

Ch.2 *Linear Time-Invariant Systems*

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Outline

- **Discrete-Time LTI Systems**
- **Continuous-Time LTI Systems**
- **Properties of LTI Systems**
- **Differential or Difference Equations**

Part I *Discrete-Time LTI Systems*

Outline

- The Representation of Discrete-Time Signals in Terms of Impulses
- The Discrete-Time Unit Impulse Response
- The Convolution-Sum Representation of LTI Systems

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Linear Systems

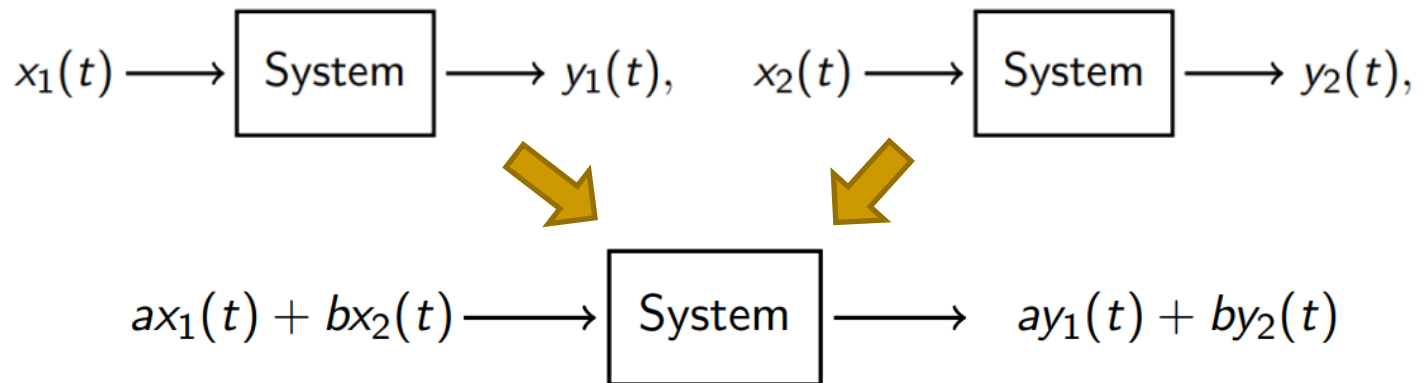
- A system is linear if the following condition holds for any two inputs $x_1(t)$ and $x_2(t)$:

- **Additivity**

The response to $x_1(t) + x_2(t)$ is $y_1(t) + y_2(t)$

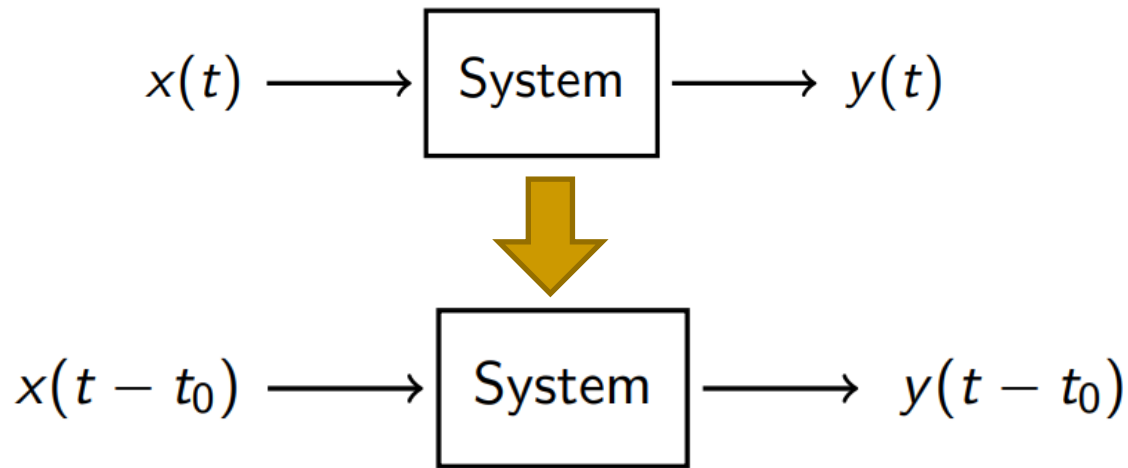
- **Scaling/Homogeneity**

The response to $a \cdot x_1(t)$ is $a \cdot y_1(t)$



Time Invariance

- A system is time-invariant if a time-shift (advance or delay) at the input causes **an identical shift** at the output.



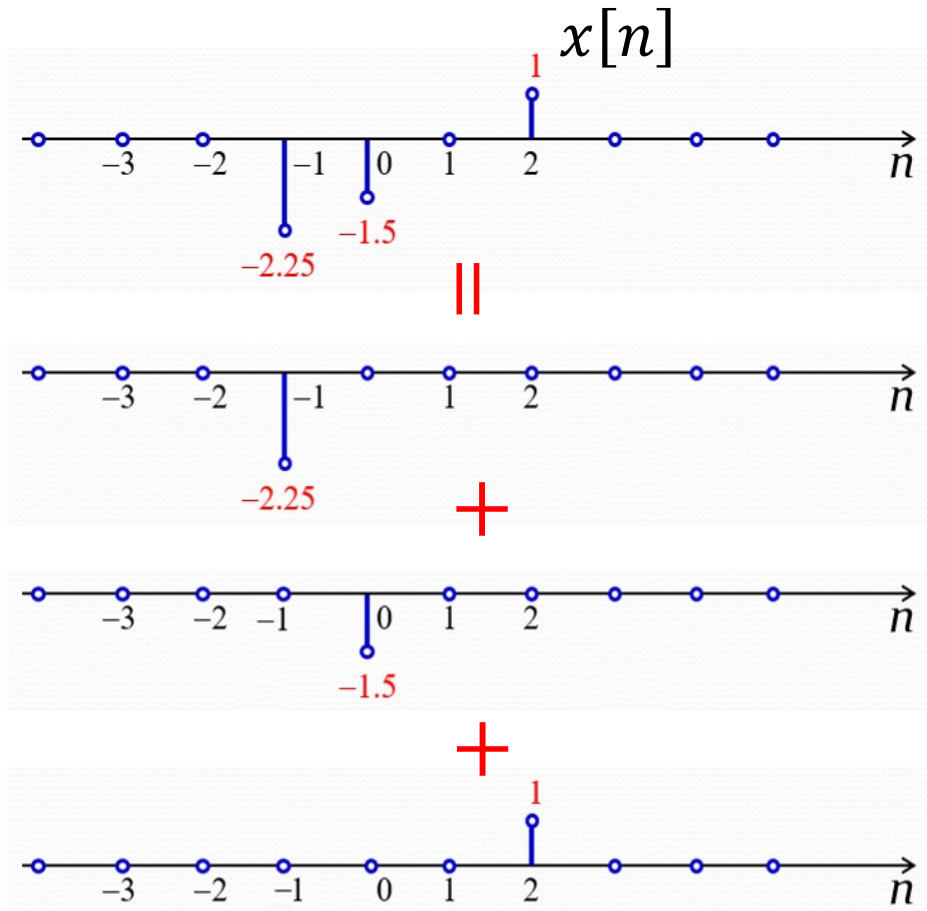
- The system has no internal way to keep time

Linear Time-Invariant Systems (LTI)

- LTI system: A system satisfying both the **linearity** and the **time-invariance** properties.
- Many physical processes possess the properties of linearity and time invariance.
- Highly useful signal processing algorithms have been developed utilizing this class of systems.
- LTI systems are mathematically easy to analyze and characterize, and consequently, easy to design. They possess **superposition** theorem.

If we represent the input to an LTI system in terms of **linear combination of a set of basic signals**, we can then use superposition to compute the output of the system in terms of responses to these basic signals.

Representation of Discrete-Time Signals in Terms of Impulse



$$x[n]$$

||

$$\Leftrightarrow x_1[n] = -2.25 \times \delta[n + 1]$$

+

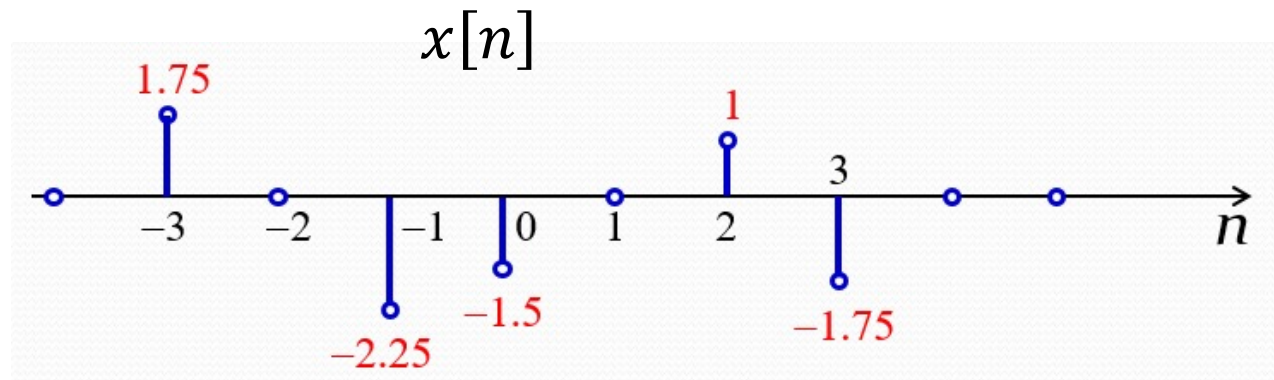
$$\Leftrightarrow x_2[n] = -1.5 \times \delta[n]$$

+

$$\Leftrightarrow x_3[n] = 1 \times \delta[n - 2]$$

Representation of Discrete-Time Signals in Terms of Impulse

- **An arbitrary sequence can be represented as the weighted sum of shifted unit impulses**



$$x[n] = 1.75\delta[n+3] - 2.25\delta[n+1] - 1.5\delta[n] + \delta[n-2] - 1.75\delta[n-3]$$

Representation of Discrete-Time Signals in Terms of Impulse

- Discrete-time unit impulse can be used to construct any discrete-time signal, because

$$x[k]\delta[n - k] = \begin{cases} x[k], & n = k \\ 0, & n \neq k \end{cases}$$

- **A general form**

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

Sifting property of $\delta[n]$

- Note: for any value of n , only one of the terms on the right-hand side (RHS) is nonzero.

Outline

- The Representation of Discrete-Time Signals in Terms of Impulses
- **The Discrete-Time Unit Impulse Response**
- The Convolution-Sum Representation of LTI Systems

Discrete-Time Unit Impulse Response

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



Discrete-Time Unit Impulse Response

- **How to calculate the impulse response of a system?**

- **Examples:** a system is defined as

$$y[n] = a_1 x[n] + a_2 x[n - 1] + a_3 x[n - 2] + a_4 x[n - 3]$$

its impulse response $h[n]$ is

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

Discrete-Time Unit Impulse Response

- **How to calculate the impulse response of a system?**

- **Examples:** a system is defined as

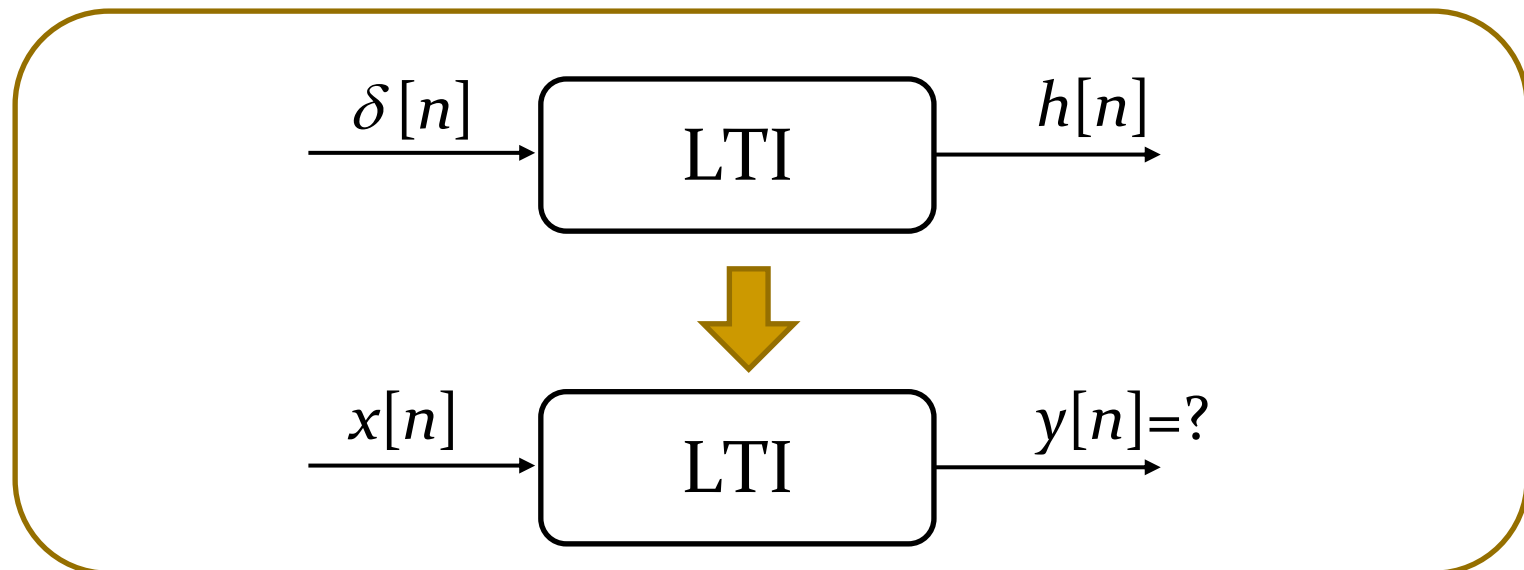
$$y[n] = \sum_{k=-\infty}^n x[k]$$

its impulse response $h[n]$ is

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

Discrete-Time Unit Impulse Response

- An LTI discrete system is **completely** characterized by its **impulse response**.
- In other words, knowing the impulse response one can compute the output of the LTI system for an arbitrary input.



Discrete-Time Unit Impulse Response

- Recall, an arbitrary input $x[n]$ can be expressed as a linear combination of shifted unit impulses

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

Sifting property of $\delta[n]$

- For any $k = k_0$

$\delta[n] \rightarrow$

LTI

 $\rightarrow h[n]$
- Time-invariance $\delta[n - k_0] \rightarrow$

LTI

 $\rightarrow h[n - k_0]$
- Scaling $x[k_0] \delta[n - k_0] \rightarrow$

LTI

 $\rightarrow x[k_0] h[n - k_0]$



Additivity

$$\begin{aligned}
 & x[n] \rightarrow \text{LTI} \rightarrow y[n] \\
 & = \sum_{k=-\infty}^{\infty} x[k] \delta[n - k] \quad = \sum_{k=-\infty}^{\infty} x[k] h[n - k]
 \end{aligned}$$

Outline

- The Representation of Discrete-Time Signals in Terms of Impulses
- The Discrete-Time Unit Impulse Response
- The Convolution-Sum Representation of LTI Systems

Discrete-Time Unit Impulse Response and the Convolution-Sum

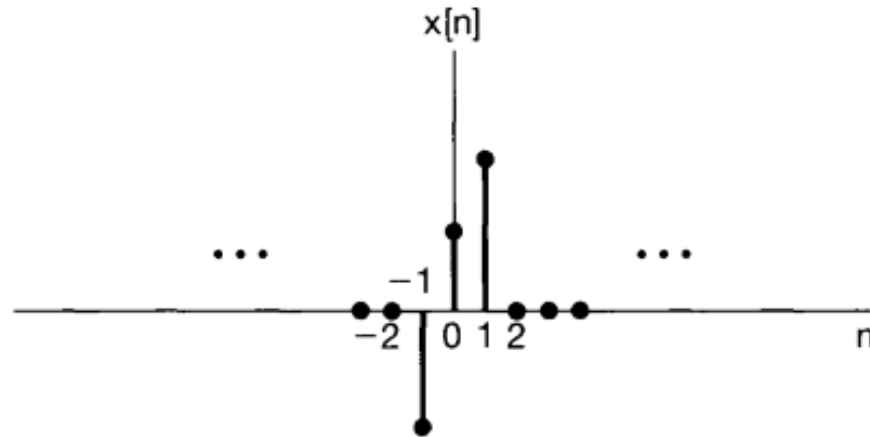
- Unit impulse response completely characterizes an LTI system.

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

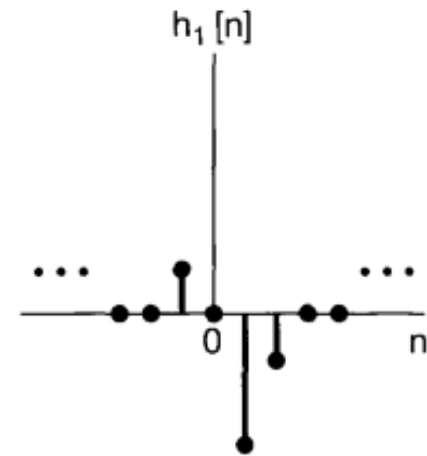
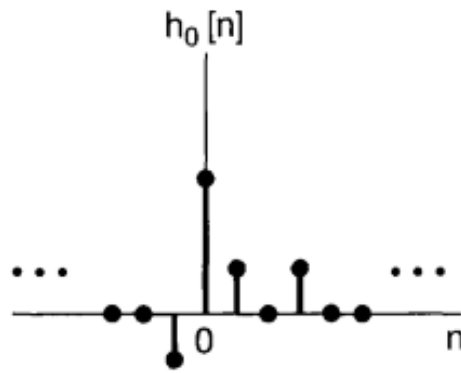
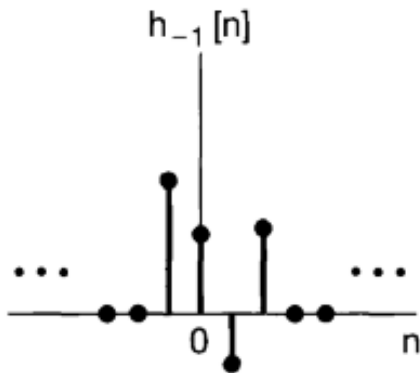
- $y[n] = \sum_{k=-\infty}^{\infty} x[k] h[n - k]$ is referred to as the **convolution-sum**, denoted by $y[n] = x[n] * h[n]$

$$x[n] \longrightarrow \text{LTI} \longrightarrow y[n] = x[n] * h[n]$$

Graphical Interpretation



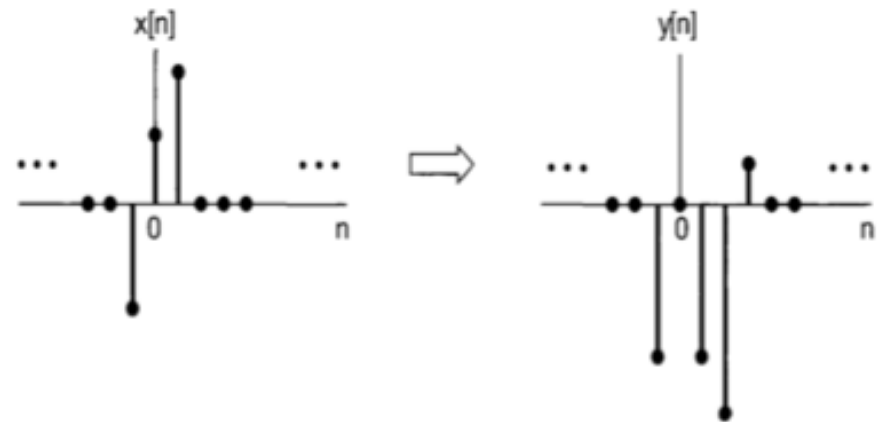
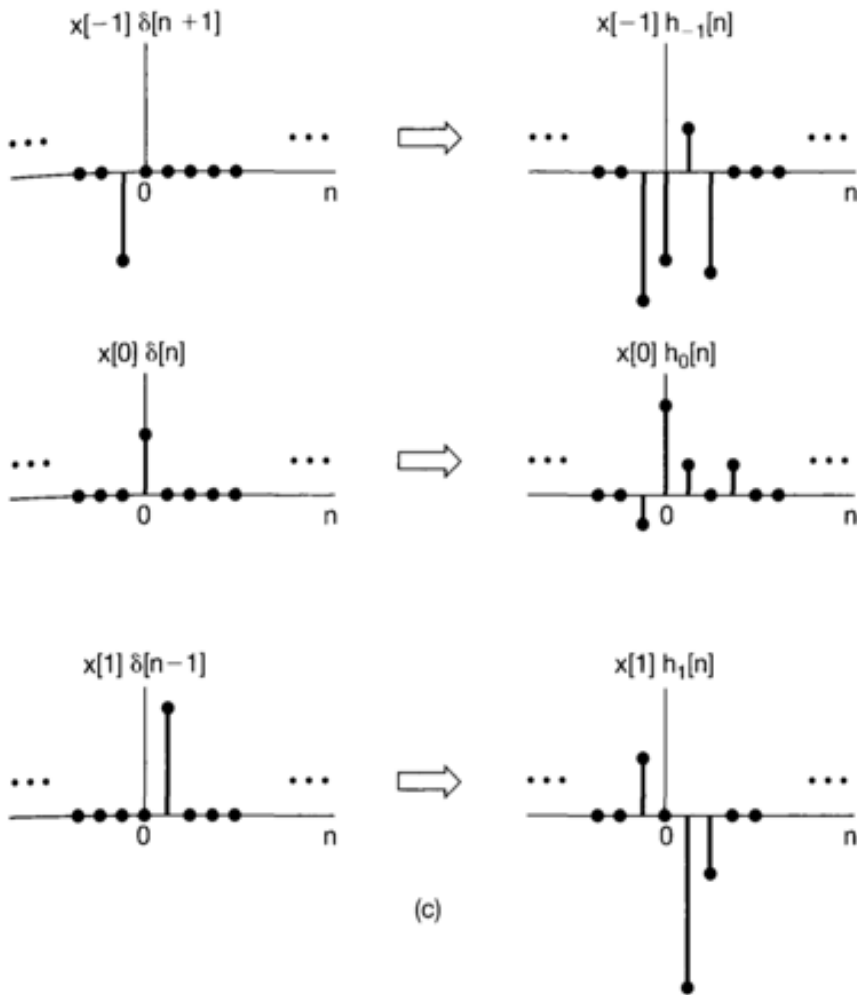
(a)



(b)

Graphical Interpretation

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

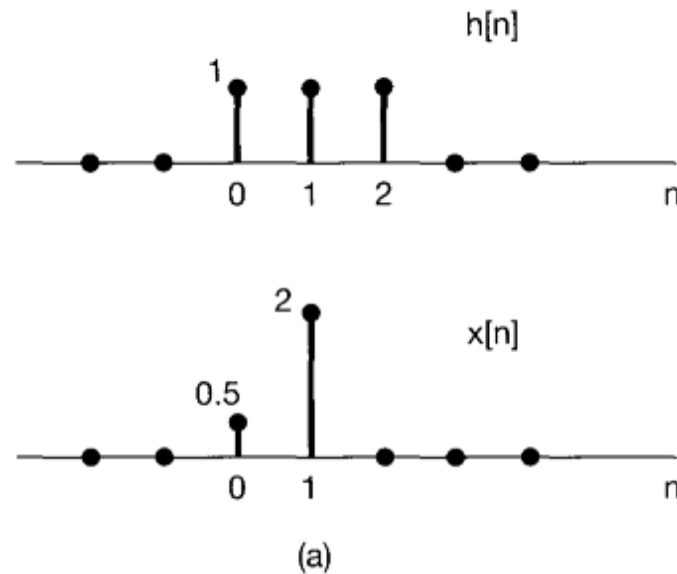


The Convolution-Sum

$$\begin{aligned} x[n] &\longrightarrow \boxed{\text{LTI}} \longrightarrow y[n] \\ &= \sum_{k=-\infty}^{\infty} x[k]h[n-k] \\ &= x[n] * h[n] \end{aligned}$$

How to calculate the convolution sum?

- Example 1: Consider an LTI system with impulse response $h[n]$ and input $x[n]$ given below, calculate $y[n]$.

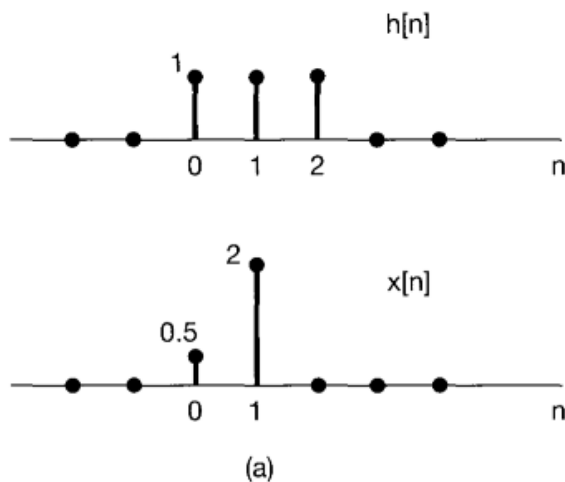


The Convolution-Sum

$$\begin{aligned} x[n] &\longrightarrow \boxed{\text{LTI}} \longrightarrow y[n] \\ &= \sum_{k=-\infty}^{\infty} x[k]h[n-k] \\ &= x[n] * h[n] \end{aligned}$$

How to calculate the convolution sum?

➤ **Method 1: View as functions of n (n : variable, k : constant)**



- **Step 1:** determine the range of k

$$k \in \{0, 1\}$$

- **Step 2:** determine the range of n

$$[n - k] \in \{0, 1, 2\} \leftrightarrow n \in \{0, 1, 2, 3\},$$

For other n , $y[n]=0$

- **Step 3:** calculate $y[n]$ for each n

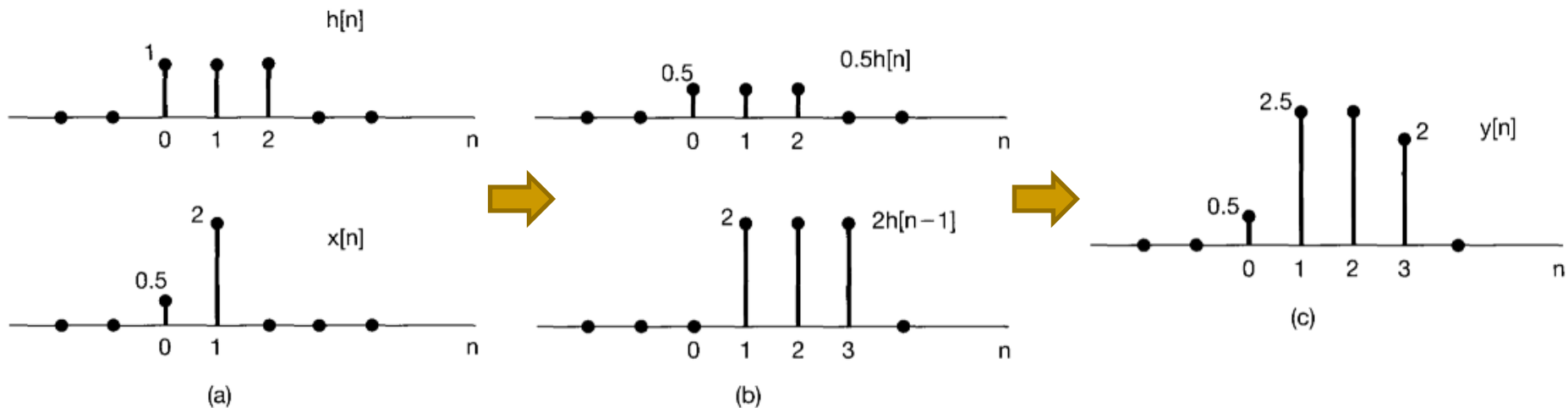
$$\begin{aligned} y[n] &= x[0]h[n-0] + x[1]h[n-1] \\ &= 0.5h[n] + 2h[n-1], \quad n \in \{0, 1, 2, 3\} \end{aligned}$$

The Convolution-Sum

How to calculate the convolution sum?

- Method 1: View as functions of n (n : variable, k : constant)

$$y[n] = x[0]h[n - 0] + x[1]h[n - 1] = 0.5h[n] + 2h[n - 1], \quad n \in \{0, 1, 2, 3\}$$



The Convolution-Sum

How to calculate the convolution sum?

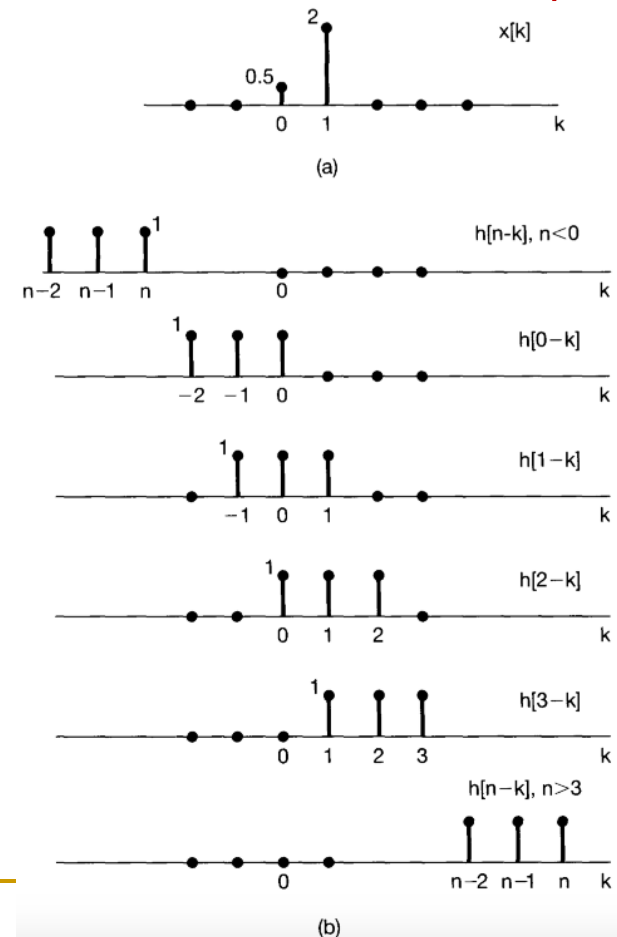
➤ Method 2: View as functions of k (n : constant, k : variable):

For each n :

- **Step 1:** change time variables
 $x[n] \rightarrow x[k]$, $h[n] \rightarrow h[k]$,
and reverse $h[k] \rightarrow h[-k]$
- **Step 2:** Shift $h[-k] \rightarrow h[n - k]$,
 n is considered as a constant
- **Step 3:** multiply $x[k] \cdot h[n - k]$
- **Step 4:** Summation

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

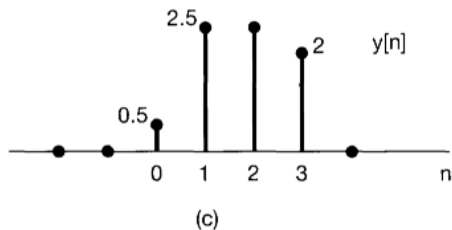
Change n , repeat step 1 to 4, calculate another $y[n]$



The Convolution-Sum

How to calculate the convolution sum?

- Method 2: View as functions of k (n : constant, k : variable):



If the lengths of the two sequences are M and N , then the sequence generated by the convolution is of length $M+N-1$

$$y[n] = 0, \text{ for } n < 0$$

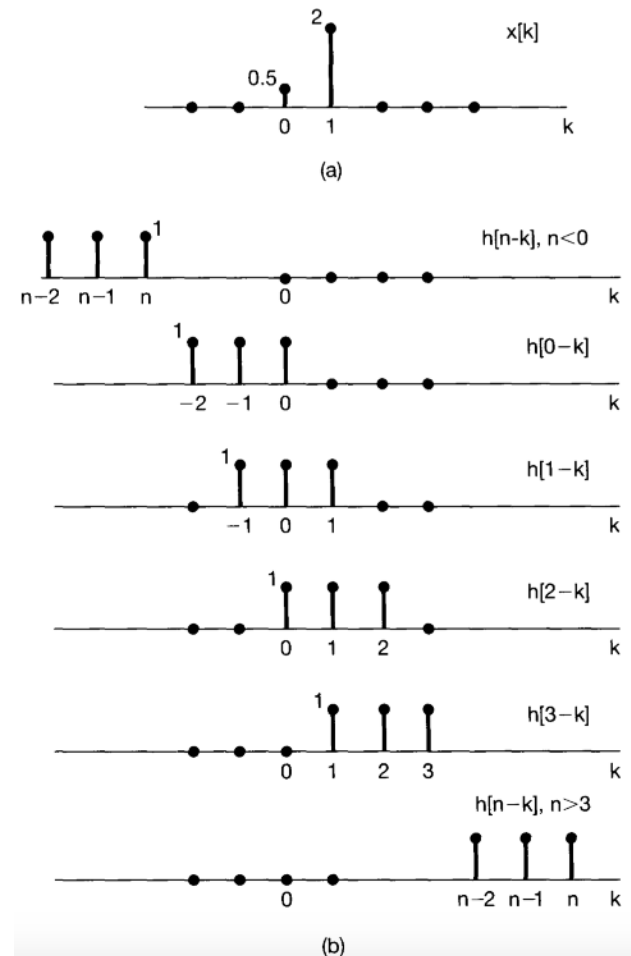
$$y[0] = \sum_{k=0}^1 x[k]h[0-k]$$

$$y[1] = \sum_{k=0}^1 x[k]h[1-k]$$

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

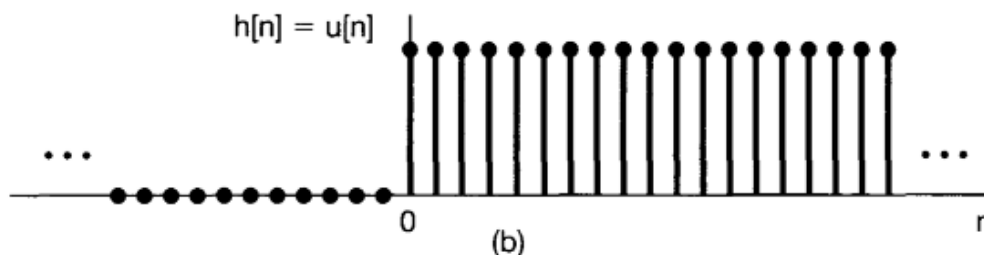
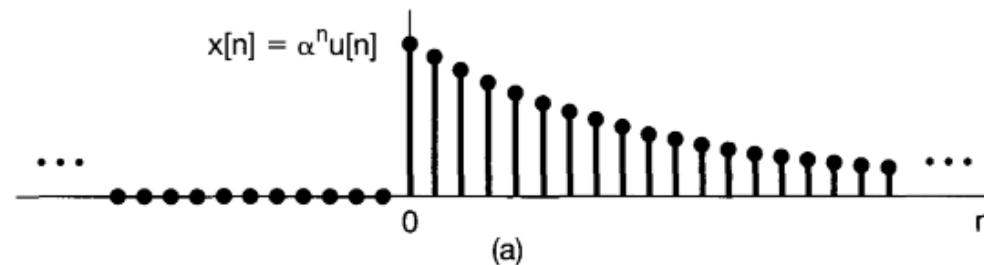
$$y[3] = \sum_{k=0}^1 x[k]h[3-k]$$

$$y[n] = 0, \text{ for } n > 3$$



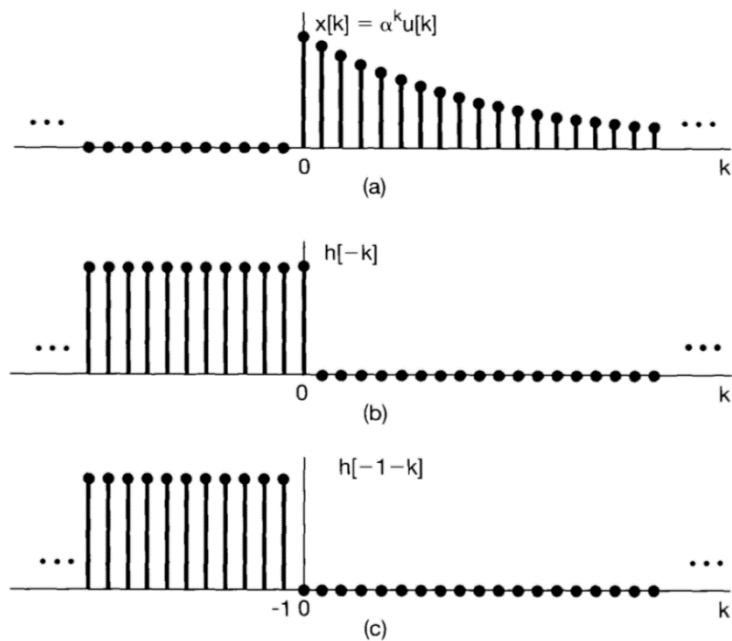
The Convolution-Sum

- Example 2: Consider an input $x[n]$ and a unit impulse response $h[n]$ given by $x[n] = \alpha^n u[n]$ and $h[n] = u[n]$ with $0 < \alpha < 1$.



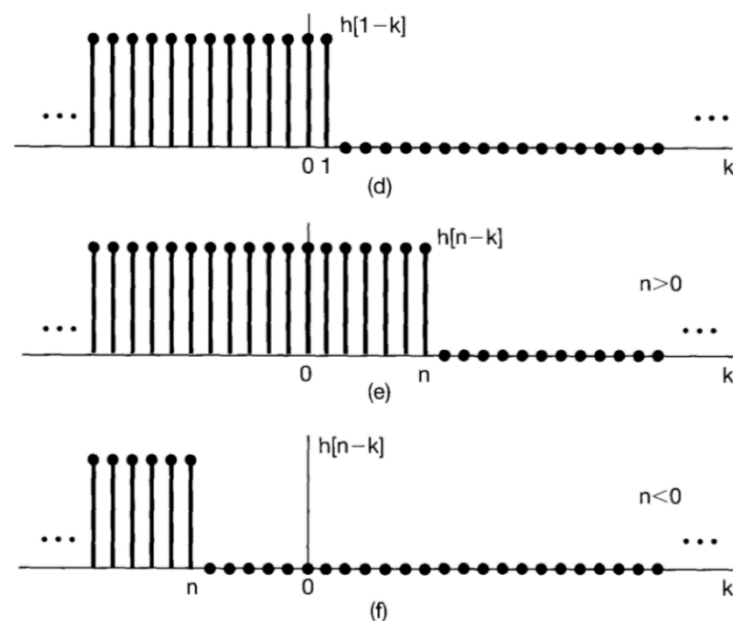
The Convolution-Sum

- Example 2: Consider an input $x[n]$ and a unit impulse response $h[n]$ given by $x[n] = \alpha^n u[n]$ and $h[n] = u[n]$ with $0 < \alpha < 1$.



For $n < 0$,

we have $x[k]h[n-k] = 0, \forall k$



For $n \geq 0$,

we have $x[k]h[n-k] = \begin{cases} \alpha^k, & 0 \leq k \leq n \\ 0, & \text{otherwise} \end{cases}$

The Convolution-Sum

- Example 2: Consider an input $x[n]$ and a unit impulse response $h[n]$ given by $x[n] = \alpha^n u[n]$ and $h[n] = u[n]$ with $0 < \alpha < 1$.

For $n < 0$,

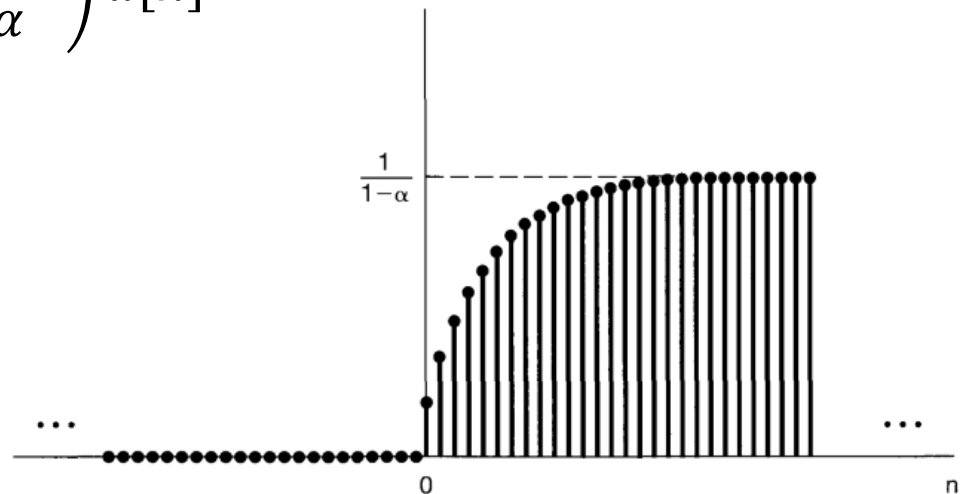
We have $y[n] = 0, n < 0$

Hence, we have

For $n \geq 0$,

we have $y[n] = \sum_{k=0}^n \alpha^k = \frac{1-\alpha^{n+1}}{1-\alpha}, n \geq 0$

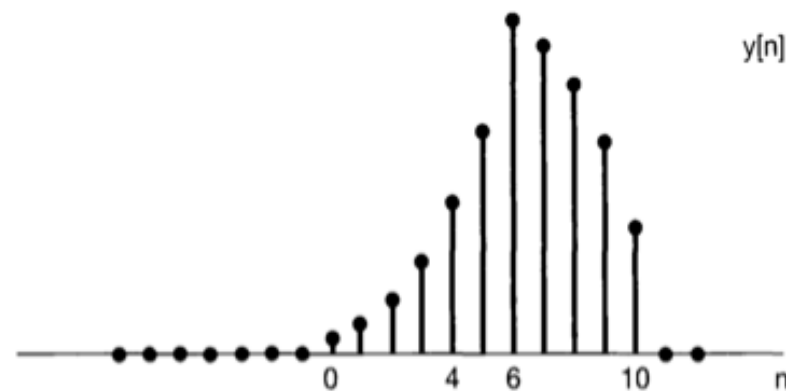
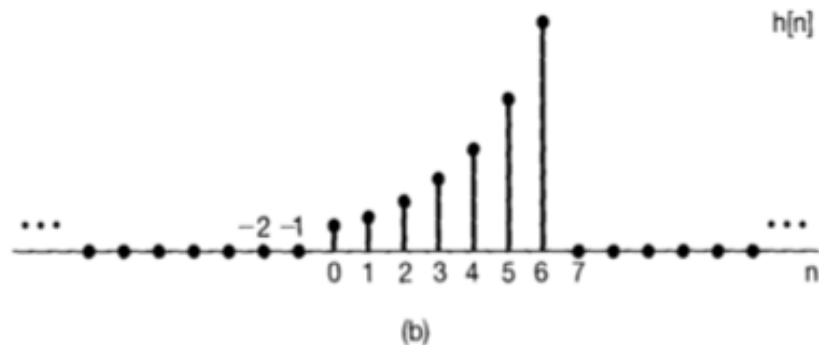
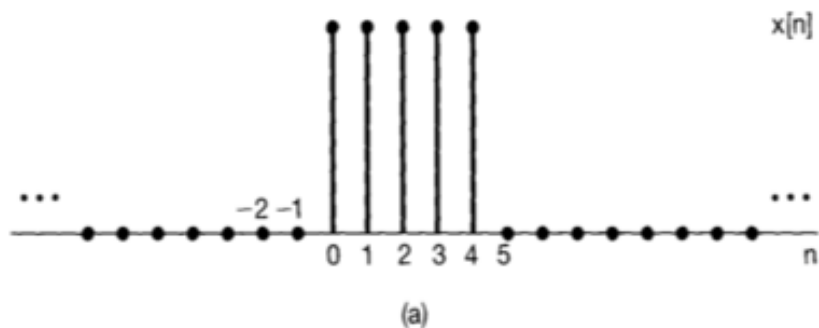
$$y[n] = \left(\frac{1 - \alpha^{n+1}}{1 - \alpha} \right) u[n]$$



The Convolution-Sum

- Example 3: Consider two sequences

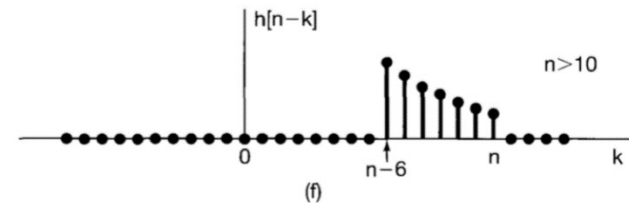
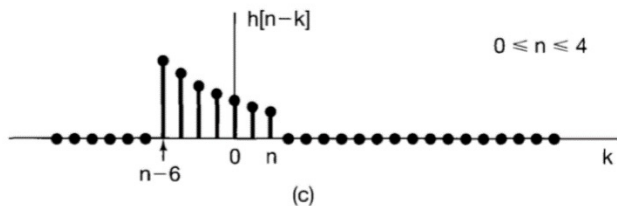
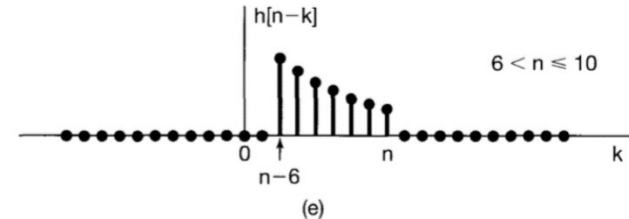
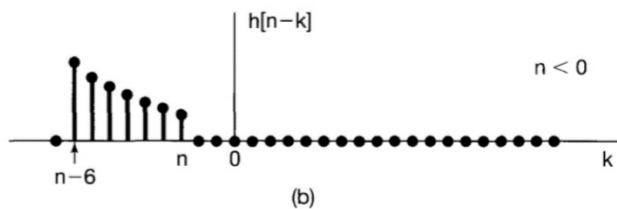
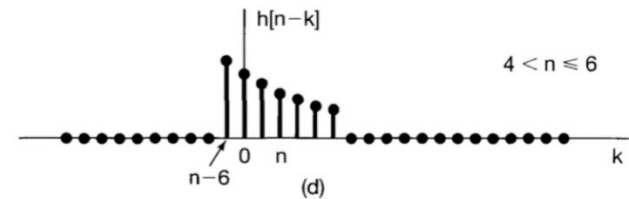
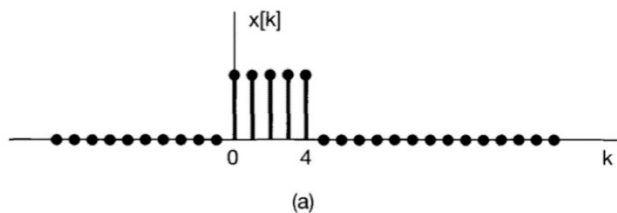
$$x[n] = \begin{cases} 1, & 0 \leq n \leq 4 \\ 0, & \text{otherwise} \end{cases} \quad \text{and} \quad h[n] = \begin{cases} \alpha^n, & 0 \leq n \leq 6 \\ 0, & \text{otherwise} \end{cases}$$



The Convolution-Sum

- Example 3: Consider two sequences

$$x[n] = \begin{cases} 1, & 0 \leq n \leq 4 \\ 0, & \text{otherwise} \end{cases} \quad \text{and} \quad h[n] = \begin{cases} \alpha^n, & 0 \leq n \leq 6 \\ 0, & \text{otherwise} \end{cases}$$



The Convolution-Sum

- Example 4: Consider an LTI system with input $x[n]$ and unit impulse response $h[n]$ specified as follows

$$x[n] = 2^n u[-n] \quad \text{and} \quad h[n] = u[n]$$

Summary

- The Representation of Discrete-Time Signals in Terms of Impulses
- The Discrete-Time Unit Impulse Response
- The Convolution-Sum Representation of LTI Systems
- Reference in textbook:
 - 2.1