

## INPUT-OUTPUT LINEARIZATION OF SISO Systems

Example ①

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

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

$$y = x_1$$

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

$$\ddot{y} = \ddot{x}_1 = a \cos x_2 \quad \dot{x}_2 = -2x_1 \cos x_2 + a \cos x_2 u$$

$$u = -(\cos x_2)^{-1} [a x_1^2 \cos x_2 + u]$$

$$\Rightarrow \ddot{y} = u, \text{ Take } u = \ddot{y}_d - k_1(y - \dot{y}_d) - k_2(y - y_d)$$

$$(\ddot{y} - \ddot{y}_d) + k_1(y - \dot{y}_d) + k_2(y - y_d) = 0$$

$$\Rightarrow y \rightarrow y_d$$

$$\text{if } y_d \equiv 0, \text{ then } y \rightarrow 0$$

Example ②

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

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

$$y = x_2$$

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

$$\text{Take } u = x_1^2 + \dot{y}_d - k(y - y_d)$$

$$\Rightarrow \dot{y} = \dot{y}_d - k(y - y_d) \Rightarrow (y - y_d) \rightarrow 0$$

Notice  $x_1$  dynamics if  $y_d \equiv 0 \Rightarrow y \rightarrow 0$   
(e.g. for regulation control, i.e.  $y_d \equiv 0$ ), then

$y \rightarrow 0$  (i.e.  $x_2 \rightarrow 0$ ) independent of  $x_1$

then  $a \sin x_2 \rightarrow 0$

i.e.  $\dot{x}_1 \rightarrow 0$

$$\text{or precisely } x_2 = x_{2(0)} e^{-kt}$$

$$\therefore \dot{x}_1 = a \sin(x_{2(0)} e^{-kt})$$

$$x_1 = \int_0^t a \sin(x_{2(0)} e^{-kr}) dr$$

example ③

$$\dot{x}_1 = 2x_1$$

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

$$y = x_2$$

from example ②,  $u = x_1^2 - ky$ , so  $y \rightarrow 0$

but  $x_1 \rightarrow \infty$

example ④

$$\begin{aligned} \dot{x}_1 &= 2x_1 & y &= x_1 \\ \dot{x}_2 &= -x_1^2 + u & \dot{y} &= \dot{x}_1 = 2x_1, \ddot{y} = 2\dot{x}_1 = 4x_1, \dots \text{etc.} \end{aligned}$$

Systan

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

$$y = h(x)$$

$x \in R^n$ ,  $u$  and  $y$  scalars

$$\begin{aligned}\dot{x}_1 &= \sin x_2 + (x_2+1)x_3 \\ \dot{x}_2 &= x_1^5 + x_3 \\ \dot{x}_3 &= x_1^2 + u\end{aligned}\quad \left. \begin{array}{l} \\ \\ \end{array} \right\} n=3$$

$$y = x_1$$

$$\dot{y} = \dot{x}_1 = \sin x_2 + (x_2+1)x_3$$

$$\ddot{y} = f_1(x) + (x_2+1)u \quad \text{--- (1)}$$

where  $f_1(x) = (x_1^5 + x_3)(x_3 + \cos x_2) + (x_2+1)x_1^2$

$$u = \frac{1}{x_2+1}(v - f_1)$$

$$\Rightarrow \ddot{y} = v$$

$$v = \ddot{y}_d - k_2(y - y_d) - k_1(y - y_d)$$

$$\Rightarrow (y - y_d) \rightarrow 0$$

- control law defined everywhere except at  $x_2 = -1$
- full state feedback
- relative degree = 2 (2 differentiations) ref. degree =  $n$

(if you differentiate  $y$   $n$  times and still don't get  $u$ , then system is not controllable)

$\therefore$  ref. degree  $< n$ ,  $\therefore$  one state is unobservable by (1)

we can use  $e, \dot{e}$  and  $x_3$  as the three states ( $e = y - y_d$ )  
internal dynamics (dynamics of the unobservable state)

$$\dot{x}_2 = x_1^2 + \frac{1}{x_2+1}(\ddot{y}_d - k_1 e - k_2 \dot{e} + f_1)$$

we want stable internal dynamics

Example

$$\dot{x}_1 = x_2^3 + u \quad \text{--- (1)}$$

$$\dot{x}_2 = u \quad \text{--- (2)}$$

$$y = x_1$$

$$\dot{y} = \dot{x}_1 = x_2^3 + u$$

$$\therefore \boxed{u = -x_2^3 - e(t) + y_d(t)} \quad \text{--- (3)}$$

where  $e(t) = y_d(t) - y(t)$

makes  $x_1 \rightarrow 0$  by obtaining  $\dot{e} + e = 0$

However, dynamics of  $x_2$  are : (using (3) in (2))

$$\dot{x}_2 = -x_2^3 - e(t) + y_d(t)$$

Since  $e(t) \rightarrow 0$  exponentially and  $y_d$  is bounded, we have

$$|y_d(t) - e(t)| < d \quad (\text{a positive number})$$

$$\therefore \dot{x}_2 + x_2^3 = y_d - e \Rightarrow$$

if  $x_2 > d^{1/3}$ , then  $\dot{x}_2 < 0$

if  $x_2 < -d^{1/3}$  then  $\dot{x}_2 > 0$

$$\therefore |x_2(t)| \leq d^{1/3}$$

$\therefore$  internal dynamics keep  $x_2(t)$  bounded.

In general, difficult to analyze internal dynamics.

Look at  $\begin{cases} \dot{x}_1 = x_2 + u \\ \dot{x}_2 = u \\ y = x_1 \end{cases}$

$$\frac{Y(s)}{U(s)} = \frac{s+1}{s^2}$$

stable internal dynamics.

Look at  $\begin{cases} \dot{x}_1 = x_2 + u \\ \dot{x}_2 = -u \\ y = x_1 \end{cases}$

$$\frac{Y(s)}{U(s)} = \frac{s-1}{s^2}$$

unstable internal dynamics

$\therefore$  for linear systems internal dynamics related to zeros.

Zero Dynamics; is the internal dynamics of the system when the system output is kept at zero by the input.

$$\text{eg for } \dot{x}_1 = x_2^3 + u$$

$$\dot{x}_2 = u$$

$$y = x_1$$

$$\therefore \text{here } u = -x_2^3$$

internal dynamics become

$$\dot{x}_2 + x_2^3 = 0$$

(stable using  $v = x_2^2$ )

Result: For nonlinear systems for tracking problem, local exponential stability of the zero dynamics guarantees stable internal dynamics if  $y_d, \dots, y_d^{(n-1)}$  have small magnitudes.

For regulation problems, local asymptotic stability of zero dynamics  $\Rightarrow$  local asymptotic stability of internal dynamics.

### INPUT / OUTPUT LINEARIZATION

$$\dot{x} = f + g u \quad (\text{non linear})$$

$$y = h(x)$$

$$\dot{y} = \nabla h \cdot \dot{x} = \nabla h(f + g u) = L_f h(x) + L_g h(x)u$$

if  $L_g h(x) \neq 0$ , then  $u = \frac{1}{L_g h}(-L_f h + v)$  gives control,

otherwise

$$\ddot{y} = L_f^2 h(x) + L_g L_f h(x)u$$

if  $L_g L_f h(x)$  is again zero, then keep differentiating till we find  $r$  ( $< n$ ) s.t

$r$  : relative degree.

$$L_g L_f^{r-1} h(x) \neq 0$$

$$\text{then take } u = \frac{1}{L_g L_f^{r-1} h}(-L_f^r h + v)$$

$$\text{giving } y^{(r)} = v$$

## NORMAL FORM

$$\dot{x} = b + gu$$

$$y = h(x)$$

with relative degree  $r \leq n$  can be written as

$$\frac{d}{dt} \begin{bmatrix} \xi_1 \\ \vdots \\ \xi_{r-1} \\ \xi_r \end{bmatrix} = \begin{bmatrix} \xi_2 \\ \vdots \\ \xi_r \\ a(\xi, \eta) + b(\xi, \eta) u \end{bmatrix}$$

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

$$y = \xi_1$$

$$\text{where } \xi = [h \ L_f h \ \dots \ L_f^{r-1} h]^T$$

$$\eta = [\eta_1 \ \eta_2 \ \dots \ \eta_{n-r}]^T$$

example  $\dot{x} = \begin{bmatrix} -x_1 \\ x_1 x_2 \\ x_2 \end{bmatrix} + \begin{bmatrix} e^{x_2} \\ 1 \\ 0 \end{bmatrix} u$

$$y = h(x) = x_3$$

$$\dot{y} = x_2$$

$$\ddot{y} = \dot{x}_2 = x_1 x_2 + u$$

$$\therefore \xi_1 = x_3 = h(x)$$

$$\xi_2 = L_f h(x) = x_2$$

$\eta(x)$  should satisfy

$$L_f \eta = \frac{\partial \eta}{\partial x_1} e^{x_2} + \frac{\partial \eta}{\partial x_2} = 0$$

$$\text{e.g. } \eta(x) = 1 + x_1 - e^{x_2}$$

$$\therefore z = (\xi_1, \xi_2, \eta)^T$$

MIMO Sysns:  $y_i = L_f h_j + \sum_{i=1}^m (L_g \cdot h_i) u_i$

(take each output + differentiate successively to get input)

we can set

$$\begin{bmatrix} y_1^{(k_1)} \\ \vdots \\ y_m^{(k_m)} \end{bmatrix} = \begin{bmatrix} L_f^{k_1} h_1(x) \\ \vdots \\ L_f^{k_m} h_m(x) \end{bmatrix} + E(x) \begin{bmatrix} u_1 \\ \vdots \\ u_m \end{bmatrix}$$

decoupling matrix