

# Assignments for Chapter 1

- 1.20
- 1.21 (c) (f)
- 1.24 (a)
- 1.26
- 1.27 (a) (f)
- 1.41

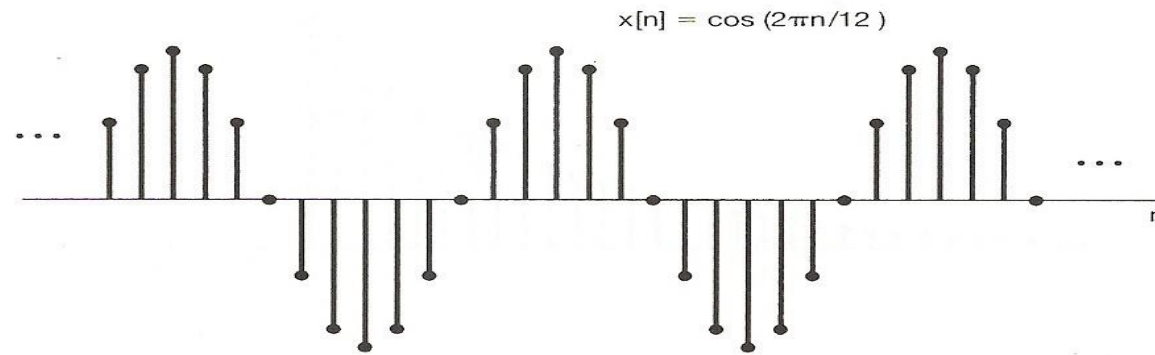
# Periodicity of DT Complex Exponentials

Important difference between  $e^{j\omega_0 n}$  and  $e^{j\omega_0 t}$ :

- $e^{j\omega_0 n}$  is a periodic signal only when  $\frac{\omega_0}{2\pi}$  is a rational number

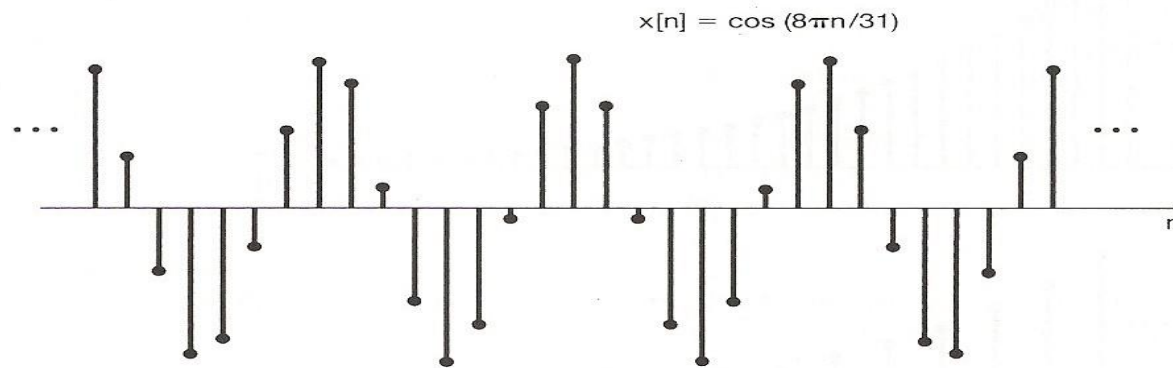
$$e^{j\omega_0 n} = e^{j\omega_0 (n+N)} \rightarrow e^{j\omega_0 N} = 1 \rightarrow \omega_0 N = 2\pi m$$

$$\text{Hence, } \frac{\omega_0}{2\pi} = \frac{m}{N}$$



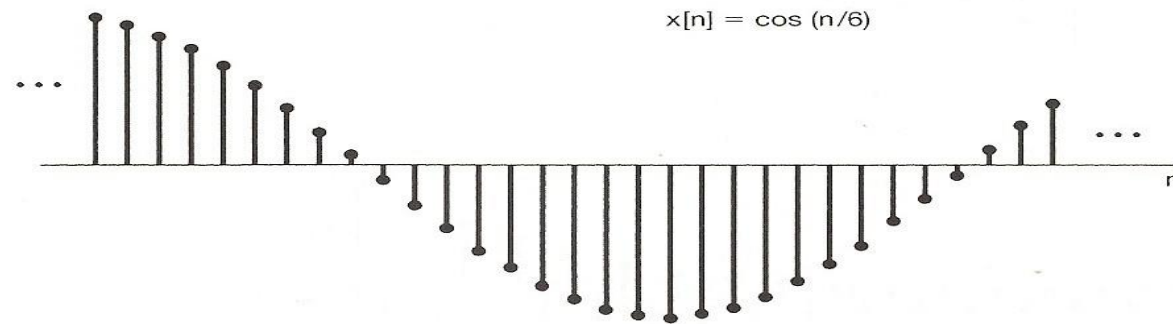
(a)

$$\frac{\omega_0}{2\pi} = \frac{2\pi/12}{2\pi} = \frac{1}{12}$$



(b)

$$\frac{\omega_0}{2\pi} = \frac{8\pi/31}{2\pi} = \frac{4}{31}$$



(c)

$$\frac{\omega_0}{2\pi} = \frac{1/6}{2\pi} = \frac{1}{12\pi}$$

**Figure 1.25** Discrete-time sinusoidal signals.

## How to determine the fundamental period of $e^{j\omega_0 n}$ ?

### Solution:

- Let  $N$  be the fundamental period, then
$$e^{j\omega_0(n+N)} = e^{j\omega_0 n} \rightarrow e^{j\omega_0 N} = 1.$$

- $\exists$  integer  $m$ ,  $\omega_0 N = 2\pi m$ .
- Therefore,

$$N = \frac{2\pi}{\omega_0} m.$$

- Hence,  $N$  is the minimum positive integer in the set  $\{\frac{2\pi}{\omega_0} m | \forall \text{ integer } m\}$ .

## Example

- What is the fundamental period of  $e^{j\frac{6}{5}\pi n}$ ?

$$\begin{aligned}\left\{\frac{2\pi}{\omega_0} m \mid \forall \text{ integer } m\right\} &= \left\{\frac{5}{3} m \mid \forall \text{ integer } m\right\} \\ &= \left\{\dots, 0, \frac{5}{3}, \frac{10}{3}, 5, \frac{20}{3}, \dots\right\}\end{aligned}$$

Hence, the fundamental period is 5 and  
fundamental frequency is  $\frac{2\pi}{5}$ .

# DT Harmonically Related Set

Harmonically related discrete-time signal sets

$$\{\phi_k[n] = e^{jk(\frac{2\pi}{N})n}, \quad k = 0, \pm 1, \pm 2, \dots\}$$

all with common period  $N$

**There are only  $N$  elements in the above set.**

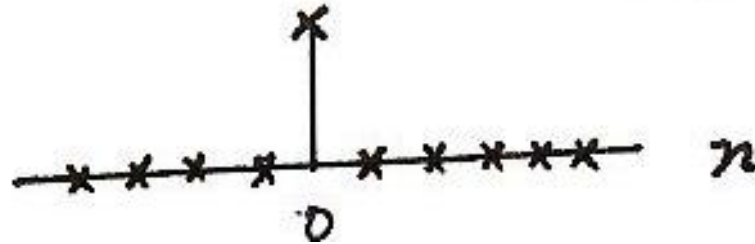
**Proof:** 
$$\phi_{k+N}[n] = e^{j(k+N)(\frac{2\pi}{N})n} = e^{jk(\frac{2\pi}{N})n} \cdot e^{j2\pi n} = e^{jk(\frac{2\pi}{N})n} = \phi_k[n]$$

This is different from continuous case. Only  $N$  distinct signals in this set.

# DT Unit Impulse Function

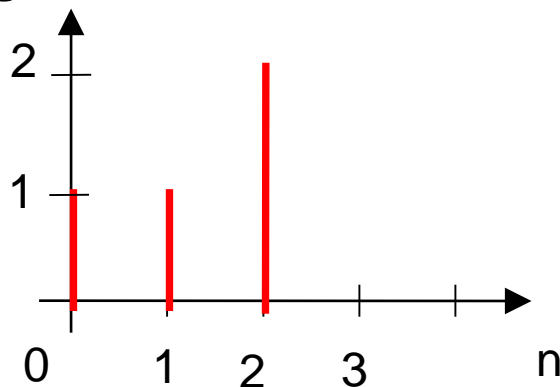
Discrete-time

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



- As a basic building function, we can use unit impulse function to represent other different signals.

e.g.1



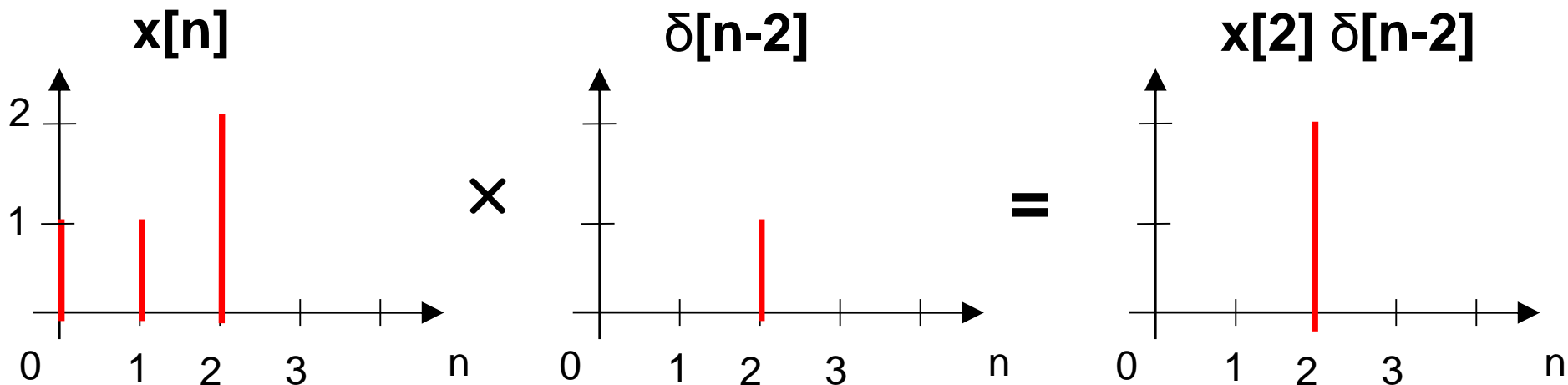
$$= \delta[n] + \delta[n - 1] + 2\delta[n - 2]$$

# DT Unit Impulse Function (cont.)

- Sampling property

$$x[n] \delta[n] = x[0] \delta[n]$$

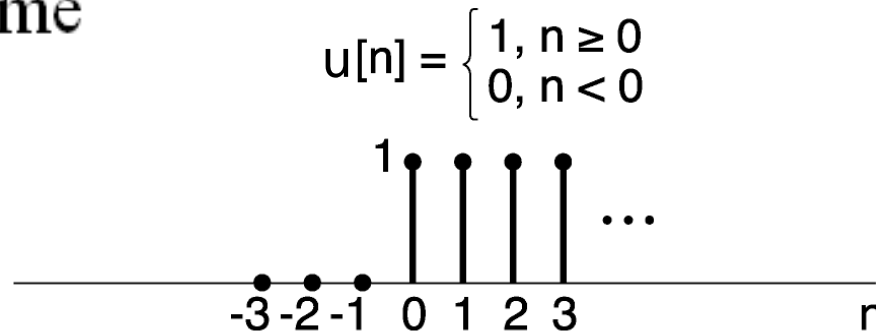
$$x[n] \delta[n-n_0] = x[n_0] \delta[n-n_0]$$





# DT Unit Step Function

Discrete-time



Relation between unit impulse and unit step functions

– First difference

$$\delta[n] = u[n] - u[n-1]$$

– Running Sum

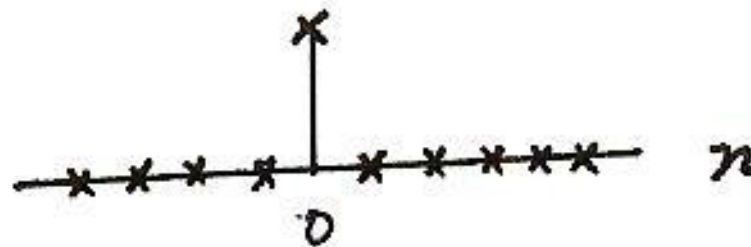
$$u[n] = \sum_{m=-\infty}^n \delta[m] \quad \left\{ \begin{array}{l} =0, \quad n < 0 \\ =1, \quad n \geq 0 \end{array} \right.$$

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

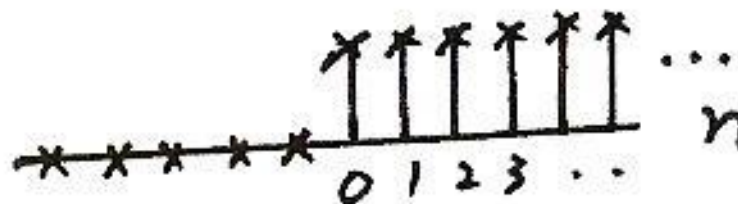
# DT Unit Step Function: First Difference

Discrete-time

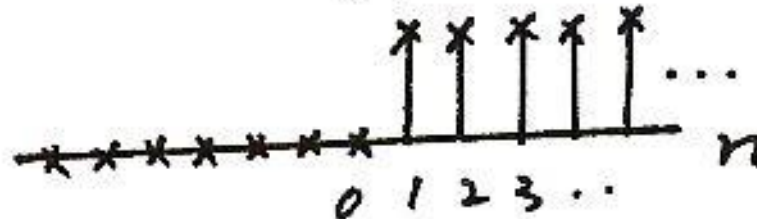
$\delta[n]$



$u[n]$



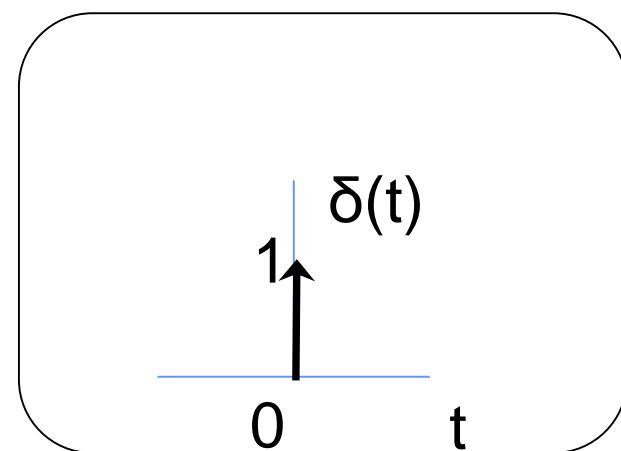
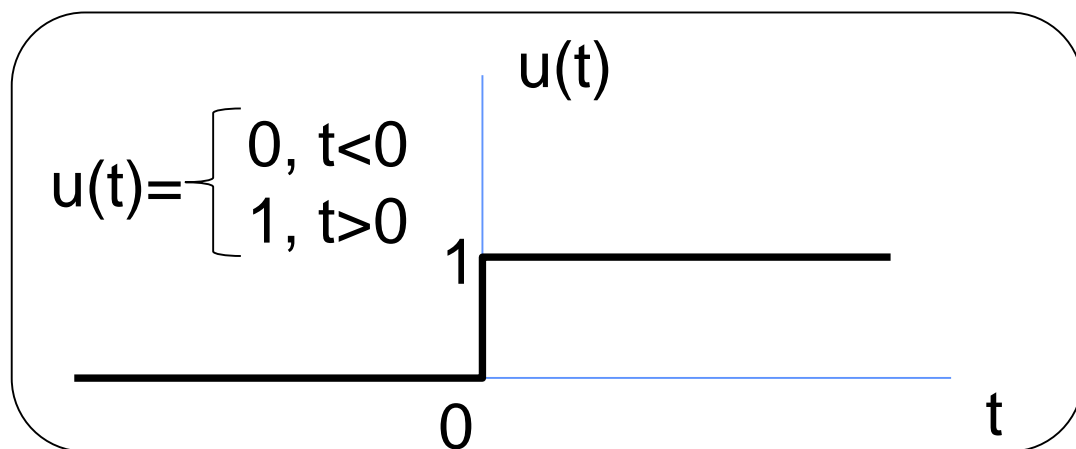
$u[n-1]$



$$\delta[n] = u[n] - u[n-1]$$

# CT Unit Impulse and Unit Step Functions

## Continuous-time



**Relation** between unit impulse and unit step functions

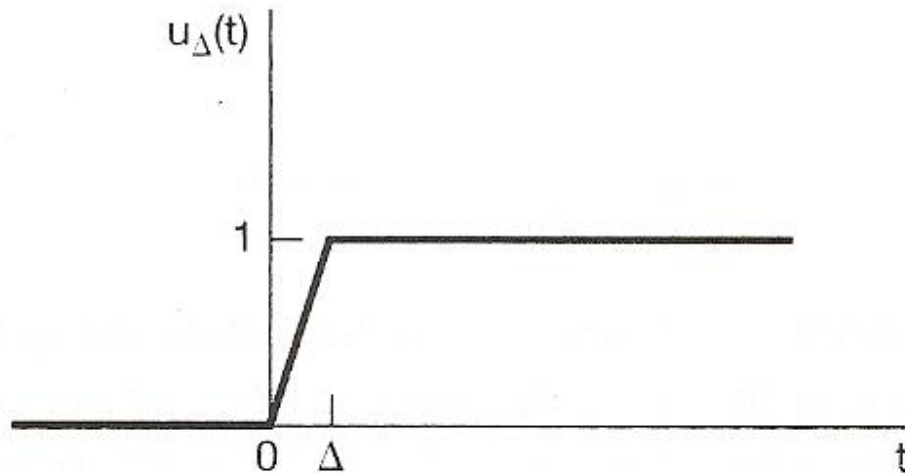
– First Derivative

$$\delta(t) = \frac{du(t)}{dt}$$

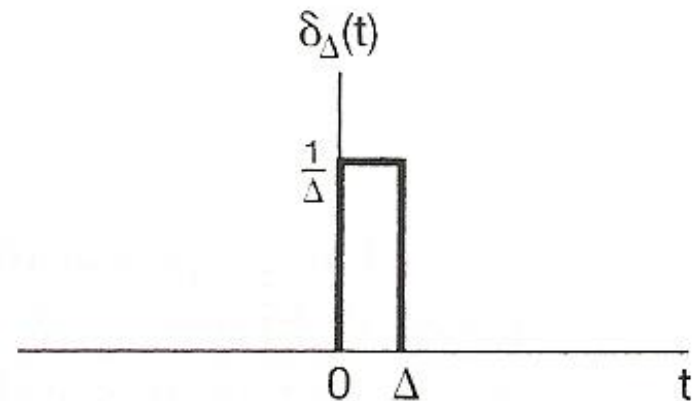
– Running Integral

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

# CT Unit Impulse and Unit Step Functions: Asymptotic View



**Figure 1.33** Continuous approximation to the unit step,  $u_{\Delta}(t)$ .



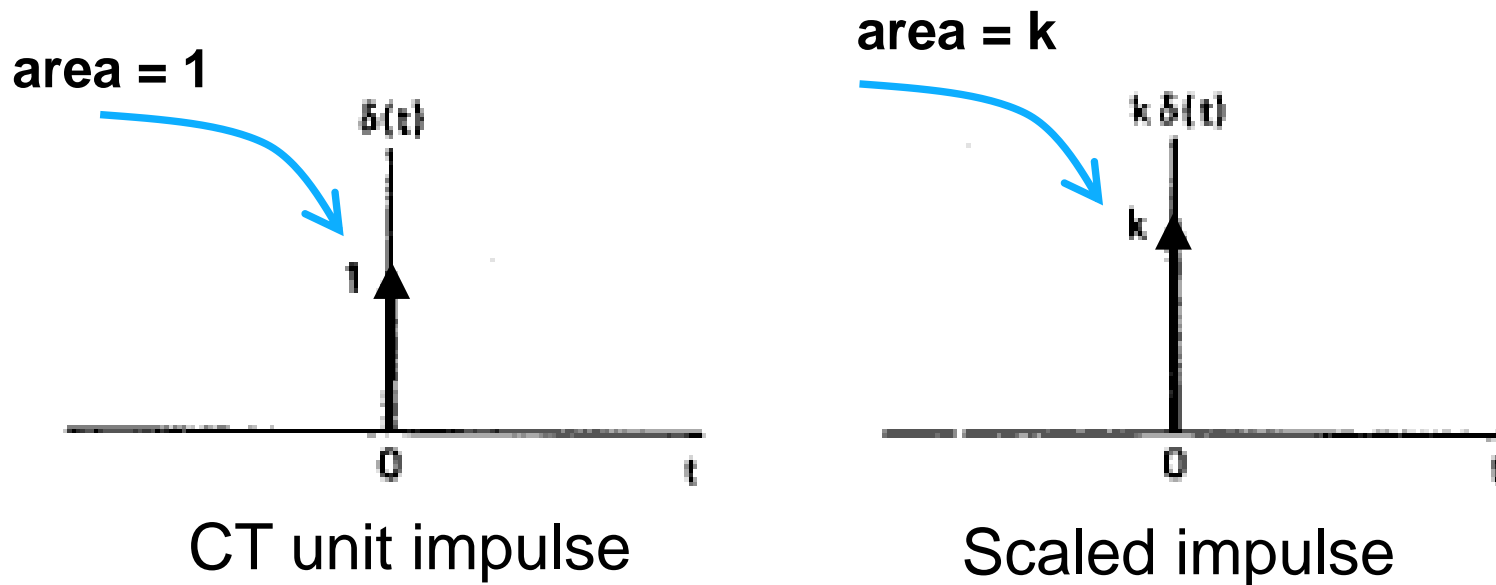
**Figure 1.34** Derivative of  $u_{\Delta}(t)$ .

$$u(t) = \lim_{\Delta \rightarrow 0} u_{\Delta}(t)$$

$$\delta(t) = \lim_{\Delta \rightarrow 0} \delta_{\Delta}(t)$$

## More on CT unit impulse function:

- $\delta(t)$  has in effect no duration, but unit area.



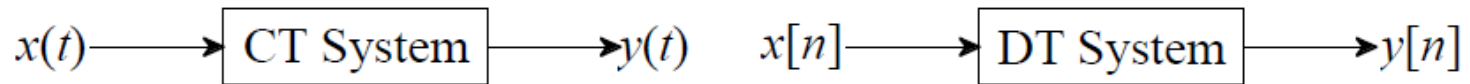
- Or the integration of CT unit impulse function is unit.  $\int_{-\infty}^{\infty} \delta(t) dt = 1$

## Sampling Property

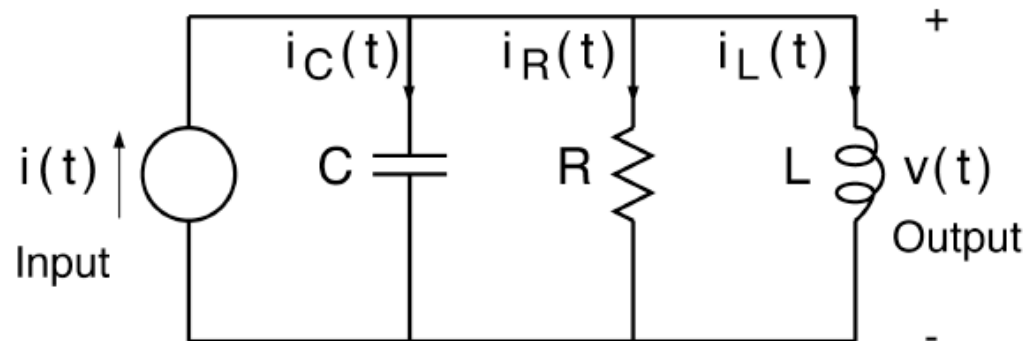
- Sampling property

$$x(t) \delta(t - t_0) = x(t_0) \delta(t - t_0)$$

# System Examples

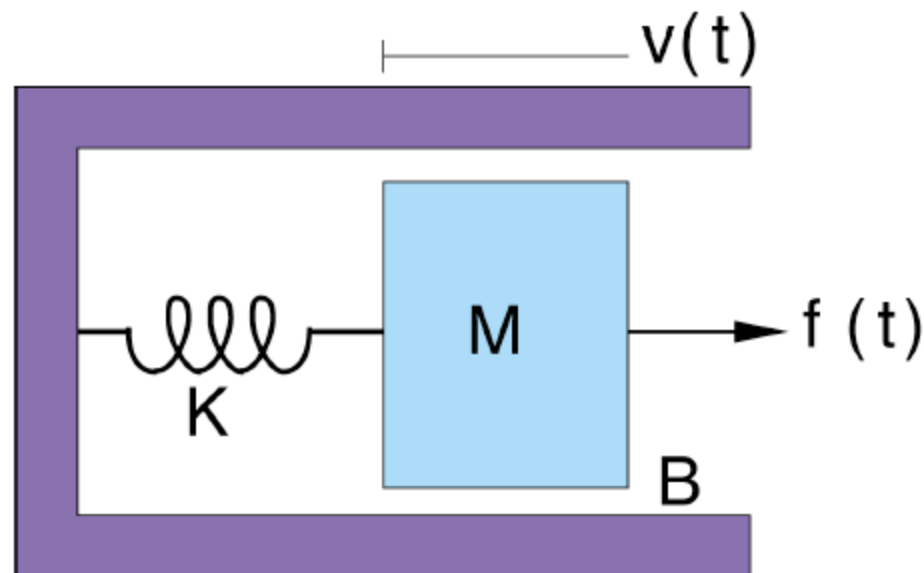


## Ex. #1 RLC circuit — an electrical system



$$i(t) = \underbrace{C \frac{dv(t)}{dt}}_{\text{capacitance}} + \underbrace{\frac{v(t)}{R}}_{\text{resistance}} + \underbrace{\frac{1}{L} \int_{-\infty}^t v(\tau) d\tau}_{\text{inductance}} .$$

## Ex. #2 A shock absorber – a mechanical system



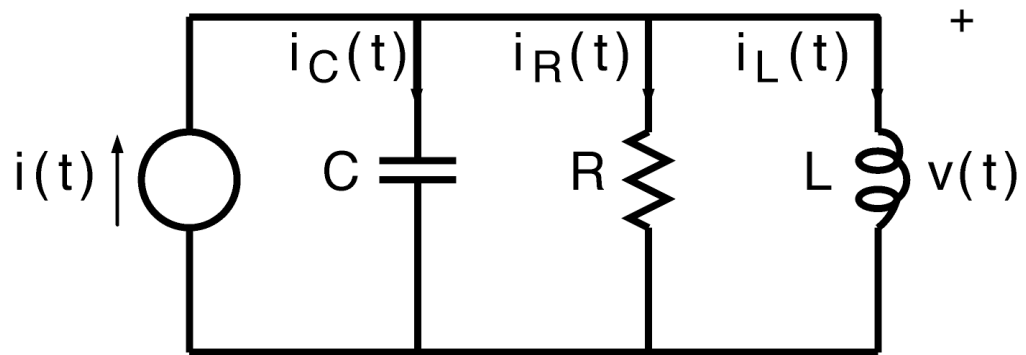
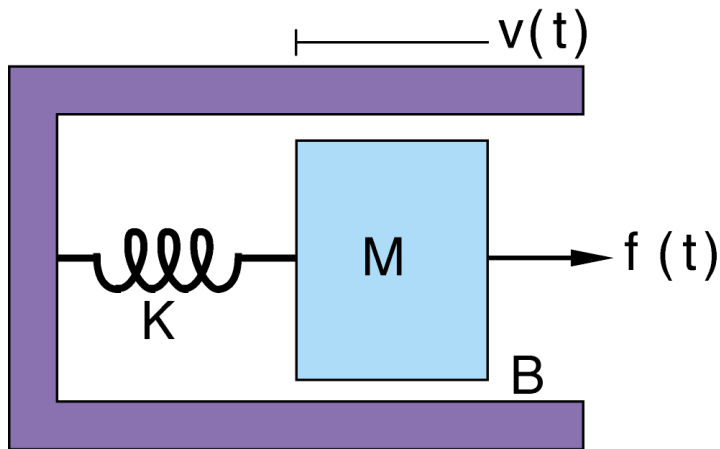
Force Balance:

$$f(t) = \underbrace{M \frac{dv(t)}{dt}}_{\text{inertial force}} + \underbrace{Bv(t)}_{\text{friction}} + \underbrace{K \int_{-\infty}^t v(\tau) d\tau}_{\text{spring force}} .$$

This equation looks quite familiar, we just saw it earlier.



- Observation: **different systems** could be described by **the same input/output relations**
- In this course, we focus on the mathematical relation between input and output

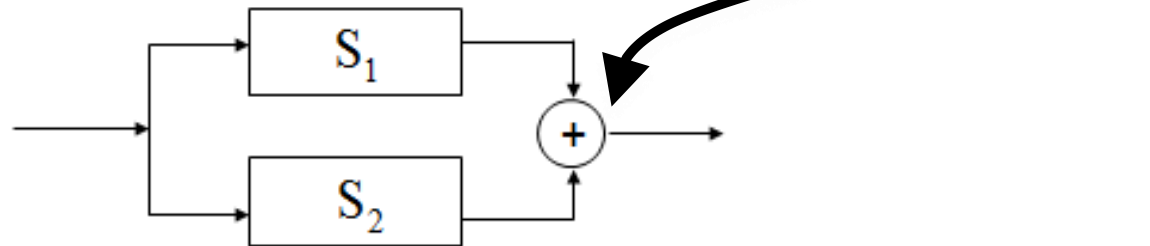


# Interconnection of Systems

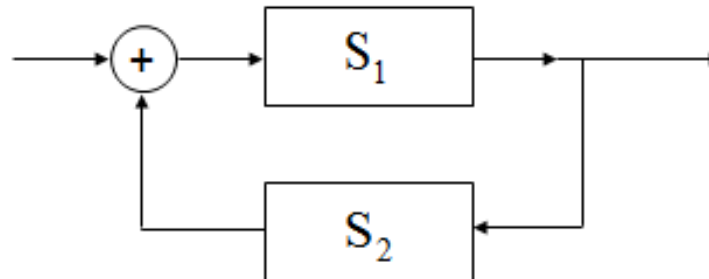
- **Series (cascade)**



- **Parallel**



- **Feedback**



# System Properties :

## 1) Memoryless or With Memory

Memoryless : output at a given time depends only on the input at the same time

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

With Memory

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

**summer or accumulator**

## 2) Invertability

invertible : distinct inputs lead to distinct outputs, i.e.  
an inverse system exists



No information loss

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

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

### 3) Causality

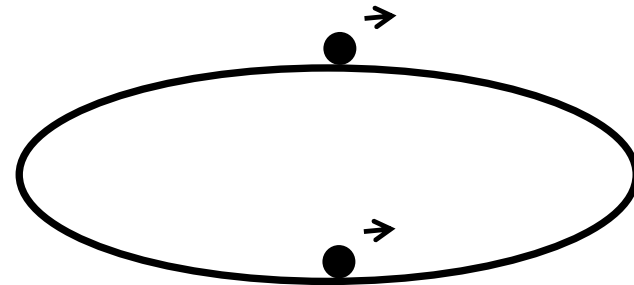
- **Causality**: A system is causal if the output does **not** anticipate future values of the input, i.e., if the output at any time depends only on values of the input up to that time.
- **All real-time** physical systems are **causal**, because time only moves forward, and effect occurs after cause.  
(Imagine if you own a noncausal system whose output depends on tomorrow's stock price.)
  - ◆ Do **not** apply to spatially varying signals. (We can move both left and right, up and down.)
  - ◆ Do **not** apply to systems processing *recorded* (or *non-realtime*) signals, e.g. taped sports games vs. live broadcast.

# Causal or Non-causal?

- $y(t) = x^2(t-1)$
- $y(t) = x(t+1)$
- $y(t) = x(t) \cos(t+1)$
- $y[n] = x[-n]$
- $y[n] = (1/2)^{n+1} x^3[n-1]$

## 4) Stability

- If the input to a stable system is bounded, the output must also be bounded.
- e.g.:  
 $S_1: y(t) = t x(t)$   
 $S_2: y(t) = e^{x(t)}$



## 5) Time Invariance (TI)

- DT: A system  $x[n] \rightarrow y[n]$  is TI if for *any* input  $x[n]$  and *any* time shift  $n_0$

$$\begin{array}{ll} \text{If} & x[n] \rightarrow y[n] \\ \text{then} & x[n - n_0] \rightarrow y[n - n_0] . \end{array}$$

- Similarly for CT time-invariant system

$$\begin{array}{ll} \text{If} & x(t) \rightarrow y(t) \\ \text{then} & x(t - t_0) \rightarrow y(t - t_0) . \end{array}$$



# Time-invariant or Time-varying?

- Steps:

- 1) Calculate  $y_1(t) \leftarrow x_1(t)$
- 2) Calculate  $y_2(t) \leftarrow x_2(t) = x_1(t-t_0)$
- 3) Does  $y_1(t-t_0)$  equal  $y_2(t)$ ?

e.g.:  $y[n] = \left(\frac{1}{2}\right)^{n+1} x^3[n-1]$

$$\textcircled{1} y_1[n] = \left(\frac{1}{2}\right)^{n+1} x_1^3[n-1]$$

$$\textcircled{2} x_2[n] = x_1[n-n_0]$$

$$\begin{aligned} y_2[n] &= \left(\frac{1}{2}\right)^{n+1} x_2^3[n-1] \\ &= \left(\frac{1}{2}\right)^{n+1} x_1^3[n-n_0-1] \end{aligned}$$

$$\textcircled{3} y_1[n-n_0] = \left(\frac{1}{2}\right)^{n-n_0+1} x_1^3[n-n_0-1]$$

$$\therefore y_1[n-n_0] \neq y_2[n]$$

$\therefore$  Time-varying

## Now we can deduce something:

- If the input to a TI system is periodic, then the output is also periodic with the same period (Problem 1.43 (a)).

**Proof:** Suppose  $x(t + T) = x(t)$   
 and  $x(t) \rightarrow y(t)$

Then by TI

$$x(t + T) \rightarrow y(t + T)$$



But these are  
the same input!



So these must be  
the same output,  
*i.e.*,  $y(t) = y(t+T)$

# Linearity

**Suppose  $x_1(t) \rightarrow y_1(t)$  and  $x_2(t) \rightarrow y_2(t)$ , such system is linear, if**

**1) Additivity property:  $x_1(t) + x_2(t) \rightarrow y_1(t) + y_2(t)$**

**2) Scaling (or homogeneity) property:**

$$a x_1(t) \rightarrow a y_1(t)$$

**where  $a$  is a complex number**

**e.g.:  $y(t) = 2 x(t)$**

**$y(t) = x^2(t)$**

# Linear system or not?

- **Steps**

- 1) Have  $y_1(t)$  and  $y_2(t)$  as output signals to  $x_1(t)$  and  $x_2(t)$
- 2) Have  $y_3(t)$  as output signal to  $x_3(t) = a x_1(t) + b x_2(t)$
- 3) Does  $y_3(t)$  equal “ $a y_1(t) + b y_2(t)$ ”?

**More examples on textbook**

**Read Example 1.17 ~ 1.20**

## Linearity (cont.)

- Superposition

If  $x_k[n] \xrightarrow{\text{Linear System}} y_k[n] \quad k=1,2,3,\dots$

Then  $\sum_k a_k x_k[n] \xrightarrow{\text{Linear System}} \sum_k a_k y_k[n]$

- This property seems to be almost trivial now, but it is one of the most important ones

# Linear Time-invariant (LTI) Systems

- LTI: Linear + Time invariant
- A basic fact: If we know the response of an LTI system to **some** inputs, we actually know the response to **many** inputs.

# Example: DT LTI System

