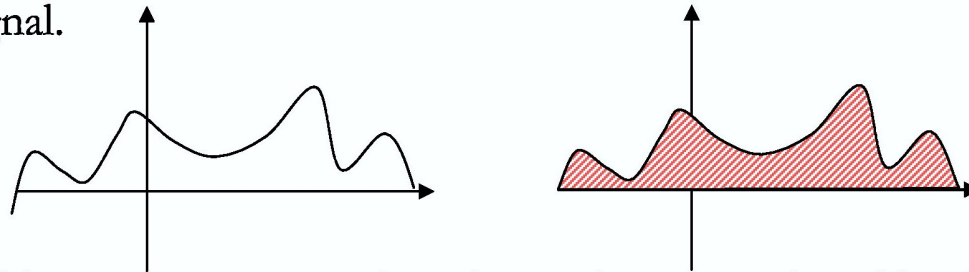


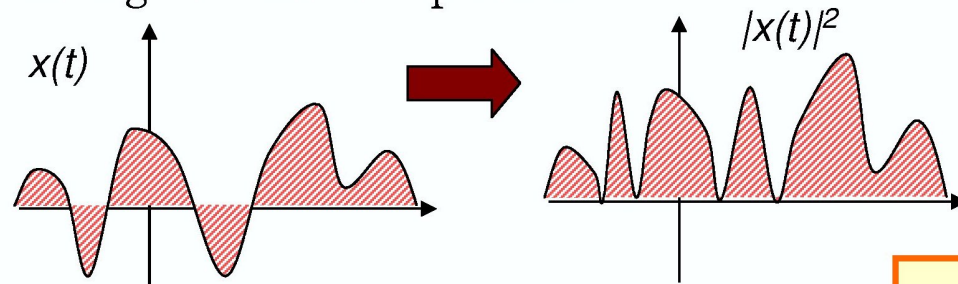
ENERGY & POWER IN SIGNALS

➤ It is often useful to define the “size” or “strength” of a signal. That is, we would like to be able to use a single number that represents the average strength of the signal. How would we do that?

↪ A reasonable answer would be to use the area under the curve. The larger the area, the stronger the signal.

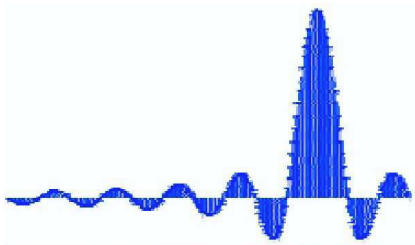


↪ But if the signal has negative areas, then the total area is reduced by the negative parts. Yet, a negative signal is not necessarily a weaker signal. In fact, -110 V will jolt you as much as +110 V will. So, we need another approach. Calculating the area under the “square of the absolute value of the signal” solves this problem



↪ This area is defined as the energy of the (continuous time) signal.

$$E_x = \sum_{n=-\infty}^{\infty} (x[n])^2$$



ENERGY & POWER IN SEQUENCES

⇒ Total **energy** of a (discrete) sequence $x[n]$ is similarly defined as

$$\mathcal{E}_x = \sum_{n=-\infty}^{\infty} x[n] \cdot x^*[n] = \sum_{n=-\infty}^{\infty} |x[n]|^2$$

⇒ An infinite length sequence with finite sample values may or may not have finite energy (see example 2.6, page 54, Mitra)

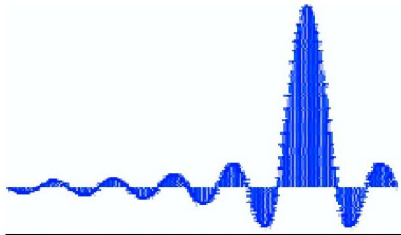
⇒ A finite length sequence with finite sample values always has finite energy

⇒ We define the energy of a sequence $x[n]$ over a finite interval $-K < n < K$ as

$$\mathcal{E}_{x,K} = \sum_{n=-K}^K |x[n]|^2$$

⇒ However, if the signal is infinitely long, say, such as the 60Hz mains, the energy becomes infinite. Hence we define the **average power** of an **aperiodic** sequence as the “energy per unit time”

$$P_x = \lim_{K \rightarrow \infty} \frac{1}{2K+1} \sum_{n=-K}^K |x[n]|^2$$



ENERGY & POWER IN SEQUENCES

⇒ Then

$$P_x = \lim_{K \rightarrow \infty} \frac{1}{2K+1} \sum_{n=-K}^K |x[n]|^2$$



$$P_x = \lim_{K \rightarrow \infty} \frac{1}{2K+1} \mathcal{E}_{x,K}$$

⇒ The average power of a *periodic* sequence $\tilde{x}[n]$ with a period N is given by

$$P_x = \frac{1}{N} \sum_{n=0}^{N-1} (\tilde{x}[n])^2$$

- ↪ The average power of an infinite-length sequence may be finite or infinite
- ↪ A signal with finite energy is called an **energy signal**. Energy signals has zero power!
- ↪ A signal with finite (and nonzero) power is called a **power signal**. A power signal, with nonzero power, has infinite energy.

Examples of Finite and Infinite Energy Signals

Infinite length sequence: $x_1[n] = \begin{cases} \frac{1}{n} & n \geq 1 \\ 0 & n \leq 0 \end{cases}$

Energy is: $\mathcal{E}_{x_1} = \sum_{n=1}^{\infty} \left(\frac{1}{n}\right)^2 = \frac{\pi^2}{6} \Rightarrow \text{finite}$

Infinite length sequence: $x_2[n] = \begin{cases} \frac{1}{\sqrt{n}} & n \geq 1 \\ 0 & n \leq 0 \end{cases}$

Energy is: $\mathcal{E}_{x_2} = \sum_{n=1}^{\infty} \left(\frac{1}{n}\right) \Rightarrow \text{infinite}$

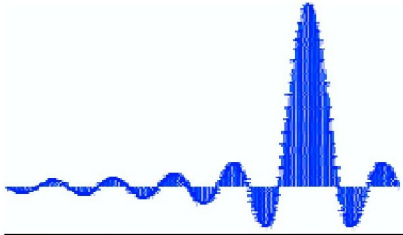
Example of a Power Signal

Infinite length sequence: $x[n] = \begin{cases} 3(-1)^n & n \geq 0 \\ 0 & n < 0 \end{cases}$

Energy : $\mathcal{E}_x = \sum_{n=0}^{\infty} 3^2 (-1)^{2n} = \sum_{n=0}^{\infty} 9(1)^n \Rightarrow \textit{infinite}$

Average Power : $P_x = \lim_{K \rightarrow \infty} \frac{1}{2K+1} 9 \sum_{k=0}^K 1^k = \lim_{K \rightarrow \infty} \frac{9(K+1)}{2K+1} = 4.5$

$\Rightarrow \textit{finite}$

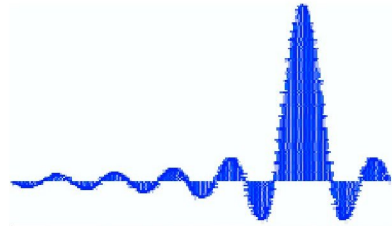


ENERGY & POWER SEQUENCES

Recap:

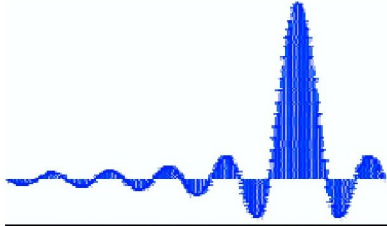
- A sequence with finite average power is called a *power signal*. Unless the power is zero, a power signal has infinite energy
 - ↳ A periodic sequence has a finite average power but infinite energy

- A sequence with finite energy is called an *energy signal*. An energy signal has zero average power.
 - ↳ A finite-length sequence has finite energy but zero average power



Discrete Time Systems

- ➔ Basic Operations on Discrete Sequences
- ➔ Discrete Systems
- ➔ Classification of Discrete Systems
 - ↪ Linearity
 - ↪ Shift-invariance
 - ↪ Causality
 - ↪ Memory
 - ↪ Stability
- ➔ Characterization of Discrete Systems
 - ↪ Impulse Response
 - ↪ Constant coefficient linear difference equations (CCLDE)
 - ↪ FIR systems
 - ↪ IIR systems



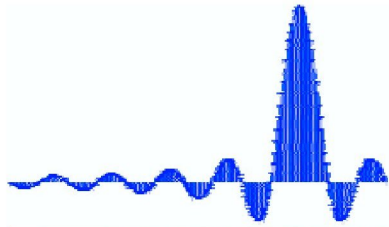
DISCRETE SYSTEMS

- ⇒ A discrete-time system processes a given input sequence $x[n]$ to generates an output sequence $y[n]$ with more desirable properties
- ⇒ In most applications, the discrete-time system is a single-input, single-output system:



$$y[n] = \mathfrak{R}\{x[n]\}$$

↗
The operator that acts on the input sequence

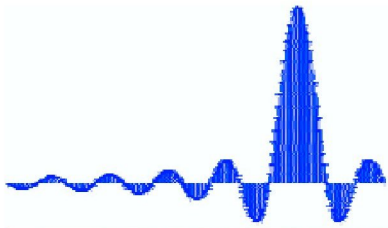


CLASSIFICATION OF SYSTEMS

➔ Discrete time systems can be classified based on their properties

- ↪ Discrete vs. continuous
- ↪ Linear vs. non-linear
- ↪ Shift-invariant or shift-variant
- ↪ Causal vs. noncausal
- ↪ Memoryless vs. with memory
- ↪ Stable vs. unstable

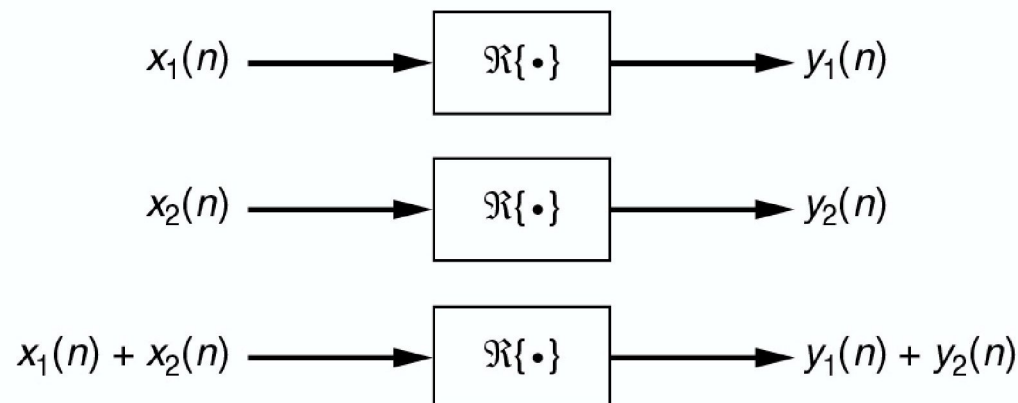
LINEARITY



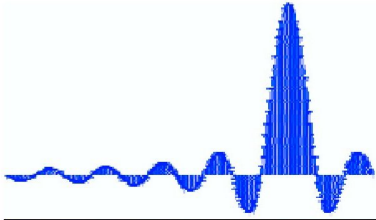
- ➔ Let $y_1[n]$ be the output due to an input $x_1[n]$ and $y_2[n]$ be the output due to an input $x_2[n]$. A system is said to be **linear**, if the following superposition & homogeneity properties are satisfied

$$x[n] = \alpha x_1[n] + \beta x_2[n] \iff y[n] = \alpha y_1[n] + \beta y_2[n]$$

$$\Re\{\alpha x_1[n] + \beta x_2[n]\} = \alpha \Re\{x_1[n]\} + \beta \Re\{x_2[n]\}$$



- ➔ This property must hold for any arbitrary constants α and β , and for all possible inputs $x_1[n]$ and $x_2[n]$, and can also be generalized to any arbitrary number of inputs



AN EXAMPLE - ACCUMULATOR

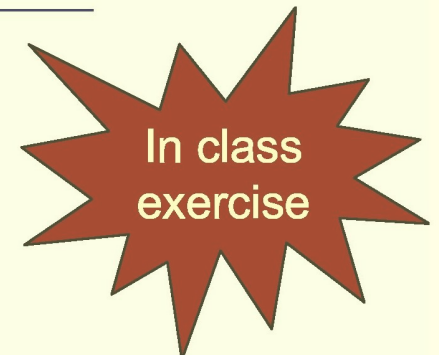
⇒ A discrete system whose input / output relationship is given as

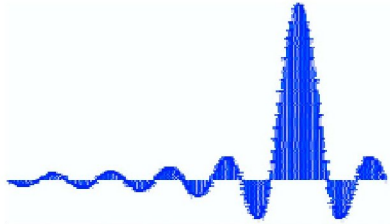
$$y[n] = \sum_{\ell=-\infty}^n x[\ell]$$
$$= y[-1] + \sum_{\ell=0}^n x[\ell],$$

The second form is used, if the signal is causal, in which case $y[-1]$ is the initial condition

is known as an **accumulator**. The output at any given time, is simply the sum of all inputs up to that time.

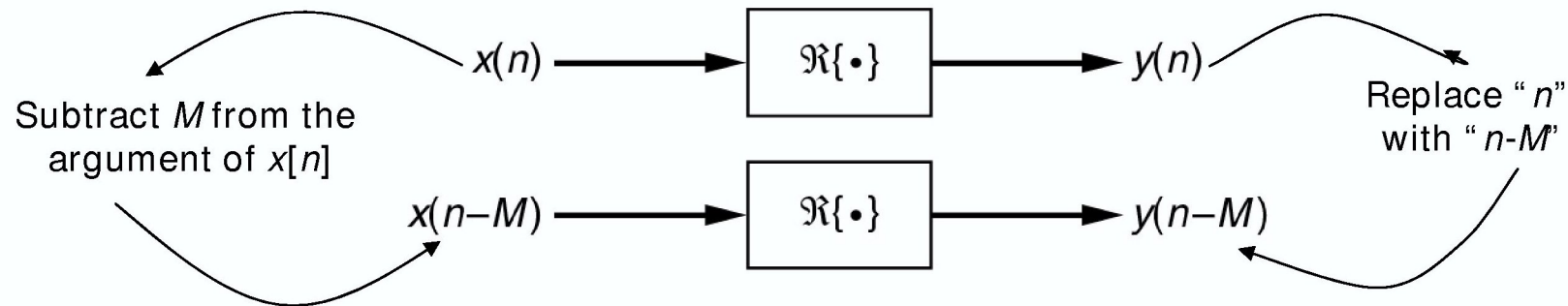
- ⇒ Is the accumulator linear? _____ ?
- ⇒ How about the accumulator for the causal systems? _____ ?
- ⇒ How about the system $y[n]=ax_1[n]+b$ _____ ?
- ⇒ As an exercise, try $y[n] = x^2[n] - x[n-1]x[n+1]$





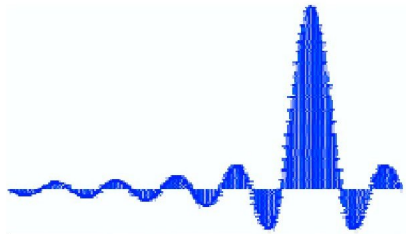
SHIFT - INVARIANCE

⇒ A system is said to be **shift-invariant** if $\mathfrak{R}(x[n-M])=y[n-M]$, for all n,m .



⇒ For sequences and systems where the index n is related to discrete instants of time, this property is also called the **time-invariance** property

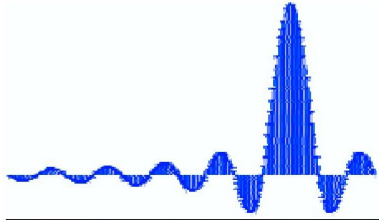
⇒ Time-invariance property ensures that for a specified input, the output is independent of the time the input is being applied



LINEAR TIME-INVARIANT SYSTEMS

- ➔ A system that satisfies both the linearity and the time (shift) invariance properties is called a ***linear time (shift) invariant*** system, LTI (LSI).

- ➔ We will see that these group of systems play a particularly important role in signal processing.
 - ↳ They are easy to characterize and analyze, and hence easy to design
 - ↳ Efficient algorithms have been developed over the years for such systems



CAUSALITY

- ⇒ A system is said to be **causal**, if the output at time n_0 does not depend on the inputs that come after n_0 .
- ⇒ In other words, in a causal system, the n_0^{th} output sample $y[n_0]$ depends only on input samples $x[n]$ for $n \leq n_0$ and does not depend on input samples for $n > n_0$.
- ⇒ Here are some examples: Which systems are causal?

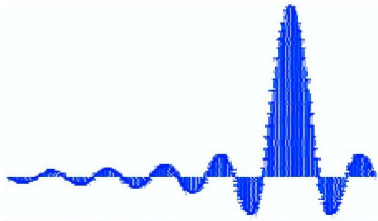
$$y[n] = \alpha_1 x[n] + \alpha_2 x[n-1] + \alpha_3 x[n-2] + \alpha_4 x[n-3]$$

$$y[n] = b_0 x[n] + b_1 x[n-1] + b_2 x[n-2] \\ + a_1 y[n-1] + a_2 y[n-2]$$

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

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

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



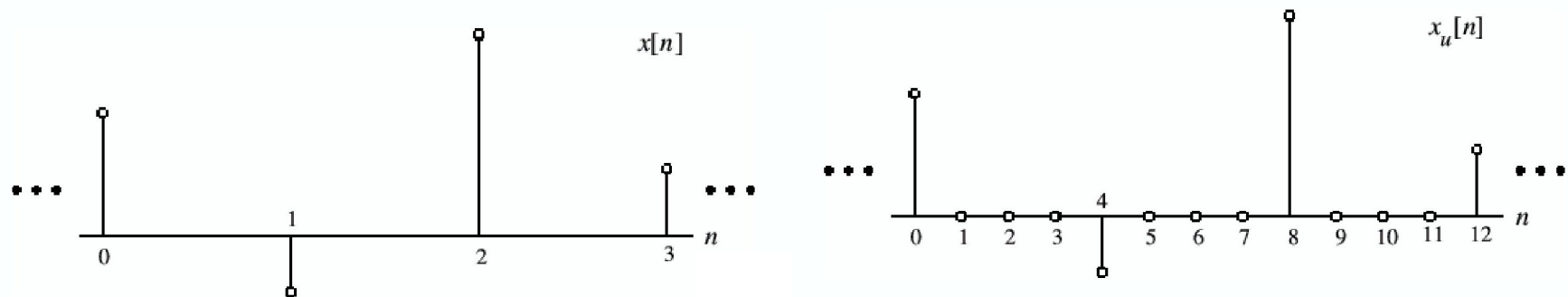
AN EXAMPLE - UPSAMPLER

⇒ A system whose input / output characteristics can be written as

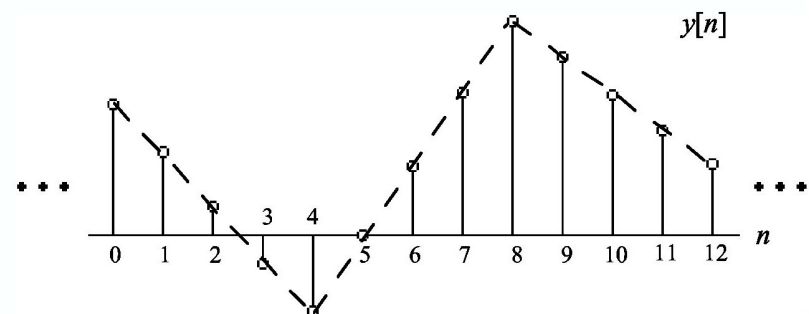
$$x_u[n] = \begin{cases} x[n/L], & n = 0, \pm L, \pm 2L, \dots \\ 0, & \text{otherwise} \end{cases}$$

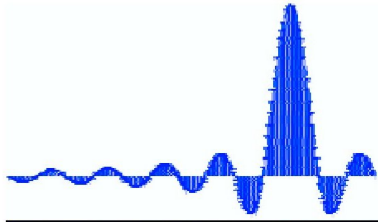
is known as an upsampler.

⇒ This system inserts L zeros between every sample. If the samples are inserted based on their amplitudes, then the system is called an interpolator.



Is the upsampler a time-invariant system?





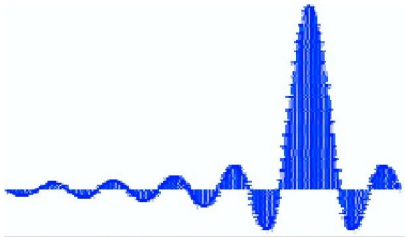
AN EXAMPLE - DOWNSAMPLER

- ⇒ A system whose input-output characteristic satisfies $y[n] = x[Mn]$ where M is a (+) integer, is called a ***downsampler*** or a ***decimator***.
- ↳ Such a system reduces the number of samples by a factor of M by removing M samples from between every sample.

⇒ Is this system

- ↳ Linear?
- ↳ Time (shift) invariant?
- ↳ Causal?





MEMORY

⇒ A system is said to be *memoryless* if the output depends only on the current input, but not on any other past (or future) inputs. Otherwise, the system is said to have memory.

⇒ Which of the following systems have memory?



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

$$y[n] = x[Mn]$$

$$x_u[n] = \begin{cases} x[n/L], & n = 0, \pm L, \pm 2L, \dots \\ 0, & \text{otherwise} \end{cases}$$

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

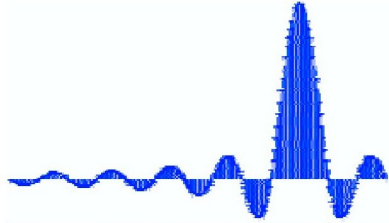


STABILITY

- ➔ There are several definitions of stability, which is of utmost importance in filter design. We will use the definition of stability in the **BIBO** sense

- ➔ A system is said to be stable in the **bounded input bounded output** sense, if the system produces a finite (bounded) output for any finite (bounded) input, that is,
 - ↳ If $y[n]$ is the response to an input $x[n]$ that satisfies $|x[n]| \leq B_x < \infty$, and $y[n]$ satisfies $|y[n]| \leq B_y < \infty$, then the system is said to be stable in the BIBO sense.

- ➔ A system (filter) that is not stable is rarely of any practical use (except for very specialized applications), and therefore, most filters are designed to be BIBO stable.



AN EXAMPLE – MOVING AVERAGE FILTER

⇒ An M – point ***moving average system*** (filter) is defined as

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

- ⇒ What would you use such a system for?
- ⇒ Is this system stable?

Matlab exercise

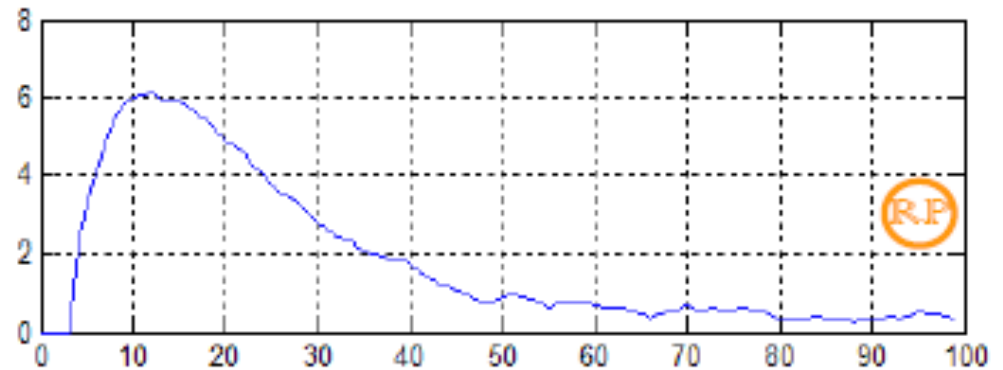
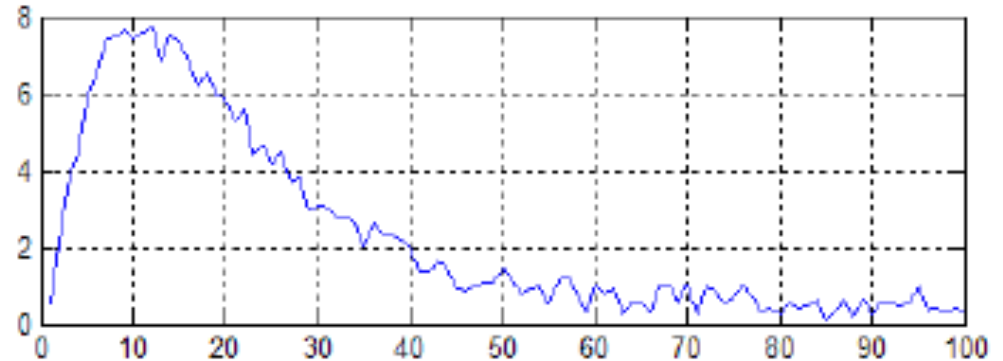
```

n=0:99;
s=2*(n.*(0.9).^n);
d=rand(1, 100);
x=s+d;
subplot(211)
plot(x); grid

for i=7:100;
y(i)=(1/7)*sum(x(i-1)+x(i-2)+x(i-3)+x(i-4)+x(i-5)+x(i-6));
end

subplot(212)
plot(n,y); grid
    
```

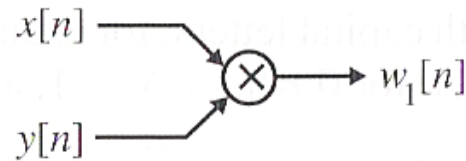
7 point moving average



Basic Operations in schematic form

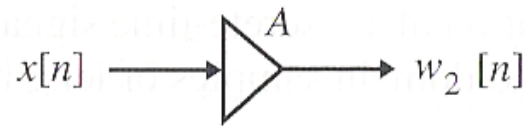
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Chapter 2: Discrete-Time Signals and Systems



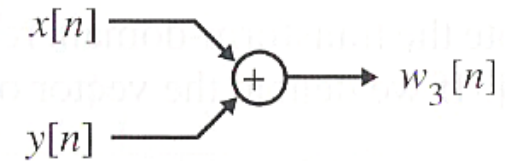
$$w_1[n] = x[n]y[n]$$

(a)



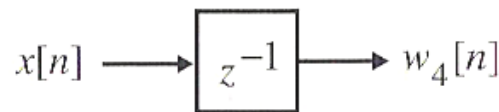
$$w_2[n] = Ax[n]$$

(b)



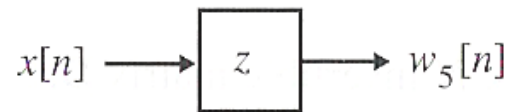
$$w_3[n] = x[n] + y[n]$$

(c)



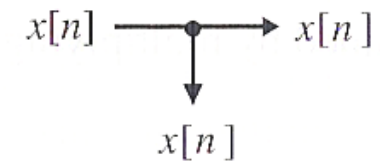
$$w_4[n] = x[n-1]$$

(d)



$$w_5[n] = x[n+1]$$

(e)



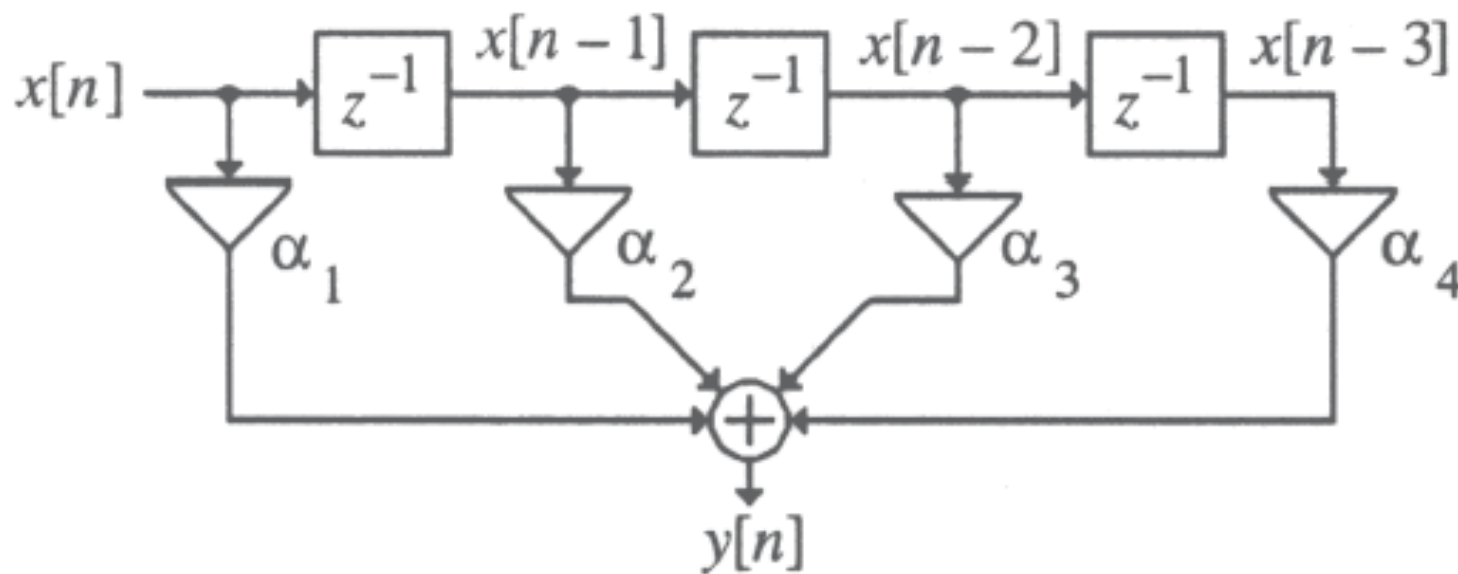
(f)

Figure 2.5: Schematic representations of basic operations on sequences: (a) multiplier, (b) multiplier, (c) adder, (d) unit delay, (e) unit advance, and (f) pick-off node.



BLOCK DIAGRAM IMPLEMENTATION

➔ What does this system do...?



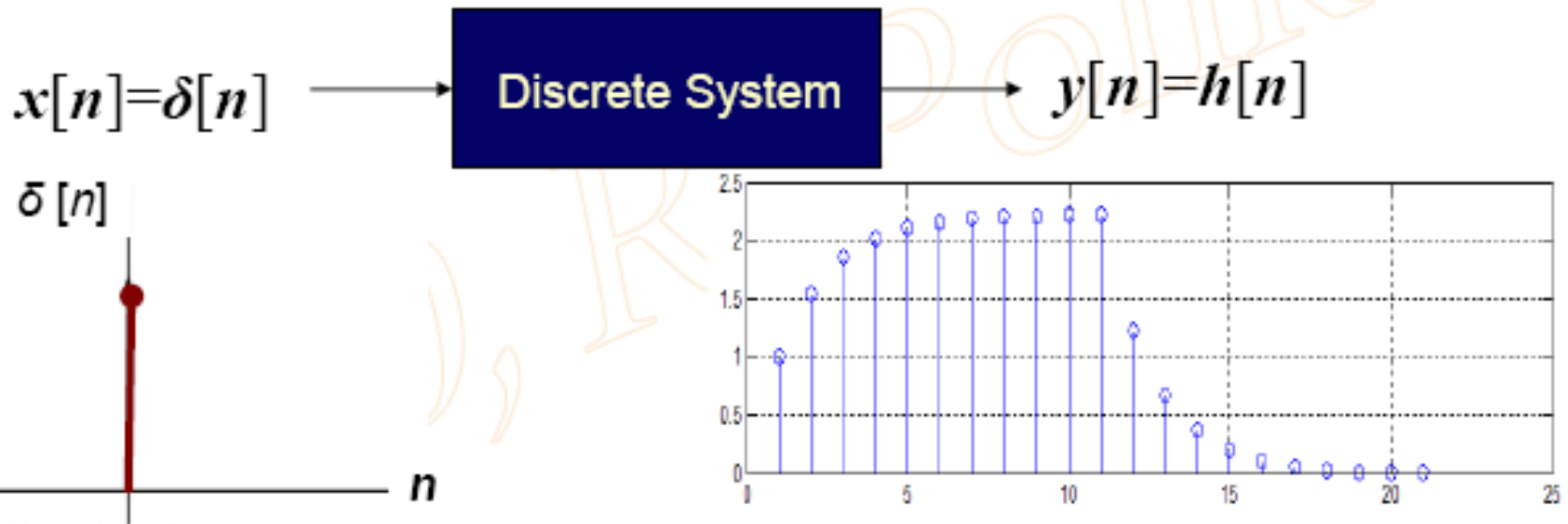
air

$y[n]=\dots$

$$\alpha_1 x[n] + \alpha_2 x[n-1] + \alpha_3 x[n-2] + \alpha_4 x[n-3]$$

IMPULSE RESPONSE

- ➔ The response of a discrete system to a unit impulse sequence $\delta[n]$ is called the *impulse response* of the system, and it is typically denoted by $h[n]$



$$h[n] = h[n] * \delta[n] = \sum_{m=-\infty}^{\infty} h[m] \cdot \delta[n-m]$$

➔ Ex: Consider the following system

$$y[n] = \alpha_1 x[n] + \alpha_2 x[n-1] + \alpha_3 x[n-2] + \alpha_4 x[n-3]$$

↳ The impulse response of this system can be obtained by setting the input to the impulse sequence ➔

$$h[n] = \alpha_1 \delta[n] + \alpha_2 \delta[n-1] + \alpha_3 \delta[n-2] + \alpha_4 \delta[n-3]$$



$$\begin{aligned} h[n] &= \sum_{m=-\infty}^{\infty} h[m] \cdot \delta[n-m] \\ &= \dots h[-1] \delta[n+1] + h[0] \delta[n] + h[1] \delta[n-1] + \dots \end{aligned}$$

$$\Rightarrow \{h[n]\} = \{\alpha_1, \alpha_2, \alpha_3, \alpha_4\}$$

↑

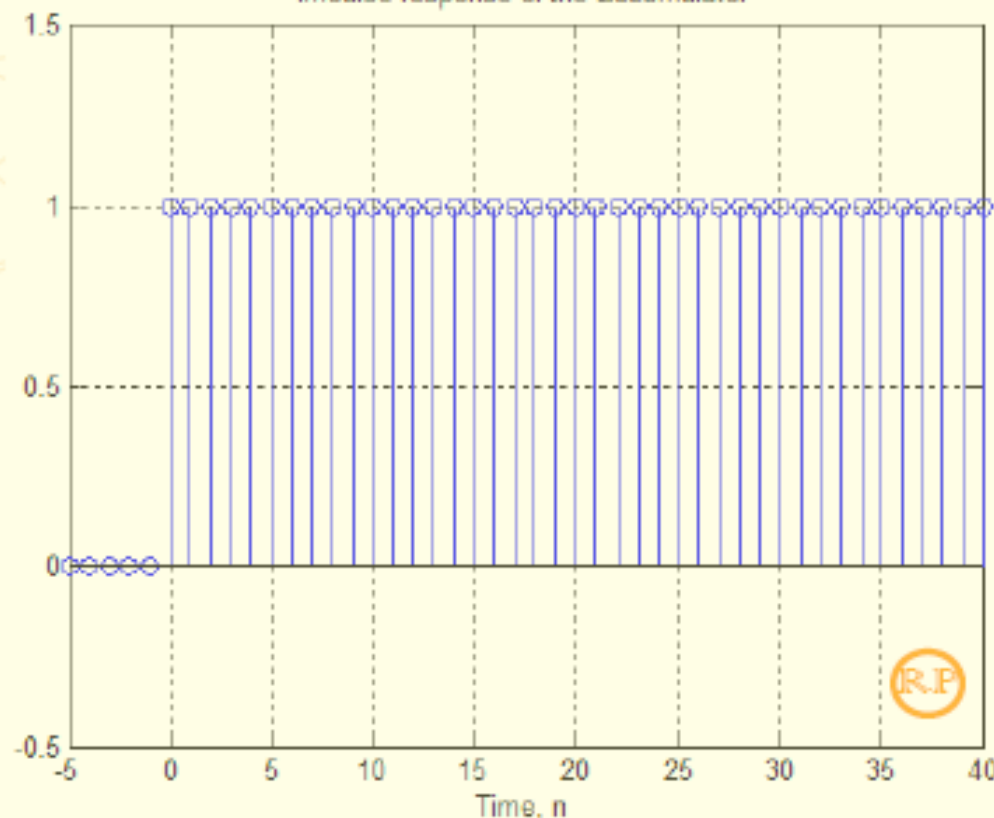
➔ Recall the discrete time accumulator:

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

↳ What is the impulse response of this system?

$h[n]=\dots$

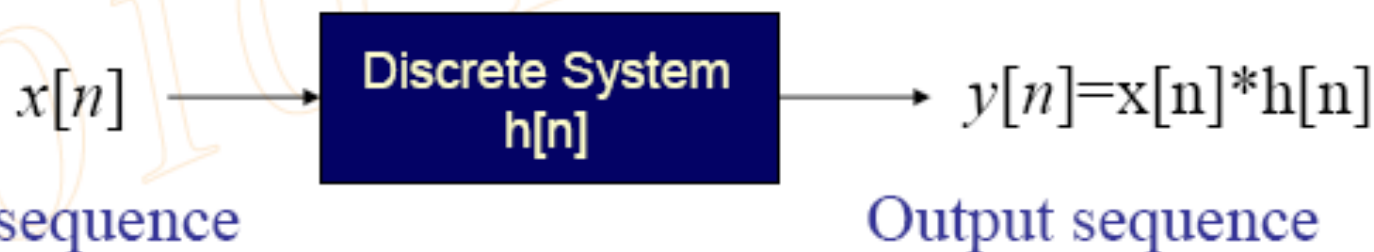
Impulse response of the accumulator





IMPULSE RESPONSE

- ⇒ So, what is the big deal?
- ⇒ The impulse response plays a *monumental* role in characterization of LTI systems.
 - ↳ In fact, if you know the impulse response of a discrete LTI system, then you know the response of the system to any arbitrary input!
 - ↳ You tell me $h[n]$, I will tell you the response to *any* $x[n]$



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



⇒ There are several useful properties of the convolution and impulse response characterization:

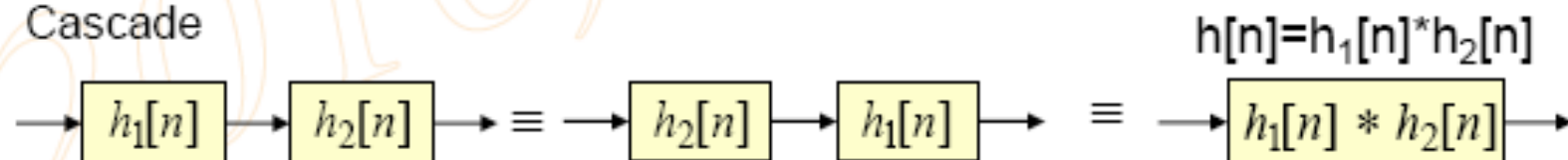
↳ An LTI system is BIBO stable, if its impulse response is absolutely summable:

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

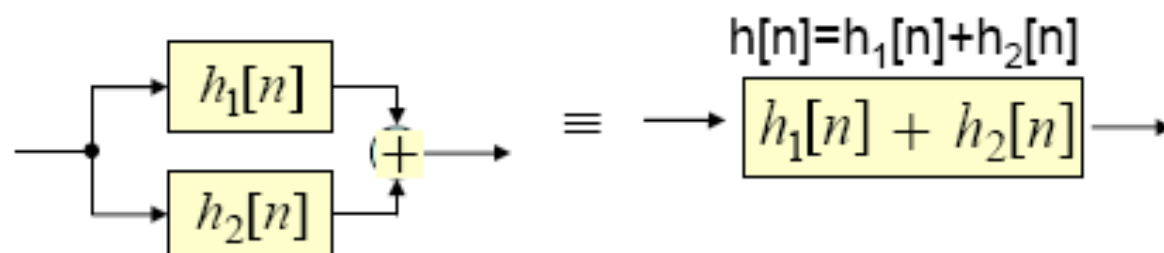
↳ An LTI system is causal, if its impulse response is a causal sequence, i.e.,
 $h[n]=0, n<0$

↳ If more than one system are connected by series or parallel, the effective system impulse response can be obtained as follows:

- Cascade



- Parallel





EXAMPLE

➔ The system, known as factor-of-2 interpolator is given as

$$h[n] = \delta[n] + \frac{1}{2}(\delta[n-1] + \delta[n+1])$$

that is, its impulse response is $h[n] = [0.5 \ 1 \ 0.5]$, $-1 < n < 1$.

➔ Is this system causal and stable?

↳ Not causal. Because the impulse response is not a causal sequence

↳ Stable, since summation of the absolute values of all $h[n]$ values is finite

➔ A causal version of this system can be obtained, however, simply by delaying the output by one sample.

$$h[n] = \delta[n-1] + \frac{1}{2}(\delta[n-2] + \delta[n])$$

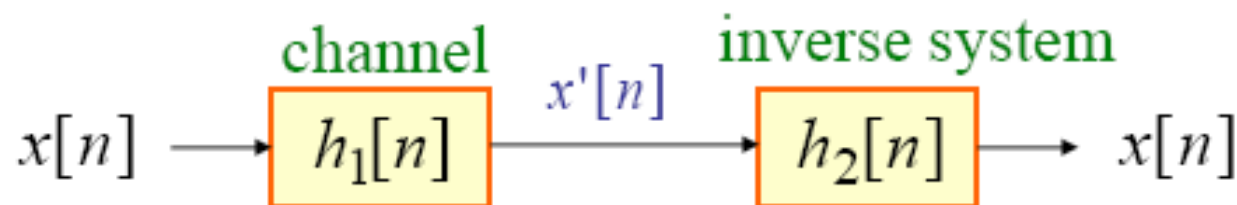


- ➔ A cascade connection of two stable systems is always stable
- ➔ If a cascade system satisfies the following condition

$$h_1[n] * h_2[n] = \delta[n]$$

then h_1 and h_2 are called **inverse systems**

- ➔ An application of the inverse system concept is in the recovery of a signal $x[n]$ from its distorted version $x'[n]$ appearing at the output of a transmission channel
- ➔ If the impulse response of the channel is known, then $x[n]$ can be recovered by designing an inverse system of the channel



$$h_1[n] * h_2[n] = \delta[n]$$





➔ Consider the discrete-time system where

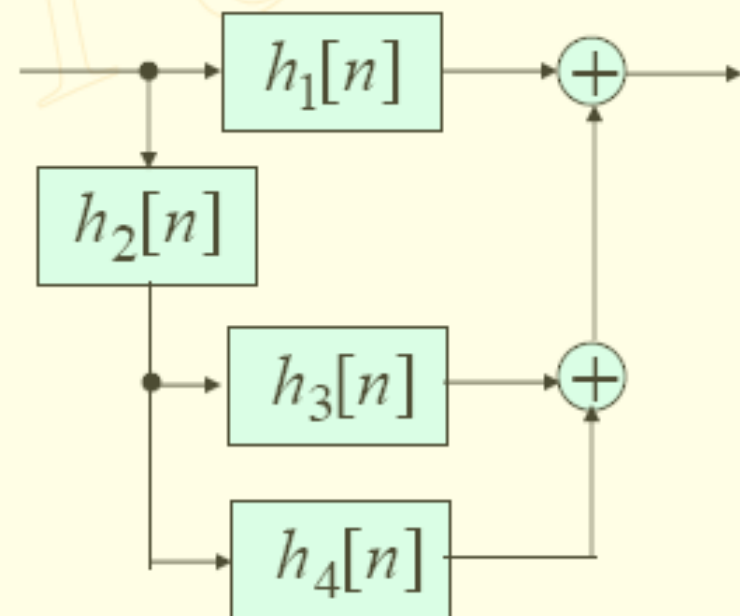
$$h_1[n] = \delta[n] + 0.5\delta[n-1],$$

$$h_2[n] = 0.5\delta[n] - 0.25\delta[n-1],$$

$$h_3[n] = 2\delta[n],$$

$$h_4[n] = -2(0.5)^n u[n]$$

What is the overall system response?





FINITE IMPULSE RESPONSE SYSTEMS

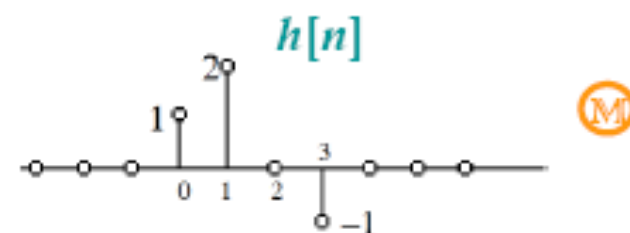
- If the impulse response $h[n]$ of a system is of finite length, that system is referred to as a *finite impulse response (FIR)* system

$$h[n] = 0 \text{ for } n < N_1 \text{ and } n > N_2, \quad N_1 < N_2$$

- ↳ The output of such a system can then be computed as a finite convolution sum

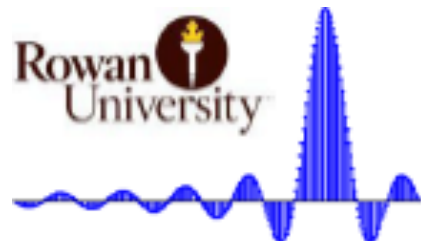
$$y[n] = \sum_{k=N_1}^{N_2} h[k]x[n-k]$$

- ↳ E.g., $h[n] = [1 \ 2 \ 0 \ -1]$ is a FIR system (filter)



- ↳ FIR systems are also called *nonrecursive systems* (for reasons that will later become obvious), where the output can be computed from the current and past input values only – without requiring the values of *previous outputs*.





INFINITE IMPULSE RESPONSE SYSTEMS

➔ If the impulse response is of infinite length, then the system is referred to as an ***infinite impulse response (IIR)*** system. These systems cannot be characterized by the convolution sum due to infinite sum.

↳ Instead, they are typically characterized by constant coefficient linear difference equations (CCLDEs), as we will see later.

↳ Recall accumulator and note that it can have an alternate – and more compact representation that makes the current output a function of previous inputs and outputs

$$y[n] = \sum_{\ell=-\infty}^n x[\ell] \quad \longrightarrow \quad y[n] = y[n-1] + x[n]$$

↳ The impulse response of this system (which is of infinite length), cannot be represented with a finite convolution sum. Note that, since the current output depends on the previous outputs, this is also called a ***recursive system***



CONSTANT COEFFICIENT LINEAR DIFFERENCE EQUATIONS

- ⇒ All discrete systems can also be represented using *constant coefficient, linear difference equations*, of the form

$$y[n] + a_1y[n-1] + a_2y[n-2] + \dots + a_Ny[n-N] = b_0x[n] + b_1x[n-1] + \dots + b_Mx[n-M]$$

Outputs $y[n]$ Inputs $x[n]$

$$\sum_{i=0}^N a_i y[n-k] = \sum_{i=0}^M b_i x[n-k]$$

Constant coefficients

- ↪ Constant coefficients a_i and b_i are called *filter coefficients*
- ↪ Integers M and N represent the maximum *delay* in the input and output, respectively. The larger of the two numbers is known as the *order of the filter*.
- ↪ Any LTI system can be represented as two finite sum of products!





- Note that the expression indicates the most general form of an LTI system:

$$\sum_{i=0}^N a_i y[n-i] = \sum_{j=0}^M b_j x[n-j], \quad a_0 = 1$$

- ↳ If the current output $y[n]$ does not depend on previous outputs $y[n-i]$, that is if all $a_i=0$ (except $a_0=1$), then we have *no recursion* – such systems are FIR (*non-recursive*) systems

$$y[n] = \sum_{j=0}^M b_j x[n-j]$$

- ↳ Note that the impulse response of an FIR system can easily be obtained from its CCLDE representation:

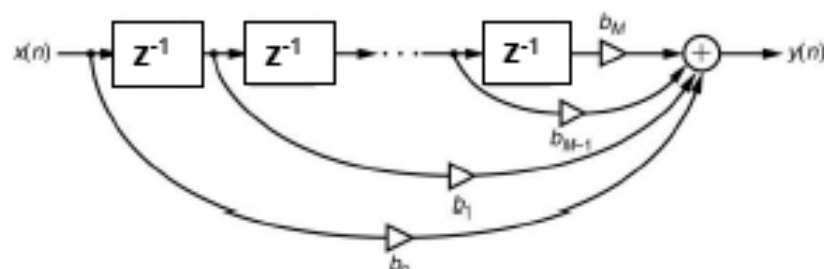
$$\begin{aligned} y[n] = \sum_{j=0}^M b_j x[n-j] &\Rightarrow h[n] = \sum_{j=0}^M b_j \delta[n-j] \\ &= b_0 \delta[n] + b_1 \delta[n-1] + \dots + b_M \delta[n-M] \end{aligned}$$

- ↳ The sum of finite numbers will always be finite, therefore, the impulse response of this system will be finite, hence, finite impulse response.
- ↳ Finite Impulse Response \leftrightarrow Nonrecursive



$$y[n] = \sum_{j=0}^M b_j x[n-j] = b_0 x[n] + b_1 x[n-1] + \dots + b_M x[n-M]$$

- Note that this representation looks similar to the definition of convolution. In fact, $y[n] = b_n * x[n]$, that is the system output of an FIR filter is simply the convolution of input $x[n]$ with the filter coefficients b_n
- Since we already know that the output of a system is the convolution of its input with the system impulse response, it follows that *filter coefficients b_n is the impulse response of an FIR filter!*
- The CCLDE representation of an FIR system can schematically be represented using the following diagram, known as the “*filter structure*”



(B)

The hardware implementation follows this structure exactly, using delay elements, adders and multipliers.

