, t)u - \ddot{x}_d + \lambda\dot{\tilde{x}} = -\dot{\tilde{x}} - \lambda\tilde{x} \Leftrightarrow$$

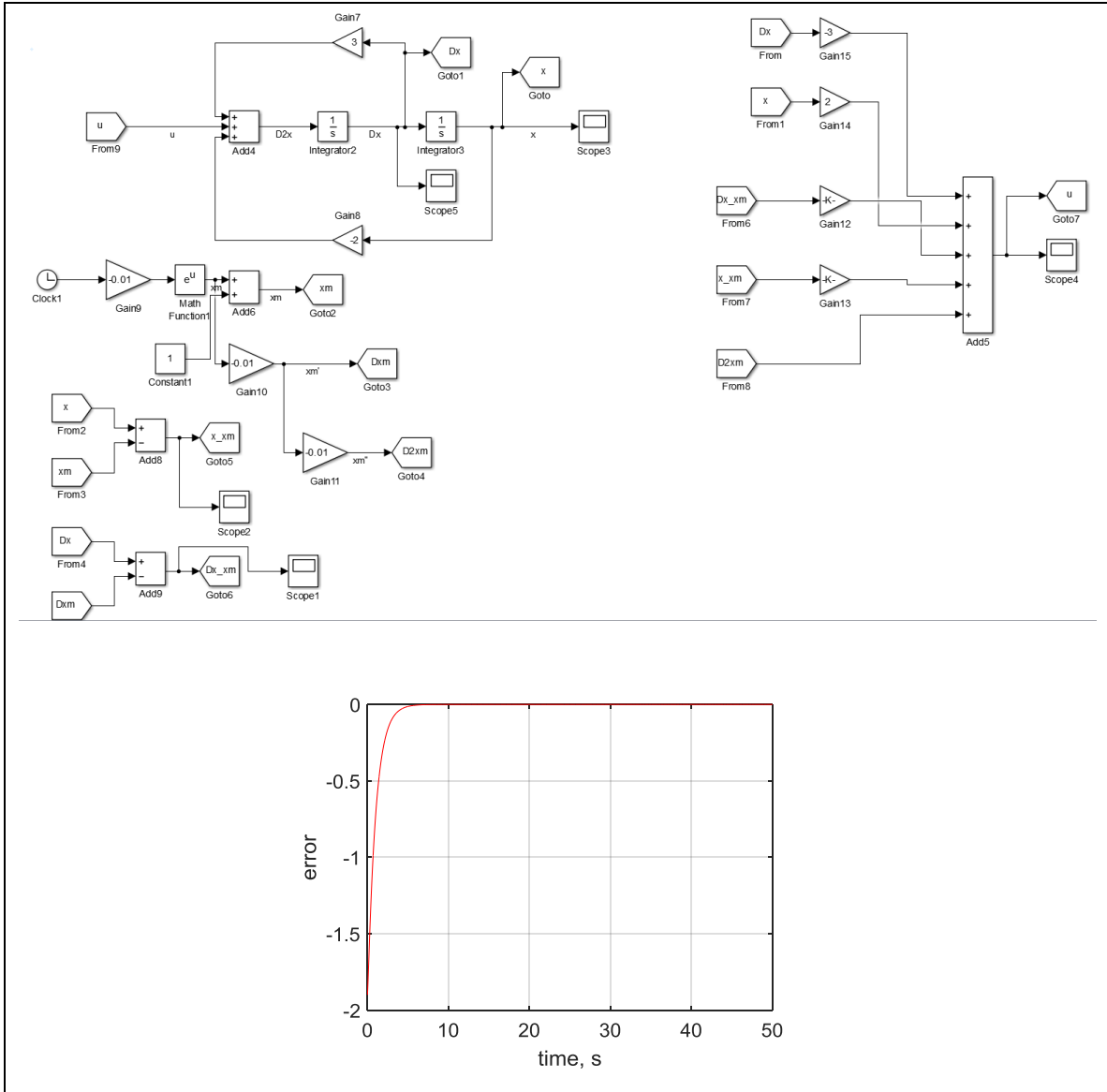
$$u = \frac{1}{g(x, \dot{x}, t)} \left( -\dot{\tilde{x}} - \lambda\tilde{x} + \ddot{x}_d - \lambda\dot{\tilde{x}} - f(x, \dot{x}, t) \right)$$

In this case we have:

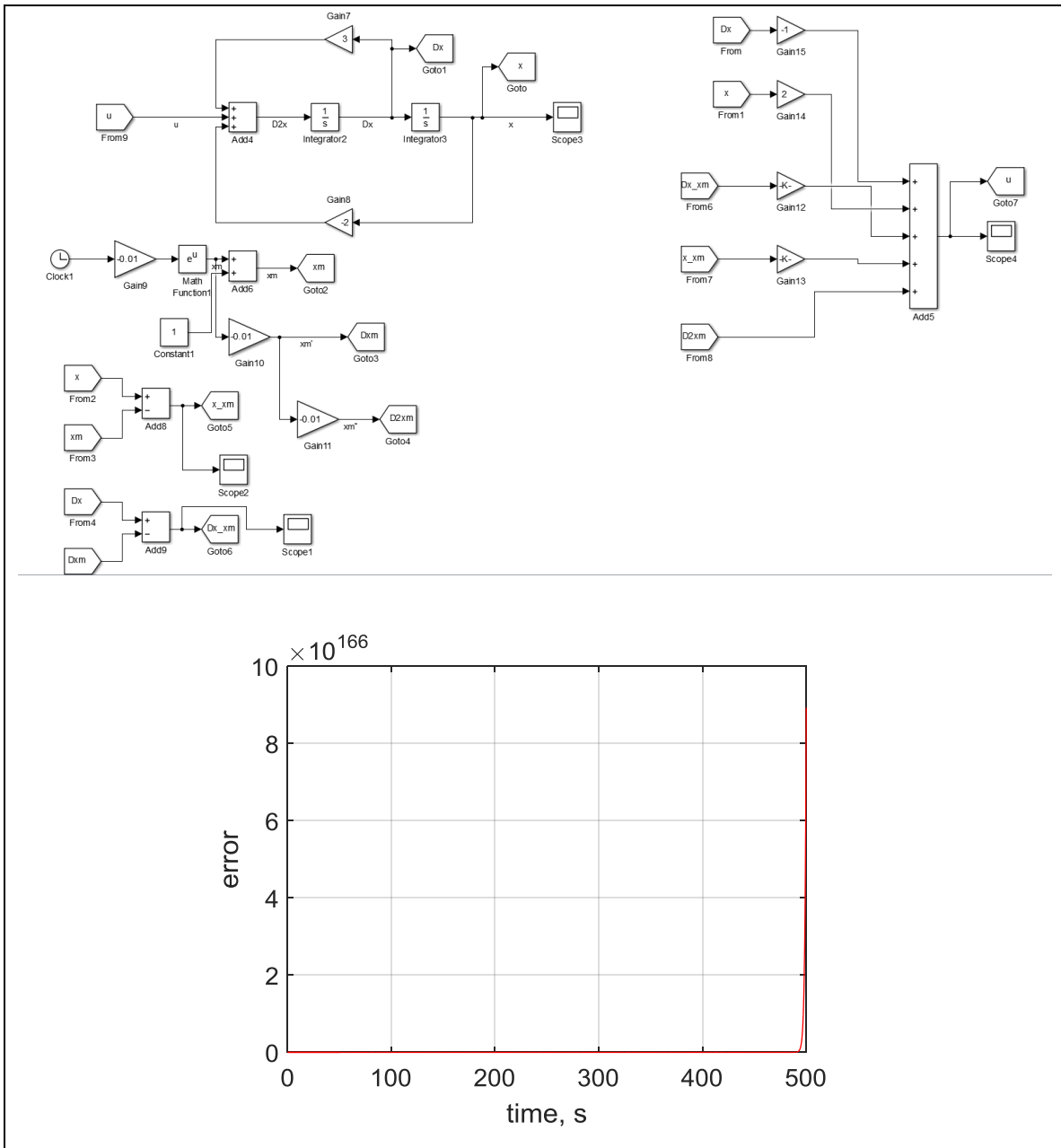
$$\begin{aligned} \ddot{x} &= f(x, \dot{x}, t) + g(x, \dot{x}, t)u \\ &= f(x, \dot{x}, t) + g(x, \dot{x}, t) \frac{1}{g(x, \dot{x}, t)} \left( -\dot{\tilde{x}} - \lambda\tilde{x} + \ddot{x}_d - \lambda\dot{\tilde{x}} - f(x, \dot{x}, t) \right) \\ &= -\dot{\tilde{x}} - \lambda\tilde{x} + \ddot{x}_d - \lambda\dot{\tilde{x}} \end{aligned}$$

$$\text{Hence, } \ddot{x} + \dot{\tilde{x}} + \lambda\tilde{x} - \ddot{x}_d + \lambda\dot{\tilde{x}} = 0 \Leftrightarrow \ddot{\tilde{x}} + \dot{\tilde{x}}(1 + \lambda) + \lambda\tilde{x} = 0$$

And hence we have a homogeneous ODE and with the appropriate choice of  $\lambda$  we can make sure that  $\tilde{x} \rightarrow 0$



But if there is an imperfection in the system then the convergence will not happen:



## 2.2 Nonsmooth control law

In order to make the controller more robust to parameter changes we impose a different condition on the Lyapunov function:

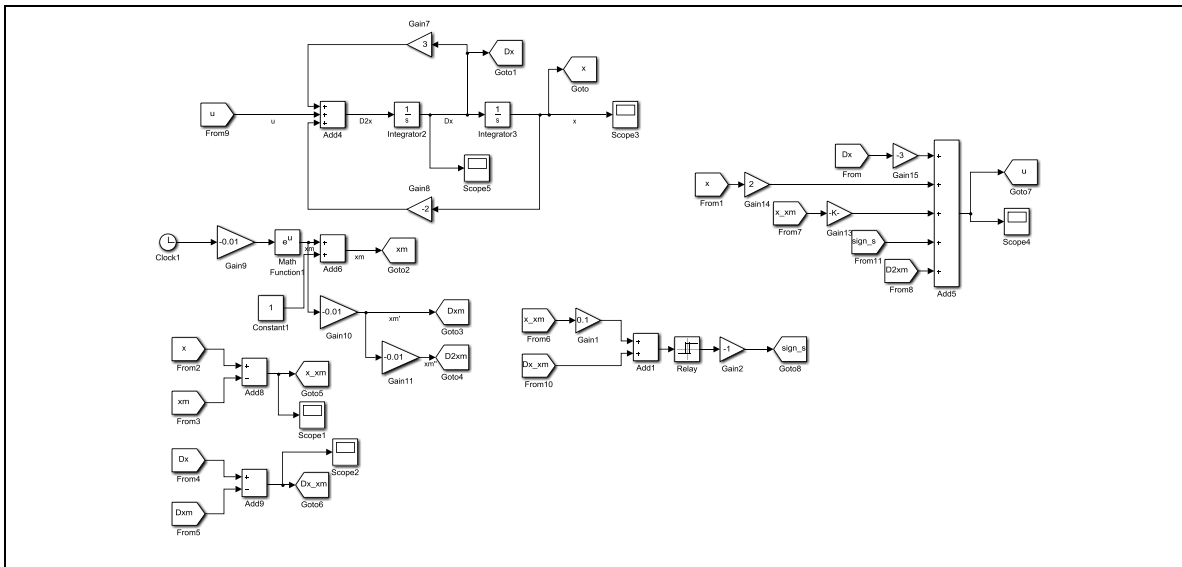
$$\dot{s} = -k \cdot \text{sign}(s)$$

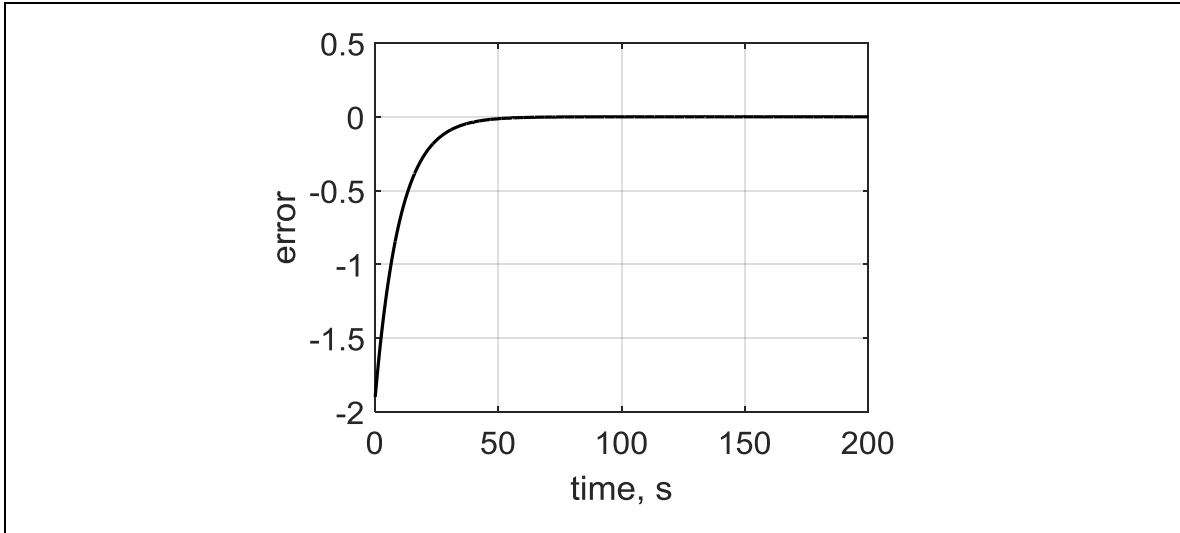
In this case  $\frac{dV(s)}{dt} = s\dot{s} = -s \cdot k \cdot \text{sign}(s) = -k \cdot |s|$

We know that:

$$\dot{s} = f(x, \dot{x}, t) + g(x, \dot{x}, t)u - \ddot{x}_d + \lambda \dot{\tilde{x}} = -k \cdot \text{sign}(s) \Leftrightarrow$$

$$u = \frac{1}{g(x, \dot{x}, t)} \left( -k \cdot \text{sign}(s) - f(x, \dot{x}, t) + \ddot{x}_d - \lambda \dot{\tilde{x}} \right)$$





### 2.3 Uncertainty

Assume that we have a system:

$$\ddot{x} = f(x, \dot{x}, t) + g(x, \dot{x}, t)u \quad (9)$$

with  $f$  and  $g$  being two **NOT “COMPLETELY”** known functions.

Our model based on an estimation of  $f$  and  $g$  is written as:

$$\ddot{x} = \hat{f}(x, \dot{x}, t) + \hat{g}(x, \dot{x}, t)u \quad (10)$$

And we only know that the difference between the real and the estimated functions is bounded:

$$|\hat{f} - f| \leq F, |\hat{g} - g| \leq G \quad (11)$$

To simplify the analysis we assume that  $g(x, \dot{x}, t) = \hat{g}(x, \dot{x}, t) = 1$

As before we can find when  $\dot{s} = 0$  but now

$$u = -\hat{f}(x, \dot{x}, t) + \ddot{x}_d - \lambda \dot{\tilde{x}} - k \cdot \text{sign}(s) \quad (12)$$

Hence,

$$\begin{aligned} \dot{s} &= f(x, \dot{x}, t) + u - \ddot{x}_d + \lambda \dot{\tilde{x}} = f(x, \dot{x}, t) + \left(-\hat{f}(x, \dot{x}, t) + \ddot{x}_d - \lambda \dot{\tilde{x}} - k \cdot \text{sign}(s)\right) - \ddot{x}_d + \lambda \dot{\tilde{x}} = \\ &= f(x, \dot{x}, t) - \hat{f}(x, \dot{x}, t) - k \cdot \text{sign}(s) \end{aligned}$$

$$\text{Or } \frac{dV(s)}{dt} = s \left( f(x, \dot{x}, t) - \hat{f}(x, \dot{x}, t) - k \cdot \text{sign}(s) \right)$$

So if we choose  $k > F$  we can be sure that  $\frac{dV(s)}{dt} \leq 0$

## 2.4 Finite Convergence

We have seen that  $\frac{dV(s)}{dt} = s\dot{s} \leq 0$ . We can impose a stricter condition to

ensure that  $s$  becomes zero in finite time:  $\frac{dV(s)}{dt} = s\dot{s} = -\eta|s| < 0, \eta > 0$

This is happening as if at some instant  $t=t_0$  we have  $s(t_0) > 0$ :

$$s\dot{s} = -\eta|s| \Leftrightarrow s\dot{s} = -\eta s \Leftrightarrow s = -\eta t + s(t_0)$$

which means that  $s$  will decrease until it becomes zero at

$$\eta t = s(t_0) \Leftrightarrow t = \frac{s(t_0)}{\eta}$$

Similarly if  $s(t_0) < 0$ :

$$s\dot{s} = -\eta|s| \Leftrightarrow s\dot{s} = \eta s \Leftrightarrow s = \eta t + s(t_0) \Leftrightarrow t = -\frac{s(t_0)}{\eta} > 0, s(t_0) < 0$$

Hence, we have

$$\frac{dV(s)}{dt} = s(f(x, \dot{x}, t) - \hat{f}(x, \dot{x}, t) - k \cdot \text{sign}(s))$$

And

$$\frac{dV(s)}{dt} \leq -\eta|s| \leq 0$$

Which is true if  $k = F + \eta$

### 3. Model Reference Adaptive Control

Assume the following system:

$$\ddot{x} = f(x, \dot{x}, t) + u \quad (13)$$

with

$$f(x, \dot{x}, t) = f_1(x, \dot{x}, t)p_1 + f_2(x, \dot{x}, t)p_2 + \dots \quad (14)$$

for example:

- $f(x, \dot{x}, t) = \dot{x}p_1 + xp_2$  (linear 2<sup>nd</sup> order ODE)
- $f(x, \dot{x}, t) = \dot{x}^2 p_1 + x|x||\dot{x}|p_2 + \cos(\sqrt{x})$  (nonlinear but time invariant)
- $f(x, \dot{x}, t) = \cos(t)\dot{x}\cos(x)p_1$  (nonlinear and time varying)

Now, we assume that  $f_1(x, \dot{x}, t), f_2(x, \dot{x}, t) \dots$  are known functions of the state vector, while the  $p_1, p_2 \dots$  are unknown **constants**.

Now as in the previous section let's choose a Lyapunov function such as:

$$V(s) = \frac{1}{2}s^2 \quad (15)$$

where  $s = \dot{\tilde{x}} + \lambda\tilde{x}$

and hence:  $\dot{s} = \ddot{\tilde{x}} + \lambda\dot{\tilde{x}} = \ddot{x} - \ddot{x}_d + \lambda\dot{\tilde{x}} = f(x, \dot{x}, t) + u - \ddot{x}_d + \lambda\dot{\tilde{x}}$

from (14) we have that



$$\dot{s} = u + (f_1(x, \dot{x}, t)p_1 + f_2(x, \dot{x}, t)p_2 + \dots) - \ddot{x}_d + \lambda \dot{\tilde{x}}$$

Now this can be written as:

$$\begin{aligned} \dot{s} &= u + (f_1(x, \dot{x}, t)p_1 + f_2(x, \dot{x}, t)p_2 + \dots - \ddot{x}_d + \lambda \dot{\tilde{x}}) \\ &= u + F \cdot p \end{aligned}$$

$$\text{So: } \dot{V}(s) = s\dot{s} = s(u + F \cdot p)$$

$$\text{with } F = [f_1(x, \dot{x}, t) \quad f_2(x, \dot{x}, t) \quad \dots \quad \ddot{x}_d \quad \dot{\tilde{x}}] \text{ \& } p = [p_1 \quad p_2 \quad \dots \quad -1 \quad \lambda]^T$$

Hence, if we knew the vector  $p$  we can choose:  $u = -F \cdot p - ks, k > 0$

which would have given:

$$\dot{V}(s) = -ks^2 < 0$$

Unfortunately we do not the vector  $p$  but we can have an estimate  $\hat{p}(t)$  and

hence we can define the parameter error  $\tilde{p}(t) = p - \hat{p}(t)$

So actually our control signal is:  $u = -F \cdot \hat{p} - ks$

$$\text{So } \dot{V}(s) = s\dot{s} = s(u + F \cdot p) = s(-F \cdot \hat{p} - ks + F \cdot p) = -ks^2 + sF \cdot \tilde{p} \quad (16)$$

Previously when we had only the tracking error we used as  $V(s) = \frac{1}{2}s^2$ , now

we also have the parameter error and hence we can use:

$$V(s) = \frac{1}{2}s^2 + \frac{1}{2}\tilde{p}_1^2 + \frac{1}{2}\tilde{p}_2^2 + \dots$$

As not all parameter errors are equally important:

$$V(s) = \frac{1}{2}s^2 + h_1 \frac{1}{2} \tilde{p}_1^2 + h_2 \frac{1}{2} \tilde{p}_2^2 + \dots$$

And in a matrix form<sup>2</sup>:

$$V(s) = \frac{1}{2}s^2 + \frac{1}{2} \tilde{p}^T H \tilde{p}$$

$$\text{Now, } \frac{d\left(\frac{1}{2} \tilde{p}^T H \tilde{p}\right)}{dt} = \frac{1}{2} \dot{\tilde{p}}^T H \tilde{p} + \frac{1}{2} \tilde{p}^T H \dot{\tilde{p}} = \dot{\tilde{p}}^T H \tilde{p} \quad (\text{or } \tilde{p}^T H \dot{\tilde{p}})$$

$$\text{Also, } \tilde{p}(t) = p - \hat{p}(t) \Rightarrow \dot{\tilde{p}}(t) = \dot{p} - \dot{\hat{p}}(t)$$

$$\text{So } \frac{d\left(\frac{1}{2} \tilde{p}^T H \tilde{p}\right)}{dt} = -\dot{\hat{p}}^T H \tilde{p}$$

Hence the time derivative of the chosen Lyapunov function is:

$$\dot{V}(s) = -ks^2 + sF \cdot \tilde{p} - \dot{\hat{p}}^T H \tilde{p}$$

$$\text{Hence, if we set } sF \cdot \tilde{p} = \dot{\hat{p}}^T H \tilde{p} \Leftrightarrow sF = \dot{\hat{p}}^T H \Leftrightarrow \dot{\hat{p}}^T = sFH^{-1}$$

$$\text{So the adaptation law is: } \dot{\hat{p}}^T = sFH^{-1} \quad (17)$$

$$\text{And the control law is: } u = -F \cdot \hat{p} - ks \quad (18)$$

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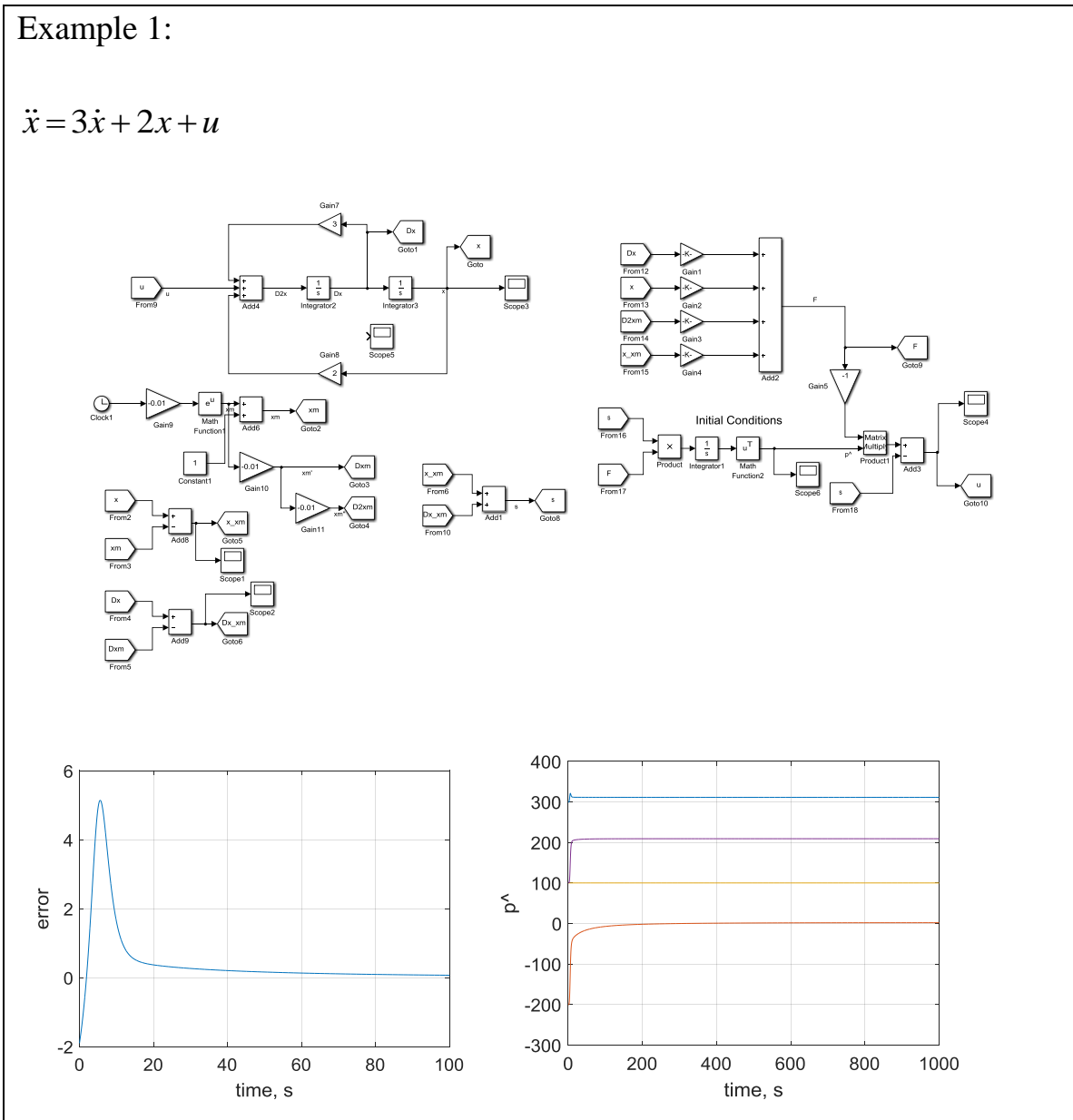
<sup>2</sup> In general the matrix H does not have to be diagonal, but only symmetric.

Another way to see this is to place (xx) into (xx):

$$\hat{p} = \left( \int sFH^{-1}dt \right)^T \Rightarrow u = \underbrace{-F \cdot \left( \int sFH^{-1}dt \right)^T}_{\text{Integral term}} - ks \quad (19)$$

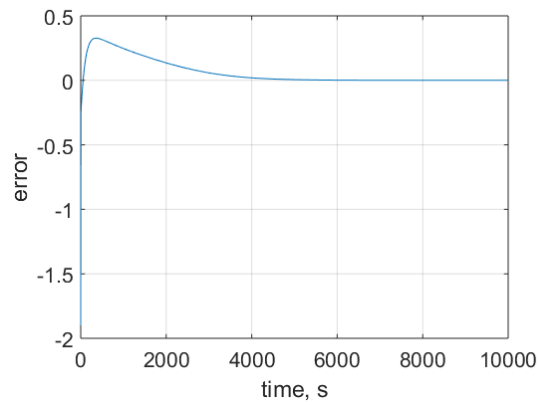
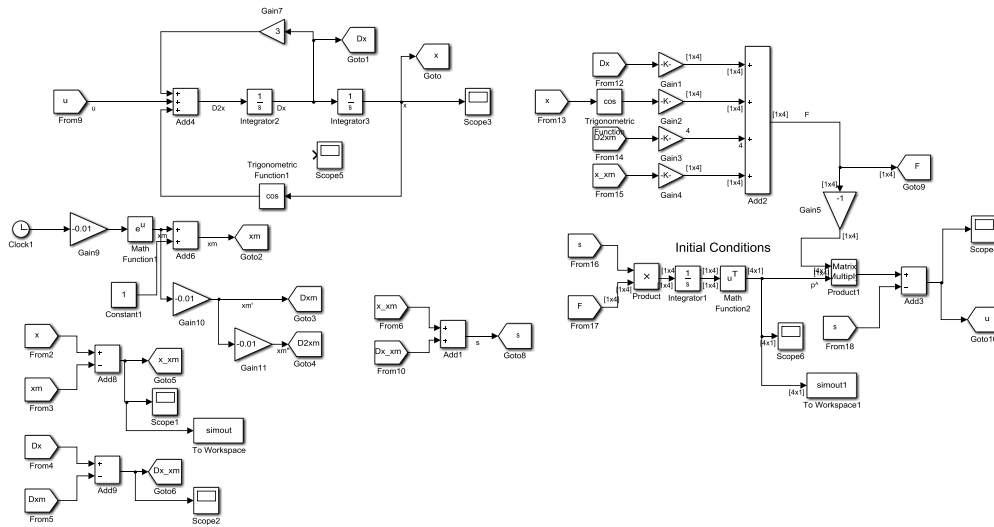
*PD term*

i.e. an adaptive PID controller.



Example 2:

$$\ddot{x} = 3\dot{x} + \cos(x) + u$$



## **Matlab Based Exercises for EEE8086**

Reproduce all Simulink files of chapter 5.