

### Course of "Automatic Control Systems" 2022/23

# Laplace domain analysis of LTI systems

#### Prof. Francesco Montefusco

Department of Economics, Law, Cybersecurity, and Sports Sciences Università degli studi di Napoli Parthenope francesco.montefusco@uniparthenope.it

Team code: uxbsz19

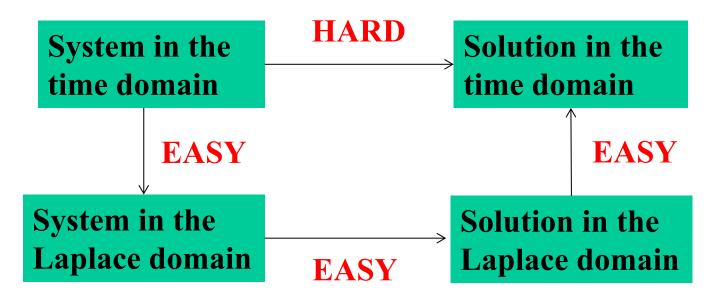


▲ Let us consider a Linear Time Invariant (LTI) system in the state space form

$$\dot{x}(t) = Ax(t) + Bu(t), \quad x(t_0) = x_0$$
 (1.a)

$$y(t) = Cx(t) + Du(t)$$
(1.b)

A The Evaluation of an LTI system response in a transformed domain is convenient only if





- ▲ Let us indicate with X(s), U(s) and Y(s) the Laplace transforms of the signals x(t), u(t) and y(t).
- ▲ Transforming both the sides of the equation (1), we have  $L(\dot{x}(t)) = L(Ax(t) + Bu(t))$  L(y(t)) = L(Cx(t) + Du(t))
- ▲ Using the time domain derivation property of the Laplace transform, a linear system in the Laplace domain can be written has

$$X(s) = (sI - A)^{-1}x_0 + (sI - A)^{-1}BU(s)$$
(2.1)

$$Y(s) = C(sI - A)^{-1}x_0 + C(sI - A)^{-1}BU(s) + DU(s)$$
(2.2)

▲ Note that in the Laplace domain the dependency of the state variables X(s) from the input U(s) is expressed by a matrix product instead of a convolution



- ▲ The matrix function  $\Phi(s) = (sI A)^{-1}$  is called *Transition matrix* whose dimension is given by the dimension of the A matrix.
- ▲ The matrix function  $W(s) = C(sI A)^{-1}B + D$  is called *Transfer function*.

$$X(s) = \Phi(s)x_0 + \Phi(s)BU(s)$$
(3.1)

$$Y(s) = C\Phi(s)x_0 + W(s)U(s)$$
(3.2)

- ▲ For Single Input Single Output (SISO) systems the transfer function W(s) is a scalar function;
- ▲ For *Multiple Input Multiple Output (MIMO)* systems the transfer function W(s) is a matrix whose element  $W(s)_{ij}$  will connect the output *i* with the input *j*.



## **Transfer function**

▲ For SISO systems the scalar *transfer function* is given by the ratio of two polynomial functions

$$W(s) = \frac{N(s)}{D(s)} = \frac{a_m s^m + a_{m-1} s^{m-1} + \dots + a_1 s + a_0}{b_n s^n + a_{n-1} s^{n-1} + \dots + b_1 s + b_0}$$

where  $m \leq n$ .

- ▲ If m < n the system is said strictly proper. It happens when the *D* matrix of the LTI system in the state space is zero.
- ▲ If m = n the system is said proper. It happens when the *D* matrix of the LTI system in the state space is different from zero zero.

$$W(s) = C(sI - A)^{-1}B + D$$



### Transition matrix

▲ Given a *transfer function* 

$$W(s) = \frac{N(s)}{D(s)} = \frac{a_m s^m + a_{m-1} s^{m-1} + \dots + a_1 s + a_0}{b_n s^n + a_{n-1} s^{n-1} + \dots + b_1 s + b_0}$$

- A The roots of the N(s) are said *zeros*.
- A The roots of the D(s) are said *poles*.
- A The polynomial D(s) is defined as D(s) = det(sI A), hence
  - ✤ D(s) coincides with the characteristic polynomial of the system
  - *the poles coincide with the eigenvalues of the system* except for possible pole-zero cancellation



▲ For SISO systems the free evolution in the Laplace domain is given by the ratio of polynomial functions

$$Y_{free}(s) = C\Phi(s)x_0$$

- ▲ This is also true for the forced evolution in case we restrict our attention to the case of polynomial and sinusoidal inputs  $Y_{forced}(s) = W(s)U(s)$
- A It is convenient to antitransform Y(s) by reducing the ratio of high degree polynomial functions to the sum of selected signals transform such as

$$L(e^{\alpha t}\cos(\omega t)\cdot 1(t)) = \frac{s-\alpha}{(s-\alpha)^2 + \omega^2} \qquad L(e^{\alpha t}sen(\omega t)\cdot 1(t)) = \frac{\omega}{(s-\alpha)^2 + \omega^2}$$
$$L(e^{\alpha t}1(t)) = \frac{1}{s-\alpha}$$



# Laplace antitransform

▲ Different methods can be used to reduce the ratio of high degree polynomial functions to the sum polynomial functions of degree one or two, such as *the residual method* (see the book for details).

#### Residual method for real and distinct eigenvalues

(see the book for the other cases)

$$W(s) = \frac{N(s)}{D(s)} = \frac{N(s)}{\prod_{i=1}^{n} (s - p_i)} \qquad p_i \neq p_j \text{ for } i \neq j$$

A In case of real and distinct eigenvalues, W(s) can be also written as

$$W(s) = \sum_{i=1}^{n} \frac{A_i}{s - p_i}$$

where 
$$A_k = \lim_{s \to p_k} (s - p_k)W(s)$$
. Hence

$$w(t) = \sum_{i=1}^{n} A_i e^{p_i t}$$



#### CASE 1: real and distinct eigenvalues

$$Y_{free}(s) = C\Phi(s)x_0 = \frac{s - 10}{(s + 2)(s + 5)}$$

▲ Appling the residual method we have

$$Y_{free}(s) = \frac{A_1}{(s+2)} + \frac{A_2}{(s+5)}$$

with

$$A_{1} = \lim_{s \to -2} (s+2)Y_{free}(s) = \lim_{s \to -2} \frac{s-10}{s+5} = -4$$
$$A_{2} = \lim_{s \to -5} (s+5)Y_{free}(s) = \lim_{s \to -5} \frac{s-10}{s+2} = 5$$

Hence

$$y_{free}(t) = (-4e^{-2t} + 5e^{-5t}) \cdot 1(t)$$



CASE 2: real multiple eigenvalues

$$Y_{forced}(s) = W(s)U(s) = \frac{s+18}{(s+3)^2}U(s)$$
 with  $u(t) = 1(t)$ 

 $\checkmark$  This function can be written as the sum of three terms

$$Y_{forced}(s) = \frac{s+18}{s(s+3)^2} = \frac{A_1}{s} + \frac{A_2}{(s+3)} + \frac{A_3}{(s+3)^2}$$

A The residual method can be applied to evaluate  $A_1$  and  $A_3$ , while  $A_2$  can be evaluated by substitution

$$A_1 = \lim_{s \to 0} sY_{forced}(s) = 2 \qquad A_3 = \lim_{s \to -3} (s+3)^2 Y_{forced}(s) = -5$$

while  $A_2 = -2$ . (By residual method for pole with multiplicity *r*, the l-th residual  $K_{l_i}$ l=1,...,r, by  $K_l = \frac{1}{(r-l)!} \frac{d^{r-l}}{ds^{r-l}} (s - p_i)^r Y(s)|_{s=p_i}$ , then  $A_2 = \frac{d(s+3)^2 Y_{forced}(s)}{ds}|_{s=-3}$ )

Hence, 
$$y_{forced}(t) = (2 - 2e^{-3t} - 5te^{-3t}) \cdot 1(t)$$



CASE 3: complex conjugate eigenvalues

$$Y_{free}(s) = C\Phi(s)x_0 = \frac{100}{(s+1)(s^2+4s+13)}$$

 $\checkmark$  This function can be written as the sum of two terms

$$Y_{free}(s) = \frac{A_1}{(s+1)} + \frac{A_2s + A_3}{(s^2 + 4s + 13)}$$

A The residual method can be applied to evaluate  $A_1$ , while  $A_2$  and  $A_3$  can be evaluated by substitution.  $A_1 = 10$ ,  $A_2 = -10$ ,  $A_3 = -30$ .

A Hence, 
$$Y_{free}(s) = \frac{10}{(s+1)} - 10 \frac{s+2}{(s+2)^2+9} - \frac{10}{3} \frac{3}{(s+2)^2+9}$$
 and

$$y_{free}(t) = \left(10e^{-t} - 10e^{-2t}\cos(3t) - \frac{10}{3}e^{-2t}\sin(3t)\right) \cdot 1(t)$$



Appendix 1

# INVERSE OF A MATRIX N×N

Prof. Francesco Montefusco



4

▲ Given a quadratic and invertible matrix

$$A = \begin{pmatrix} x_{1,1} & \dots & x_{1,j} \\ \vdots & \ddots & \vdots \\ x_{i,1} & \dots & x_{i,j} \end{pmatrix}$$

its inverse is defined as

$$A^{-1} = \frac{1}{\det(A)} \begin{pmatrix} \operatorname{cof}(A, x_{1,1}) & \dots & \operatorname{cof}(A, x_{1,j}) \\ \vdots & \ddots & \vdots \\ \operatorname{cof}(A, x_{i,1}) & \dots & \operatorname{cof}(A, x_{i,j}) \end{pmatrix}^T$$

where the cofactor is

$$\operatorname{cof}(A,i,j) = (-1)^{i+j} \operatorname{det}(\operatorname{minor}(A,i,j))$$

and the minor (i, j) is the determinant of the matrix obtained excluding the row i and the column j.



## Inverse of a 2×2 matrix

▲ Given a matrix

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$$

its inverse is

$$A^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$



## Inverse of a 3×3 matrix

#### $\blacktriangle$ Given a matrix

$$A = \begin{pmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \\ A_{31} & A_{32} & A_{33} \end{pmatrix}$$

its inverse is

$$\frac{1}{\det(A)} \begin{pmatrix} + \begin{vmatrix} A_{22} & A_{23} \\ A_{32} & A_{33} \end{vmatrix} & - \begin{vmatrix} A_{12} & A_{13} \\ A_{32} & A_{33} \end{vmatrix} & + \begin{vmatrix} A_{12} & A_{13} \\ A_{22} & A_{23} \end{vmatrix}} \\ - \begin{vmatrix} A_{21} & A_{23} \\ A_{31} & A_{33} \end{vmatrix} & + \begin{vmatrix} A_{11} & A_{13} \\ A_{31} & A_{33} \end{vmatrix} & - \begin{vmatrix} A_{11} & A_{13} \\ A_{21} & A_{23} \end{vmatrix}} \\ + \begin{vmatrix} A_{21} & A_{22} \\ A_{31} & A_{32} \end{vmatrix} & - \begin{vmatrix} A_{11} & A_{12} \\ A_{31} & A_{32} \end{vmatrix} & + \begin{vmatrix} A_{11} & A_{12} \\ A_{31} & A_{32} \end{vmatrix}$$



Appendix 2

# **EIGENVALUES AND EIGENVECTORS**



▲ Given a matrix  $A \in \mathbb{R}^{n \times n}$ , a scalar  $\lambda \in \mathbb{C}$  is said *eigenvalue* of the matrix A if there exists a vector  $v \in \mathbb{C}^n$ , said *eigenvector*, such that

$$Av = \lambda v$$

A Taking into account account that eigenvalues and eigenvectors of a matrix verify the equation

$$(A-\lambda I)v=0.$$

The eigenvalues can be found evaluating the roots of the *characteristic polynomial*  $p(\lambda)$  defined as

$$p(\lambda) = det(A - \lambda I).$$

A The **poles of W(s)** coincide to the eigenvalues of the matrix A.



# Examples

$$A = \begin{pmatrix} 1 & 2 \\ 3 & -4 \end{pmatrix}$$

$$p(s) = s^2 + 3s - 10$$

Eigenvalues 
$$\lambda_1 = 2, \lambda_2 = -5$$

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & -5 & 6 \\ 7 & -8 & 9 \end{pmatrix}$$
$$p(s) = s^{3} - 5s^{2} - 22s - 24$$
Eigenvalues
$$\lambda_{1} = 8.09$$
$$\lambda_{2} = -1.54 + j0.765$$
$$\lambda_{3} = -1.54 - j0.765$$