

## Lecture 3: Time Response

⇒ Let us consider a LTI first order system:

$$\begin{aligned} \dot{x}(t) &= ax(t) + bu(t) \\ y(t) &= cx(t) + du(t) \end{aligned}$$

here,  $x(t), y(t), u(t) \in \mathbb{R}$   
 $\& a, b, c, d \in \mathbb{R}$

⇒ Let us denote the above system by  $\Sigma$ :

$$\text{So } \Sigma: u \rightarrow y$$

⇒ Time Response:  ~~$y(t)$  given  $u(t)$~~   
 $\text{as } x(0) \neq t \geq 0$ .

⇒ Let us break  $u$  into two parts:

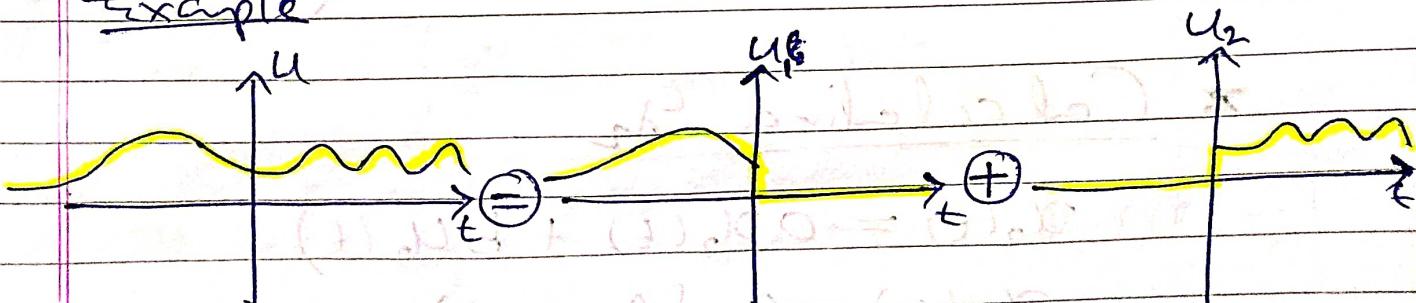
$$u = (P_0 u) + (u - P_0 u)$$

$$u = (P_0 u) + (u - P_0 u)$$

$$\Rightarrow u = u_1 + u_2$$

$\left. \begin{array}{l} P_T \text{ is an operator} \\ \text{which truncate } u \\ \text{after time } T \end{array} \right\}$

Example



$$\Rightarrow \sum u = \sum (u_1 + u_2)$$

$$\Rightarrow y = \sum u_1 + \sum u_2 \quad \left\{ \begin{array}{l} \text{As the System is} \\ \text{Linear} \end{array} \right.$$

$$y = y_1 + y_2$$

### \* Calculating $y_1$

$$\dot{x}_1(t) = ax_1(t) + bu_1(t)$$

$$\forall t \geq 0$$

$$\dot{x}_1(t) = ax_1(t)$$

$$\frac{dx_1(t)}{dt} = a x_1(t) \Rightarrow \int \frac{dx_1(t)}{x_1(t)} = \int a dt$$

$$\Rightarrow x_1(t) = x_1(0) e^{at} \quad \left\{ \begin{array}{l} \text{As } \Sigma \text{ is causal} \\ \text{i.e. } \end{array} \right.$$

$$\Rightarrow x_1(t) = \cancel{x_1(0)} e^{at}$$

$$\Rightarrow y_1(t) = c x_1(0) e^{at}$$

$$\left. \begin{aligned} & \sum P_T u(t) \\ & = P_T (\sum u(t)) \end{aligned} \right\} \forall t \leq T$$

### \* Calculating $y_2$

$$\dot{x}_2(t) = ax_2(t) + bu_2(t)$$

$$x_2(0) = 0 \quad \left\{ \begin{array}{l} \text{Assuming System is Stable} \end{array} \right.$$

$\Rightarrow$  Multiplying both sides by  $e^{-at}$ .

$$\Rightarrow e^{-at} \dot{x}_2(t) dt = e^{-at} ax_2(t) dt + e^{-at} bu_2(t) dt$$

$$\Rightarrow \int_0^t e^{-a\tau} \dot{x}_2(\tau) d\tau = \int_0^t e^{-a\tau} ax_2(\tau) d\tau + \int_0^t e^{-a\tau} bu_2(\tau) d\tau$$

$$\Rightarrow \text{LHS} = e^{-a\tau} x_2(\tau) \Big|_0^t - \int_0^t -ae^{-a\tau} x_2(\tau) d\tau$$

$\left. \begin{array}{l} \text{Integration by parts} \\ \{ \int u(\tau) v'(\tau) d\tau \\ = u(\tau) v(\tau) \Big|_0^t - \int u'(\tau) v(\tau) d\tau \end{array} \right\}$

$$\Rightarrow \text{LHS} = e^{-at} x_2(t) + a \int_0^t e^{-a\tau} \dot{x}_2(\tau) d\tau \quad \text{--- (1)}$$

$$\Rightarrow \text{RHS} = a \int_0^t e^{-a\tau} \dot{x}_2(\tau) d\tau + \int_0^t e^{-a\tau} bu_2(\tau) d\tau \quad \text{--- (2)}$$

$$\Rightarrow \dot{x}_2(t) = \int_0^t e^{a(t-\tau)} bu_2(\tau) d\tau, \quad \forall t \geq 0$$

$$\Rightarrow x_2(t) = \int_0^t e^{a(t-\tau)} bu_2(\tau) d\tau$$

$$\Rightarrow y_2(t) = c \int_0^t e^{a(t-\tau)} bu_2(\tau) d\tau + du(t)$$

## \* Calculating $y$

$$\Rightarrow y = y_1 + y_2$$

$$\Rightarrow y(t) = c\alpha(0)e^{at} + c \int_0^t e^{a(t-\tau)} bu(\tau)d\tau + du(t)$$

Let  $\phi(t) = e^{at}$

so  $e^{a(t-\tau)} = \phi(t-\tau)$

$$\Rightarrow y(t) = c\phi(t)x(0) + c \int_0^t \phi(t-\tau)bu(\tau)d\tau + du(t)$$

## \* The matrix exponential

$\Rightarrow$  For a matrix  $A$  Let

$$e^{At} = I + At + \frac{1}{2}(At)^2 + \dots + \frac{1}{m!}(At)^m + \dots$$

{ This overloading of the exponential operation is inspired by Taylor expansion of  $e^{at}$  where  $a \in \mathbb{R}$  }

$$e^{at} = 1 + at + \frac{1}{2}(at)^2 + \dots + \frac{1}{m!}(at)^m + \dots$$

$$\Rightarrow \frac{de^{At}}{dt} = A + \frac{1}{2!} 2A^2 t + \frac{1}{3!} 3A^3 t^2 + \dots$$

$$\Rightarrow A \left( I + At + \frac{1}{2!} (At)^2 + \dots \right)$$

$$\boxed{\frac{de^{At}}{dt} = A e^{At}} \quad \left. \begin{array}{l} \text{Similar to } e^{at} \forall a \in \mathbb{R} \end{array} \right\}$$

$\Rightarrow$  For  $n^{\text{th}}$  order LTI System ~~will be~~, will be

$$\dot{x}_1(t) = Ax_1(t)$$

$\Rightarrow$  ~~the~~  $x_1(t) = e^{At} K$  is solution of the above ~~the~~ differential equation.

$$x_1(0) = e^{A0} K = K$$

$$\left. \begin{array}{l} K \in \mathbb{R}^{n \times 1} \end{array} \right\}$$

$\Rightarrow$  Verifying the Solution.

$$\cancel{\frac{d}{dt} e^{At} x_1(0)} \quad \frac{d}{dt} e^{At} x_1(0) = A e^{At} x_1(0)$$

$$\Rightarrow \frac{d}{dt} x_1(t) = A x_1(t)$$

$$\text{So } \boxed{x_1(t) = e^{At} x(0) = \Phi(t) x(0)}$$

$$\text{So } \boxed{y_1(t) = C e^{At} x(0)}$$

$\Rightarrow$  Similar arguments can be followed for  $y_2(t)$  also of first order system.

$$\Rightarrow \dot{X}_2(t) = AX_2(t) + BU_2(t)$$

$$\Rightarrow e^{-At} \dot{X}_2(t) = e^{-At} AX_2(t) + e^{-At} BU_2(t)$$

$$\Rightarrow \int_0^t e^{-Ar} \dot{X}_2(r) dr = \int_0^t e^{-Ar} AX_2(r) dr + \int_0^t e^{-Ar} BU_2(r) dr$$

$$\Rightarrow LHS = e^{-At} X_2(t) \Big|_0^t + \int_0^t e^{-Ar} AX_2(r) dr$$

$$\Rightarrow RHS = \int_0^t e^{-Ar} AX_2(r) dr + \int_0^t e^{-Ar} BU_2(r) dr$$

$$\Rightarrow e^{-At} X_2(t) = \int_0^t e^{-Ar} BU_2(r) dr$$

$$\Rightarrow e^{At} e^{-At} X_2(t) = e^{At} \int_0^t e^{-Ar} BU_2(r) dr$$

$$\left\{ e^A e^B = e^{A+B} \text{ iff } AB = BA \right\}$$

$$\Rightarrow X_2(t) = \int_0^t e^{A(t-\tau)} B U_2(\tau) d\tau$$

$$\Rightarrow Y_2(t) = C \int_0^t e^{A(t-\tau)} B U(\tau) d\tau + D U(t)$$

$$\Rightarrow \text{So } Y(t) = Y_1(t) + Y_2(t)$$

$$Y(t) = C e^{At} X(0) + C \int_0^t e^{A(t-\tau)} B U(\tau) d\tau + D U(t)$$

$$Y(t) = C \phi(t) X(0) + C \int_0^t \phi(t-\tau) B U(\tau) d\tau + D U(t)$$

## ★ How to compute a matrix exponential

1. Bonita face: Compute many terms of Taylor expansion.

2. Find a generalization of the system (i.e. a special choice of the state vector  $\mathbf{x}$ ) such that the matrix  $A$  is either:

→ Diagonal  $\begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_m \end{bmatrix}$

→ Jordan form

## \* Jordan Form

A Square matrix is said to be in Jordan form if it is block diagonal where each block is a Jordan block.

$\Rightarrow$  A Jordan block of size  $K \times K$  is a matrix of form:

$$\begin{pmatrix} \lambda & 1 & & \\ & \ddots & & \\ & & \lambda & 1 \\ & 0 & & \ddots & 1 \\ & & & & \lambda \end{pmatrix}$$

## \* Qualitative behavior of a LTI System

1) **Lyapunov stability:** A system is called Lyapunov stable if, for any bounded initial condition, and zero input, the state remains bounded.

$$\|x_0\| < \epsilon \text{ and } u=0 \Rightarrow \|x(t)\| < \delta \quad t \geq 0$$

2) **Asymptotic stability:** A system is called asymptotically stable if, for any bounded initial condition and zero input, the state converges to zero.

$$\|x_0\| < \epsilon \text{ & } u=0 \Rightarrow \lim_{t \rightarrow +\infty} \|x(t)\| = 0$$

### 3) Bounded-Input Bounded-Output Stability (BIBO)

A System is called BIBO-stable if, for any bounded input, the output remains bounded

$$\|u(t)\| < \infty \quad \forall t \geq 0 \quad \& \quad x(0) = 0$$



$$\|y(t)\| < \infty \quad \forall t \geq 0$$

4) Unstable: A system is called unstable if it is not stable by any of the above criteria.

\* Stability analysis of a general Linear time-invariant System

⇒ For Lyapunov and Asymptotic stability check  
 $u=0$ .

$$\begin{array}{l} \boxed{\dot{x} = Ax} \\ \quad \quad \quad \textcircled{a} \\ \boxed{y = cx} \\ \quad \quad \quad \textcircled{b} \end{array}$$

①  $A$  has distinct Eigen values

⇒ Let  $v_1, v_2, \dots, v_m$  be the eigen vectors corresponding to the eigen values  $\lambda_1, \lambda_2, \dots, \lambda_m$  of  $A$ .

⇒ Let nxn matrix  $T = [v_1, v_2, \dots, v_m] \quad \text{--- } \textcircled{1}$

$$\Rightarrow \cancel{AT} = T\Lambda$$

where,  $\Lambda = \begin{bmatrix} \lambda_1 \\ \lambda_2 \\ \vdots \\ \lambda_n \end{bmatrix}$

$$\Rightarrow A = T\Lambda T^{-1} \text{ or } \Lambda = T^{-1}AT$$

$$\Rightarrow \text{Let } X = T\tilde{X} \quad \text{--- (2)}$$

~~So~~  $\Rightarrow$  Substituting eq(2) in eq(1) we get:

$$\Rightarrow T\tilde{X} = AT\tilde{X}$$

$$\Rightarrow \tilde{X} = T^{-1}AT\tilde{X} = \Lambda\tilde{X}$$

$$\Rightarrow \tilde{X} = \Lambda\tilde{X}$$

$\Rightarrow$  Solution of the above equation is

$$\boxed{\tilde{X} = e^{\Lambda t}\tilde{X}(0)} \quad \text{--- (3)}$$

$\Rightarrow$  Substituting eq(2) in eq(3) to eliminate  $\tilde{X}$

$$\Rightarrow T^{-1}X = e^{\Lambda t}T^{-1}X(0)$$

$$\Rightarrow \boxed{X = T e^{\Lambda t} T^{-1} X(0)}$$

$$\Rightarrow \text{Let } T^{-1} = \begin{bmatrix} w_1^T \\ w_2^T \\ \vdots \\ w_n^T \end{bmatrix}$$

$\Rightarrow \omega_1^T, \omega_2^T, \dots, \omega_n^T$  are the left eigen vectors corresponding to  $\lambda_1, \lambda_2, \dots, \lambda_n$ .

$$\boxed{\begin{aligned} & AT = T \Lambda \\ & \Rightarrow T^{-1}AT = \Lambda \\ & \Rightarrow T^{-1}A = \Lambda T^{-1} \\ & \Rightarrow \begin{bmatrix} \omega_1^T \\ \vdots \\ \omega_n^T \end{bmatrix} A = \begin{bmatrix} \lambda_1 & & \\ & \ddots & \\ & & \lambda_n \end{bmatrix} \begin{bmatrix} \omega_1^T \\ \vdots \\ \omega_n^T \end{bmatrix} \\ & \Rightarrow \omega_1^T A = \lambda_1 \omega_1^T \\ & \omega_n^T A = \lambda_n \omega_n^T \end{aligned}}$$

$$\begin{aligned} & \Rightarrow X(t) = [V_1, V_2, \dots, V_n] \begin{bmatrix} e^{\lambda_1 t} & & \\ & e^{\lambda_2 t} & \\ & & \ddots & \\ & & & e^{\lambda_n t} \end{bmatrix} \begin{bmatrix} \omega_1^T \\ \omega_2^T \\ \vdots \\ \omega_n^T \end{bmatrix} X(0) \\ & = [V_1, V_2, \dots, V_n] \begin{bmatrix} e^{\lambda_1 t} \omega_1^T \\ e^{\lambda_2 t} \omega_2^T \\ \vdots \\ e^{\lambda_n t} \omega_n^T \end{bmatrix} X(0) \\ & = (V_1 e^{\lambda_1 t} \omega_1^T + V_2 e^{\lambda_2 t} \omega_2^T + \dots + V_n e^{\lambda_n t} \omega_n^T) X(0) \end{aligned}$$

$$\Rightarrow X(t) = \left( e^{\lambda_1 t} v_1 w_1^T + e^{\lambda_2 t} v_2 w_2^T + \dots + e^{\lambda_n t} v_n w_n^T \right) X(0)$$

$\Rightarrow$  If all the eigen values are stable, then all the terms of  $X(t)$ , are linear combination of form  $e^{\lambda_i t}$  ( $\forall t > 0$  and  $i = \{1, 2, \dots, n\}$ ).

$\Rightarrow$  As  $Y$  is just a linear combination of  $X$ , it also has the same form.

$\Rightarrow$  So for Lyapunov stability:

$$\lambda_i \leq 0 \quad \forall i \in \{1, 2, \dots, n\}$$

$\Rightarrow$  For Asymptotic stability:

$$\lambda_i < 0 \quad \forall i \in \{1, 2, \dots, n\}$$

$\Rightarrow$  If some of the eigen value are complex conjugate pairs:

$\Rightarrow$  Let  $\lambda$  and  $\bar{\lambda}$  be two eigen values.



$\Rightarrow$  Let  $V$  and  $W^T$  be the right and left eigen vector corresponding to  $\lambda$ .

$$AV = \lambda V \Rightarrow A\bar{V} = \bar{\lambda}\bar{V}$$

$$\omega^T A = \lambda \omega^T \Rightarrow \bar{\omega}^T A = \bar{\lambda} \bar{\omega}^T$$

$\Rightarrow$  So  $\bar{V}$  and  $\bar{\omega}^T$  are the right and left eigenvectors corresponding to  $\bar{\lambda}$ .

$\Rightarrow e^{\sigma t} V \omega^T + e^{\bar{\lambda} t} \bar{V} \bar{\omega}^T$  is the term contributed in the solution by  $\lambda$  and  $\bar{\lambda}$ .

$$\Rightarrow \text{Let, } \lambda = \sigma + j\omega \Rightarrow \bar{\lambda} = \sigma - j\omega$$

$$, V \omega^T = \cancel{E + jF} \Rightarrow \bar{V} \bar{\omega}^T = \cancel{E - jF}$$

$$\Rightarrow e^{(\sigma+j\omega)t} (E + jF) + e^{(\sigma-j\omega)t} (E - jF)$$

$$= e^{\sigma t} e^{j\omega t} (E + jF) + e^{\sigma t} e^{-j\omega t} (E - jF)$$

$$\Rightarrow e^{\sigma t} \left( (\cos(\omega t) + j \sin(\omega t)) (E + jF) + ((\cos(\omega t) - j \sin(\omega t)) (E - jF) \right)$$

$$\Rightarrow e^{\sigma t} \left( \cos(\omega t) E - F \sin(\omega t) + E \cos(\omega t) - F \sin(\omega t) + j \sin(\omega t) F + j \cos(\omega t) F - j \sin(\omega t) E - j \cos(\omega t) F \right)$$

$$\Rightarrow e^{\sigma t} (2 \cos(\omega t) E - 2 \sin(\omega t) F)$$

$\Rightarrow$  So this contains sinusoidal terms with exponential terms.

$\Rightarrow$  So, if complex conjugate eigen values are available, it will introduce terms of type  $e^{\sigma t} \sin(\omega t + \phi)$  in the solution ( $y$ ).

$\Rightarrow$  For Lyapunov stability:

$$\operatorname{Re}(\lambda_i) \leq 0 \quad \forall i \in \{1, 2, \dots, n\}$$

$\Rightarrow$  For Asymptotic stability:

$$\operatorname{Re}(\lambda_i) < 0 \quad \forall i \in \{1, 2, \dots, n\}$$

② A has repeated Eigenvalues

$\Rightarrow$  If the matrix A is not diagonalizable (i.e. algebraic multiplicity of eigenvalues is not equal to its geometric multiplicity) then it can be converted to Jordan form.

$\Rightarrow$  Let T be the similarity transform that will convert matrix A into Jordan form

$$T^{-1}AT = J \text{ where, } J = \begin{bmatrix} J_1 & & & \\ & J_2 & & \\ & & \ddots & \\ & & & J_p \end{bmatrix}$$

where  $J_1, J_2, \dots, J_p$  are Jordan blocks.

$\Rightarrow$  From previous we know solution is

$$X(t) = T e^{Jt} T^{-1} X(0)$$

$$e^{Jt} = \begin{bmatrix} e^{\lambda_1 t} & & \\ & e^{\lambda_2 t} & \\ & & e^{\lambda_p t} \end{bmatrix}$$

$$\Rightarrow \text{If } J_i = \begin{pmatrix} \lambda_1 & & \\ & \lambda_1 & \\ & & -\lambda_1 \end{pmatrix}$$

$$\text{then } e^{J_i t} = e^{\lambda_1 t} \begin{pmatrix} 1 & \frac{t}{1!} & \frac{t^2}{2!} & \dots \\ & 1 & \frac{t}{1!} & \dots \\ & & \ddots & \\ & & & 1 \end{pmatrix}$$

$\Rightarrow$  So from above we know that  $e^{Jt}$  contains terms of form

$$\left\{ e^{\lambda_1 t} + t e^{\lambda_1 t} + t^2 e^{\lambda_1 t} + \dots \right\}$$

$\Rightarrow$  So Solution will also terms of above form.

$\Rightarrow$  If some of the eigen values are complex then term of form given below will appear.

$$t^m e^{\lambda t} \sin(\omega t + \phi)$$

⇒ For asymptotic stability:

$$\lambda_i < 0 \quad \forall i \in \{1, \dots, n\}$$

⇒ FACT: For linear system asymptotic stability = BIBO stability.