

# Control Theory

Course ID: 322MI – Spring 2023

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322MI –Spring 2023

Lecture 2: Solutions to linear systems

## Continuous-time linear systems

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Consider a *homogeneous* (i.e. having no input) continuous-time linear time-invariant system:

$$\dot{x}(t) = Ax(t), \quad x(0) = x_0, \quad (1)$$

where  $x \in \mathbb{R}^n$  and the initial time is 0 with no loss of generality.

We want to find the solution  $x(t)$ ,  $t \geq 0$ . To this aim, let's consider the scalar case (i.e.,  $n = 1$ ) first:

$$\dot{x}(t) = ax(t), \quad x(0) = x_0, \quad x, a \in \mathbb{R}. \quad (2)$$

The solution is easily proven to be:

$$x(t) = e^{at}x_0.$$

In the general case (i.e.,  $n \geq 1$ ), the solution takes the same form, as shown in the following.

Motivated by the analogy with the scalar

$$e^a = \sum_{k=0}^{\infty} \frac{a^k}{k!},$$

we can define the *matrix exponential* of a given  $n \times n$  matrix  $A$  by

$$e^A = \sum_{k=0}^{\infty} \frac{A^k}{k!}.$$

The above is a convergent series, since the factorial dominates the exponential for  $k \rightarrow \infty$ . Now, it is easy to prove that

$$x(t) = e^{At} x_0$$

is the solution of (1).

Indeed, it satisfies the initial condition:

$$x(0) = e^{A0}x_0 = Ix_0 = x_0;$$

moreover, by taking the derivative, we get:

$$\frac{d}{dt} \left( e^{At}x_0 \right) = \frac{d}{dt} \left( \sum_{k=0}^{\infty} \frac{A^k t^k}{k!} \right) = \sum_{k=1}^{\infty} A \frac{A^{k-1} t^{k-1}}{(k-1)!} = A \left( e^{At}x_0 \right),$$

thus it satisfies the differential equation.

Consider a *nonhomogeneous* continuous-time linear time-invariant system:

$$\dot{x}(t) = Ax(t) + Bu(t), \quad x(0) = x_0 \quad (3)$$

It can be verified that the solution of (3) is:

$$x(t) = e^{At}x_0 + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau.$$

Indeed, by taking the Laplace transform of (3), we get

$$sX(s) - x_0 = AX(s) + BU(s) \implies (sI - A)X(s) = x_0 + BU(s),$$

and, solving for  $X(s)$ :

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

which is the solution in terms of Laplace transforms.

Now, recalling that  $\mathcal{L}(e^{At}) = (sI - A)^{-1}$ , and that the Laplace transform of the convolution of two functions is the product of the individual Laplace transforms, we have

$$x(t) = \mathcal{L}^{-1}[X(s)] = \mathcal{L}^{-1}[(sI - A)^{-1}x_0] + \mathcal{L}^{-1}[(sI - A)^{-1}BU(s)] = e^{At}x_0 + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau.$$

We can thus state the following (where the initial time is now  $t_0$ )

### Theorem

The solution of

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

takes the form:

$$x(t) = \varphi(t, t_0, x_0, u(\cdot)) = e^{A(t-t_0)}x_0 + \int_{t_0}^t e^{A(t-\tau)}Bu(\tau)d\tau.$$



In the right side of the expression

$$x(t) = e^{A(t-t_0)}x_0 + \int_{t_0}^t e^{A(t-\tau)}Bu(\tau)d\tau$$

the first term depends on the initial state  $x_0$ , but not on the input, while the second depends on the input  $u(\cdot)$ , but not on the initial state. Thus, the whole solution can be decomposed as follows:

- *Natural (state) response*, i.e. the solution when the input is zero:

$$u(t) = 0, \forall t \geq t_0 \implies x_N(t) = e^{A(t-t_0)}x_0$$

- *Forced (state) response*, i.e. the solution when the initial state is zero:

$$x_0 = 0 \implies x_F(t) = \int_{t_0}^t e^{A(t-\tau)}Bu(\tau)d\tau$$

The whole solution is indeed

$$x(t) = x_N(t) + x_F(t).$$

Taking into account the output equation, we have:

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t), & x(t_0) = x_0 \\ y(t) = Cx(t) + Du(t) \end{cases}$$

and by substituting the state response  $x(t)$  in the output equation, we get:

$$y(t) = Ce^{A(t-t_0)}x_0 + \int_{t_0}^t Ce^{A(t-\tau)}Bu(\tau)d\tau + Du(t)$$

- *Natural response.* By setting  $u(t) = 0$ ,  $t \geq t_0$  we get:

$$y(t) = y_N(t) = Ce^{A(t-t_0)}x_0$$

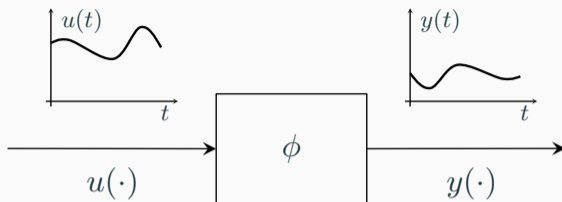
- *Forced response.* By setting  $x_0 = 0$  we get:

$$y(t) = y_F(t) = \int_{t_0}^t Ce^{A(t-\tau)}Bu(\tau)d\tau + Du(t)$$

The whole output response is thus given by:

$$y(t) = y_N(t) + y_F(t).$$

# Input-output representation



A continuous-time linear system can be represented as linear *operator*  $\phi$  mapping input signals to output signals. That representation is the *input-output representation* of linear systems:

$$\phi : \mathbb{U} \longrightarrow \mathbb{Y}$$

where  $\mathbb{U}$  is a vector space of input signals

$$u(\cdot) : \mathbb{R} \longrightarrow \mathbb{R}^m$$

and  $\mathbb{Y}$  is a vector space of output signals

$$y(\cdot) : \mathbb{R} \longrightarrow \mathbb{R}^p$$

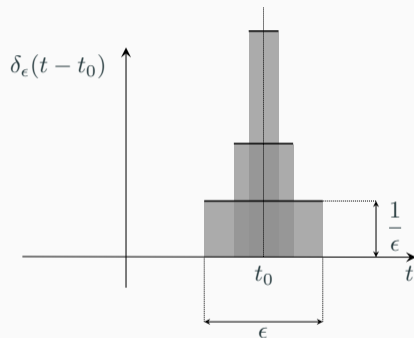
The operator (and thus, the system) can be characterized by the *impulse response*, i.e. the response of the system to a particular input called the impulse. Although the concept is far more general, in the following we consider only the case of causal linear time-invariant systems.

# The Dirac delta

An impulse is a phenomenon with high intensity and very short duration. To represent it mathematically, we can consider a function  $\delta_\epsilon(t)$  defined as

$$\delta_\epsilon(t) = \begin{cases} 0 & \text{if } t < -\frac{\epsilon}{2} \\ \frac{1}{\epsilon} & \text{if } -\frac{\epsilon}{2} \leq t \leq \frac{\epsilon}{2} \\ 0 & \text{if } t > \frac{\epsilon}{2} \end{cases}$$

The support of the function (namely, the interval where the function is non-zero) is  $[-\epsilon/2, \epsilon/2]$ . For decreasing  $\epsilon$ , the interval becomes increasingly small, while the value taken by the function, i.e.  $1/\epsilon$ , becomes increasingly large. Note that the integral of the function remains equal to 1.



## The Dirac delta (cont.)

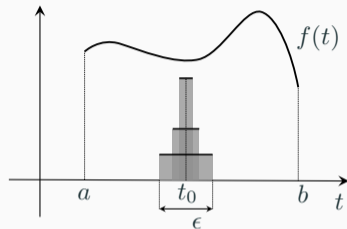
Let  $t_0$  be a point in the interior of  $[a, b]$  and  $\epsilon$  be such that  $[t_0 - \epsilon/2, t_0 + \epsilon/2] \subset [a, b]$ . The impulse in  $t_0$ ,  $\delta(t - t_0)$ , can be seen as the the “limit” for  $\epsilon \rightarrow 0$  of the function  $\delta_\epsilon(t - t_0)$ . Intuitively (a formal treatment can be found in Antsaklis and Michel (2006)), consider the integral

$$\int_a^b f(t) \delta_\epsilon(t - t_0) dt,$$

where  $f$  is a continuous function. Then

$$\begin{aligned} \int_a^b f(t) \delta_\epsilon(t - t_0) dt &= \int_{t_0 - \frac{\epsilon}{2}}^{t_0 + \frac{\epsilon}{2}} f(t) \delta_\epsilon(t - t_0) dt = \\ &= \int_{t_0 - \frac{\epsilon}{2}}^{t_0 + \frac{\epsilon}{2}} f(t) \frac{1}{\epsilon} dt = \frac{1}{\epsilon} f(\tau) \epsilon = f(\tau) \end{aligned}$$

where  $\tau$  (which exists for the mean value theorem) belongs to the interval  $[t_0 - \frac{\epsilon}{2}, t_0 + \frac{\epsilon}{2}]$ .



## The Dirac delta (cont.)

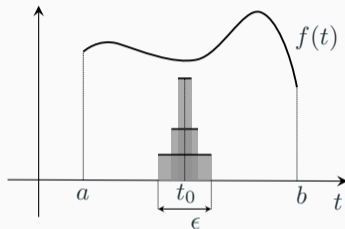
Being  $f$  continuous, when  $\epsilon \rightarrow 0$  we have

$$\int_a^b f(t) \delta_\epsilon(t - t_0) dt \rightarrow f(t_0).$$

The *Dirac delta distribution*  $\delta(t - t_0)$  is defined as the “function” such that for every continuous function  $f$  defined in  $[a, b]$  containing  $t_0$ , we have that

$$\int_a^b f(t) \delta(t - t_0) dt = f(t_0). \quad (4)$$

Eq. (4) is called the *sifting property* of the impulse (or the *sampling property* of the impulse).



Consider the SISO system

$$\begin{cases} \dot{x}(t) &= Ax(t) + Bu(t) \\ y(t) &= Cx(t) + Du(t) \end{cases}$$

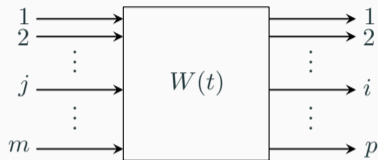
and assume that  $x(0) = 0$ . By applying the unit impulse  $\delta(t)$ , we get

$$y(t) = \int_0^t Ce^{A(t-\tau)} B \delta(\tau) d\tau + D\delta(t) = Ce^{At} B + D\delta(t)$$

where the second equality follows from the sifting property. The function

$$W(t) \doteq Ce^{At} B + D\delta(t)$$

is called the *impulse response* of the system.



In the MIMO case,  $W(t)$  is a  $p \times m$  matrix: each element  $w_{ij}(t)$  represents the ensuing response of the  $i$ th output at time  $t$ , due to an impulse applied at time 0 to the  $j$ th input, for zero initial condition.



## Discrete-time linear systems

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Consider a homogeneous discrete-time linear time-invariant system:

$$x(k+1) = Ax(k), \quad x(k_0) = x_0$$

where  $x \in \mathbb{R}^n$ .

Clearly,  $x(k)$ ,  $k > k_0$  can be determined by iterating the state equation:

$$\begin{aligned}x(k_0) &= x_0 \\x(k_0 + 1) &= Ax(k_0) \\x(k_0 + 2) &= Ax(k_0 + 1) = A^2x(k_0) \\&\vdots \\x(k) &= A^{k-k_0}x(k_0)\end{aligned}$$

thus we have:

$$x(k) = A^{k-k_0}x_0.$$

Now, consider a nonhomogeneous linear discrete-time system:

$$x(k+1) = Ax(k) + Bu(k), \quad x(k_0) = x_0$$

Clearly:

$$\begin{aligned}x(k_0) &= x_0 \\x(k_0 + 1) &= Ax(k_0) + Bu(k_0) \\x(k_0 + 2) &= Ax(k_0 + 1) + Bu(k_0 + 1) \\&= A[Ax(k_0) + Bu(k_0)] + Bu(k_0 + 1) \\&= A^2x(k_0) + ABu(k_0) + Bu(k_0 + 1) \\x(k_0 + 3) &= Ax(k_0 + 2) + B(k_0 + 2)u(k_0 + 2) \\&= A^3x(k_0) + A^2Bu(k_0) + ABu(k_0 + 1) + Bu(k_0 + 2) \\&\vdots \\x(k) &= A^{(k-k_0)}x(k_0) + \sum_{j=k_0}^{k-1} A^{(k-1-j)}Bu(j)\end{aligned}$$

We can thus state the following

### Theorem

The solution of

$$x(k+1) = Ax(k) + Bu(k), \quad x(k_0) = x_0$$

takes the form:

$$x(k) = \varphi(k, k_0, x_0, u(\cdot)) = A^{(k-k_0)}x_0 + \sum_{j=k_0}^{k-1} A^{(k-1-j)}Bu(j)$$

Taking into account the output equation, we have:

$$\begin{cases} x(k+1) = Ax(k) + Bu(k), & x(k_0) = x_0 \\ y(k) = Cx(k) + Du(k) \end{cases}$$

By substituting the state response  $x(k)$  in the output equation we get:

$$y(k) = CA^{(k-k_0)}x_0 + \sum_{j=k_0}^{k-1} CA^{k-1-j}Bu(j) + Du(k), \quad k \geq k_0$$

- *Natural response.* By setting  $u(k) = 0, \forall k \geq k_0$ , we get:

$$y(k) = y_N(k) = CA^{(k-k_0)}x_0, \quad k \geq k_0$$

- *Forced response.* By setting  $x_0 = 0$ , we get:

$$y(k) = y_F(k) = \sum_{j=k_0}^{k-1} CA^{k-1-j}Bu(j) + Du(k), \quad k \geq k_0$$

The whole response is thus given by:

$$y(k) = y_N(k) + y_F(k).$$

A discrete-time linear system can be represented as linear *operator*  $\phi$  mapping input signals to output signals. That representation is the input-output representation of linear systems:

$$\phi : \mathbb{U} \longrightarrow \mathbb{Y}$$

where  $\mathbb{U}$  is a vector space of input signals

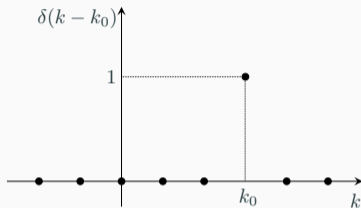
$$u(\cdot) : \mathbb{Z} \longrightarrow \mathbb{R}^m$$

and  $\mathbb{Y}$  is a vector space of output signals

$$y(\cdot) : \mathbb{Z} \longrightarrow \mathbb{R}^p$$

The operator (and thus, the system) can be characterized by the *impulse response*, i.e. the response of the system to a particular input called the impulse.

Although the concept is far more general, in the following we consider only the case of causal linear time-invariant systems.



In the discrete-time case, the unit impulse at time  $k_0$  is simply:

$$\delta(k - k_0) = \begin{cases} 0, & k \neq k_0, k \in \mathbb{Z} \\ 1, & k = k_0 \end{cases}$$

Consider the SISO system

$$\begin{cases} x(k+1) &= Ax(k) + Bu(k) \\ y(k) &= Cx(k) + Du(k) \end{cases}$$

and assume that  $x(0) = 0$ . By applying the unit impulse  $\delta(k)$  we get

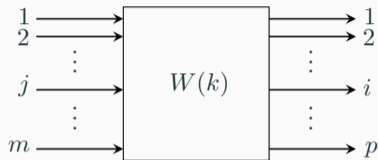
$$y(k) = \sum_{j=0}^{k-1} CA^{k-1-j} B\delta(j) + D\delta(k)$$

where the summation is assumed to be zero for  $k = 0$ . The function

$$W(k) \doteq \begin{cases} CA^{k-1}B, & k > 0 \\ D, & k = 0 \\ 0, & k < 0 \end{cases}$$

is called the *impulse response* of the system.





In the MIMO case,  $W(k)$  is a  $p \times m$  matrix: each element  $w_{ij}(k)$  represents the ensuing response of the  $i$ th output at time  $k$ , due to an impulse applied at time 0 to the  $j$ th input, for zero initial condition.

## Modal analysis (continuous-time)

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We have seen that the state response of the system

$$\dot{x}(t) = Ax(t) + Bu(t)$$

takes the form:

$$x(t) = e^{A(t-t_0)}x_0 + \int_{t_0}^t e^{A(t-\tau)}Bu(\tau)d\tau$$

Without loss of generality, we can take  $t_0 = 0$  thus obtaining

$$x(t) = e^{At}x_0 + \int_0^t e^{A(t-\tau)}Bu(\tau)d\tau$$

The matrix  $A$  plays a fundamental role and is responsible of the qualitative behavior of the response. In the following we will analyze the qualitative behavior, starting with the simple case of  $A$  being diagonalizable.

If  $A$  is diagonalizable by a similarity transformation we can write:

$$\begin{cases} A = T\Lambda T^{-1} \\ \Lambda = T^{-1}AT \end{cases}$$

where  $\Lambda$  is a diagonal matrix having diagonal elements  $\{\lambda_1, \lambda_2, \dots, \lambda_n\}$ , where  $\lambda_i$  is the  $i$ th eigenvalue of  $A$ . The columns of the matrix  $T$  are eigenvectors  $t_i$  of matrix  $A$ . The inverse of  $T$ ,  $S = T^{-1}$  can be partitioned row-wise

$$T = [t_1 \ t_2 \ \dots \ t_n], \quad T^{-1} = S = \begin{bmatrix} s_1^\top \\ s_2^\top \\ \vdots \\ s_n^\top \end{bmatrix}.$$

Thus,  $A$  may be rewritten as:

$$A = [t_1 \ t_2 \ \dots \ t_n] \begin{bmatrix} \lambda_1 & 0 & 0 & 0 & \dots \\ 0 & \lambda_2 & 0 & 0 & \dots \\ 0 & 0 & \lambda_3 & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_n \end{bmatrix} \begin{bmatrix} s_1^\top \\ s_2^\top \\ \vdots \\ s_n^\top \end{bmatrix}. \quad (5)$$

## Modal analysis, diagonalizable $A$ (cont.)

As a consequence,  $e^{At}$  can be written as:

$$\begin{aligned} e^{At} &= \sum_{k=0}^{\infty} A^k \frac{t^k}{k!} = \sum_{k=0}^{\infty} (T\Lambda T^{-1})^k \frac{t^k}{k!} = \sum_{k=0}^{\infty} \underbrace{(T\Lambda T^{-1}T\Lambda T^{-1} \dots T\Lambda T^{-1})}_{k \text{ times}} \frac{t^k}{k!} \\ &= \sum_{k=0}^{\infty} T\Lambda^k T^{-1} \frac{t^k}{k!} = T \left( \sum_{k=0}^{\infty} \Lambda^k \frac{t^k}{k!} \right) T^{-1} \end{aligned}$$

thus

$$e^{At} = T e^{\Lambda t} T^{-1}$$

By expliciting the columns  $t_i$  of  $T$  e the rows  $s_i^\top$  of  $S = T^{-1}$  we get:

$$e^{At} = [t_1 \ t_2 \ \dots \ t_n] \begin{bmatrix} e^{\lambda_1 t} & 0 & 0 & 0 & \dots \\ 0 & e^{\lambda_2 t} & 0 & 0 & \dots \\ 0 & 0 & e^{\lambda_3 t} & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & 0 \\ 0 & 0 & 0 & \dots & e^{\lambda_n t} \end{bmatrix} \begin{bmatrix} s_1^\top \\ s_2^\top \\ \vdots \\ s_n^\top \end{bmatrix} = \sum_{i=1}^n t_i s_i^\top e^{\lambda_i t}.$$

## Modal analysis, diagonalizable $A$ (cont.)

By defining  $n$  matrices of size  $n \times n$

$$Z_i = t_i s_i^\top, \quad i = 1, \dots, n$$

we can state the following

### Property

If  $A$  is diagonalizable, the state transition matrix  $e^{At}$  can be written as the sum of constant matrices  $Z_i$ , each multiplied by the function  $e^{\lambda_i t}$

$$e^{At} = \sum_{i=1}^n Z_i e^{\lambda_i t}. \quad (6)$$

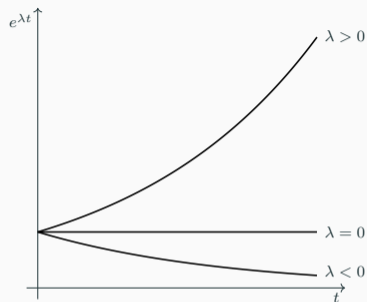
The natural state response can thus be expressed as

$$x_N(t) = \sum_{i=1}^n Z_i e^{\lambda_i t} x(0) = \sum_{i=1}^n t_i (s_i^\top x(0)) e^{\lambda_i t} = \sum_{i=1}^n t_i \alpha_i(x(0)) e^{\lambda_i t} \quad (7)$$

where  $\alpha_i(x(0)) = s_i^\top x(0)$ ,  $i = 1, 2, \dots, n$  are scalars obtained as the dot product of each left eigenvector and the initial condition  $x_0$ . The functions

$$e^{\lambda_i t}$$

are the *modes* of the system.



If  $\lambda \in \mathbb{R}$ , the mode  $e^{\lambda t}$  is an exponential mode that, for increasing  $t$ , has the following behavior:

- if  $\lambda > 0$  the mode diverges;
- if  $\lambda < 0$  the mode vanishes;
- if  $\lambda = 0$  the mode is constant.

## Modal analysis, diagonalizable $A$ (cont.)

In general,  $A$  may have complex eigenvalues (i.e. eigenvalues whose imaginary part is non zero). It is well-known that if  $\lambda, v$  is an eigenpair of  $A$ , the complex conjugate pair  $\lambda^*, v^*$  is an eigenpair of  $A$  too.

Without loss of generality, assume that  $\lambda_1, \lambda_2, \dots, \lambda_r$ , where  $r \leq n$ , are real numbers and  $\lambda_{r+1}, \dots, \lambda_n$  are complex numbers, ordered pairwise  $\lambda_{i+1} = \lambda_i^*$ , or

$$\sigma(A) = \{ \underbrace{\lambda_1, \lambda_2, \dots, \lambda_r}_{\text{real}}, \underbrace{\lambda_{r+1}, \lambda_{r+1}^*}_{\text{conjugate}}, \underbrace{\lambda_{r+3}, \lambda_{r+3}^*}_{\text{conjugate}}, \dots, \underbrace{\lambda_{n-1}, \lambda_{n-1}^*}_{\text{conjugate}} \}.$$

Then it follows that  $Z_i$  e  $Z_{i+1}$  are conjugate if  $\lambda_i$  and  $\lambda_{i+1}$  are. As a consequence:

$$e^{At} = \sum_{i=1}^r Z_i e^{\lambda_i t} + \sum_{r+1}^{n-1} (Z_i e^{\lambda_i t} + Z_i^* e^{\lambda_i^* t})$$

Decomposing  $\lambda_i$  and  $Z_i$  in real and imaginary part we get:

$$\begin{aligned} \lambda_i &= \mu_i + j\omega_i \\ Z_i &= M_i + jN_i. \end{aligned}$$



From Euler's formula

$$e^{\lambda_i t} = e^{\mu_i t} [\cos(\omega_i t) + j \sin(\omega_i t)]$$

and its simple to check that the imaginary contributions cancel each other thus

$$e^{At} = \sum_{i=1}^r Z_i e^{\mu_i t} + \\ + \sum_{r+1}^{n-1} 2e^{\mu_i t} [M_i \cos(\omega_i t) - N_i \sin(\omega_i t)].$$

$r+1$  (step 2)

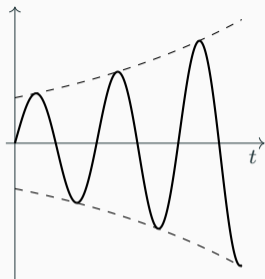
The following fundamental property holds.

### Property

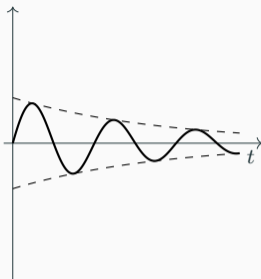
Each conjugate pair of eigenvalues  $\lambda = \mu + j\omega$  and  $\lambda^* = \mu - j\omega$  produces the complex modes  $e^{\lambda t}$  and  $e^{\lambda^* t}$ , that result in real modes of the form:

$$e^{\mu t} \cos(\omega t) \quad \text{and} \quad e^{\mu t} \sin(\omega t).$$

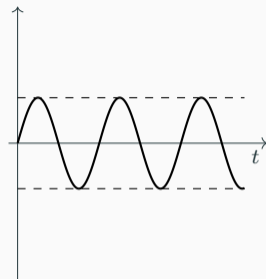
## Modal analysis, diagonalizable $A$ (cont.)



$$e^{\mu t} \sin(\omega t) \text{ for } \mu > 0.$$



$$e^{\mu t} \sin(\omega t) \text{ for } \mu < 0.$$



$$e^{\mu t} \sin(\omega t) \text{ for } \mu = 0.$$

The state transition matrix is thus governed by real exponential terms associated to real eigenvalues and oscillating (“*pseudo-periodic*”) modes associated to the conjugate pairs of eigenvalues. Depending on the real part of the conjugate pairs the following behaviors may occur:

- if  $\mu > 0$  the amplitude of the oscillation diverges;
- if  $\mu < 0$  the amplitude of the oscillation vanishes;
- if  $\mu = 0$  the amplitude of the oscillation is constant.

If  $A$  is non-diagonalizable, we need to resort to the following

## Theorem (Jordan normal form)

For every matrix  $A \in \mathbb{C}^{n \times n}$ , there exists a non-singular change of basis matrix  $T \in \mathbb{C}^{n \times n}$  such that

$$J = T^{-1}AT = \begin{bmatrix} J_1 & 0 & \dots & 0 \\ 0 & J_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & J_s \end{bmatrix}, \quad \text{where}$$

$$J_i = \begin{bmatrix} \lambda_i & 1 & 0 & \dots \\ 0 & \lambda_i & \ddots & 0 \\ \vdots & \vdots & \ddots & 1 \\ 0 & \dots & \dots & \lambda_i \end{bmatrix}$$

is a *Jordan block* with each  $\lambda_i$  an eigenvalue of  $A$  and  $s$  equal to the number of independent eigenvectors of  $A$ . The matrix  $J$  is unique up to a reordering of the blocks and is called the *Jordan normal form* of  $A$ .

## Note

There may be several Jordan blocks associated to the same eigenvalue. Sometimes the  $J_i$  are referred to as the Jordan *mini*-blocks, and the diagonal block composed of all the mini-blocks associated to the same eigenvalue is called a Jordan block.

Notice that the  $i$ th block  $J_i \in \mathbb{R}^{\nu_i \times \nu_i}$  may be written as

$$J_i = \underbrace{\begin{bmatrix} \lambda_i & 0 & 0 & 0 & \dots \\ 0 & \lambda_i & 0 & 0 & \dots \\ 0 & 0 & \lambda_i & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & 0 \\ 0 & 0 & 0 & \dots & \lambda_i \end{bmatrix}}_{=\lambda_i I} + \underbrace{\begin{bmatrix} 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & 1 & 0 & \dots \\ 0 & 0 & 0 & 1 & \dots \\ \vdots & \vdots & \vdots & \ddots & 1 \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix}}_{\doteq J_{i0}}$$

thus

$$J_i = \Lambda_i + J_{i0}.$$

Since  $J = T^{-1}AT$ , the state transition matrix may be written as

$$e^{At} = Te^{Jt}T^{-1} \quad (8)$$

where

$$e^{Jt} = \begin{bmatrix} e^{J_1t} & 0 & 0 & 0 & \dots \\ 0 & e^{J_2t} & 0 & 0 & \dots \\ 0 & 0 & e^{J_3t} & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & e^{J_st} \end{bmatrix}.$$

Now, let us consider the block  $e^{J_it} = e^{(\Lambda_it + J_{i0}t)}$ . It is easy to check that, if the square matrices  $M$  and  $N$  are such that  $MN = NM$ , i.e. if they commute, then  $e^{(M+N)} = e^M e^N$ . From the definition of  $\Lambda_i$  and  $J_{i0}$  it follows that they commute:  $\Lambda_i J_{i0} = J_{i0} \Lambda_i$ , hence

$$e^{J_it} = e^{\lambda_i I t} e^{J_{i0}t}. \quad (9)$$

## Modal analysis, non-diagonalizable $A$ (cont.)

The powers of  $J_{i0}$  are obtained, by “moving upwards the 1s”, for instance:

$$J_{i0} = \begin{bmatrix} 0 & 1 & 0 & 0 & \dots \\ 0 & 0 & 1 & 0 & \dots \\ 0 & 0 & 0 & 1 & \dots \\ \vdots & \vdots & \vdots & \ddots & 1 \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix}, \quad J_{i0}^2 = \begin{bmatrix} 0 & 0 & 1 & \dots & 0 \\ 0 & 0 & 0 & 1 & \vdots \\ 0 & 0 & 0 & \dots & \vdots \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix} \dots$$
$$J_{i0}^{\nu_i-1} = \begin{bmatrix} 0 & 0 & 0 & \dots & 1 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix}, \quad J_{i0}^{\nu_i} = \begin{bmatrix} 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ 0 & 0 & 0 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & 0 \end{bmatrix}.$$

Moreover,  $J_{i0}^p = 0$ ,  $\forall p \geq \nu_i$ . Thus, the series corresponding to  $e^{J_{i0}t}$  is actually a sum of a finite number of terms

$$e^{J_{i0}t} = \sum_{j=0}^{\nu_i-1} \frac{1}{j!} J_{i0}^j t^j.$$

## Modal analysis, non-diagonalizable $A$ (cont.)

By inspecting the form of each of the terms of  $e^{J_{i0}t} = \sum_{j=0}^{\nu_i-1} \frac{1}{j!} J_{i0}^j t^j$ , it is easy to check that

$$e^{J_{i0}t} = \begin{bmatrix} 1 & t & \frac{t^2}{2!} & \frac{t^3}{3!} & \cdots & \frac{t^{\nu_i-1}}{(\nu_i-1)!} \\ 0 & 1 & t & \frac{t^2}{2!} & \cdots & \frac{t^{\nu_i-2}}{(\nu_i-2)!} \\ 0 & 0 & 1 & t & \cdots & \frac{t^{\nu_i-3}}{(\nu_i-3)!} \\ \vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \ddots & t \\ 0 & 0 & 0 & 0 & \cdots & 1 \end{bmatrix}$$

On the other hand:

$$e^{\lambda_i I t} = \begin{bmatrix} e^{\lambda_i t} & 0 & 0 & \cdots \\ 0 & e^{\lambda_i t} & 0 & \cdots \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & e^{\lambda_i t} \end{bmatrix}.$$

## Modal analysis, non-diagonalizable $A$ (cont.)

Thus, the  $i$ th block of  $e^{Jt}$  takes the form

$$e^{J_i t} = \sum_{j=0}^{\nu_i-1} J_{i0}^j \frac{t^j}{j!} e^{\lambda_i t} = e^{\lambda_i t} \begin{bmatrix} 1 & t & \frac{t^2}{2!} & \frac{t^3}{3!} & \cdots & \frac{t^{\nu_i-1}}{(\nu_i-1)!} \\ 0 & 1 & t & \frac{t^2}{2!} & \cdots & \frac{t^{\nu_i-2}}{(\nu_i-2)!} \\ 0 & 0 & 1 & t & \cdots & \frac{t^{\nu_i-3}}{(\nu_i-3)!} \\ \vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \ddots & t \\ 0 & 0 & 0 & 0 & \cdots & 1 \end{bmatrix} \quad (10)$$

or

$$e^{J_i t} = \begin{bmatrix} e^{\lambda_i t} & te^{\lambda_i t} & \frac{t^2}{2!} e^{\lambda_i t} & \frac{t^3}{3!} e^{\lambda_i t} & \cdots & \frac{t^{\nu_i-1}}{(\nu_i-1)!} e^{\lambda_i t} \\ 0 & e^{\lambda_i t} & te^{\lambda_i t} & \frac{t^2}{2!} e^{\lambda_i t} & \cdots & \frac{t^{\nu_i-2}}{(\nu_i-2)!} e^{\lambda_i t} \\ 0 & 0 & e^{\lambda_i t} & te^{\lambda_i t} & \cdots & \frac{t^{\nu_i-3}}{(\nu_i-3)!} e^{\lambda_i t} \\ \vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\ 0 & 0 & 0 & 0 & \ddots & te^{\lambda_i t} \\ 0 & 0 & 0 & 0 & \cdots & e^{\lambda_i t} \end{bmatrix} \quad (11)$$



## Modal analysis, non-diagonalizable $A$ (cont.)

Back to the exponential matrix, letting  $S = T^{-1}$ , by partitioning  $T$  (column-wise) and  $S$  (row-wise) – according to the size of the diagonal blocks – we get:

$$e^{At} = [T_1 \ T_2 \ \dots \ T_s] \begin{bmatrix} e^{J_1 t} & 0 & 0 & 0 & \dots \\ 0 & e^{J_2 t} & 0 & 0 & \dots \\ 0 & 0 & e^{J_3 t} & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & e^{J_s t} \end{bmatrix} \begin{bmatrix} S_1^\top \\ S_2^\top \\ \vdots \\ S_s^\top \end{bmatrix}, \quad (12)$$

where  $T_i \in \mathbb{C}^{n \times \nu_i}$  e  $S_i^\top \in \mathbb{C}^{\nu_i \times n}$ .

Thus

$$e^{At} = \sum_{i=1}^s [T_i e^{J_i t} S_i^\top], \quad (13)$$

and, by using (10),

$$e^{At} = \sum_{i=1}^s \sum_{j=0}^{\nu_i-1} T_i J_{i0}^j S_i^\top \frac{t^j}{j!} e^{\lambda_i t}.$$

Finally, by letting  $Z_{ij} = T_i J_{i0}^j S_i^\top \frac{1}{j!}$  we obtain

$$e^{At} = \sum_{i=1}^s \sum_{j=0}^{\nu_i-1} Z_{ij} t^j e^{\lambda_i t}. \quad (14)$$

If there exist Jordan blocks of size greater than one, associated to the eigenvalue  $\lambda$ , in the matrix  $e^{At}$ , the following modes will appear:

$$e^{\lambda t}, te^{\lambda t}, t^2 e^{\lambda t}, \dots, t^{(\nu(\lambda)-1)} e^{\lambda t}$$

where  $\nu(\lambda)$  denotes the *degree* of the eigenvalue  $\lambda$ , i.e. the size of the largest Jordan block associated to  $\lambda$ .

**Example.** Let the Jordan form of a matrix  $A$  be

$$\begin{bmatrix} 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 2 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 2 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 2 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 5 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 5 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 5 \end{bmatrix}$$

Then, its eigenvalues are 2 and 5, having degree  $\nu(2) = 3$  and  $\nu(5) = 2$ , respectively. The modes of  $e^{At}$  are:

$$e^{2t}, te^{2t}, t^2e^{2t}, e^{5t}, te^{5t}.$$

## Modal analysis (discrete-time)

---

We have seen that the state response of the system

$$x(k+1) = Ax(k) + Bu(k)$$

takes the form:

$$x(k) = A^{(k-k_0)}x_0 + \sum_{j=k_0}^{k-1} A^{(k-1-j)}Bu(j)$$

Without loss of generality we can take  $k_0 = 0$  thus obtaining

$$x(k) = A^k x_0 + \sum_{j=0}^{k-1} A^{(k-1-j)} Bu(j)$$

The transition matrix  $A^k$  can be computed simply as a matrix product repeated  $k - 1$  times, but this is of little interest. Instead, to reveal the properties of the response, we can perform a modal analysis, similarly to the continuous-time case. We will first consider the case of diagonalizable  $A$ .

If  $A$  is diagonalizable by a similarity transformation, we can write:

$$\begin{cases} A = T\Lambda T^{-1} \\ \Lambda = T^{-1}AT \end{cases}$$

where  $\Lambda$  is a diagonal matrix having diagonal elements  $\{\lambda_1, \lambda_2, \dots, \lambda_n\}$ , where  $\lambda_i$  is the  $i$ th eigenvalue of  $A$ . The columns of the matrix  $T$  are eigenvectors  $t_i$  of matrix  $A$ . The inverse of  $T$ ,  $S = T^{-1}$  can be partitioned row-wise

$$T = [t_1 \ t_2 \ \dots \ t_n], \quad T^{-1} = S = \begin{bmatrix} s_1^\top \\ s_2^\top \\ \vdots \\ s_n^\top \end{bmatrix}.$$

Thus,  $A$  may be rewritten as:

$$A = [t_1 \ t_2 \ \dots \ t_n] \begin{bmatrix} \lambda_1 & 0 & 0 & 0 & \dots \\ 0 & \lambda_2 & 0 & 0 & \dots \\ 0 & 0 & \lambda_3 & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_n \end{bmatrix} \begin{bmatrix} s_1^\top \\ s_2^\top \\ \vdots \\ s_n^\top \end{bmatrix}. \quad (15)$$

Since

$$A^k = \underbrace{T\Lambda T^{-1}T\Lambda T^{-1} \dots T\Lambda T^{-1}}_{k \text{ times}} = T\Lambda^k T^{-1}, \quad (16)$$

we get

$$A^k = [t_1 \ t_2 \ \dots \ t_n] \begin{bmatrix} \lambda_1^k & 0 & 0 & 0 & \dots \\ 0 & \lambda_2^k & 0 & 0 & \dots \\ 0 & 0 & \lambda_3^k & 0 & \dots \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & 0 & \dots & \lambda_n^k \end{bmatrix} \begin{bmatrix} s_1^\top \\ s_2^\top \\ \vdots \\ s_n^\top \end{bmatrix}. \quad (17)$$

By defining the matrices

$$Z_h = t_h s_h^\top,$$

we can state the following

### Property

If  $A$  is diagonalizable, the state transition matrix  $A^k$  can be written as the sum of constant matrices  $Z_h$ , each multiplied by the discrete mode  $\lambda_h^k$

$$A^k = \sum_{h=1}^n t_h s_h^\top \lambda_h^k = \sum_{h=1}^n Z_h \lambda_h^k. \quad (18)$$

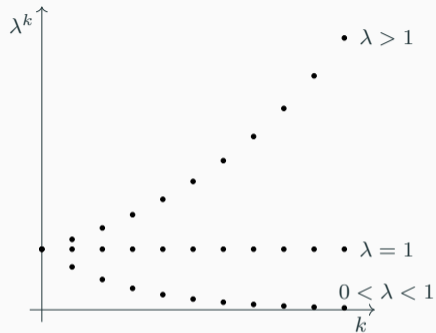
As in the continuous-time case, we can distinguish the two cases of the eigenvalue  $\lambda$  being real or complex.



## Modal analysis, diagonalizable $A$ (cont.)

For  $\lambda \in \mathbb{R}$ , the mode  $\lambda^k$  has the following behavior:

- if  $|\lambda| > 1$  the mode diverges;
- if  $|\lambda| < 1$  the mode vanishes;
- if  $|\lambda| = 1$  the mode has constant amplitude.



Based on the sign of  $\lambda$ , there is the further distinction:

- if  $\lambda > 0$  the mode is positive;
- if  $\lambda < 0$  the mode has **alternated sign**;
- if  $\lambda = 0$  the mode is null.

## Modal analysis, diagonalizable $A$ (cont.)

If some eigenvalue is complex we can, as before, order the eigenvalues:

$$\sigma(A) = \{ \underbrace{\lambda_1, \lambda_2, \dots, \lambda_r}_{\text{real}}, \underbrace{\lambda_{r+1}, \lambda_{r+1}^*}_{\text{conjugate}}, \underbrace{\lambda_{r+3}, \lambda_{r+3}^*}_{\text{conjugate}}, \dots, \underbrace{\lambda_{n-1}, \lambda_{n-1}^*}_{\text{conjugate}} \}$$

By taking the real and imaginary part of  $Z_h$  and expressing  $\lambda_h$  in polar form we have

$$\begin{aligned}\lambda_h &= \rho_h e^{j\theta_h} \\ Z_h &= M_h + jN_h.\end{aligned}$$

From Euler's formula

$$\lambda_h^k = \rho_h^k e^{j\theta_h k} = \rho_h^k [\cos(\theta_h k) + j \sin(\theta_h k)]$$

it can be obtained (with a rather long maths)

$$A^k = \sum_{h=1}^r Z_h \lambda_h^k + \sum_{r+1}^{n-1} \text{(step 2)} 2\rho_h^k [M_h \cos(\theta_h k) - N_h \sin(\theta_h k)]$$

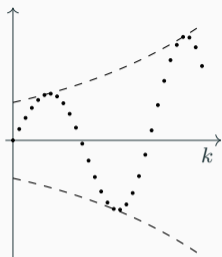
The following important property follows:

### Property

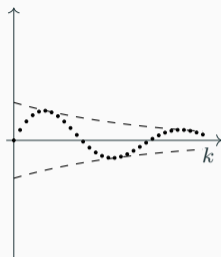
Each conjugate pair of eigenvalues  $\lambda$  and  $\lambda^*$  produces complex modes that result in real sequences of the form:

$$\rho^k \cos(\theta k) \quad \text{and} \quad \rho^k \sin(\theta k).$$

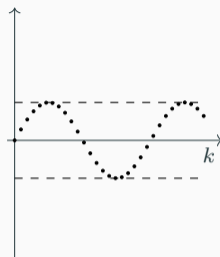
## Modal analysis, diagonalizable $A$ (cont.)



$\rho^k \sin(\theta k)$  for  $\rho > 1$ .



$\rho^k \sin(\theta k)$  for  $0 < \rho < 1$ .



$\rho^k \sin(\theta k)$  for  $\rho = 1$ .

The state transition matrix is thus governed by real exponential terms associated to real eigenvalues and oscillating (“pseudo-periodic”) modes associated to the conjugate pairs of eigenvalues. Depending on the modulus of the eigenvalue, the following behaviors may occur:

- if  $\rho > 1$  the amplitude of the oscillation diverges;
- if  $\rho < 1$  the amplitude of the oscillation vanishes;
- if  $\rho = 1$  the amplitude of the oscillation is constant.

As for the continuous-time case, if  $A$  is non-diagonalizable, we may resort to the Jordan form

$$A = TJT^{-1} \implies A^k = TJ^kT^{-1} \quad (19)$$

where

$$J = \text{diag}\{J_1, J_2, \dots, J_s\}$$

is the Jordan normal form of  $A$ .

Recalling the definition of the binomial coefficient  $\binom{k}{i} = \frac{k!}{i!(k-i)!}$ , the  $k$ th power of the block  $J_h$  can be written as

$$\begin{aligned} J_h^k &= (\lambda_h I + J_{h0})^k = \\ &= \lambda_h^k I + \binom{k}{1} \lambda_h^{k-1} J_{h0} + \binom{k}{2} \lambda_h^{k-2} J_{h0}^2 + \dots + \binom{k}{k-1} \lambda_h^1 J_{h0}^{k-1} + J_{h0}^k = \\ &= \sum_{i=0}^k \binom{k}{i} \lambda_h^{k-i} J_{h0}^i. \end{aligned}$$

## Modal analysis, non-diagonalizable $A$ (cont.)

Recall that  $J_{h0}^k = 0 \forall k \geq \nu_h$  ( $\nu_h$  being the size of the block  $J_{h0}$ ). Moreover, by definition,  $\binom{k}{i} = 0$  if  $k < i$ . Then

$$J_h^k = \sum_{i=0}^{\nu_h-1} \binom{k}{i} \lambda_h^{k-i} J_{h0}^i. \quad (20)$$

Observe that, for  $k \geq i$ , the binomial coefficient

$$\binom{k}{i} = \frac{k!}{i!(k-i)!} = \frac{k(k-1)(k-2)\dots(k-i+1)}{i!} \doteq p_i(k),$$

is a polynomial of degree  $i$  in the variable  $k$ . Thus, similarly to the continuous-time case, we get:

$$A^k = \sum_{h=1}^s \sum_{i=0}^{\nu_h-1} Z_{hi} p_i(k) \lambda_h^{k-i}, \quad (21)$$

where  $Z_{hi} = T_h J_{h0}^i S_h^\top$ .

If there exist Jordan blocks of size  $\geq 1$ , associated to  $\lambda$ , in the matrix  $A^k$ , the following modes will appear:

$$\lambda^k, \binom{k}{1} \lambda^{k-1}, \binom{k}{2} \lambda^{k-2}, \dots, \binom{k}{\nu-1} \lambda^{k-\nu+1}$$

where  $\nu = \nu(\lambda)$  is the degree of  $\lambda$ .

**Example.** Let the Jordan form of a matrix  $A$  be

$$\begin{bmatrix} 3 & 1 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 & 0 \\ 0 & 0 & 3 & 1 & 0 \\ 0 & 0 & 0 & 3 & 1 \\ 0 & 0 & 0 & 0 & 3 \end{bmatrix}$$

Then, the sole eigenvalue  $\lambda = 3$  has a degree  $\deg(\lambda) = 3$  and, as a consequence, the modes of  $A^k$  are:

$$3^k, \quad \binom{k}{1} 3^{k-1}, \quad \binom{k}{2} 3^{k-2}$$

## Transfer function (continuous-time)

---



Consider the time-invariant dynamic system:

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

and let  $x(0) = x_0$ . By applying the Laplace transform to both sides of the first equation we get:

$$sX(s) - x_0 = AX(s) + BU(s) \implies (sI - A)X(s) = x_0 + BU(s)$$

which implies

$$X(s) = (sI - A)^{-1}x_0 + (sI - A)^{-1}BU(s) \quad (22)$$

By substituting  $X(s)$  in the Laplace transform of the output equation we get

$$Y(s) = C(sI - A)^{-1}x_0 + [C(sI - A)^{-1}B + D]U(s) \quad (23)$$

Letting  $x_0 = 0$ , it follows that:

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

and  $W(s)$  is called the *transfer function*.

Let's analyze the structure of the transfer function:

$$W(s) = \begin{bmatrix} w_{11}(s) & \cdots & w_{1m}(s) \\ \vdots & & \vdots \\ w_{i1}(s) & \cdots & w_{im}(s) \\ \vdots & & \vdots \\ w_{p1}(s) & \cdots & w_{pm}(s) \end{bmatrix}$$

$W(s)$  is a  $p \times m$  matrix. If  $x_0 = 0$ , the  $i$ th component of the output vector is given by:

$$Y_i(s) = \sum_{r=1}^m w_{ir}(s)U_r(s) = w_{i1}(s)U_1(s) + w_{i2}(s)U_2(s) + \cdots$$

Thus:

$$\begin{matrix} x(0) = 0 \\ u_r(t) = 0, \quad r \neq j \end{matrix} \implies w_{ij}(s) = \frac{Y_i(s)}{U_j(s)}$$

In particular, if we take  $u_j(t) = \delta(t)$ , we have

$$U_j(s) = \mathcal{L}[u_j(t)] = \mathcal{L}[\delta(t)] = 1$$

hence

$$w_{ij}(s) = \frac{Y_i(s)}{U_j(s)} = Y_i(s)$$

In other words,  $w_{ij}(s)$  is the Laplace transform of the  $i$ th component of the output response to the unit impulse applied to the  $j$ th input. Thus

$$w_{ij}(s) = \mathcal{L}[w_{ij}(t)]$$

where  $w_{ij}(t)$  is the  $ij$ th element of the impulse response matrix  $W(t)$ .

Since the above holds for any pair  $i, j$ , it follows that

$$W(s) = \mathcal{L}[W(t)]$$

hence, the transfer function is the Laplace transform of the impulse response.

### Impulse response and transfer function

The impulse response and transfer function of the system

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

are given, respectively, by

$$W(t) = \mathcal{L}^{-1} [C(sI - A)^{-1}B + D]$$

and

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

In the following, we show that the entry  $w_{ij}(s)$  of a transfer function is a **proper rational function** (a rational function is a ratio of polynomials; it is *proper* if the degree of the numerator is less than or equal to the degree of the denominator; it is *strictly proper* if strict inequality holds).

Indeed:

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

and

$$(sI - A)^{-1} = \begin{bmatrix} s - a_{11} & -a_{12} & \cdots & -a_{1n} \\ -a_{21} & s - a_{22} & & \vdots \\ \vdots & & \ddots & \\ -a_{n1} & \cdots & & s - a_{nn} \end{bmatrix}^{-1}$$

The inverse can be expressed as:

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

where  $K(s)$  is the matrix of the algebraic complements (each of which is the determinant of an  $(n - 1) \times (n - 1)$  minor of  $sI - A$ ).

Clearly:

- $\varphi(s) \doteq \det(sI - A)$  is a polynomial of degree  $n$  (the characteristic polynomial of  $A$ )
- $K(s) = [k_{ij}(s)]$ ,  $i, j = 1, \dots, n$ , where  $k_{ij}(s)$  is a polynomial of degree  $< n$ ,  $\forall i, j$

As a consequence,

$$(sI - A)^{-1} = \frac{K(s)}{\varphi(s)}$$

is an  $n \times n$  matrix of strictly proper rational functions.

Therefore:

$$W(s) = C (sI - A)^{-1} B + D = C \frac{K(s)}{\varphi(s)} B + D = \frac{M(s)}{\varphi(s)} + D = \frac{N(s)}{\varphi(s)}$$

where the entries of  $N(s)$  are polynomials of degree  $\leq n$ :

$$\deg(n_{ij}(s)) \leq n$$

The strict inequality holds if and only if the corresponding entry of  $D$  is zero, i.e.  $d_{ij} = 0$ .

In summary,  $W(s)$  is strictly proper (all its entries are strictly proper) if and only if  $D = 0$  (i.e. the system is strictly proper).

Given

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases}$$

let  $\hat{x} \doteq T^{-1}x$ , where  $T \in \mathbb{R}^{n \times n}$  is a generic non-singular  $n \times n$  matrix ( $\det(T) \neq 0$ ). Then, an equivalent state-space description is given by:

$$\begin{cases} \dot{\hat{x}}(t) = T^{-1}\dot{x}(t) = T^{-1}AT\hat{x}(t) + T^{-1}Bu(t) = \hat{A}\hat{x}(t) + \hat{B}u(t) \\ y(t) = CT\hat{x}(t) + Du(t) = \hat{C}\hat{x}(t) + Du(t) \end{cases}$$

In other words:

$$\begin{cases} \dot{x}(t) = Ax(t) + Bu(t) \\ y(t) = Cx(t) + Du(t) \end{cases} \iff \begin{cases} \dot{\hat{x}}(t) = \hat{A}\hat{x}(t) + \hat{B}u(t) \\ y(t) = \hat{C}\hat{x}(t) + Du(t) \end{cases}$$



$$\begin{aligned}\hat{W}(s) &= \hat{C}(sI - \hat{A})^{-1}\hat{B} + \hat{D} \\ &= (CT) (sI - T^{-1}AT)^{-1} (T^{-1}B) + D \\ &= CT (sT^{-1}T - T^{-1}AT)^{-1} T^{-1}B + D \\ &= CT [T^{-1}(sI - A)T]^{-1} T^{-1}B + D \\ &= CT [T^{-1}(sI - A)^{-1}T] T^{-1}B + D \\ &= C (sI - A)^{-1} B + D \\ &= W(s)\end{aligned}$$

Hence, the transfer function is invariant to change of basis.

## Transfer function (discrete-time)

---

Consider the time-invariant dynamic system:

$$\begin{cases} x(k+1) = Ax(k) + Bu(k) \\ y(k) = Cx(k) + Du(k) \end{cases}$$

and let  $x(0) = x_0$ . By applying the  $\mathcal{Z}$ -transform to both sides of the first equation we get:

$$z[X(z) - x_0] = AX(z) + BU(z) \implies (zI - A)X(z) = z x_0 + BU(z)$$

which implies

$$X(z) = (zI - A)^{-1} z x_0 + (zI - A)^{-1} BU(z) \quad (24)$$

By substituting  $X(z)$  in the  $\mathcal{Z}$ -transform of the output equation we get

$$Y(z) = C(zI - A)^{-1} z x_0 + [C(zI - A)^{-1} B + D] U(z) \quad (25)$$

Letting  $x_0 = 0$ , it follows that:

$$Y(z) = [C(zI - A)^{-1} B + D] U(z) = W(z) U(z)$$

and  $W(z)$  is called the *transfer function*.

Let's analyze the structure of the transfer function:

$$W(z) = \begin{bmatrix} w_{11}(z) & \cdots & w_{1m}(z) \\ \vdots & & \vdots \\ w_{i1}(z) & \cdots & w_{im}(z) \\ \vdots & & \vdots \\ w_{p1}(z) & \cdots & w_{pm}(z) \end{bmatrix}$$

$W(z)$  is a  $p \times m$  matrix. If  $x_0 = 0$ , the  $i$ th component of the output vector is given by:

$$Y_i(z) = \sum_{r=1}^m w_{ir}(z)U_r(z) = w_{i1}(z)U_1(z) + w_{i2}(z)U_2(z) + \cdots$$

Thus:

$$\begin{matrix} x(0) = 0 \\ u_r(k) = 0, \quad r \neq j \end{matrix} \implies w_{ij}(z) = \frac{Y_i(z)}{U_j(z)}$$

In particular, if we take  $u_j(k) = \delta(k)$ , we have

$$U_j(z) = \mathcal{Z}[u_j(k)] = \mathcal{Z}[\delta(k)] = 1$$

hence

$$w_{ij}(z) = \frac{Y_i(z)}{U_j(z)} = Y_i(z)$$

In other words,  $w_{ij}(z)$  is the  $\mathcal{Z}$ -transform of the  $i$ th component of the output response to the unit impulse applied to the  $j$ th input. Thus

$$w_{ij}(z) = \mathcal{Z}[w_{ij}(k)]$$

where  $w_{ij}(k)$  is the  $ij$ th element of the impulse response matrix  $W(k)$ .

Since the above holds for any pair  $i, j$ , it follows that

$$W(z) = \mathcal{Z}[W(k)]$$

hence, the transfer function is the  $\mathcal{Z}$ -transform of the impulse response.

### Impulse response and transfer function

The impulse response and transfer function of the system

$$\begin{cases} x(k+1) = Ax(k) + Bu(k) \\ y(k) = Cx(k) + Du(k) \end{cases}$$

are given, respectively, by

$$W(k) = \mathcal{Z}^{-1} [C(zI - A)^{-1}B + D]$$

and

$$W(z) = C(zI - A)^{-1}B + D$$

In the following, we show that the entry  $w_{ij}(z)$  of a transfer function is a proper rational function (a rational function is a ratio of polynomials; it is *proper* if the degree of the numerator is less than or equal to the degree of the denominator; it is *strictly proper* if strict inequality holds).

Indeed:

$$W(z) = C(zI - A)^{-1}B + D$$

and

$$(zI - A)^{-1} = \begin{bmatrix} z - a_{11} & -a_{12} & \cdots & -a_{1n} \\ -a_{21} & z - a_{22} & & \vdots \\ \vdots & & \ddots & \\ -a_{n1} & \cdots & & z - a_{nn} \end{bmatrix}^{-1}$$

The inverse can be expressed as:

$$(zI - A)^{-1} = \frac{1}{\det(zI - A)} K(z)$$

where  $K(z)$  is the matrix of the algebraic complements (each of which is the determinant of an  $(n - 1) \times (n - 1)$  minor of  $zI - A$ ).

Clearly:

- $\varphi(z) \doteq \det(zI - A)$  is a polynomial of degree  $n$  (the characteristic polynomial of  $A$ )
- $K(z) = [k_{ij}(z)]$ ,  $i, j = 1, \dots, n$ , where  $k_{ij}(z)$  is a polynomial of degree  $< n$ ,  $\forall i, j$

As a consequence,

$$(zI - A)^{-1} = \frac{K(z)}{\varphi(z)}$$

is an  $n \times n$  matrix of strictly proper rational functions.



Therefore:

$$W(z) = C(zI - A)^{-1}B + D = C \frac{K(z)}{\varphi(z)}B + D = \frac{M(z)}{\varphi(z)} + D = \frac{N(z)}{\varphi(z)}$$

where the entries of  $N(z)$  are polynomials of degree  $\leq n$ :

$$\deg(n_{ij}(z)) \leq n$$

The strict inequality holds if and only if the corresponding entry of  $D$  is zero, i.e.  $d_{ij} = 0$ .

In summary,  $W(z)$  is strictly proper (all its entries are strictly proper) if and only if  $D = 0$  (i.e. the system is strictly proper).

Given

$$\begin{cases} x(k+1) = Ax(k) + Bu(k) \\ y(k) = Cx(k) + Du(k) \end{cases}$$

Let  $\hat{x} := T^{-1}x$ , where  $T \in \mathbb{R}^{n \times n}$  is a generic non-singular  $n \times n$  matrix ( $\det(T) \neq 0$ ). Then, the equivalent state-space description is given by:

$$\begin{cases} \hat{x}(k+1) = T^{-1}x(k+1) = T^{-1}AT\hat{x}(k) + T^{-1}Bu(k) = \hat{A}\hat{x}(k) + \hat{B}u(k) \\ y(k) = CT\hat{x}(k) + Du(k) = \hat{C}\hat{x}(k) + Du(k) \end{cases}$$

In other words:

$$\begin{cases} x(k+1) = Ax(k) + Bu(k) \\ y(k) = Cx(k) + Du(k) \end{cases} \iff \begin{cases} \hat{x}(k+1) = \hat{A}\hat{x}(k) + \hat{B}u(k) \\ y(k) = \hat{C}\hat{x}(k) + Du(k) \end{cases}$$

$$\begin{aligned}\hat{W}(z) &= \hat{C}(zI - \hat{A})^{-1}\hat{B} + \hat{D} \\ &= (CT) (zI - T^{-1}AT)^{-1} (T^{-1}B) + D \\ &= CT (zT^{-1}T - T^{-1}AT)^{-1} T^{-1}B + D \\ &= CT [T^{-1}(zI - A)T]^{-1} T^{-1}B + D \\ &= CT [T^{-1}(zI - A)^{-1}T] T^{-1}B + D \\ &= C (zI - A)^{-1} B + D \\ &= W(z)\end{aligned}$$

Hence, the transfer function is invariant to change of basis.

Given the LTI discrete-time system, having two inputs and one output:

$$\begin{cases} x(k+1) = \begin{bmatrix} 0 & 1 \\ -1 & -2 \end{bmatrix} x(k) + \begin{bmatrix} 0 & -1/2 \\ 1 & 1/2 \end{bmatrix} u(k) \\ y(k) = [-3 \ 3] x(k) \end{cases}$$

the transfer function is a  $1 \times 2$  matrix:

$$\begin{aligned} W(z) &= [-3 \ 3] \begin{bmatrix} z & -1 \\ 1 & z+2 \end{bmatrix}^{-1} \begin{bmatrix} 0 & -1/2 \\ 1 & 1/2 \end{bmatrix} \\ &= [-3 \ 3] \frac{1}{(z+1)^2} \begin{bmatrix} z+2 & 1 \\ -1 & z \end{bmatrix} \begin{bmatrix} 0 & -1/2 \\ 1 & 1/2 \end{bmatrix} \\ &= \begin{bmatrix} -\frac{3}{z+1} & \frac{3(z-1)}{(z+1)^2} \end{bmatrix} \begin{bmatrix} 0 & -1/2 \\ 1 & 1/2 \end{bmatrix} = \begin{bmatrix} \frac{3(z-1)}{(z+1)^2} & \frac{3}{z+1} \end{bmatrix} \end{aligned}$$

## References

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Antsaklis, P. J. and Michel, A. N. (2006). *Linear Systems*. Springer Science & Business Media.

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Lecture 2  
Solutions to linear systems

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