

BLG 354E Signals & Systems

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Recap: Any periodic DT signal: $\tilde{\omega}_0$

$$x[n] = \sum_{k=0}^{N-1} a_k e^{j\frac{2\pi}{N} k n}$$

↑ coefficients
↑ basis func_k

Sequence $x(n)$ is periodic w/ fundamental period N .

You have to determine this

Fourier synthesis

Fourier analysis

$$a_k = \frac{1}{N} \sum_{n=0}^{N-1} x[n] \cdot e^{-j\frac{2\pi}{N} k n}$$

w/ a_k periodic w/ N .
 $; a_{k+N} = a_k$

↑
sum over time index

DTFS \longleftrightarrow CTF S

finite $\Rightarrow N$

distinct harmonic components

$$e^{j\frac{2\pi}{N} k n}$$

$$\sum_{k=-\infty}^{\infty} a_k e^{t j 2\pi f_0 k t}$$

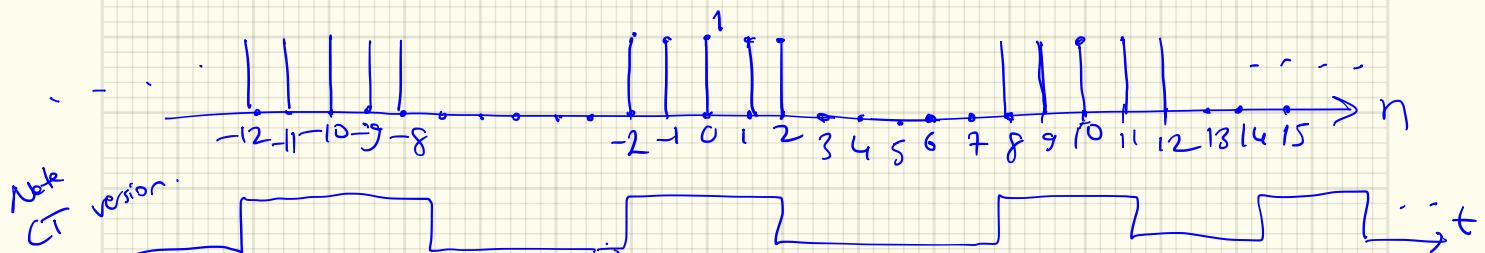
Can have
only many distinct a_k 's.
↓
= harmonic component

Def: $|a_k|^2$ → its graph vs $\hat{f} = \frac{k}{N}$ or $\hat{\omega} = 2\pi \frac{k}{N}$, or simply k ,
 is known as the Power Spectrum of the periodic signal $x[n]$

Ex: Rectangular Pulse Train Sequence: , $N=10$; ^{say}

In a period : $x(n) = \begin{cases} 1 & , -2 \leq n \leq 2 \\ 0 & , \text{o/w} \end{cases}$

$$x[n+10] = x[n] \checkmark$$



F.S. $a_k = \frac{1}{N} \sum_{n=-L}^{L=2} 1 \cdot e^{-j \frac{2\pi}{N} kn}$, $a_0 = \frac{2L+1}{N} = \frac{5}{10} = \frac{1}{2}$ q.

$$a_k = \frac{1}{10} \sum_{n=-2}^2 1 e^{-j \frac{2\pi}{10} kn} \Rightarrow a_1 = \frac{1}{10} \sum_{n=-2}^2 e^{-j \frac{2\pi}{10} n}$$

$\Rightarrow a_1 = \frac{1}{10} \left(e^{j \frac{4\pi}{10}} + e^{j \frac{6\pi}{10}} + 1 + e^{j \frac{8\pi}{10}} + e^{-j \frac{4\pi}{10}} \right) = \frac{1}{10} \left(2\cos\left(\frac{4\pi}{10}\right) + 1 + 2\cos\left(\frac{2\pi}{10}\right) \right)$

$$a_2 = a_{11} \dots$$

$$a_{k+N} = a_k$$

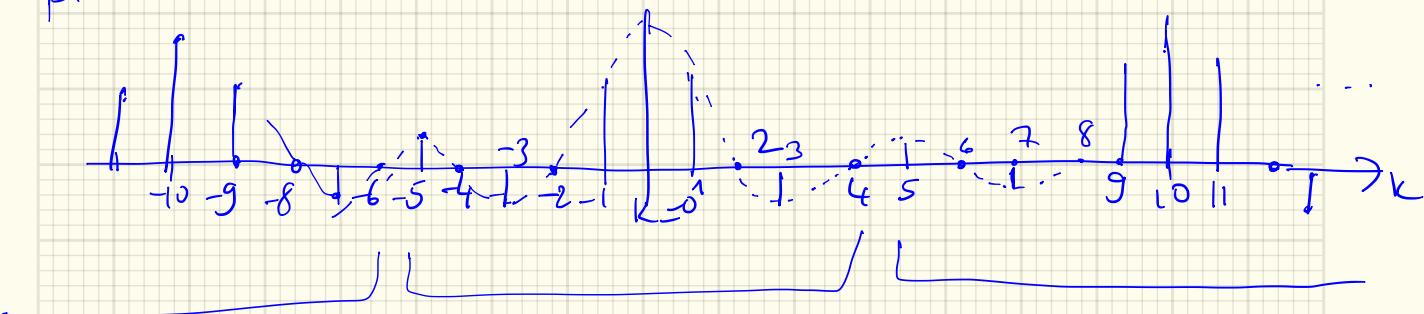
$$\Rightarrow a_k =$$

for signals
 $a_k \neq 0 \forall k$

$$a_k = \begin{cases} \frac{2L+1}{N}, & k = 0, \pm N, \pm 2N, \dots \\ \frac{1}{N} \left[\frac{\sin \frac{2\pi}{N} k(L+y_2)}{\sin(\frac{2\pi}{N} k \cdot \frac{1}{2})} \right], & \text{otherwise} \end{cases}$$

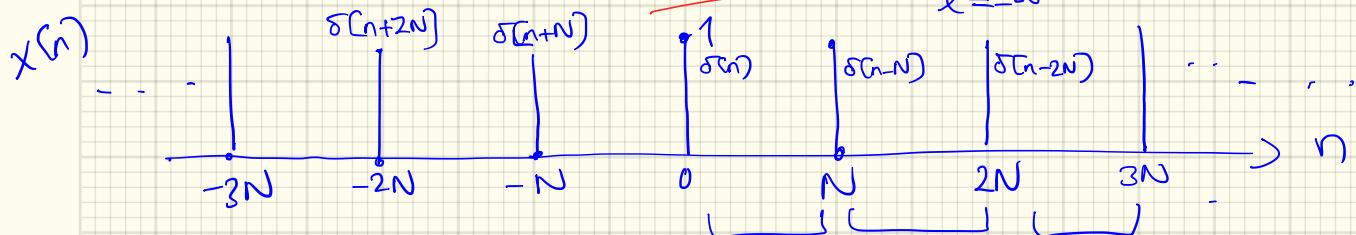
Exercise
derive this
expression

plot the spectrum for $N=10, L=2$



See the periodic form.

Ex: Periodic Impulse Train $x[n] = \sum_{l=-\infty}^{\infty} \delta[n - lN]$



DT Fourier series coeff of $x[n] = ?$

$$a_k = \frac{1}{N} \sum_{n=0}^{N-1} x[n] e^{-j \frac{2\pi}{N} k n}$$

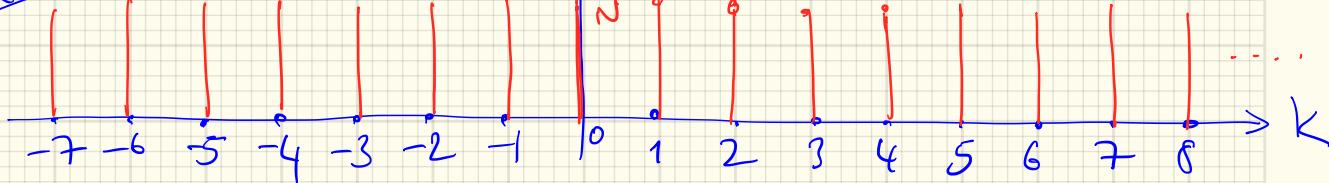
$\underbrace{x[n]}_{\delta[n]=1} \quad \underbrace{e^{-j \frac{2\pi}{N} k n}}_{=1}$

$$= \frac{1}{N} \sum_{n=0}^{N-1} e^{-j \frac{2\pi}{N} k n}$$

$\underbrace{\sum_{n=0}^{N-1} e^{-j \frac{2\pi}{N} k n}}_{=1} = \boxed{\frac{1}{N}} = a_k$

sum over 1 period

Spectrum of $x[n]$



use Fourier synthesis

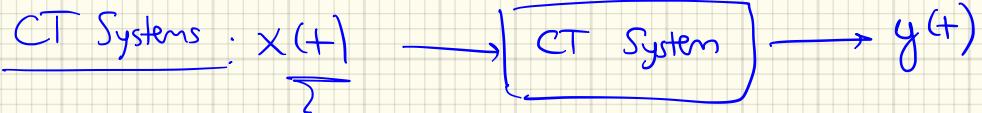
$$x[n] = \sum_{k=0}^{N-1} a_k \frac{1}{N} e^{j \frac{2\pi}{N} k n} \quad \forall n$$

Chapter 5

SYSTEMS

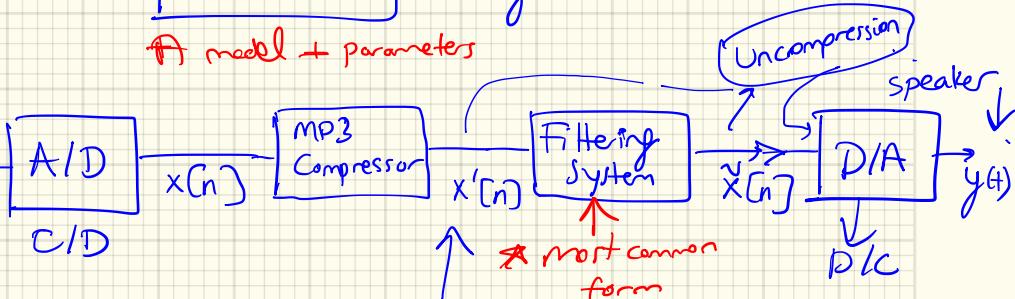
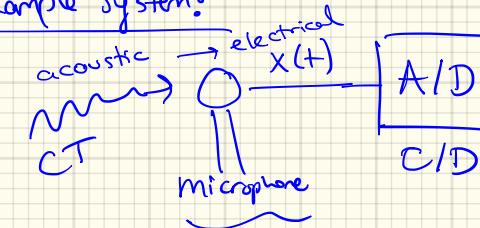
: Maps an input (signal/data) to an output (signal)

\approx Algorithms



A model + parameters

Example System:



C/D
D/A
 we will cover these
 after we cover Fourier
 transform.

① Systems for removing unwanted features such as noise etc. from signal.

Filtering :



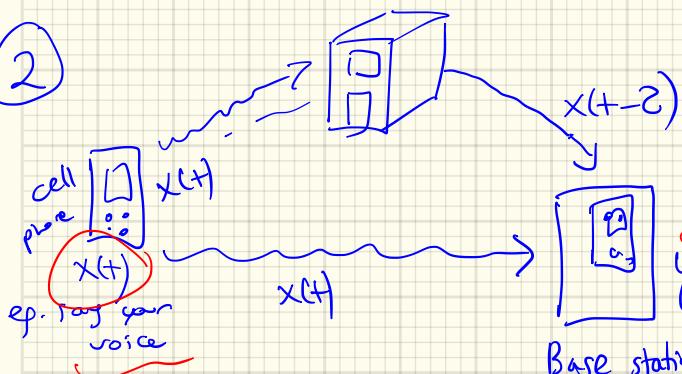
We believe

input is corrupted w/ noise

or input needs to be simplify

filter w/ your belief /assumption about the corruption or unnecessary components

②



ep. say your voice

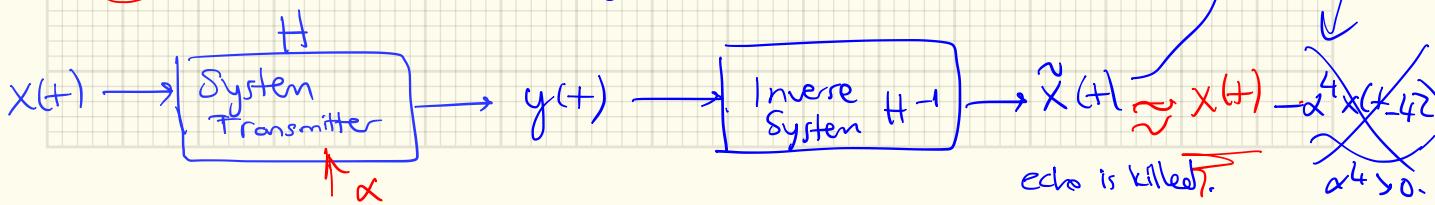
$$\begin{aligned} \tilde{x} &= y(t) - \alpha y(t-2) \\ &+ \alpha^2 y(t-2^2) \\ &- \alpha^3 y(t-3^2) \end{aligned}$$

insert

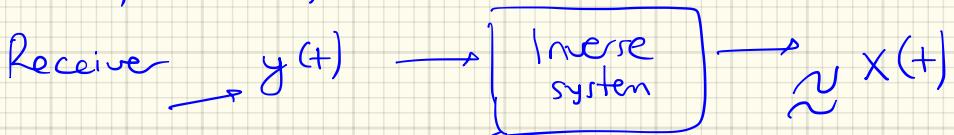
$$y(t) = X(t) + \alpha \cdot X(t-2)$$

received signal echo

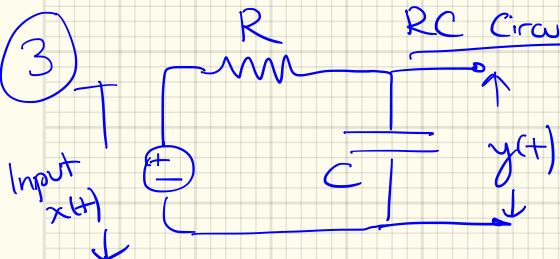
$$\alpha \ll 1$$



→ This is an example of an inverse system ; tries to eliminate unwanted parts ("echoes") in the signal.



③



RC Circuit System

: CT System can be represented by a

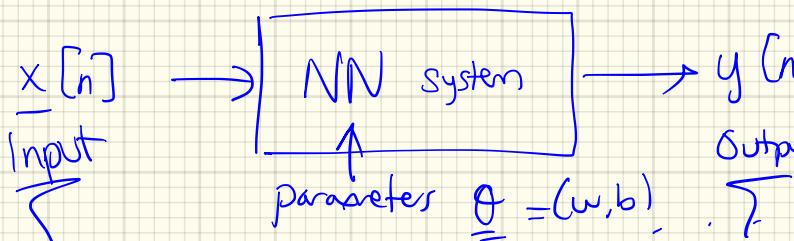
$$\frac{dy}{dt} + \frac{1}{RC} y(t) = \frac{1}{RC} x(t)$$

Differential Equations.

R, C : parameters of the system

④

NN : Neural Networks : Nonlinear System example



parameters $\underline{\theta} = (w, b)$

Deep NNs: millions of parameters

$$y[n] = g(\underline{\theta} \underline{x}) = g(g(g(g(\underline{\theta} \underline{x}))))$$

↑ nonlinearity

- Systems
- 1) model the effects of a physical phenomenon
 - 2) Implement a desired effect on the signal (that we recorded)

e.g. $\frac{ML}{\downarrow}$ Machine Learning : extract features from the data.

Discrete-time Systems : (DT)

DT Input Signal $x[n] \rightarrow$ [DT System] $\rightarrow y[n]$

T : operator that models the system
(mapping)

$$y[n] = T[x[n]]$$

models
maps the input x to the output y

I/O (Input/output) relation

Ex: $x[n] \xrightarrow{S} y[n]$: I/O relation of the system S

is given by

$$y[n] = \frac{1}{3}x[n] + \frac{1}{3}x[n-1] + \frac{1}{3}x[n-2]$$

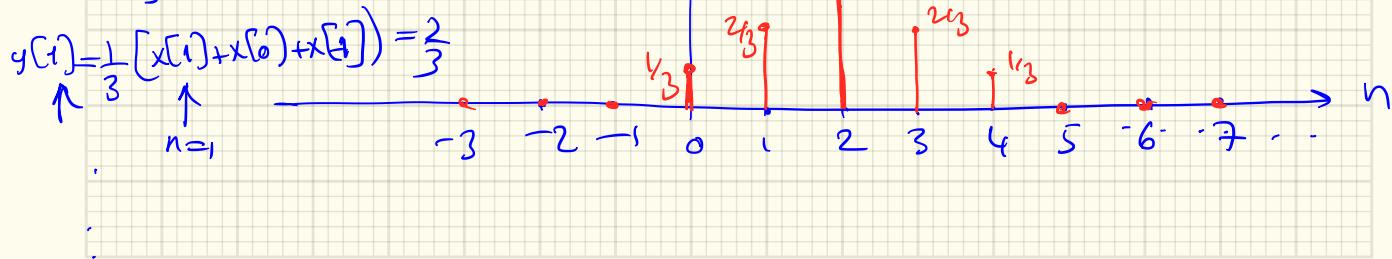
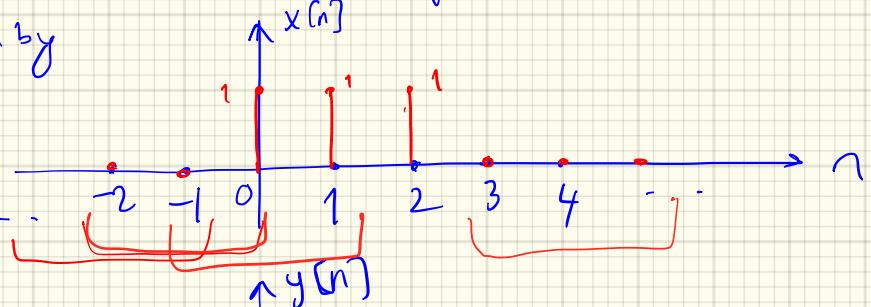
3-pt Average System.

(map) that says output at time n is average of 3 consecutive inputs.

e.g. say $x[n]$ is given by

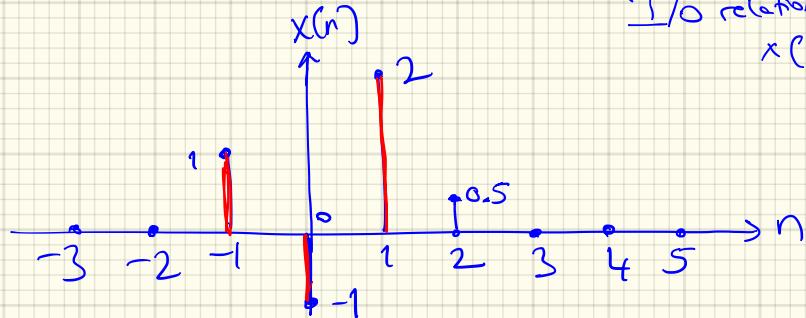
$$y[0] = \frac{1}{3}(x[0] + x[-1] + x[-2]) \\ = \frac{1}{3}$$

$$y[1] = \frac{1}{3}(x[1] + x[0] + x[-1]) = \frac{2}{3}$$



$$\text{Ex: } y[n] = (x[n])^2$$

e.g.

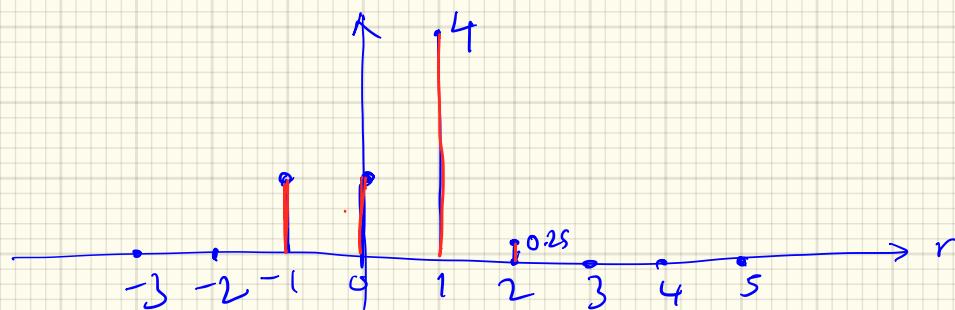


Square system

I/O relation of

$$x(n) \rightarrow [S]$$

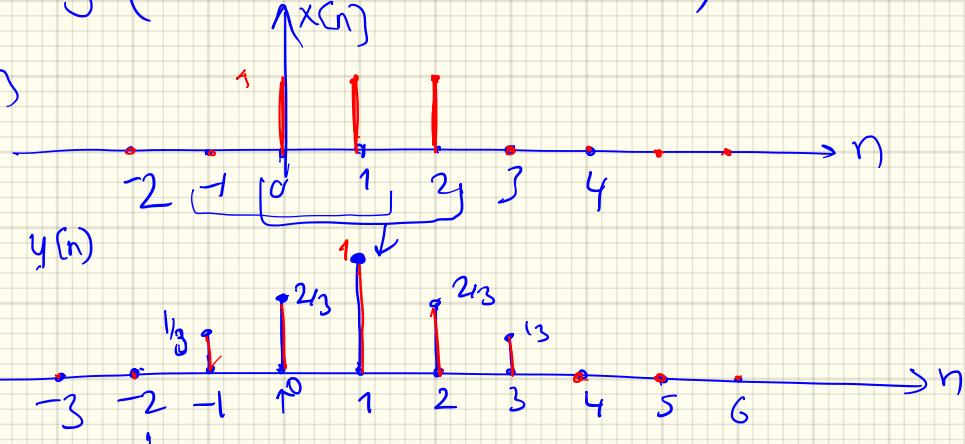
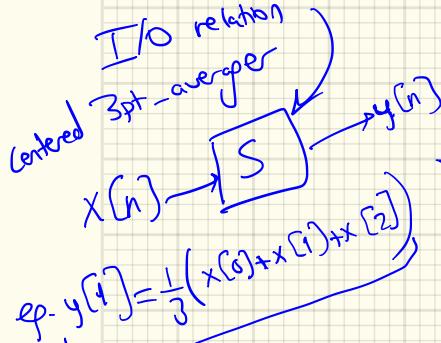
$$y[n] = (x[n])^2$$



Note: The value of the output depends just at the value of the input at the same time point.

(∴ Causal system)

$$\text{Ex: } y[n] = \frac{1}{3} (x[n-1] + x[n] + x[n+1])$$



System Property:

Output of the system depends on ONLY

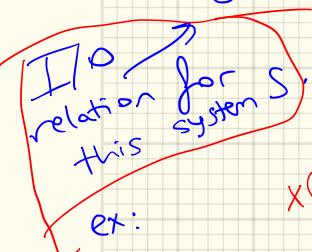
Causal System (Def): the current & past values of the input.

eg. $y[1]$: output at time $n=1$ \rightarrow depends on $x[n]$ at $n=0$ ✓
 $n=1$ ✓
 $n=2$ ✗

\therefore This past example (centered averger) is NOT a causal system.

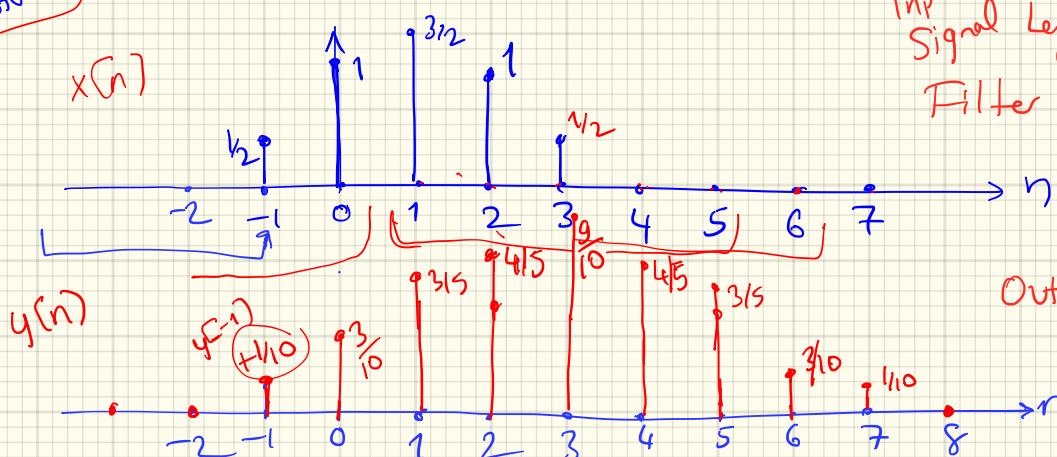
 Causality is important for real-time ("online") implementation of a system.

$$\text{Ex: } y[n] = \frac{1}{5} (x[n] + x[n-1] + x[n-2] + x[n-3] + x[n-4])$$


 If D relation for system S ,
 this
 ex:



Input Signal Length = $5 = N$
 Filter: $5 = L$



5-pt
 Running
 Average
 Filter:

Exercise
 fill in all $y[n]$

$$\begin{aligned}
 y[4] &= \frac{1}{5} [x[4] + x[3] + x[2] + x[1] + x[0]] \\
 &= \frac{1}{5} [0 + \frac{1}{2} + 1 + \frac{3}{2} + 1] = \frac{4}{5}
 \end{aligned}$$

In general RAF (Running Average Filter) : $(m+1)p +$.

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

\downarrow
 b_k : filter weights

eg. 3 pt RAF

$$y[n] = \sum_{k=0}^2 \frac{1}{3} x[n-k]$$

Generalize to :FIR Filter (Finite impulse response)

$$y[n] = \sum_{k=0}^m b_k x[n-k]$$

b_k : filter coefficients



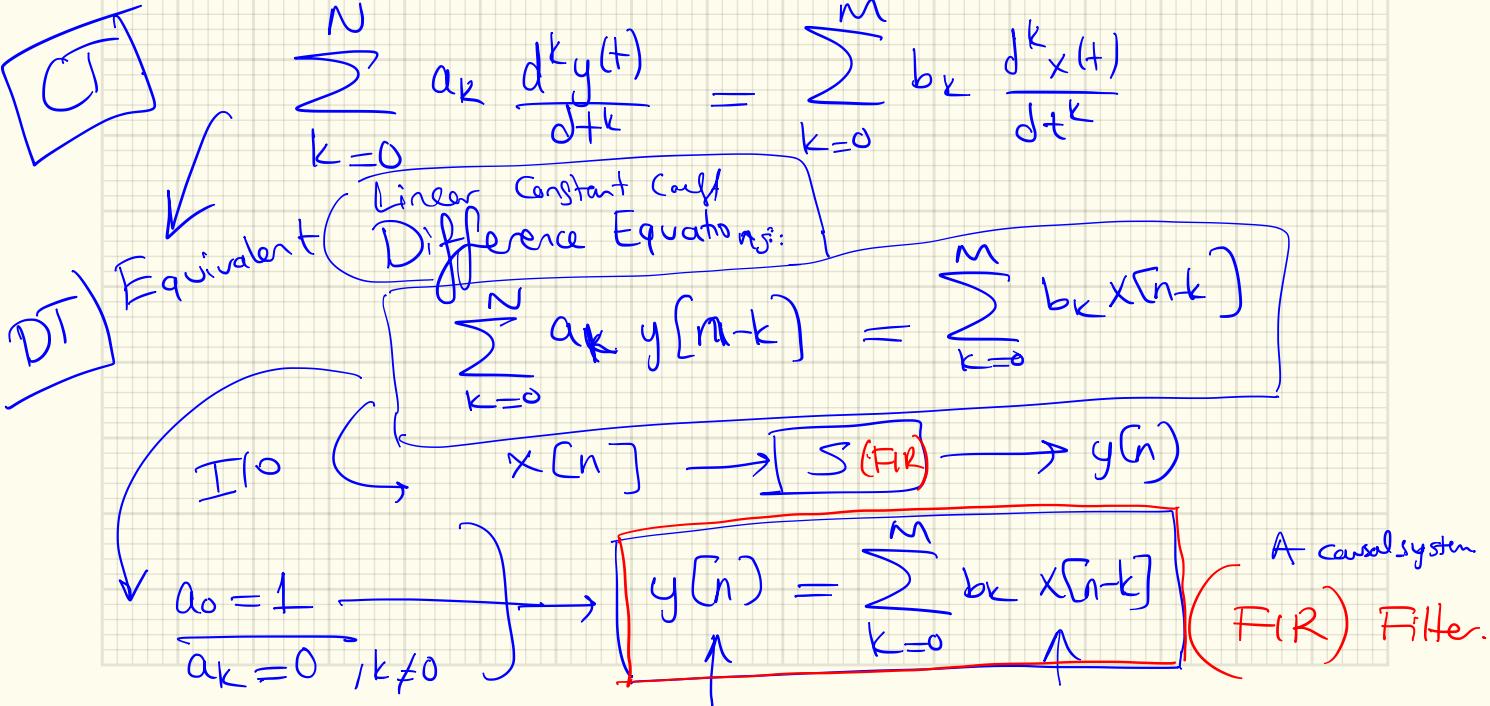
5pt RAF

$$y[n] = \sum_{k=0}^4 \frac{1}{5} x[n-k]$$

→ Linear Constant Coefficient Differential Equations

e.g.

$$\frac{d^4 y(t)}{dt^4} + a_3 \frac{d^3 y(t)}{dt^3} + a_2 \frac{d^2 y(t)}{dt^2} + a_1 \frac{dy(t)}{dt} + a_0 y_0(t) = \sum_{k=0}^m b_k \frac{d^k x(t)}{dt^k}$$



$$\rightarrow y[n] = b_0 \times [n] + b_1 \times [n-1] + \dots + b_m \times [n-m] \rightarrow \{b_k\}_{k=0}^m$$

$x[n] \xrightarrow{\text{FIR}} y[n]$

filter coeff.

Ex: $b_k = \{+3, -1, 2, 1\}$

$\begin{matrix} \uparrow & \uparrow & \uparrow & \uparrow \\ b_0 & b_1 & b_2 & b_3 \end{matrix}$

Q1. I/O relation of this system?

Q2. Causal system or not?

A1: $y[n] = \sum_{k=0}^3 b_k \times [n-k] = 3x[n] - x[n-1] + 2x[n-2] + x[n-3]$

A2: Causal system

Filter Length = $M+1 = L = 4$

Filter order = M : 3rd-order filter.

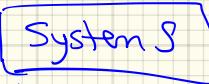
Given $x[n]$ of length N (N -pt signal)

Q: Length of $y[n]$: $N+L-1$. ✓

System Properties :

Causality ✓

① Linearity: Satisfies the property of superposition

i. Sum of Inputs \rightarrow  \rightarrow Sum of outputs

ii. Scale the input \rightarrow  \rightarrow Doubled output.
e.g. double the input
(Scaled)

$$a \cdot x_1[n] + b \cdot x_2[n] \rightarrow \boxed{S} \rightarrow a \cdot y_1[n] + b \cdot y_2[n]$$

$$\left. \begin{array}{l} a \cdot x_1[n] \rightarrow \boxed{S} \rightarrow a \cdot y_1[n] \\ b \cdot x_2[n] \rightarrow \boxed{S} \rightarrow b \cdot y_2[n] \end{array} \right\} \text{equal.}$$

$$a \cdot y_1[n] + b \cdot y_2[n]$$

Testing for Linearity of a System S :

I. $x_1[n] \rightarrow \boxed{S} \rightarrow y_1[n]$ $\xrightarrow{\times a} a \cdot y_1[n]$
 $x_2[n] \rightarrow \boxed{S} \rightarrow y_2[n]$ $\xrightarrow{\times b} b \cdot y_2[n]$

$$y'[n] = a \cdot y_1[n] + b \cdot y_2[n]$$

$y[n] = y'[n]$?

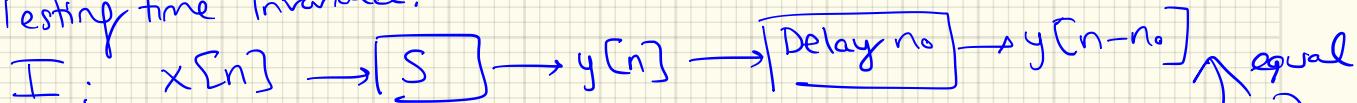
II. $x_1[n] \xrightarrow{\times a} a \cdot x_1[n]$ $x_2[n] \xrightarrow{\times b} b \cdot x_2[n]$

$$a \cdot x_1[n] + b \cdot x_2[n] \rightarrow \boxed{S} \rightarrow y[n]$$

If yes $\rightarrow S$ is Linear

② Time Invariance of a System : System behavior does not change w/ time.

Testing time invariance:



$| \Rightarrow z[n] == y[n-n_0]$ If yes, S is Time Inv.

No S is Not Time Inv.

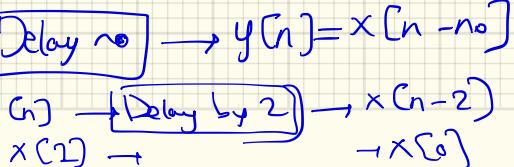
Def(LTI) Linear Time Invariant Systems :

A system that is both Linear & Time Invariant

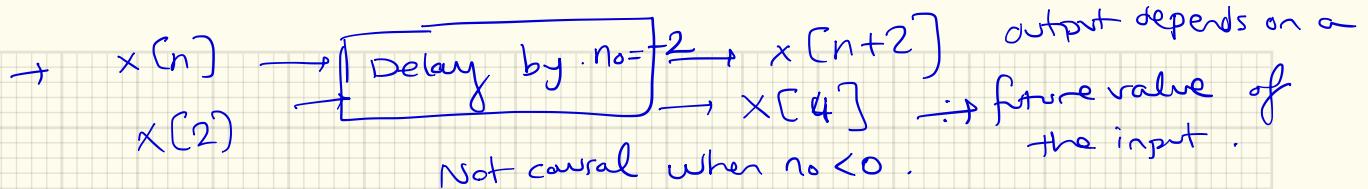
is an LTI system.

Note: (Delay system): $x(n) \rightarrow [Delay n_0] \rightarrow y(n) = x[n-n_0]$

Def Causal?



Causal $n_0 > 0$
not slw.



Ex: $y[n] = (x[n])^2$ Is this an LTI system?

(1) Time Inv? I. $x[n] \rightarrow [S] \rightarrow (x[n])^2 = y[n] \rightarrow [Delay] \rightarrow (x[n-n_0])^2$

II. $x[n] \rightarrow [Delay] \rightarrow x[n-n_0] \rightarrow [S] \rightarrow w[n]$

Yes.
Time Inv

(2) Linear?

I. $x_1[n] \rightarrow [S] \rightarrow (x_1[n])^2 \otimes a \rightarrow a y_1(n) + b y_2(n)$
 $x_2[n] \rightarrow [S] \rightarrow (x_2[n])^2 \otimes b \rightarrow b y_1(n) + a y_2(n) = a(x_1[n])^2 + b(x_2[n])^2$

II. $\underline{ax_1[n] + bx_2[n]} \rightarrow [S] \rightarrow (ax_1[n] + bx_2[n])^2 = ax_1^2[n] + bx_2^2[n] + 2abx_1[n]x_2[n]$

\Rightarrow Not a Linear System : \therefore Not an LTI system.

Ex: $y[n] = x[-n]$ System's I/O relation is given.
 $x[n] \rightarrow [S] \rightarrow y[n]$ Q. Is this an LTI system?

Linearity? I. $x_1[n] \rightarrow [S] \rightarrow y_1[n] = x_1[-n]$

I. $x_2[n] \rightarrow [S] \rightarrow y_2[n] = x_2[-n]$

$\downarrow a$ \oplus $\downarrow b$

$a x_1[-n] + b x_2[-n]$

$\uparrow =$

II. $a x_1[n] + b x_2[n] \rightarrow [S] \rightarrow a x_1[-n] + b x_2[-n]$ ✓

Yes, Linear System.

Time Inv? I. $x[n] \rightarrow [S] \rightarrow x[-n] \rightarrow [\text{Delay } n_0] \rightarrow z[n] = x[-(n-n_0)] = x[-n+n_0]$

II. $x[n] \rightarrow [\text{Delay } n_0] \rightarrow x[n-n_0] \rightarrow [S] \rightarrow y[n] = x[-n-n_0]$

Not a Time-Invariant System

↑
system does a time-axis scaling by -1!

∴ S is not an LTI system.