Also, the linear CT system must satisfy superposition condition (additive and homogeneity). That is :

```
If S[x_1(t)] = y_1(t) and S[x_2(t)] = y_2(t)
then S[ax_1(t) + bx_2(t)] = ay_1(t) + by_2(t)
```

Example: check the linearity of the following system y(t) =x(t)?

Solution

```
Let x (t) = x_1(t) + x_2(t) then y(t) = a_1 y_1(t) + a_2 y_2(t).
```

```
y_1(t) = x_1(t)
```

```
y_2(t) = x_2(t)
```

Then the output y(t) corresponding to the input x(t) is y(t) = $[a_1 x_1(t) + a_2 x_2(t)] = a_1 y_1(t) + a_2 y_2(t)$. Linear system.

Example/ is y(t) = 5t x(t) is linear or not? $y_1(t) = 5t x_1(t)$ $y_2(t) = 5t x_2(t)$ Scaling them to get $y(t) = a_1 y_1(t) + a_2 y_2(t) = a_1 5t x_1(t) + a_2 5t x_2(t)$ 1

$$y(t) = 5t [a_1 x_1(t) + a_2 x_2(t)]$$
 2
.... 2
.... 2

$$x_{1}(t) \rightarrow \times 5t \rightarrow y_{1}(t) = 5t \ x_{1}(t) + y_{1}(t) = a_{1} y_{1}(t) + a_{2} y_{2}(t)$$

$$x_{2}(t) \rightarrow \times 5t \rightarrow y_{2}(t) = 5t \ x_{2}(t) - y_{2}(t) = 2t \ x_{2}(t) - y_{2}(t) - y_{2}(t) = 2t \ x_{2}(t) - y_{2}(t) - y_{2}(t) = 2t \ x_{2}(t) - y_{2}(t) - y_{2}(t) - y_{2}(t) = 2t \ x_{2}(t) - y_{2}(t) - y_{2}$$

Note that the resistor and the capacitor mentioned before are linear systems.

Example

Let a digital amplifier,

If the inputs are:

Outputs will be:

y(n) = 10x(n)

 $x_1(n) = u(n)$ and $x_2(n) = \delta(n)$

 $y_1(n) = 10u(n)$ and $y_2(n) = 10\delta(n)$, respectively.





If we apply combined input to the system: $x(n) = 2x_1(n) + 4x_2(n) = 2u(n) + 4\delta(n)$

The output will be:

$$y(n) = 10x(n) = 10(2u(n) + 4\delta(n)) = 20u(n) + 40\delta(n)$$
Individual outputs:

$$2y_1(n) = 2\times 10x_1(n) = 20u(n)$$

$$4y_2(n) = 4\times 10x_2(n) = 40\delta(n)$$

$$2x_1(n) = 2u(n)$$

$$x = 10$$

$$2y_1(n) = 20u(n)$$

$$4x_2(n) = 4\delta(n)$$

$$4y_2(n) = 40\delta(n)$$

$$4x_2(n) = 4\delta(n)$$

$$4y_2(n) = 40\delta(n)$$



$$x_1(n) = u(n)$$
System
$$y_1(n) = u^2(n)$$

$$x_2(n) = \delta(n)$$
System
$$y_2(n) = \delta^2(n)$$

If the input is: $x(n) = 4x_1(n) + 2x_2(n)$

Then the output is: $y(n) = x^2(n) = (4x_1(n) + 2x_2(n))^2$ = $(4u(n) + 2\delta(n))^2 = 16u^2(n) + 16u(n)\delta(n) + 4\delta^2(n)$ = $16u(n) + 20\delta(n)$.

Individual outputs: $4y_1(n) = 4 \times x_1^2(n) = 4u(n)$ $2y_2(n) = 2 \times x_2^2(n) = 2\delta(n)$ + Non Linear System

Example: is $y(t) = x^2(t)$ linear or nonlinear system ? **Solution:**

 $y_1(t) = x_1^2(t)$ $y_2(t) = x_2^2(t)....$

The summation and scaling of these sub-o/p is

 $y(t) = a_1 y_1(t) + a_2 y_2(t) = a_1 x_1(t)^2 + a_2 x_2(t)^2 \dots 1$ then the output y(t) corresponding to the input $x(t) = [a_1 x_1(t) + a_2 x_2(t)]$ is $y(t) = [a_1 x_1(t) + a_2 x_2(t)]^2 = a_1^2 x_1(t)^2 + a_2^2 x_2(t)^2 + 2 a_1 x_1(t) a_2 x_2(t) \dots 2$ Which is not equal to $a_1 y_1(t) + a_2 y_2(t)$. Hence the system is said to be **nonlinear**.

Example: is y(t) = cos(x(t)) linear or nonlinear system ? **Solution**

```
y_1(t) = cos(x_1(t)), and y_2(t) = cos(x_2(t))
```

Then the output y(t) corresponding to the input x(t) is

 $y(t) = \cos(\alpha x_1(t) + \beta x_2(t))$

```
= \cos(\alpha x_1(t)) \cos(\beta x_2(t)) - \sin(\alpha x_1(t)) \sin(\beta x_2(t))
```

```
\neq \alpha \cos(x_1(t)) + \beta \cos(x_2(t))
```

The system is NOT linear.

A system is said to be time variant if its input and output characteristics vary with time. Otherwise, the system is considered as time invariant.

On another words, If the input $x(t - \tau)$ produces a response $y(t - \tau)$ where τ is any real constant, the system is called a **timeinvariant system**. In other words, a system is called *time-invariant if a time shift (delay or advance) in the input signal causes the same time shift in the output signal*. Thus, for a continuous-time system, the system is time-invariant if

$$f(x(t-\tau)) = y(t-\tau)$$

For any real value of τ . For a discrete-time system, the system is time-invariant (or shift-invariant) if

$$f(x(n-k)) = y(n-k)$$

For any integer **k**. A system which does not satisfy above conditions is called a time-varying system.



We have a system y(t) = H(x(t)). The response of the system to a shifted input x(t-k) should be the same as if the output y(t) has been shifted by k, i.e., y(t-k):

Time-invariant system is one whose parameters do not change with time:



Example : The system *y*[*n*]=K where K is constant is time invariant because it does not depend on time.

Example: check the time invariance of the system y(t) = x(-t)**Solution :** The output due to shifting the input is $y_1(t) = T[x(t-T)] = x(-t-T)$

The output due to shifting the equation (replace each t by (t-T) is $y_2(t-T) = x(-(t-T)) = x(-t+T)$ $\therefore y_1 \neq y_2$. Hence, the system is time variant.

Time-Invariant and Time-Varying Systems

Example : is y(t) = x(t - 2) + x(2 - t) time variant or not? **Solution**: Find output due to shifting in input which is $x(t - to) \rightarrow y_1(t) = x (t - to - 2) + x(2 - t - to)$ Find $y_2(t - to)$ which corresponding to output shifting (replacing each t by t-to) $y_2(t - to) = x(t - to - 2) + x(2 - t + to)$. $y_1 \neq y_2$: Hence system is time-variant.

Example: Examine y[n] = x[n] + n x[n+1] for time invariance. **Solution:** Notice that the equation has a **time-varying coefficient**, *n*. The output y[n] corresponding to x[n] is already given above. Delaying y[n] by *k* gives $y_1(n-k) = x(n-k) + (n-k) x(n-k+1) \rightarrow (A)$ Compare with $y_2(n) = T[x(n-k)] = x(n-k) + n x(n-k+1) \rightarrow (B)$ Since (A) rightarrow (B), the system is time varying.

Note that in applying the time-invariance test, we time-shift the input signal only, not the coefficients.

Exercise: Which of the following system is time-invariant? (a) y(t) = 3x(t) (b) y(t) = t x(t)

Linear Time-Invariant (LTI) Systems

In previous sections, a number of basic system properties are introduced. Two of these, linearity and time invariance, play a fundamental role in analysis and simulation of systems because of the many physical processes that can be modeled by linear time-invariant (**LTI**) systems. It will be shown that the input-output relationship for LTI systems is described in terms of a **convolution operation.** The importance of the convolution operation in LTI systems comes from the fact that knowledge of the response of an LTI system to the **unit impulse input** allows to find its output to any input signals. Specifying the input-output relationships for LTI systems by differential and difference equations will also be discussed.



Continuous LTI System

The *impulse response* completely characterizes the input-output behavior of LTI systems. Hence, properties of the system are related to the system's impulse response.

The Impulse Response:

The response to a **delta function** of an arbitrary linear system defined by an operator **T** is defined as

$$h(t) = \mathbf{T}\{\delta(t)\}$$

The function h(t) is called the *impulse response*. For LTI systems

$$h(t-\tau) = \mathbf{T}\{\delta(t-\tau)\}$$

Continuous LTI System

The Convolution Integral :

A convolution is an integral that expresses the amount of overlap of one function as it is shifted over another function. The response of a LTI system to an arbitrary input *x*(*t*) is given by

$$\mathbf{y(t)} = \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau$$

Above equation is referred to as the **convolution integral** or the **superposition integral** and indicates that a continuoustime LTI system is completely characterized by its impulse response h(t). The convolution of the signal x(t) and the impulse response h(t) is denoted as

$$y(t) = x(t) * h(t)$$

Properties of the Convolution Integral:

The most important properties of the convolution are a consequence of the superposition properties of linear timeinvariant systems.

1- The first basic property is that it is a commutative operation

$$x(t) * h(t) = h(t) * x(t)$$

2- A second useful property is that it is associative

$$x(t) * [h_1(t) * h_2(t)] = [x(t) * h_1(t)] * h_2(t)$$

3- The third is the distributive property

 $y(t) = x(t) * [h_1(t) + h_2(t)] = x(t) * h_1(t) + x(t) * h_2(t)$

Continuous LTI System

Convolution Integral Operation

Applying the commutative property of convolution to below equation

$$\mathbf{y(t)} = \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau$$

We obtain

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

The convolution integral operation involves the following four steps:

1. First, we time-reverse $h(\tau)$ and shift it to form $h(t - \tau)$. Note: the independent variable of $h(t - \tau)$ is τ , not t. The variable t is the shift parameter, i.e. the function $h(\tau)$ is shifted by an amount t.

2. Next, fix t and multiply $x(\tau)$ with $h(t - \tau)$ for all values of τ .

3. Then integrate $x(\tau) h(t - \tau)$ over all τ to get y(t) which is a single value that depends on t. Remember that τ is the

integration variable and that t is treated like a constant when doing the integral.

4. Finally, repeat for all values of t to produce the entire output y(t).



Continuous LTI System



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Discrete LTI System

The Impulse Response: The impulse response (or unit sample response) h[n] of a discrete-time LTI system (represented by T) is defined to be the response of the system when the input is $\delta[n]$, that is,

$$h[n] = T\{\delta[n]\}$$

The output response **y** [**n**] of LTI system to an arbitrary input **x** [**n**] can be expressed as

$$\mathbf{y[n]} = \sum_{k=-\infty}^{\infty} x[k] h[n-k]$$

The Convolution Sum:

The convolution sum of two sequences **x** [**n**] and **h** [**n**] is defined by

$$\mathbf{y[n]} = x(n) * h(n) = \sum_{k=-\infty}^{\infty} x[k] h[n-k]$$

Thus, again, we have the fundamental result that the output of any discrete-time LTI system is the convolution of the input **x**[**n**] with the impulse response **h**[**n**] of the system.

Discrete LTI System

Properties of the Convolution Sum

The following properties of the convolution sum are analogous to the convolution integral properties

1- Commutative: $x [n]^* h [n] = h [n]^* x [n]$

2- Associative: $\{x[n] * h_1[n]\} * h_2[n] = x[n] * \{h_1[n] * h_2[n]\}$

3- Distributive: $x[n] * \{h_1[n]\} + h_2[n]\} = x[n] * \{h_1[n] + x[n] * h_2[n]\}$

Convolution Sum Operation:

Similar to the continuous-time case, the convolution sum operation involves the following four steps: 1. The impulse response h[k] is time-reversed (that is, reflected about the origin) to obtain h[-k] and then shifted by n to form h[n-k] = h[-(k-n)] which is a function of k with parameter n.

2. Two sequences x [k] and h[n-k] are multiplied together for all values of k with n fixed at some value.

- 3. The product x [k] h [n-k] is summed over all k to produce a single output sample y[n].
- 4. Steps 1 to 3 are repeated as *n* varies over $-\infty$ to ∞ to produce the entire output y[n].