

EE 640 - The Transfer Function

Note Title

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The Transfer Function of a Realization

We have seen that a LTI system (i.e. a ODE) may be represented by

- A transfer function
- State Variable Representations

Q) How are these two related?

The transfer function of a state Representation:

Suppose we have a realization:

$$* \begin{cases} \dot{x} = Ax + Bu \\ y = Cx \end{cases} \text{ using the usual notation.}$$

Define $X(s) = \mathcal{L}\{x(t)\} = \int_0^{\infty} x(t)e^{-st} dt$
One-sided Laplace Transform of $x(t)$

Similarly $U(s) = \mathcal{L}\{u(t)\} = \int_0^{\infty} u(t)e^{-st} dt$

From (*) we have:

$$\mathcal{L}\{\dot{x}(t)\} = A \mathcal{L}\{x(t)\} + B \mathcal{L}\{u(t)\}$$

$$\text{or } sX(s) - X(0^-) = AX(s) + BU(s)$$

$$\text{or } (sI - A)X(s) = X(0) + BU(s)$$

$$** \therefore \boxed{X(s) = (sI - A)^{-1} X(0) + (sI - A)^{-1} BU(s)}$$

(Note that we are conveniently bypassing questions about invertibility of $(sI - A)$!)

Now consider the L.T. of the output equation:

$$Y(s) = \mathcal{L}\{y(t)\} = \mathcal{L}\{Cx(t)\} = C X(s)$$

Using $(*)$,

$$Y(s) = C (sI - A)^{-1} x(0) + C (sI - A)^{-1} B U(s)$$

For getting the transfer function matrix or transfer matrix, set $x(0) = 0$

$$\text{Then } Y(s) = C (sI - A)^{-1} B U(s)$$

$$\text{or } Y(s) = G(s) U(s)$$

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

For a SISO system $G(s)$ is a scalar transfer function.

It should be clear that the state space representation $(*)$ and the equation $(**)$ may be equivalently studied to answer any questions about the original system.

For example: if we are interested in computing the trajectories of

$x(t)$ it can be calculated directly from $(*)$ or through $(**)$ and then use an Inverse L.T.

Let us consider SISO systems for the rest of this section

Then
$$\frac{Y(s)}{U(s)} = G(s) = C(sI - A)^{-1}B$$

$$= \frac{1}{\det(sI - A)} \cdot C [adj(sI - A)] B$$

\downarrow \downarrow \downarrow \downarrow
 1×1 $1 \times n$ $n \times n$ $n \times 1$

Digression :

Let B be a $n \times n$ matrix

$$B = \begin{bmatrix} b_{11} & b_{12} & \dots & b_{1n} \\ b_{21} & b_{22} & \dots & b_{2n} \\ \vdots & & & \\ b_{n1} & \dots & & b_{nn} \end{bmatrix}$$

A minor determinant of B is the determinant of the $(n-1) \times (n-1)$ matrix obtained when one row & one column of B are deleted.

The minor determinant

$$m_{ij} = \det [B \text{ minus row } i, \text{ column } j]$$

The cofactor B_{ij} of the entry b_{ij} of B is

$$B_{ij} = (-1)^{i+j} m_{ij}$$

The adjoint matrix of B is

$$\text{adj}(B) = \begin{bmatrix} B_{11} & B_{12} & \dots & B_{1n} \\ B_{21} & B_{22} & \dots & B_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ B_{n1} & B_{n2} & \dots & B_{nn} \end{bmatrix}^T$$

Then,
$$B^{-1} = \frac{1}{\det B} \text{adj}(B)$$

NOTE that the term $\det(sI - A)$ occurs in the denominator of the transfer function and happens to be of special interest to us.

In fact we give it a name:

Characteristic Polynomial

Let A be a $n \times n$ matrix. The characteristic polynomial of A

is $a(s) = \det(sI - A)$

$$= s^n + a_1 s^{n-1} + a_2 s^{n-2} + \dots + a_n$$

[Exercise: This answers the doubt we had earlier about inverting $(sI - A)$. How?]

We can factor $a(s)$ over the complex numbers (but not always over the reals)

$$a(s) = (s - \lambda_1)(s - \lambda_2) \dots (s - \lambda_n)$$

Then $(\lambda_1, \lambda_2, \dots, \lambda_n)$ are the eigenvalues of A.

Exercise: An eigenvalue of A is also defined as a scalar corresponding to which there is a non-zero vector v so that

$$Av = \lambda v$$

Here v is an eigenvector of A corresponding to the eigenvalue λ .

How are these two definitions same?

So, we have a fairly good idea about the denominators of $\begin{pmatrix} X & X \\ X & X \end{pmatrix}$.
 What about the numerators.

The resolvent formula

$$\begin{aligned} \text{adj}(sI - A) &= A^{n-1} + (s + a_1)A^{n-2} \\ &+ (s^2 + a_1s + a_2)A^{n-3} + \dots + \dots \\ &\dots + (s^{n-1} + a_1s^{n-2} + \dots + a_{n-1})I \end{aligned}$$

OR, re-arranging the terms:

$$\begin{aligned} \text{adj}(sI - A) &= s^{n-1}I + s^{n-2}(A + a_1I) \\ &+ s^{n-3}(A^2 + a_1A + a_2I) + \dots + \dots \\ &\dots + (A^{n-1} + a_1A^{n-2} + \dots + a_{n-1}I) \end{aligned}$$

Using this formula let us simplify equation $\begin{pmatrix} X & X \\ X & X \end{pmatrix}$.

$$\begin{aligned} G(s) &= \frac{1}{\det(sI - A)} C[\text{adj}(sI - A)]B \\ &= \frac{1}{s^n + a_1s^{n-1} + \dots + a_n} C \left[s^{n-1}I + s^{n-2}(A + a_1I) \right. \\ &\quad \left. + \dots + (A^{n-1} + \dots + a_{n-1}I) \right] B \end{aligned}$$

Note : $CB \equiv \text{scalar}$
 $CA^i B \equiv \text{scalar}$

Hence,

$$G(s) = \frac{s^{n-1}(CB) + s^{n-2}(CAB + a_1 CB) + \dots}{s^n + a_1 s^{n-1} + \dots + a_n}$$

= polynomial of degree $\leq n-1$
polynomial of degree = n .

$$=: \frac{b(s)}{a(s)} \quad \left(\begin{array}{l} \text{i.e. name the numerators} \\ b(s) \end{array} \right)$$

Now cancel all the common polynomial factors between $b(s)$ and $a(s)$ to get

$$G(s) = \frac{b_r(s)}{a_r(s)}$$

Defⁿ : The roots of $a_r(s) = 0$ are the poles of $G(s)$

Defⁿ : The roots of $b_r(s) = 0$ are the zeros of $G(s)$

Since $a_r(s)$ is a factor of the characteristic polynomial, the poles of $G(s)$ are eigenvalues of the matrix A .

If $a_n(s) \neq a(s)$, then

$$\deg(a_n(s)) < \deg(a(s))$$

The degree of $a_n(s)$ gives us the minimum number of integrators needed to implement $G(s)$.

So if $\deg a_n(s) < \deg a(s)$, then the same transfer function could have been implemented with less than n integrators.

If $\deg a_n(s) = \deg a(s)$ [i.e. there are no cancellations] then we say that the realization is a minimal realization.

Impulse Response of $G(s)$

We have seen that the transfer function $G(s)$ has already helped us to identify the minimum number of integrators needed to implement the underlying differential equation.

Another important use of $G(s)$ is to analytically calculate time domain responses of the realization.

Let us further investigate the quantity

$(sI - A)^{-1}$ in the formula for $G(s)$

Digression: Geometric Series

$$(1 - \alpha)^{-1} = 1 + \alpha + \alpha^2 + \alpha^3 + \dots$$

In this form the geometric series expansion is valid for a $n \times n$ matrix.

$$\begin{aligned}(sI - A)^{-1} &= \frac{1}{s} \left(I - \frac{1}{s} A \right)^{-1} \\ &= \frac{1}{s} \left[I + \frac{1}{s} A + \frac{1}{s^2} A^2 + \dots \right] \\ &= \frac{1}{s} I + \frac{1}{s^2} A + \frac{1}{s^3} A^2 + \dots\end{aligned}$$

Now apply the inverse Laplace Transform

$$\begin{aligned}\mathcal{L}^{-1}(sI - A)^{-1} &= \mathcal{L}^{-1}\left(\frac{1}{s}\right)I + \mathcal{L}^{-1}\left(\frac{1}{s^2}\right)A \\ &\quad + \mathcal{L}^{-1}\left(\frac{1}{s^3}\right)A^2 + \dots\end{aligned}$$

$$\left[\text{Recall that } \mathcal{L}^{-1}\left(\frac{1}{s^{n+1}}\right) = \frac{t^n}{n!}, t \geq 0 \right]$$

$$= I + tA + \frac{t^2}{2!} A^2 + \frac{t^3}{3!} A^3 + \dots$$

Now where have we seen a similar form? Recall the series expansion of e^{α} for a scalar α .

$$e^x = 1 + x + \frac{x^2}{2!} + \frac{x^3}{3!} + \dots$$

This looks so similar to the above formula, that we DEFINE the matrix exponential:

$$\begin{aligned} e^{At} &:= I + At + A^2 \frac{t^2}{2!} + A^3 \frac{t^3}{3!} + \dots \\ &= \sum_{i=0}^{\infty} \frac{t^i}{i!} A^i \end{aligned}$$

From the above formula, for a $n \times n$ matrix A , e^{At} is also a $n \times n$ matrix.

$$\boxed{\mathcal{L}^{-1}\{(sI - A)^{-1}\} = e^{At} \quad t \geq 0}$$

The Impulse Response: Using the above formula for $G(s)$, the impulse response follows easily:

$$\mathcal{L}\{f(t)\} = 1$$

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

$$y(t) = \mathcal{L}^{-1}\{Y(s)\} = \mathcal{L}^{-1}\{C(sI - A)^{-1} B\}$$

$$= C \mathcal{L}^{-1}\{(sI - A)^{-1}\} B$$

$$= C e^{At} B$$

The impulse response is often denoted by $h(t)$. So

$$h(t) = ce^{At}B$$

Invariants under Similarity Transformation

We are now in a position to answer a question we had asked in the previous chapter: namely, what remains the same for two similar realizations. We will check some of the quantities we have learned until now.

Claim 1: Similar realizations have the same characteristic polynomial.

Proof: Consider the two realizations:

$$\textcircled{1} \left(\begin{array}{l} \dot{\bar{x}} = \bar{A} \bar{x} + \bar{B}u \\ y = \bar{C} \bar{x} \end{array} \right) \quad \left| \quad \begin{array}{l} \dot{x} = Ax + Bu \\ y = Cx \end{array} \right) \textcircled{2}$$

Connected by a similarity transform T
i.e. $x = T\bar{x} \Leftrightarrow \bar{A} = T^{-1}AT$

Consider the characteristic polynomial of $\textcircled{1}$:

$$c(s) = \det(sI - \bar{A})$$

$$\begin{aligned}
&= \det (sI - T^{-1}AT) \\
&= \det (sT^{-1}IT - T^{-1}AT) \\
&= \det [T^{-1}(sI - A)T] \\
&= (\det T^{-1})(\det (sI - A))(\det T) \\
&= \underbrace{(\det T^{-1})(\det T)}_{=1} \det (sI - A) \\
&= \det (sI - A)
\end{aligned}$$

$= a(s) \equiv$ the characteristic polynomial of realization (2).

Claim 2: The transfer function (of the realization) is invariant under a similarity transformation

$$\begin{aligned}
\text{Let } G_1(s) &= C(sI - A)^{-1}B \\
G_2(s) &= \bar{C}(sI - \bar{A})^{-1}\bar{B}
\end{aligned}$$

$$\begin{aligned}
\text{where } \bar{A} &= T^{-1}AT ; \bar{B} = T^{-1}B \\
\bar{C} &= CT
\end{aligned}$$

$$\begin{aligned}
G_2(s) &= \bar{C}(sI - \bar{A})^{-1}\bar{B} \\
&= \bar{C} \left[\frac{1}{s}I + \frac{1}{s^2}\bar{A} + \frac{1}{s^3}\bar{A}^2 + \dots \right] \bar{B} \\
&= CT \left[\frac{1}{s}I + \frac{1}{s^2}(T^{-1}AT) + \frac{1}{s^3}(T^{-1}AT)^2 + \dots \right]
\end{aligned}$$

$$\dots \dots \dots] T^{-1} B$$

Note that:

$$(T^{-1} A T)^i = \underbrace{(T^{-1} A T)(T^{-1} A T) \dots \dots (T^{-1} A T)}_{i \text{ times}}$$

$$= T^{-1} A^i T$$

Hence

$$\begin{aligned} H(s) &= C T \left[\frac{1}{s} I + \frac{1}{s^2} (T^{-1} A T) + \frac{1}{s^3} (T^{-1} A^2 T) \right. \\ &\quad \left. + \dots \dots \dots \right] T^{-1} B \\ &= C \left[\frac{1}{s} (T T^{-1}) + \frac{1}{s^2} T (T^{-1} A T) T^{-1} + \dots \dots \right] T^{-1} B \\ &= C \left[\frac{1}{s} I + \frac{1}{s^2} A + \frac{1}{s^3} A^2 + \dots \dots \right] B \\ &= C (sI - A)^{-1} B \equiv G_2(s) \end{aligned}$$

Some Nomenclature : Markov Parameters

Consider the formula for

$$\begin{aligned} G(s) &= C \left[\frac{1}{s} I + \frac{1}{s^2} A + \frac{1}{s^3} A^2 + \dots \right] B \\ &= \frac{1}{s} (CB) + \frac{1}{s^2} (CAB) + \frac{1}{s^3} (CA^2 B) + \dots \dots \end{aligned}$$

The scalar sequence :

$$CB, CAB, CA^2B, \dots$$

are called Markov Parameters m_i

$$m_i = CA^{i-1}B \quad i=1, 2, \dots$$

Exercise : Prove that the Markov parameters are invariant under similarity transformation.

Claim 3 : The impulse response $h(t)$ is invariant under similarity transformation.

HINT : The impulse response

$$\begin{aligned} h(t) &= \mathcal{L}^{-1} \{ G(s) \cdot 1 \} \\ &= \mathcal{L}^{-1} \left[C \left(\frac{1}{s} I + \frac{1}{s^2} A + \frac{1}{s^3} A^2 + \dots \right) B \right] \end{aligned}$$

Some ^{related} Linear Algebra FACTS

1) A (strange) property of the characteristic polynomial.

Suppose we have a polynomial

$$p(s) = p_0 s^m + p_1 s^{m-1} + \dots + p_{m-1} s + p_m$$

We CAN substitute the $n \times n$ matrix A for s :

$$p(A) = p_0 A^m + p_1 A^{m-1} + \dots + p_{m-1} A + p_m I$$

FACT (Cayley Hamilton Theorem):

Let A be a $n \times n$ matrix with characteristic polynomial $a(s)$. Then

$$a(A) = O_{n \times n}$$

namely, if $a(s) = s^n + a_1 s^{n-1} + \dots + a_n$
 then $a(A) = A^n + a_1 A^{n-1} + \dots + a_n I = O_{n \times n}$

Proof: let $B = \text{adj}(sI - A)$

$$(sI - A) \cdot B = \det(sI - A) I = a(s) I$$

We have seen that (Recall the resolvent form)

$$B = \sum_{i=0}^{n-1} s^i B_i^0$$

$$\text{Hence, } a(s) I = (sI - A) B$$

$$\begin{aligned} &= (sI - A) \sum_{i=0}^{n-1} s^i B_i^0 \\ &= \sum_{i=0}^{n-1} s^{i+1} B_i^0 - \sum_{i=0}^{n-1} s^i A B_i^0 \end{aligned}$$

$$\begin{aligned}
&= s^n B_{n-1} + \sum_{i=1}^{n-1} s^i (B_{i-1} - AB_i) - AB_0 \\
&= s^n B_{n-1} + s^{n-1} (B_{n-2} - AB_{n-1}) + \dots \\
&\quad + s (B_0 - AB_1) - AB_0 \quad \text{--- (1)}
\end{aligned}$$

Now, $a(s) = s^n + a_1 s^{n-1} + \dots + a_n$

as $a(s)I = s^n I + s^{n-1} a_1 I + \dots + a_n I$ (2)

Matching coefficients in (1) & (2),

$$\left. \begin{aligned}
B_{n-1} &= I \\
B_{n-2} - AB_{n-1} &= a_1 \\
B_{n-3} - AB_{n-2} &= a_2 \\
&\vdots \\
B_0 - AB_1 &= a_{n-1} \\
-AB_0 &= a_n
\end{aligned} \right\} \begin{aligned}
&\times A^n \\
&\times A^{n-1} \\
&\times A^{n-2} \\
&\vdots \\
&\times A \\
&\times I
\end{aligned}$$

$$0 = A^n + a_1 A^{n-1} + a_2 A^{n-2} + \dots + a_n I$$

An Immediate Consequence:

ANY power of A is a linear combination of $A^{n-1}, A^{n-2}, \dots, A, I$

Proof: $A^n = -a_1 A^{n-1} - \dots - a_n I$

$$\begin{aligned}
\text{Then } A^{n+1} &= AA^n \\
&= A \left[-a_1 A^{n-1} - \dots - a_n I \right] \\
&= -a_1 A^n - a_2 A^{n-1} - \dots - a_n A \\
&= -a_1 \left[-a_1 A^{n-1} - \dots - a_n I \right] - a_2 A^{n-1} - \dots - a_n A
\end{aligned}$$

Similarly for any powers of A .

2) The minimal polynomial

We saw that:

$$(sI - A)^{-1} = \underbrace{\frac{1}{\det(sI - A)}}_{\text{polynomial}} \underbrace{\text{adj}(sI - A)}_{\text{polynomial matrix}}$$

In general, we could have factors of the characteristic polynomial that are common to all entries of the polynomial matrix $\text{adj}(sI - A)$.

Cancelling all these common factors, we get

$$(sI - A)^{-1} = \underbrace{\frac{1}{\mu(s)}}_{\text{polynomial}} \underbrace{\Gamma(s)}_{\text{polynomial matrix}}$$

$\mu(s) =:$ The minimal polynomial of A .

FACT: The minimal polynomial $\mu(s)$ is the polynomial of least degree so that

$$\mu(A) = O_{n \times n}$$

FACT: By Cayley Hamilton thm, the minimal polynomial always a factor of the characteristic polynomial.