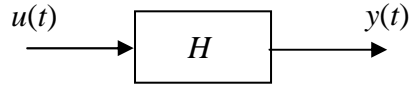


07. Linear Time-Invariant Systems

A system is depicted on the right which consists of input $u(t)$, output $y(t)$ and the system H itself. In reality, if $y(t)$ is a result generated by the input $u(t)$, then it is impossible that $y(t)$ is received earlier than time that $u(t)$ sent into the system. For example, $y(t)=0$ for $t<0$ if $u(t)$ starts to excite the system at $t=0$. This kind of system is called the causal system. A system operating in real time must be a causal system. However, it is not necessary for a system which is not processed in real time. For example, the process of an image taken at a specific time is not causal.



In system engineering, the most common system is the linear time-invariant (LTI) system, which satisfies the superposition and is independent to the time shifting. For simplicity in explanation, the system operation is often described as below:

$$y(t) = H[u(t)] \quad (1)$$

which implies the output $y(t)$ is a function of $u(t)$. Let $y_1(t)$ and $y_2(t)$ be the outputs of $u_1(t)$ and $u_2(t)$, respectively, i.e.,

$$y_1(t) = H[u_1(t)] \quad (2)$$

$$y_2(t) = H[u_2(t)] \quad (3)$$

If the system satisfies

$$\begin{aligned} H[\alpha_1 u_1(t) + \alpha_2 u_2(t)] &= \alpha_1 H[u_1(t)] + \alpha_2 H[u_2(t)] \\ &= \alpha_1 y_1(t) + \alpha_2 y_2(t) \end{aligned} \quad (4)$$

for any constant α_1 and α_2 , then the system is a linear system. The property in (4) is the so-called superposition. If the number of inputs is more than two, (4) can be further extended as

$$H\left[\sum_{i=1}^m \alpha_i u_i(t)\right] = \sum_{i=1}^m \alpha_i H[u_i(t)] = \sum_{i=1}^m \alpha_i y_i(t) \quad (5)$$

where $y_i(t) = H[u_i(t)]$, $i=1,2,\dots,m$.

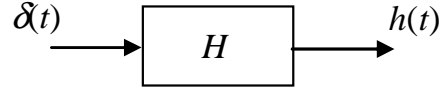
As for the time-invariant property, it states that the output of a system is shifted in time by τ when the input is shifted by τ , i.e.,

$$y(t-\tau) = H[u(t-\tau)] \quad (6)$$

Clearly, the operation of a time-invariant system does not depend on the time it is excited.

To sum up, an LTI system must satisfy the superposition (5) and the time-invariant property (6).

It is known that an LTI system is often represented by its impulse response $h(t)$, i.e., $y(t)=h(t)$ for $u(t)=\delta(t)$, or



$$h(t) = H[\delta(t)] \tag{7}$$

Before introducing the relation of $u(t)$ and $y(t)$, let's express $u(t)$ in terms of $\delta(t)$ as

$$u(t) = \int_{-\infty}^{\infty} u(\tau)\delta(t-\tau)d\tau \tag{8}$$

which can be further changed into the following expression:

$$u(t) = \lim_{\Delta\tau \rightarrow 0} \sum_{k=-\infty}^{\infty} u(k\Delta\tau)\delta(t-k\Delta\tau)\Delta\tau \tag{9}$$

or

$$u(t) = \lim_{\Delta\tau \rightarrow 0} \sum_{k=-\infty}^{\infty} \alpha_k u_k(t) \tag{10}$$

Note that the coefficients $\alpha_k = u(k\Delta\tau)\Delta\tau$ do not depend on t and $u_k(t) = \delta(t-k\Delta\tau)$ is the k -th component of $u(t)$. Clearly, $u(t)$ has been decomposed into infinite impulse functions.

Now, let's show that for an LTI system the output $y(t)$ equals the convolution of the input $u(t)$ and the impulse response $h(t)$, i.e.,

$$y(t) = h(t)*u(t) = u(t)*h(t) \tag{11}$$

where the convolution is commutative. First, from (1), (8) and (10) we have

$$y(t) = H[u(t)] = H\left[\int_{-\infty}^{\infty} u(\tau)\delta(t-\tau)d\tau\right] = \lim_{\Delta\tau \rightarrow 0} H\left[\sum_{k=-\infty}^{\infty} \alpha_k u_k(t)\right] \tag{12}$$

Since the system is linear and thus

$$y(t) = \lim_{\Delta\tau \rightarrow 0} \sum_{k=-\infty}^{\infty} \alpha_k H[u_k(t)] \tag{13}$$

with $\alpha_k = u(k\Delta\tau)\Delta\tau$ and $u_k(t) = \delta(t-k\Delta\tau)$. Hence,

$$y(t) = \lim_{\Delta\tau \rightarrow 0} \sum_{k=-\infty}^{\infty} u(k\Delta\tau)H[\delta(t-k\Delta\tau)]\Delta\tau \tag{14}$$

which leads to the integral form as

$$y(t) = \int_{-\infty}^{\infty} u(\tau)H[\delta(t-\tau)]d\tau \tag{15}$$

Further applying the time-invariant property (6), we have

$$h(t-\tau) = H[\delta(t-\tau)] \tag{16}$$

where $h(t)$ is the impulse response given in (7). Now, Substituting (16) into (15) yields

$$y(t) = \int_{-\infty}^{\infty} u(\tau)h(t-\tau)d\tau = u(t) * h(t) = h(t) * u(t) \tag{17}$$

which proves (11).

According to the property of convolution in time of Fourier transform, if $Y(\omega)$, $H(\omega)$ and $U(\omega)$ are the Fourier transforms of $y(t)$, $h(t)$ and $u(t)$, then

$$Y(\omega) = H(\omega)U(\omega) \tag{18}$$

which has been widely used in filter design.

Consider the case that $u(t)$ starts to excite the system at $t=0$, i.e., $u(t)=0$ for $t<0$. Since the impulse response $h(t)$ is a special output related to the input $\delta(t)$, we also know that $h(t)=0$ for $t<0$ based on the causality. As a result, the convolution in (17) is rewritten as

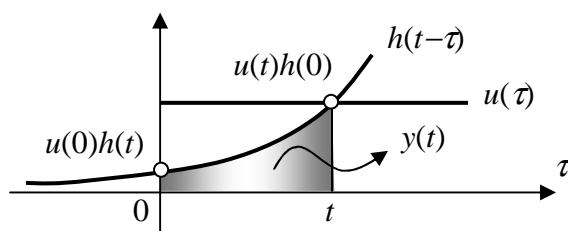
$$y(t) = \int_0^t u(\tau)h(t-\tau)d\tau \tag{19}$$

which is a kind of running integral. Clearly, the output $y(t)$ only depends on the input $u(\tau)$ for $0 \leq \tau \leq t$. Indeed, (19) represents a causal system. Most importantly, $y(t)$ requires the history of $u(t)$ and therefore the system must be possessed of ‘memory’ components to record the past of $u(t)$.

Later, we will learn that the impulse response of an LTI system often consists of decreased exponential function, such as e^{-at} where $a>0$. To demonstrate the calculation of convolution, let’s use the the following example with $u(t)=1$ and $h(t)=e^{-2t}$ for $t \geq 0$. Hence, the output will be obtained as

$$y(t) = \int_0^t u(\tau)h(t-\tau)d\tau = \int_0^t e^{-2(t-\tau)} d\tau = 0.5(1 - e^{-2t}) \tag{20}$$

which is depicted in the following figure.



To explain the convolution in the above figure, according to (14) we change the output in (20) as

$$y(t) = \lim_{\Delta\tau \rightarrow 0} \sum_{k=0}^n u(k\Delta\tau)h(t-k\Delta\tau)\Delta\tau \quad (21)$$

where $t=n\Delta\tau$. That means

$$y(t) = u(0)h(n\Delta\tau)\Delta\tau + \cdots + u(k\Delta\tau)h((n-k)\Delta\tau)\Delta\tau \quad (22)$$

$$+ u((k+1)\Delta\tau)h((n-(k+1))\Delta\tau)\Delta\tau + \cdots + u(n\Delta\tau)h(0)\Delta\tau$$

i.e., $y(t)$ is composed of terms $u(k\Delta\tau)h((n-k)\Delta\tau)\Delta\tau$. Because $h((n-k)\Delta\tau)$ is less than $h((n-(k+1))\Delta\tau)$, we can conclude that $y(t)$ is more affected by $u((k+1)\Delta\tau)$ than $u(k\Delta\tau)$. In other words, the effect caused by the past $u(t)$ is faded away as time advances on.

Exercise:

Find the convolution of the following functions:

$$(A) \quad f(t) = \begin{cases} 1, & 0 \leq t \leq 2 \\ 0, & \text{elsewhere} \end{cases} \quad \text{and} \quad g(t) = \begin{cases} e^{-3t}, & t \geq 0 \\ 0, & \text{elsewhere} \end{cases}$$

$$(B) \quad f(t) = \begin{cases} 1, & 0 \leq t \leq 2 \\ 0, & \text{elsewhere} \end{cases} \quad \text{and} \quad g(t) = \begin{cases} \sin \frac{1}{2}t, & t \geq 0 \\ 0, & \text{elsewhere} \end{cases}$$