

Nonlinear Systems

Lecture 18

04/02/13

Last time

Ex 1

$$y = \psi^T \theta$$

↓
regressor

→ unknown constant parameters

Convergence of gradient-based parameter estimator

Ex 2

$$\dot{x}_1 = -\alpha x_1 + W^T(t) x_2$$

$$\dot{x}_2 = -W(t) x_1$$

$W(t)$: persistently exciting $\oplus \alpha > 0 \Rightarrow$ uniform asymptotic stability of the origin

Example Model Reference Adaptive Control (MRAC)

$$\dot{y} = a y + u \quad \dots (1)$$

$$\left. \begin{array}{l} u(\cdot) \quad \text{input signal} \\ y(\cdot) \quad \text{output signal} \end{array} \right\} \text{scalars} \quad \begin{array}{l} u(t) \in \mathbb{R} \\ y(t) \in \mathbb{R} \end{array}$$

a : constant but unknown parameter

Reference Model :

$$\dot{y}_m(t) = -a_m y_m(t) + r(t) \quad \dots \quad (2)$$

$$a_m > 0 \quad (\text{fixed})$$

$r(\cdot)$: reference signal

$$\begin{aligned} \text{If } a \text{ was known, } u(t) &= -ay(t) - a_m y(t) + r(t) \\ &= -(a + a_m) y(t) + r(t) \\ &= -K y(t) + r(t) \quad (3) \end{aligned}$$

would give us $(3) \rightarrow (1)$

$$\dot{y}(t) = -a_m y(t) + r(t) \quad (4)$$

Error variable : $e(t) = y(t) - y_m(t)$

$$(4) - (2) \Rightarrow$$

$$\dot{e}(t) = -a_m e(t) \Rightarrow e(t) = \exp(-a_m t) e(0)$$

Summary: Works well when a is known

Challenge: What to do when a is unknown

Idea: use,

$$u(t) = -\hat{K}(t)y(t) + r(t) \quad (5)$$

$\dot{\hat{K}}(t)$ to be determined

Write:

$$K = \hat{K}(t) + \tilde{K}(t)$$

\curvearrowright estimation error

$$\dot{\hat{K}}(t) = K - \tilde{K}(t) \quad (6)$$

(6) \rightarrow (5) : objective: find $\dot{\hat{K}} = -\tilde{K}$?

$$u(t) = -K y(t) + \tilde{K}(t)y(t) + r(t) \quad (7)$$

(7) \rightarrow (1)

$$\dot{y}(t) = -a_m y(t) + \tilde{K}(t)y(t) + r(t) \quad (8)$$

$$(8) - (2) \Rightarrow \dot{e}(t) = -a_m e(t) + y(t)\tilde{K}(t)$$

$\dot{\tilde{K}}(t) = ?$

We need to find $\tilde{K}(t)$ st. $\begin{bmatrix} e \\ \tilde{K} \end{bmatrix}$ system is UAS!

↓
Uniformly Asymptotically Stable

Propose a Lyapunov function:

$$V(e, \tilde{K}) = \frac{1}{2} \dot{e}^2(t) + \frac{1}{2} \tilde{K}^2(t)$$

$$\dot{V} = e \cdot \dot{e} + \tilde{K} \cdot \dot{\tilde{K}} = e[-a_m e + y \cdot \tilde{K}] + \tilde{K} \cdot \tilde{K}$$

$$= -a_m e^2 + \tilde{K} [\tilde{K} + e \cdot y]$$

choose $\tilde{K}(t) = -e(t)y(t)$

$$\Rightarrow \boxed{\dot{V}(e, \tilde{K}) = -a_m e^2} \quad \begin{array}{l} \text{if } * \text{ negative semi-definite} \\ (\dot{V} = -a_m e^2 - 0 \tilde{K}^2 \leq 0) \end{array}$$

\Rightarrow US

$$\dot{e} = -a_m e + y(t) \tilde{K}$$

$$\dot{\tilde{K}} = -y(t)e + 0 \cdot \tilde{K}$$

2)* If $r(t)$ is bounded $\Rightarrow e(t) \xrightarrow{t \rightarrow \infty} 0$ Why?

↳ Barbalat's lemma allows you to conclude this.

$$\left\{ \begin{array}{l} \dot{e} = -a_m e + y(t) \tilde{K} \\ \dot{\tilde{K}} = -y(t) e + \omega \tilde{K} \end{array} \right.$$

\Rightarrow

$$\begin{aligned} \dot{x}_1 &= -a_m x_1 - w(t) x_2 \\ \dot{x}_2 &= w(t) x_1 \end{aligned}$$

3) * $y(t)$: persistently exciting \Rightarrow UAS

Output of the system depends on $r(t)$

$$\begin{aligned} y(t) &= e(t) + y_m(t) \\ \dot{y}_m(t) &= -a_m y(t) + r(t) \end{aligned}$$

So based on this even though $r(t)$ was just considered to be bounded (no need to be rich enough) at the end

$$\tilde{K} \xrightarrow{t \rightarrow \infty} K$$

Summary :

$$u(t) = -\hat{K}(t)y(t) + r(t)$$

measured given reference

and ...

$$\hat{K}(t) = y(t)(y(t) - y_m(t))$$

Aside

$$\tilde{K}(t) = K - \hat{K}(t) \Rightarrow \tilde{K}(t) = -\hat{K}(t)$$

↓ const. unknown

therefore the control will be : (Implementation)

$$\begin{cases} u(t) = -\hat{K}(t)y(t) + r(t) \\ \hat{K}(t) = y(t)(y(t) - y_m(t)) \\ \dot{y}_m(t) = -a_m y_m(t) + r(t) \end{cases}$$

Dynamical controller

Backstepping —

Consider

$$\dot{x} = \underbrace{f(x)}_{\substack{\text{vector} \\ \text{field}}} + \underbrace{g(x) \cdot u}_{\substack{\text{vector} \\ \text{field}}} \quad \xrightarrow{\text{scalar control input}}$$

$$x(t) \in \mathbb{R}^n$$

Assume that there is

$$u = \alpha(x)$$

st. with $V(x)$: pos. definite radially unbounded

$$\frac{\partial V}{\partial x} [f(x) + g(x)\alpha(x)] \leq -W(x) < 0 \Rightarrow \text{GAS}$$

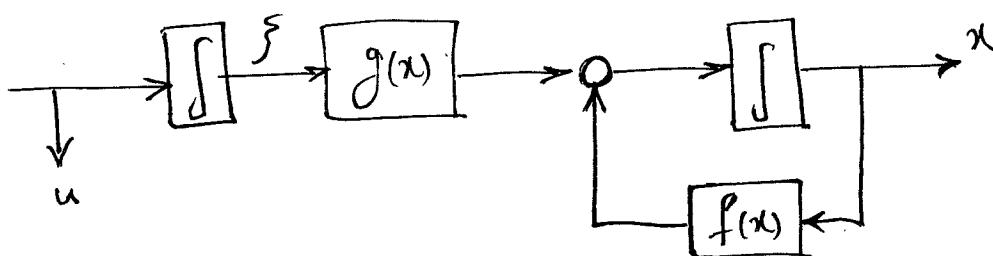
Now consider

$$\dot{x} = f(x) + g(x)\xi \quad (\text{I})$$

$$\dot{\xi} = u \quad (\text{II})$$

with $\xi = \alpha(x)$ satisfying the above conditions.

Q. Can we design $\xi = \beta(x, \xi)$ st. the origin
of (I), (II) is GAS?



ξ : not control

Think of it as a virtual control, penelige the difference between ξ and $\alpha(x)$ and figure out u so that we guarantee GAS of (I), (II)

Augment $V(x)$

$$\frac{\partial V}{\partial x} [f(x) + g(x)\alpha(x)] = -W(x) < 0$$

with $(\xi - \alpha(x))^2$

$$V_a(x, \xi) = V(x) + \frac{1}{2} (\xi - \alpha(x))^2$$

$$Z = \xi - \alpha(x)$$

$$\dot{z} = \dot{\xi} - \frac{\partial \alpha}{\partial x} \dot{x} = u - \frac{\partial \alpha}{\partial x} (f(x) + g(x)\xi)$$

$$V_a(x, z) = V(x) + \frac{1}{2} z^2$$

$$\begin{aligned} \dot{V}_a &= \dot{V} + z \dot{z} = \frac{\partial V}{\partial x} [f(x) + g(x) \underbrace{(\alpha(x) + Z)}_{\xi}] + \\ &\quad + Z \left[u - \frac{\partial \alpha}{\partial x} (f(x) + g(x)\xi) \right] \end{aligned}$$

$$\dot{V}_a = \underbrace{\frac{\partial V}{\partial x} [f(x) + g(x)\alpha(x)]}_{-W(x)} + z \left[u - \frac{\partial \alpha}{\partial x} [f(x) + g(x)\xi] + \frac{\partial V}{\partial x} g(x) \right]$$

+ ↓

Since we can measure z
there is no need to set this
~~to zero!~~

~~$-Kz^2$~~

Made possible by :

$$u = \frac{\partial \alpha}{\partial x} [f(x) + g(x)\xi] - \frac{\partial V}{\partial x} g(x) - Kz$$

$K > 0$