

Chapter 2

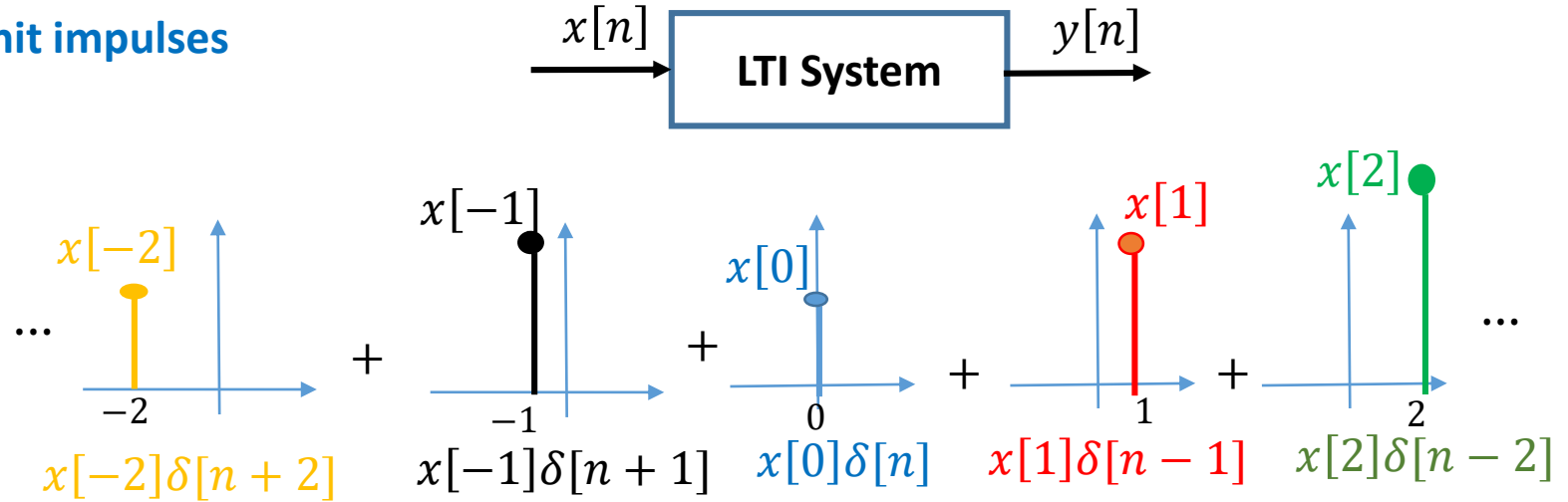
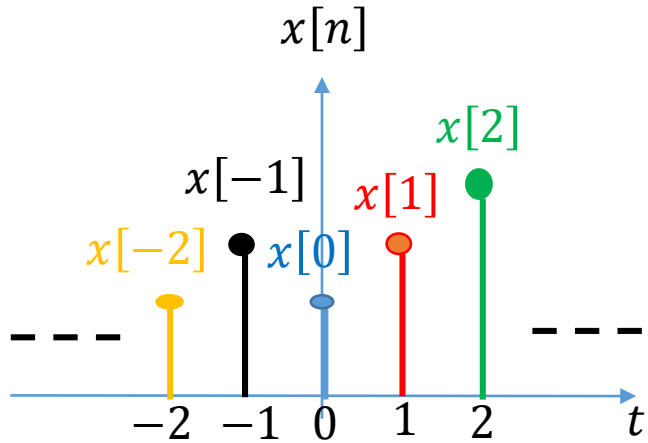
Linear Time-Invariant Systems

(LTI Systems)

Linear Time-Invariant Systems

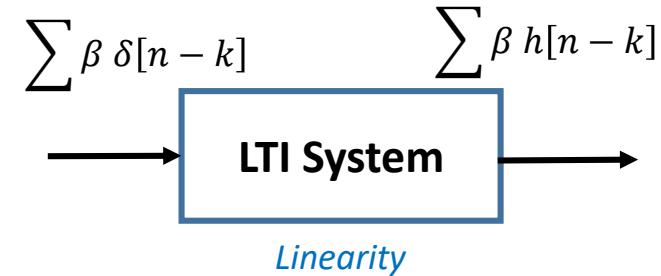
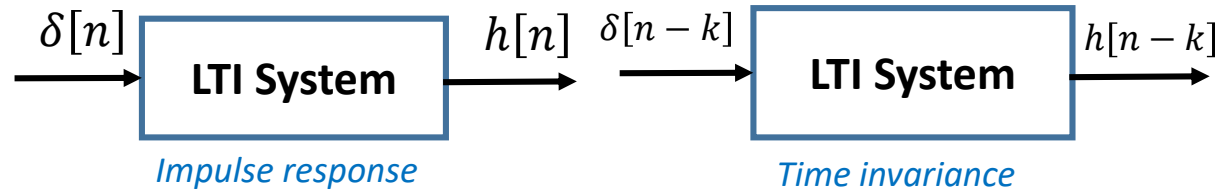
- A system is said to be Linear Time-Invariant (*LTI*) if it possesses the basic system properties of *linearity* and *time-invariance*.
- The input-output relationship for LTI systems is described in terms of a *convolution* operation.

DT Signal Decomposition in terms of shifted unit impulses



$$x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n-k]$$

Sum of scaled impulses



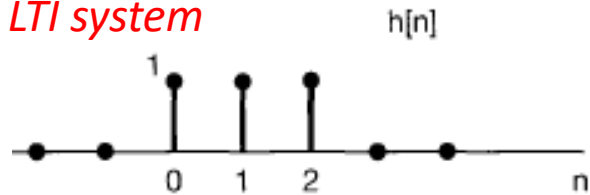
$$x[n] \rightarrow \text{LTI System} \rightarrow y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k] \quad \Rightarrow \quad y[n] = x[n] * h[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$

Convolution sum

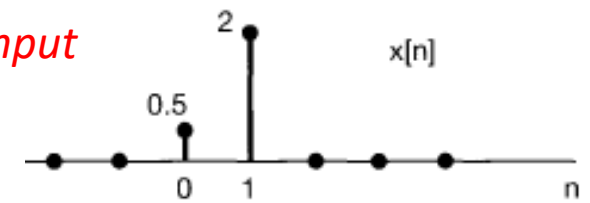
Convolution 1

Example:

Impulse response of an LTI system

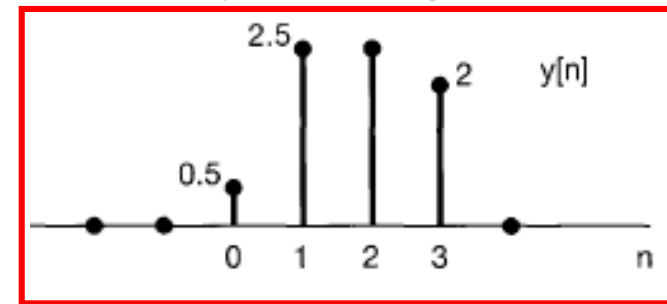
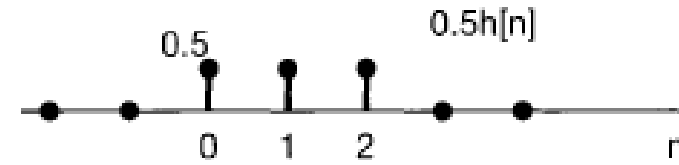


input



Find output

Convolution



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

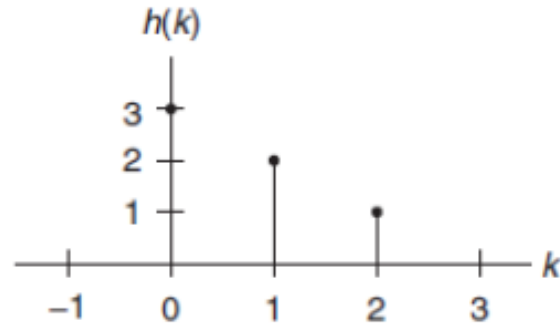
There are only two non-zero values for the input

$$y[n] = x[0]h[n-0] + x[1]h[n-1] = 0.5h[n] + 2h[n-1]$$

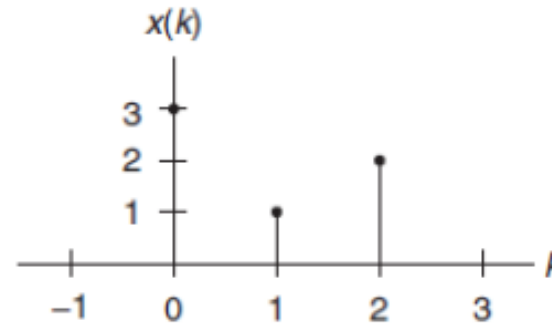
Convolution 2

Example:

$$x[n] = \begin{cases} 3 & n = 0 \\ 2 & n = 1 \\ 1 & n = 2 \\ 0 & \text{elsewhere} \end{cases}$$



Length = 3



Length = 3

$$h[n] = \begin{cases} 3 & n = 0 \\ 1 & n = 1 \\ 2 & n = 2 \\ 0 & \text{elsewhere} \end{cases}$$

Solution:

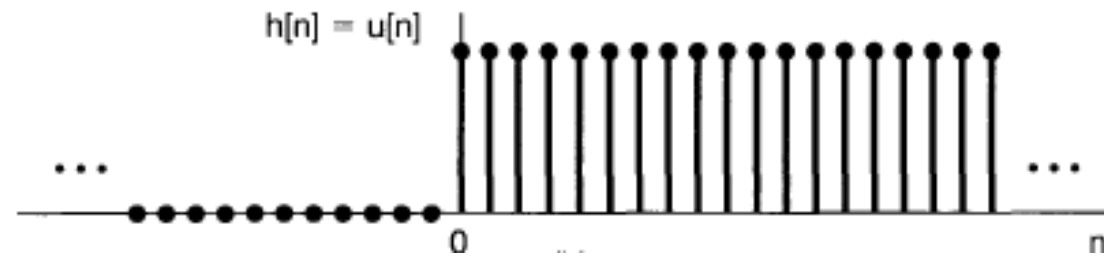
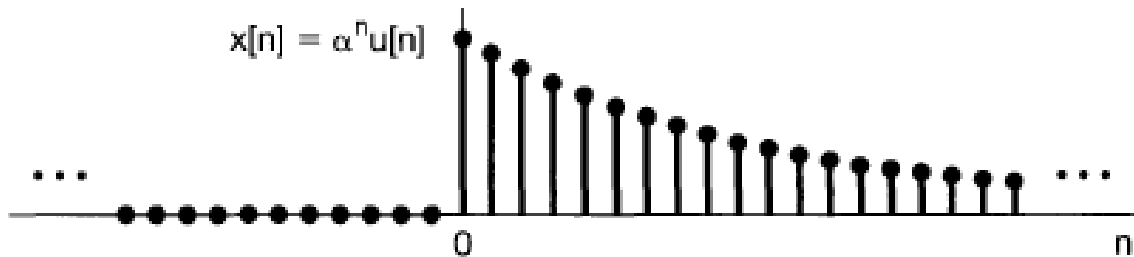
Convolution sum using the table method.

$k:$	-2	-1	0	1	2	3	4	5	
$x(k):$			3	1	2				
$h(-k):$	1	2	3						$y(0) = 3 \times 3 = 9$
$h(1-k)$		1	2	3					$y(1) = 3 \times 2 + 1 \times 3 = 9$
$h(2-k)$			1	2	3				$y(2) = 3 \times 1 + 1 \times 2 + 2 \times 3 = 11$
$h(3-k)$				1	2	3			$y(3) = 1 \times 1 + 2 \times 2 = 5$
$h(4-k)$					1	2	3		$y(4) = 2 \times 1 = 2$
$h(5-k)$						1	2	3	$y(5) = 0$ (no overlap)

$$\text{Convolution Length} = N_1 + N_2 - 1 = 3 + 3 - 1 = 5$$

Convolution 3

Example: Find the output of an LTI system having a unit impulse response $h[n] = u[n]$, for the input $x[n] = \alpha^n u[n]$



$$x[k]h[n-k] = \begin{cases} \alpha^k & 0 \leq k \leq n \\ 0 & \text{otherwise} \end{cases}$$

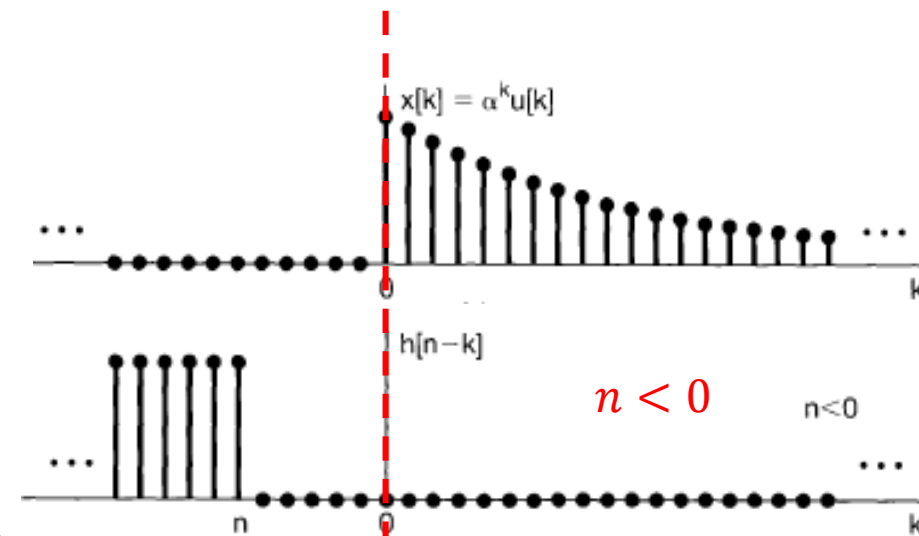
$$\begin{aligned} k > n &\rightarrow h[n-k] = 0 \\ k < 0 &\rightarrow x[k] = 0 \end{aligned}$$

Thus, for $n \geq 0$,

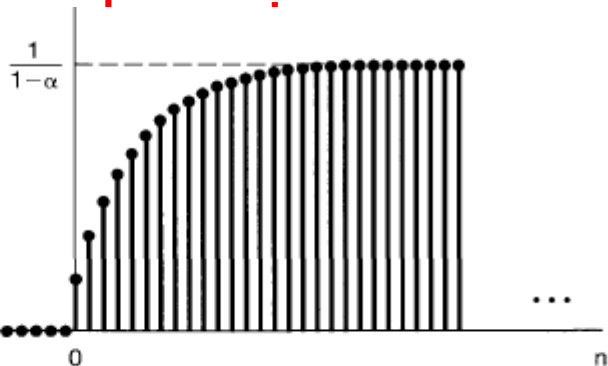
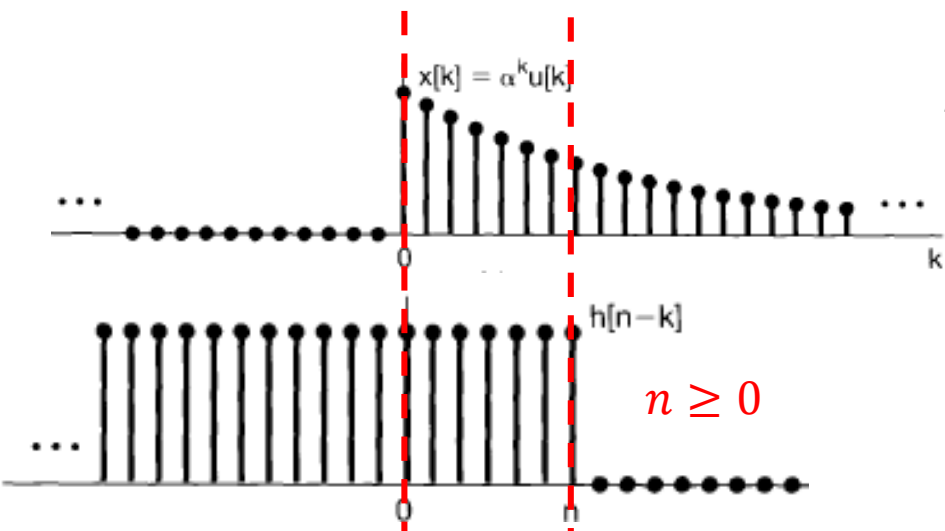
$$y[n] = \sum_{k=0}^n \alpha^k = \frac{1-\alpha^{n+1}}{1-\alpha}$$

Thus, for all n ,

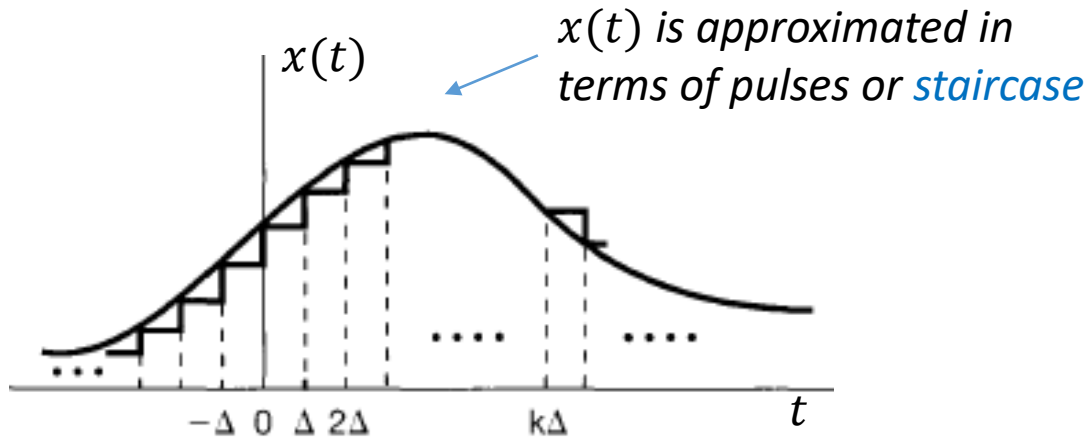
$$y[n] = \left(\frac{1-\alpha^{n+1}}{1-\alpha} \right) u[n]$$



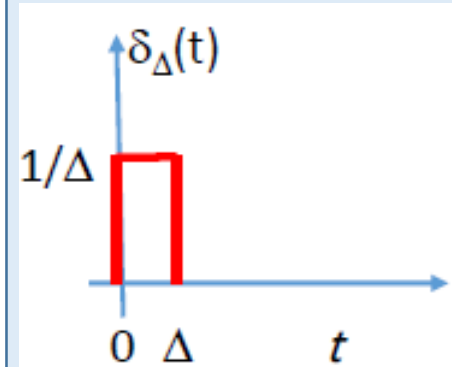
$$\begin{aligned} s &= \sum_{k=0}^n \alpha^k = 1 + \alpha + \alpha^2 + \dots + \alpha^n \\ \alpha s &= \alpha + \alpha^2 + \dots + \alpha^{n+1} \end{aligned} \quad \rightarrow \quad s - \alpha s = 1 - \alpha^{n+1}$$



The Representation of Continuous-Time Signals in Terms of Impulses



Defining:



$$\delta_{\Delta}(t) = \begin{cases} 1/\Delta & 0 \leq t \leq \Delta \\ 0 & \text{otherwise} \end{cases}$$

$$\Delta \delta_{\Delta}(t) = \begin{cases} 1 & 0 \leq t \leq \Delta \\ 0 & \text{otherwise} \end{cases}$$

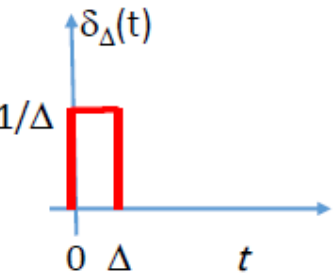
$\Delta \delta_{\Delta}(t)$ has a unit amplitude

The complete pulse/staircase approximation of $x(t)$ is the sum

$$\hat{x}(t) = \dots + \hat{x}(-\Delta) + \hat{x}(0) + \hat{x}(\Delta) + \dots$$

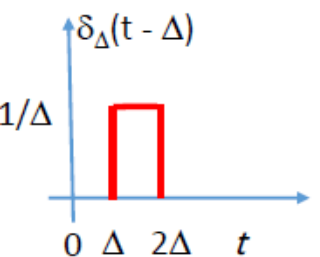
$$\hat{x}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta) \delta_{\Delta}(t - k\Delta) \Delta$$

$\hat{x}(t)$: the approximation of $x(t)$



At $t = 0$

$$\hat{x}(0) = x(0) \Delta \delta_{\Delta}(t) = \begin{cases} x(0) & 0 \leq t \leq \Delta \\ 0 & \text{otherwise} \end{cases}$$



At $t = \Delta$

$$\hat{x}(\Delta) = x(\Delta) \Delta \delta_{\Delta}(t - \Delta) = \begin{cases} x(\Delta) & \Delta \leq t \leq 2\Delta \\ 0 & \text{otherwise} \end{cases}$$

In general At $t = k\Delta$

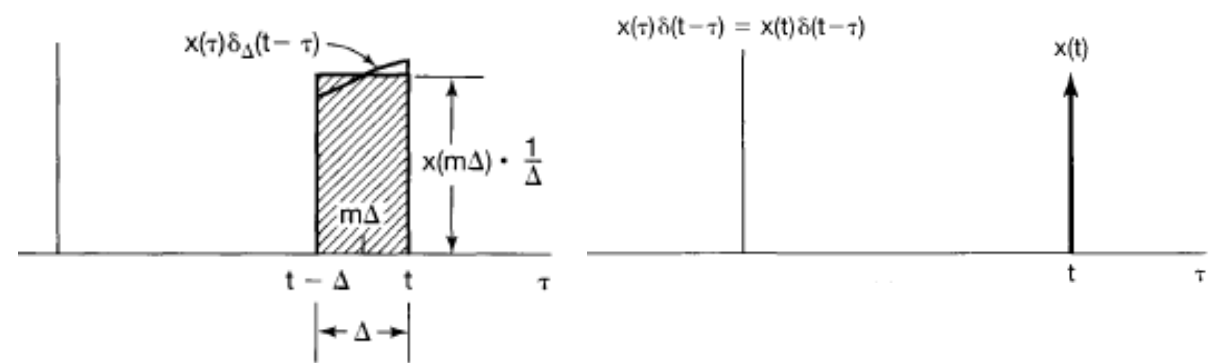
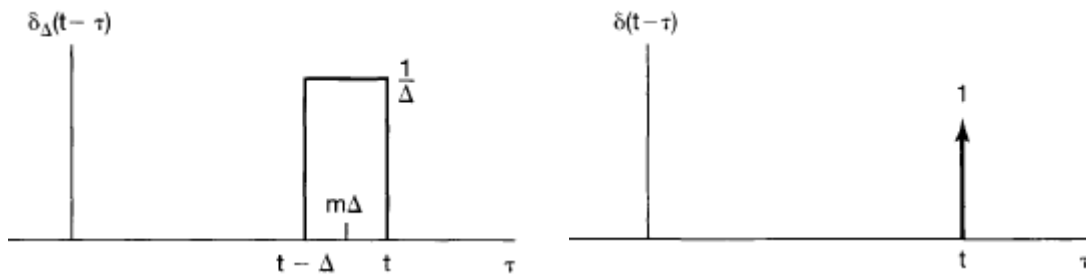
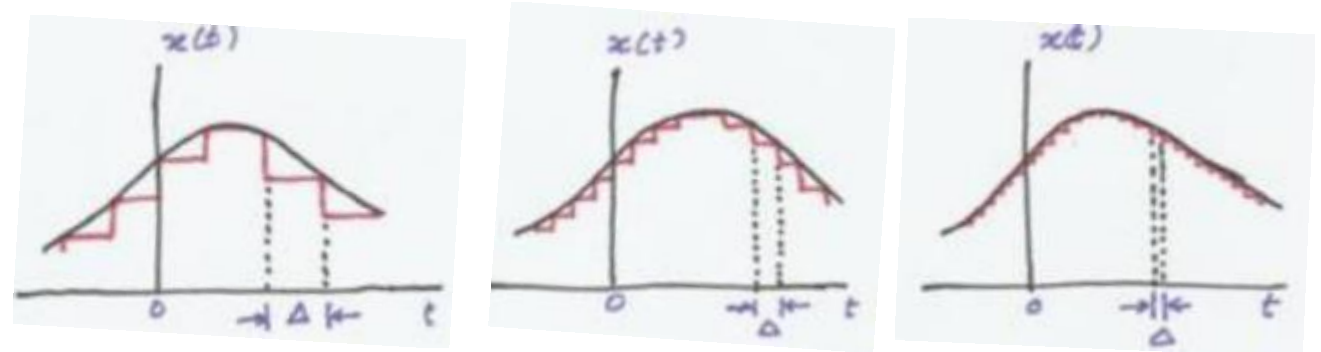
$$\hat{x}(k\Delta) = x(k\Delta) \Delta \delta_{\Delta}(t - k\Delta) = \begin{cases} x(k\Delta) & k\Delta \leq t \leq (k+1)\Delta \\ 0 & \text{otherwise} \end{cases}$$

The Representation of CT Signals in Terms of Impulses₂

If we let Δ approach 0 $\hat{x}(t)$ becomes closer and closer and equals $x(t)$ in the limit of 0

$$x(t) = \lim_{\Delta \rightarrow 0} \hat{x}(t) = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{\infty} x(k\Delta) \delta_{\Delta}(t - k\Delta) \Delta$$

In the limiting case the sum approaches integral (area):



$$\Delta \rightarrow 0; \delta_{\Delta}(t) \rightarrow \delta(t)$$

$$\sum_{k=-\infty}^{\infty} \dots \Delta \rightarrow \int_{\tau=-\infty}^{\infty} \dots d\tau$$

Consequently:

$$x(t) = \int_{\tau=-\infty}^{\infty} x(\tau) \delta(t - \tau) d\tau$$

The Convolution Integral

- The impulse response $h(t)$ of a continuous-time LTI system S

$$h(t) = S\{\delta(t)\}$$



Sum (Integral) of weighted shifted impulses

For the input $x(t)$:

$$y(t) = S\{x(t)\} = S\left\{\int_{\tau=-\infty}^{\infty} \underset{\substack{\text{weight} \\ \uparrow}}{x(\tau)} \underset{\substack{\text{impulse} \\ \uparrow}}{\delta(t-\tau)} d\tau\right\} \stackrel{\text{linearity}}{=} \int_{\tau=-\infty}^{\infty} x(\tau) S\{\delta(t-\tau)\} d\tau$$

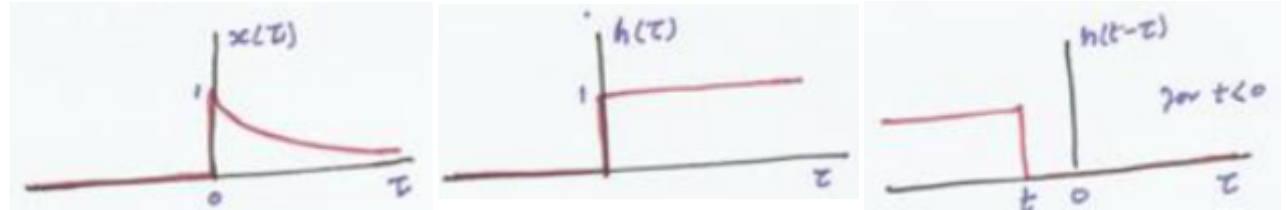
- Since the system is time-invariant:

$$S\{\delta(t-\tau)\} = h(t-\tau) \quad \text{Time-invariance} \quad \longrightarrow$$

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

Example 1

Let, the input $x(t)$ to an LTI system with unit impulse response $h(t)$ be given as $x(t) = e^{-at}u(t)$ for $a > 0$ and $h(t) = u(t)$. Find the output $y(t)$ of the system.

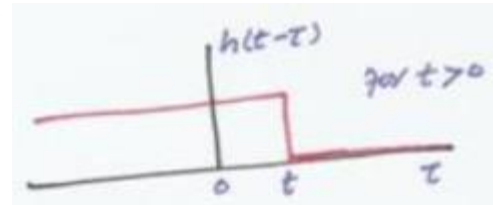


$$y(t) = x(t) * h(t) = \int_{\tau=-\infty}^{\infty} x(\tau) h(t-\tau) d\tau$$

$x(t) \neq 0$ for $t \geq 0$
 $h(t) = u(t)$.

$$= \int_0^{\infty} e^{-a\tau} h(t-\tau) d\tau \quad \text{for } t > 0$$

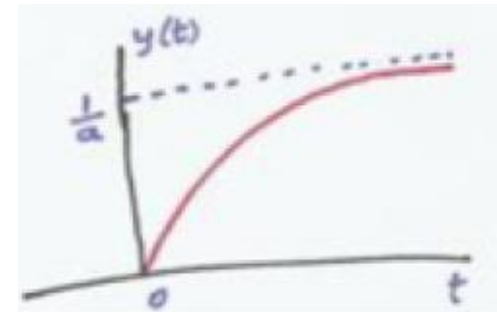
$$h(t-\tau) = \begin{cases} 1 & 0 < \tau < t \\ 0 & \tau > t \end{cases}$$



$$= \int_0^t e^{-a\tau} d\tau = \frac{1}{-a} e^{-a\tau} \Big|_0^t = \frac{1}{a} (1 - e^{-at})$$

Thus, for all t , we can write

$$y(t) = \frac{1}{a} (1 - e^{-at})$$



The Convolution Integral₂

Example 2

Find $y(t) = x(t) * h(t)$, where

$$\begin{cases} x(t) = e^{2t}u(-t) \\ h(t) = u(t-3) \end{cases}$$

The system response is $y(t) = \int_{-\infty}^{\infty} x(\tau) h(t-\tau) d\tau$

these two signals have regions of nonzero overlap

For $t - 3 \leq 0$: nonzero overlap for $-\infty \leq \tau \leq t - 3$

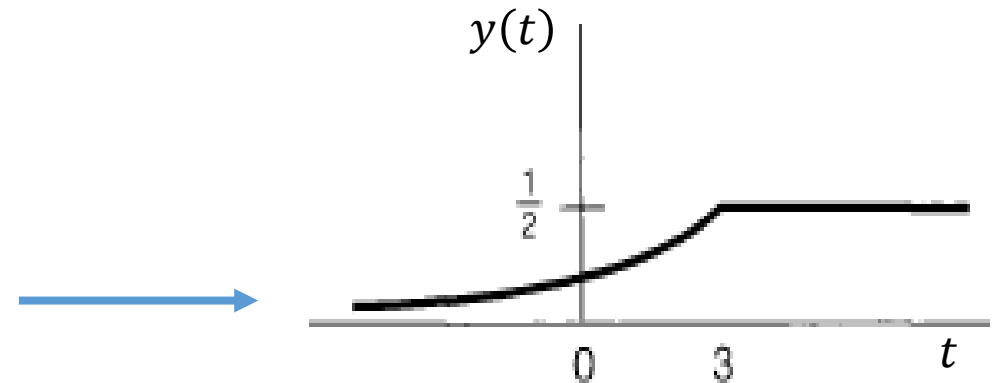
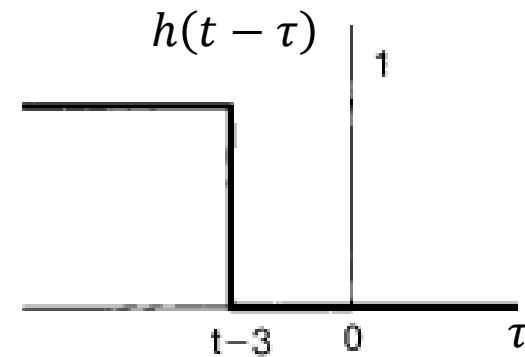
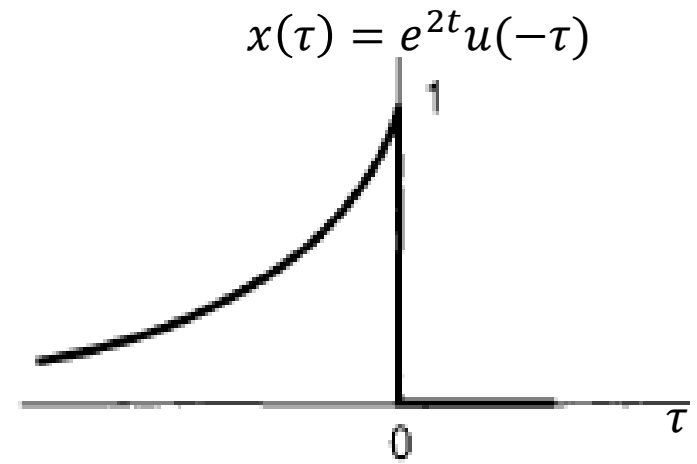
$$y(t) = \int_{-\infty}^{t-3} e^{2\tau} d\tau = \frac{1}{2} e^{2(t-3)} \quad \text{For } t \leq 3$$

For $t - 3 \geq 0$: nonzero overlap for $-\infty \leq \tau \leq 0$

$$y(t) = \int_{-\infty}^0 e^{2\tau} d\tau = \frac{1}{2} \quad \text{For } t \geq 3$$

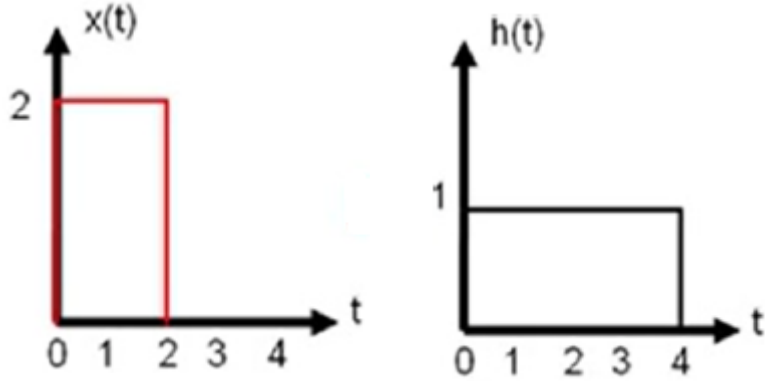
→

$$y(t) = \begin{cases} \frac{1}{2} e^{2(t-3)} & \text{For } t \leq 3 \\ 1/2 & \text{For } t \geq 3 \end{cases}$$

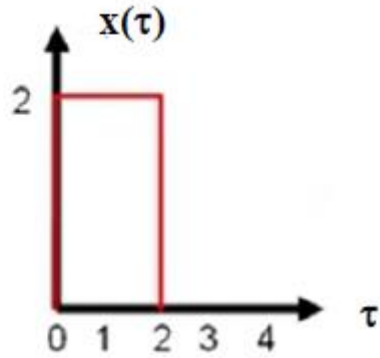
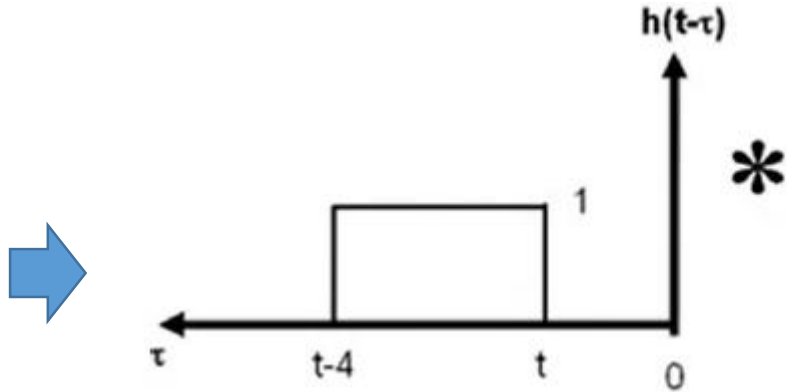
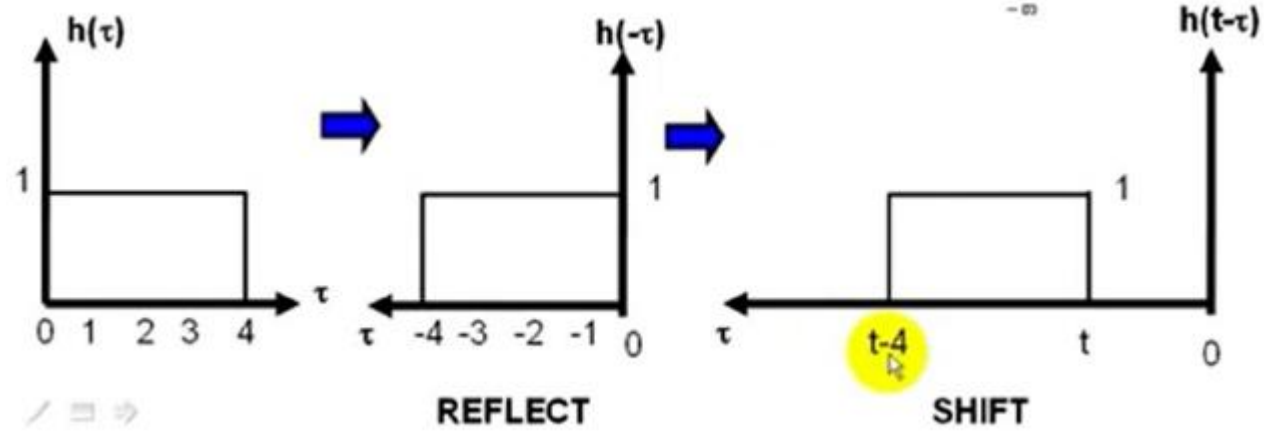


$$y(t) = \int_{\tau=-\infty}^{\infty} x(\tau) h(t - \tau) d\tau$$

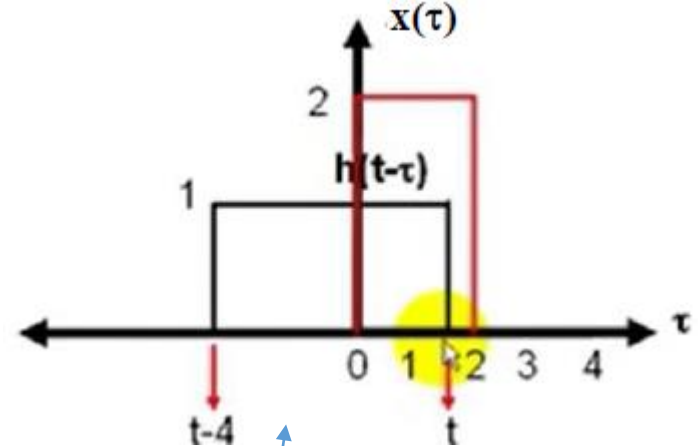
The Convolution Integral₃



Convolution



Shift $h(t - \tau)$



Aggregate the overlapping

$t < 0$ No overlapping

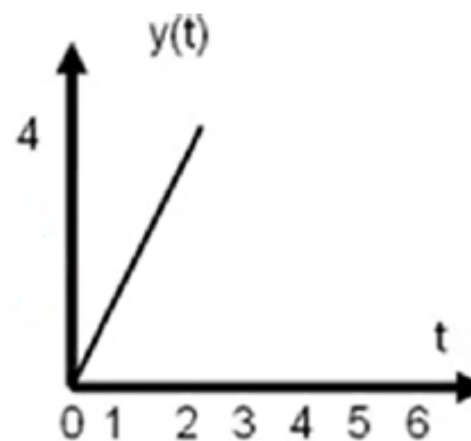
$y(t) = 0$

$y(t) = \int_0^t 2 \cdot 1 d\tau \quad 0 < \tau < t$

$y(t) = 2 \left(\tau \Big|_0^t \right)$

$y(t) = 2t$

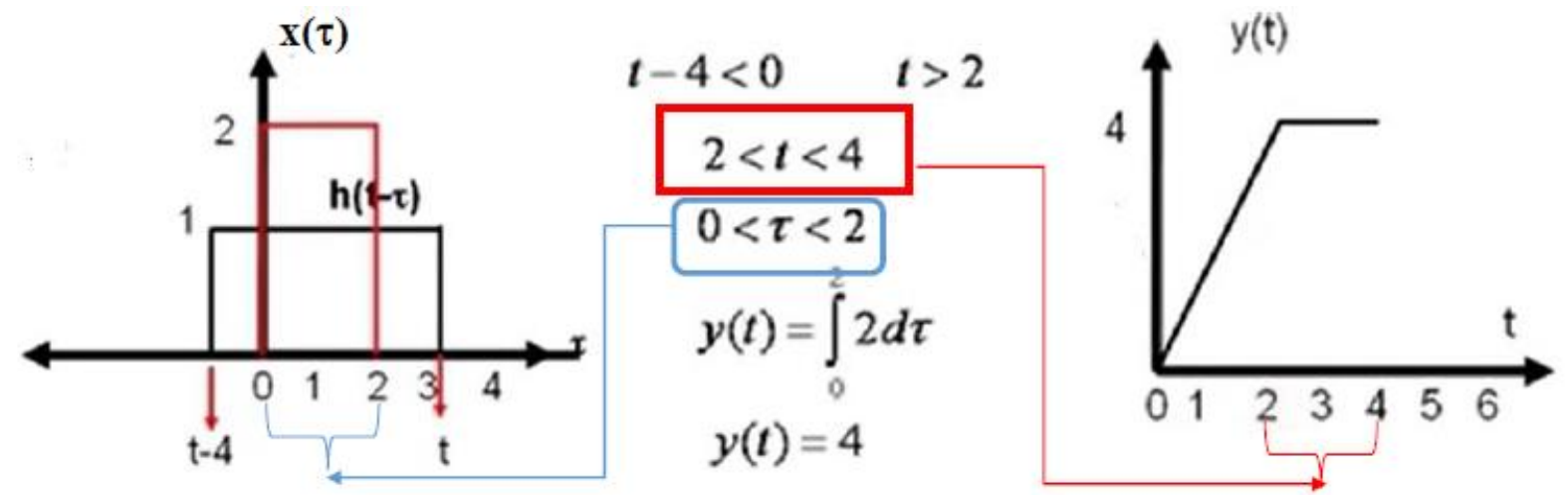
$0 < t < 2$



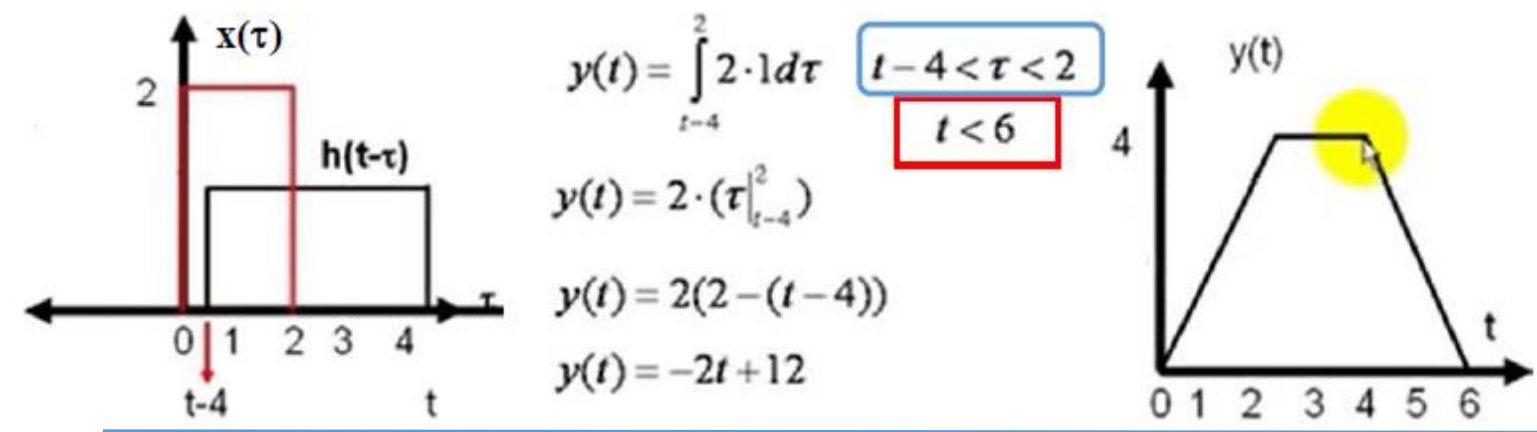
$0 < t < 2$

The Convolution Integral₄

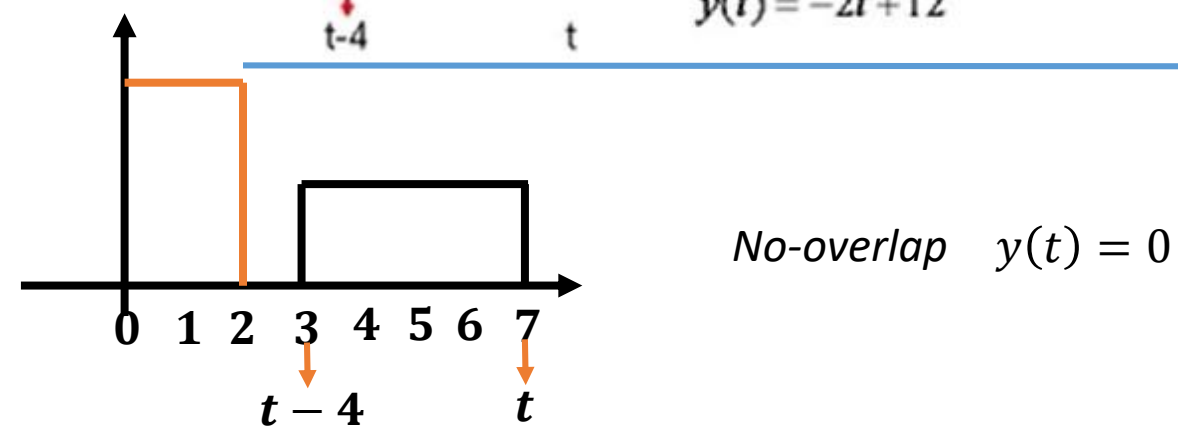
$2 < t < 4$ →



$4 < t < 6$ →



$t > 6$ →



Properties of LTI Systems

- The characteristics of an LTI system are completely determined by its *impulse response*. This property holds in general *only* for *LTI systems* only.
- The unit impulse response of a *nonlinear system* does not completely characterize the behavior of the system.

Consider a discrete-time system with unit impulse response:

$$h[n] = \begin{cases} 1, & n = 0, 1 \\ 0, & \text{otherwise} \end{cases}$$

If the system is *LTI*, we get the system output (by convolution): $y[n] = x[n] + x[n-1]$

There is only one such LTI system for the given $h[n]$.

However, there are many nonlinear systems with the same response, $h[n]$.

Two different Non-Linear systems with same impulse response

$$y[n] = (x[n] + x[n-1])^2$$



$$h[n] = (\delta[n] + \delta[n-1])^2$$



$$h[n] = \begin{cases} 1, & n = 0, 1 \\ 0, & \text{otherwise} \end{cases}$$

$$y[n] = \max(x[n], x[n-1])$$



$$h[n] = \text{Max}(\delta[n], \delta[n-1])$$



$$h[n] = \begin{cases} 1, & n = 0, 1 \\ 0, & \text{otherwise} \end{cases}$$



1 Commutative Property

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

$$x[n] * h[n] = h[n] * x[n]$$

Proof: (discrete domain)

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

$$\text{Put } r = n - k \Rightarrow k = n - r$$

$$x[n] * h[n] = \sum_{r=-\infty}^{\infty} x[n-r]h[r] = \sum_{r=-\infty}^{\infty} h[r]x[n-r] = h[n] * x[n]$$

Similarly, we can prove it for continuous domain.

2 Distributive Property

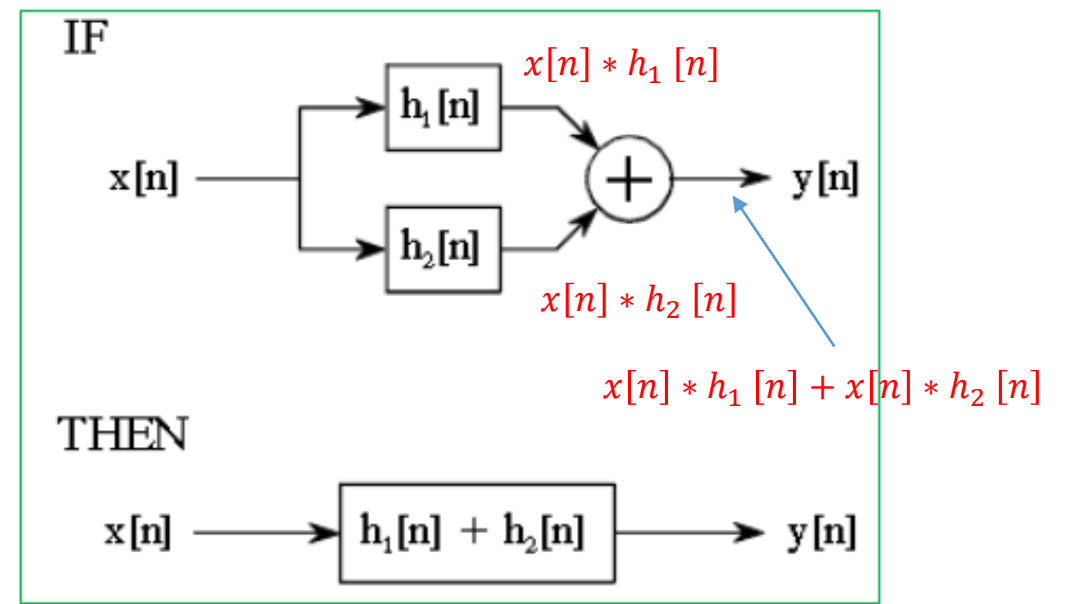
Convolution is distributive over addition,

in discrete time

$$x[n] * (h_1[n] + h_2[n]) = x[n] * h_1[n] + x[n] * h_2[n]$$

in continuous time

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



Example:

$$y[n] = x[n] * h[n] \quad x[n] = \left(\frac{1}{2}\right)^n u[n] + 2^n u[-n] \quad \text{and} \quad h[n] = u[n]$$

$x[n]$ is nonzero for entire n , so direct convolution is difficult. Therefore, we will use commutative property.

$$y[n] = x[n] * h[n] = (x_1[n] + x_2[n]) * h[n] = (x_1[n] * h[n] + x_2[n] * h[n]) = y_1[n] + y_2[n]$$

$$y_1[n] = x_1[n] * h[n] = \sum_{k=-\infty}^{\infty} x_1[k] h[n-k] = \sum_{k=-\infty}^{\infty} \left(\frac{1}{2}\right)^k u[k] u[n-k] = \left(\frac{1-(1/2)^{n+1}}{1-(1/2)}\right) u[n] = 2(1-(1/2)^{n+1})u[n]$$

\uparrow 1 for $k \geq 0$ \uparrow 1 for $k \leq n$

$$y_2[n] = x_2[n] * h[n] = \sum_{k=-\infty}^{\infty} 2^k u[-k] u[n-k] = 2^{n+1} \sum_{l=-n}^{\infty} \left(\frac{1}{2}\right)^l = 2^n \sum_{m=0}^{\infty} \left(\frac{1}{2}\right)^m = 2^{n+1}$$

\uparrow 1 for $k \leq 0$ \uparrow 1 for $k \leq n$

$$y[n] = y_1[n] + y_2[n] = 2(1-(1/2)^{n+1})u[n] + 2^{n+1}$$

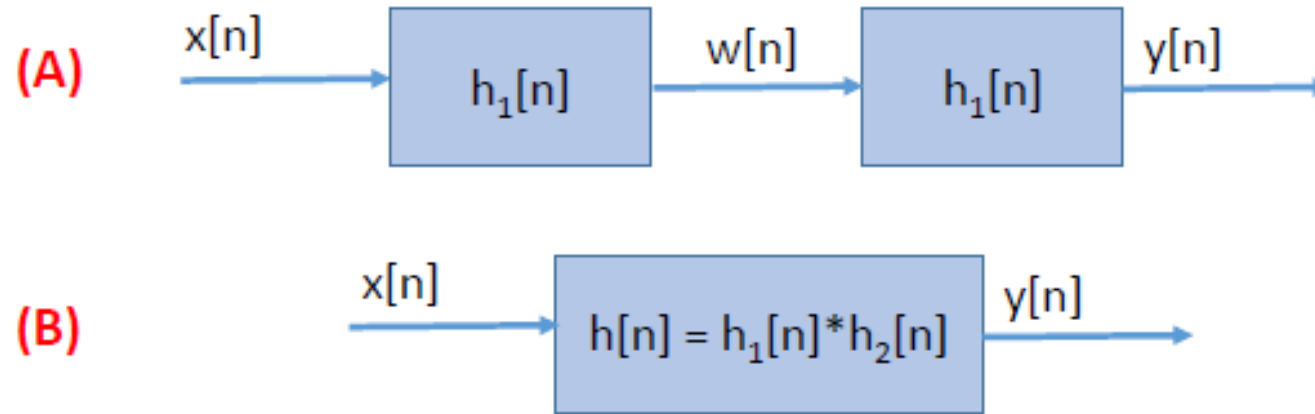
3 Associative Property

in continuous time

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

In discrete time

$$x[n] * [h_1[n] * h_2[n]] = [x[n] * h_1[n]] * h_2[n]$$



Proof:

From (A), $y[n] = w[n] * h_2[n] = (x[n] * h_1[n]) * h_2[n]$

From (B), $y[n] = x[n] * h[n] = x[n] * (h_1[n] * h_2[n])$

4 LTI Systems With and Without Memory

A system is *memory-less* if its output at any time depends only on the value of the input at that same time.

System output:
$$y[n] = \sum_{k=-\infty}^{+\infty} x[k]h[n-k]$$
 ← $y[n]$ depends on only $x[n]$ only if $k = n$, so for $h[n] = 0$ for $n \neq 0$

A discrete-time LTI system can be memory-less if only: $h[n] = 0, \text{ for } n \neq 0$ ← impulse response $x[n] = \delta[n]$
← $y[n] = x[n] h[0] = K\delta[n]$

Thus, the impulse response have the form: $h[n] = K\delta[n], \text{ with } K = h[0] \text{ is a constant}$

the convolution sum reduces to $y[n] = Kx[n]$

If $K = 1$, then the system is called *identity system*.

Similarly for continuous LTI systems.

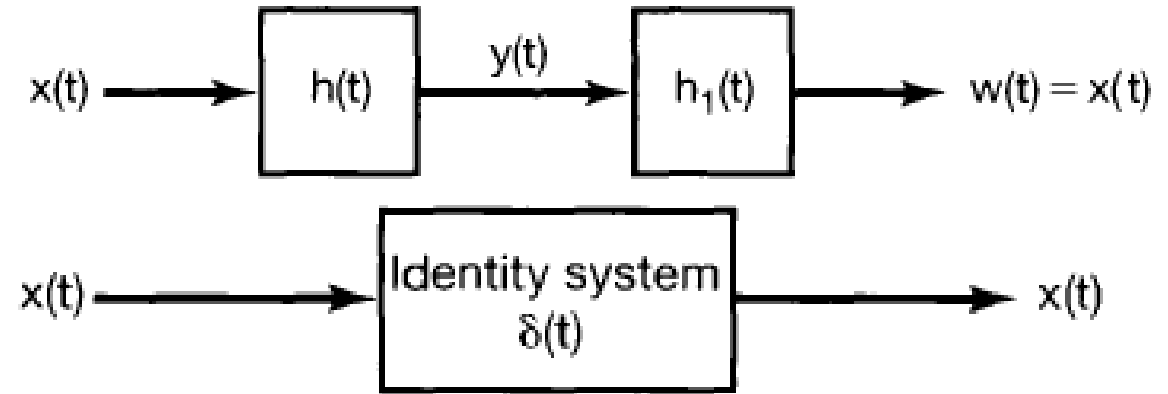
a continuous-time LTI system is memory-less if $h(t) = 0 \text{ for } t \neq 0,$
 $h(t) = K\delta(t). \quad y(t) = Kx(t)$

5 Invertibility of LTI Systems

A system is invertible only if *an inverse system exists*

The system with impulse response $h_1(t)$ is inverse of the system with impulse response $h(t)$ if:

$$h(t) * h_1(t) = \delta(t)$$



Example:

Consider the LTI system consisting of a pure time shift $y(t) = x(t - t_0)$

$t_0 > 0$ *delay*

$t_0 < 0$ *advance*

The *impulse response* for the system (for $x(t) = \delta(t)$): $h(t) = \delta(t - t_0)$ ← impulse response $x(t) = \delta(t)$

the system's output (*the convolution*): $y(t) = x(t) * h(t) = x(t) * \delta(t - t_0) = x(t - t_0)$

To recover the input from the output (*invert the system*), all that is required is to shift the output back.

The inverse system impulse response: $h_1(t) = \delta(t + t_0)$

then $h(t) * h_1(t) = \delta(t - t_0) * \delta(t + t_0) = \delta(t)$ ← *identity system* ($y(t) = x(t) * \delta(t) = x(t)$)

Invertibility of LTI Systems: Example 2

Consider an LTI system with impulse response: $h[n] = u[n]$

Response of this system (convolution sum): $y[n] = \sum_{k=-\infty}^{\infty} x[k] u[n-k]$ $\xrightarrow{u[n-k]=0 \text{ for } k > n}$ $y[n] = \sum_{k=-\infty}^n x[k]$

summer or accumulator

$\xrightarrow{\text{first difference equation}}$ $y[n] = x[n] + \sum_{k=-\infty}^{n-1} x[k]$ $\xrightarrow{\text{first difference equation}}$ $y[n] = x[n] + y[n-1]$ $\xrightarrow{\text{first difference equation}}$ $x[n] = y[n] - y[n-1]$ $\xrightarrow{\text{Inverse system}}$ $y[n] = x[n] - x[n-1]$

Impulse response ($x[n] = \delta[n]$): $h_1[n] = \delta[n] - \delta[n-1]$

Verification: $h[n] * h_1[n] = \delta[n]$

$$\begin{aligned} h[n] * h_1[n] &= u[n] * \{ \delta[n] - \delta[n-1] \} \\ &= \{ u[n] * \delta[n] \} - \{ u[n] * \delta[n-1] \} \\ &= u[n] - u[n-1] \\ &= \delta[n] \end{aligned} \quad \xrightarrow{\text{Verification}} \quad h[n] * h_1[n] = \delta[n]$$

the impulse responses are inverses of each other

6 Causality of LTI Systems

- The output of a *causal system* depends only on the present and past values of the input to the system.

$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k] \quad y[n] \text{ must not depend on } x[k] \text{ for } k > n \text{ to be } \textit{causal}$$

Therefore, for a discrete-time LTI system to be causal: $x[k]h[n-k] = 0$ for $k > n$ \Rightarrow $h[n-k] = 0$ for $k > n$

for $k > n \rightarrow n - k < 0 \quad \Rightarrow \quad \boxed{h[n] = 0 \quad \text{for } n < 0}$

Causality for LTI system is equivalent to the condition of initial rest (output must be 0 before applying the input)

for $k > n$ $h[n-k] = 0$

for $k < 0$ $h[k] = 0$

$$y[n] = \sum_{k=-\infty}^n x[k]h[n-k] = \sum_{k=0}^{\infty} h[k]x[n-k]$$

Both the accumulator ($h[n] = u[n]$) and its inverse ($h[n] = \delta[n] - \delta[n-1]$) are causal.

Inverse system of the accumulator

$$\begin{aligned} h[n] * h_1[n] &= u[n] * \{\delta[n] - \delta[n-1]\} \\ &= u[n] * \delta[n] - u[n] * \delta[n-1] \\ &= u[n] - u[n-1] \\ &= \delta[n]. \end{aligned}$$

- Similarly, for a continuous-time LTI system to be causal:

$$y(t) = \int_0^{\infty} h(\tau)x(t-\tau)d\tau$$

7 Stability of LTI Systems

- A *system is stable* if every bounded input produces a bounded output (*BIBO*).

Consider, an input $x[n]$ to an LTI system that is bounded in magnitude:

$$|x[n]| < B, \quad \text{for all } n$$

Suppose that we apply this to the LTI system with impulse response $h[n]$.

$$|y[n]| = \left| \sum_{k=-\infty}^{\infty} h[k]x[n-k] \right| \leq \sum_{k=-\infty}^{\infty} |h[k]| |x[n-k]| \leq B \sum_{k=-\infty}^{\infty} |h[k]| \quad \text{for all } n$$

We take $x[n] = B$

Therefore, if $\sum_{k=-\infty}^{\infty} |h[k]| < \infty$, then $|y[n]| < \infty$

The system is stable if the *impulse response $h[n]$ is absolutely summable.*

- Similar case in continuous-time LTI system.

the system is stable if the impulse response is absolutely integrable.

Example:

An LTI system with pure time shift is stable.

$$\sum_{n=-\infty}^{\infty} |h[n]| = \sum_{n=-\infty}^{\infty} |\delta[n - n_0]| = 1$$

An accumulator (DT domain) system is unstable.

$$\sum_{n=-\infty}^{\infty} |h[n]| = \sum_{n=-\infty}^{\infty} |u[n]| = \sum_{n=0}^{\infty} |u[n]| = \infty$$

Similarly, an integrator (CT domain) system is unstable.

8 Unit Step Response of An LTI System

- the *unit step response*, $s[n]$ or $s(t)$, the output corresponding to the input $x[n] = u[n]$ or $x(t) = u(t)$.
- it is worthwhile relating the *unit step response* to the impulse response

commutative property

$$s[n] = u[n] * h[n] = h[n] * u[n]$$

Response to the input $h[n]$ of a LTI system with unit impulse response $u[n]$.

$u[n]$ is the *unit impulse response* of the *accumulator*.

impulse response of an accumulator

$$h[n] = \sum_{k=-\infty}^n \delta[k] = \begin{cases} 1 & n \geq 0 \\ 0 & n < 0 \end{cases} = u[n]$$



Discrete-time domain



$$\Rightarrow s[n] = \sum_{k=-\infty}^n h[k]$$



Running Sum

$$\Rightarrow h[n] = s[n] - s[n-1]$$



First Difference

Continuous-time domain



$$s(t) = \int_{-\infty}^t h(\tau) d\tau$$



Running Integral

$$h(t) = \frac{ds(t)}{ds} = s'(t)$$



First Derivative

LTI Systems Described by Differential Equation

(Linear Constant-Coefficient Differential Equation)

A general N^{th} -order linear constant-coefficient differential equation that relates the input $x(t)$ to the output $y(t)$ is given by

$$\sum_{k=0}^N a_k \frac{d^k y(t)}{dt^k} = \sum_{k=0}^M b_k \frac{d^k x(t)}{dt^k}$$

Example 1:

consider a first-order differential equation $\frac{dy(t)}{dt} + 2y(t) = x(t)$

where the input to the system is: $x(t) = Ke^{3t}u(t)$

Solve for $y(t)$.

The complete solution is $y(t) = y_p(t) + y_h(t)$

$y_p(t)$ the particular solution

$y_h(t)$ The homogeneous solution,

- Finding the particular solution $y_p(t)$ (signal of the same form as the input)

forced response $y_p(t) = Ye^{3t}$ ← Determine Y

From differential equation:

$$3Ye^{3t} + 2Ye^{3t} = Ke^{3t} \Rightarrow 3Y + 2Y = K \Rightarrow Y = \frac{K}{5}$$

→ $y_p(t) = \frac{K}{5}e^{3t}, K \text{ real and } t > 0$

- Finding the homogeneous solution (hypothesize a solution)

$y_h(t) = Ae^{st}$ ← Determine s and A

From differential equation:

$$sAe^{st} + 2Ae^{st} = 0 \Rightarrow A(s + 2)e^{st} = 0 \Rightarrow s = -2$$

$$y_h(t) = Ae^{-2t}$$

Complete solution:



$$y(t) = \frac{K}{5}e^{3t} + Ae^{-2t}$$

Example_contd

- To find A suppose that the auxiliary condition is $y(0) = 0$, i.e., at $t = 0, y(t) = 0$

Using this condition into the complete solution, we get:

$$y(t) = \frac{K}{5} e^{3t} + A e^{-2t} \quad \text{with } y(0) = 0$$

$$\Rightarrow 0 = \frac{K}{5} + A \Rightarrow A = -\frac{K}{5}$$



$$y(t) = \frac{K}{5} [e^{3t} - e^{-2t}], \quad t > 0$$
$$= \frac{K}{5} [e^{3t} - e^{-2t}] u(t)$$

Example2:

Find $y[n]$ of the system with the difference equation $y[n] - \frac{1}{2}y[n-1] = x[n]$

We have the output $y[n] = x[n] + \frac{1}{2}y[n-1]$

Consider the input $x[n] = k \delta[n]$ and initial condition $y[-1] = 0$ (rest)

$$y[0] = x[0] + \frac{1}{2}y[-1] = k$$

$$y[1] = x[1] + \frac{1}{2}y[0] = \frac{1}{2}k$$

$$y[2] = x[2] + \frac{1}{2}y[1] = \left(\frac{1}{2}\right)^2 k$$

\vdots

$$y[n] = x[n] + \frac{1}{2}y[n-1] = \left(\frac{1}{2}\right)^n k$$

Setting $k = 1$ we obtain the *impulse response* for the system

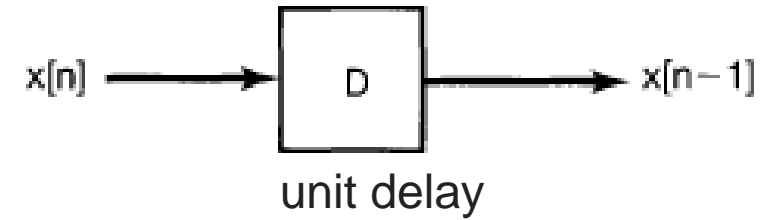
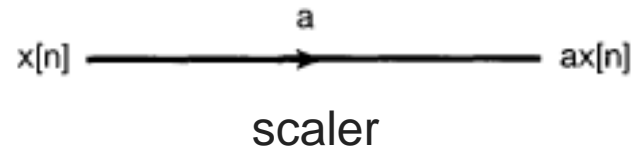
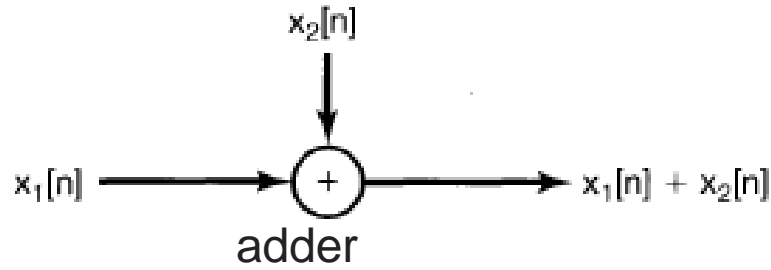
$$h[n] = \left(\frac{1}{2}\right)^n u[n]$$

impulse response with infinite duration

\Rightarrow *infinite impulse response (IIR) systems.*

Block Diagram Representations of Systems

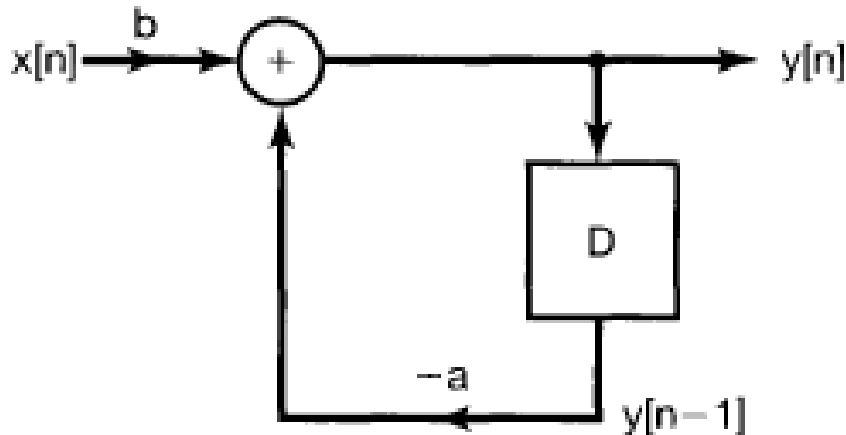
systems described by linear constant-coefficient difference and differential equations can be represented in terms of block diagram interconnections of elementary operations (adder, scaler, unit delay).



Example:

Consider the causal system described by the first-order difference equation

$$y[n] + a y[n-1] = b x[n]$$



Consider the causal continuous-time system described by a first-order differential equation

$$\frac{dy(t)}{dt} + ay(t) = b x(t)$$

