

# Class 3

Lecture 5

Solutions to LTV Systems

Lecture 6

Solutions to LTI Systems

# Lecture 5

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## Outline

Fundamental Theorem of Differential Equations

State Transition Matrix

Solutions of D.E.'s - LTV Systems

Homogeneous Solution ( $u(t) \equiv 0$ )

$$\begin{cases} \dot{x}(t) = A(t)x(t) \\ x(t_0) = x_0 \end{cases}$$

Nonhomogeneous Solution

$$\begin{cases} \dot{x}(t) = A(t)x(t) + B(t)u(t) \\ x(t_0) = x_0 \end{cases}$$

## Lecture 5

Fundamental Theorem of Differential Equations

Given the differential equation

$$\begin{cases} \dot{x}(t) = f(x, t) \\ x(t_0) = x_0 \end{cases}$$

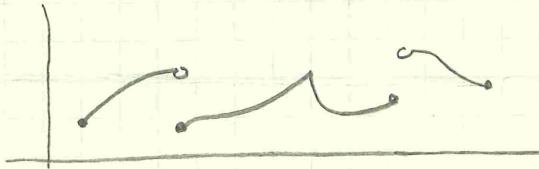
Let  $f: \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}^n$  satisfy

- (a)  $\forall x \in \mathbb{R}^n, t \mapsto f(x, t)$  is piecewise continuous  
with its set of discontinuity points  $= D$  may be zero

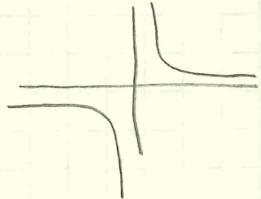
Note Piecewise continuous

A function is piecewise continuous if it consists of a finite number of continuous pieces with finite discontinuities between them.

e.g.,

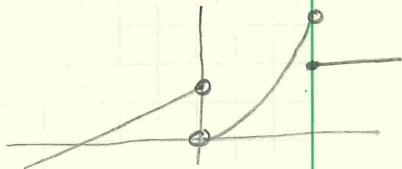


e.g.,  $f(x) = \frac{1}{x}$  is not p.c.



b/c discontinuity  
between two  
continuous pieces  
is not finite

e.g.,  $f(x) = \begin{cases} x+4 & x < 0 \\ x^2 & 0 \leq x < 5 \\ 7 & x \geq 5 \end{cases}$



(b)  $f(x, t)$  satisfies the Lipschitz Condition:

$\forall t \in \mathbb{R}^+, \forall \xi, \xi' \in \mathbb{R}^n, \exists L(t)$  piecewise continuous fct

$$\text{s.t. } \|f(\xi, t) - f(\xi', t)\| \leq L(t) \|\xi - \xi'\|$$



norm = length of vector

e.g.,  $x = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix} \quad \|x\| = \sqrt{x_1^2 + x_2^2 + \dots + x_n^2}$

Then

If (1) Existence

$\forall x_0 \in \mathbb{R}^n, \forall t_0 \in \mathbb{R}^+, \exists \phi : \mathbb{R}^+ \rightarrow \mathbb{R}^n$  continuous s.t.

$$\begin{cases} \dot{\phi}(t) = f(\phi(t), t) & \forall t \in \mathbb{R}^+, t \neq t_0 \\ \phi(t_0) = x_0 \end{cases}$$

(2) Uniqueness

$\phi$  is unique

(3)  $\phi$  depends on  $x_0$  continuously

Then  $\Rightarrow \phi$  is called the solution of the differential equation.

Remark Importance of Fundamental Theorem

- \* Gives a large class of problems for which we can speak of the unique solution
- \* We rely on the Theorem to prove equalities e.g. to show two functions are equal

$$L(t) = R(t) \quad \forall t \in \mathbb{R}^+$$

i.e., We may observe that

$$(i) L(t_0) = R(t_0)$$

(ii)  $L(\cdot) = R(\cdot)$  are solutions to

the same d.e., which itself satisfies the conditions of the Fundamental Theorem

## Sketch of Proof

### Proof of Existence

(Picard Iteration Scheme) (by construction)

On  $[0, \infty)$ , construct a sequence of functions

$$x_{m+1}(t) = x_0 + \int_{t_0}^t f(x_m(t'), t') dt' \quad \forall t \in [0, \infty)$$

Show that on every finite interval,  $[t_0, t_1]$  say,

$\{x_i\}_{i=0}^{\infty}$  converges uniformly to a continuous

function that satisfies the d.e. wherever  $t \notin D$ .

### Proof of Uniqueness (by contradiction)

Suppose  $\exists$  two solutions  $\phi(t)$  and  $\psi(t)$  satisfying

$$\begin{cases} \dot{x} = f(x, t) \\ x(t_0) = x_0 \end{cases}$$

By integrating the d.e. and subtracting,

$$\phi(t) - \psi(t) = \int_{t_0}^t [f(\phi(\tau), \tau) - f(\psi(\tau), \tau)] d\tau \quad \forall t \in \mathbb{R}^+$$

Restricting our attention to a finite interval  $[t_0, t_1]$ ,

$$\|\phi(t) - \psi(t)\| \leq \int_{t_0}^t \|f(\phi(\tau), \tau) - f(\psi(\tau), \tau)\| d\tau$$

$$\text{Lipschitz Condition} \quad \leq L \int_{t_0}^t \|\phi(\tau) - \psi(\tau)\| d\tau \quad \forall t \in [t_0, t_1]$$

Then, for any  $c_1 \geq 0$ ,

$$\|\phi(t) - \psi(t)\| \leq c_1 + L \int_{t_0}^t \|\phi(\tau) - \psi(\tau)\| d\tau \quad L \text{ constant}$$

$$\text{Bellman-Gronwall Lemma} \quad \leq c_1 e^{L(t-t_0)} \quad \forall t \in [t_0, t_1]$$

Holds  $\forall c_1 \geq 0$ . Thus, for  $c_1 = 0$ , we must have

$$\|\phi(t) - \psi(t)\| = 0 \quad \forall t \in [t_0, t_1]$$

$$\Rightarrow \phi(t) = \psi(t) \quad \forall t \in [t_0, t_1]$$

$\Rightarrow$  Solution is unique on  $[t_0, t_1]$

Extend domain of uniqueness  
step-by-step to all  $t \in \mathbb{R}^+$

**Bellman-Gronwall Lemma**

$c_1 \geq 0$  constant.

$u(\cdot), k(\cdot)$  real-valued p.c.  
fcts on  $\mathbb{R}^+$ ,  $k(\cdot)$  non-negative.

If  $u(t) \leq c_1 + \int_{t_0}^t k(\tau) u(\tau) d\tau$

Then  $u(t) \leq c_1 \exp \int_{t_0}^t k(\tau) d\tau$   
 $\forall t \in \mathbb{R}^+$

## Def. Dynamical System

A model of a physical system is a dynamical system iff we can associate with it a quintuple (list of 5 objects)

$$D = (U, \Sigma, Y, S, R)$$

satisfying two axioms.

Let  $U = \{u(t), t \in T \subset \text{TR}\}$  input set

$\Sigma = \{x(t), t \in T\}$  state set

$Y = \{y(t), t \in T\}$  output set

$S = \text{state transition function}$

$$x(t_1) = S(t_1, t_0, x_0, u) \quad t_1 \geq t_0, x_0 \in \Sigma$$

"the state at  $t_1$  reached from  $x_0$  at  $t_0$  as a result of  $u$ "

$$t_0, t_1 \in T$$

$R = \text{read-out function}$

$$y(t) = R(t, x(t), u(t)) \quad t \in T, x \in \Sigma, u \in U$$

"the response of the system at  $t$  when it is in state  $x(t)$  at  $t$  and the input is  $u(t)$ "

State Transition Axiom  $\forall t_0, t_1 \in T, \forall x_0 \in \Sigma$

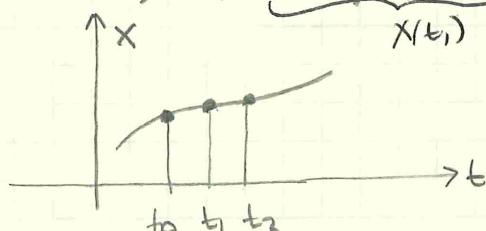
If  $u, \bar{u} \in U$  and  $u(t) = \bar{u}(t) \forall t \in [t_0, t_1] \cap T$

Then  $S(t_1, t_0, x_0, u) = S(t_1, t_0, x_0, \bar{u}) = x(t_1)$

i.e.,  $x(t_1)$  depends only on  $u[t_0, t_1]$

State Composition Axiom  $\forall t_0 \leq t_1 \leq t_2 \in T, \forall x_0 \in \Sigma, \forall u \in U$

$$S(t_2, t_1, S(t_1, t_0, x_0, u), u) = S(t_2, t_0, x_0, u) = x(t_2)$$



aka Semi-group Axiom

## LTV System

$$\textcircled{1} \begin{cases} \dot{x}(t) = A(t)x(t) + B(t)u(t) & x \in \mathbb{R}^n, u \in \mathbb{R}^k \\ y(t) = C(t)x(t) + D(t)u(t) & y \in \mathbb{R}^m \end{cases}$$

$A(\cdot)$ ,  $B(\cdot)$ ,  $C(\cdot)$ ,  $D(\cdot)$  matrices w/ elements piecewise cts  
 $n \times n$     $n \times k$     $m \times n$     $m \times k$  real-valued fcts defined on  $\mathbb{R}_+$

Refer to  $\textcircled{1}$  as the system representation  $R = [A(\cdot), B(\cdot), C(\cdot), D(\cdot)]$

Recall: a given model of a physical system may have many representations.  $\Rightarrow$  carefully distinguish between properties of the model & of its representation.

Theorem  $R = [A(\cdot), B(\cdot), C(\cdot), D(\cdot)]$  represents a linear dynamical system.

Proof (sketch)

Let  $U = \text{Set of piecewise cts functions } : \mathbb{R}_+ \rightarrow \mathbb{R}^k$   
 $X = " : \mathbb{R}_+ \rightarrow \mathbb{R}^n$   
 $S = \mathbb{R}^n$

From the Fund. Thm. of d.e.,  $R$  represents a dynamical system.

Show 3 Properties are satisfied:

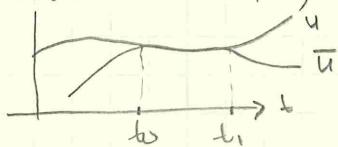
1) Unique Solution

$\forall u(\cdot)$ ,  $\textcircled{1}$  has a unique solution denoted by

$$S(t, t_0, x_0, u) \quad \left\{ \begin{array}{l} \frac{d}{dt} S(t, t_0, x_0, u) = A(t)S(t, t_0, x_0, u) + B(t)u \\ S(t_0, t_0, x_0, u) = x_0 \end{array} \right.$$

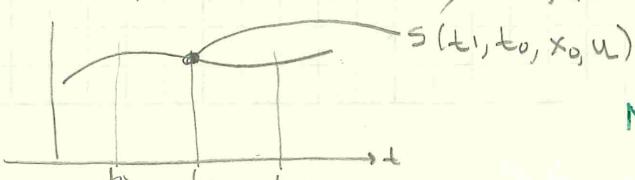
2) State Transition Axiom

$$u(t) = \bar{u}(t), t \in [t_0, t_1] \Rightarrow S(t_1, t_0, x_0, u) = S(t_1, t_0, x_0, \bar{u})$$



3) Semi-group Axiom

$$S(t_2, t_1, S(t_1, t_0, x_0, u_{[t_0, t_1]}), u_{[t_2, t_1]}) = S(t_2, t_0, x_0, u_{[t_0, t_2]})$$



Note  $u_{[t_0, t_2]} = u_{[t_0, t_1]} \cup u_{[t_1, t_2]}$

## Solution to Homogeneous LTV System

$$u(t) \equiv 0$$

D.E.  $\begin{cases} \dot{x}(t) = A(t)x(t) & x \in \mathbb{R}^n \\ y(t) = C(t)x(t) \\ x(t_0) = x_0 \end{cases}$

Find solution to D.E.

$$s(t, t_0, x_0, u) = \phi(t, t_0, x_0) \quad (\text{no } u(t))$$

which will satisfy the D.E.

$$\begin{cases} \frac{d}{dt} \phi(t, t_0, x_0) = A(t) \phi(t, t_0, x_0) \\ \phi(t_0, t_0, x_0) = x_0 \end{cases}$$

Note  $\phi$  is a linear map of  $x_0$

to the solution of the D.E.

$$\phi(t, t_0, \cdot) : \mathbb{R}^n \rightarrow \mathbb{R}^n$$

$$x_0 \rightarrow x(t) = \phi(t, t_0, x_0)$$

## Definition: Basis

A set of vectors  $\{v_1, v_2, \dots, v_n\}$  is a basis of a linear space  $V \subset \mathbb{R}^n$  iff

1)  $\{v_1, v_2, \dots, v_n\}$  is a linearly independent set

$$\text{so, } \sum_{i=1}^n \alpha_i v_i = 0 \Rightarrow \alpha_1 = \alpha_2 = \dots = \alpha_n = 0$$

2)  $\text{Span } \{v_1, v_2, \dots, v_n\} = V$

Span is the set of all linear combinations of elements in  $V$

Thus, if  $x \in V$  and  $\{v_1, v_2, \dots, v_n\}$  is a basis of  $V$

then  $\exists!$  set of scalars  $\{\alpha_1, \dots, \alpha_n\}$  s.t.

$$x = \sum_{i=1}^n \alpha_i v_i$$

$\alpha = \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{bmatrix}$  is the "representation" of  $x$

w.r.t. the basis  $\{v_1, v_2, \dots, v_n\}$

Remark: A basis is not unique.

Example  $\mathbb{R}_3[s]$  real polynomials of degree  $< 3$

Basis #1

$$\{v_1, v_2, v_3\} = \{1, s, s^2\}$$

Basis #2

$$\{\bar{v}_1, \bar{v}_2, \bar{v}_3\} = \{1, s+1, s^2+s\}$$

$$x = 5s^2 + 4s + 7 = [v_1 \ v_2 \ v_3] \begin{bmatrix} 7 \\ 4 \\ 5 \end{bmatrix} \leftarrow \alpha$$

$$= [\bar{v}_1 \ \bar{v}_2 \ \bar{v}_3] \begin{bmatrix} 8 \\ -1 \\ 5 \end{bmatrix} \leftarrow \bar{\alpha}$$

## Fundamental Matrix of Homogeneous LTV System

Let the Fundamental Matrix of the homogeneous LTV D.E. be defined as the matrix representation of the linear map  $\phi(t, t_0, x_0)$  w.r.t. the basis  $\{v_i\}$

$$x_0 = x(t_0) = \sum_{i=1}^n \alpha_i v_i \quad \text{initial state}$$

The Fundamental Matrix is

$$\underline{\mathcal{X}}(t) : x_0 \mapsto x(t) = \phi(t, t_0, x_0) = \begin{bmatrix} \phi_1(t, t_0, x_0) \\ \vdots \\ \phi_n(t, t_0, x_0) \end{bmatrix}$$

or

$$x(t) = \underline{\mathcal{X}}(t) x_0$$

$$= \begin{bmatrix} \phi(t_0, v_1) & \phi(t_0, v_2) & \dots & \phi(t_0, v_n) \end{bmatrix} \begin{bmatrix} \alpha_1 \\ \alpha_2 \\ \vdots \\ \alpha_n \end{bmatrix}$$

Properties of F.M.:

$$1) \quad \underline{\mathcal{X}}(t_0) = [v_1 \quad v_2 \quad \dots \quad v_n] \quad \begin{array}{l} \text{nonsingular} \\ \text{b.c. } \{v_i\} \text{ is a basis} \end{array}$$

$$2) \quad \begin{cases} \dot{\underline{\mathcal{X}}}(t) = A(t) \underline{\mathcal{X}}(t) \\ \underline{\mathcal{X}}(t_0) = \underline{\mathcal{X}}_0 \end{cases} \quad \text{satisfies D.E.}$$

Remarks  $\underline{\mathcal{X}}(t)$  is not unique

$\underline{\mathcal{X}}(t)$  is nonsingular.

## State Transition Matrix

Choose  $\{v_i\}$  as the standard basis  $\{e_i\}$ .

Then

$$\underline{\Phi}(t, t_0) \stackrel{\Delta}{=} \begin{bmatrix} \phi(t, t_0, e_1) & \phi(t, t_0, e_2) & \cdots & \phi(t, t_0, e_n) \end{bmatrix}$$

is called the State Transition Matrix.

Unique

## Properties of the STM

1) By definition, the unique solution to the d.e.

$$\text{is } \phi(t, t_0, x_0) = \boxed{x(t) = \underline{\Phi}(t, t_0) x_0}$$

$$\begin{cases} \dot{x} = A(t)x(t) \\ x(t_0) = x_0 \end{cases}$$

ZIR

$$y = C(t)x(t)$$

2)  $\underline{\Phi}(t, t_0)$  is the unique solution to

$$\begin{cases} \frac{d}{dt} \underline{\Phi}(t, t_0) = A(t) \underline{\Phi}(t, t_0) \\ \underline{\Phi}(t_0, t_0) = I \end{cases}$$

Matrix d.e.

3)  $\underline{\Phi}(t, t_0)$  is nonsingular  $\forall t$

$$4) \underline{\Phi}(t, t_0) = [\underline{\Phi}(t_0, t)]^{-1} \quad \text{invertible}$$

5) Semigroup Property

$$\underline{\Phi}(t, t_0) = \underline{\Phi}(t, t_1) \underline{\Phi}(t_1, t_0)$$

## How to Compute STM

1)  $i^{\text{th}}$  column of  $\underline{\Phi}(t, t_0)$  is the unique solution to

$$\begin{cases} \dot{x} = Ax \\ x(t_0) = e_i \end{cases}$$

$\{e_i\}$  standard basis

$$2) \underline{\Phi}(t, t_0) = \underline{\mathcal{X}}(t) \underline{\mathcal{X}}^{-1}(t_0)$$

$\underline{\mathcal{X}}$  = Fundamental Matrix

of  $\dot{x} = Ax$

Homogeneous System

Example (Chap 10 T)

$$\begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \dot{\mathbf{x}}(t) = \begin{bmatrix} 0 & 0 \\ t & 0 \end{bmatrix} \mathbf{x}(t) = A(t) \mathbf{x}(t)$$

For every initial state  $\mathbf{x}_i(t_0)$ ,  $i=1, 2$ ,  $\exists!$  solution  $\mathbf{x}_i(t)$ ,  $i=1, 2$ .

Arrange  $n$  solutions as columns of  $\mathbf{X}(t) = [x_1 | x_2]$

Then we will have  $\dot{\mathbf{X}}(t) = A(t) \mathbf{X}(t)$

$$\dot{x}_1(t) = 0 \quad \xrightarrow{\text{solution}} \text{for } t_0=0 \quad x_1(t) = x_1(0)$$

$$\dot{x}_2(t) = t x_1(t) \quad \rightarrow \quad x_2(t) = \int_0^t \tau x_1(\tau) d\tau + x_2(0) \\ = \frac{1}{2} t^2 x_1(0) + x_2(0)$$

Look at two initial conditions (arbitrary)

$$\mathbf{x}(0) = \begin{bmatrix} x_1(0) \\ x_2(0) \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \Rightarrow \mathbf{x}(t) = \begin{bmatrix} 1 \\ \frac{1}{2} t^2 \end{bmatrix}$$

$$\mathbf{x}(0) = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \Rightarrow \mathbf{x}(t) = \begin{bmatrix} 1 \\ \frac{1}{2} t^2 + 2 \end{bmatrix}$$

linearly independent

$$\Rightarrow \mathbf{X}(t) = \begin{bmatrix} 1 & 1 \\ \frac{1}{2} t^2 & \frac{1}{2} t^2 + 2 \end{bmatrix} \quad \text{A Fundamental Matrix (nonsingular)}$$

To find STM:

$$[\mathbf{X}(t)]^{-1} = \begin{bmatrix} \frac{1}{4} t^2 + 1 & -\frac{1}{2} \\ -\frac{1}{4} t^2 & \frac{1}{2} \end{bmatrix}$$

$$\Rightarrow \Phi(t, t_0) = \mathbf{X}(t) \mathbf{X}^{-1}(t_0) = \begin{bmatrix} 1 & 0 \\ \frac{1}{2} (t^2 - t_0^2) & 1 \end{bmatrix} \quad 2 \times 2$$

$$\Rightarrow \boxed{\mathbf{x}(t) = \Phi(t, t_0) \mathbf{x}_0} \quad \text{is the unique solution to}$$

$$\text{the d.e. } \begin{cases} \dot{\mathbf{x}} = A\mathbf{x} \\ \mathbf{x}(t_0) = \mathbf{x}_0 \end{cases}$$

for this  $\mathbf{x}_0$

## Solution to Nonhomogeneous Linear System

d.e.  $\begin{cases} \dot{x}(t) = A(t)x(t) + B(t)u(t) \\ x(t_0) = x_0 \\ y(t) = C(t)x(t) + D(t)u(t) \end{cases}$

The solution to this d.e. = ZIR + ZSR

"complete  
solution"

$$x(t) = \underline{\mathbb{E}}(t, t_0)x_0 + \int_{t_0}^t \underline{\mathbb{E}}(t, \tau)B(\tau)u(\tau)d\tau$$

ZIR                                    ZSR

To see this, can check

$$x(t_0) = x_0 \quad \checkmark$$

$$\begin{aligned} \dot{x}(t) &= \frac{d}{dt} \underline{\mathbb{E}}(t, t_0)x_0 + \frac{d}{dt} \int_{t_0}^t \underline{\mathbb{E}}(t, \tau)B(\tau)u(\tau)d\tau \\ &\quad \downarrow \frac{d}{dt}\underline{\mathbb{E}} = A\underline{\mathbb{E}} \qquad \frac{d}{dt} \int f(t, s)ds = f(t, t) \\ &= A(t)\underline{\mathbb{E}}(t, t_0)x_0 + \underline{\mathbb{E}}(t, t)B(t)u(t) \\ &\quad + \int_{t_0}^t \frac{d}{dt} \underline{\mathbb{E}}(t, \tau)B(\tau)u(\tau)d\tau \\ &= A(t)\underline{\mathbb{E}}(t, t_0)x_0 + B(t)u(t) + A(t) \int_{t_0}^t \underline{\mathbb{E}}(t, \tau)B(\tau)u(\tau)d\tau \\ &= A(t) \left[ \underline{\mathbb{E}}(t, t_0)x_0 + \int_{t_0}^t \underline{\mathbb{E}}(t, \tau)B(\tau)u(\tau)d\tau \right] + B(t)u(t) \\ &= A(t)x(t) + B(t)u(t) \quad \checkmark \text{ unique} \end{aligned}$$

## Connection Between I/O & State-Space Descriptions

$$\begin{aligned} \text{ZSR } y(t) &= C(t) \int_{t_0}^t \underline{\mathbb{E}}(t, \tau)B(\tau)u(\tau)d\tau + D(t)u(t) \\ &= \int_{t_0}^t [C(t)\underline{\mathbb{E}}(t, \tau)B(\tau) + D(t-\tau)]u(\tau)d\tau \\ &= \int_{t_0}^t G(t, \tau)u(\tau)d\tau \quad \text{Impulse Response Matrix } G \end{aligned}$$

$$\begin{aligned} \Rightarrow G(t, \tau) &= C(t)\underline{\mathbb{E}}(t, \tau)B(\tau) + D(t)S(t-\tau) \\ &= C(t)\underline{\mathbb{X}}(t)\underline{\mathbb{X}}^{-1}(\tau)B(\tau) + D(t)S(t-\tau) \end{aligned}$$

How to solve d.e. (Chen p. 110)

Solving for

$$x(t) = \underline{\Phi}(t, t_0) x_0 + \int_{t_0}^t \underline{\Phi}(t, \tau) B(\tau) u(\tau) d\tau$$

$$\text{and } y(t) = C(t) x(t) + D(t) u(t)$$

rely on solving

$$\dot{x}(t) = A(t) x(t) \quad \text{or} \quad \frac{d}{dt} \underline{\Phi}(t, t_0) = A(t) \underline{\Phi}(t, t_0)$$

→ very difficult!

Special Cases:

1) If  $A(t)$  is triangular

$$\dot{x}(t) = \begin{bmatrix} a_{11}(t) & 0 \\ a_{21}(t) & a_{22}(t) \end{bmatrix} x$$

$$\text{solve } \dot{x}_1(t) = a_{11}(t) x_1(t)$$

saw this in  
above example

$$\& \text{ substitute into } \dot{x}_2(t) = a_{22}(t) x_2(t) + a_{21}(t) x_1(t)$$

2) If  $A(t)$  is diagonal (or constant) and has the commutative property

$$A(t) \left( \int_{t_0}^t A(\tau) d\tau \right) = \left( \int_{t_0}^t A(\tau) d\tau \right) A(t) \quad \forall t, t_0$$

Then the solution to  $\begin{cases} \frac{d}{dt} \underline{\Phi}(t, t_0) = A(t) \underline{\Phi}(t, t_0) \\ \underline{\Phi}(t_0, t_0) = I \end{cases}$

$$\text{is } \underline{\Phi}(t, t_0) = \exp \int_{t_0}^t A(\tau) d\tau$$

Note - We want  $A$  to be in a "nice" form

- will see how to do this

Outline

Solution to LTI System

State Transition Matrix for LTI

Matrix Exponential & Properties

Cayley - Hamilton Theorem

Characteristic Polynomial

Discriminants

## Lecture 6

LTI System Representation

Consider the LTI System:

$$\textcircled{1} \quad \begin{cases} \dot{x}(t) = Ax(t) + Bu(t) & x \in \mathbb{R}^n \quad u \in \mathbb{R} \\ x(t_0) = x_0 & x_0 \in \mathbb{R}^n \\ y(t) = Cx(t) + Du(t) & y \in \mathbb{R}^n \end{cases}$$

System representation  $R = [A, B, C, D]$

Solutions to Homogeneous LTI System  $u(t) \equiv 0$

$$\textcircled{2} \quad \begin{cases} \dot{x}(t) = Ax(t) & t \geq 0 \\ x(t_0) = x_0 \\ y(t) = Cx(t) \end{cases}$$

The unique solution to  $\textcircled{2}$  is given by the  
State Transition Matrix:

$$x(t) = \Phi(t, t_0) x_0$$

solution

where

$$\begin{aligned} \Phi(t, t_0) &= \sum_{k=0}^{\infty} \frac{(t-t_0)^k}{k!} A^k \\ &= e^{A(t-t_0)} \end{aligned}$$

$$\text{Thus, } x(t) = e^{A(t-t_0)} x_0$$

solution to  $\textcircled{2}$

Proof

The STM satisfies the d.e.  $\textcircled{2}$

$$\begin{cases} \frac{d}{dt} \Phi(t, t_0) = A \Phi(t, t_0) & t \geq 0 \\ \Phi(t_0, t_0) = I \end{cases}$$

Integrating, we get

$$\Phi(t, t_0) = I + \int_{t_0}^t A \Phi(\tau, t_0) d\tau$$

Use Picard Iteration:

$$\mathbb{E}_0(t, t_0) = I$$

$$\mathbb{E}_1(t, t_0) = I + \int_{t_0}^t A \mathbb{E}_0(\tau, t_0) d\tau = I + A(t - t_0)$$

$$\mathbb{E}_2(t, t_0) = I + \int_{t_0}^t A \mathbb{E}_1(\tau, t_0) d\tau$$

$$= I + \int_{t_0}^t A(I + A(\tau - t_0)) d\tau$$

$$= I + At + \frac{A^2 t^2}{2!}$$

⋮

$$\mathbb{E}(t, t_0) = I + At + \frac{A^2 t^2}{2!} + \frac{A^3 t^3}{3!} + \dots = \sum_{k=0}^{\infty} \frac{(t-t_0)^k}{k!} A^k$$

$$= e^{A(t-t_0)} \quad \text{by definition (Taylor expansion series of exponential)}$$

This is called the Matrix Exponential

$$e^M := \sum_{k=0}^{\infty} \frac{1}{k!} M^k$$

### Solution to Nonhomogeneous LTI System

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) & x \in \mathbb{R}^n, u \in \mathbb{R}^m \\ x(t_0) = x_0 & x_0 \in \mathbb{R}^n \\ y(t) = Cx(t) + Du(t) & y \in \mathbb{R}^m \end{cases}$$

As in LTV case, we have the solution

$$x(t) = \mathbb{E}(t, t_0)x_0 + \int_{t_0}^t \mathbb{E}(t, \tau)B u(\tau) d\tau$$

$$\Rightarrow \begin{cases} x(t) = e^{A(t-t_0)}x_0 + \int_{t_0}^t e^{A(t-\tau)}B u(\tau) d\tau & \text{unique solution} \\ y(t) = Cx(t) + Du(t) & \text{response} \end{cases}$$

\* Note  $e^x = \sum_{k=0}^{\infty} \frac{x^k}{k!} = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots$  Taylor Series expansion

## Properties of the Matrix Exponential (STM)

$$e^{At} = \sum_{k=0}^{\infty} \frac{1}{k!} (At)^k = I + At + A^2 \frac{t^2}{2!} + A^3 \frac{t^3}{3!} + \dots$$

1)  $e^{At}$  is the unique solution to

$$\begin{cases} \frac{de^{At}}{dt} = A e^{At} & t \geq 0 \\ e^{A \cdot 0} = I \end{cases} \quad \text{b/c it is the STM}$$

2)  $i^{th}$  column of  $e^{At}$  is unique solution to

$$\begin{cases} \dot{x}(t) = Ax(t) & t \geq 0 \\ x(0) = e_i \end{cases}$$

$e_i = i^{th}$  vector of standard basis for  $\mathbb{R}^n$

$$\begin{bmatrix} 0 \\ 0 \\ \vdots \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

3)  $e^{At} e^{A\tau} = e^{A(t+\tau)}$   $\forall t, \tau \in \mathbb{R}$

4)  $e^{At}$  is nonsingular  $\forall t \in \mathbb{R}$

$$(e^{At})^{-1} = e^{-At}$$

5) If  $AB = BA$ ,  $e^{A+B} = e^A e^B$

6)  $e^{At} = \mathcal{L}^{-1} [ (sI - A)^{-1} ]$  useful formula

7)  $A \cdot e^{At} = e^{At} \cdot A$

8)  $e^{At} = \sum_{i=0}^{n-1} \alpha_i(t) A^i$   $\forall t \in \mathbb{R}$

$\alpha_i$ ,  $i=0, \dots, n-1$  scalar functions

e.g.,  $n=2$ ,  $e^{At} = \alpha_0(t) I + \alpha_1(t) A$

Cayley - Hamilton Theorem

Used to prove property #8 above

Useful to compute  $e^{At}$ , hence STM, hence solution  $x(t)$ .

Theorem: For every  $n \times n$  matrix  $A$ ,

$$\Delta(A) = A^n + \alpha_1 A^{n-1} + \dots + \alpha_n I = 0_{n \times n}$$

where

$$\Delta(s) = s^n + \alpha_1 s^{n-1} + \dots + \alpha_n = \det(SI - A)$$

is the characteristic polynomial of  $A$ .

Proof (by construction)

$$\begin{cases} \frac{d}{dt} \Phi(t) = A \Phi(t) \\ \Phi(0) = I \end{cases}$$

STM satisfies d.e.

Let  $t_0 = 0$  b/c time-invariant  
Response only depends on  $t - t_0$

Take Laplace Transform:

$$s \hat{\Phi}(s) - I = A \hat{\Phi}(s)$$

$$\hat{\Phi}(s) = (SI - A)^{-1}$$

$$= \frac{1}{\det(SI - A)} [\text{adj}(SI - A)]'$$

$\text{o}(n)$   
polynomial       $\text{o}(n-1)$  or less  
polynomial

adjoint matrix of  $(SI - A)$   
(cofactors)

$$(SI - A)^{-1} = \frac{1}{s^n + \alpha_1 s^{n-1} + \dots + \alpha_n} [s^{n-1} \beta_0 + s^{n-2} \beta_1 + \dots + s \beta_{n-2} + \beta_{n-1}]$$

$$(s^n + \alpha_1 s^{n-1} + \dots + \alpha_n) I = [s^{n-1} \beta_0 + s^{n-2} \beta_1 + \dots + \beta_{n-1}] (SI - A)$$

$$\stackrel{\uparrow \Delta(s)}{=} s^n \beta_0 + s^{n-1} (\beta_1 - \beta_0 A) + \dots + \beta_{n-1} A$$

Assuming we know  $\Delta(s)$  characteristic polynomial,  
we can calculate  $\{\beta_i\}$  by equating like powers

$$\left\{ \begin{array}{l} \beta_0 = I \\ \beta_1 = \beta_0 A + \alpha_1 I \\ \beta_2 = \beta_1 A + \alpha_2 I \\ \vdots \\ \beta_{n-1} = \beta_{n-2} A + \alpha_{n-1} I \\ \Omega_{n \times n} = \beta_{n-1} A + \alpha_n I \end{array} \right.$$

Then substitute

$$\begin{aligned} \Omega_{n \times n} &= \beta_{n-1} A + \alpha_n I = (\beta_{n-2} A + \alpha_{n-1} I) A + \alpha_n I \\ &= \beta_{n-2} A^2 + \alpha_{n-1} A + \alpha_n I \\ &\vdots \\ &= A^n + \alpha_1 A^{n-1} + \dots + \alpha_n I \\ &= \Delta(A) \end{aligned}$$

Example (Chen p. 89)

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

Find  $x(t)$  solution for  $t_0 = 0$

$$x(t) = e^{At} x(0) + \int_0^t e^{A(t-\tau)} B u(\tau) d\tau$$

$$e^{At} = \mathcal{L}^{-1} \left[ (sI - A)^{-1} \right]$$

$$\begin{aligned} (sI - A)^{-1} &= \left( \begin{bmatrix} s & 0 \\ 0 & s \end{bmatrix} - \begin{bmatrix} 0 & -1 \\ 1 & -2 \end{bmatrix} \right)^{-1} \\ &= \begin{bmatrix} s & 1 \\ -1 & s+2 \end{bmatrix}^{-1} = \frac{1}{s(s+2)+1} \begin{bmatrix} s+2 & -1 \\ 1 & s \end{bmatrix} \\ &= \frac{1}{s^2+2s+1} \begin{bmatrix} s+2 & -1 \\ 1 & s \end{bmatrix} = \frac{1}{(s+1)^2} \begin{bmatrix} s+2 & -1 \\ 1 & s \end{bmatrix} \\ &= \begin{bmatrix} \frac{s+2}{(s+1)^2} & \frac{-1}{(s+1)^2} \\ \frac{1}{(s+1)^2} & \frac{s}{(s+1)^2} \end{bmatrix} \end{aligned}$$

$$\mathcal{L}^{-1} \left[ (sI - A)^{-1} \right] = \mathcal{L}^{-1} \left[ \begin{bmatrix} \frac{s+2}{(s+1)^2} & \frac{-1}{(s+1)^2} \\ \frac{1}{(s+1)^2} & \frac{s}{(s+1)^2} \end{bmatrix} \right] = \begin{bmatrix} (1+t)e^{-t} & -te^{-t} \\ te^{-t} & (1-t)e^{-t} \end{bmatrix}$$

Then

$$\begin{aligned} x(t) &= \begin{bmatrix} (1+t)e^{-t} & -te^{-t} \\ te^{-t} & (1-t)e^{-t} \end{bmatrix} x(0) + \int_0^t \begin{bmatrix} (1+t-\tau)e^{-(t-\tau)} & -(t-\tau)e^{-(t-\tau)} \\ (t-\tau)e^{-(t-\tau)} & (1-(t-\tau))e^{-(t-\tau)} \end{bmatrix} \begin{bmatrix} 0 \\ 1 \end{bmatrix} u(\tau) d\tau \\ &= \begin{bmatrix} (1+t)e^{-t} & -te^{-t} \\ te^{-t} & (1-t)e^{-t} \end{bmatrix} x(0) + \begin{bmatrix} - \int_0^t (t-\tau)e^{-(t-\tau)} u(\tau) d\tau \\ \int_0^t (1-(t-\tau))e^{-(t-\tau)} u(\tau) d\tau \end{bmatrix} \end{aligned}$$

Note

$$\mathcal{L}^{-1} \left[ \frac{1}{s} \right] = 1$$

$$\mathcal{L}^{-1} \left[ \frac{n!}{s^{n+1}} \right] = t^n$$

see Handout 3

$$\mathcal{L}^{-1} \left[ \frac{1}{s^2} \right] = t$$

$$\mathcal{L}^{-1} \left[ \frac{n!}{(s-a)^{n+1}} \right] = t^n e^{at}$$

$$\mathcal{L}^{-1} \left[ \frac{1}{s-a} \right] = e^{at}$$

## SYSTEM

 $\rightarrow$ 

Solution

Summary

$$\text{LTIV} \left\{ \begin{array}{l} \dot{x} = A(t)x + B(t)u \\ y = C(t)x + D(t)u \\ x(t_0) = x_0 \end{array} \right. \xrightarrow{\int} \left\{ \begin{array}{l} x(t) = \mathbb{E}(t, t_0)x_0 + \int_{t_0}^t \mathbb{E}(t, \tau)B(\tau)u(\tau)d\tau \\ y(t) = C(t)\mathbb{E}(t, t_0)x_0 + C(t)\int_{t_0}^t \mathbb{E}(t, \tau)B(\tau)u(\tau)d\tau + D(t)u(t) \end{array} \right.$$

$$x(t_0) = x_0$$

$$y(t) = \int_{t_0}^t G(t, \tau)u(\tau)d\tau$$

convolution

$$\xrightarrow{\text{realization}} G(t, \tau) = C(t)\mathbb{X}^{-1}(\tau)B(\tau) + D(t)\mathcal{S}(t - \tau)$$

$$\xrightarrow{\text{discretize}} y(k\Delta) = \sum_{m=0}^{k\Delta} G(k\Delta, m\Delta) u(m\Delta) \Delta t$$

step size

compute numerically

$$\text{LTI} \left\{ \begin{array}{l} \dot{x} = Ax + Bu \\ y = Cx + Du \\ x(t_0) = x_0 = 0 \end{array} \right.$$

$$\xrightarrow{\mathcal{L}} \left\{ \begin{array}{l} \hat{x}(s) = (sI - A)^{-1}B\hat{u}(s) \\ \hat{y}(s) = C(sI - A)^{-1}B\hat{u}(s) + Du(s) \end{array} \right.$$

$$\xrightarrow{\mathcal{L}^{-1}} y(t) = \mathcal{F}^{-1}[\hat{y}(s)]$$

Solve with  
Matlabbut hard if  
multiple poles

$$\hat{y}(s) = \hat{G}(s)\hat{u}(s) \rightarrow y(t) = \mathcal{F}^{-1}[\hat{G}(s)\hat{u}(s)]$$

$$\hat{G}(s) = \frac{\text{num}(s)}{\text{den}(s)}$$

STM

$$\hat{G}(s) = \frac{\text{num}(s)}{\text{den}(s)} \xrightarrow{\text{realization}} \text{LTI} \xrightarrow{\text{state-space model}} \text{state-space model}$$

realization

$$\hat{G}(s) = C(sI - A)^{-1}B$$