# **Kuwait University** College of Engineering and Petroleum



# جامعة الكويت KUMAIT UNIVERSITY

# **ME417 CONTROL OF MECHANICAL SYSTEMS** PART II: CONTROLLER DESIGN VIA STATE-SPACE LECTURE 1: STATE-SPACE REPRESENTATION

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# Lecture Plan

- Objectives:
	- Introduction to modeling in the time domain
	- Treating the general state-space representation
	- Converting from state-space to transfer function and back
- Reading:
	- *Nise: 3.1-3.6*
- Practice Problems Included





- But why not just use transfer functions?
- Transfer functions are powerful. They help us understand dynamic systems behavior and they have a good set of tools (root-locus, bode plots)
- However, they are limited to
	- 1. Linear systems
		- Real world systems are often nonlinear
	- 2. Time-invariant systems
		- Systems may change their properties over time
- High order systems produce mathematically complex transfer functions
- A transfer function can only relate one input to one output
- Multi-degree of freedom systems must be decoupled if T.F. representation is desired



• The general state-space representation is

$$
\dot{x}(t) \in \mathcal{R}^{n \times 1} = f(x, t, u) = A(x, t) + B(x, t)
$$

$$
y(t) = C(x, t) + D(x, t)
$$

 $f(x, t, u)$  is called the system model, which is a function of the system state x, time t and the input to the system  $u$ 

- Note that  $f(x, t, u)$  correspond to the model function we use to simulate a general dynamic system numerically.
- The state-space representation can represent
	- Time-varying systems
	- Nonlinear dynamics
	- Multiple input and multiple outputs
	- Matrix representation for complex systems
		- Increasing system order increases matrix dimensions, rather than high order polynomial



• For linear time-invariant systems (the scope of this course) the representation is simplified to

> $\dot{x} = Ax + Bu$  $y = Cx + Du$

- $x \in \mathbb{R}^n$ : State vector,  $\dot{x} \in \mathbb{R}^n$ : derivative of state vector,
- $\mathbf{A} \in \mathcal{R}^{n \times n}$ : Constant System matrix,  $\mathbf{B} \in \mathcal{R}^{n \times p}$ : Constant Input matrix
- $\boldsymbol{u} \in \mathcal{R}^p$ : Input vector,
- $y \in \mathcal{R}^m$ : Output vector,  $\mathbf{C} \in \mathcal{R}^{m \times n}$ : Constant Output matrix,
- $\mathbf{D} \in \mathcal{R}^{n \times p}$ : Constant Feedforward matrix
- $\in \mathbb{R}^{n \times n}$ : denotes the matrix size is  $n \times n$  and its values are real
- $\blacksquare \in \mathcal{R}^n$ : denotes the vector size is  $n$  and its values are real
- n: Number of state variables, p: Number of inputs, m: Number of outputs



Linear Algebra Flash Refresher

• Matrix Addition/Subtraction

$$
\mathbf{A}_1 + \mathbf{A}_2 = \mathbf{B} \Rightarrow \begin{bmatrix} 1 & -2 & 3 \\ 0 & 2 & -5 \\ 2 & 2 & 7 \end{bmatrix} + \begin{bmatrix} -1 & 1 & 1 \\ 1 & 4 & 5 \\ 2 & -2 & 7 \end{bmatrix} = \begin{bmatrix} 0 & -1 & 4 \\ 1 & 6 & 0 \\ \frac{4}{3x3} & \frac{1}{3x3} \end{bmatrix}
$$

• Matrix Scaling

$$
c\mathbf{A} = \mathbf{B} \Rightarrow 3 \begin{bmatrix} 1 & 2 & 1 \\ 3 & 2 & 3 \\ 4 & 3 & 2 \end{bmatrix} = \begin{bmatrix} 3 & 6 & 3 \\ 9 & 6 & 9 \\ 12 & 9 & 6 \end{bmatrix}
$$

• Matrix Multiplication

$$
\mathbf{A}x = y \Rightarrow \begin{bmatrix} 1 & 2 & 1 \\ 3 & 2 & 3 \\ 4 & 3 & 2 \end{bmatrix} \begin{bmatrix} 3 \\ 4 \\ 2 \end{bmatrix} = \begin{bmatrix} 13 \\ 23 \\ 28 \end{bmatrix}, \quad \mathbf{A}\mathbf{B} = \mathbf{C} \Rightarrow \begin{bmatrix} 1 & 2 & 3 \\ 0 & 4 & 2 \\ 2 & 1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 2 & 3 \\ 4 & 2 & 0 \\ 1 & 2 & 0 \end{bmatrix} = \begin{bmatrix} 12 & 12 & 3 \\ 18 & 12 & 0 \\ 8 & 10 & 6 \end{bmatrix}
$$



• Matrix Inverse

$$
\mathbf{A} \in \mathcal{R}^{n \times n}, \quad \mathbf{A}^{-1} = \frac{adj(\mathbf{A})}{det(\mathbf{A})}, \quad \mathbf{A}\mathbf{A}^{-1} = \mathbf{I}
$$

• Matrix Adjugate

$$
for \mathbf{A} \in \mathbb{R}^{2 \times 2}, \quad \mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, \quad \text{adj}(\mathbf{A}) = \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}
$$

• Matrix Determinant

for 
$$
\mathbf{A} \in \mathbb{R}^{2 \times 2}
$$
,  $\mathbf{A} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$ ,  $det(\mathbf{A}) = |\mathbf{A}| = ab - cd$ 

• Matrix Eigenvalues

 $\lambda_i \in \mathcal{C}^n$  are the special constants of a matrix  $\mathbf{A} \in \mathcal{R}^{n \times n}$ , called the eigenvalues  $\mathbf{A}\mathbf{x} = \lambda_i \mathbf{x}, A\mathbf{x} - \lambda_i \mathbf{x} = (\mathbf{A} - \lambda_i \mathbf{I})\mathbf{x} = 0$ 

The trivial solution is  $x = 0$ , while the nontrivial solution occurs if the determinant vanishes  $\det(\mathbf{A} - \lambda_i \mathbf{I}) = 0, i = 1, 2, ..., n$ 

# The state-space representation

- Consider the mass-spring-damper mechanical system
- The equation of motion is:  $M\ddot{x} + f_p\dot{x} + Kx = f(t) = u(t)$
- This is a second-order system, requiring 2 state variables
	- Corresponding to the two derivatives

$$
\boldsymbol{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x \\ \dot{x} \end{bmatrix}
$$

$$
\dot{x} = \begin{bmatrix} \dot{x} \\ \dot{x} \end{bmatrix} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ (u(t) - f_v \dot{x} - Kx)/M \end{bmatrix} = \begin{bmatrix} x_2 \\ -f_v/Mx_2 - K/Mx_1 \end{bmatrix} + \begin{bmatrix} 0 \\ u(t)/M \end{bmatrix}
$$





# The state-space representation

$$
\dot{\boldsymbol{x}} = \begin{bmatrix} 0 & 1 \\ -K/M & -f_v/M \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \\ 1/M \end{bmatrix} u(t) = \mathbf{A}\boldsymbol{x} + \mathbf{B}\boldsymbol{u}
$$

$$
y = x = x_1 = \begin{bmatrix} 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} 0 \end{bmatrix} u(t) = \mathbf{C} x + \mathbf{D} u
$$



$$
\mathbf{A} \in \mathcal{R}^{2 \times 2} = \begin{bmatrix} 0 & 1 \\ -K/M & -f_v/M \end{bmatrix}, \qquad \mathbf{B} \in \mathcal{R}^{2 \times 1} = \begin{bmatrix} 0 \\ 1/M \end{bmatrix}, \qquad \frac{F(s)}{\mathcal{M}^{s^2 + f_v s + K}} \longrightarrow \frac{X(s)}{\mathcal{M}^{s^2 + f_v s + K}}
$$

 $C \in \mathcal{R}^{1 \times 2} = [1 \quad 0], \quad D \in \mathcal{R} = 0$ 

• In this course  $\mathbf{D} = 0$ 





# Derive the dynamic model of the system **Example 1** Example 1 shown in state-space form



# Example 1 - Continue $•1966$ **جامعة الكويت**<br>KUWAIT UNIVERSITY



• Given an n<sup>th</sup> order differential equation

$$
a_n \frac{d^n x}{dt^n} + a_{n-1} \frac{d^{n-1} x}{dt^{n-1}} + \dots + a_1 \frac{dx}{dt} + a_0 x = f(t)
$$

• We can write it as n simultaneous first-order differential equations:

$$
\frac{dx_1}{dt} = a_{11}x_1 + a_{12}x_2 + \dots + a_{1n-1}x_{n-1} + a_{1n}x_n + b_1f(t)
$$
  
\n
$$
\vdots
$$
  
\n
$$
\frac{dx_i}{dt} = a_{i1}x_1 + a_{i2}x_2 + \dots + a_{in-1}x_{n-1} + a_{in}x_n + b_i f(t)
$$
  
\n
$$
\vdots
$$
  
\n
$$
\frac{dx_n}{dt} = a_{n1}x_1 + a_{n2}x_2 + \dots + a_{nn-1}x_{n-1} + a_{nn}x_n + b_nf(t)
$$



• In state-space form, a general  $n<sup>th</sup>$ -order set of differential equations can be represented as:

$$
\dot{\boldsymbol{x}} = \begin{bmatrix} \dot{x}_1 \\ \vdots \\ \dot{x}_n \end{bmatrix} = \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{n1} & \cdots & a_{nn} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix} f(t)
$$

• The output y depends on what we are measuring. If state  $x_1$  is the output then:

$$
\mathbf{y} = \begin{bmatrix} 1 & 0 & \cdots & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} + 0
$$

• Note that we use  $y(t)$  or  $c(t)$  interchangeably, to express the output. In statespace, it is common convention to use  $y(t)$  for the output.



# From Transfer Function to State-Space

- To convert a transfer function into state space, we first convert it to differential equation form
- Given:  $G(s) = \frac{C(s)}{U(s)}$  $U(s)$  $=\frac{b_0}{\sqrt{a_0^2 + a_0^2 + \cdots + a_n^2}}$  $a_n s^n + a_{n-1} s^{n-1} + \dots + a_1 s + a_0$ , converting to diff. eq. • Gives:  $a_n \frac{d^n c(t)}{dt^n}$  $\frac{n_{c(t)}}{dt^n} + a_n \frac{d^n c(t)}{dt^n}$  $\frac{n_{c(t)}}{dt^n} + \cdots + a_1 \frac{dc(t)}{dt}$  $\frac{c(t)}{dt} + a_0 c(t) = b_0 u(t)$
- An n<sup>th</sup> order diff. eq. gives  $n$  states

$$
x = \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} c & dc/dt & \dots & d^{n-1}c/dt^{n-1} & d^{n}c/dt^{n}]^T \\ 0 & 1 & 0 & \dots & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \dots & \dots & \dots & 0 & 1 & 0 \\ \dots & \dots & \dots & \dots & 0 & 1 \\ a_0/a_n & a_1/a_n & \dots & \dots & a_{n-2}/a_n & a_{n-1}/a_n \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} + \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 1/a_n \end{bmatrix} u(t)
$$
  

$$
y = \begin{bmatrix} 1 & 0 & \cdots & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} + 0
$$



- The state-space form retrieved from the transfer function, is called a phase variable form
	- Note the off-diagonal identity matrix, and how all the coefficients are grouped in the nth row

$$
A = \begin{bmatrix} 0 & 1 & 0 & \dots & \dots & \dots \\ \dots & 0 & 1 & 0 & \dots & \dots \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ \dots & \dots & \dots & 0 & 1 & 0 \\ \dots & \dots & \dots & \dots & 0 & 1 \\ \dots & \dots & \dots & \dots & 0 & 1 \\ a_0/a_n & a_1/a_n & \dots & \dots & a_{n-2}/a_n & a_{n-1}/a_n \end{bmatrix}
$$

- Unlike the state-space form retrieved from the equations of motion directly.
- Both the standard, and phase variable forms are valid representations







From State Space to Transfer Function

• Given a state space representation of an LTI system, there is an analytical solution to expressing the system in transfer function form

• Given

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

- Take the Laplace transform of both sides, we get  $sX(s) = AX(s) + BU(s)$  $Y(s) = CX(s) + DU(s)$
- Solving for  $X(s)$

 $sI - A$ ) $X(s) = BU(s) \Rightarrow X(s) = (sI - A)^{-1}BU(s)$ , where I is the identity matrix



• Substituting  $X(s)$  in  $Y(s)$ 

$$
Y(s) = C((sI - A)^{-1}BU(s)) + DU(s)
$$

$$
Y(s) = [C((sI - A)^{-1}B) + D]U(s)
$$

- The term  $\left[\textbf{C}\big((s\textbf{I}-\textbf{A})^{-1}\textbf{B}\big)+\textbf{D}\right]$ , is the *transfer function matrix*  $G(s) =$  $Y(s)$  $U(s)$  $= \left[ C((sI - A)^{-1}B) + D \right]$
- The inverse term  $(s\bm{I}-\mathbf{A})^{-1}$ , can be computed as

$$
(sI - A)^{-1} = \frac{adj(sI - A)}{det(sI - A)}
$$



Example 3 From the following system given in state-space form, find the transfer function

$$
\dot{x} = \begin{bmatrix} 5 & 3 \\ 2 & 3 \end{bmatrix} x + \begin{bmatrix} 1 \\ 0 \end{bmatrix} u(t)
$$

$$
y = \begin{bmatrix} 1 & 0 \end{bmatrix} x
$$



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- Note that the term  $det(sI A)$ , is the denominator of the transfer function.
- In other words: it's the characteristic equation for the transfer function.
	- The roots of which are the poles of the system
- Note that the poles of the system are a function of the system matrix **A**
- So, stability can be evaluated by knowing the matrix **A**
- Remember that stability is defined in the context of the natural response (no input):  $\dot{x} = Ax$
- Also note that the poles of the system are the eigenvalues  $\lambda_i$  of the matrix **A**



# Is the following system stable? Example 5

$$
\dot{x} = \begin{bmatrix} 5 & 3 \\ 8 & 3 \end{bmatrix} x + \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(t)
$$

$$
\mathbf{y} = \begin{bmatrix} 1 & 0 \end{bmatrix} x
$$



