

- linear Time Invariant Systems. (LTI systems)

(Time Domain Representation)

Discrete time case

One of the crucial aspects of unit impulse i.e. the sifting property can be used to represent any signal $x[n]$ in terms of a train of ~~unit~~ impulses of varying magnitudes. Then a response of an LTI system to this unit impulse can be used to ~~def~~ represent a LTI system. This representation goes under the name "Impulse response".

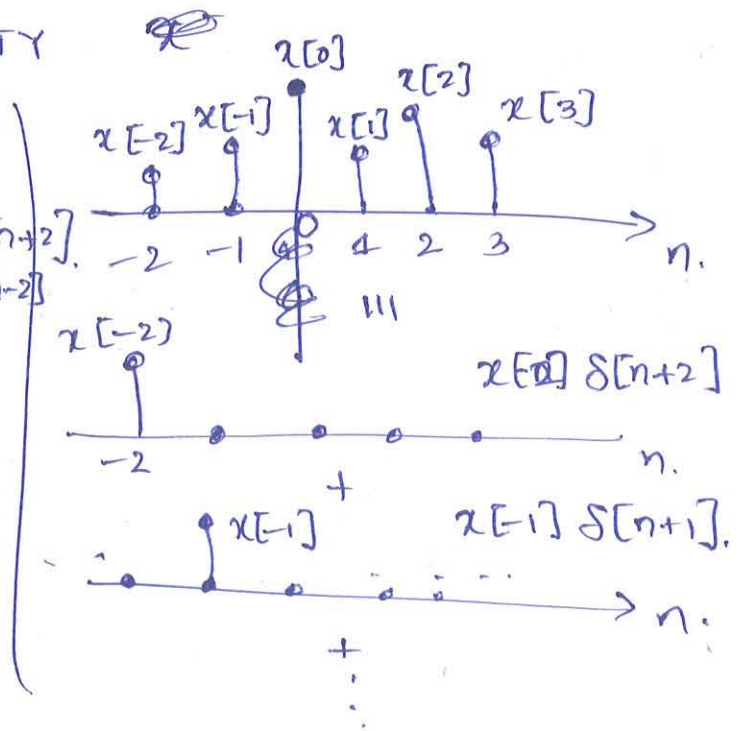
- SIFTING PROPERTY

Consider $x[n]$

then $x[n] \delta[n+2] = x[-2] \delta[n+2]$
 and so on. $x[n] \delta[n-2] = x[2] \delta[n-2]$

thus

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



(*)

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

$\delta[n]$

$$= x[\dots] + x[-2] \delta[n+2] + x[-1] \delta[n+1] + x[0] \delta[n] + x[1] \delta[n-1] + \dots$$

(*) is the representation of $x[n]$ as a train of impulses with varying magnitudes.

~~The~~ An impulse response of a System H

Now let us consider a system

$$y[n] = H\{x[n]\}$$

$$\text{Then } y[n] = H\left\{\sum_{k=-\infty}^{\infty} x[k] \delta[n-k]\right\}$$

If H is linear then by additivity and homogeneity we get

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

$$\text{Let } h_k[n] = H\{\delta[n-k]\}$$

$$\text{then } y[n] = \sum_{k=-\infty}^{\infty} h_k[n] x[k]$$

~~Q~~ Q

The signal $h_k[n]$ is the ~~input~~ response of H to a unit impulse $\delta[n-k]$.

In case we ~~were considering~~ ^{of} time varying systems ~~these~~, $h_k[n]$ for all $k \in \mathbb{Z}$ are unrelated (or no direct relation is seen)

But, in case of time invariant systems.
 $h_k[n] = h_0[n-k]$ (By definition of time invariance)
where $h_0[n]$ is the response of H to unit impulse $\delta[n]$.

And thus for any input signal $x[n]$ we obtain the response of H as follows.

$$y[n] = H\{x[n]\} = \sum_{k=-\infty}^{\infty} x[k] h_0[n-k]$$

Knowledge of $h_0[n]$ is sufficient to determine response of H to any signal $x[n]$.



The operation between two signals
~~is~~ $x[n]$ and $h_0[n]$ given by

$$\sum_{k=-\infty}^{\infty} x[k] h_0[n-k] := x[n] * h_0[n]$$

is ~~defined~~ defined to be the convolution
of $x[n]$ with $h_0[n]$. " $*$ " is used
to denote this binary operation.

Thus, for any signal $x[n]$
the response of a system H is ~~the~~
the convolution of signal $x[n]$ with
~~the~~ the impulse response of H

$$H\{x[n]\} = x[n] * h_0[n]$$

Remark: The impulse response $h_0[n]$ completely characterizes H (LTI system)

one more property of unit impulse.

$$H\{\delta[n]\} = \delta[n] * h_0[n] = h_0[n]$$

$$\Rightarrow \delta[n] * h_0[n] = h_0[n]$$

Convolution of unit impulse $\delta[n]$
any signal $x[n]$ results in $x[n]$,
itself.

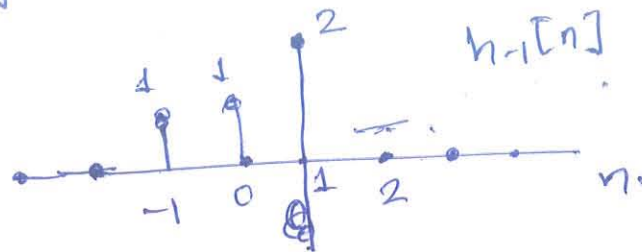
Stability -

Example: Time Varying System

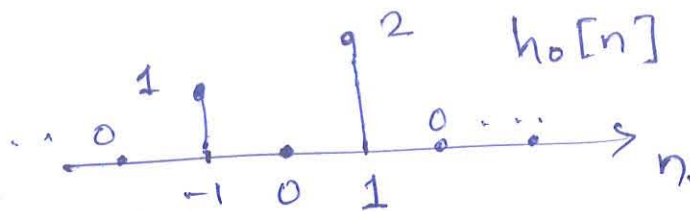
Can be specified by responses to $\delta[n-k]$ for $k \in \mathbb{Z}$

i.e. $H\{\delta[n-k]\} = h_k[n]$ for $k \in \mathbb{Z}$

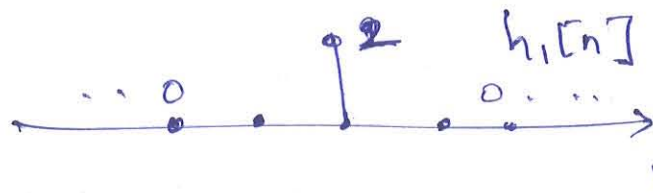
Let $h_{-1}[n]$



$h_0[n]$

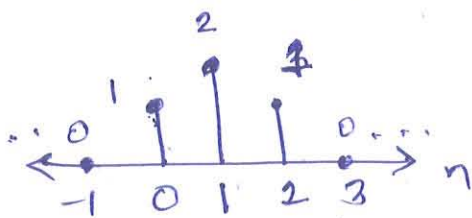


$h_1[n]$



and $h_k[n] \equiv 0$ for $k \in \mathbb{Z} - \{-1, 0, 1\}$.
(all k except $-1, 0, 1$)

then $x[n] \rightarrow \boxed{H} \rightarrow y[n] = H\{x[n]\}$



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

$$= x[0] \delta[n-0] + x[1] \delta[n-1] + x[2] \delta[n-2]$$

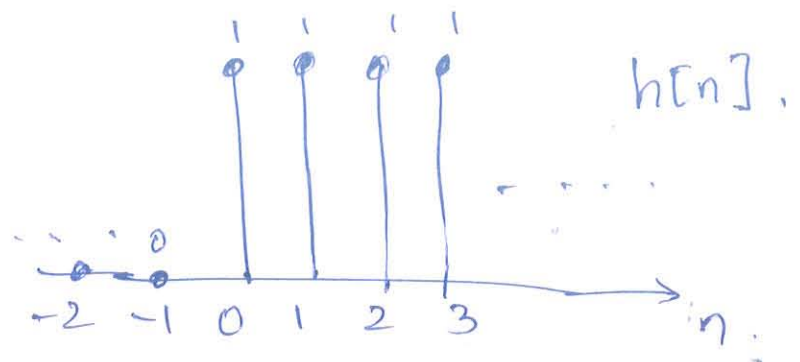
$$H\{x[n]\} = x[0] H\{\delta[n-0]\} + x[1] H\{\delta[n-1]\} + x[2] H\{\delta[n-2]\}$$

$$= x[0] h_0[n] + x[1] h_1[n] + x[2] h_2[n]$$

$$= h_0[n] + 2h_1[n] + h_2[n]. \text{ (pointwise addition)}$$

Example ¹ Time Invariant System

Consider $h[n] = u[n] = \begin{cases} 1, & n \geq 0 \\ 0, & n < 0 \end{cases}$



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

And so

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

$$= \dots + x[-1] h[n+1] + x[0] h[n] + x[1] h[n-1] + \dots$$

Let $x[n] \equiv 0$ for $n < 0$. but be non-zero for $n \geq 0$.

then

$$y[n] = x[0] h[n] + x[1] h[n-1] + \dots + x[n] h[n-n] + \dots$$

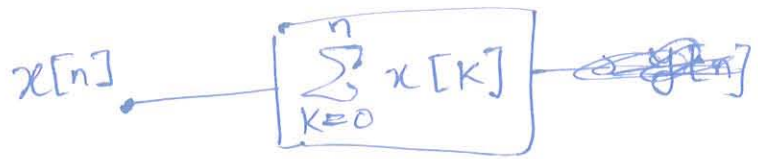
This gives us.

$$y[0] = x[0], \quad y[1] = x[0] + x[1]$$

$$y[2] = x[0] + x[1] + x[2], \quad \dots, \quad y[n] = \sum_{k=0}^n x[k], \quad \text{and so on.}$$

The impulse response

$u[n]$ represent a system that is an accumulator.



But let us spend some discussion on convolution operation

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

Now we can expand this sum and say that output is just the weighted and summed impulse responses shifted by amount k .

This is only one ~~way~~ interpretation.

There is another useful way to look at this sum.

At a particular instance 'n' consider $x[k] h[n-k]$ as a sequence in k . i.e., fix the value of 'n' and treat $x[k]$ as a sequence in k and $h[n-k]$ too.

Notice in earlier cases we needed $x[k]$ to be value of $x[n]$ at fixed k .

Let $g_n[k] = x[k] h[n-k]$ be a sequence.

Then $y[n]$ at our fixed value n

$$\text{is } \sum_{k=-\infty}^{\infty} g[k].$$

For demonstration. Let $n=0$ be fixed.

Then $g_0[k] = x[k] h[-k]$ is a sequence in k at $n=0$

$$\text{and } y[0] = \sum_{k=-\infty}^{\infty} g_0[k].$$

Similarly for $n=1$

$$g_1[k] = x[k] h[1-k]$$

$$\text{and } y[1] = \sum_{k=-\infty}^{\infty} g_1[k].$$

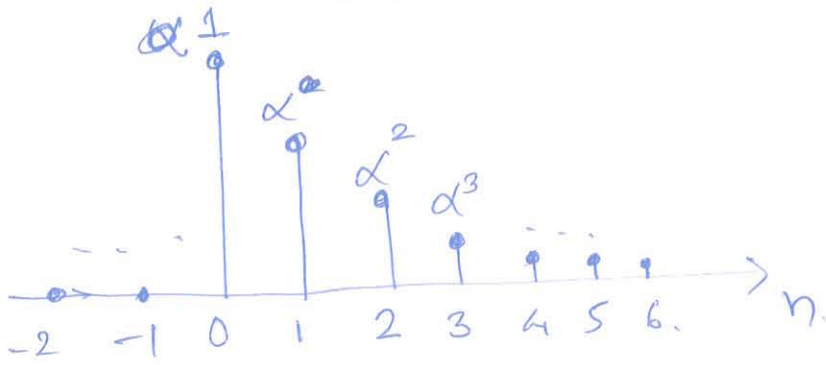
and so on. for $n=r$

$$g_r[k] = x[k] h[r-k]$$

$$\text{and } y[r] = \sum_{k=-\infty}^{\infty} g_r[k].$$

Example $x[n] = \alpha^n u[n]$

$h[n] = u[n]$.



Now we expect to get cumulative

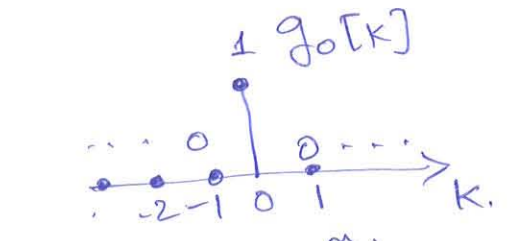
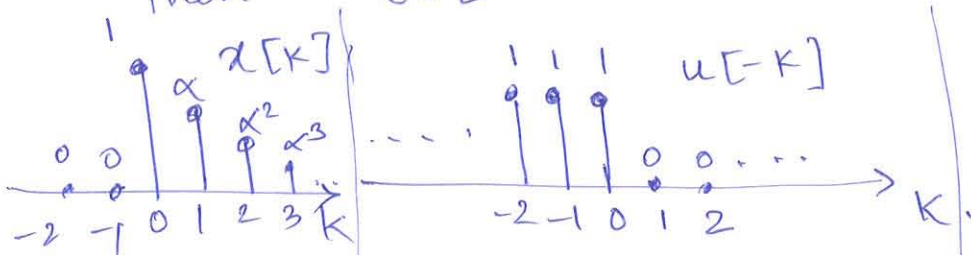
Sums. i.e. $y[k] = 1 + \alpha + \alpha^2 + \dots + \alpha^k$
 $= \frac{1 - \alpha^{k+1}}{1 - \alpha}$

for $k \geq 0$ since $x[n]$ ~~starts~~ is non-zero from $n=0$ onwards and is zero before.

Let $n=0$.

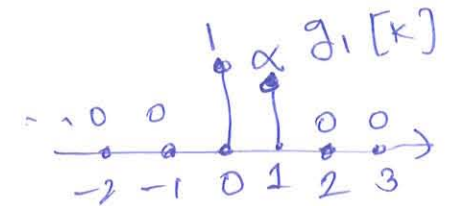
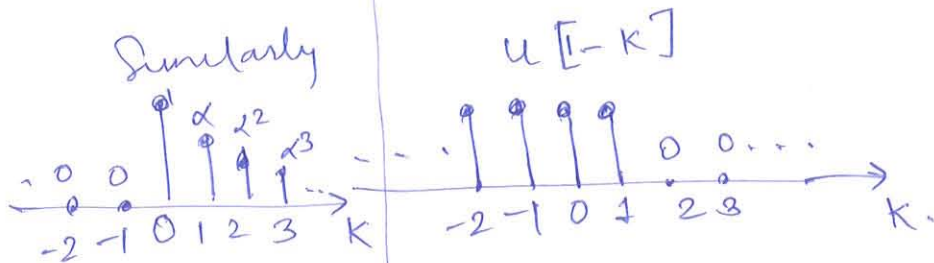
then

$$g_0[k] = x[k] h[-k] = x[k] u[-k]$$



and $y_0[k] = \sum_{k=-\infty}^{\infty} g_0[k]$
 $= 1$

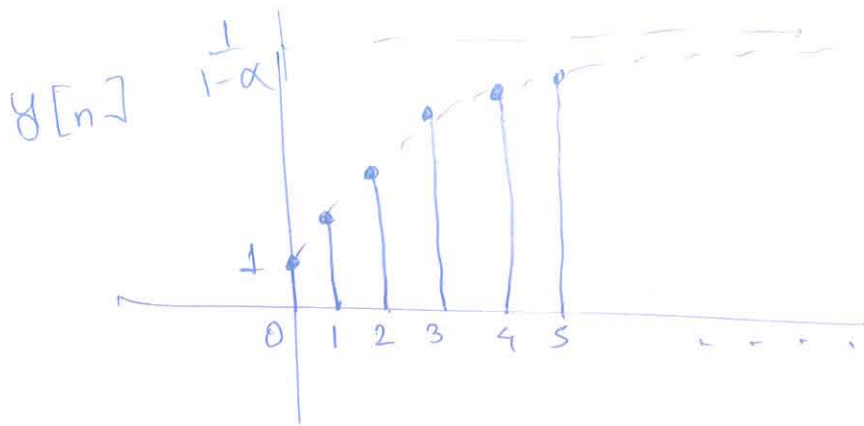
Similarly



and $y_1[k] = \sum_{k=-\infty}^{\infty} g_1[k]$

$= 1 + \alpha$.

and so on.



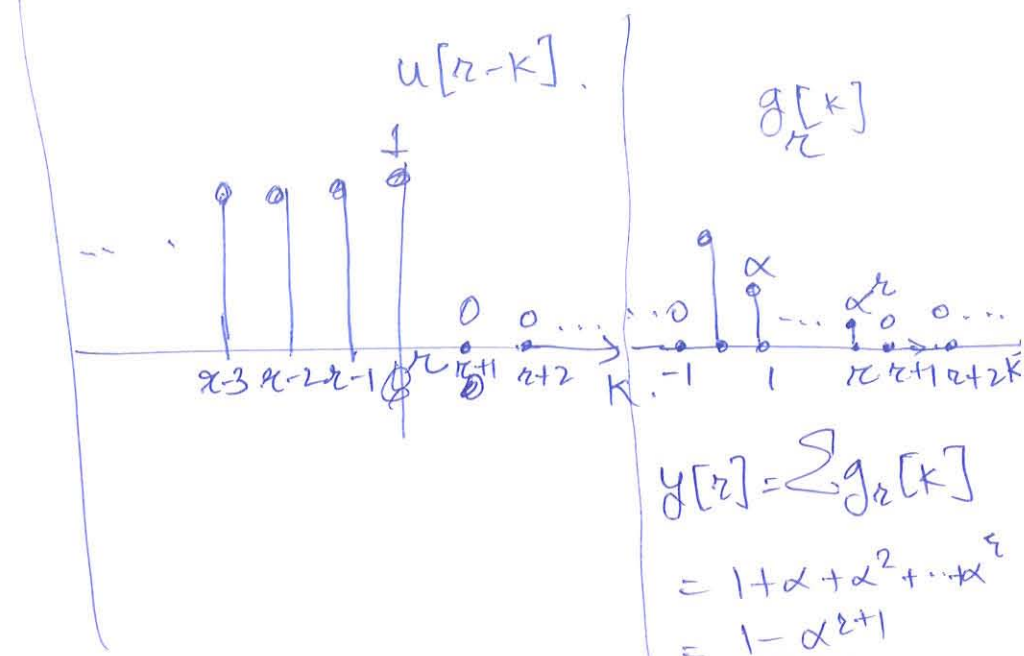
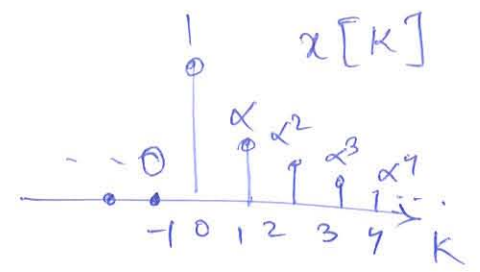
$$y[n] = \begin{cases} \frac{1-\alpha^{n+1}}{1-\alpha} & n \geq 0 \\ 0 & \text{otherwise} \end{cases}$$

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

and, as $n \rightarrow \infty$, $y[n] \rightarrow \frac{1}{1-\alpha}$

~~as $y[n]$~~

for $y[z]$.



$$y[z] = \sum g_k[k]$$

$$= 1 + \alpha + \alpha^2 + \dots + \alpha^z$$

$$= \frac{1 - \alpha^{z+1}}{1 - \alpha}$$

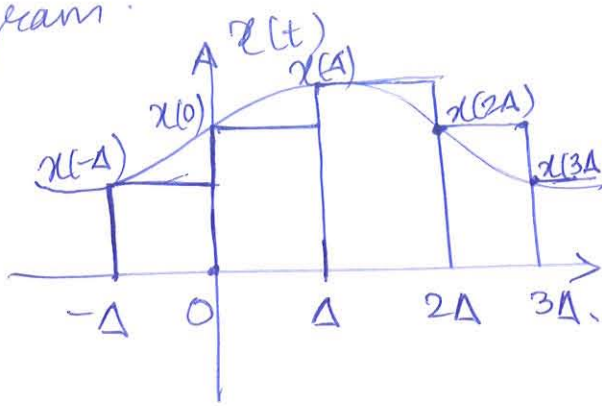
- Continuous time ~~non-linear~~ LTI system.

(Time Domain Representation)

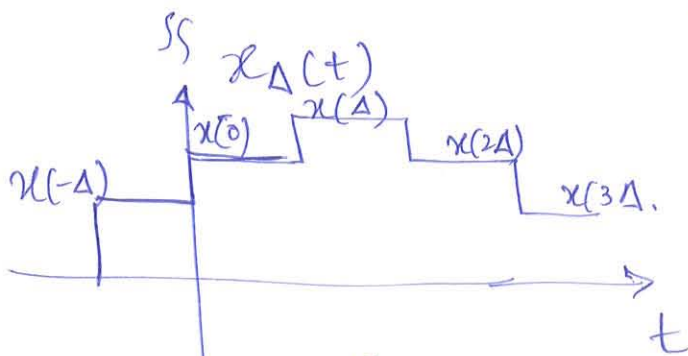
- Signal Representation

Any signal, $x(t)$ can be decomposed into a continuum of Dirac-delta $\delta(t)$ (which we saw briefly in earlier class).

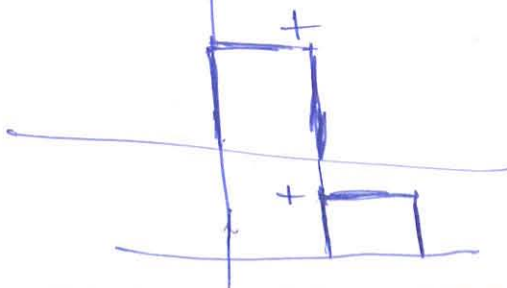
Signal $x(t)$ can be approximated by ~~discretizing~~ a sample and hold procedure as depicted in the following diagram.



- Sample $x(t)$ at $t = -\Delta, 0, \Delta, 2\Delta, \dots$ and hold the value in interval $[-\Delta, 0)$, $[0, \Delta)$, $[\Delta, 2\Delta)$ and so on.



||| Sum of individual pulses of width Δ and height $x(t)$.



Now let us consider the approximation to

Dirac-delta $\delta(t)$ (considered earlier) which

$$\text{is } \delta_{\Delta}(t) := \begin{cases} \frac{1}{\Delta}, & 0 \leq t < \Delta \\ 0, & \text{otherwise} \end{cases}$$

$$\delta_{\Delta}(t-\Delta) := \begin{cases} \frac{1}{\Delta}, & -\Delta \leq t < 0 \\ 0, & \text{otherwise} \end{cases}$$

and so on

$$\delta_{\Delta}(t-k\Delta) := \begin{cases} \frac{1}{\Delta}, & k\Delta \leq t < (k+1)\Delta \\ 0, & \text{otherwise.} \end{cases}$$

Using $\delta_{\Delta}(t)$ we can represent

$x_{\Delta}(t)$ as follows.

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

Now as $\Delta \rightarrow 0$, $x_{\Delta}(t) \rightarrow x(t)$

and $\delta_{\Delta}(t) \rightarrow \delta(t)$

and this summation can be replaced
by integration

~~Consider~~ i.e.,

$$\begin{aligned} x(t) &= \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{\infty} x(k\Delta) \delta_{\Delta}(t-k\Delta) \Delta \\ &= \int_{-\infty}^{\infty} x(\tau) \delta(t-\tau) d\tau \end{aligned}$$

This gives ~~the~~ ~~to~~ us a way to represent any signal $x(t)$ in terms of Dirac-delta $\delta(t)$ as follows.

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

Analogous to what we did for Discrete time systems now we can characterize LTI system by its impulse response.

For that first consider the response of system H to a pulse $\delta_{\Delta}(t)$.

$$H \{ \delta_{\Delta}(t - k\Delta) \} = h_{k\Delta}(t)$$

Then response of this system to the approximate signal $x_{\Delta}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta) \delta_{\Delta}(t - k\Delta)$

would be

$$y_{\Delta}(t) = H \left\{ x_{\Delta}(t) \right\} = H \left\{ \sum_{k=-\infty}^{\infty} x(k\Delta) \delta_{\Delta}(t - k\Delta) \Delta \right\}$$

$$= \sum_{k=-\infty}^{\infty} x(k\Delta) H \left\{ \delta_{\Delta}(t - k\Delta) \right\} \Delta = \sum_{k=-\infty}^{\infty} x(k\Delta) h_{k\Delta}(t) \Delta$$

Now as $\Delta \rightarrow 0$, $\delta_{\Delta}(t) \rightarrow \delta(t)$, $h_{k\Delta}(t) \rightarrow h(t)$
 $x_{\Delta}(t) \rightarrow x(t)$, $y_{\Delta}(t) \rightarrow y(t)$

4

$$\begin{aligned}
 y(t) &= \lim_{\Delta \rightarrow 0} \sum_{\Delta} y_{\Delta}(t) \\
 &= \lim_{\Delta \rightarrow 0} H \{ x_{\Delta}(t) \} \\
 &= \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{\infty} x(k\Delta) h_{k\Delta}(t) \Delta \\
 &= \int_{-\infty}^{\infty} x(\tau) h_{\tau}(t) d\tau
 \end{aligned}$$

~~This~~ For time varying system $h_{\tau}(t)$ would be the impulse response at various τ

In case of Time invariant systems

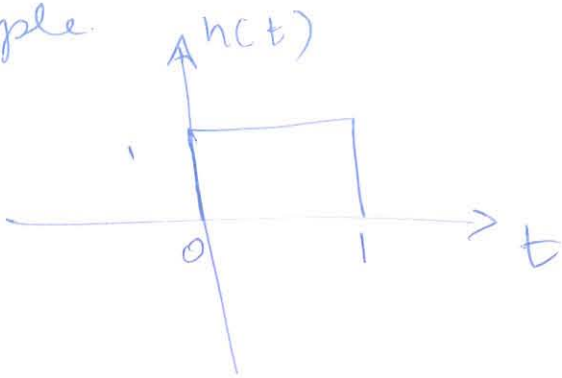
$$h_{\tau}(t) = h_0(t - \tau) \quad \left[\begin{array}{l} \text{Since} \\ h_0(t) = H \{ \delta(t) \} \\ \text{and} \\ h_{\tau}(t) = H \{ \delta(t - \tau) \} \end{array} \right]$$

and thus for LTI system the impulse response $h_0(t - \tau)$ gives a way to calculate output $y(t)$ for any input $x(t)$ as follows.

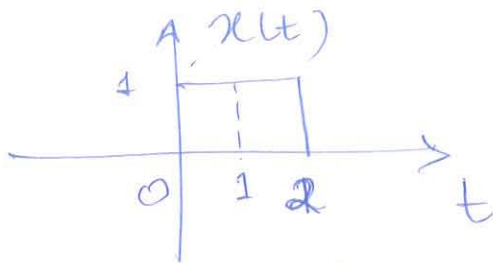
$$y(t) = \int_{-\infty}^{\infty} x(\tau) h_0(t - \tau) d\tau$$

This operation of $x(t)$ and $h_0(t)$ is ~~defined~~ as continuous time convolution

Example



$$h(t) = \begin{cases} 0, & t < 0 \\ 1, & t \in [0, 1] \\ 0, & t > 1 \end{cases}$$



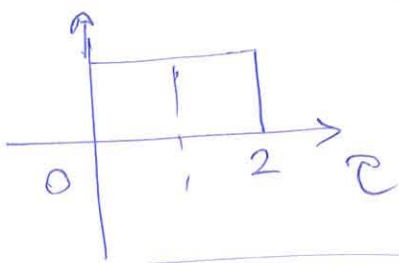
$$x(t) = \begin{cases} 0, & t < 0 \\ 1, & t \in [0, 2] \\ 0, & t > 2 \end{cases}$$

for

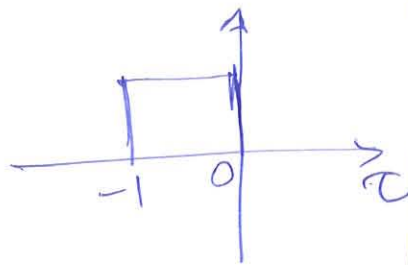
$t \leq 0$, $x(\tau)h(t-\tau) = 0$ because $x(\tau) = 0$ for $t < 0$

Consider

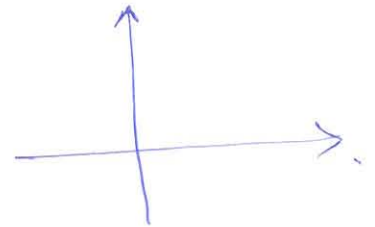
~~$x(\tau)$~~
 $x(\tau)$



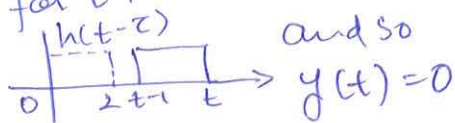
$h(-\tau)$



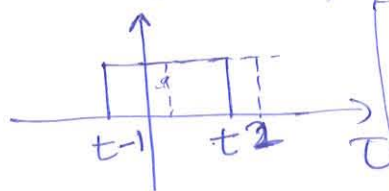
$x(\tau)h(-\tau)$



for $t > 3$

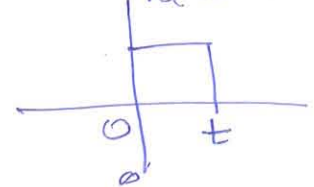


$h(t-\tau)$, $t \in [0, 1]$



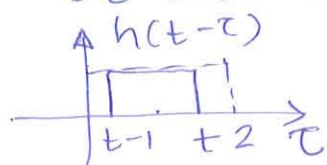
$t \in [0, 1]$

$x(\tau)h(t-\tau)$



$$y(t) = \int_{-\infty}^{\infty} x(\tau)h(t-\tau) d\tau = \int_0^t 1 d\tau = t$$

for $t \in [1, 2]$



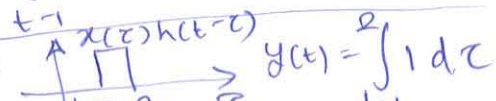
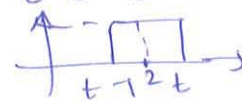
$x(\tau)h(t-\tau)$



$$y(t) = \int_{-\infty}^{\infty} x(\tau)h(t-\tau) d\tau = \int_{t-1}^t 1 d\tau = t - (t-1) = 1$$

for

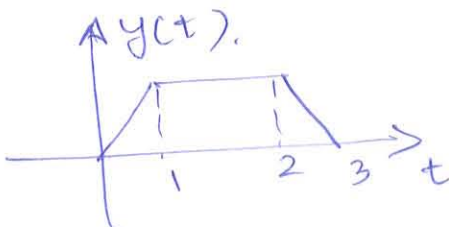
$t \in [2, 3]$



$$y(t) = \int_{t-1}^2 1 d\tau = 2 - (t-1) = 3 - t$$

Thus

$$y(t) = \begin{cases} 0, & t < 0 \\ t, & t \in [0, 1] \\ 1, & t \in [1, 2] \\ 3-t, & t \in [2, 3] \\ 0, & t > 3 \end{cases}$$



Example.

$$y(t) = x(t - t_0)$$

$$h(t) = \delta(t - t_0) \quad \text{delay by } t_0$$

Inverse of $h(t)$. $\rightarrow h_1(t) = \delta(t + t_0)$ advance by t_0

$$\begin{aligned} h(t) * h_1(t) &= \int_{-\infty}^{\infty} \delta(\tau - t_0) \delta(t + t_0 - \tau) d\tau \\ &= \int_{-\infty}^{\infty} \delta(\tau - t_0) \delta(t - (\tau - t_0)) d\tau \\ &= \int_{-\infty}^{\infty} \delta(\omega) \delta(t - \omega) d\omega \\ &= \delta(t). \quad \text{[Sifting Property]} \end{aligned}$$

Properties of ^{LTI} Systems

- Memorylessness (LTI system).

Since output depends on current instance!
input $y[n] = kx[n]$

Impulse response. $h[n] = k\delta[n]$

$$y(t) = kx(t)$$

Impulse response is $h(t) = k\delta(t)$

- If $h[n]$ (or respectively $h(t)$) is non-zero at $n \neq 0$ (respectively $t \neq 0$) then system has memory.

Properties of LTI systems / Convolution.

(i) Commutative.

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

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

Proof is left as exercise.

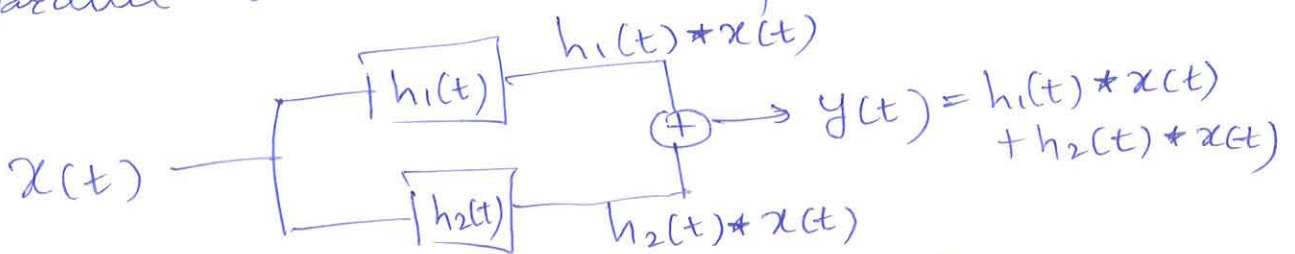
(ii) Distributive Property.

$$x[n] * \{h_1[n] + h_2[n]\} = x[n] * h_1[n] + x[n] * h_2[n]$$

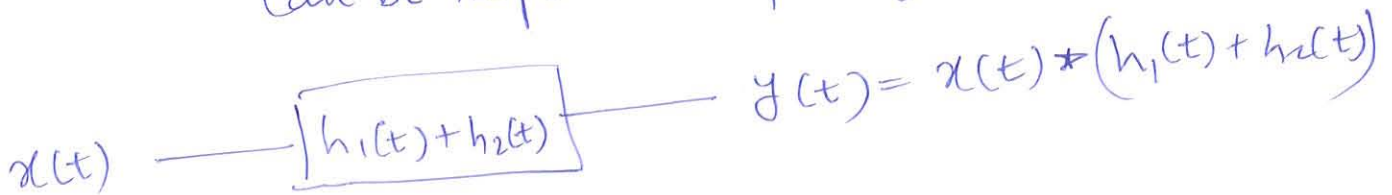
$$x(t) * \{h_1(t) + h_2(t)\} = x(t) * h_1(t) + x(t) * h_2(t)$$

[Proof is left as exercise].

This property helps in simplifying parallel interconnection of ~~linear~~ LTI systems.



can be replaced by single block.



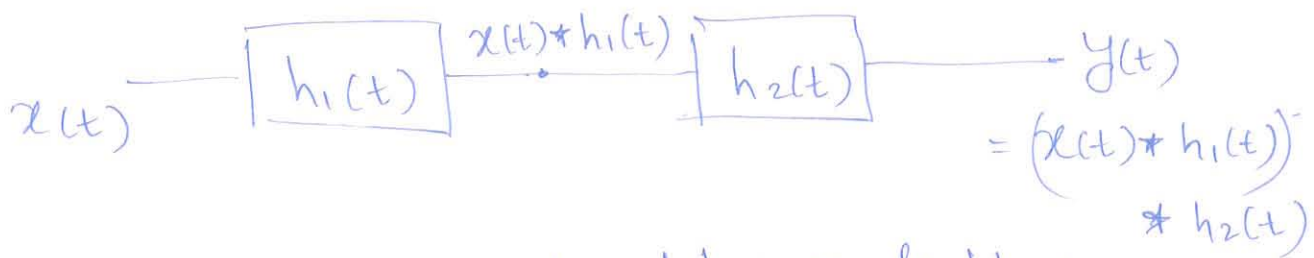
(iii) Associative Property.

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

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

$$x(t) * (h_1(t) * h_2(t)) = (x(t) * h_1(t)) * h_2(t).$$

This property helps in simplifying cascade interconnection of LTI systems.



Can be replaced by single block



(iv) Invertibility -

A system $h_1(t)$ is called as the inverse system of $h(t)$ if for any signal

$$x(t), \quad [x(t) * h(t)] * h_1(t) = x(t)$$

In particular, for $x(t) = \delta(t)$

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

Similarly for discrete time systems

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

$\delta(t)$, $\delta[n]$ are ~~xxxx~~ identity elements

in the space of continuous time & discrete time signals respectively.

Example

$$y(t) = x(t - t_0)$$

$$h(t) = \delta(t - t_0) \text{ impulse response}$$

(delay by t_0)

$$h_1(t) = \delta(t + t_0) \text{ is the inverse system.}$$

of $h(t)$.

Note.

$$h(t) * h_1(t) = \int_{-\infty}^{\infty} \delta(\tau - t_0) \delta(t + t_0 - \tau) d\tau$$

$$\stackrel{\infty}{=} \int_{-\infty}^{\infty} \delta(w) \delta(t - w) dw$$

Let $w = \tau - t_0$, then $dw = d\tau$

$$= \delta(t) \text{ [by sifting property]}$$

Causality of LTI System.

Causal if $y[n]$ does not depend on $x[k]$ for $k > n$ (future inputs)

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

Independence of $y[n]$ from $x[k]$ for $k > n$ if
and only if Impulse response $h[n-k]$ does not
pick non-zero $x[k]$ with $k > n$

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

$$\Rightarrow h[n] = 0 \text{ for } n < 0.$$

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

Analogously.

in Continuous time

Causal if $h(t) = 0$ for $t < 0$.

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

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

- Stability of systems.