

EE-220 Signals and Systems

Session 2019

Week 3

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Last Week

- Exponential Signals
 - Continuous and Discrete Time
- Impulse Function
- Unit Step Function
- System
 - Interconnection of Systems

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System Properties

- A system is said to be **memoryless** if its output for each value of the independent variable at a given time is **dependent on the input** at only that same time.
- A system is with **memory** if its output for each value of the independent variable at a given time is dependent on the **input and previous values**.

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

$$y(t) = x(t)$$

$$y(t) = Rx(t)$$

$$y(t) = x(t - 1)$$

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

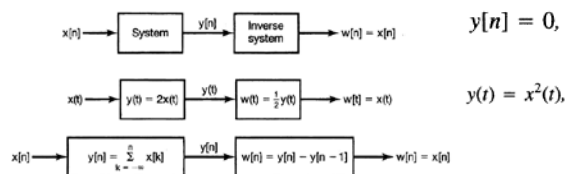
$$y(t) = \frac{1}{C} \int_{-\infty}^t x(\tau) d\tau$$

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System Properties

□ Invertibility and Inverse Systems:

- A system is said to be **invertible** if distinct inputs lead to distinct outputs.
- If a system is invertible, then an **inverse system** exists



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System Properties

- A system is **causal** if the output at any time depends on values of the input at only the present and past times.
- Such a system is often referred to as being **non anticipative**
- All memory less systems are causal, since the output responds only to the current value of the input.

$$y(t) = x(t - 1) \quad y[n] = x[n] - x[n + 1]$$

$$y[n] = \sum_{k=-\infty}^{\infty} x[k] \quad y(t) = x(t + 1)$$

Example 1.12

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

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System Properties

- A **stable system** is one in which small inputs lead to responses that do not diverge
 - A bounded input must produce bounded output signals
- Example 1.13
- **Time Invariance:**
 - A system is time invariant if the behavior and characteristics of the system are fixed over time.
- Example 1.14-1.16



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System Properties

- **Linear:** A *linear system*, in continuous time or discrete time, is a system that possesses the important property of superposition
 - *additivity* property
 - *scaling or homogeneity* property
1. The response to $x_1(t) + x_2(t)$ is $y_1(t) + y_2(t)$.
 2. The response to $ax_1(t)$ is $ay_1(t)$, where a is any complex constant.

□ 1.17-1.20

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Chapter 2

LINEAR TIME-INVARIANT SYSTEMS



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The Representation of Discrete-Time Signals in Terms of Impulses

- Discrete-time unit impulse can be used to construct any discrete-time signal. Discrete-time signal as a sequence of individual impulses.

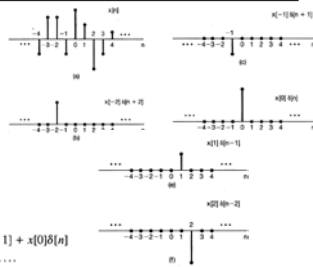
$$x[-1]\delta[n+1] = \begin{cases} x[-1], & n = -1 \\ 0, & n \neq -1 \end{cases}$$

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

$$x[1]\delta[n-1] = \begin{cases} x[1], & n = 1 \\ 0, & n \neq 1 \end{cases}$$

- Therefore, the sum of the five sequences in the figure equals $x[n]$ for $-2 < n < 2$.

$$x[n] = \dots + x[-3]\delta[n+3] + x[-2]\delta[n+2] + x[-1]\delta[n+1] + x[0]\delta[n] + x[1]\delta[n-1] + x[2]\delta[n-2] + x[3]\delta[n-3] + \dots$$



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Sifting Property of DT Impulse

$$x[n] = \dots + x[-3]\delta[n+3] + x[-2]\delta[n+2] + x[-1]\delta[n+1] + x[0]\delta[n] + x[1]\delta[n-1] + x[2]\delta[n-2] + x[3]\delta[n-3] + \dots$$

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

- Equation is called the **sifting property** of the discrete-time unit impulse

$$u[n] = \sum_{k=0}^{+\infty} \delta[n-k].$$

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The Discrete-Time Unit Impulse Response and the Convolution-Sum Representation of LTI Systems

- The importance of the sifting property lies in the fact that it represents $x[n]$ as a superposition of scaled versions of shifted unit impulses.
- The response of a linear system to $x[n]$ will be the superposition of the scaled responses of the system to each of these shifted impulses.
- Moreover, the property of time in variance tells us that the responses of a time-invariant system to the time-shifted unit impulses are simply time-shifted versions of one another.
- The convolution-sum representation for discrete-time systems that are *both* linear and time invariant results from putting these two basic facts together.

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Convolution

- For $x[n]$ we can represent the input as a linear combination of shifted unit impulses.

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

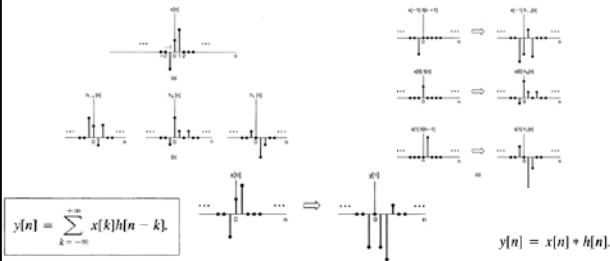
- Let $h_k[n]$ denotes the response of the linear system to the shifted unit impulse $\delta[n-k]$.
- Specifically, since $\delta[n-k]$ is a time-shifted version of $\delta[n]$, the response $h_k[n]$ is a time-shifted version of $h_0[n]$

$$h_k[n] = h_0[n-k]. \quad h[n] = h_0[n].$$

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

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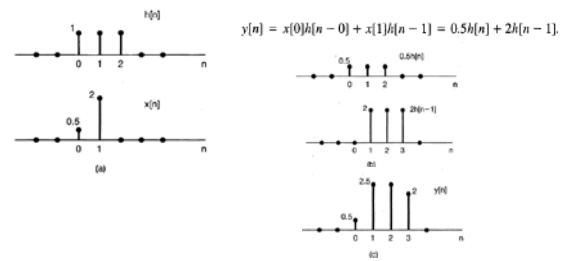
Convolution Sum or Superposition Sum



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Convolution

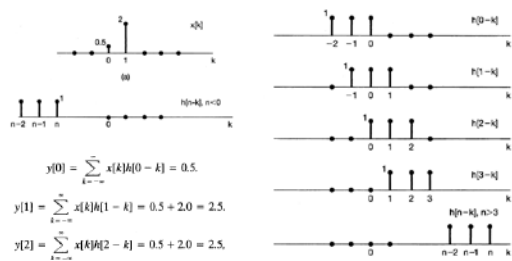
$$y[n] = \sum_{k=-\infty}^{\infty} x[k]h[n-k]$$



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Convolution

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



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Important Math relations

Sum of Geometric Series:

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

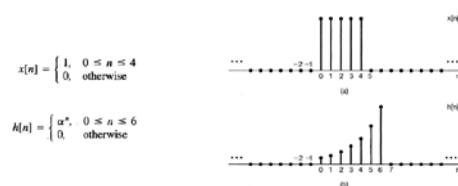
$$\sum_{n=x}^y \alpha^n = \frac{\alpha^x - \alpha^{y+1}}{1 - \alpha}$$

$$\sum_{n=0}^{\infty} \alpha^n = \frac{1}{1 - \alpha}$$

$$\sum_{n=0}^{\infty} n \alpha^n = \frac{\alpha}{1 - \alpha^2}$$

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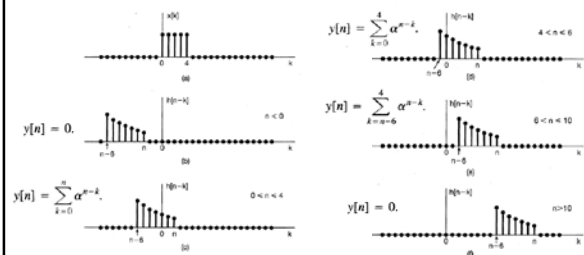
Convolution



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Convolution

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

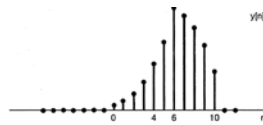


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Convolution

- Using formula of geometric series

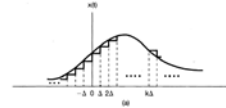
$$y[n] = \begin{cases} 0, & n < 0 \\ \frac{1 - \alpha^{n+1}}{1 - \alpha}, & 0 \leq n \leq 4 \\ \frac{\alpha^{n-4} - \alpha^{n+1}}{1 - \alpha}, & 4 < n \leq 6 \\ \frac{\alpha^{n-4} - \alpha^7}{1 - \alpha}, & 6 < n \leq 10 \\ 0, & 10 < n \end{cases}$$



- Practice example 2.5

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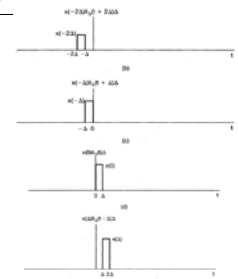
The Representation of Continuous-Time Signals in Terms of Impulses



$$\hat{x}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta)\delta_{\Delta}(t - k\Delta)\Delta$$

$$x(t) = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{\infty} x(k\Delta)\delta_{\Delta}(t - k\Delta)\Delta$$

$$x(t) = \int_{-\infty}^{\infty} x(\tau)\delta(t - \tau)d\tau$$



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Convolution Integral Representation of LTI Systems

$$\hat{y}(t) = \sum_{k=-\infty}^{\infty} x(k\Delta)\hat{h}_{k\Delta}(t)\Delta$$

$$y(t) = \lim_{\Delta \rightarrow 0} \sum_{k=-\infty}^{\infty} x(k\Delta)\hat{h}_{k\Delta}(t)\Delta$$

$$y(t) = \int_{-\infty}^{\infty} x(\tau)\hat{h}_\tau(t)d\tau$$

$$y(t) = \int_{-\infty}^{\infty} x(\tau)h(t - \tau)d\tau$$

$$y(t) = x(t) * h(t)$$



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