

Ch.2 *Time Domain Representations of Linear Time-Invariant Systems (I)*

Prof. Shaoshi Yang

E-mail: shaoshi.yang@bupt.edu.cn

Web: <https://teacher.bupt.edu.cn/yangshaoshi>

School of Information & Communication Engineering

BUPT

Outline

- Linear Time-invariant systems (LTI)
 - Introduction
 - Discrete time LTI systems: Convolution Sum
 - Convolution Sum Evaluation Procedure

Linear Time-invariant systems (LTI)

- LTI system: A system satisfying both the **linearity** and the **time-invariance** property
- 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.
- Highly useful signal processing algorithms have been developed utilizing this class of systems.

Representation of LTI systems

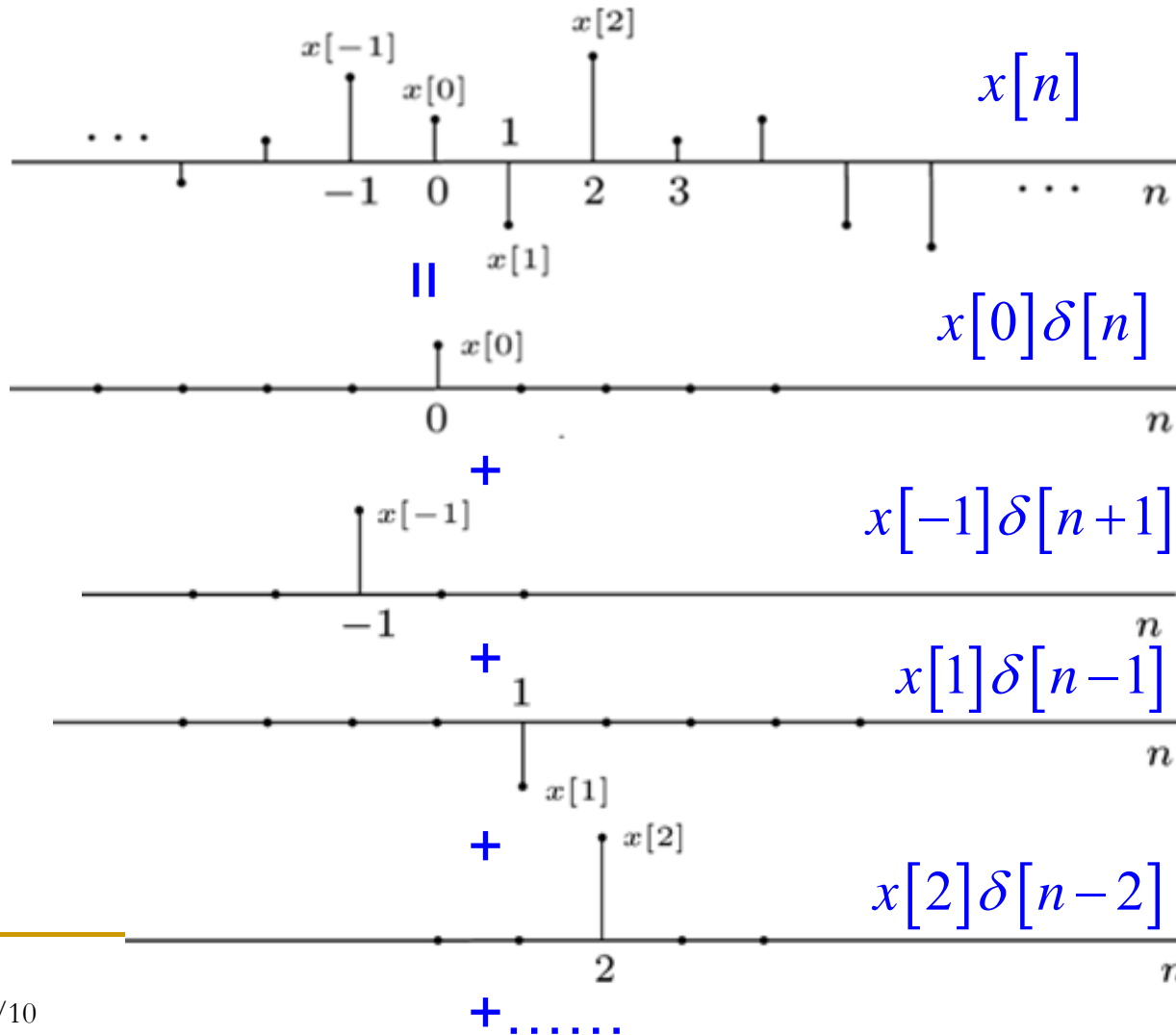
- Any linear time-invariant system (LTI) system, continuous-time or discrete-time, can be uniquely characterized by its
 - **Impulse response**(冲激响应): response of system to an impulse
 - **Frequency response**(频率响应): response of system to a complex exponential $e^{j2\pi ft}$ for all possible frequencies f .
 - **Transfer function**(传递函数): Laplace transform of impulse response
- Given one of the three, we can find other two provided that they exist.

Significance of unit impulse

- Every signal whether large or small can be represented in terms of **linear combination of delayed impulses**(延迟冲激).
- For DT or CT case, there are two natural choices for these two basic building blocks
 - For CT: shifted unit impulses **$\delta(t)$** .
 - For DT: shifted unit samples(impulse sequences) **$\delta[n]$** .

Superposition Sum for DT Systems

- Representing DT Signals with Sums of Unit Samples



Superposition Sum for DT Systems

- Representing DT Signals with a linear combination of delayed and advanced unit sample sequences

$$x[n] = \cdots + x[-2]\delta[n+2] + x[-1]\delta[n+1] + x[0]\delta[n] \\ + x[1]\delta[n-1] + x[2]\delta[n-2] + \cdots$$



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

Coefficients

Basic Signals

- $x[n]$: entire signal;
- $x[k]$: specific value of the signal $x[n]$ at time k .

Superposition Sum for DT Systems

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

- The response of the LTI system to an input will be

input

output

$$x[k] \delta[n-k] \quad \Longrightarrow \quad x[k] H \{ \delta[n-k] \} = x[k] h[n-k]$$

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

For time-invariant system:

$$H \{ \delta[n] \} = h[n] \quad \Longrightarrow \quad H \{ \delta[n-k] \} = h[n-k]$$

- The response of a linear system to $x[n]$ will be the superposition of the scaled responses $h[n]$ of the system to each of these shifted unit impulses.

Time-Domain Characterization of LTI DT System

- Knowing $h[n]$, one can compute the output of the system for any arbitrary input

- Ex: $x[n] = 0.5\delta[n+2] + 1.5\delta[n-1] - \delta[n-2] + 0.75\delta[n-5]$

- Since the system is time-invariant:

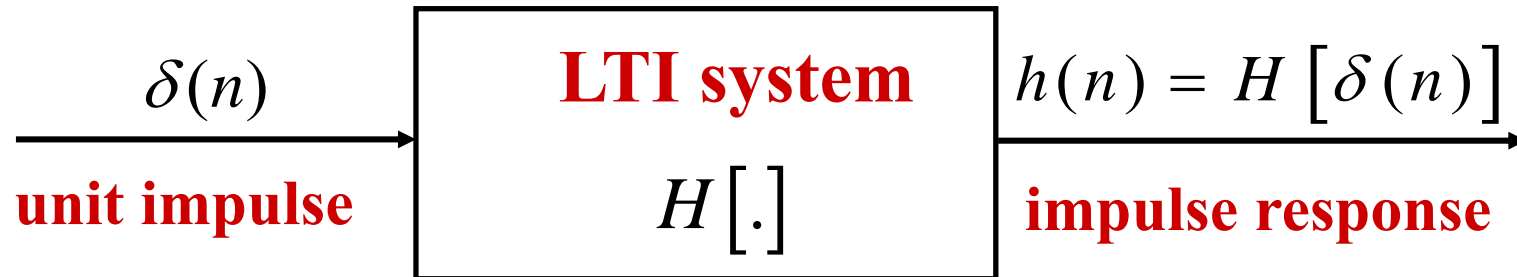
input	output
$\delta[n] \rightarrow h[n]$	$\delta[n+2] \rightarrow h[n+2]$
\Rightarrow	$\delta[n-1] \rightarrow h[n-1]$
	$\delta[n-2] \rightarrow h[n-2]$
	$\delta[n-5] \rightarrow h[n-5]$

- As the system is linear: $0.5\delta[n+2] \rightarrow 0.5h[n+2], \dots$

$$y[n] = 0.5h[n+2] + 1.5h[n-1] - h[n-2] + 0.75h[n-5]$$

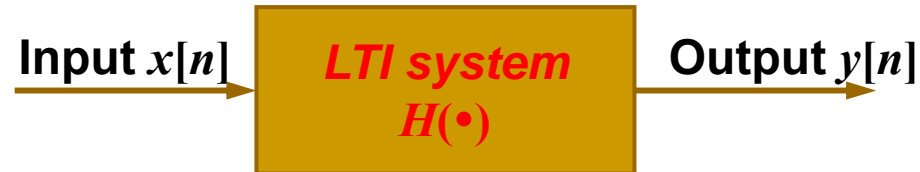
Impulse Response (冲激响应 / 脉冲响应)

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



- **Input-Output Relationship** - A consequence of the linear, time-invariance property is that an LTI discrete-time system is **completely characterized** by its impulse response.

Discrete time LTI systems: Convolution Sum



- Input: $x[n]$ as a weighted sum of time-shifted impulses

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

- Impulse response of LTI system: $H(\bullet)$ or $h[n]$

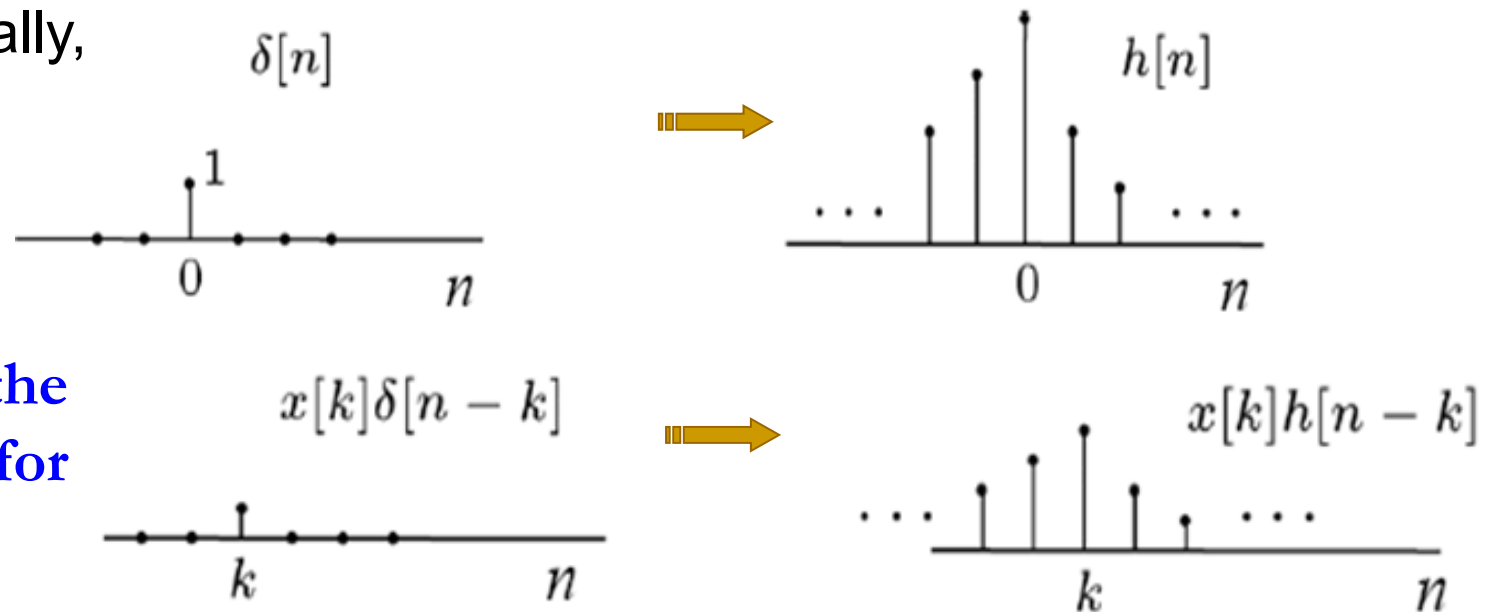
- Output:

$$\begin{aligned} y[n] &= H\{x[n]\} = H\left\{\sum_{k=-\infty}^{\infty} x[k] \delta[n-k]\right\} = \sum_{k=-\infty}^{\infty} H\{x[k] \delta[n-k]\} \\ &= \sum_{k=-\infty}^{\infty} x[k] H\{\delta[n-k]\} = \sum_{k=-\infty}^{\infty} x[k] h[n-k] \end{aligned}$$

Convolution Sum (卷积和)

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

- **Convolution sum:** the output of an LTI system is a weighted sum of the response of the system to time-shifted impulses. The sum is denoted by the symbol $*$.
- Graphically,



Sum up all the responses for all k 's.

Discrete time LTI systems: Convolution Sum

- Illustration of the convolution sum

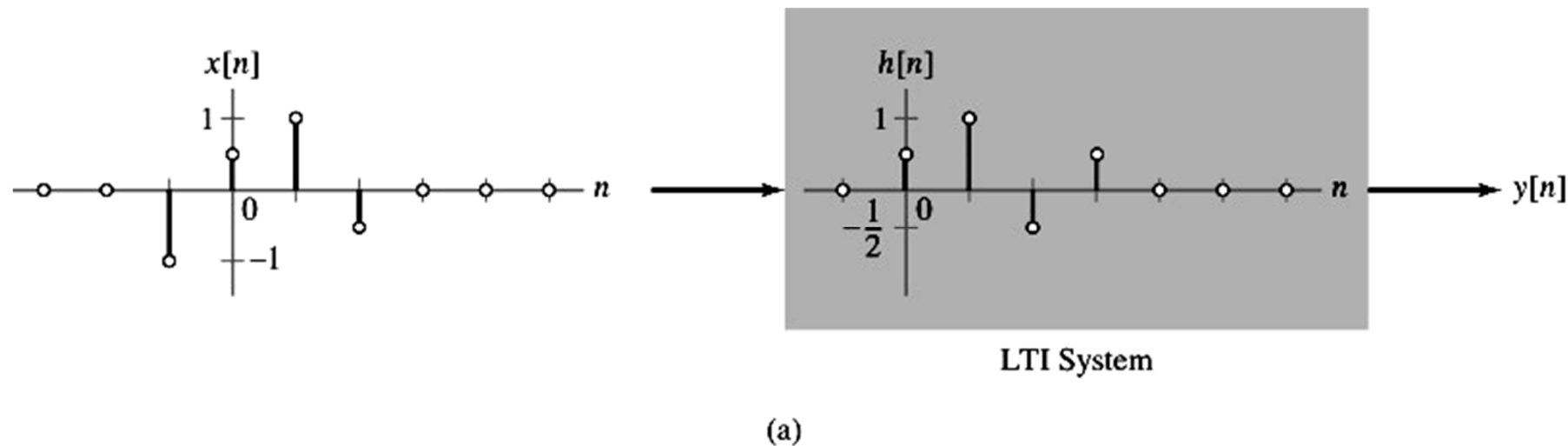
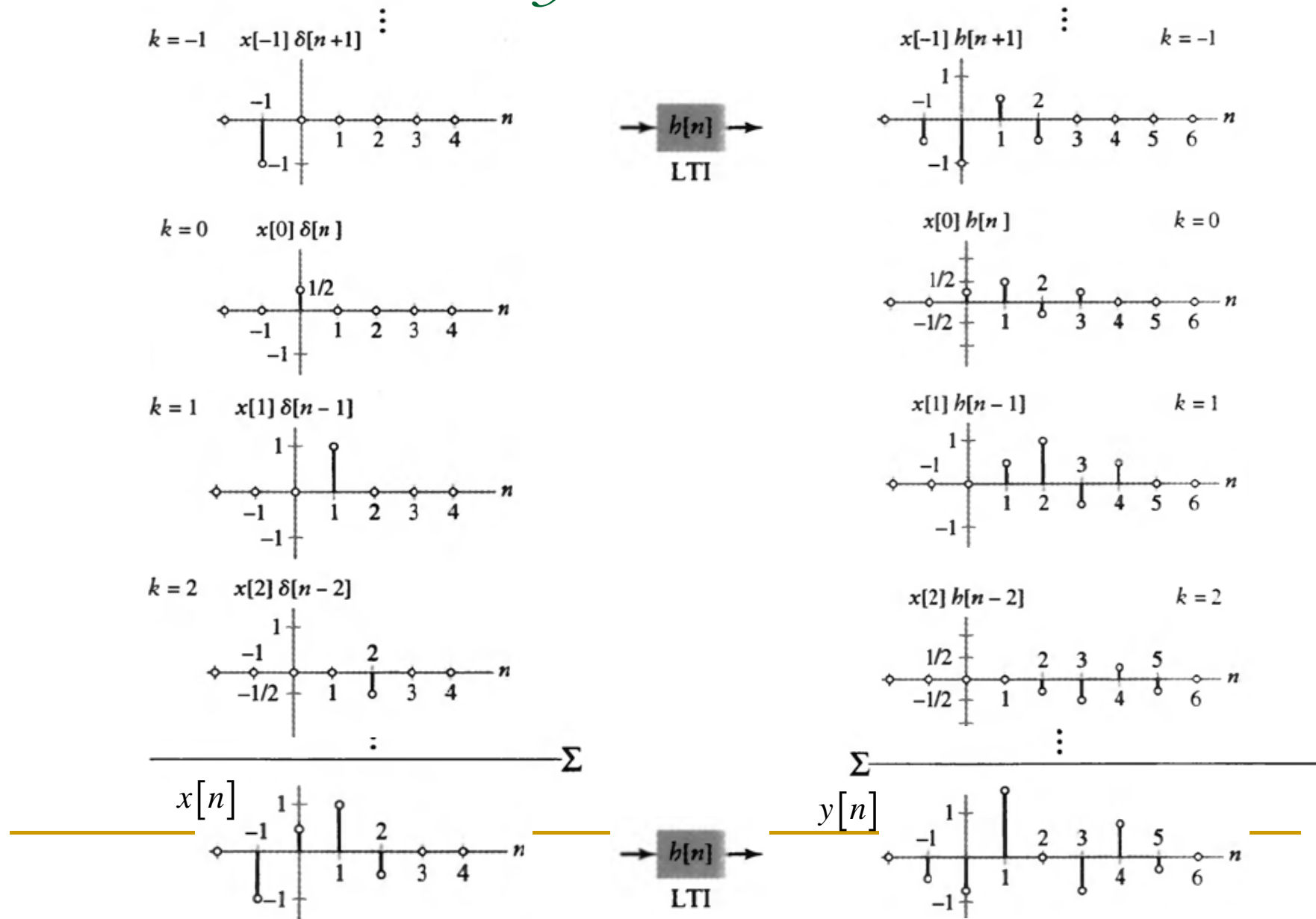


Figure 2.2a Illustration of the convolution sum. (a) LTI system with impulse response $h[n]$ and input $x[n]$.

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

Discrete time LTI systems: Convolution Sum



Discrete time LTI systems: Convolution Sum

Example 2.1 Multipath Communication Channel: Direct Evaluation of the Convolution Sum

Consider the discrete-time LTI system model representing a two-path propagation channel. If the strength of the indirect path is $a = 1/2$, then

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

Letting $x[n] = \delta[n]$, we find the impulse response is

$$h[n] = \begin{cases} 1, & n = 0 \\ \frac{1}{2}, & n = 1 \\ 0, & \text{otherwise} \end{cases}$$

Determine the output of this system in response to the input

$$x[n] = \begin{cases} 2, & n = 0 \\ 4, & n = 1 \\ -2, & n = 2 \\ 0, & \text{otherwise} \end{cases}$$

Discrete time LTI systems: Convolution Sum

<Sol.>

□ input: $x[n] = \begin{cases} 2, & n = 0 \\ 4, & n = 1 \\ -2, & n = 2 \\ 0, & \text{otherwise} \end{cases}$

$$= 2\delta[n] + 4\delta[n-1] - 2\delta[n-2]$$

□ system: $h[n] = \begin{cases} 1, & n = 0 \\ \frac{1}{2}, & n = 1 \\ 0, & \text{otherwise} \end{cases}$

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

□ output: $y[n] = 2h[n] + 4h[n-1] - 2h[n-2]$

$$= 2\left\{\delta[n] + \frac{1}{2}\delta[n-1]\right\} + 4\left\{\delta[n-1] + \frac{1}{2}\delta[n-2]\right\} - 2\left\{\delta[n-2] + \frac{1}{2}\delta[n-3]\right\}$$

$$= 2\delta[n] + 5\delta[n-1] - \delta[n-3] = \begin{cases} 0, & n < 0 \\ 2, & n = 0 \\ 5, & n = 1 \\ 0, & n = 2 \\ -1, & n = 3 \\ 0, & n \geq 4 \end{cases}$$

Convolution Sum Evaluation Procedure


- An alternative approach to evaluate the convolution sum

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

- Define the intermediate signal:

$$w_n[k] = x[k]h[n-k] \quad \text{where } k = \text{independent variable} \\ n = \text{constant.}$$

where $h[n-k] = h[-(k-n)]$
~ a reflected and time-shifted version of $h[k]$.


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

The time shift n determines the time at which we evaluate the output of the system.

Convolution Sum Evaluation Procedure

Example 2.2 Convolution Sum Evaluation by using Intermediate Signal

Consider a system with impulse response $h[n] = \left(\frac{3}{4}\right)^n u[n]$

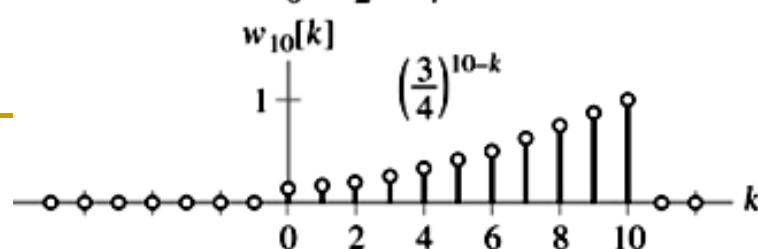
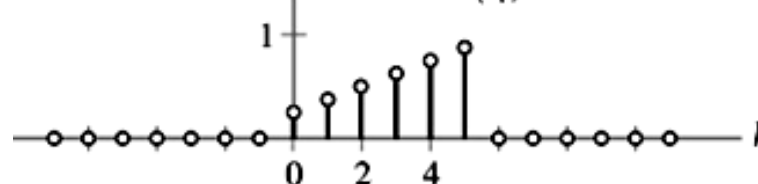
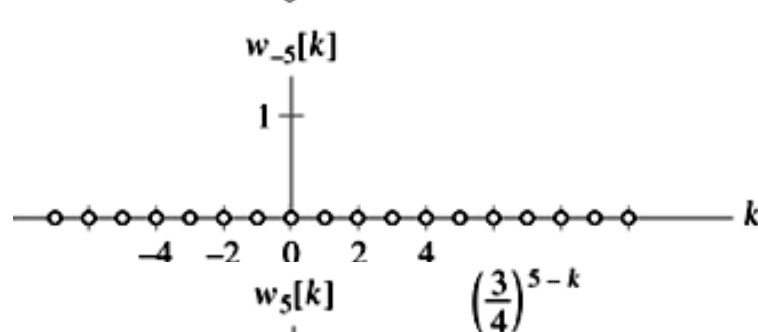
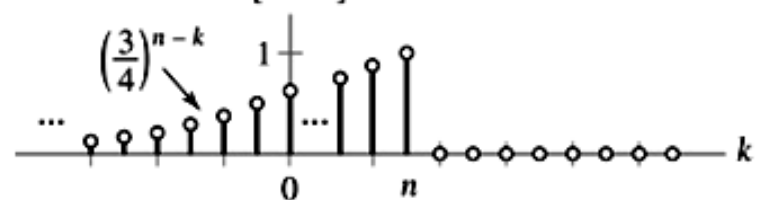
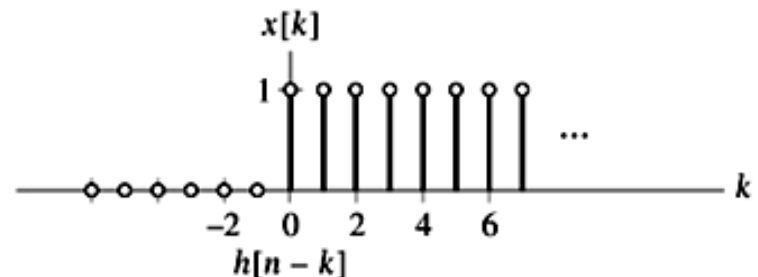
determine the output at time $n = -5, 5, 10$ when the input is $x[n] = u[n]$.

□ intermediate signal: $w_n[k] = x[k]h[n-k]$

$$h[n-k] = \left(\frac{3}{4}\right)^{n-k} u(n-k) = \begin{cases} \left(\frac{3}{4}\right)^{n-k}, & k \leq n \\ 0, & \text{otherwise} \end{cases} \quad \Rightarrow$$

$$w_n[k] = u(k) \left(\frac{3}{4}\right)^{n-k} u(n-k) = \begin{cases} \left(\frac{3}{4}\right)^{n-k}, & 0 \leq k \leq n \\ 0, & \text{otherwise} \end{cases}$$

Convolution Sum Evaluation Procedure



□ $n=-5: w_{-5}[k] = 0$

$$y[-5] = \sum_{k=-\infty}^{\infty} w_{-5}[k] = 0$$

□ $n=5: w_5[k] = \left(\frac{3}{4}\right)^{5-k}, \quad 0 \leq k \leq 5$

$$y[5] = \sum_{k=0}^5 w_5[k] = \left(\frac{3}{4}\right)^5 \sum_{k=0}^5 \left(\frac{4}{3}\right)^k$$

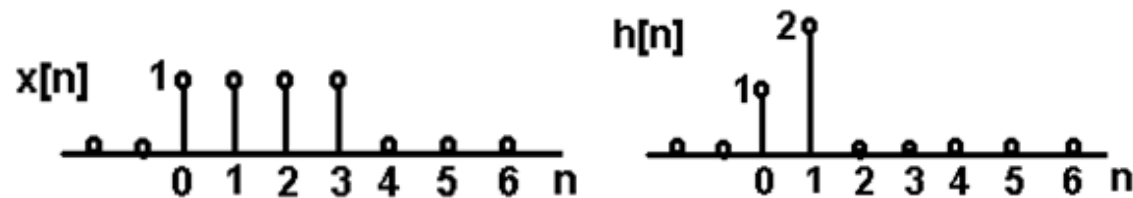
$$= \left(\frac{3}{4}\right)^5 \frac{1 - \left(\frac{4}{3}\right)^6}{1 - \left(\frac{4}{3}\right)} = 3.288$$

□ $n=10: w_{10}[k] = \left(\frac{3}{4}\right)^{10-k}, \quad 0 \leq k \leq 10$

$$y[10] = \sum_{k=0}^{10} w_{10}[k] = 3.831$$

Convolution Sum Evaluation Procedure

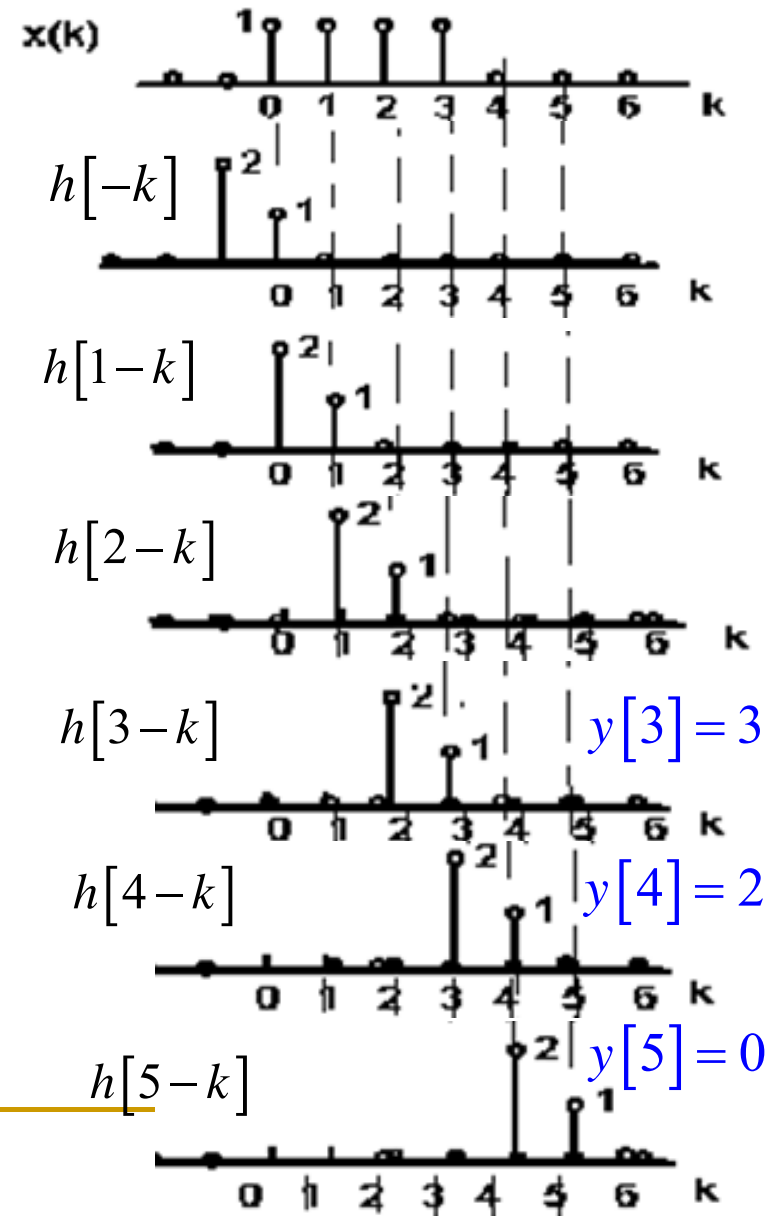
Ex. Develop the sequence $y[n]$ generated by the convolution of $x[n]$ and $h[n]$.



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

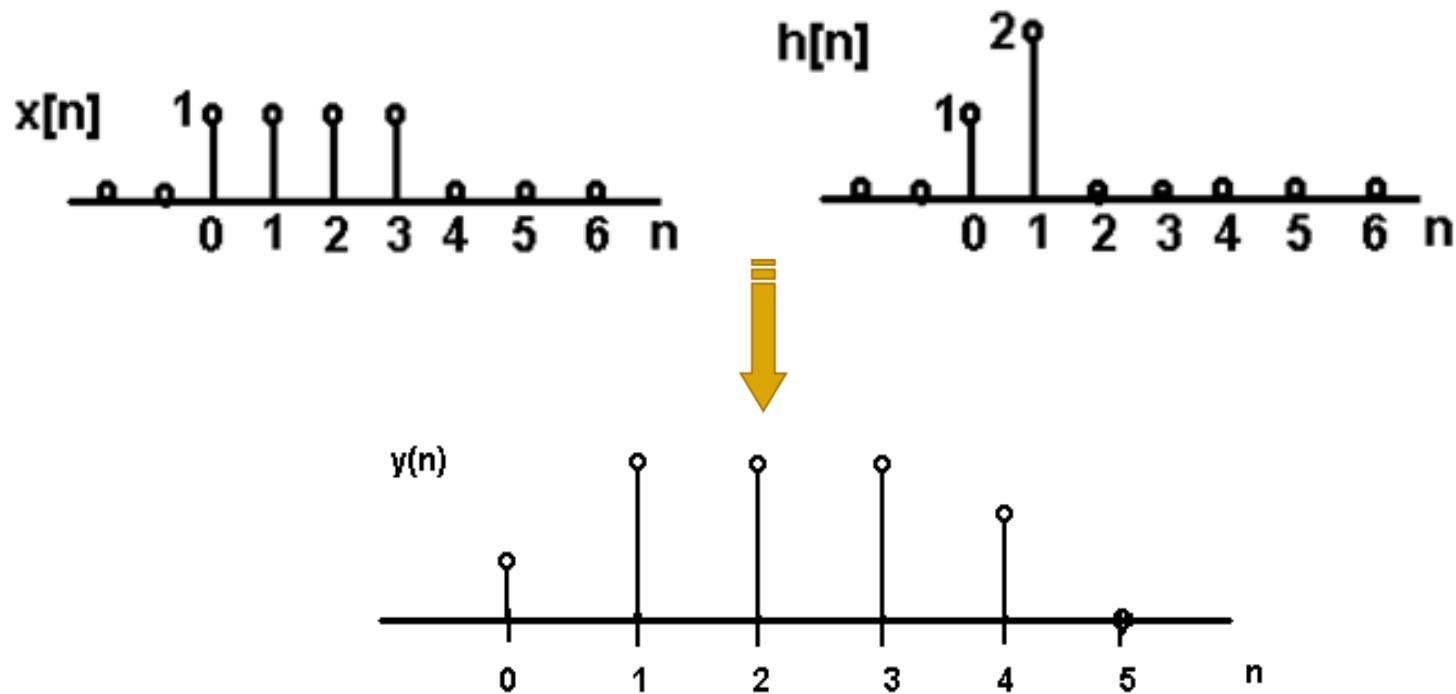
$$n = 1: y[1] = \sum_{k=-\infty}^{\infty} x[k]h[1-k] \\ = x[0]h[1] + x[1]h[0] = 3$$

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



Convolution Sum Evaluation Procedure

Ex. Develop the sequence $y[n]$ generated by the convolution of $x[n]$ and $h[n]$.



The length of $x[n]$: M

The length of $h[n]$: N



The length of $y[n]$: $M + N - 1$

Convolution Sum Evaluation Procedure

■ Procedure 2.1: Reflect and Shift Convolution Sum Evaluation

- Graph both $x[k]$ and $h[n-k]$ as a function of the independent variable k . To determine $h[n-k]$, first reflect $h[k]$ about $k=0$ to obtain $h[-k]$. Then shift by $-n$.
- Begin with n large and negative. That is, shift $h[-k]$ to the far left on the time axis.
- Write the mathematical representation for the intermediate signal $w_n[k]$.
- Increase the shift n (i.e., move $h[n-k]$ toward the right) until the mathematical representation for $w_n[k]$ changes. The value of n at which the change occurs defines the end of the current interval and the beginning of a new interval
- Let n be in the new interval. Repeat step 3 and 4 until all intervals of times shifts and the corresponding mathematical representations for $w_n[k]$ are identified. This usually implies increasing n to a very large positive number.
- For each interval of time shifts, sum all the values of the corresponding $w_n[k]$ to obtain $y[n]$ on that interval.

Convolution Sum Evaluation Procedure

Example 2.3 Moving-Average System

The output $y[n]$ of the four-point moving-average system is related to the input $x[n]$ according to the formula

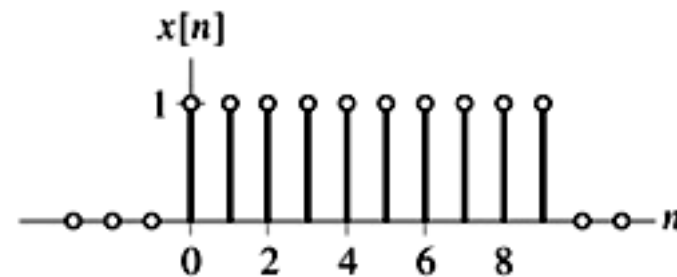
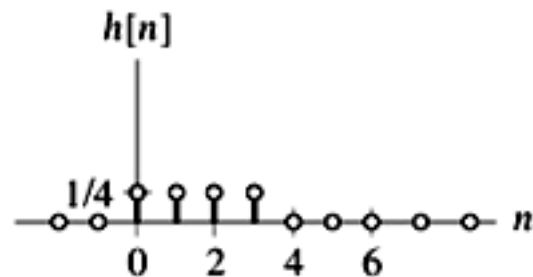
$$y[n] = \frac{1}{4} \sum_{k=0}^3 x[n-k]$$

The impulse response $h[n]$ of this system is obtained by letting $x[n] = \delta[n]$, which yields

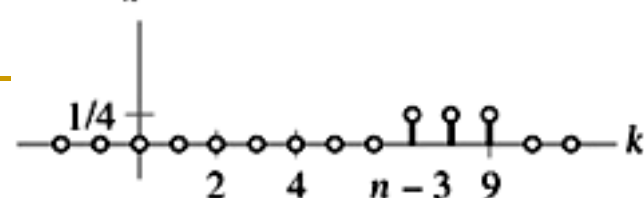
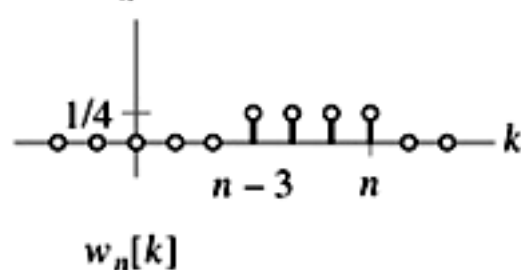
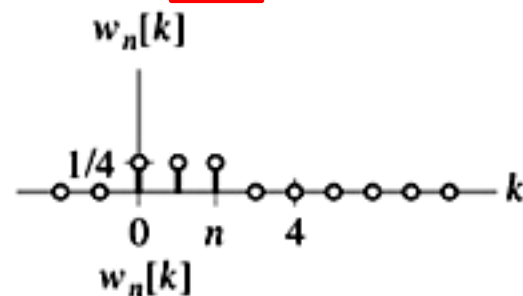
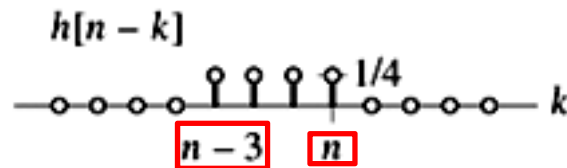
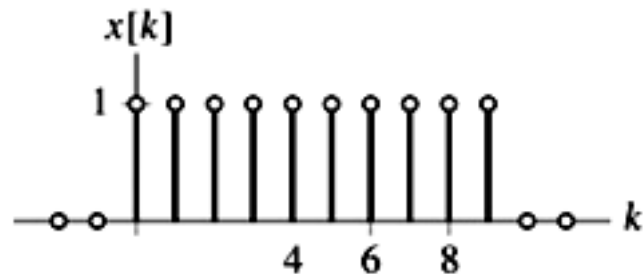
$$h[n] = \frac{1}{4} (\delta[n] + \delta[n-1] + \delta[n-2] + \delta[n-3]) = \frac{1}{4} (u[n] - u[n-4])$$

Determine the output of the system when the input is the rectangular pulse defined as

$$x[n] = u[n] - u[n-10]$$



Convolution Sum Evaluation Procedure



■ Five intervals

□ $n < 0$: $w_n[k] = 0$ $y[n] = 0$

□ $0 \leq n \leq 3$: $w_n[k] = 1/4$, $0 \leq k \leq n$

$$y[n] = \sum_{k=0}^n 1/4 = \frac{n+1}{4}$$

□ $3 < n \leq 9$: $w_n[k] = 1/4$, $n-3 \leq k \leq n$

$$y[n] = \sum_{k=n-3}^n 1/4 = 1$$

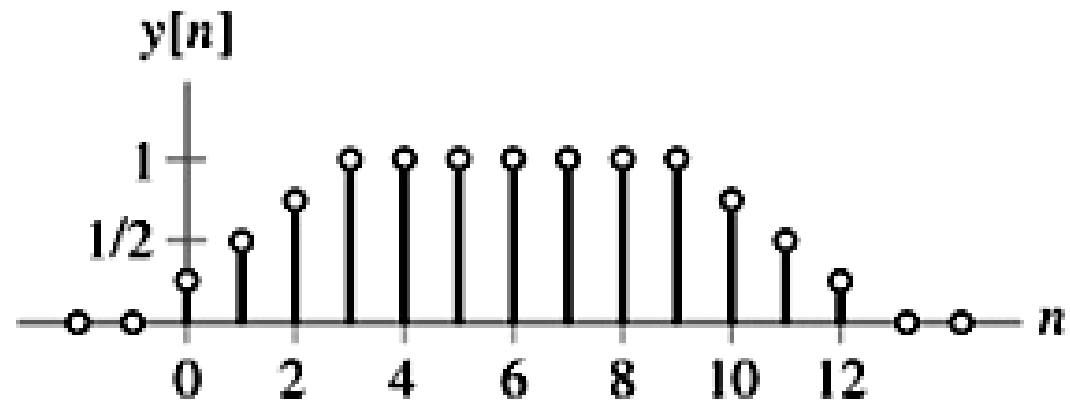
□ $9 < n \leq 12$: $w_n[k] = 1/4$, $n-3 \leq k \leq 9$

$$y[n] = \sum_{k=n-3}^9 1/4 = \frac{13-n}{4}$$

□ $n > 12$: $w_n[k] = 0$ $y[n] = 0$

Convolution Sum Evaluation Procedure

$$y[n] = \begin{cases} 0, & n < 0 \\ \frac{n+1}{4}, & 0 \leq n \leq 3 \\ 1, & 3 < n \leq 9 \\ \frac{13-n}{4}, & 9 < n \leq 12 \\ 0, & n > 12 \end{cases}$$



Convolution Sum Evaluation Procedure

Example 2.4 First-order Recursive System:

The input-output relationship for the first-order recursive system is given by

$$y[n] - \rho y[n-1] = x[n]$$

Let the input be given by $x[n] = b^n u[n+4]$

We use convolution to find the output of this system, assuming that $b \neq \rho$ and that the system is causal.

<Sol.>

- Impulse response: $h[n] = \rho h[n-1] + \delta[n]$

Since the system is causal, we have $h[n] = 0$ for $n < 0$.

$$h[0] = \delta[0] = 1, \quad h[1] = \rho, \quad h[2] = \rho^2, \dots, \quad \Rightarrow \quad h[n] = \rho^n u[n]$$

- Graph both $x[k]$ and $h[n-k]$ as a function of k .

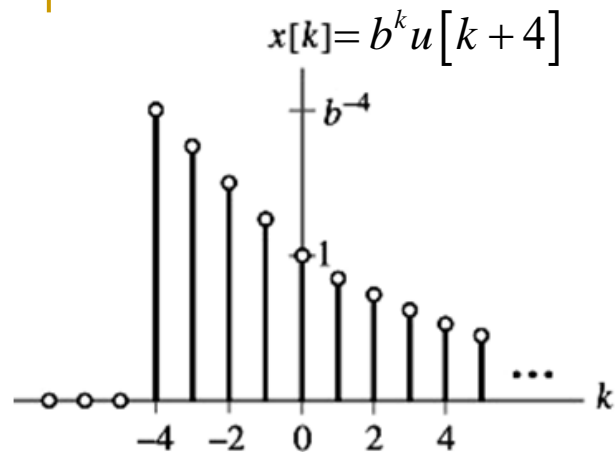
$$x[k] = b^k u[k+4]$$

$$= \begin{cases} b^k, & k \geq -4 \\ 0, & \text{otherwise} \end{cases}$$

$$h[n-k] = \rho^{n-k} u[n-k]$$

$$= \begin{cases} \rho^{n-k}, & k \leq n \\ 0, & \text{otherwise} \end{cases}$$

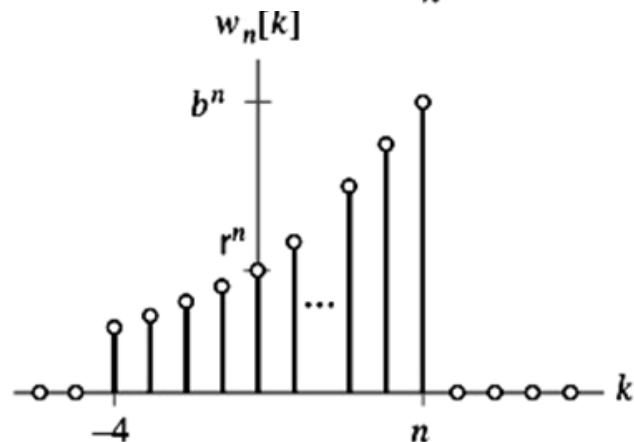
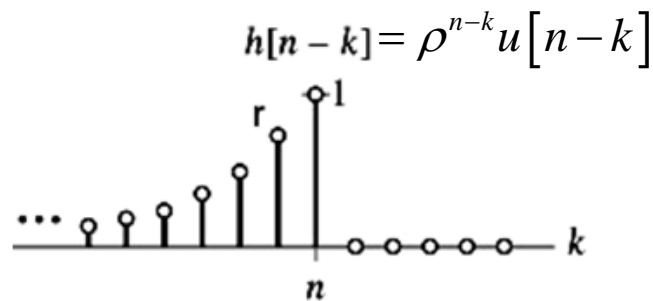
Convolution Sum Evaluation Procedure



Two intervals

■ $n < -4$: $w_n[k] = 0$ $y[n] = 0$

■ $n \geq -4$: $w_n[k] = \begin{cases} b^k \rho^{n-k}, & -4 \leq k \leq n \\ 0, & \text{otherwise} \end{cases}$



$$y[n] = \sum_{k=-4}^n b^k \rho^{n-k} = \rho^n \sum_{k=-4}^n \left(\frac{b}{\rho} \right)^k$$

$$= \rho^n \left(\frac{\rho}{b} \right)^4 \sum_{m=0}^{n+4} \left(\frac{b}{\rho} \right)^m$$

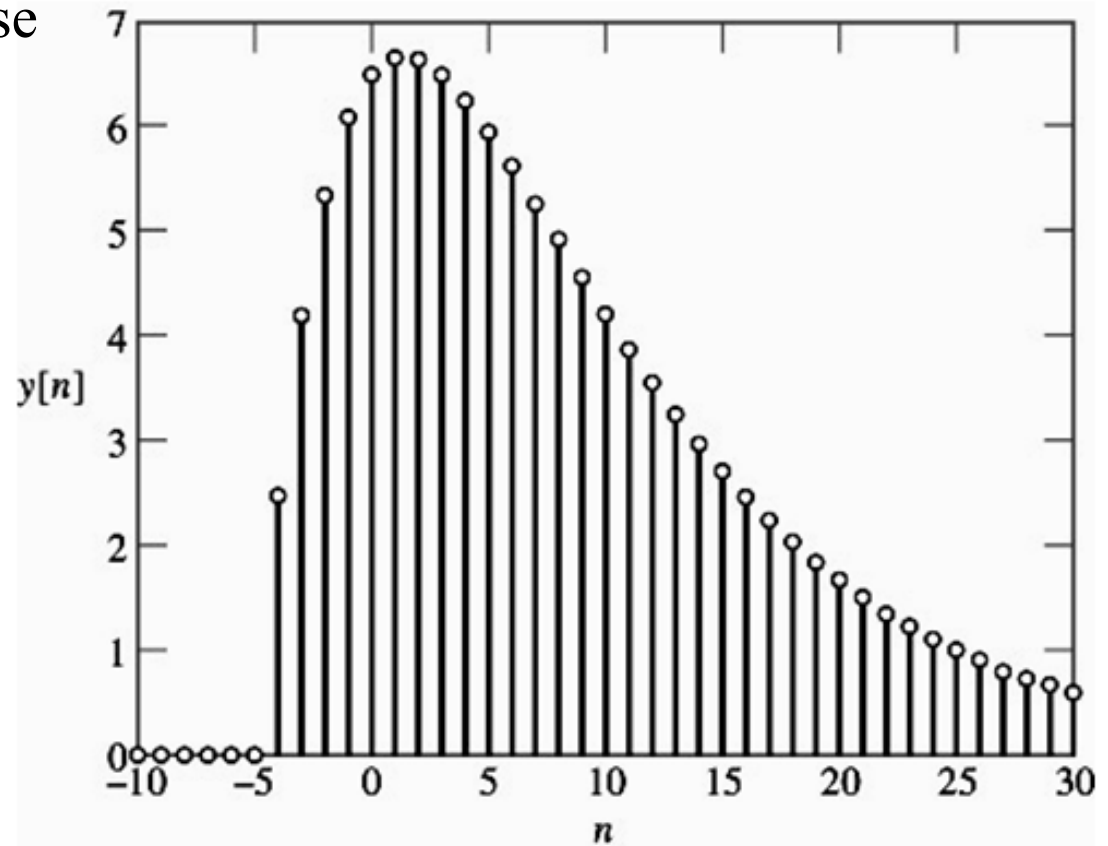
$$= \rho^n \left(\frac{\rho}{b} \right)^4 \frac{1 - (b/\rho)^{n+5}}{1 - (b/\rho)}$$

$$= b^{-4} \left(\frac{\rho^{n+5} - b^{n+5}}{\rho - b} \right)$$

Convolution Sum Evaluation Procedure

$$y[n] = \begin{cases} b^{-4} \left(\frac{\rho^{n+5} - b^{n+5}}{\rho - b} \right), & n \geq -4 \\ 0, & \text{otherwise} \end{cases}$$

for $\rho = 0.9$, $b = 0.8$



Convolution Sum Evaluation Procedure

Example 2.5 Investment Computation

The first-order recursive system is used to describe the value of an investment earning compound interest at a fixed rate of $r\%$ per period if we set $\rho = 1+(r/100)$. Let $y[n]$ be the value of the investment at the start of period n . If there are no deposits or withdrawals, then the value at time n is expressed in terms of the value at the previous time as $y[n] = \rho y[n - 1]$. Now, suppose $x[n]$ is the amount deposited ($x[n] > 0$) or withdrawn ($x[n] < 0$) at the start of period n . In this case, the value of the amount is expressed by the first-order recursive equation

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

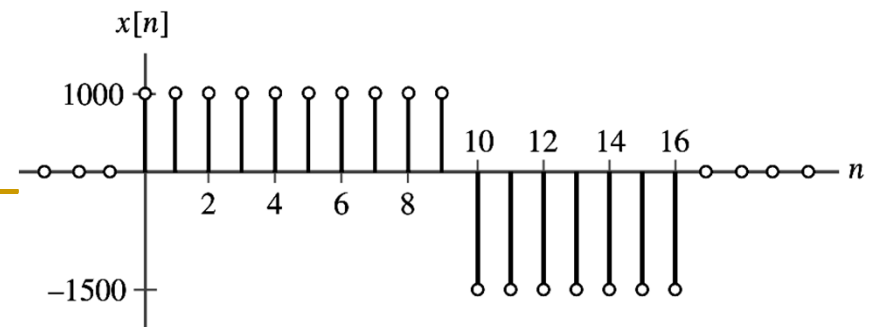
We use convolution to find the value of an investment earning **8%** per year if \$1000 is deposited at the start of each year for **10** years and then \$1500 is withdrawn at the start each year for **7** years.

<Sol.>

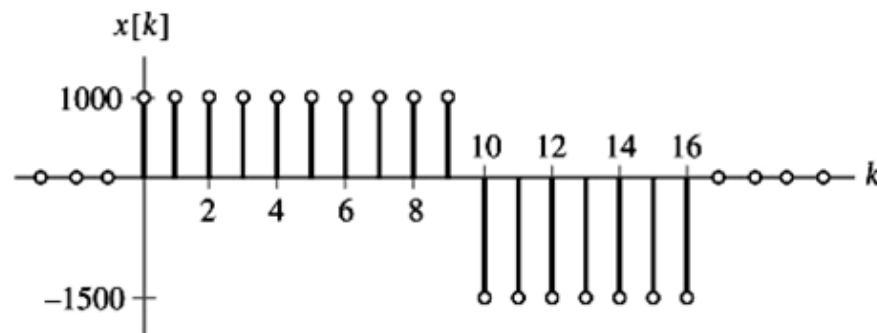
□ Impulse response : $h[n] = \rho h[n - 1] + \delta[n] \implies h[n] = \rho^n u[n], \rho = 1.08$

□ input:

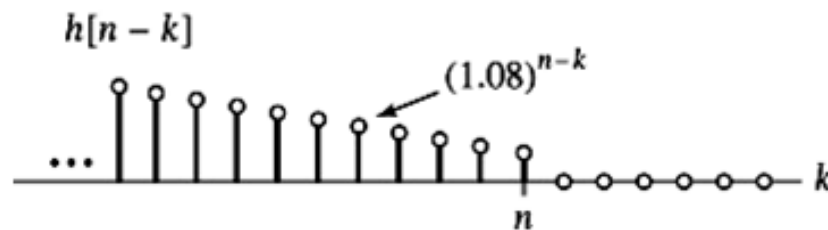
$$x[n] = \begin{cases} 1000, & 0 \leq n \leq 9 \\ -1500, & 10 \leq n \leq 16 \\ 0, & \text{otherwise} \end{cases}$$



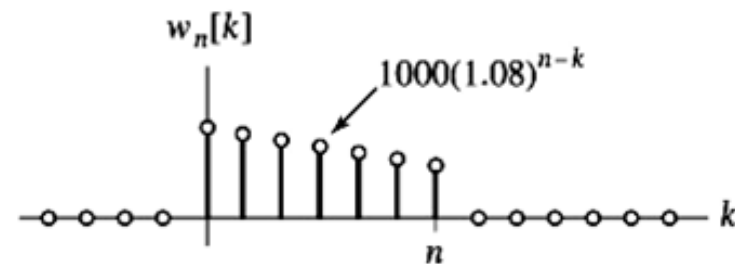
Convolution Sum Evaluation Procedure



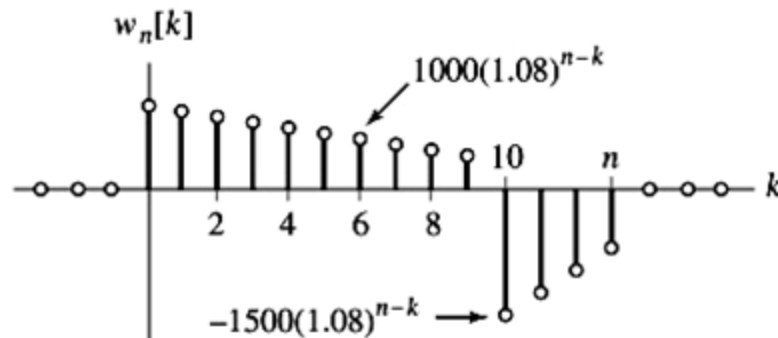
$$h[n-k] = 1.08^{n-k} u[n-k] = \begin{cases} 1.08^{n-k}, & k \leq n \\ 0, & \text{otherwise} \end{cases}$$



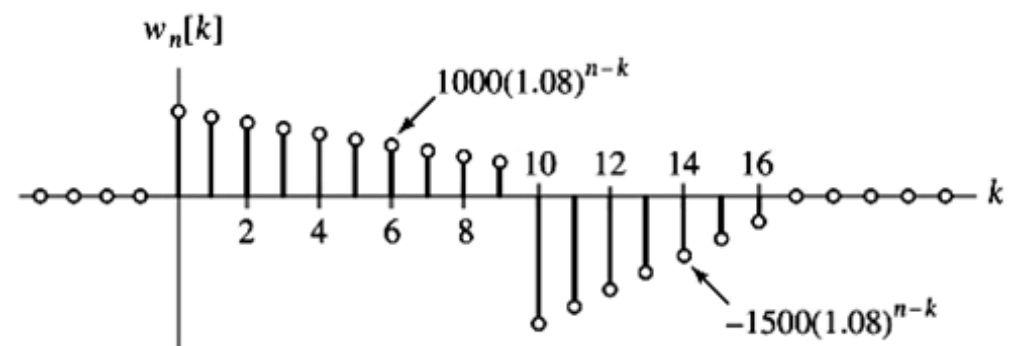
□ $0 \leq n \leq 9$:



□ $10 \leq n \leq 16$:



□ $n \geq 17$:



Convolution Sum Evaluation Procedure

■ **Four intervals**

$$h[n-k] = \begin{cases} 1.08^{n-k}, & k \leq n \\ 0, & \text{otherwise} \end{cases} \quad x[n] = \begin{cases} 1000, & 0 \leq n \leq 9 \\ -1500, & 10 \leq n \leq 16 \\ 0, & \text{otherwise} \end{cases}$$

□ **$n < 0$:** $w_n[k] = 0$ $y[n] = 0$

□ **$0 \leq n \leq 9$:** $w_n[k] = \begin{cases} 1000 * 1.08^{n-k}, & 0 \leq k \leq n \\ 0, & \text{otherwise} \end{cases} \Rightarrow y[n] = \sum_{k=0}^n 1000 * 1.08^{n-k}$

□ **$10 \leq n \leq 16$:** $w_n[k] = \begin{cases} 1000 * 1.08^{n-k}, & 0 \leq k \leq 9 \\ -1500 * 1.08^{n-k}, & 10 \leq k \leq n \\ 0, & \text{otherwise} \end{cases}$

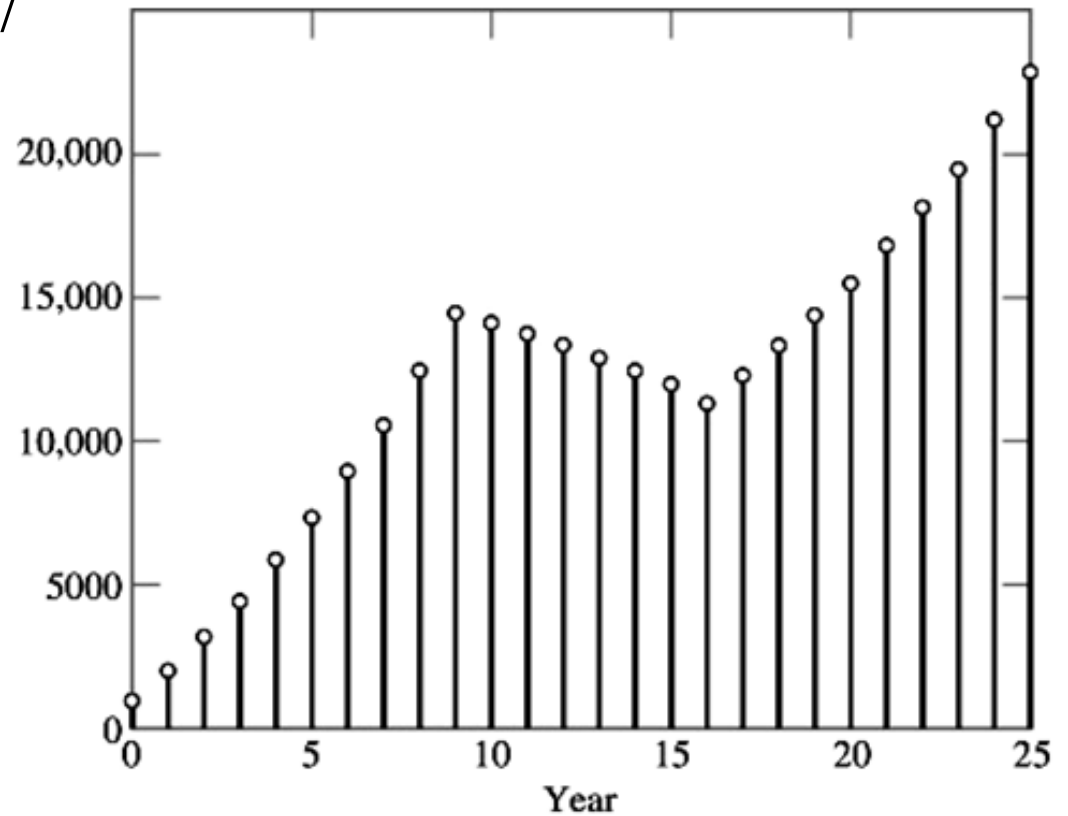
$\Rightarrow y[n] = \sum_{k=0}^9 1000 * 1.08^{n-k} - \sum_{k=10}^n 1500 * 1.08^{n-k}$

□ **$n \geq 17$:** $w_n[k] = \begin{cases} 1000 * 1.08^{n-k}, & 0 \leq k \leq 9 \\ -1500 * 1.08^{n-k}, & 10 \leq k \leq 16 \\ 0, & \text{otherwise} \end{cases}$

$\Rightarrow y[n] = \sum_{k=0}^9 1000 * 1.08^{n-k} - \sum_{k=10}^{16} 1500 * 1.08^{n-k}$

Convolution Sum Evaluation Procedure

$$y[n] = \begin{cases} 12500(1.08^{n+1} - 1), & 0 \leq n \leq 9 \\ 7246.89(1.08)^n - 18750(1.08^{n-9} - 1), & 10 \leq n \leq 16 \\ 3340.17(1.08)^n, & n \geq 17 \\ 0, & \text{otherwise} \end{cases}$$



Summary

- Linear Time-invariant systems (LTI)
 - Introduction
 - Discrete time LTI systems: Convolution Sum
 - Convolution Sum Evaluation Procedure
- Reference in textbook: 2.1, 2.2, 2.3
- Homework: 2.32, 2.34(a,c,d), 2.35