

What is adaptive control?

- Now, although the purpose of the class is to learn this, we can contemplate an example that illuminates some of the questions and hopefully some of the challenges.

Start w/a dynamical system to control,

$$\dot{x} = f(x, u)$$

↑ controller
↑ system state

where designed controller $u(t)$ leads to desired behavior in $x(t)$.

In reality, the system has some unknown or incorrectly estimated components/parameters

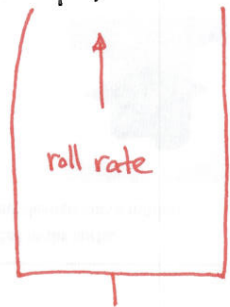
$$\dot{x} = f(x, u; \mu)$$

where $\mu \in \mathcal{P}$ and the true values are $\mu^* \in \mathcal{P}$ for $\mathcal{P} \subset \mathbb{R}^p$ compact.

- CAN CONTROL DESIGN USING μ STILL STABILIZE FOR $\mu \neq \mu^*$?
- HOW LARGE CAN DOMAIN \mathcal{P} BE?
- HOW QUICKLY CAN WE COMPENSATE FOR $\mu \neq \mu^*$?
- CAN WE STILL MEET PERFORMANCE SPECS?

Example: Uncertain Roll Dynamics

$$\dot{p} = L_p p + L_s \delta$$



aileron position
(our control variable)

unknown damping & aileron effectivenesses.

⇒

$$\dot{p} = L_s \left(\frac{L_p}{L_s} p + \delta \right)$$

$$\dot{p} = L_s (ap + \delta)$$

$$\sim \dot{x} = ax + u$$

Scenario: suppose we choose state feedback to stabilize p .

⇒

$$\delta(t) = -k p(t)$$

⇒

$$\dot{p} = L_s (a - k) p = -L_s (k - a) p$$

Q: What choice of k will stabilize the system?

A: $k > a$

Q: What if true value $a^* \in \mathcal{P} \equiv [a_{\min}^*, a_{\max}^*]$ is unknown?

A: $k > a_{\max}^*$

Q: Now, what if at best, conservative estimates of a_{\min}^* & a_{\max}^* existed, e.g., $\hat{a}_{\min} \leq \hat{a}_{\max}$ such that $[a_{\min}^*, a_{\max}^*] \subset [\hat{a}_{\min}, \hat{a}_{\max}]$?

A: $k > \hat{a}_{\max}$

PROBLEMATIC !!! Why?

1] Whole procedure based on an estimate.

- What if \hat{a}_{\max} estimate ~~is~~^{is} not large enough and it was that $a_{\max}^* > \hat{a}_{\max}$?

System could be unstable.

2] Procedure is conservative.

- What if a^* ~~was~~^{was} equal to a_{\min}^* 95% of the time?

Then too much control effort is being provided.

This is wasteful.

↳ mechanical wear
power issues (electric, fuel, etc.)
⋮

Adaptive control seeks to eliminate this problem while gauranteesting stability

↑ this is tricky part!

but, robust control seeks to handle uncertainty or unmodeled components while guaranteeing stability. What's the difference?

anyhow, back to the problem of handling $a \neq a^*$ in an adaptive fashion. Here are two similar solutions:

1 select the following controller and adaptive law

$$\delta(t) = \delta_{ad}(t) \equiv k(t)p(t)$$

where

$$\dot{k}(t) = -\gamma p^2(t), \quad \gamma > 0$$

↑
called: adaptation rate

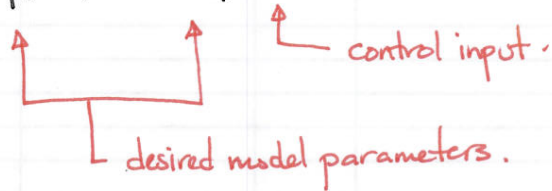
We get the following cleverly written closed-loop system:

$$\dot{p} = L_s a_m p + L_s (k(t) + a - a_m) p, \quad a_m < 0$$

- CAN SHOW STABILITY OF THIS SYSTEM.
- a_m IS CHOSEN TO MODEL DESIRED BEHAVIOR.
- SIGNALS BOUNDED IN THE PRESENCE OF UNCERTAINTIES.
- NOTICE THAT A NONLINEAR ADAPTIVE SYSTEM WAS OBTAINED FROM A LINEAR UNCERTAIN SYSTEM.

2 our second system begins with a desired model,

$$\dot{p}_m = L_{p,m} p_m + L_{s,m} \delta$$



(*)

→ we want the roll-rate tracking error,

$$e_p(t) \equiv p(t) - p_m(t)$$

to vanish, e.g., $e_p(t) \rightarrow 0$ in time

So, goal is to design an adaptive control law $\delta_{ad}(t)$ such that

$$\dot{p} = L_p p + L_s \delta_{ad}$$

tracks ~~the~~ (*) under the control input $\delta(t)$.

One such ~~ad~~ adaptive law is:

$$\delta_{ad}(t) = \hat{k}_p(t) p(t) + \hat{k}_s \delta(t)$$

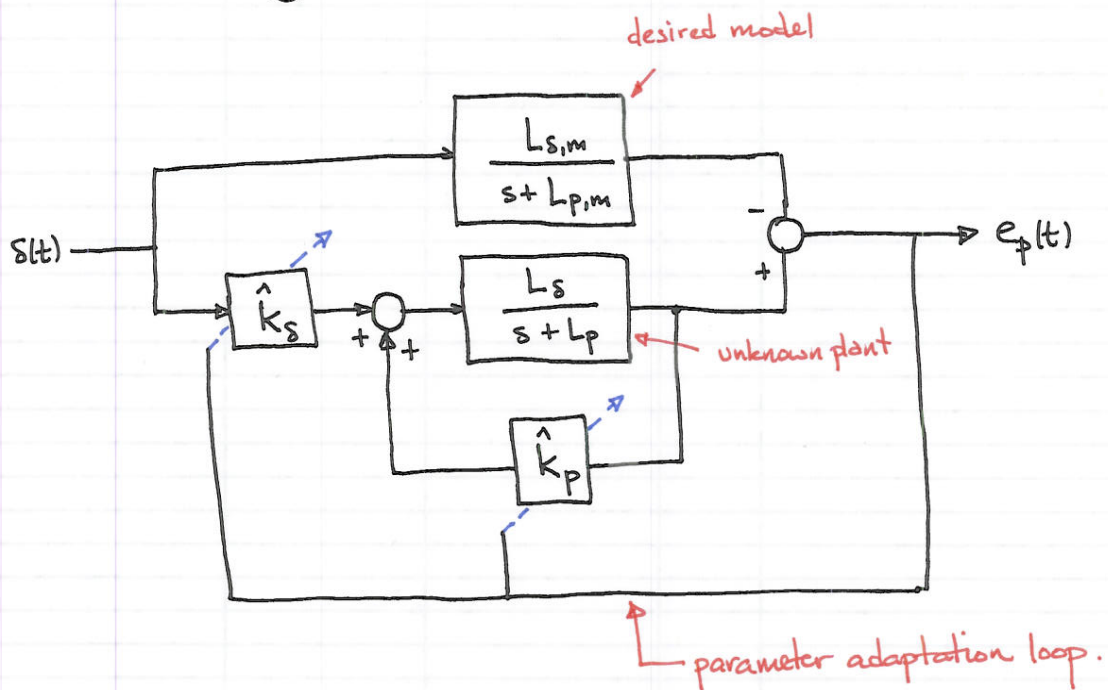
$$\dot{\hat{k}}_p(t) = -\gamma_p p(t) (p(t) - p_m(t))$$

$$\gamma_p > 0$$

$$\dot{\hat{k}}_s(t) = -\gamma_s \delta(t) (p(t) - p_m(t))$$

$$\gamma_s > 0$$

As a block diagram:



- YIELDS ASYMPTOTIC TRACKING WITH ALL REMAINING SIGNALS BOUNDED IN THE PRESENCE OF UNCERTAINTIES.