

### Linearization problem

Consider a non-linear time-invariant system of the form

$$\begin{cases} \dot{x}(t) = f(x(t), u(t)), \\ y(t) = g(x(t), u(t)) \end{cases} \quad (1)$$

such that  $x \in \mathbb{R}^n$ ,  $u \in \mathbb{R}^m$ ,  $y \in \mathbb{R}^p$ , and

Slide 1

A:  $\frac{\partial f(x,u)}{\partial x}$ ,  $\frac{\partial f(x,u)}{\partial u}$ ,  $\frac{\partial g(x,u)}{\partial x}$ , and  $\frac{\partial g(x,u)}{\partial u}$  exist and are continuous.

The question is:

✗ How to approximate system (1) by a linear model?

### The simplest example

Consider the static time-invariant system

$$y(t) = ku(t) + \eta. \quad (2)$$

This system is nonlinear, but intuitively it seems to be “almost linear.” Indeed, this system can easily be made linear. To this end notice that for any *constant*  $u(t) = u_e$  the output  $y(t)$  is also constant and given by  $y_e \doteq ku_e + \eta$ . Let's call the pair  $(y_e, u_e)$  the *equilibrium* point.

Slide 2

Define now new signals:

$$y_\delta(t) \doteq y(t) - y_e \quad \text{and} \quad u_\delta(t) \doteq u(t) - u_e,$$

which are deviations of  $y$  and  $u$  from the equilibrium point. Then

$$y_\delta(t) = ku_\delta(t), \quad (2_\delta)$$

which is *linear*. Thus, nonlinear system (2) can be made linear by an appropriate (cf. equilibrium) shift of its input and output signals.

### A little bit more complicated example

Consider now the system

$$\begin{cases} \dot{x}(t) = -ax(t) + bu(t) + \eta_x, \\ y(t) = x(t) + \eta_y. \end{cases} \quad (3)$$

Assume again that the input  $u(t) = u_e$  is constant. Then

$$x(t) = \int_0^t e^{-a(t-s)}(bu_e + \eta_x)dt = \frac{1}{a}(1 - e^{-at})(bu_e + \eta_x).$$

As  $\lim_{t \rightarrow \infty} e^{-at} = 0$ , system (3) approaches its equilibrium point<sup>a</sup>  $(x_e, u_e)$ , where

$$x_e = \frac{1}{a}(bu_e + \eta_x).$$

---

<sup>a</sup>This is called the *steady-state*.

Slide 3

### A little bit more complicated example (contd)

Let's write now system (3) in terms of the deviations

$$x_\delta(t) \doteq x(t) - x_e \quad \text{and} \quad u_\delta(t) \doteq u(t) - u_e.$$

We get:

$$\begin{cases} \dot{x}_\delta(t) = -ax_\delta(t) + bu_\delta(t), \\ y_\delta(t) = x_\delta(t), \end{cases} \quad (3_\delta)$$

where  $y_\delta(t) \doteq y(t) - x_e - \eta_y$ . Equation  $(3_\delta)$  is thus a linear equivalent of the nonlinear system (3).

Slide 4

## Equilibrium

The pair  $(x_e, u_e)$  is called *equilibrium* if

$$f(x_e, u_e) = 0.$$

Let  $(x_e, u_e)$  be an equilibrium of (1). Then  $x(t)$ ,  $y(t)$ , and  $u(t)$  can be presented as follows:

$$x(t) = x_e + x_\delta(t), \quad y(t) = y_e + y_\delta(t), \quad \text{and} \quad u(t) = u_e + u_\delta(t),$$

where  $y_e \doteq g(x_e, u_e)$ .

Eqn. (1) can then be rewritten as follows:

$$\begin{cases} \dot{x}_\delta(t) = f(x_e + x_\delta(t), u_e + u_\delta(t)), \\ y_\delta(t) = g(x_e + x_\delta(t), u_e + u_\delta(t)) - y_e \end{cases} \quad (1_\delta)$$

Slide 5

## Tailor series expansion

One can expand  $f(\cdot)$  and  $g(\cdot)$  as follows<sup>a</sup>:

$$\begin{aligned} f(x, u) &= f(x_e, u_e) + \left. \frac{\partial f}{\partial x} \right|_{x_e, u_e} x_\delta + \left. \frac{\partial f}{\partial u} \right|_{x_e, u_e} u_\delta + o(x_\delta, u_\delta), \\ g(x, u) &= g(x_e, u_e) + \left. \frac{\partial g}{\partial x} \right|_{x_e, u_e} x_\delta + \left. \frac{\partial g}{\partial u} \right|_{x_e, u_e} u_\delta + o(x_\delta, u_\delta), \end{aligned}$$

where  $o(\eta)$  is any function of a variable  $\eta$  such that  $\lim_{\|\eta\| \rightarrow 0} \frac{\|o(\eta)\|}{\|\eta\|} = 0$ .

Slide 6

<sup>a</sup>Given an  $n$ -dimensional function  $\phi(\eta)$  of an  $m$ -dimensional argument  $\eta$ , the partial derivative of  $\phi$  with respect to  $\eta$  (the *Jacobian matrix*) is defined as the following  $n \times m$  matrix:

$$\frac{\partial \phi(\eta)}{\partial \eta} \doteq \begin{bmatrix} \frac{\partial \phi_1(\eta)}{\partial \eta_1} & \cdots & \frac{\partial \phi_1(\eta)}{\partial \eta_m} \\ \vdots & & \vdots \\ \frac{\partial \phi_n(\eta)}{\partial \eta_1} & \cdots & \frac{\partial \phi_n(\eta)}{\partial \eta_m} \end{bmatrix}.$$

### Linearization around $(x_e, u_e)$

If the deviations  $x_\delta$  and  $u_\delta$  are “small enough,” then Eqn. (1 <sub>$\delta$</sub> ) can be approximated by the following *linear* model:

$$\begin{cases} \dot{x}_\delta(t) = A_\delta x_\delta(t) + B_\delta u_\delta(t), \\ y_\delta(t) = C_\delta x_\delta(t) + D_\delta u_\delta(t), \end{cases} \quad (1_{\text{linearized}})$$

where

$$A_\delta \doteq \left. \frac{\partial f}{\partial x} \right|_{x_e, u_e},$$

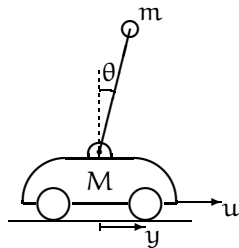
$$B_\delta \doteq \left. \frac{\partial f}{\partial u} \right|_{x_e, u_e},$$

$$C_\delta \doteq \left. \frac{\partial g}{\partial x} \right|_{x_e, u_e},$$

$$D_\delta \doteq \left. \frac{\partial g}{\partial u} \right|_{x_e, u_e}.$$

Slide 7

### Inverted pendulum



$M$ : mass of the cart  
 $m$ : mass of the pendulum  
 $l$ : length of the pendulum

$y$ : position of the cart

$\theta$ : angular rotation

$u$ : force on the cart

$\mu$ : friction coefficient

$g$ : acceleration of gravity

Nonlinear model:

$$\begin{cases} (M + m)\ddot{y} + ml\ddot{\theta}\cos\theta - ml(\dot{\theta})^2\sin\theta + \mu\dot{y} = u \\ l\ddot{\theta} - g\sin\theta + \ddot{y}\cos\theta = 0 \end{cases}$$

or, equivalently,

$$\begin{bmatrix} M + m & ml\cos\theta \\ \cos\theta & l \end{bmatrix} \begin{bmatrix} \ddot{y} \\ \ddot{\theta} \end{bmatrix} = \begin{bmatrix} ml(\dot{\theta})^2\sin\theta - \mu\dot{y} \\ g\sin\theta \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u.$$

Slide 8

### Inverted pendulum: the state-space realization

The state vector:

$$x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} y \\ \theta \\ \dot{y} \\ \dot{\theta} \end{bmatrix}$$

Slide 9

The state equation:

$$\frac{d}{dt} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \underbrace{\begin{bmatrix} x_3 \\ x_4 \\ [M+m \quad ml \cos x_2]^{-1} \left( \begin{bmatrix} mlx_4^2 \sin x_2 - \mu x_3 \\ g \sin x_2 \end{bmatrix} + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u \right) \end{bmatrix}}_{f(x,u)}. \quad (4)$$

where  $\Lambda(x_2) \doteq \begin{bmatrix} M+m & ml \cos x_2 \\ \cos x_2 & l \end{bmatrix}$  is nonsingular ( $\det \Lambda(x_2) = l(M + m \sin^2 x_2)$ ).

### Inverted pendulum: equilibriums

Equilibrium points should correspond to  $\dot{x} \equiv 0$ . Then, from Eq. (4) one can get that  $x_3 \equiv 0$ ,  $x_4 \equiv 0$ , and

$$\Lambda(x_2)^{-1} \begin{bmatrix} u \\ g \sin x_2 \end{bmatrix} = 0.$$

Therefore, the equilibrium points are:

$$(x_e, u_e) = \left( \begin{bmatrix} x_1 \\ k\pi \\ 0 \\ 0 \end{bmatrix}, 0 \right), \quad \text{for every } x_1 \in \mathbb{R} \text{ and } k = 0, \pm 1, \pm 2, \dots$$

Slide 10

### Inverted pendulum: Jacobian matrices

Jacobian matrices<sup>a</sup>:

$$\frac{\partial f_{1,2}}{\partial \mathbf{x}} = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix},$$

$$\frac{\partial f_{3,4}}{\partial \mathbf{x}} = \Lambda(\mathbf{x}_2)^{-1} \begin{bmatrix} 0 & mlx_4^2 \cos x_2 + \phi_1(x_2, x_3, x_4) \sin x_2 & -\mu & 2mlx_4 \sin x_2 \\ 0 & g \cos x_2 + \phi_2(x_2, x_3, x_4) \sin x_2 & 0 & 0 \end{bmatrix},$$

where  $\phi_1$  and  $\phi_2$  are some algebraic functions of  $x_2$ ,  $x_3$ , and  $x_4$ ,

$$\frac{\partial f}{\partial \mathbf{u}} = \frac{1}{l(M+m \sin^2 x_2)} \begin{bmatrix} 0 \\ 0 \\ 1 \\ -\cos x_2 \end{bmatrix}.$$

<sup>a</sup>Given two matrix functions  $M_1(\eta)$  and  $M_2(\eta)$ , then

$$\frac{\partial}{\partial \eta} (M_1(\eta)^{-1} M_2(\eta)) = M_1(\eta)^{-1} \left( \frac{\partial M_2(\eta)}{\partial \eta} - \frac{\partial M_1(\eta)}{\partial \eta} M_1(\eta)^{-1} M_2(\eta) \right).$$

Slide 11

### Inverted pendulum: linearization around (0,0)

It's readily verified that

$$\left. \frac{\partial f}{\partial \mathbf{x}} \right|_0 = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ 0 & -\frac{m}{M}g & -\frac{\mu}{M} & 0 \\ 0 & \frac{m+M}{Ml}g & \frac{\mu}{Ml} & 0 \end{bmatrix}$$

and

$$\left. \frac{\partial f}{\partial \mathbf{u}} \right|_0 = \begin{bmatrix} 0 \\ 0 \\ \frac{1}{M} \\ -\frac{1}{Ml} \end{bmatrix},$$

which yield the "A" and "B" matrices of the linearized realization.

Slide 12

## Linearization of discrete-time systems

---

The linearization problem for discrete-time system

$$\begin{cases} x_{k+1} = f(x_k, u_k), \\ y_k = g(x_k, u_k) \end{cases}$$

where  $x \in \mathbb{R}^n$ ,  $u \in \mathbb{R}^m$ ,  $y \in \mathbb{R}^p$ , has exactly the same solution as in the continuous time, except that the *equilibrium* is obtained from

$$f(x_e, u_e) = x_e.$$

Slide 13