

## Lecture # 16

### Integrator Backstepping Thm.

Consider  $\begin{cases} \dot{\underline{\eta}} = \underline{f}(\underline{\eta}) + \underline{g}(\underline{\eta})\underline{\xi} & \dots (*) \\ \dot{\underline{\xi}} = u & \dots (** ) \end{cases}$   $\begin{cases} \underline{\eta} \in \mathbb{R}^n \\ \underline{\xi}, u \in \mathbb{R} \end{cases}$

Let  $\underline{\xi} = \phi(\underline{\eta})$  be a stabilizing state feedback control for  $(*)$  with  $\phi(0) = 0$ , & let  $V(\underline{\eta})$  be a Lyapunov f<sup>or</sup> this feedback control, i.e.,

$$\frac{\partial V}{\partial \underline{\eta}} (\underline{f}(\underline{\eta}) + \underline{g}(\underline{\eta}) \phi(\underline{\eta})) \leq - \underbrace{w(\underline{\eta})}_{\text{p.d.}} \quad \forall \underline{\eta} \in \mathcal{D}. \quad \left( \begin{array}{l} \text{If } \mathcal{D} \subseteq \mathbb{R}^n \\ \text{the c.A.S.} \end{array} \right)$$

Then the state feedback controller

$$u = \frac{\partial \phi}{\partial \underline{\eta}} (\underline{f}(\underline{\eta}) + \underline{g}(\underline{\eta})\underline{\xi}) - \frac{\partial V}{\partial \underline{\eta}} g(\underline{\eta}) - k(\underline{\xi} - \phi(\underline{\eta}))$$

makes the origin in  $\mathbb{R}^{n+1}$  A.S. with Lyap. f<sup>or</sup>

$$V_c(\underline{\eta}, \underline{\xi}) = V(\underline{\eta}) + \frac{1}{2}(\underline{\xi} - \phi(\underline{\eta}))^2.$$

Example : ( Integrator Backstepping)

$$\left\{ \begin{array}{l} \dot{x}_1 = x_1^2 - x_1^3 + u \\ \dot{x}_2 = u \end{array} \right. \quad g \equiv 1 \quad \begin{array}{l} x \in \mathbb{R}^2 \\ u \in \mathbb{R} \end{array}$$

Let

$$\begin{cases} \eta = x_1 \\ \xi = x_2 \end{cases}$$

Start with this subsystem

Design  $x_2 = \phi(x_1)$  such that  $x_1 = 0$  stable

Can do this by choosing

e.g.

$$x_2 = \phi(x_1) = -x_1^2 - x_1$$

killing unfriendly  
nonlinearity

This makes the first subsystem:

$$\dot{x}_1 = -x_1 - x_1^3$$

$$\& V(x_1) = \frac{1}{2} x_1^2 \Rightarrow \dot{V} = -x_1^2 - x_1^4 \leq -x_1^2 \quad \forall x_1 \in \mathbb{R}$$

$\therefore$  origin of the subsystem

$x_1 = -x_1 - x_1^3$  is GES (As needs LaSalle)

To backstep, let

$$z_2 := x_2 - \phi(x_1)$$

$$= x_2 + x_1 + x_1^2$$

to get new system:

$$\dot{x}_1 = -x_1 - x_1^3 + z_2$$

$$\dot{z}_2 = u + (1+2x_1)(-x_1 - x_1^3 + z_2)$$

Take  $V_c(x) = \frac{1}{2}x_1^2 + \frac{1}{2}z_2^2$

$$= \frac{1}{2}x_1^2 + \frac{1}{2}(x_2 + x_1 + x_1^2)^2$$

$$\Rightarrow \dot{V}_c = -x_1^2 - x_1^4 + z_2 \left\{ x_1 + (1+2x_1)(-x_1 - x_1^3 + z_2) + u \right\}$$

Take:

$$u = -x_1 - (1+2x_1)(-x_1 - x_1^3 + z_2) - z_2$$

$$\Rightarrow \dot{V}_c = -x_1^2 - x_1^4 - z_2^2 \Rightarrow \text{origin } \alpha(A) \text{ ES}$$

$$\begin{array}{l}
 \text{HW 5 P 2} \\
 \left. \begin{array}{l}
 \dot{x}_1 = x_1^2 - x_1^3 + x_2 \\
 \dot{x}_2 = x_3 \\
 \dot{x}_3 = u
 \end{array} \right\} \\
 \left. \begin{array}{l}
 x_3 = -x_1 - (1+2x_1) \\
 \quad \quad \quad (x_1^2 - x_1^3 + x_2) \\
 \quad \quad \quad - (x_2 + x_1 + x_1^2)
 \end{array} \right\} \\
 =: \phi(x_1, x_2)
 \end{array}$$

$$\& V(x_1, x_2)$$

$$\begin{array}{l}
 \text{we need to} \\
 \text{further backstep:}
 \end{array}
 \quad
 \begin{aligned}
 &= \frac{1}{2} x_1^2 + \frac{1}{2} (x_2 + x_1 + x_1^2)^2
 \end{aligned}$$

$$z_3 = x_3 - \phi(x_1, x_2)$$

$$\begin{array}{l}
 \text{Find } u = u(x_1, x_2, x_3) \\
 \text{and } V_c(x_1, x_2, x_3)
 \end{array}$$

Can work out same idea for slightly more general form:

$$\left\{ \begin{array}{l} \dot{\underline{\eta}} = \underline{f}(\underline{\eta}) + \underline{g}(\underline{\eta}) \underline{\xi} \\ \dot{\underline{\xi}} = f_a(\underline{\eta}, \underline{\xi}) + g_a(\underline{\eta}, \underline{\xi}) u \end{array} \right. \quad \left. \begin{array}{l} \underline{\xi} \in \mathbb{R}, \quad u \in \mathbb{R} \\ \underline{\eta} \in \mathbb{R}^n \\ \text{assumption: } g_a(\underline{\eta}, \underline{\xi}) \neq 0 \end{array} \right.$$

Can simply call "this whole thing" as new control  $u_a$ , i.e.,  
the input transformation  $u \mapsto u_a$  given by

$$u = \frac{1}{g_a(\underline{\eta}, \underline{\xi})} (u_a - f_a(\underline{\eta}, \underline{\xi}))$$

brings this system back to  $(*)-(**)$  form in the Thm.  
Can do backstepping on that to get:

$$\begin{aligned} u &= \phi_c(\underline{\eta}, \underline{\xi}) \sim \text{stabilizing controller} \\ &= \frac{1}{g_a(\underline{\eta}, \underline{\xi})} \left\{ \frac{\partial \phi}{\partial \underline{\eta}} (\underline{f}(\underline{\eta}) + \underline{g}(\underline{\eta}) \underline{\xi}) - \frac{\partial V}{\partial \underline{\eta}} \underline{g}(\underline{\eta}) \right. \\ &\quad \left. - K(\underline{\xi} - \phi(\underline{\eta})) - f_a(\underline{\eta}, \underline{\xi}) \right\} \end{aligned}$$

$$V_c(\underline{\eta}, \underline{\xi}) = V(\underline{\eta}) + \frac{1}{2} (\underline{\xi} - \phi(\underline{\eta}))^2$$

(overall Lyap. fn)

We can generalize further: Can apply backstepping recursively for any nonlinear system of the form:

$$\dot{\underline{x}} = \underline{f}_0(\underline{x}) + \underline{g}_0(\underline{x}) z_1$$

overall state vector  
 $\begin{pmatrix} \underline{x} \\ z \end{pmatrix} \in \mathbb{R}^{n+k}$

$$\dot{z}_1 = f_1(\underline{x}, z_1) + g_1(\underline{x}, z_1) z_2$$

$$\dot{z}_2 = f_2(\underline{x}, z_1, z_2) + g_2(\underline{x}, z_1, z_2) z_3$$

⋮

$$\dot{z}_{k-1} = f_{k-1}(\underline{x}, z_1, \dots, z_{k-1}) + g_{k-1}(\underline{x}, z_1, \dots, z_{k-1}) z_k$$

$$\dot{z}_k = f_k(\underline{x}, z_1, \dots, z_k) + g_k(\underline{x}, z_1, \dots, z_k) u$$

Here

$$\underline{x} \in \mathbb{R}^n, \quad z_i \in \mathbb{R} \quad \forall i = 1, \dots, k, \quad \underline{f}_0 : \mathbb{R}^n \mapsto \mathbb{R}^n$$

$$\underline{g}_0 : \mathbb{R}^n \mapsto \mathbb{R}^n$$

$$f_i(0) = 0 \quad \forall i = 0, 1, \dots, k$$

Assumption is that  $g_i(\underline{x}, z_1, \dots, z_j) \neq 0 \quad \forall 1 \leq j \leq k$

Same strategy: start from top-most eq<sup>n</sup> & then back-step.

Block BackStepping : (Even further generalization)

$$\dot{\underline{\eta}} = \underbrace{f(\underline{\eta})}_{n \times 1} + \underbrace{G(\underline{\eta}) \underline{\xi}}_{n \times m} \quad \dots \quad (1)$$

$$\dot{\underline{\xi}} = \underbrace{f_a(\underline{\eta}, \underline{\xi})}_{-} + G_a(\underline{\eta}, \underline{\xi}) u \quad \dots \quad (2)$$
$$\underline{\eta} \in \mathbb{R}^n, \quad \underline{\xi} \in \mathbb{R}^m, \quad u \in \mathbb{R}^m, \quad m > 1$$

Suppose (1) can be stabilized by

Here overall state vector  
 $(\begin{smallmatrix} \underline{\eta} \\ \underline{\xi} \end{smallmatrix}) \in \mathbb{R}^{n+m}$

$$\underline{\xi} = \phi(\underline{\eta}) \text{ with } \phi(0) = 0$$

& let  $V(\underline{\eta})$  be the Lyap. f<sup>u</sup> for that subsystem.

$$f_u \in \mathbb{R}^m$$

$$G_a \in \mathbb{R}^{n \times m},$$

Assumption is that  
 $G_a$  is non-singular

The overall stabilizing state feedback is

$$\frac{u}{m \times 1} = \underbrace{\underline{G_a}^{-1}_{m \times m}}_{m \times n} \left[ \underbrace{\frac{\partial \phi}{\partial \eta}}_{n \times n}^{\text{mat}} \underbrace{\left( \underline{f}_{n \times 1} + \underline{G}_{n \times m} \underline{\xi}_{m \times 1} \right)}_{n \times 1} \right]$$

$$V_c : \mathbb{R}^{m+n} \mapsto \mathbb{R}_+$$

$$= \underbrace{\underline{G}^T_{m \times n} \frac{\partial V}{\partial \eta}_{n \times 1}}_{m \times 1} - \frac{\underline{f}_a}{m \times 1} - K \left( \underline{\xi}_{m \times 1} - \underline{\phi}_{m \times 1} \right)$$

for some  $K > 0$

&  $\underbrace{V_c(\underline{\xi}, \underline{\eta})}_{\text{overall}} = V(\underline{\eta}) + \frac{1}{2} (\underline{\xi} - \underline{\phi}(\underline{\eta}))^\top (\underline{\xi} - \underline{\phi}(\underline{\eta}))$

Lyap - f<sup>-1</sup>

# Geometric Control Theory

## Design Idea #5

### Feedback linearization

Motivation: SISO system

(will later generalize to MIMO)

$$\begin{cases} \dot{\underline{x}} = f(\underline{x}) + g(\underline{x}) u \\ y = h(\underline{x}) \end{cases} \quad \begin{array}{l} \underline{x} \in \mathbb{R}^n \\ u, y \in \mathbb{R} \end{array}$$

(SISO\_original)

Question: Does there exist change-of-variable

$$\underline{\zeta} := \begin{bmatrix} \eta \\ \underline{x} \end{bmatrix} = \underline{\zeta}(\underline{x}), \quad \text{where } \underline{\zeta}(\cdot) \text{ is a diffeomorphism}$$

and a control

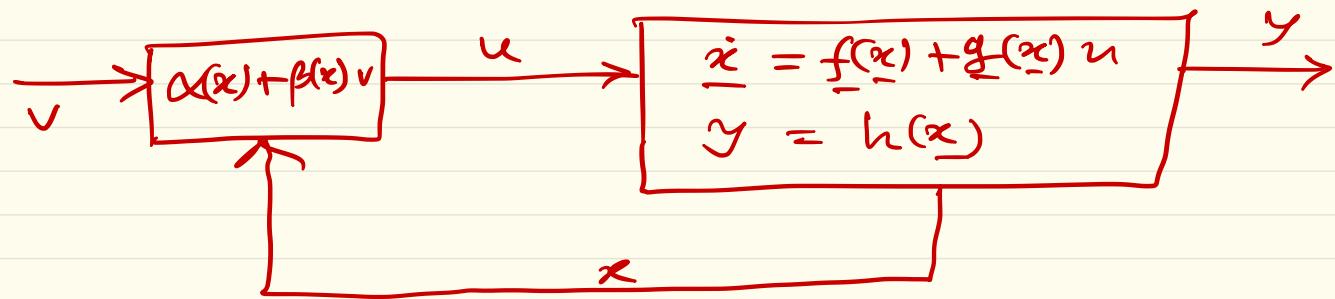
$$u = \alpha(\underline{x}) + \beta(\underline{x}) v$$

$\underline{\zeta}$  is  $C^1$   
 $\underline{\zeta}^{-1}$  is also  $C^1$

such that  $(\gamma(\cdot), \alpha(\cdot), \beta(\cdot))$  transforms  
(SISO-original) into

either (Form 2)

$$\dot{\underline{z}} = A \underline{z} + B v$$



or (Form 2)

$$\dot{\xi} = f_0(\eta, \xi)$$

$$\dot{\xi} = A\xi + Bv$$

$$y = C\xi$$

If the ANSWER is Yes for Form 1,  
then we say the original nonlinear  
System (SISO-original) is

" feedback linearizable "

If Yes for Form 2, we say  
(SISO-original) is "Input - Output Linearizable"  
(State)  
Feedback linearization :

$$\underline{z} := \begin{bmatrix} n \\ \underline{x} \end{bmatrix} = \tau(\underline{x})$$

Static ( $\tau(x)$ )

$$u = \alpha(x) + \beta(x)v$$

Dynamic ( $\tau(x, \beta, \gamma, \delta)$ )

$$\underline{z} := \begin{bmatrix} n \\ \underline{x} \end{bmatrix} = \tau(\underline{x})$$

$$\dot{\underline{x}}_c = \gamma(\underline{x}, \underline{x}_c) + \delta(\underline{x}, \underline{x}_c)v$$

$$u = \alpha(\underline{x}, \underline{x}_c) + \beta(\underline{x}, \underline{x}_c)v$$

Example : Pendulum in air

$$x_1 = \theta$$

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

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

$$\dot{x}_2 = -\alpha \sin x_1 - \beta x_2 + u \quad \left. \right\} \alpha, \beta > 0$$

"Cancel the nonlinearity & get stable linear closed-loop system" approach:

$$u = \underbrace{+\alpha \sin x_1}_{\alpha(x)} + \underbrace{1 \cdot v}_{\beta(x)} = \alpha(x) + \beta(x) v$$

Then we get:

$$\begin{aligned} \dot{x}_1 &= x_2 \\ \dot{x}_2 &= -\beta x_2 + v \end{aligned} \quad \left. \right\} \text{LT) control sys.}$$

Car design

$$v = -k_1 x_1 - k_2 x_2 \Rightarrow \begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{bmatrix} 0 & 1 \\ -k_1 & -k_2 - \beta \end{bmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Choose  $k_1, k_2$  s.t. Humanize even if  $\beta = 0$

Cannot always  
kill nonlinearities

Example

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

$$\dot{x}_2 = -x_1^2 + u$$

Cannot choose "u" to cancel " $a \sin x_2$ ".

But consider

$$\begin{pmatrix} z_1 \\ z_2 \end{pmatrix} = \tilde{\gamma}(\underline{x}) = \begin{Bmatrix} x_1 \\ a \sin x_2 \end{Bmatrix} = \begin{Bmatrix} x_1 \\ \dot{x}_1 \end{Bmatrix}$$

Then dynamics in new variables:

$$\dot{z}_1 = z_2$$

$$\dot{z}_2 = a(\cos x_2) \dot{x}_1$$

$$= a(\cos x_2) (-x_1^2 + u)$$

Now can kill nonlinearity by control

$$u = \frac{x_1^2}{\alpha(x)} + \frac{1}{a \cos x_2}$$

$v$  (well defined for  
 $-\pi/2 < x_2 < +\pi/2$ )

$$\begin{cases} \dot{z}_1 = z_2 \\ \dot{z}_2 = v \end{cases}$$

This is  
feedback  
linearized