

# Discrete-Time Systems

EE4015 Digital Signal Processing

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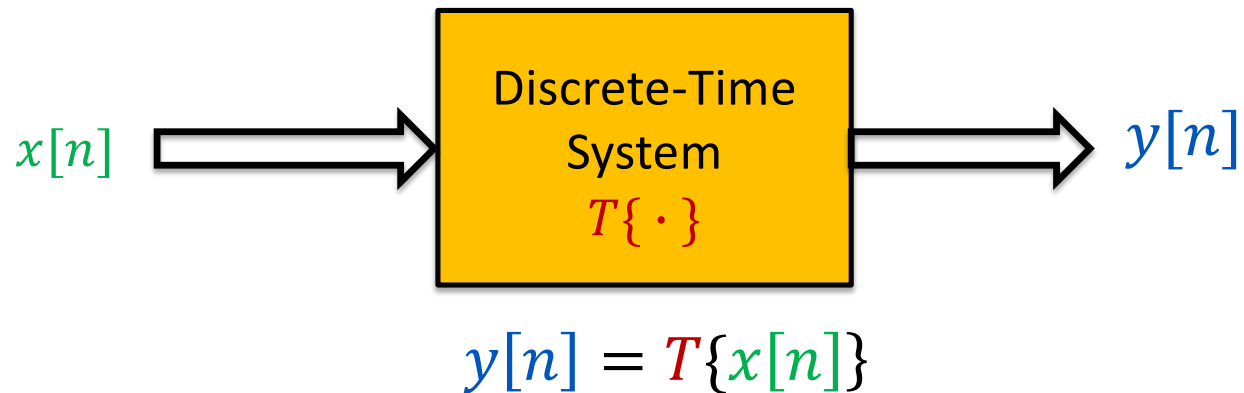
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# Content

- Definition of Discrete-Time (DT) Systems
- Classification of DT Systems
  - Memory, Time-Invariant, Linear, Causal and Stable
- Linear Time-Invariant (LTI) System
  - Unit sample Response (Impulse Response)
  - Convolution
- Block Diagram Representation
- Difference Equations
- Finite Impulse Response (FIR) System
- Infinite Impulse Response (IIR) System
- Stability of LTI System

# Discrete-Time Systems

- A discrete-time system is a device or algorithm that operates on a discrete-time signal (or sequence)  $x[n]$  called the input to produce another discrete-time signal called the output or response  $y[n]$ .



- where the symbol  $T$  denotes the **transformation** or processing performed by the system on  $x[n]$  to produce  $y[n]$

# Classification of Discrete-Time Systems

- In the system analysis, it is desirable to classify the systems according to their general properties.
- General Categories of DT Systems are:
  - **Memoryless Systems**
  - **Time-Invariant Systems**
  - **Linear Systems**
  - **Causal Systems**
  - **Stable Systems**

# Memoryless Systems

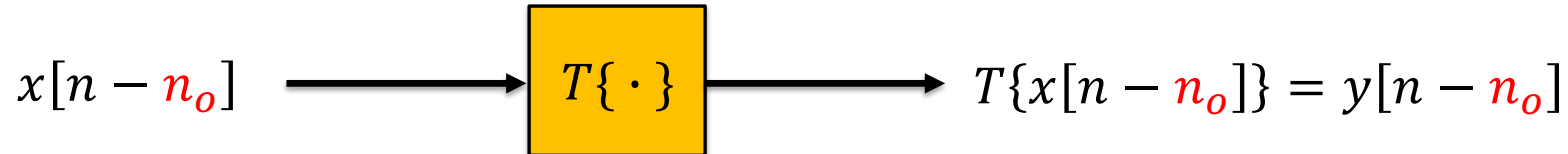
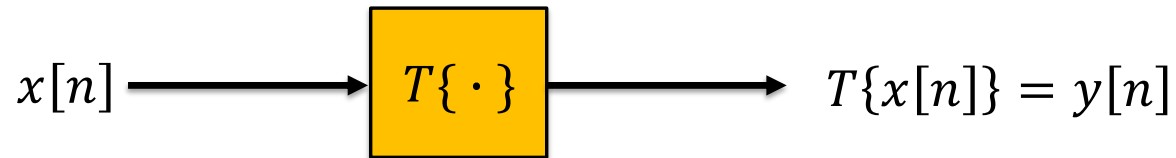
- A discrete-time system is called **memoryless** if its output at any instant  $n$  **depends at most on the input sample at the same time**, but not on past or future samples of the input.
- Example memoryless systems:
  - $y[n] = a x[n]$
  - $y[n] = n x[n] + b(x[n])^2$
- The output of these systems  **$y[n]$  are only depends on  $x[n]$**
- They are all memoryless systems

# Memory Systems

- On the other hand, the systems described by the following input/output relations, such as
  - $y[n] = 2x[n] + 3x[n - 1]$
  - $y[n] = \sum_{k=0}^N x[n - k]$
- These are systems with memory as their **outputs depend on previous input samples**.

# Time-Invariant Systems

- A time-invariant system is defined as follows:



- Specifically, a system is time invariant if **a time shift in the input signal results in an identical time shift in the output signal.**

# Time-Invariant System Example 1

- Determine if the system is time variant or time invariant.
  - $y[n] = T\{x[n]\} = |x[n]|$
- The response of this system to  $x[n - k]$  is
  - $T\{x[n - k]\} = |x[n - k]|$
- Now if we delay  $y[n]$  by  $k$  units in time, we obtain
  - $y[n - k] = |x[n - k]|$
- This system is **time-invariant**, since
  - $T\{x[n - k]\} = |x[n - k]| = y[n - k]$

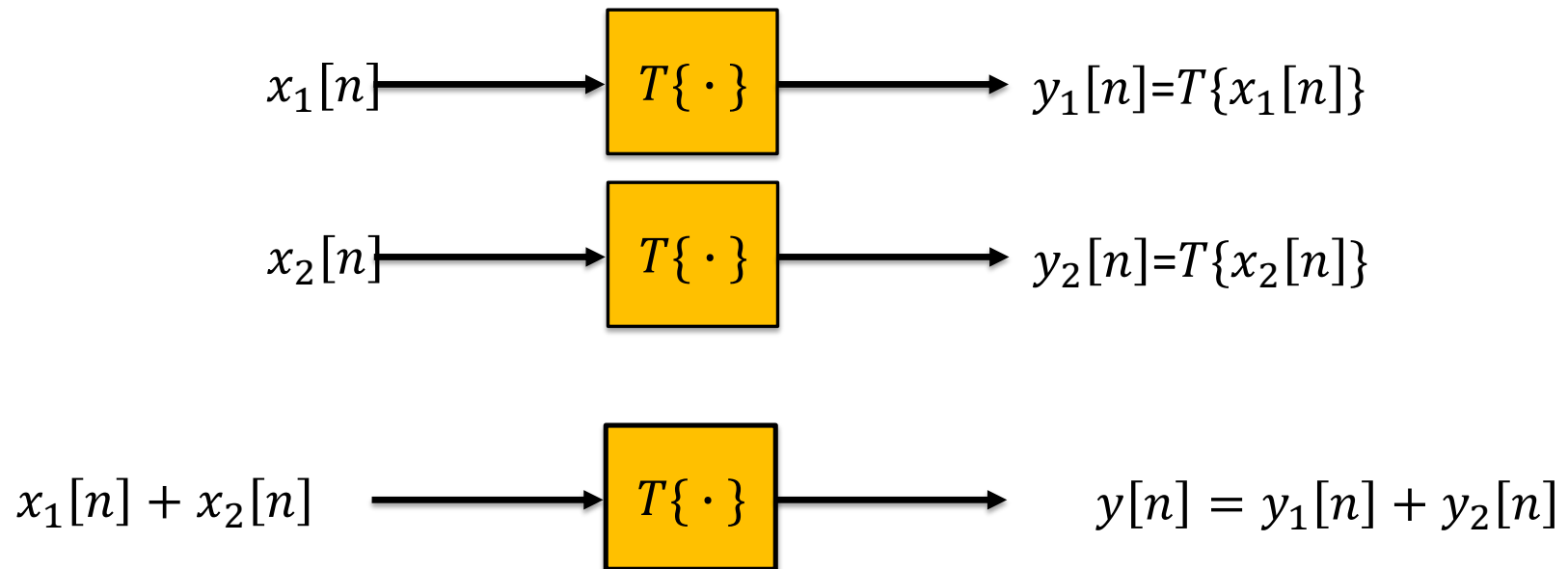


## Time-Invariant System Example 2

- Determine if the system is time variant or time invariant.
  - $y[n] = T\{x[n]\} = nx[n]$
- The response of this system to  $x[n - k]$  is
  - $T\{x[n - k]\} = nx[n - k]$
- Now if we delay  $y[n]$  by  $k$  units in time, we obtain
  - $y[n - k] = (n - k)x[n - k]$
- This system is **time-variant**, since
  - $T\{x[n - k]\} = nx[n - k] \neq y[n - k] = (n - k)x[n - k]$

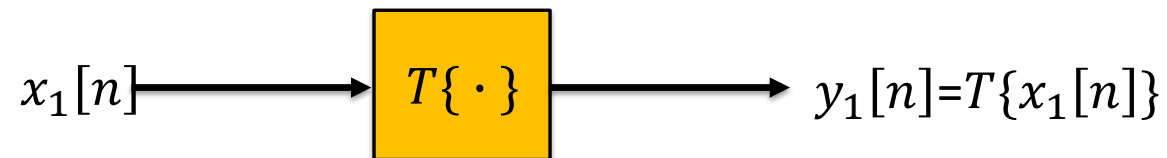
# Superposition Condition

- A superposition system is defined as follows:

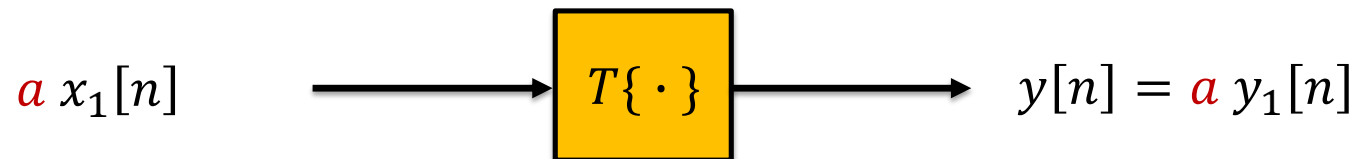


# Homogeneity Condition

- A homogeneity system is defined as follows:

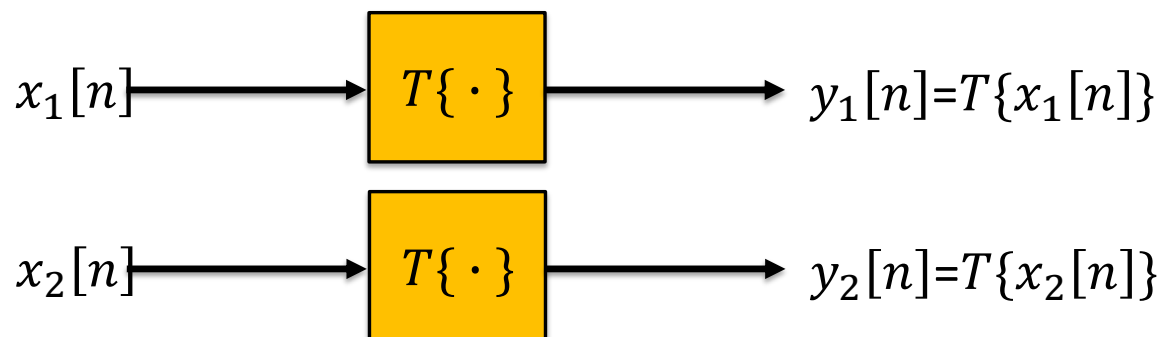


- For arbitrary constants  $a$ :

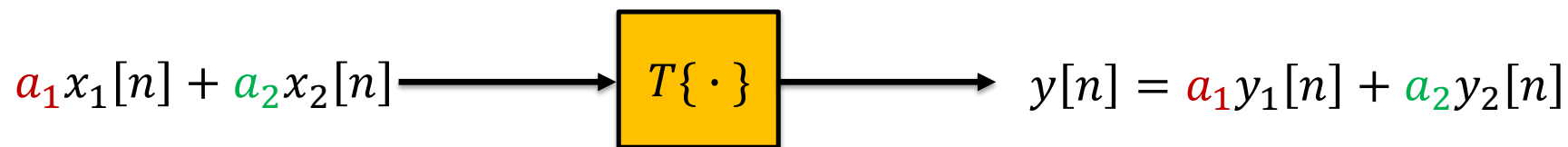


# Linear Systems : Superposition and Homogeneity

- Linear systems satisfy both superposition and homogeneous conditions:



- For arbitrary constants  $a_1$  and  $a_2$  :



# Linear System Example 1

- Determine the 3-sample average system is linear or Not.

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

- The response of this system to  $\{a_1x_1[n] + a_2x_2[n]\}$  is  $T\{a_1x_1[n] + a_2x_2[n]\}$

$$= \frac{1}{3}(a_1x_1[n] + a_2x_2[n] + a_1x_1[n - 1] + a_2x_2[n - 1] + a_1x_1[n - 2] + a_2x_2[n - 2])$$

$$= \frac{1}{3}(a_1x_1[n] + a_1x_1[n - 1] + a_1x_1[n - 2]) + \frac{1}{3}(a_2x_2[n] + a_2x_2[n - 1] + a_2x_2[n - 2])$$

$$= a_1y_1[n] + a_2y_2[n]$$

- The 3-sample average is a linear system

## Linear System Example 2

- Determine the squared input system is linear or Not.
  - $y[n] = T\{x[n]\} = (x[n])^2$
- Let  $y_1[n]=T\{x_1[n]\} = (x_1[n])^2$  and  $y_2[n]=T\{x_2[n]\} = (x_2[n])^2$
- The response of this system to  $\{a_1x_1[n] + a_2x_2[n]\}$  is  $T\{a_1x_1[n] + a_2x_2[n]\}$   
 $= (a_1x_1[n] + a_2x_2[n])^2 = (a_1x_1[n])^2 + (a_2x_2[n])^2 + 2a_1a_2x_1[n]x_2[n]$
- This is NOT equal to  $a_1y_1[n] + a_2y_2[n] = a_1(x_1[n])^2 + a_2(x_2[n])^2$
- This system is non-linear.

# Linear System Example 3

- Determine this system is linear or Not.

- $y[n] = T\{x[n]\} = n x[n]$

- Let  $y_1[n]=T\{x_1[n]\} = nx_1[n]$  and  $y_2[n]=T\{x_2[n]\} = nx_2[n]$

- The response of this system to  $\{a_1x_1[n] + a_2x_2[n]\}$  is

$$H\{a_1x_1[n] + a_2x_2[n]\} = n (a_1x_1[n] + a_2x_2[n]) = a_1nx_1[n] + a_2nx_2[n]$$

- This is equal to  $a_1y_1[n] + a_2y_2[n]$

- This system is linear.

# Causal Systems

- A system is said to be **causal** if the output of the system at any time ' $n$ ' **depends only on present and past inputs** but does not depend on future inputs.
- If a system does not satisfy this definition, it is called **noncausal**.
  - The noncausal systems have outputs that **depend not only on present and past inputs but also on future inputs**.



# Causal and Noncausal System Examples

- **Causal System Examples**

- $y[n] = x[n] + 3x[n - 1]$
- $y[n] = 2x[n]$

- **Noncausal System Examples**

- $y[n] = x[n] + 3x[n + 2]$
- $y[n] = x[-n]$ 
  - Let  $n = -1 \Rightarrow y[-1] = x[1]$ , the output at  $n = -1$  depends on the input at  $n = 1$ .

# Causal System Exercise

- **Determine whether the following systems are causal or not.**
  1.  $y[n] = 0.5x[n] + 2.5x[n - 2]$ , for  $n \geq 0$
  2.  $y[n] = 0.25x[n - 1] + 2.5x[n + 2] - 0.4y[n - 1]$ , for  $n \geq 0$

# Causal System Exercise

- **Determine whether the following systems are causal or not.**
  1.  $y[n] = 0.5x[n] + 2.5x[n - 2]$ , for  $n \geq 0$
  2.  $y[n] = 0.25x[n - 1] + 2.5x[n + 2] - 0.4y[n - 1]$ , for  $n \geq 0$

## Solution

1) Causal

2) Non-causal

# Stable Systems : BIBO Stable

- A discrete signal  $x[n]$  is **bounded** if there exists a finite  $B_x$  such that  $|x[n]| < B_x$  for all  $n$ .
- A discrete-time system is **Bounded Input-Bounded Output (BIBO)** stable if every bounded input sequence  $x[n]$  produced a bounded output sequence.
  - If  $\underbrace{x[n]_{\max} \leq B_x}_{\text{Bounded Input}}$ , then  $\underbrace{y[n]_{\max} \leq B_y}_{\text{Bounded Output}}$

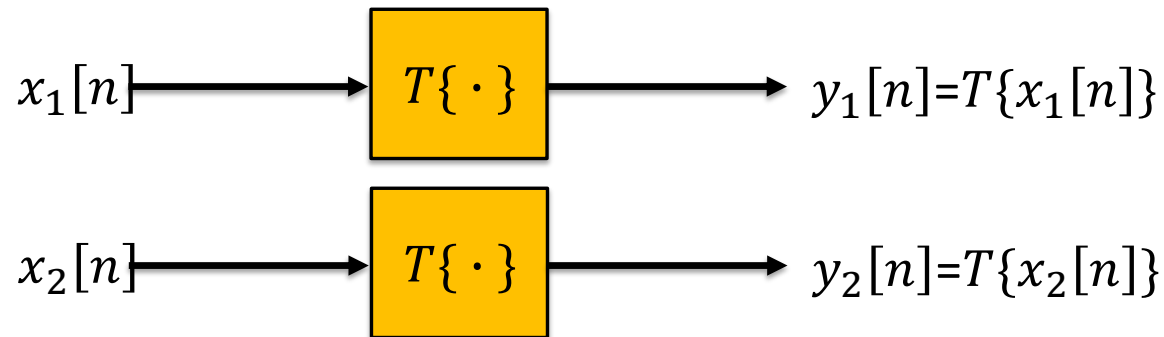
# Stable System Example

- A discrete-time system with difference equation of
  - $y[n] = ny[n - 1] + x[n], \quad n > 0$
  - The system at rest (i.e.  $y[-1] = 0$ )
- Check if the system is BIBO stable?
- If  $x[n] = u[n]$ , then  $|x[n]| \leq 1$ . But for this bounded input, the output is
  - $n = 0 \Rightarrow y[0] = x[0] = 1$
  - $n = 1 \Rightarrow y[1] = 1y[0] + x[1] = 2$
  - $n = 2 \Rightarrow y[2] = 2y[1] + x[2] = 5$
  - ...  $\Rightarrow \infty$
- The input of unit step sequence is bounded, but the output is unbounded. Hence the system is BIBO unstable.

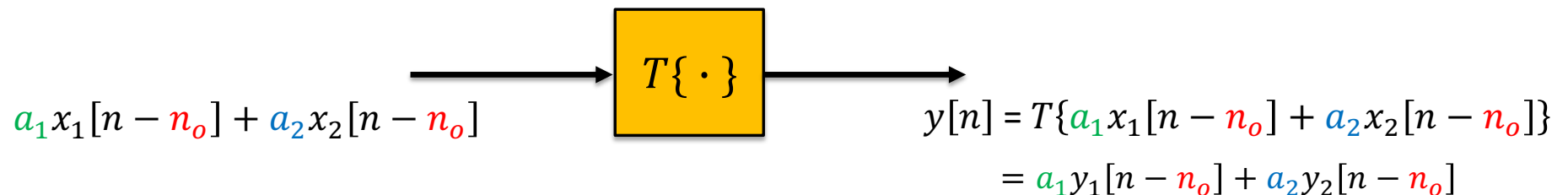
# LTI Systems

# Linear Time-Invariant (LTI) Discrete-Time Systems

- LTI discrete-time systems satisfy both Linear and Time-Invariant properties.



- For an integer  $n_o$  and arbitrary constants  $a_1$  and  $a_2$ , LTI system property is



# LTI System Examples

- Absolute Magnitude System:  $y[n] = |x[n]|$ 
  - It is Time-Invariant but **not Linear**
- Time scaling System :  $y[n] = n x[n]$ 
  - It is Linear but **not Time-Invariant**
- 3-sample average System :  $y[n] = \frac{1}{3}(x[n] + x[n - 1] + x[n - 2])$ 
  - It satisfies both **Linear and Time-Invariant** properties
  - **It is an LTI system**



# LTI System Exercise 1

**Determine whether the linear system  $y[n] = 2x[n - 5]$  is time invariant.**

## Solution

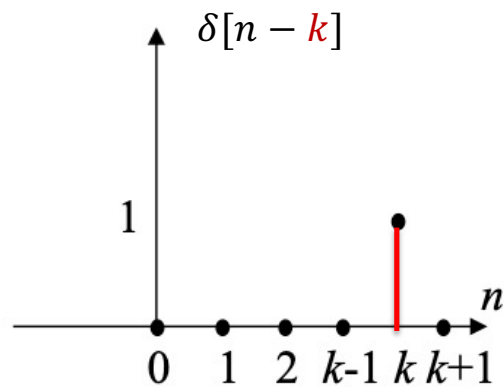
- Let the input and output be  $x_1[n]$  and  $y_1[n]$ , respectively, then the system output is
  - $y_1[n] = 2x_1[n - 5]$
- Again, let  $x_2[n] = x_1[n - n_0]$  be the shifted input and  $y_2[n]$  be the output using the shifted input can be described as
  - $y_2[n] = 2x_2[n - 5] = x_1[n - n_0 - 5] = y_1[n - n_0]$
- As  $y_2[n] = y_1[n - n_0]$ , then the system is **time-invariant**.

# Impulse Response and Convolution

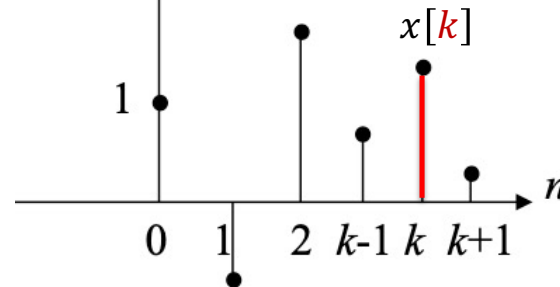
# Representation of Discrete-Time Signals by Sum of Scaled and Shifted Unit Impulse Signals (1)

- A discrete-time signal,  $x[n]$  may be shifted in time (delayed or advanced) by replacing the variables  $n$  with  $(n - k)$  where  $k > 0$  is an integer
  - $x[n - k] \Rightarrow x[n]$  delayed by  $k$  samples
  - $x[n + k] \Rightarrow x[n]$  advanced by  $k$  samples
- For example, consider a shifted version of the unit impulse function. If we multiply an arbitrary signal  $x[n]$  by this function, we obtain a signal that is zero everywhere, except at  $n = k$ .
  - $y[n] = x[n] \cdot \delta[n - k] = x[k] \cdot \delta[n - k]$

# Representation of Discrete-Time Signals by Sum of Scaled and Shifted Unit Impulse Signals (2)

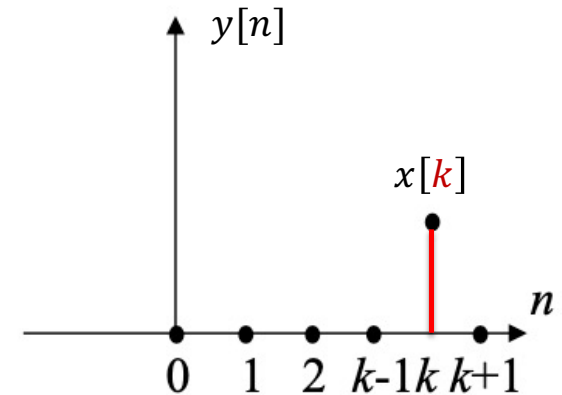


X

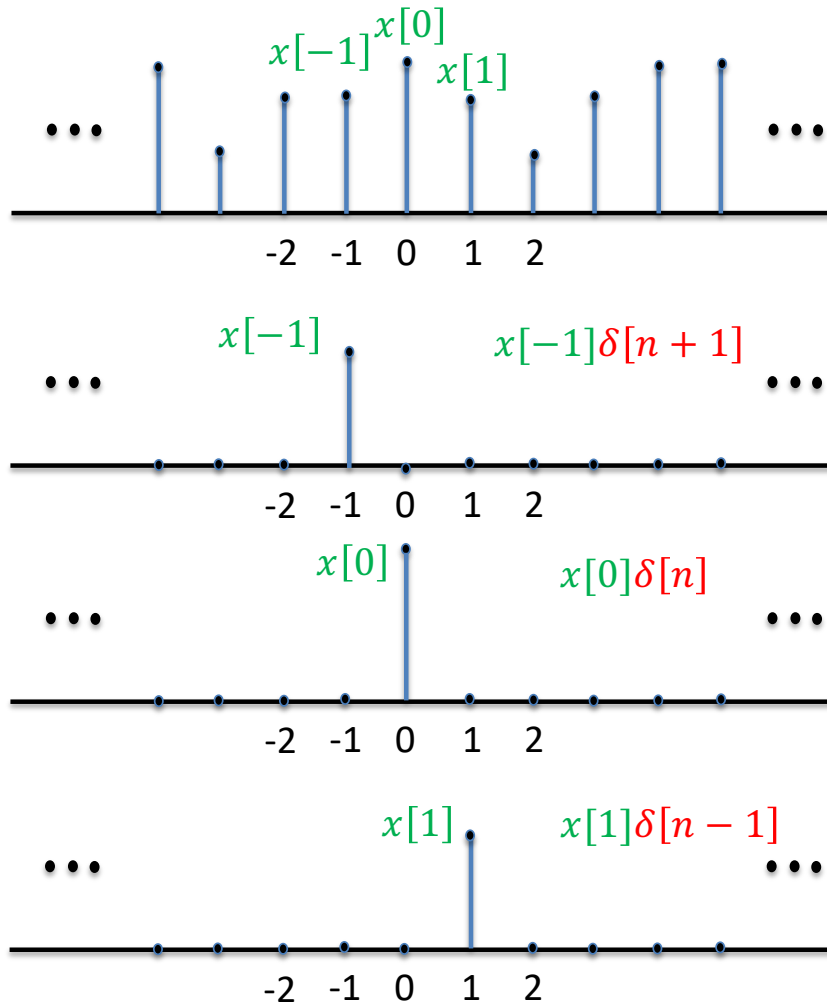


$$y[n] = x[k] \cdot \delta[n - k]$$

=



# Unit Impulse based Composite Sequence Expression



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

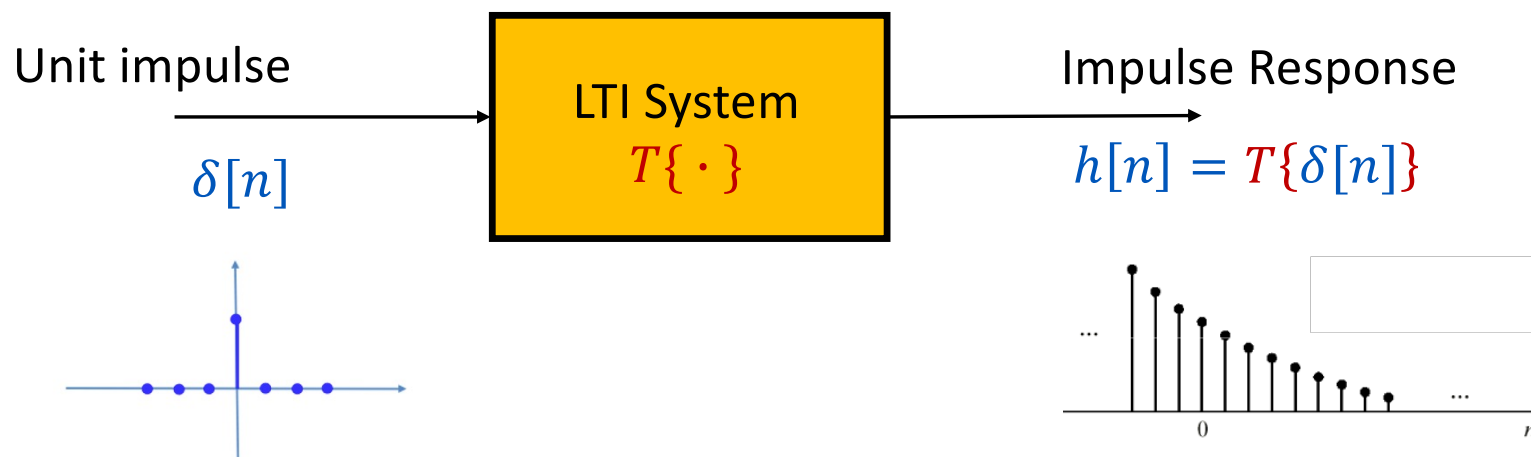


$$x[n] = \sum_{k=-\infty}^{\infty} x[k]\delta[n-k]$$

Any discrete-time signals can be expressed in terms of **summation of scaled and shifted unit sample sequences**  $x[k]\delta[n-k]$ .

# Impulse Response

- If the input is **unit impulse** (unit sample) sequence  $\delta[n]$ , the corresponding output is called the **impulse response**  $h[n]$  of the LTI system



# Convolution : Why Impulse Response is so important?

- The output of any LTI system is a convolution operation of the input signal with the unit impulse response:

$$x[n] = \sum_{k=-\infty}^{\infty} x[k] \delta[n-k] \longrightarrow \boxed{\text{LTI System } h[n]} \longrightarrow y[n] = x[n] * h[n]$$

Convolution operator

$$y[n] = T\{x[n]\} = T\left\{\sum_{k=-\infty}^{\infty} x[k] \delta[n-k]\right\} \underset{\text{Linear Property}}{=} \sum_{k=-\infty}^{\infty} x[k] T\{\delta[n-k]\} \underset{\text{Time-Invariant Property}}{=} \sum_{k=-\infty}^{\infty} x[k] h[n-k] = x[n] * h[n]$$

- Any Discrete-Time LTI system can be **completely characterized by its unit impulse response ( $h[n]$ )**.

# Performing the Convolution Algorithmically

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

- **Ingredients**

- A sequence  $x[k]$
- A second sequence  $h[k]$

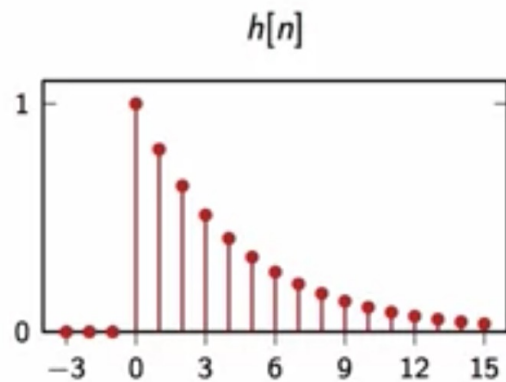
- **The recipe:**

- Time-reverse  $h[k]$
- At each step  $n$  (from  $-\infty$  to  $\infty$ ):
  - Center the time-reversed  $h[k]$  in  $n$  (i.e. shift by  $-n$ )
  - Compute the inner product

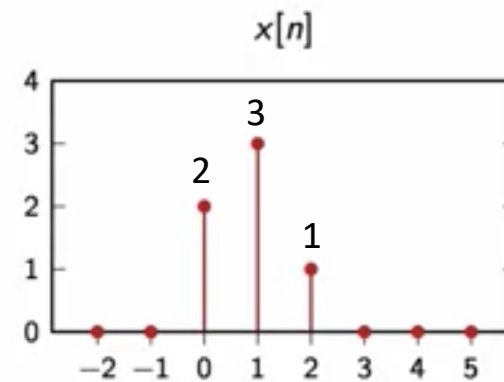


# Convolution Example 1

- Compute the  $x[n] * h[n]$

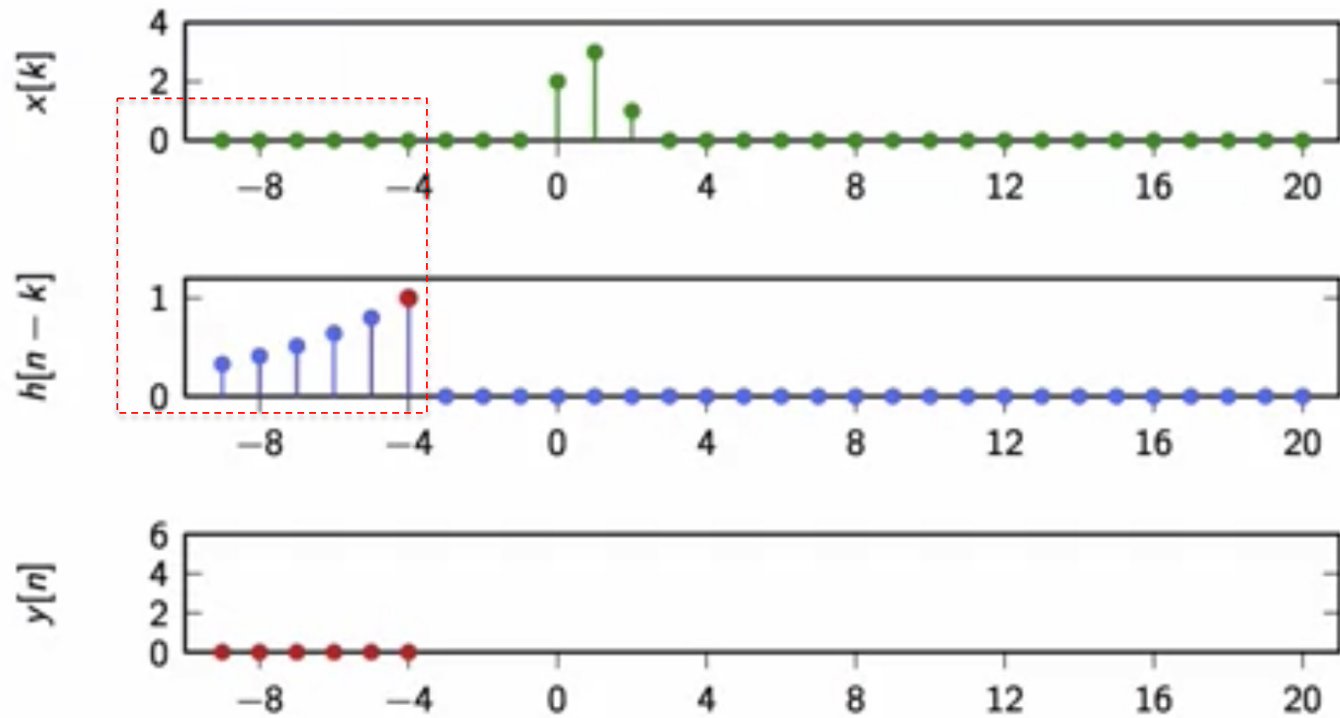


$$h[n] = \alpha^n u[n]$$



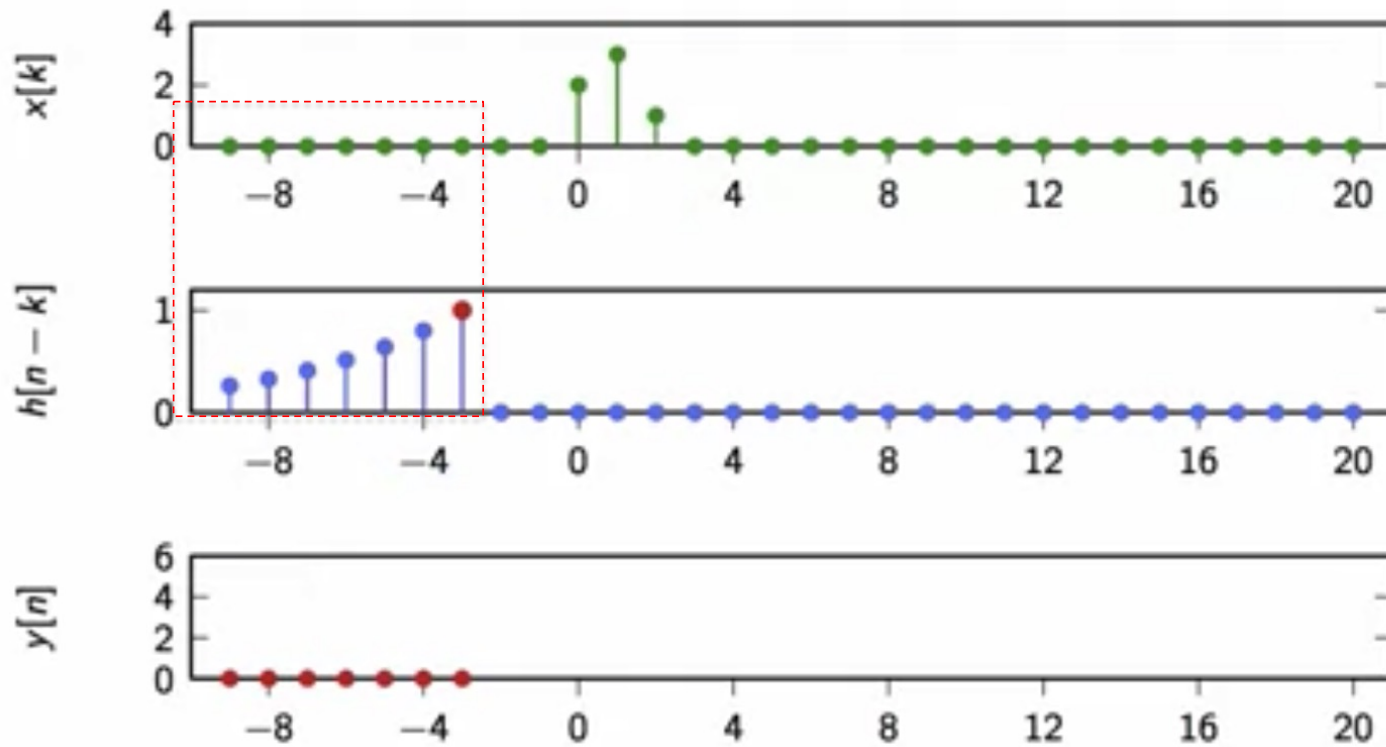
$$x[n] = \begin{cases} 2 & n = 0 \\ 3 & n = 1 \\ 1 & n = 2 \\ 0 & \text{otherwise} \end{cases}$$

$x[n] * h[n]$  when  $n = -4$



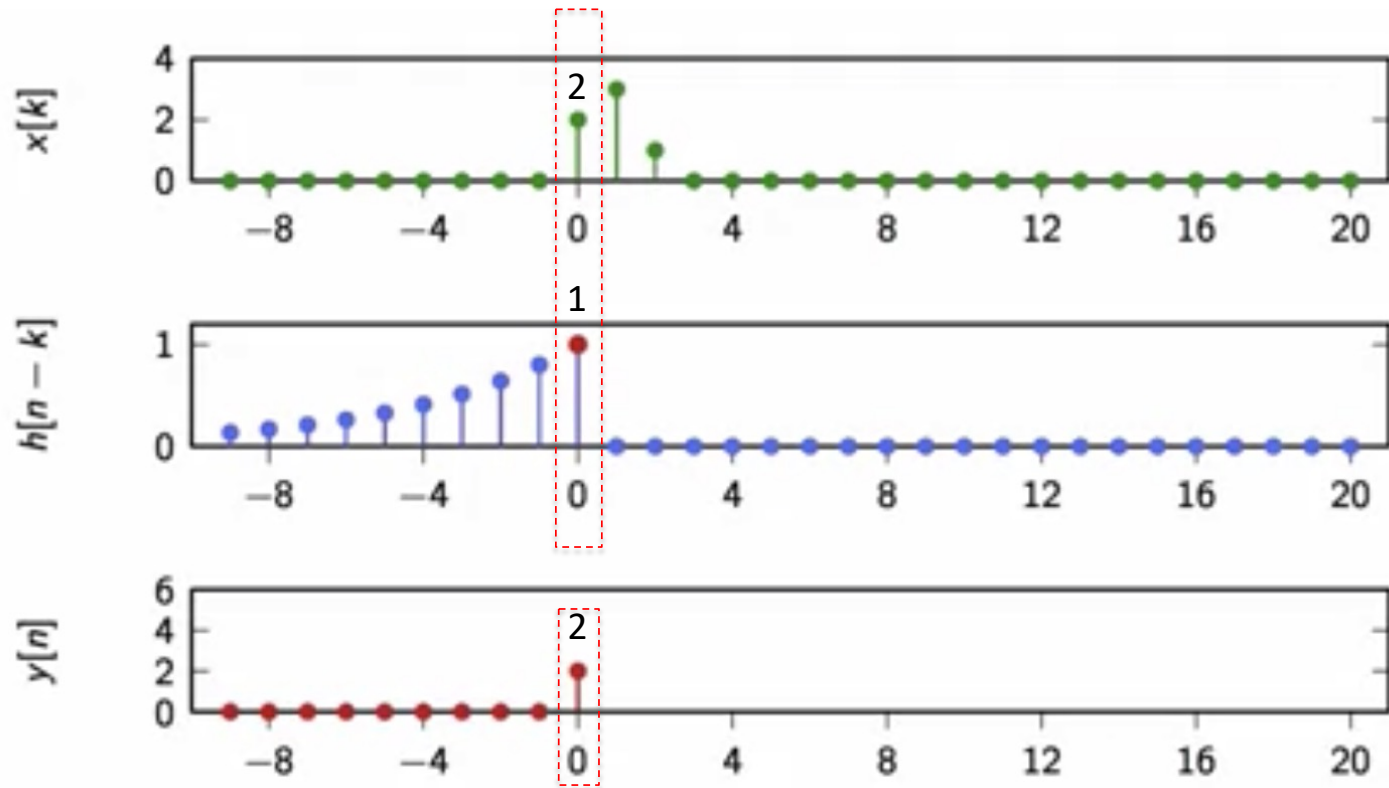
$h[-4 - k]$

$x[n] * h[n]$  when  $n = -3$



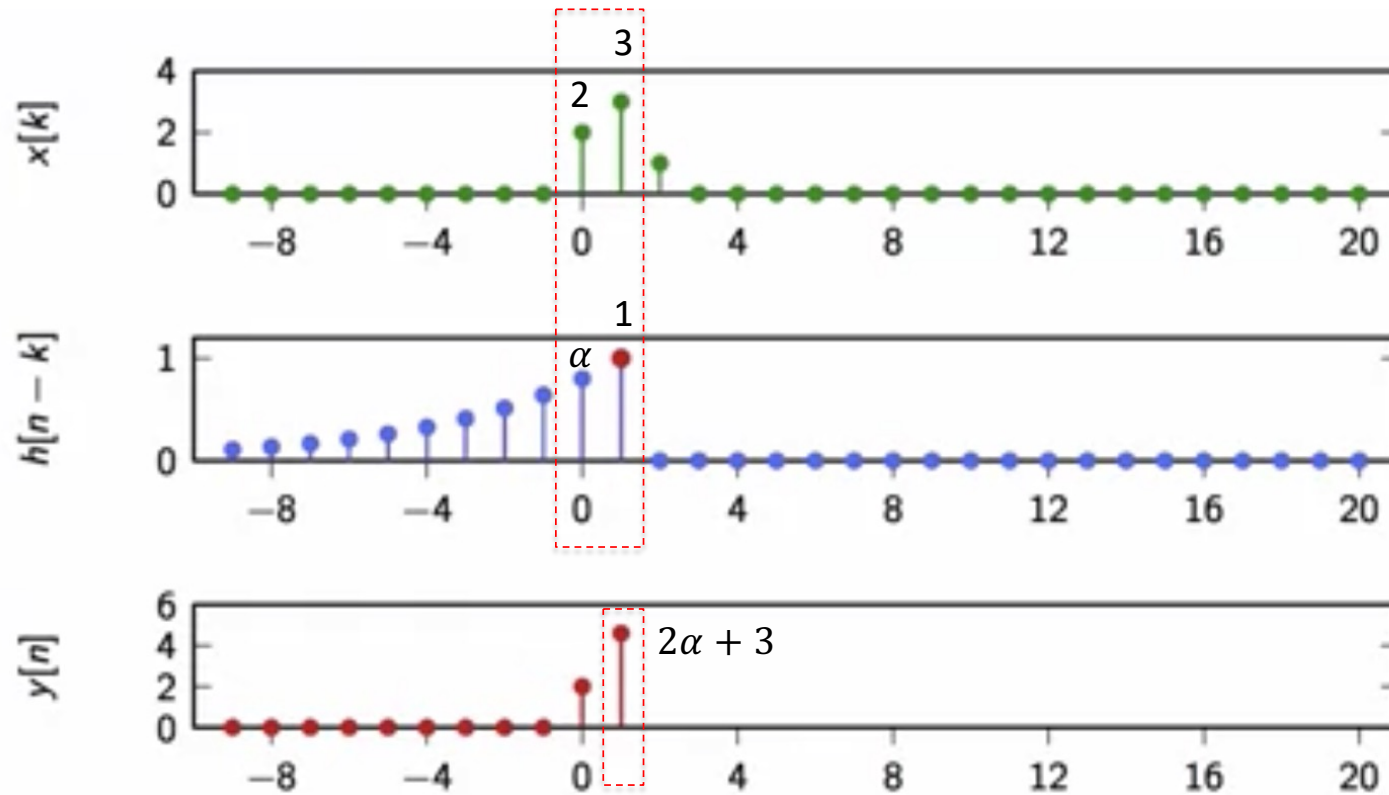
$h[-3-k]$

$x[n] * h[n]$  when  $n = 0$



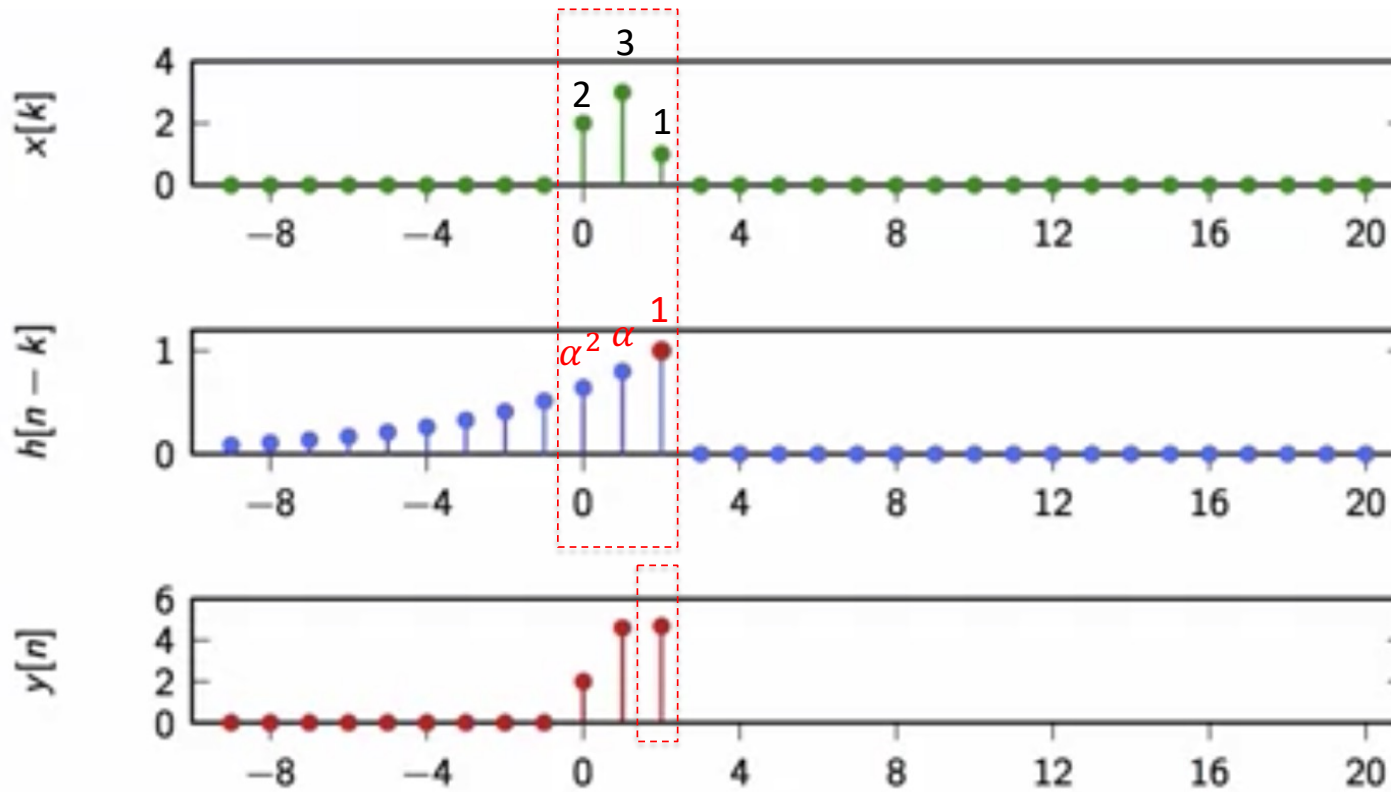
$h[-k]$

$x[n] * h[n]$  when  $n = 1$



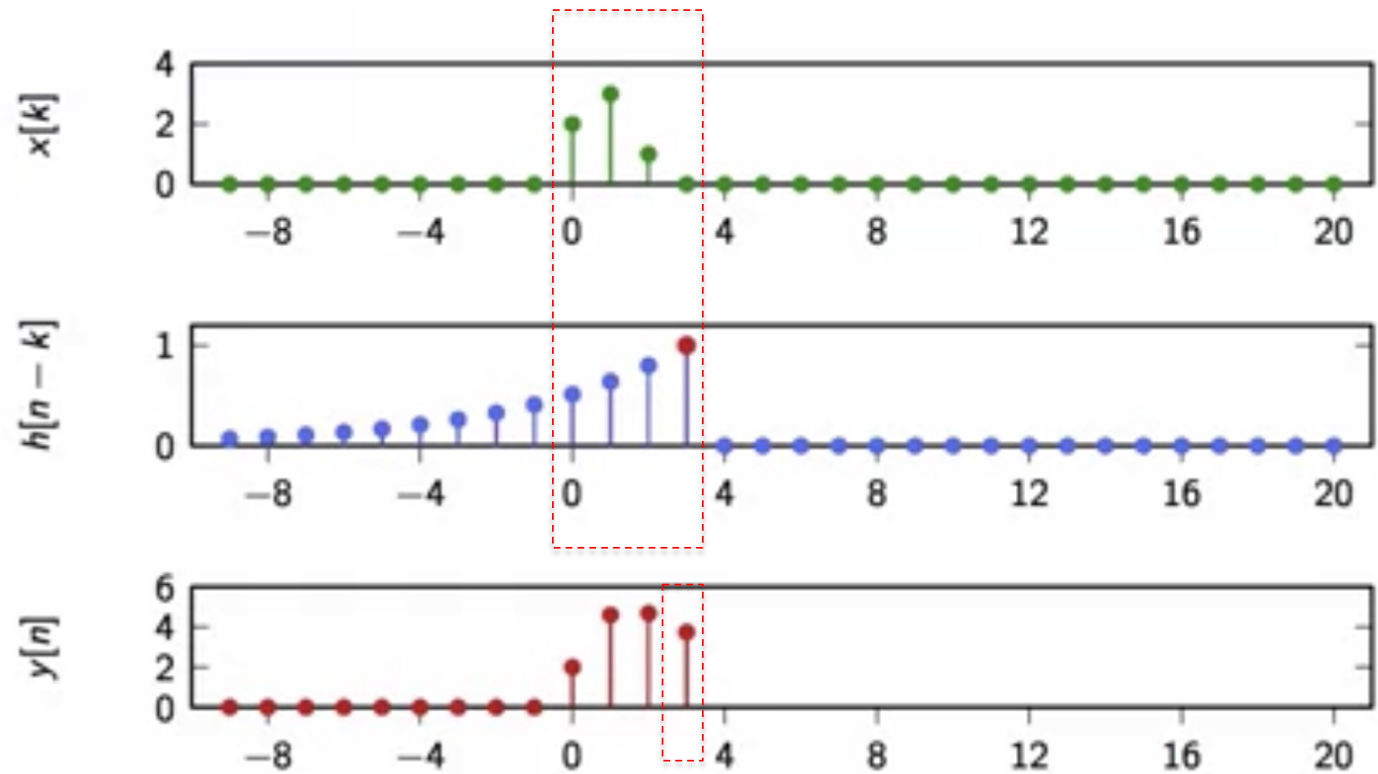
$h[1 - k]$

$x[n] * h[n]$  when  $n = 2$



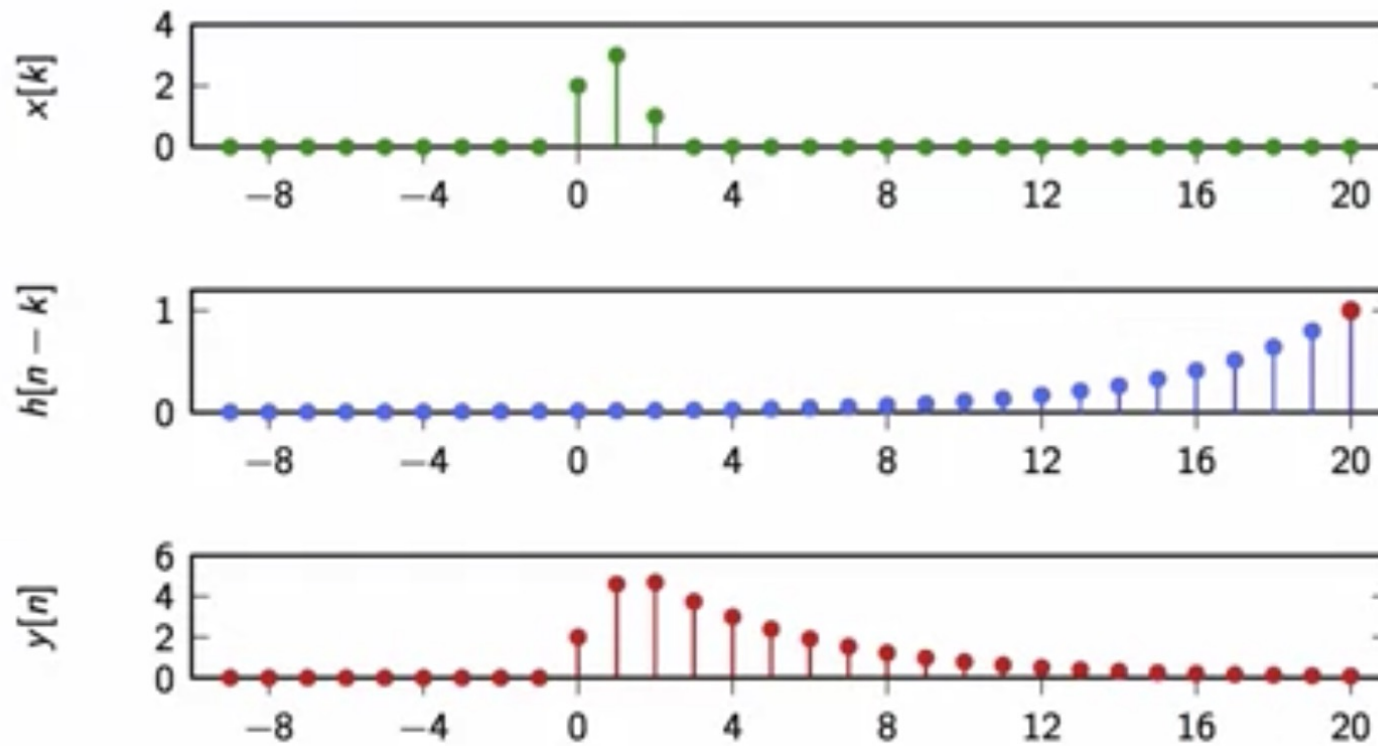
$h[2 - k]$

$x[n] * h[n]$  when  $n = 3$



$h[3 - k]$

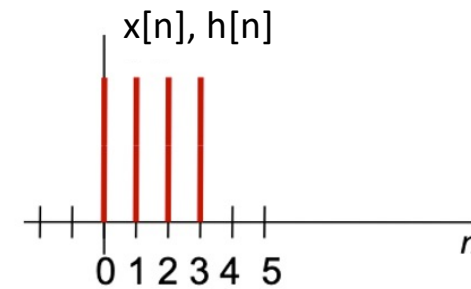
$x[n] * h[n]$  for all  $n$





# Convolution Example 2

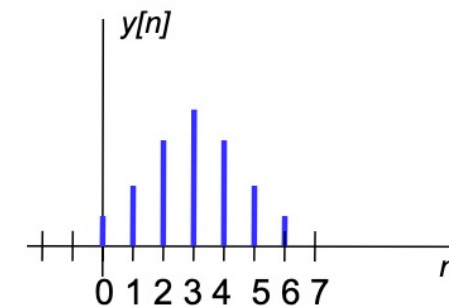
$$x[n] = \begin{cases} 1 & 0 \leq n \leq 3 \\ 0 & \text{otherwise} \end{cases} \quad h[n] = \begin{cases} 1 & 0 \leq n \leq 3 \\ 0 & \text{otherwise} \end{cases}$$

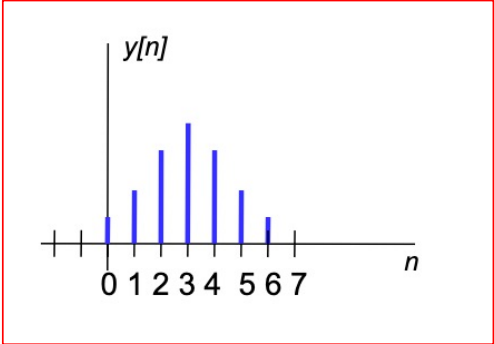
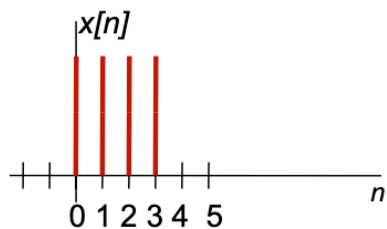
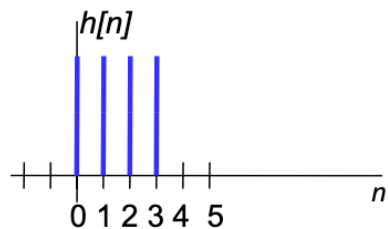


What is the output  $y[n]$  of the LTI system?

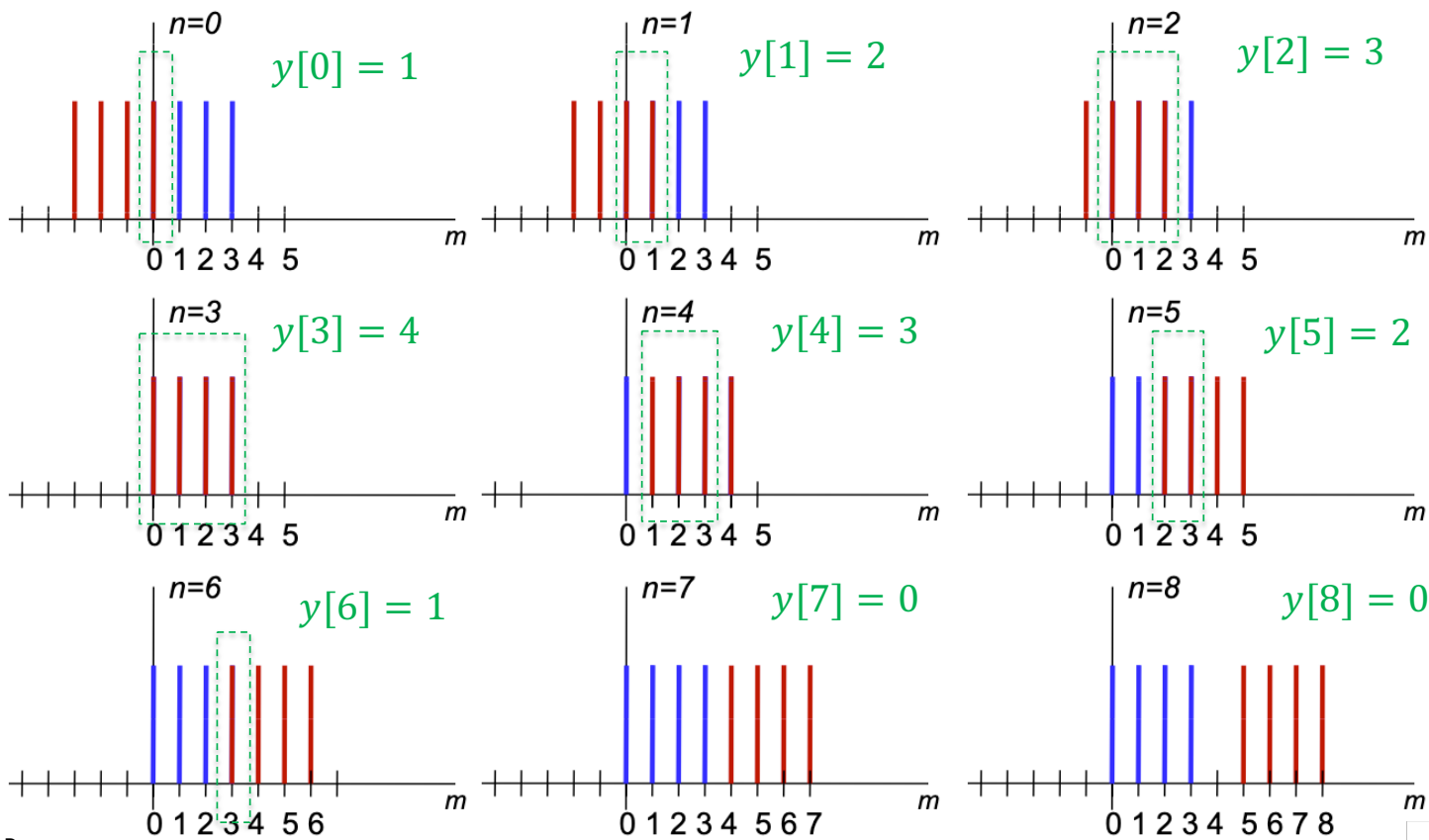
Solution:

$$y[n] = x[n] * h[n] = \sum_{m=-\infty}^{\infty} h[m] x[n-m]$$
$$= \begin{cases} \sum_{m=0}^n 1 \cdot 1 = (n+1) & 0 \leq n \leq 3 \\ \sum_{m=n-3}^3 1 \cdot 1 = (7-n) & 4 \leq n \leq 6 \\ 0 & n \leq 0, n \geq 7 \end{cases}$$





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



# Convolution Example 3

- The impulse response of an LTI system is of the form:
  - $h[n] = a^n u[n] \quad |a| < 1$
- And the input to the system is of form:
  - $x[n] = b^n u[n] \quad |b| < 1, b \neq a$
- Determine the output of the system using discrete convolution operation.

Solution:

$$\begin{aligned} y[n] &= h[n] * x[n] = \sum_{k=-\infty}^{\infty} h[k]x[n-k] = \sum_{k=-\infty}^{\infty} a^k u[k] b^{n-k} u[n-k] \\ &= \sum_{k=0}^n a^k b^{n-k} u[n] = b^n \sum_{k=0}^n a^k b^{-k} u[n] = b^n \sum_{k=0}^n \left(\frac{a}{b}\right)^k u[n] \\ &= b^n \left[ \frac{1 - \left(\frac{a}{b}\right)^{n+1}}{1 - \left(\frac{a}{b}\right)} \right] u[n] = b^n \left[ \frac{\frac{b^{n+1} - a^{n+1}}{b^{n+1}}}{\frac{b-a}{b}} \right] u[n] = \left[ \frac{b^{n+1} - a^{n+1}}{b-a} \right] u[n] \end{aligned}$$

# Convolution Example 4

- Consider a discrete-time system with finite-duration input  $x[n] = \{1,1,1,1\}$  and impulse response  $h[n] = a^k u[n]$ ,  $|a| < 1$ .
- Determine the response  $y[n]$  of this LTI system.

## Solution:

We recognize that  $x[n]$  can be written as the difference between two unit-step sequences, i.e.  $x[n] = u[n] - u[n - 4]$ . Hence, we can solve for  $y[n]$  as the difference between the output of the system with a step input and the output of the system with a delayed step input. Thus, we solve for the response to a unit step as:

$$y_1[n] = h[n] * u[n] = \sum_{k=-\infty}^{\infty} a^k u[k] u[n - k] = \sum_{k=0}^n a^k u[n] = \left[ \frac{1 - a^{n+1}}{1 - a} \right] u[n]$$

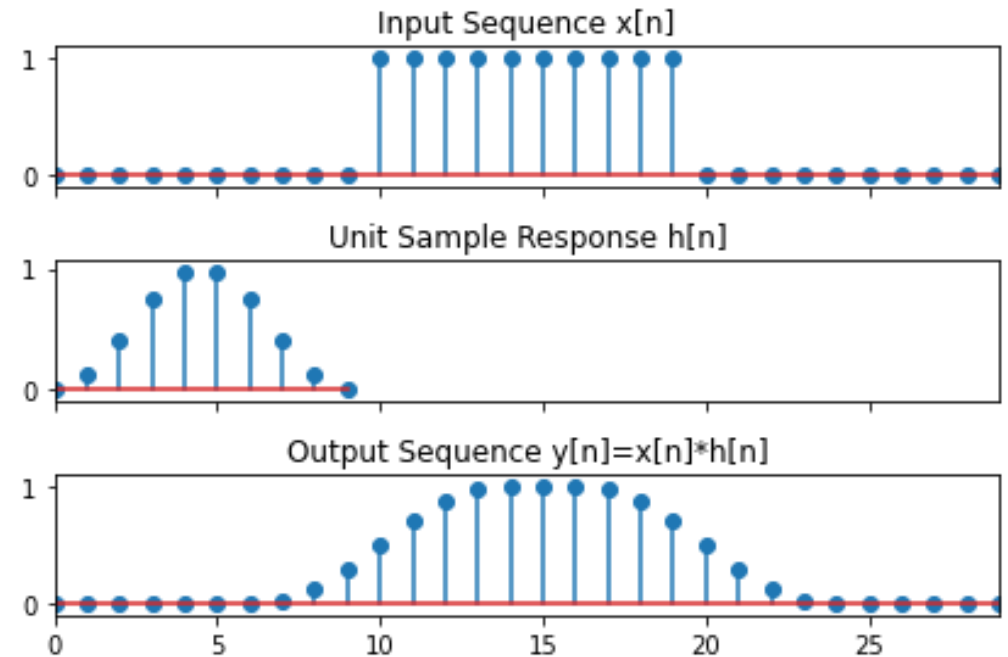
$$y[n] = h[n] * u[n] - h[n] * u[n - 4] = y_1[n] - y_1[n - 4] = \left[ \frac{1 - a^{n+1}}{1 - a} \right] u[n] - \left[ \frac{1 - a^{n-3}}{1 - a} \right] u[n - 4]$$

# Convolution : Python Code

```
import matplotlib.pyplot as plt
from scipy import signal
import numpy as np

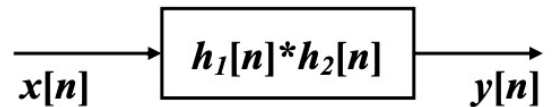
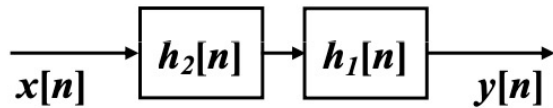
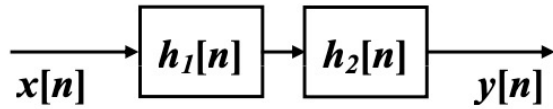
x = np.repeat([0., 1., 0.], 10)
h = signal.hann(10)
y = signal.convolve(x, h, mode='same') / sum(h)

fig, (ax_x, ax_h, ax_y) = plt.subplots(3, 1, sharex=True)
ax_x.stem(x)
ax_x.set_title('Input Sequence x[n]')
ax_x.margins(0, 0.1)
ax_h.stem(h)
ax_h.set_title('Unit Sample Response h[n]')
ax_h.margins(0, 0.1)
ax_y.stem(y)
ax_y.set_title('Output Sequence y[n]=x[n]*h[n]')
ax_y.margins(0, 0.1)
fig.tight_layout()
fig.show()
```

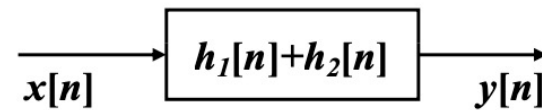
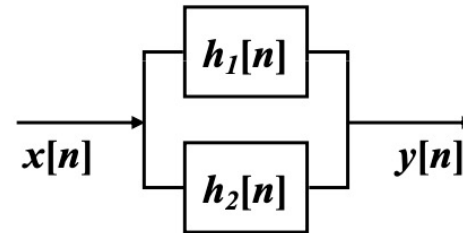


<https://docs.scipy.org/doc/scipy/reference/generated/scipy.signal.convolve.html>

# Convolution Properties

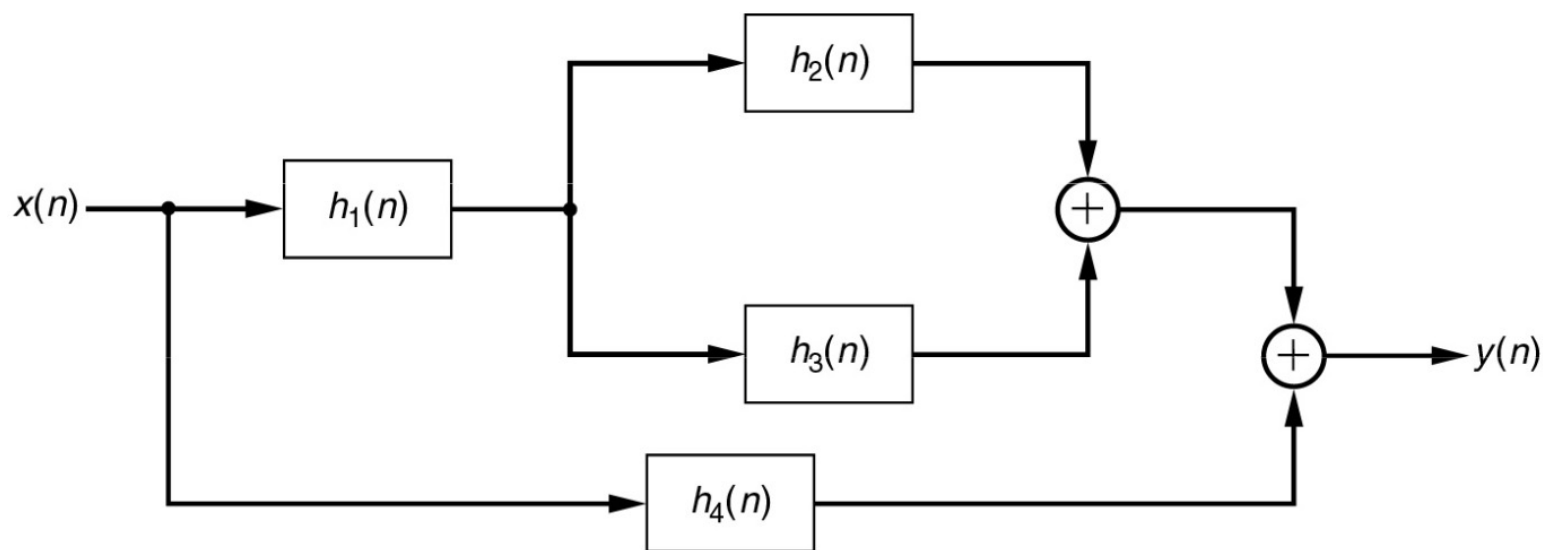


$$h_1[n]*h_2[n]=h_2[n]*h_1[n]$$



$$h_1[n]+h_2[n]=h_2[n]+h_1[n]$$

# More Complex System Interconnections



$$y[n] = x[n] * h_c[n]$$

$$h_c[n] = h_1[n] * (h_2[n] + h_3[n]) + h_4[n]$$

# Stability Analysis



# Stability Analysis of LTI Systems

- Similar to the continuous-time LTI systems, we can analyze the stability of the system based on its impulse response.
- A Discrete-Time LTI system is Bounded-Input Bounded-Output (BIBO) **stable** if and only if its **unit sample response**  $h[k]$  is **absolutely summable**.

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

- Let  $x[n]$  be a bounded input sequence {i.e.  $|x[n]| < B_x$  for all  $n$ , where  $B_x$  is a finite number}. We must show that the output is bounded when  $S$  is finite. To this end, we work again with the convolution formula.

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

- If we take the absolute value of both sides of this equation, we obtain

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

- Now, the absolute value of the sum of terms is always less than or equal to the sum of the absolute values of the terms

$$|y[n]| \leq B_x \sum_{k=-\infty}^{\infty} |h[k]| \leq B_x S$$

- Hence, since both  $B_x$  and  $S$  are finite, the output is also bounded, i.e., **an LTI system is stable if its unit sample response is absolutely summable.**

# LTI System Stability Example (1)

- Check the stability of the first-order recursive system shown below:
  - $y[n] = a y[n - 1] + x[n]$
- The impulse response of this system is
  - $h[n] = a^n u[n]$  for all formula
- Its stability factor  $S$  is

$$S = \sum_{n=-\infty}^{\infty} |h[n]| < \sum_{n=0}^{\infty} |a|^n$$

## LTI System Stability Example (2)

- It is obvious that  $S$  is unbounded for  $|a| \geq 1$ , since each term in the series are greater than or equal to 1.
- For  $|a| < 1$ , we can apply the infinite geometric sum formula, to find

$$S = \frac{1}{1 - |a|} \quad \text{for } |a| < 1$$

- Since  $S$  is finite for  $|a| < 1$ , the system is stable.

# Block Diagram Representation

# Representations of LTI Discrete-Time Systems

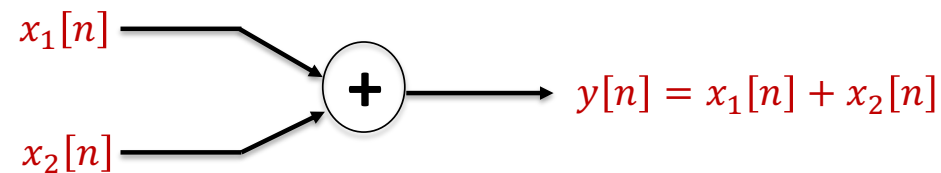
In general, LTI discrete-Time systems can be represented by

- **Unit Sample Response** :  $h[n]$
- **Difference Equations** with or without feedback :
  - $y[n] = b_0x[n] + b_1x[n - 1] + b_2x[n - 2]$
  - $y[n] = a_1y[n - 1] + a_2y[n - 2] + b_0x[n] + b_1x[n - 1] + b_2x[n - 2]$
- **Block Diagram** with basic operation elements
  - Adders, Constant Multipliers, and Unit Delay Elements
- **Transfer Function** in z-domain :
  - $H(z) = \frac{b_0 + b_1z^{-1} + b_2z^{-2}}{1 + a_1z^{-1} + a_2z^{-2}}$

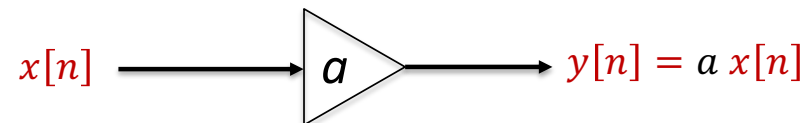
# Block Diagram Representation of DT System

Some basic blocks that can be interconnected to form complex systems.

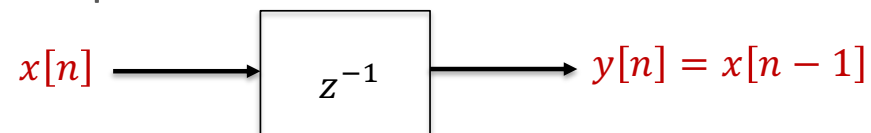
- **An Adder** : A system that performs the addition of two signal sequences to form another sequence



- **A Constant Multiplier** : This operation simply represents applying a scale factor on the input  $x[n]$ .

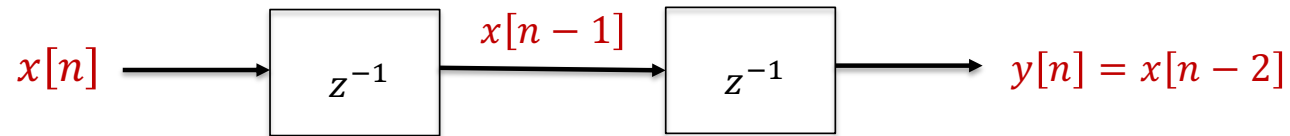


- **A Unit Delay Element** : The unit delay is a special system that simply delays the signal passing through it by one sample.

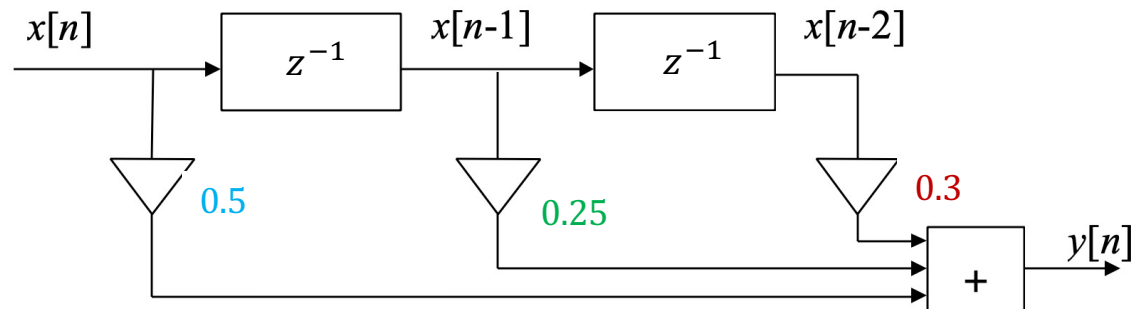


# Difference Equation Representation of DT System

- Example 1:  $y[n] = x[n - 2]$



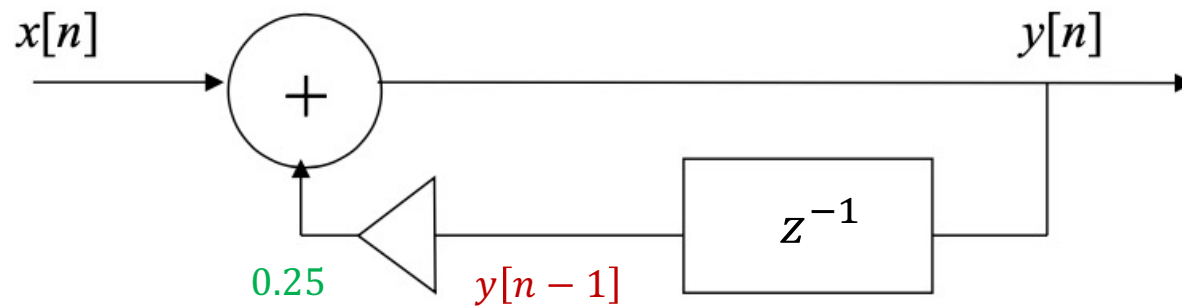
- Example 2:  $y[n] = 0.5x[n] + 0.25x[n - 1] + 0.3x[n - 2]$





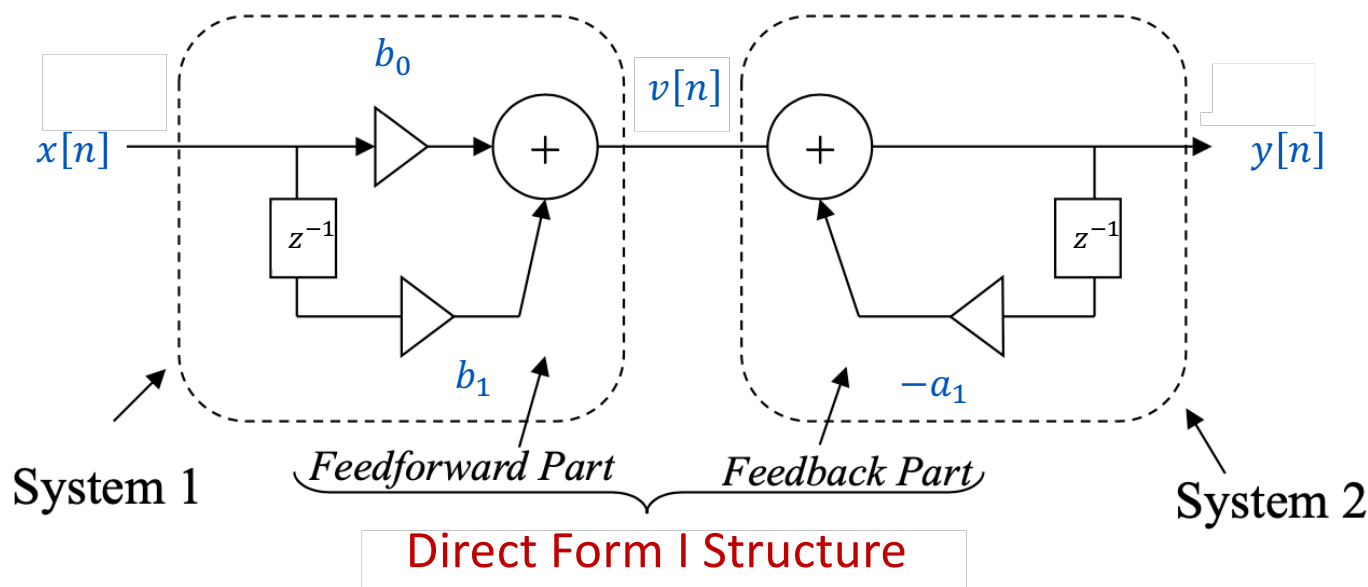
# Difference Equation with Feedback

- Example 3:  $y[n] = x[n] + 0.25y[n - 1]$



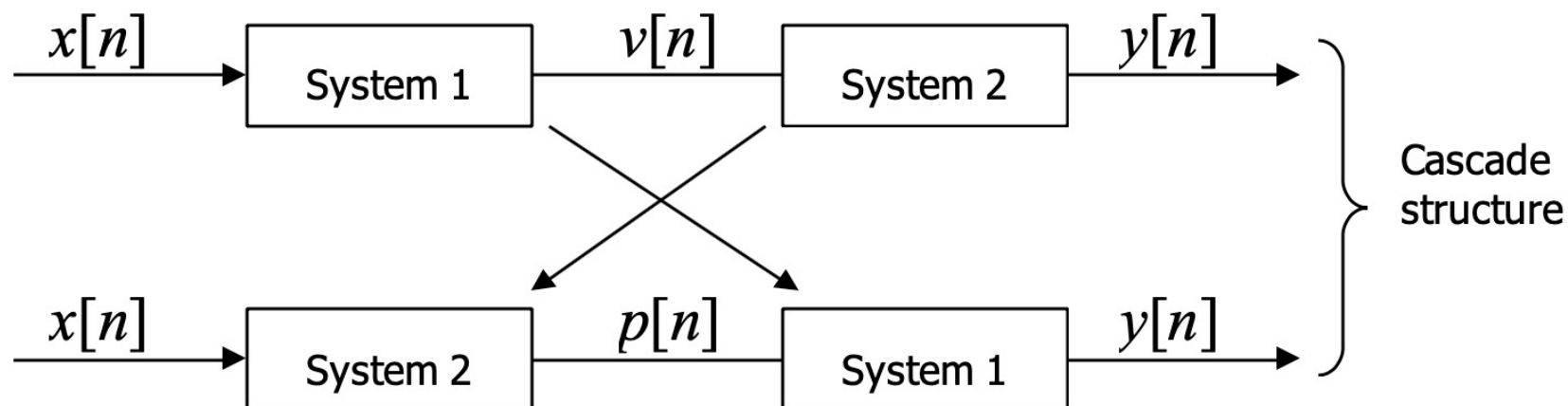
# Direct Form Structure (or Direct Form I)

- Draw a system implementation for the below difference equation.
  - $y[n] = b_0x[n] + b_1x[n - 1] - a_1y[n - 1]$
- We can write the above difference equation as a set of two equations
  - $v[n] = b_0x[n] + b_1x[n - 1]$  and  $y[n] = v[n] - a_1y[n - 1]$

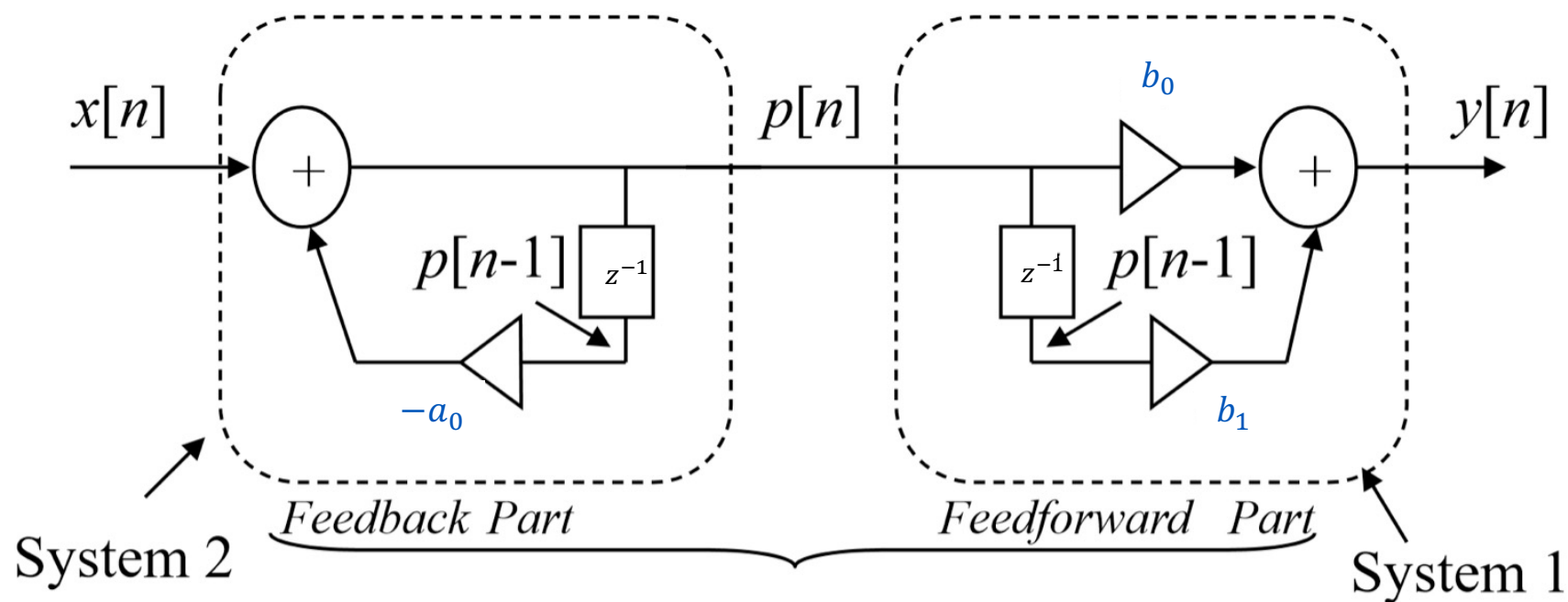


# Cascade Structure

- Without changing the input-output relationship, we can **reverse the ordering** of the two systems in the cascade representation.



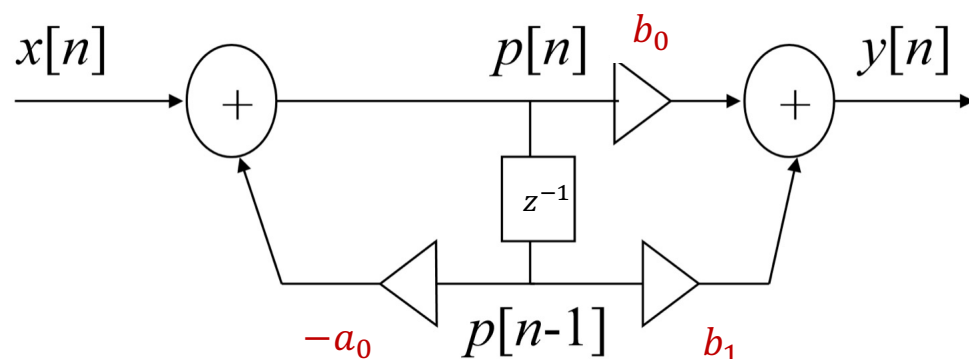
# Canonic Form Structure (or Direct Form II)



Direct Form II Structure

# Canonical Form

- There is no need for two delay operations in the Direct Form II structure, they can be **combined into a single delay**.



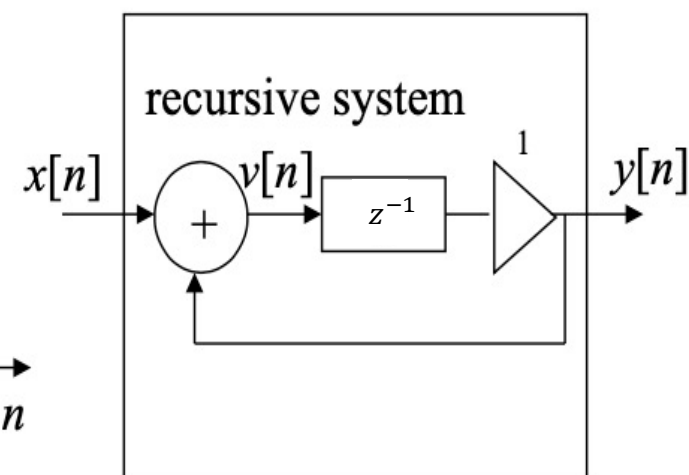
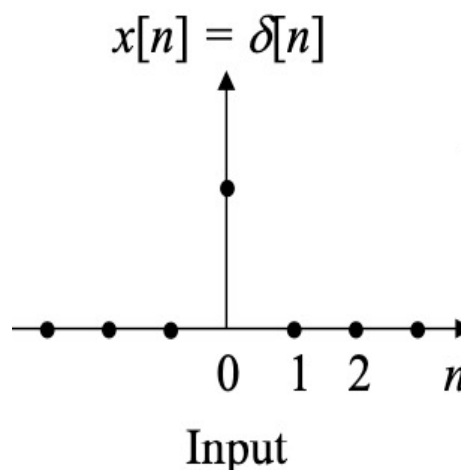
- Since delay operations are implemented with memory in a computer, this Canonical Form implementation would **require less memory** compared to the implementation of Direct Form I and II structure.

# IIR and FIR Systems

# Infinite Impulse Response (IIR) System

- If the impulse response of an LTI system is of **infinite duration**, the system is said to be an **Infinite Impulse Response (IIR) system**.
- Example

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



- If  $x[n] = \delta[n]$ , calculate  $h[n]$  for  $n = 0, 1, 2, \dots$

## Find the Impulse Response based Difference Equation

- Find the impulse response  $h[n]$  of the following first-order recursive system.

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

- To find  $h[n]$ , we let  $x[n] = \delta[n]$  and apply the zero-initial condition.

$$n = 0, y[0] = h[0] = ay[-1] + \delta[0] = 1$$

$$n = 1, y[1] = h[1] = ay[0] + \delta[1] = a$$

$$n = 2, y[2] = h[2] = ay[1] + \delta[2] = a^2$$

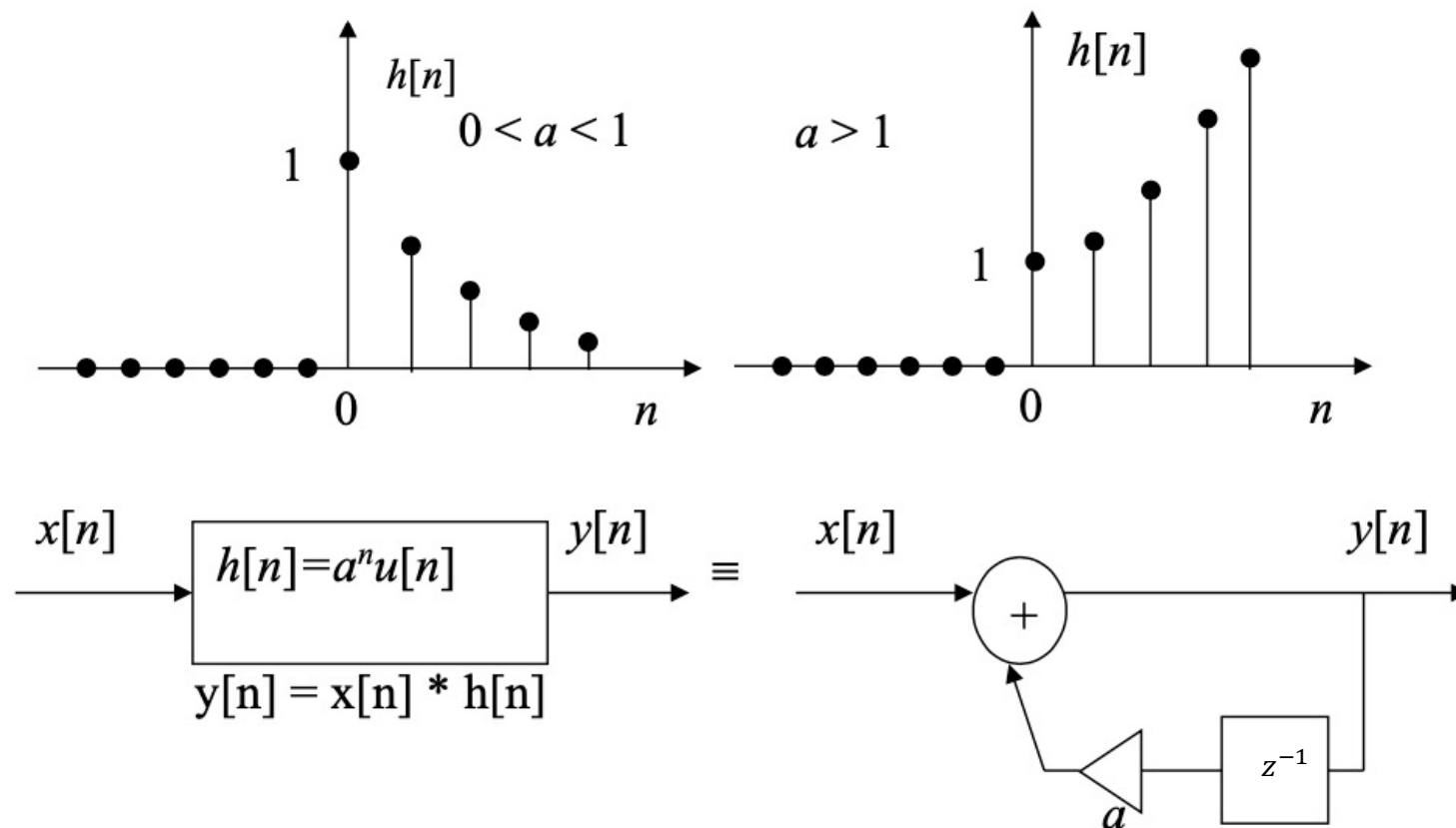
⋮

$$n = n, y[n] = h[n] = a^n \quad \text{for } n \geq 0$$

Infinite-duration unit sample response



- $y[n] = h[n] = 0$  for  $n < 0$ , because  $\delta[n]$  is zero for  $n < 0$  and  $y[-1] = 0$ .
- Hence,  $h[n] = a^n u[n]$  for all  $n$



# Finite Impulse Response (FIR) System

- Find the impulse response  $h[n]$  of the following fourth order **non-recursive** system.

$$y[n] = a_0x[n] + a_1x[n - 1] + a_2x[n - 2] + a_3x[n - 3] + a_4x[n - 4]$$

- To find  $h[n]$ , we let  $x[n] = \delta[n]$

$$n=0 \rightarrow h[0] = a_0\delta[0] + a_1\delta[-1] + a_2\delta[-2] + a_3\delta[-3] + a_4\delta[-4] = a_0$$

$$n=1 \rightarrow h[1] = a_0\delta[1] + a_1\delta[0] + a_2\delta[-1] + a_3\delta[-2] + a_4\delta[-3] = a_1$$

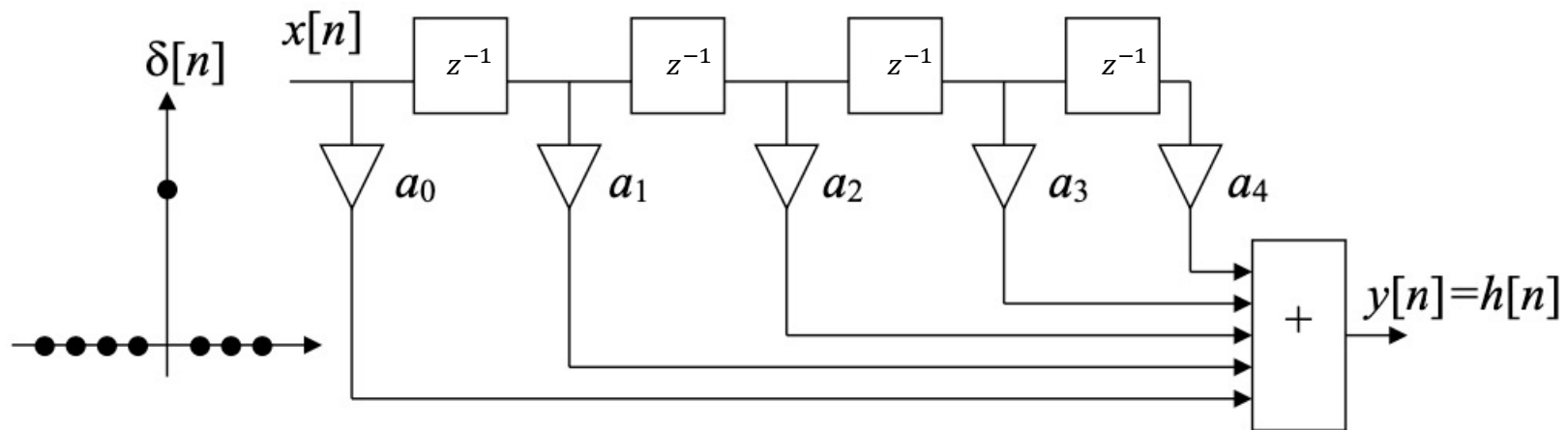
$$n=2 \rightarrow h[2] = a_0\delta[2] + a_1\delta[1] + a_2\delta[0] + a_3\delta[-1] + a_4\delta[-2] = a_2$$

$$n=3 \rightarrow h[3] = a_0\delta[3] + a_1\delta[2] + a_2\delta[1] + a_3\delta[0] + a_4\delta[-1] = a_3$$

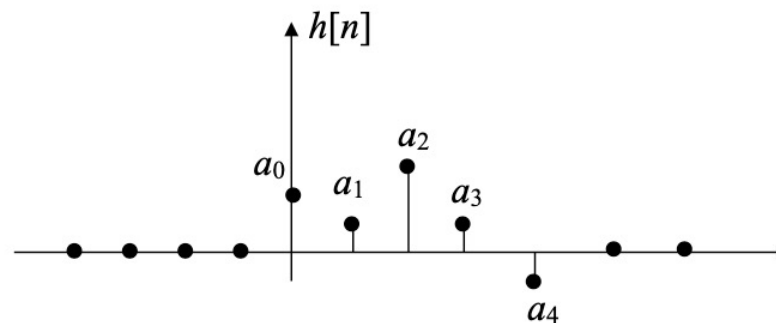
$$n=4 \rightarrow h[4] = 0 + 0 + 0 + 0 + a_4\delta[0] = a_4$$

$$n=5 \rightarrow h[5] = 0 + 0 + 0 + 0 + a_4\delta[1] = 0$$

- For  $n \geq 5$ ,  $h[n] = 0$ , since the nonzero value of  $\delta[n]$  has moved out of the memory of this system.



If the impulse response of an LTI system is of **finite duration**, the system is said to be an **finite Impulse Response (FIR) system**. In addition, non-recursive systems have finite impulse responses.



$a_0, a_1, a_2, a_3$  and  $a_4$  are called coefficients (+ or -) or constants.

# Summary (1)

- The definition of a discrete-time system
- The definition of the memory, Time Invariant, Linear, Causal and Stable Systems
- Interpretation of a discrete-time signal as a weighted sum of delayed impulses
- Definition and understanding of convolution (including hand and graphical computation of convolution)
- The impulse response of a linear and time-invariant system, and how to calculate it from a difference equation

## Summary (2)

- Basic blocks of a discrete-time system: the adder, multiplier and unit delay
- How to draw the block diagram of a discrete-time system given its difference equation
- How to write the difference equation of a discrete-time system given its block diagram
- The difference between Direct Forms I and II and Canonical Form
- The difference between an FIR and an IIR system. In particular, that the impulse responses of FIR systems have identical values to the coefficients of the difference equation.
- Given the impulse responses of two cascaded systems, be able to compute the overall impulse response.