

# Plan of the Lecture

- ▶ **Review:** frequency-domain design method.
- ▶ **Today's topic:** introduction to state-space design.

*Goal:* introduce basic notions of state-space control: different state-space realizations of the same transfer function; several canonical forms of state-space systems; controllability matrix.

*Reading:* FPE, Chapter 7

## Frequency-Domain vs. State-Space

- ▶ 90% of industrial controllers are designed using frequency-domain methods (PID is a popular architecture)
- ▶ 90% of current research in systems and control is in the state-space framework

To be able to talk to control engineers *and* follow progress in the field, we need to know both methods and *understand the connections between them.*

# State-Space Methods

- ▶ the state-space approach reveals *internal system architecture* for a given transfer function
- ▶ the mathematics is different: heavy use of *linear algebra*
- ▶ this is just a short introduction; to learn this material properly, take ECE 515

## A General State-Space Model

$$\text{state } x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix} \in \mathbb{R}^n \quad \text{input } u = \begin{pmatrix} u_1 \\ \vdots \\ u_m \end{pmatrix} \in \mathbb{R}^m$$

$$\text{output } y = \begin{pmatrix} y_1 \\ \vdots \\ y_p \end{pmatrix} \in \mathbb{R}^p$$

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

where:

$A$  – system matrix ( $n \times n$ )

$B$  – input matrix ( $n \times m$ )

$C$  – output matrix ( $p \times n$ )

$D$  – feedthrough matrix ( $p \times m$ )

## From State-Space to Transfer Function

Let us find the *transfer function* from  $u$  to  $y$  corresponding to the state-space model

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

- ▶ in the scalar case ( $x, y, u \in \mathbb{R}$ ), we took the Laplace transform
- ▶ the same idea here when working with vectors: just do it component by component

## From State-Space to Transfer Function

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

$$x = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}, \quad u = \begin{pmatrix} u_1 \\ u_2 \\ \vdots \\ u_m \end{pmatrix}, \quad y = \begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_p \end{pmatrix}$$

Recall matrix-vector multiplication:

$$\begin{aligned} \dot{x}_i &= (Ax)_i + (Bu)_i \\ &= \sum_{j=1}^n a_{ij}x_j + \sum_{k=1}^m b_{ik}u_k \end{aligned}$$

$$\begin{aligned} y_\ell &= (Cx)_\ell + (Du)_\ell \\ &= \sum_{j=1}^n c_{\ell j}x_j + \sum_{k=1}^m d_{\ell k}u_k \end{aligned}$$

## From State-Space to Transfer Function

Now we take the Laplace transform:

$$\dot{x}_i = \sum_{j=1}^n a_{ij}x_j + \sum_{k=1}^m b_{ik}u_k$$

$\downarrow \mathcal{L}$

$$sX_i(s) - x_i(0) = \sum_{j=1}^n a_{ij}X_j(s) + \sum_{k=1}^m b_{ik}U_k(s), \quad i = 1, \dots, n$$

Write down in matrix-vector form:

$$sX(s) - x(0) = AX(s) + BU(s)$$

$$(Is - A)X(s) = x(0) + BU(s) \quad (I \text{ is the } n \times n \text{ identity matrix})$$

$$X(s) = (Is - A)^{-1}x(0) + (Is - A)^{-1}BU(s)$$

## From State-Space to Transfer Function

$$y_\ell = \sum_{j=1}^n c_{\ell j} x_j + \sum_{k=1}^m d_{\ell k} u_k$$

$\downarrow \mathcal{L}$

$$Y_\ell(s) = \sum_{j=1}^n c_{\ell j} X_j(s) + \sum_{k=1}^m d_{\ell k} U_k(s), \quad \ell = 1, \dots, p$$

Write down in matrix-vector form:

$$\begin{aligned} Y(s) &= CX(s) + DU(s) \\ &= C [(Is - A)^{-1}x(0) + (Is - A)^{-1}BU(s)] + DU(s) \\ &= C(Is - A)^{-1}x(0) + [C(Is - A)^{-1}B + D] U(s) \end{aligned}$$

To find the input-output t.f., set the IC to 0:

$$Y(s) = G(s)U(s), \quad \text{where } G(s) = C(Is - A)^{-1}B + D$$



## From State-Space to Transfer Function

The transfer function from  $u$  to  $y$ , corresponding to

$$\dot{x} = Ax + Bu$$

$$y = Cx + Du$$

is given by

$$G(s) = C(Is - A)^{-1}B + D$$

Observe that  $G(s)$  contains information about the state-space matrices  $A, B, C, D!!$

## From State-Space to Transfer Function

$$\begin{aligned} \dot{x} &= Ax + Bu & Y(s) &= G(s)U(s) \\ y &= Cx + Du & &= [C(Is - A)^{-1}B + D] U(s) \end{aligned}$$

### **Important!!**

- ▶  $G(s)$  is *undefined* when the  $n \times n$  matrix  $Is - A$  is *singular* (or noninvertible), i.e., precisely when  $\det(Is - A) = 0$
- ▶ since  $A$  is  $n \times n$ ,  $\det(Is - A)$  is a *polynomial* of degree  $n$  (the *characteristic polynomial* of  $A$ ):

$$\det(Is - A) = \det \begin{pmatrix} s - a_{11} & -a_{12} & \dots & -a_{1n} \\ -a_{21} & s - a_{22} & \dots & -a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ -a_{n1} & -a_{n2} & \dots & s - a_{nn} \end{pmatrix},$$

and its roots are the *eigenvalues* of  $A$

- ▶  $G$  is (open-loop) stable if all eigenvalues of  $A$  lie in LHP.

## Example: Computing $G(s)$

Consider the state-space model in **Controller Canonical Form (CCF)\***:

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ -6 & -5 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \underbrace{\begin{pmatrix} 0 \\ 1 \end{pmatrix}}_B u, \quad y = \underbrace{\begin{pmatrix} 1 & 1 \end{pmatrix}}_C \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

— this is a *single-input, single-output* (SISO) system, since  $u, y \in \mathbb{R}$ ; the state is two-dimensional.

Let's compute the transfer function:

$$G(s) = C(Is - A)^{-1}B \quad (D = 0 \text{ here})$$

$$Is - A = \begin{pmatrix} s & -1 \\ 6 & s + 5 \end{pmatrix}$$

\* We will explain this terminology later.

## Example: Computing $G(s)$

$$Is - A = \begin{pmatrix} s & -1 \\ 6 & s + 5 \end{pmatrix} \quad \text{— how do we compute } (Is - A)^{-1}?$$

A useful formula for the inverse of a  $2 \times 2$  matrix:

$$M = \begin{pmatrix} a & b \\ c & d \end{pmatrix}, \det M \neq 0 \implies M^{-1} = \frac{1}{\det M} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

Applying the formula, we get

$$\begin{aligned} (Is - A)^{-1} &= \frac{1}{\det(Is - A)} \begin{pmatrix} s + 5 & 1 \\ -6 & s \end{pmatrix} \\ &= \frac{1}{s^2 + 5s + 6} \begin{pmatrix} s + 5 & 1 \\ -6 & s \end{pmatrix} \end{aligned}$$

## Example: Computing $G(s)$

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ -6 & -5 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \underbrace{\begin{pmatrix} 0 \\ 1 \end{pmatrix}}_B u, \quad y = \underbrace{\begin{pmatrix} 1 & 1 \end{pmatrix}}_C \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\begin{aligned} G(s) &= C(Is - A)^{-1}B \\ &= \begin{pmatrix} 1 & 1 \end{pmatrix} \frac{1}{s^2 + 5s + 6} \begin{pmatrix} s + 5 & 1 \\ -6 & s \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} \\ &= \frac{1}{s^2 + 5s + 6} \begin{pmatrix} 1 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ s \end{pmatrix} \\ &= \frac{s + 1}{s^2 + 5s + 6} \end{aligned}$$

- ▶ the above state-space model is a *realization* of this t.f.
- ▶ note how coefficients 5 and 6 appear in both  $G(s)$  and  $A$ !!

## State-Space Realizations of Transfer Functions

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ -6 & -5 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \underbrace{\begin{pmatrix} 0 \\ 1 \end{pmatrix}}_B u, \quad y = \underbrace{\begin{pmatrix} 1 & 1 \end{pmatrix}}_C \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$
$$G(s) = \frac{s + 1}{s^2 + 5s + 6}$$

— at least in this example, information about the state-space model  $(A, B, C)$  is contained in  $G(s)$ .

Is this information *recoverable*? — i.e., is there only one state-space realization of a given t.f.? Or are there many?

**Answer:** There are infinitely many!

## State-Space Realizations of Transfer Functions

Start with

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ -6 & -5 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \underbrace{\begin{pmatrix} 0 \\ 1 \end{pmatrix}}_B u, \quad y = \underbrace{\begin{pmatrix} 1 & 1 \end{pmatrix}}_C \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

and consider a new state-space model

$$\dot{x} = \bar{A}x + \bar{B}u, \quad y = \bar{C}x$$

with

$$\bar{A} = A^T = \begin{pmatrix} 0 & -6 \\ 1 & -5 \end{pmatrix}, \quad \bar{B} = C^T = \begin{pmatrix} 1 \\ 1 \end{pmatrix}, \quad \bar{C} = B^T = \begin{pmatrix} 0 & 1 \end{pmatrix}$$

This is a different state-space model!

## State-Space Realizations of Transfer Functions

**Claim:** The state-space model

$$\dot{x} = \bar{A}x + \bar{B}u, \quad y = \bar{C}x$$

with

$$\bar{A} = A^T, \quad \bar{B} = C^T, \quad \bar{C} = B^T$$

has the same transfer function as the original model with  $(A, B, C)$ .

**Proof:**

$$\begin{aligned} \bar{C}(Is - \bar{A})^{-1}\bar{B} &= B^T (Is - A^T)^{-1} C^T \\ &= B^T [(Is - A)^T]^{-1} C^T \\ &= B^T [(Is - A)^{-1}]^T C^T \\ &= [C(Is - A)^{-1}B]^T \\ &= C(Is - A)^{-1}B \end{aligned}$$



# State-Space Realizations of Transfer Functions

The state-space model

$$\dot{x} = \bar{A}x + \bar{B}u, \quad y = \bar{C}x$$

with

$$\bar{A} = A^T, \quad \bar{B} = C^T, \quad \bar{C} = B^T$$

has the same transfer function as the original model with  $(A, B, C)$ .

But the state-space model is now in the **Observer Canonical Form (OCF)**:

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} 0 & -6 \\ 1 & -5 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} 1 \\ 1 \end{pmatrix} u, \quad y = \begin{pmatrix} 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

## Even More Realizations ...

Yet another realization of  $G(s) = \frac{s+1}{s^2+5s+6}$  can be extracted from the partial-fractions decomposition:

$$G(s) = \frac{s+1}{(s+2)(s+3)} = \frac{2}{s+3} - \frac{1}{s+2}.$$

This is the **Modal Canonical Form (MCF)**:

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \begin{pmatrix} -3 & 0 \\ 0 & -2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \begin{pmatrix} 1 \\ 1 \end{pmatrix} u, \quad y = (2 \quad -1) \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\begin{aligned} \text{Then } C(Is - A)^{-1}B &= (2 \quad -1) \begin{pmatrix} s+3 & 0 \\ 0 & s+2 \end{pmatrix}^{-1} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \\ &= (2 \quad -1) \begin{pmatrix} \frac{1}{s+3} & 0 \\ 0 & \frac{1}{s+2} \end{pmatrix} \begin{pmatrix} 1 \\ 1 \end{pmatrix} \\ &= (2 \quad -1) \begin{pmatrix} \frac{1}{s+3} \\ \frac{1}{s+2} \end{pmatrix} = \frac{2}{s+3} - \frac{1}{s+2} \end{aligned}$$

## State-Space Realizations: Bottom Line

- ▶ a given transfer function  $G(s)$  can be realized using infinitely many state-space models
- ▶ certain properties make some realizations preferable to others
- ▶ one such property is *controllability*

# Controllability Matrix

Consider a single-input system ( $u \in \mathbb{R}$ ):

$$\dot{x} = Ax + Bu, \quad y = Cx \quad x \in \mathbb{R}^n$$

The **Controllability Matrix** is defined as

$$\mathcal{C}(A, B) = [B \mid AB \mid A^2B \mid \dots \mid A^{n-1}B]$$

- recall that  $A$  is  $n \times n$  and  $B$  is  $n \times 1$ , so  $\mathcal{C}(A, B)$  is  $n \times n$ ;
- the controllability matrix only involves  $A$  and  $B$ , not  $C$

We say that the above system is **controllable** if its controllability matrix  $\mathcal{C}(A, B)$  is *invertible*.

(This definition is only true for the single-input case; the multiple-input case involves the *rank* of  $\mathcal{C}(A, B)$ .)

# Controllability Matrix

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We say that the above system is **controllable** if its controllability matrix  $\mathcal{C}(A, B)$  is *invertible*.

- ▶ As we will see later, if the system is controllable, then we may assign arbitrary closed-loop poles by *state feedback* of the form  $u = -Kx$ .
- ▶ Whether or not the system is controllable depends on its state-space realization.

## Example: Computing $\mathcal{C}(A, B)$

Let's get back to our old friend:

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ -6 & -5 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \underbrace{\begin{pmatrix} 0 \\ 1 \end{pmatrix}}_B u, \quad y = \underbrace{\begin{pmatrix} 1 & 1 \end{pmatrix}}_C \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Here,  $x \in \mathbb{R}^2 \implies A \in \mathbb{R}^{2 \times 2} \implies \mathcal{C}(A, B) \in \mathbb{R}^{2 \times 2}$

$$\begin{aligned} \mathcal{C}(A, B) &= [B \mid AB] & AB &= \begin{pmatrix} 0 & 1 \\ -6 & -5 \end{pmatrix} \begin{pmatrix} 0 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ -5 \end{pmatrix} \\ \implies \mathcal{C}(A, B) &= \begin{pmatrix} 0 & 1 \\ 1 & -5 \end{pmatrix} \end{aligned}$$

Is this system controllable?

$$\det \mathcal{C} = -1 \neq 0 \quad \implies \quad \text{system is controllable}$$

## Controller Canonical Form

A single-input state-space model

$$\dot{x} = Ax + Bu, \quad y = Cx$$

is said to be in **Controller Canonical Form** (CCF) if the matrices  $A, B$  are of the form

$$A = \begin{pmatrix} 0 & 1 & 0 & \dots & 0 & 0 \\ 0 & 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & 0 & 1 \\ * & * & * & \dots & * & * \end{pmatrix}, \quad B = \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \end{pmatrix}$$

A system in CCF is *always controllable!!*

(The proof of this for  $n > 2$  uses the Jordan canonical form, we will not worry about this.)

## CCF with Arbitrary Zeros

In our example, we had  $G(s) = \frac{s+1}{s^2+5s+6}$ , with a minimum-phase zero at  $z = -1$ .

Let's consider a general zero location  $s = z$ :

$$G(s) = \frac{s-z}{s^2+5s+6}$$

This gives us a CCF realization

$$\begin{pmatrix} \dot{x}_1 \\ \dot{x}_2 \end{pmatrix} = \underbrace{\begin{pmatrix} 0 & 1 \\ -6 & -5 \end{pmatrix}}_A \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} + \underbrace{\begin{pmatrix} 0 \\ 1 \end{pmatrix}}_B u, \quad y = \underbrace{\begin{pmatrix} -z & 1 \end{pmatrix}}_C \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

Since  $A, B$  are the same,  $C(A, B)$  is the same  $\implies$  the system is still controllable.

A system in CCF is controllable for any locations of the zeros.