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# Lecture 1, 8/25/2022

[TODO]

# Lecture 2, 8/30/2022

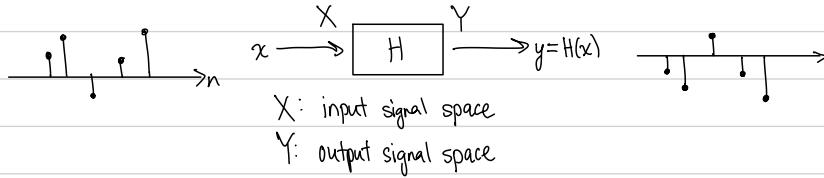
## Review

- Signals are functions

CT:  $x: \mathbb{R} \rightarrow \mathbb{R}$  or  $\mathbb{C}$  real CT signal  
complex CT signal

DT:  $x: \mathbb{Z} \rightarrow \mathbb{R}$

- Systems are functions



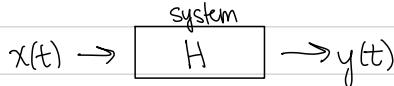
- If  $X = [\mathbb{R} \rightarrow \mathbb{R}]$  and  $Y = [\mathbb{R} \rightarrow \mathbb{R}]$ , we say that  $H$  is a CT system  
↑ set of all real-valued CT signals

- If  $X = [\mathbb{Z} \rightarrow \mathbb{R}]$  and  $Y = [\mathbb{Z} \rightarrow \mathbb{R}]$ , we say that  $H$  is a DT system  
↑ set of all real-valued DT signals

- Delta Impulse  $\delta(n)$ , Unit Step  $u(n)$

- Every DT Signal can be decomposed into a linear combo of shifted impulses  
↳  $x(n) = \sum_{k=-\infty}^{\infty} x(k) \delta(n-k)$

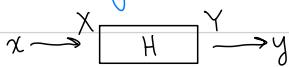
## Time-Invariant (LTI)



If  $\hat{x}(t) = x(t-T)$  yields  $\hat{y}(t) = y(t-T)$  for all  $T \in \mathbb{R}$  and all inputs  $x \in X$ , we say that  $H$  is time invariant

## Linearity

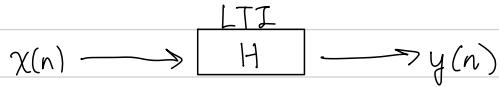
(Superposition)



For any  $x_1, x_2 \in X$  and  $\alpha_1, \alpha_2 \in \mathbb{C}$ ,  $H$  is linear if the following hold:

- Scaling:  $H(\alpha x_1) = \alpha H(x_1)$
- Additivity:  $H(x_1 + x_2) = H(x_1) + H(x_2)$

## Linear, Time-Invariant System



An LTI system satisfies both linearity and time-invariance

Claim:  $x(n) = \delta(n) \xrightarrow{\text{DT-LTI}} H \xrightarrow{} y(n) = h(n)$   
 (Impulse Response)

Knowing the impulse response  $h$  enables us to determine the output corresponding to any arbitrary input  $x$

$\hookrightarrow \text{Proof: } h(n) = H(\delta(n))$

$$h(n-k) = H(\delta(n-k))$$

$$x(k) h(n-k) = x(k) H(\delta(n-k)) = H(x(k) \delta(n-k))$$

$$\text{Thus, } \sum_{k=-\infty}^{\infty} x(k) h(n-k) = \sum_{k=-\infty}^{\infty} H(x(k) \delta(n-k)) = H\left(\sum_{k=-\infty}^{\infty} x(k) \delta(n-k)\right)$$

## Convolution

$$\text{Discrete-Time: } (x * h)(n) = \sum_{k=-\infty}^{\infty} x(k) h(n-k)$$

$$\text{Continuous-Time: } (x * h)(t) = \int_{-\infty}^{\infty} x(u) h(t-u) du$$

\* Commutative:  $x * h = h * x$

$\hookrightarrow$  Why care?

$$x(n) = \delta(n) \xrightarrow{\text{DT-LTI}} f \xrightarrow{\text{DT-LTI}} g \xrightarrow{\text{DT-LTI}} h(n) = (f * g)(n)$$

Cascade (Series)  
 Interconnection of LTI System

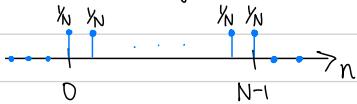
$$h(n) = (f * g)(n) = (g * f)(n) \text{ by commutativity (cascade order doesn't matter)}$$

### Ex. Simple Moving Average

$$x(n) \rightarrow [h(n)] \rightarrow y(n) = \frac{1}{N} \sum_{k=0}^{N-1} x(n-k)$$

Determine  $h(n)$  and plot it.

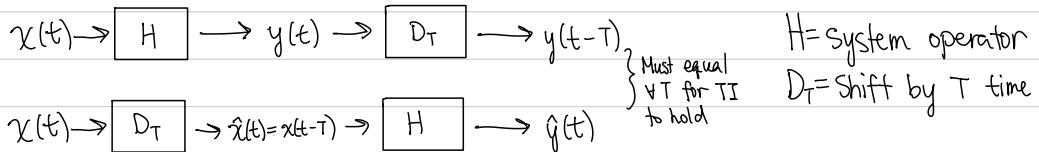
Let  $x(n) = \delta(n)$ . Then  $y(n) = h(n) = \frac{1}{N} \sum_{k=0}^{N-1} x(n-k)$



# Lecture 3, 9/1/2022

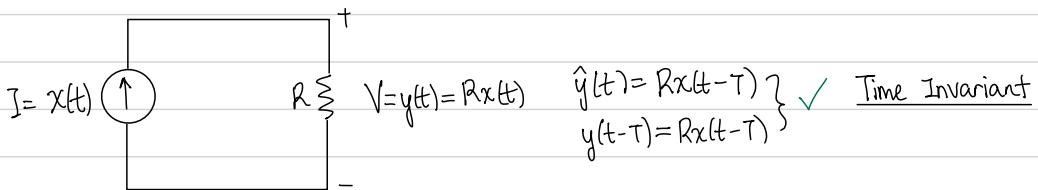
## Time Invariance (TI)

### Visualization of Time Invariance

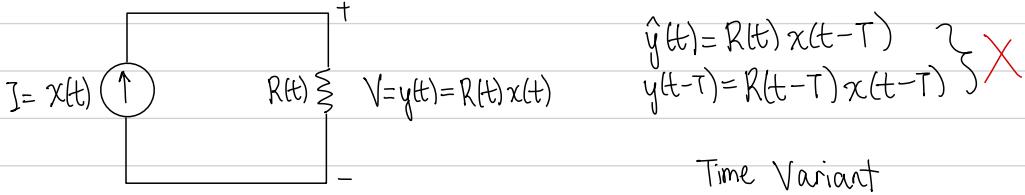


Basically, Time Invariance holds iff  $H \notin D_T$  are commutative

Ex 1.



Ex 2.



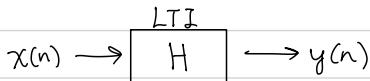
Ex 3. Two-Pointer Moving Average

$$x(n) \rightarrow [H] \rightarrow y(n) = \frac{x(n) + x(n-1)}{2}$$

• Linear?  $\hat{x}(n) = \alpha x_1(n) + \beta x_2(n) \rightarrow \hat{y}(n) = \frac{\hat{x}(n) + \hat{x}(n-1)}{2} = \frac{(\alpha x_1(n) + \beta x_2(n)) + (\alpha x_1(n-1) + \beta x_2(n-1))}{2} = \alpha y_1(n) + \beta y_2(n)$

• Time Invariant?  $\hat{x}(n) = x(n-N) \rightarrow \hat{y}(n) = \frac{\hat{x}(n) + \hat{x}(n-1)}{2} = \frac{x(n-N) + x(n-1-N)}{2}$

## Linear, Time-Invariant (LTI) Systems



Claim: If I know the response of  $H$  to the input  $x(n)=\delta(n) \forall n \in \mathbb{Z}$  (i.e. the impulse response  $h(n)=H(\delta(n))$ ) then I know the output for any arbitrary input signal  $x$ .

→ Proof/Explanation: We can express any input signal  $x(n)$  as follows:

$$x(n) = \sum_{m=-\infty}^{\infty} x(m) \delta(n-m)$$

and thus any output signal  $y(n)$  as follows (using LTI properties):

$$y(n) = \sum_{m=-\infty}^{\infty} x(m) h(n-m) = (x * h)(n) = (h * x)(n)$$

↑  
Convolutions are commutative

Cascading Systems: we can interchange the order of LTI systems in a cascade (series) interconnection



→  $y = x * f * g = x * g * f$ , though  $r \neq q$

## Convolutions

Def:  $(v * u)(n) = \sum_{k=-\infty}^{\infty} v(n) u(n-k) = \sum_{m=-\infty}^{\infty} v(n-m) u(m)$

Properties:

- Commutative:  $v * u = u * v$
- Identity Element:  $x * \underline{\delta} = x$

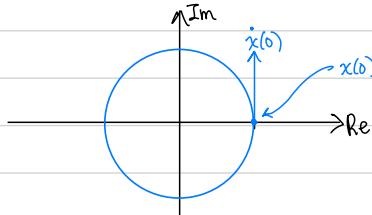
## Complex Exponentials

$$x(t) = e^t \longleftrightarrow \begin{cases} x(0) = 1 \\ \dot{x}(t) \triangleq \frac{dx(t)}{dt} = x(t) \\ \ddot{x}(t) \triangleq \frac{d^2x(t)}{dt^2} = x(t) \end{cases}$$

We use the properties of complex exponentials to solve interesting problems:

Ex.  $x(t) = e^{it}$  = the instantaneous position of a particle on the complex plane

$$\begin{aligned} x(0) &= 1 \\ \dot{x}(t) &= ie^{it} = ix(t) \\ \ddot{x}(t) &= i^2e^{it} = -x(t) \end{aligned}$$



Claim: the particle moves on the unit circle at a constant speed counter-clockwise

↳ Proof:  $x(t) = e^{it} = a(t) + ib(t)$        $a: \mathbb{R} \rightarrow \mathbb{R}, b: \mathbb{R} \rightarrow \mathbb{R}$

$$\dot{x}(t) = \dot{a}(t) + i\dot{b}(t)$$

$$\dot{x}(t) = ix(t) = -b(t) + ia(t)$$

$$\begin{aligned} b, \quad \dot{a}(t) &= -b(t) \\ a(t) &= b(t) \end{aligned}$$

$$\hookrightarrow a(t)\dot{a}(t) = -b(t)\dot{b}(t)$$

$$\hookrightarrow 2a(t)\dot{a}(t) + 2b(t)\dot{b}(t) = 0$$

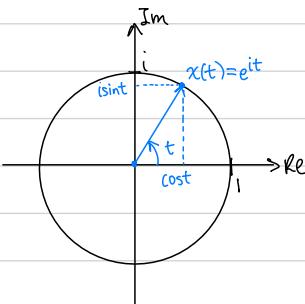
$$\hookrightarrow \frac{d}{dt} [a^2(t) + b^2(t)] = 0$$

$$\hookrightarrow a^2(t) + b^2(t) = C = 1 \quad \forall t \in \mathbb{R} \quad \therefore \text{Moves in circle of radius 1}$$

$$|\dot{x}(t)| = |ix(t)| = |x(t)| = 1 \quad \therefore \text{Moves at a constant speed at 1 rad/sec}$$

## Euler's Formula

$$e^{it} = \cos(t) + i\sin(t) \quad \forall t \in \mathbb{R}$$



# Lecture 4, 9/6/2022

## Euler's Formula

$$e^{it} = \cos t + i \sin t$$

$$e^{iwt} = \cos(wt) + i \sin(wt) \quad \text{where } w = \text{rotating speed in rad/sec}$$

$\begin{cases} > 0 & \rightarrow \text{counterclockwise} \\ < 0 & \rightarrow \text{clockwise} \end{cases}$

## Inverse Euler Formulas:

$$\begin{cases} \cos(wt) = \frac{e^{iwt} + e^{-iwt}}{2} \\ \sin(wt) = \frac{e^{iwt} - e^{-iwt}}{2i} \end{cases}$$

Note on different ways to measure rate:  $\omega = \frac{\text{rad/sec}}{\text{rad/cycle}} = 2\pi f$

$\uparrow \text{cycles/sec}$

## Periodicity

Main Idea:  $R \uparrow \longrightarrow J \downarrow$



A signal/system is periodic if it is cyclic,  
i.e. it eventually repeats

## CT Periodicity

Suppose we have a CT signal  $x: \mathbb{R} \xrightarrow{\text{or } \mathbb{C}} \mathbb{R}$ . If  $x(t+T) = x(t) \quad \forall t \in \mathbb{R}$  for some  $T \in \mathbb{R}_+$ , we say that  $x$  is periodic w/ period  $T$ . If  $T$  is the smallest positive value that satisfies the relation, then  $T$  is the fundamental period.

Ex.  $x(t) = \cos(\frac{2\pi}{5}t)$

$$x(t+T) = \cos(\frac{2\pi}{5}t + \frac{2\pi}{5}T) = x(t) = \cos(\frac{2\pi}{5}t)$$

$$\Rightarrow \frac{2\pi}{5}T = 2\pi k \quad \text{where } k \in \{1, 2, 3, \dots\}$$

$$\hookrightarrow T = 5 \text{ sec}$$

## DT Periodicity

Suppose we have a DT signal  $x: \mathbb{Z} \rightarrow \mathbb{C}$ . If  $x(n+N) = x(n) \forall n \in \mathbb{Z}$  for some  $N \in \mathbb{Z}_+$ , we say that  $x$  is periodic w/ period  $N$ . If  $N$  is the smallest such positive integer, we call  $N$  the fundamental period of  $x$ .

Ex.  $x(n) = C \quad \forall n \in \mathbb{Z}$

$$\hookrightarrow N=1, \omega_0 = \text{fundamental frequency} = \frac{2\pi}{N} = 2\pi$$

Ex.  $x(n) = e^{jn} \quad \forall n \in \mathbb{Z}$

$$x(n+N) = e^{j(n+N)} = e^{jn} e^{jN} = x(n) = e^{jn}$$

$$\hookrightarrow e^{jN} = 1$$

$$\hookrightarrow N = 2\pi \notin \mathbb{Z}$$

Thus, the fundamental period and frequency do not exist. This is called a Quasi-periodic signal.

Ex.  $x(n) = e^{j\frac{\pi}{4}n} \quad \forall n \in \mathbb{Z}$

$$x(n+N) = e^{j\frac{\pi}{4}(n+N)} = e^{j\frac{\pi}{4}n} \cdot e^{j\frac{\pi}{4}N} = x(n) = e^{j\frac{\pi}{4}n}$$

$$\hookrightarrow e^{j\frac{\pi}{4}N} = 1$$

$$\hookrightarrow N = \frac{4}{\pi} \cdot 2\pi = 8, \omega_0 = \frac{2\pi}{N} = \frac{\pi}{4}$$

## Periodicity of $x(n) = e^{jwn}$

Necessary & Sufficient conditions for  $e^{jwn}$  to be periodic in  $n$ .

$$e^{jw(n+N)} = e^{jwn}$$

$$e^{jwn} \cdot e^{jwN} = e^{jwn}$$

$$e^{jwN} = 1$$

$$wN = 2\pi k$$

$$w = \frac{2k}{N} \cdot \pi$$

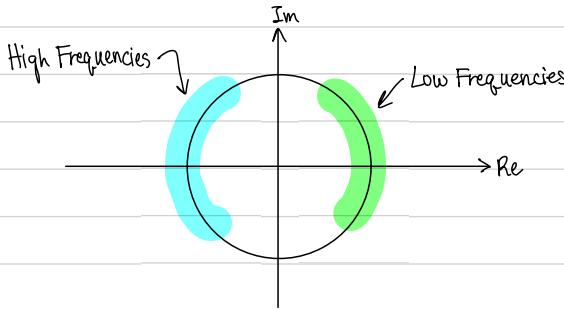
$\hookrightarrow w$  must be a rational multiple of  $\pi \Leftrightarrow N = \frac{2\pi k}{w} \in \mathbb{Z}_+$  can be found

## CT Oscillation Behavior

$x(t) = e^{i\omega t}$ . The sky is not a limit for  $\omega$ , i.e.  $\omega$  can go up to  $\infty$ . As  $\omega \rightarrow \infty$ , oscillations become progressively fast. The slowest frequency is  $\omega=0$ , which corresponds to a constant signal.

## DT Oscillation Behavior

$x(n) = e^{i\omega n}$ . The highest frequency is  $\omega=\pi$ , which corresponds to a "flip-flopping" signal. The slowest frequency is  $\omega=0$ , which corresponds to a constant signal.



## Frequency Response of a DT-LTI System

$$\text{DT-LTI} \quad x(n) = s(n) \xrightarrow{H} y(n) = h(n)$$

Impulse Response

$$x(n) = e^{i\omega n} \xrightarrow{H} y(n) = \sum_m h(m) x(n-m)$$

$$= \sum_m h(m) e^{i\omega(n-m)}$$

$$= \left( \sum_m h(m) e^{i\omega m} \right) e^{i\omega n}$$

$$= H(\omega) e^{i\omega n}$$

$$= H(\omega) x(n)$$

Frequency Response

\*  $H(e^{i\omega n}) = H(\omega) e^{i\omega n}$  is analogous to  $A\vec{v} = \lambda \vec{v}$ , and it's called the Eigenfunction Property of Complex Exponentials with respect to DT-LTI System.

\* LTI can get rid of frequency in the input signal, but it cannot create a new frequency in the output signal.

Ex.  $x(n) \rightarrow \boxed{H(\cdot)^2} \rightarrow y(n) = x^2(n)$   
 $x(n) = e^{i\omega n} \rightarrow \rightarrow y(n) = e^{i(2\omega)n}$

Thus, this system cannot be LTI because it creates a new frequency (i.e.  $2\omega$ ). It is time-invariant, but not linear.

Ex.  $x(n) \rightarrow \boxed{H} \rightarrow y(n) = e^{i\omega_0 n} x(n)$   
 $x(n) = e^{i\omega_0 n} \rightarrow \rightarrow y(n) = e^{i\omega_0 n} e^{i\omega_0 n} = e^{i(\omega_0 + \omega_0)n}$

Thus, the system cannot be LTI because it creates a new frequency (e.g.  $\omega_0 + \omega_0$ ). It is linear, but time-varying.

### Property of any LTI System

$$\alpha_0 e^{i\omega_0 n} + \alpha_1 e^{i\omega_1 n} \rightarrow \boxed{H} \stackrel{\text{LTI}}{\rightarrow} \alpha_0 H(\omega_0) e^{i\omega_0 n} + \alpha_1 H(\omega_1) e^{i\omega_1 n}$$

### Frequency Response of the Two-Point Moving Average Filter

$$x(n) \rightarrow \boxed{H} \rightarrow y(n) = \frac{x(n) + x(n-1)}{2}$$

Ways to find the frequency response  $H(\omega)$ :

Method I:  $H(\omega) = \sum_m h(m) e^{-i\omega m} = h(0) e^{-i\omega(0)} + h(1) e^{-i\omega(1)} = \frac{1+e^{-i\omega}}{2}$

Method II: Next time!

# Lecture 5, 9/8/2022

## Overview

- DT Frequency Response
  - 2 point Moving Average Filter
  - Recursive Filter

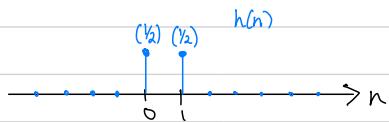
## DT Frequency Response

$$x(n) = e^{i\omega n} \rightarrow \boxed{H} \longrightarrow y(n) = H(\omega) e^{i\omega n}$$

$$H(\omega) = \sum_{n=-\infty}^{\infty} h(n) e^{-i\omega n} = \text{Freq. Response of Filter } H$$

## 2-Point Moving Average Filter

$$x(n) \longrightarrow \boxed{\frac{H}{h(n), H(\omega)}} \longrightarrow y(n) = \frac{x(n) + x(n-1)}{2}$$



$$h(n): \text{ let } x(n) = s(n) \rightarrow y(n) = h(n)$$

$$h(n) = \frac{s(n) + s(n-1)}{2}$$

## Computing $H(\omega)$ (using 2-pt moving average as example)

Method I: Use definition of  $H(\omega)$

$$H(\omega) = \sum_n h(n) e^{-i\omega n}$$

$$H(\omega) = h(0) + h(1)e^{-i\omega} = \frac{1 + e^{-i\omega}}{2}$$

Method II: Use eigenfunction property of complex exponentials wrt LTI systems

Let  $x(n) = e^{i\omega n} \rightarrow y(n) = H(\omega) e^{i\omega n}$ . Then we have:

Note: this method assumes that the frequency response exists

$y(n) = x(n) + x(n-1)$  (Linear Constant Coefficient)

$$H(\omega) e^{i\omega n} = \frac{e^{i\omega n} + e^{i\omega(n-1)}}{2} = \frac{1 + e^{-i\omega}}{2} e^{i\omega n}$$

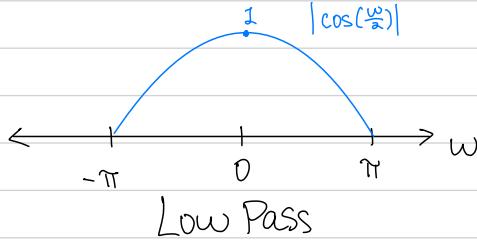
$$H(\omega) = \frac{1 + e^{-i\omega}}{2}$$

## Plotting Frequency Response (using 2-pt moving average as example)

$$H(\omega) = |H(\omega)| e^{j\Delta H(\omega)}$$

↓  
Magnitude Response      Phasor Response

$$\begin{aligned} |H(\omega)| &= \sqrt{\left(\frac{1}{2} + \frac{1}{2} \cos(\omega)\right)^2 + \left(-\frac{1}{2} \sin(\omega)\right)^2} \quad \leftarrow \text{One Way} \\ &= \left| \frac{1}{2} e^{j(\omega)} + \frac{1}{2} e^{j(-\omega)} \right| \quad \leftarrow \text{Better Way (Exploit Symmetry)} \\ &= \left| \frac{e^{j\omega_2} + e^{-j\omega_2}}{2} \right| e^{-j\omega_2} \\ &= \left| \cos\left(\frac{\omega}{2}\right) e^{-j\omega_2} \right| \\ &= \left| \cos\left(\frac{\omega}{2}\right) \right| \end{aligned}$$



\* We only consider the range  $[-\pi, \pi]$  because those are limits of a DT filter

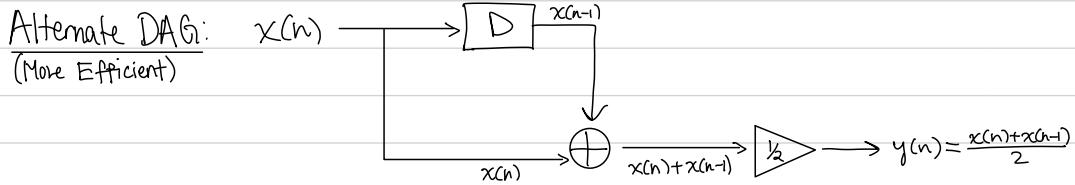
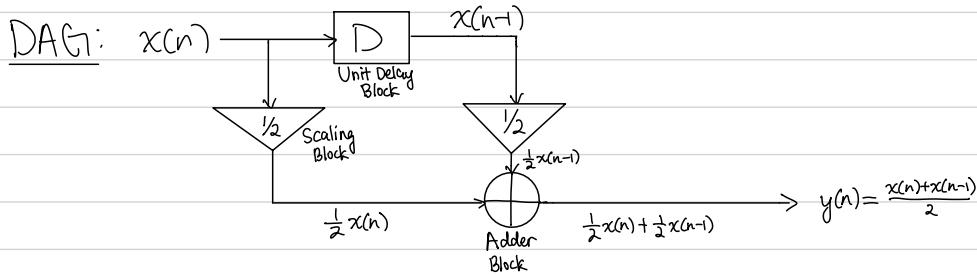
Note:  $|H(\omega)| = |H(\omega+2\pi)|$  holds for all DT filter frequency responses

## $2\pi$ -Periodicity of DT-LTI Frequency Response

$$\begin{aligned} H(\omega) &= \sum_n h(n) e^{-i\omega n} \\ H(\omega+2\pi) &= \sum_n h(n) e^{-i(\omega+2\pi)n} \\ &= \sum_n h(n) e^{-i\omega n} \underbrace{e^{-i(2\pi)n}}_{=1 \forall n \in \mathbb{Z}} \\ &= H(\omega) \end{aligned}$$

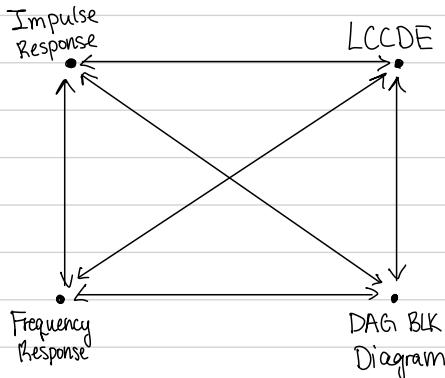
## Delay-Adder-Gain Block Diagram Implementation

Normal:  $x(n) \rightarrow \boxed{H}$   $\xrightarrow{h(n), H(w)} y(n) = \frac{x(n) + x(n-1)}{2}$



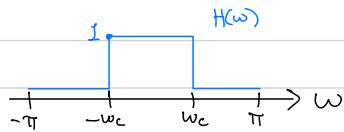
## The Big Picture

Moving between various characterizations of LTI systems

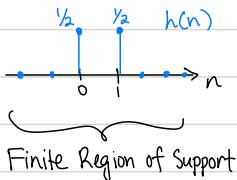


## DT-LTI System w/o LCCDE Representation

$$h(n) = \frac{\sin(\omega_c n)}{\omega_c n} \quad \forall n \in \mathbb{Z} \quad \longleftrightarrow$$



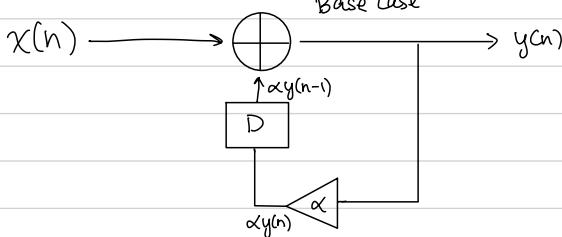
## Finite-Duration Impulse Response (FIR) Filter



## Infinite-Duration Impulse Response (IIR) Filter

$$y(n) = \alpha y(n-1) + x(n), \quad y(-\underbrace{1}) = 0, \quad |\alpha| < 1$$

Base Case



Determine  $h(n)$ :  $h(-1) = 0$

$$h(n) = \alpha h(n-1) + \delta(n)$$

$$h(0) = \alpha h(-1) + \delta(0) = 1$$

$$h(1) = \alpha h(0) + \delta(1) = \alpha$$

$$h(2) = \alpha h(1) + \delta(2) = \alpha^2$$

$$h(3) = \alpha h(2) + \delta(3) = \alpha^3$$

⋮

$$h(n) = \alpha^n \quad \forall n \geq 0$$

$\Rightarrow h(n) = \alpha^n u(n)$  where  $u(n) = \text{unit step}$

## Frequency Response for Recursive Filter

$$y(n) = \alpha y(n-1) + x(n), \quad y(-1) = 0, \quad |\alpha| < 1$$

$$h(n) = \alpha^n u(n)$$

Method I:  $H(\omega) = \sum_{n=-\infty}^{\infty} h(n) e^{-i\omega n}$

$$\begin{aligned} &= \sum_{n=-\infty}^{\infty} \alpha^n u(n) e^{-i\omega n} \\ &= \sum_{n=0}^{\infty} \alpha^n e^{-i\omega n} \\ &= \sum_{n=0}^{\infty} (\alpha e^{-i\omega})^n \\ &= \frac{1}{1 - \alpha e^{-i\omega}} \end{aligned}$$

Note:  $|\alpha e^{-i\omega}| = |\alpha| |e^{-i\omega}| = |\alpha| < 1$

Method II:  $x(n) = e^{i\omega n} \rightarrow y(n) = H(\omega) e^{i\omega n}$

$$\begin{aligned} y(n) &= \alpha y(n-1) + x(n) \\ &= \alpha H(\omega) e^{i\omega(n-1)} + e^{i\omega n} \\ &= (\alpha H(\omega) e^{-i\omega} + 1) e^{i\omega n} \\ &\equiv H(\omega) e^{i\omega n} \end{aligned}$$

$$H(\omega) = \alpha H(\omega) e^{-i\omega} + 1$$

$$1 = (1 - \alpha e^{-i\omega}) H(\omega)$$

$$H(\omega) = \frac{1}{1 - \alpha e^{-i\omega}}$$