



Signals and Systems Analysis

Course Code: (ECEG-2121)

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Course title: Signals and Systems Analysis

ECTS (Credit): 6hr (4hr)

Pre-requisite: Applied Math **III**

Academic Year: 2019/20

Course Description

- Classification of signals and their representation as function **of time and frequency domain**; Fourier Series & Fourier transform of periodic and non-periodic signals and their properties; Laplace transform of signals:- properties and their application; representation of systems:- classification, convolution integral and system modeling using impulse response & transfer function;
- **Analysis of LTI systems:-** using differential equations solution, frequency response; discrete time signals and systems:- sampled data (sequences), convolution sum, Z-transform, system analyses in Z-domain, system function and their realization.

Course Objectives

Students who **successfully complete** the course will be able to:

- **Understand and apply** the representation different signals,
- Classification characterization different signals
- analysis of signals and systems in time and frequency domain
- properties and their application; representation of systems
- Representation of frequency response; discrete and continuous time signals and systems

Text Book:

- Signals and Systems, Second Edition, **Simon Haykin** and Barry Van Veen, John, Wiley & Sons, 2003

References:

- Signals and systems, **A.P. Oppenheim**, A.S. Willsky, I.T. Young, 2001
- Roberts: Signals and Systems: Analysis using Transform Methods and MATLAB, MJ, International Edition, McGraw Hill, 2003.
- Philip Denbigh: System Analysis and Signal, 1988
- **** AMU – AmiT Digital Library Sources (additionally you can referee)
 - drs.amu.edu.et
 - ils.amu.edu.et

Evaluation and assessment

Evaluation Description		Weight/100%	Date
Lecture	Test I	15%	
	Assignment I	10%	
	Test II	20%	
	Quiz	5%	
	Final Exam	45%	

Chapters

Chapter-1 Introduction to **Signal** and **System**

Chapter-2 **Time Domain** Representations for **LTIS**

Chapter-3 **Fourier** Representations for Signals

Chapter-4 **Laplace** Transform

Chapter-5 **Z-transform**

Chapter 1

Introduction to **Signals** and **Systems**

1.1 Introduction to Signal and System

- **Signal** and **System** analysis is a fundamental course for electrical, electronics computer engineering filed.

Definition of terms

- **Mathematical model**
- **Model**
- **Signal**
- **System**

Cont.

● **Mathematical Model:** is representation of a **system or real world** problem using mathematical expression(language). **and**

- The process of developing mathematical model is known as **mathematical modeling.**

Model: is a simplified form of representation of a **system or real world** events using :

- **Mathematical equation**
- **Graphs**
- **Drawing**
- **Prototype**

Cont.

Signal: is an abstraction of any measurable quantity that is a function of one or more independent variables such as **time(t)** and **space(x,y,z)** which intended conveys information. Or

- **signal:** is a set of information of data.

A signal: is formally defined as a function of *one or more variables*, which carries information on the nature of physical phenomena.

Example of Signal

Electrical signal

- **Voltage $v(t)$ and Current $I(t)$**

Mechanical signal

- **Sound signal**

Cont ...

Light signal

- **Image**

When a signal depends on *a single variable*, the signal is said to be *one-dimensional (single valued) signal*.

- A **speech signal**: is an example of **a one dimensional** signal whose amplitude varies with time, depending on the spoken word and who speaks it.

Cont ...

When the function depends on two or more variables, the signal is said to be *multidimensional signal*.

- *An image* is an example of *a two dimensional signal*, with the horizontal and vertical coordinates of the image representing the two dimension.
- In this course we focus on signals involving *a single independent variable*.

System

A **system** is formally defined as an entity that **manipulates one or more signals** to accomplish a function, thereby **yielding new signals**. or

System: is defined as **interconnection** of **one or more networks** in a **designed manner** to perform desired task.

- The interaction between a system its associated signals is illustrated schematically in **figure 1**.

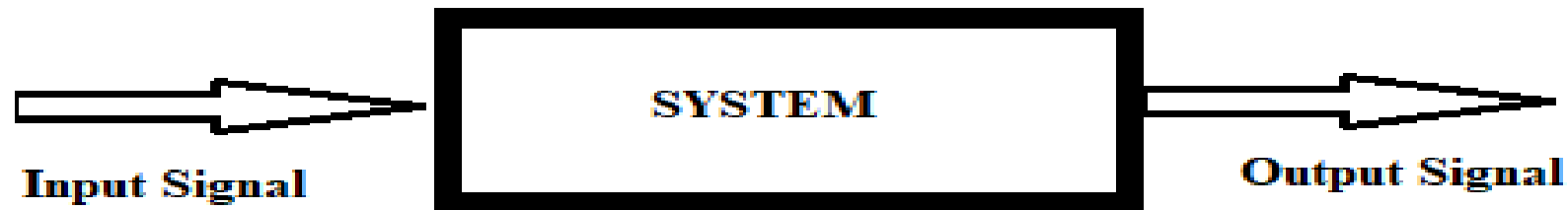


FIGURE 1: Block diagram representation of a system

Cont ...

The descriptions of the input and output signals naturally depends on the intended application of the system:

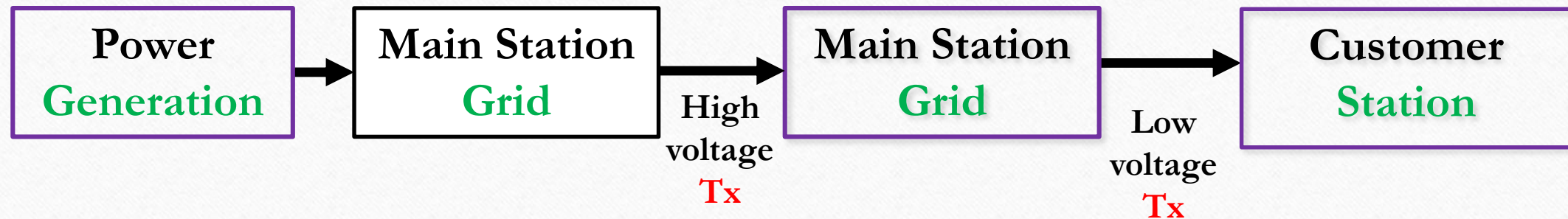
- In an automatic speaker recognition system, the **input signal** is a **speech (voice)** signal, the system is a computer and the **output signal** is the **identity** of the speaker
- In **a communication system**, the input signal could be a speech signal or computer data, the system itself is made up of the combination of a ***transmitter, channel and receiver*** and the output signal is an estimate of **the original message signal**.

Cont ...

□ Electronic Communication System



□ Electrical Power System



1.2 Classification of Signals

In this course we will restrict our attention to *one-dimensional signals* defined as *single valued functions of time*.

“single valued” means that for every instant of time there is a unique value of the function.

- This value may be a real number, in which case we speak of *a real valued signal*, or
- it may be a complex number, in which case we speak of *a complex-valued signal*.
- *In either case*, the independent variable, *namely time, is real valued*.

Cont ...

- We may identify **five methods of classifying signals** based on different features.

1. **Continuous –Time and Discrete-Time Signals**

2. **Even and Odd Signals**

3. **Period Signal and Non-Periodic Signal**

4. **Deterministic Signals and Random Signals**

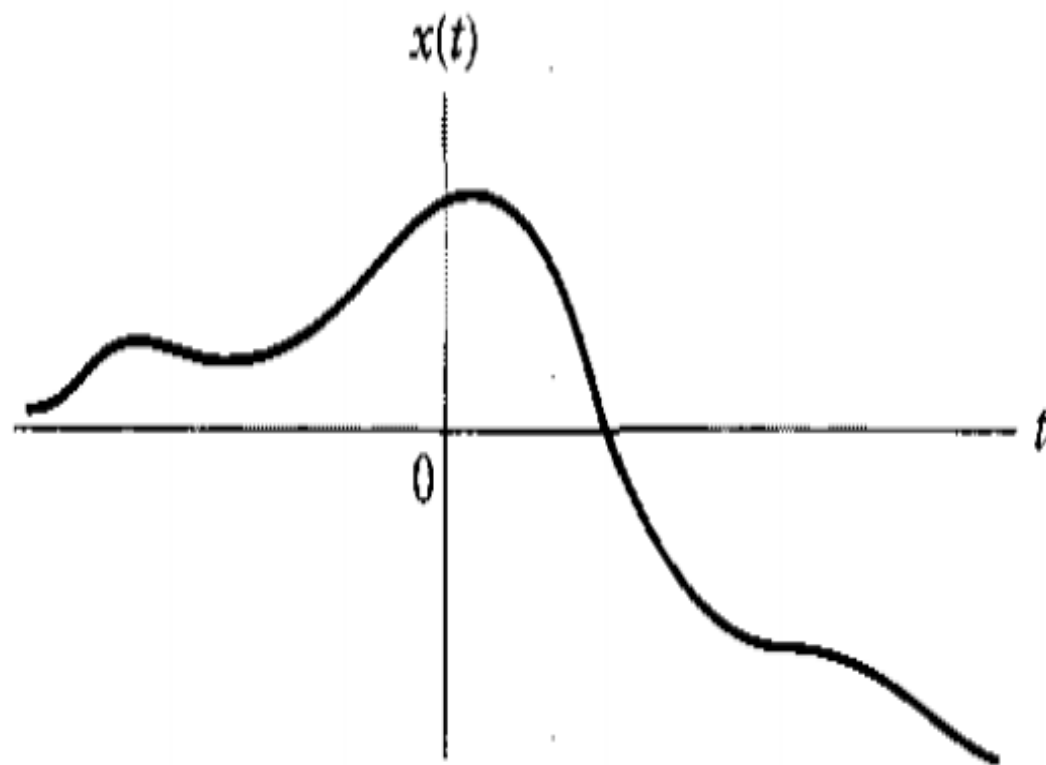
5. **Causal and Non-causal Signal**

6. **Energy Signals and Power Signals**

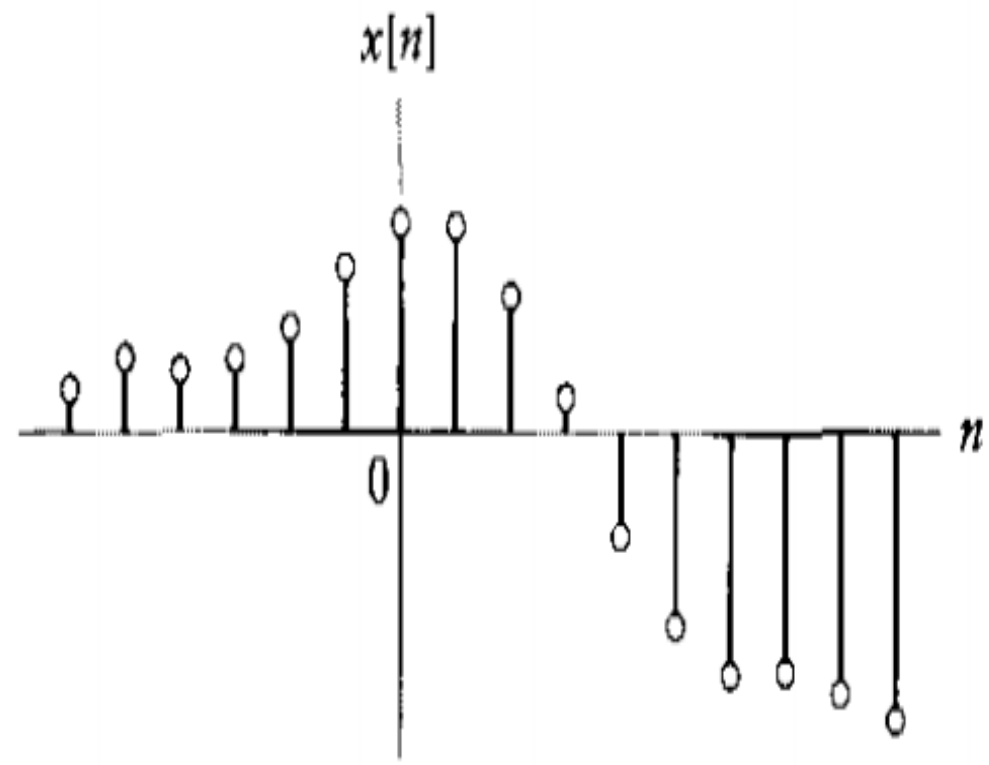
1. Continuous –Time and Discrete-Time Signals

Continuous-time signals

- A signal $x(t)$ is said to be **a continuous-time** signal if it is defined for all time 't'.
- **Figure 2(a)**, represents an example of **a continuous-time signal** whose amplitude or values varies continuously with time.
- Continuous-time signals arise naturally when a physical waveform such as **an acoustic (audio)** wave or **light wave** is converted in to an **electrical signal**.



(a)



(b)

Figure 2: (a) Continuous-time signal $x(t)$ (b) Representation of $x(t)$ as a discrete-time signal $x[n]$

Cont ...

- The conversion is effected by means of a transducer; examples include the **microphone**, which converts sound pressure variations in to corresponding *voltage or current variations*, and
- The photocell, which does the same for *light-intensity variations*.

Discrete-time

- In the case of DT-signal $x[n]$ is defined only at discrete instant of time.
- Thus, in this case, the independent variable has discrete values only, **which are usually uniformly** spaced, see **Figure 2(b) is DT-Signal**.
- A discrete-time signal is often derived from a continuous-time signal by *sampling it at a uniform rete*.

Cont ...

- Let T denote the sampling period and ' n ' denote an integer that may assume positive and negative values.
- Sampling a continuous-time signal $x(t)$ at time $t = nT$ yields a sample of values $x(nT)$.

- For convenience of presentation, we write:

$$x[n] = x(nT); \quad n = 0, \quad \pm 1, \quad \pm 2, \quad \pm 3, \quad \dots$$

- Thus, a discrete-time signal is represented by the sequence numbers

$$\dots, x[-2], x[-1], x[0], x[1], x[2], \dots$$

Cont ...

- Such a sequence of numbers is referred to **as a time series**, written as:

$$\{x[n], n = 0, \pm 1, \pm 2, \dots\} \quad \text{or simply } x[n]$$

the latter notation is used through out this course.

- Throughout this course, we use the symbol:
 - **'t'** to denote time for a **CT-signal** and the symbol **'n'** for a **DT-signal**.
 - **Parenthesis (.)** are used to denote **continuous-valued quantities** while **bracket [.]** are used to denote **discrete-valued quantities**.

2. Even and Odd Signals

- A continuous-time signal $x(t)$ is said to be an even signal if it satisfies the condition:

$$x(-t) = x(t) \quad \text{for all } t$$

- A continuous-time signal $x(t)$ is said to be an odd signal if it satisfies the condition:

$$x(-t) = -x(t) \quad \text{for all } t$$

- In other word;
 - Even signals are symmetric about the vertical axis or time origin, where as odd signals are anti-symmetric (asymmetric) about the time origin.

Notice: *Similar remarks apply to discrete-time signals.*

Example 1: Develop the **Even/Odd** decomposition of a general signal $x(t)$ by applying the definition of **even and odd** function:

Solution: Let the signal $x(t)$ be expressed as the sum of two components $x_e(t)$ and $x_o(t)$ as follows:

$$x(t) = x_e(t) + x_o(t)$$

- Define $x_e(t)$ to be even and $x_o(t)$ to be odd; that is:

$$x_e(-t) = x_e(t) \quad \text{and} \quad x_o(-t) = -x_o(t)$$

- Putting $t = -t$; in the expression for $x(t)$; we may then write

$$x(-t) = x_e(-t) + x_o(-t)$$

$$x(-t) = x_e(t) - x_o(t)$$

Cont ...

- Solving for $x_e(t)$ and $x_o(t)$, we thus obtain:

$$x_e(t) = \frac{x(t) + x(-t)}{2} \quad \text{and} \quad x_o(t) = \frac{x(t) - x(-t)}{2}$$

- The above definition of *even and odd* signals assume that the signals are real valued.
- In the case of *complex-valued signal*, we may speak of *conjugate symmetry*.
- A complex-valued signal is said to be *conjugate symmetric* if it satisfies the condition:

$$x(-t) = x^*(t)$$

Where the “**asterisk**” denotes **complex conjugate**.

Cont ...

- Let;

$$x(t) = a(t) + jb(t)$$

Where:

- $a(t)$ and $b(t)$ are the real and imaginary part of $x(t)$ respectively
- j is the square root of -1
- The complex conjugate of $x(t)$ is:
$$x^*(t) = a(t) - jb(t)$$
- From the previous equation it follows that: a complex values signal $x(t)$ is conjugate symmetric if **its real part is even** and **its imaginary part is odd**.
- Notice: similar remark applies to *a discrete-time signal*.

Example 2: consider the pair of signals shown in *figure 3*. Which of these signals is even and which one is odd?

Example 3: The signals $x_1(t)$ and $x_2(t)$ shown in *figure 3 (a) and (b)* constitute the **real and imaginary parts** of a complex-valued signal $x(t)$.

What form of **symmetry** does $x(t)$ have?

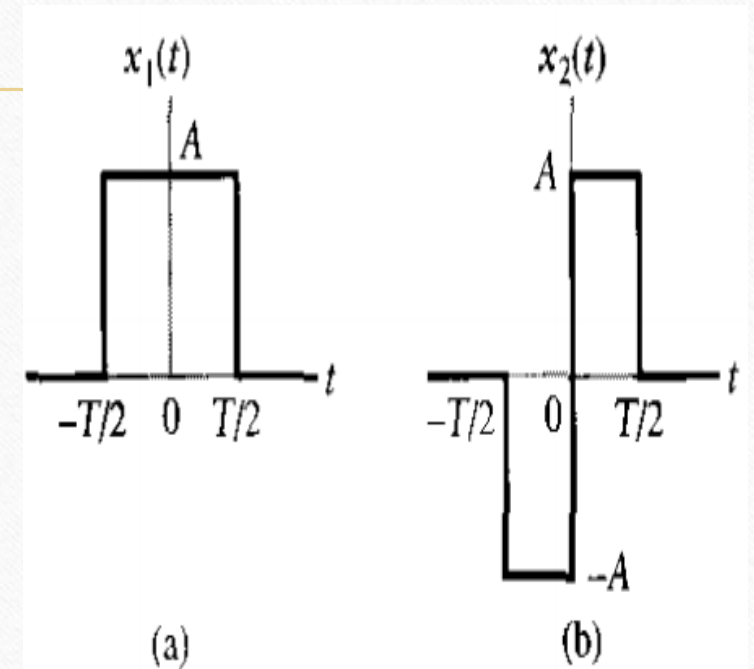


Figure 3: Examples of Continuous-Time Signal

3. Periodic Signal and Non-Periodic Signal

- A periodic signal $x(t)$ is a function that satisfies the condition:

$$x(t) = x(t + T) \quad \text{for all } t$$

Where: T is a positive constant

- Clearly, if this condition is satisfied for $T = T_o$, say, then it is also satisfied for $T = 2T_o, 3T_o, 4T_o, \dots$.
- The smallest value of T that satisfies the above equation is called the **fundamental period of $x(t)$** .
- The reciprocal of the **fundamental period T** is called the **fundamental frequency(f)** of the periodic signal $x(t)$;

Cont ...

- We thus formally write;

$$f = \frac{1}{T}$$

- The frequency f is measured in **hertz (Hz)** or **cycles per second**.
- The **angular frequency**, measured in **radians per second**, is defined by:

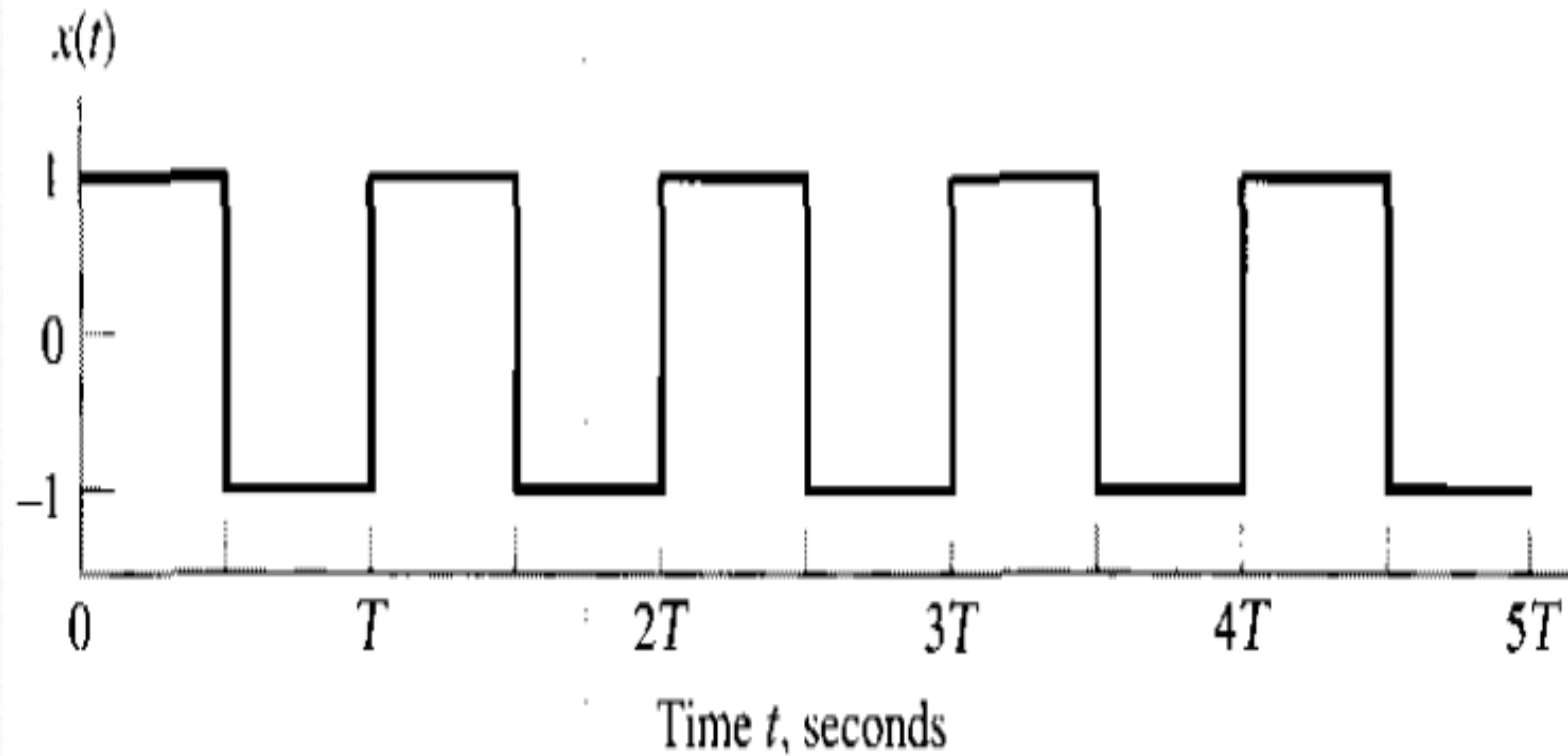
$$\omega = \frac{2\pi}{T} = 2\pi f$$

- Since there are ***2π radians in one complete cycle***.
- To simplify the terminology, **ω** is often referred to simply as frequency.

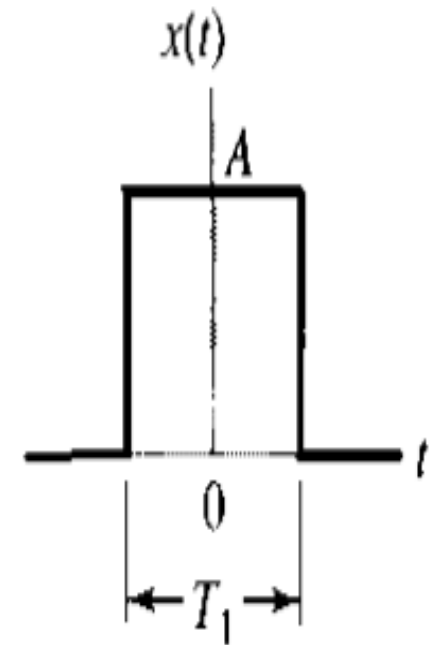
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- Any *signal $x(t)$* for which there is **no values of T** to satisfy the condition of periodic signal is called *aperiodic or non-periodic signal*.
- *Figures 4 (a) and (b)* presents examples of **periodic and non-periodic signals**, respectively.
- The periodic signal shown here represents a square wave of amplitude **$A = 1$** and period **T** and the non-periodic signal represents a rectangular pulse of amplitude **A** and duration **T_1** .

Cont ...



(a)



(b)

Figure 4: (a) Square wave with amplitude $A = 1$, and period $T = 0.2$ s (b) Rectangular pulse of amplitude A and duration T_1

Example 4:

- **Figure 5** shows a triangular wave. What is the *fundamental frequency* of this wave? Express the fundamental frequency in units of **Hz or rad/s**.

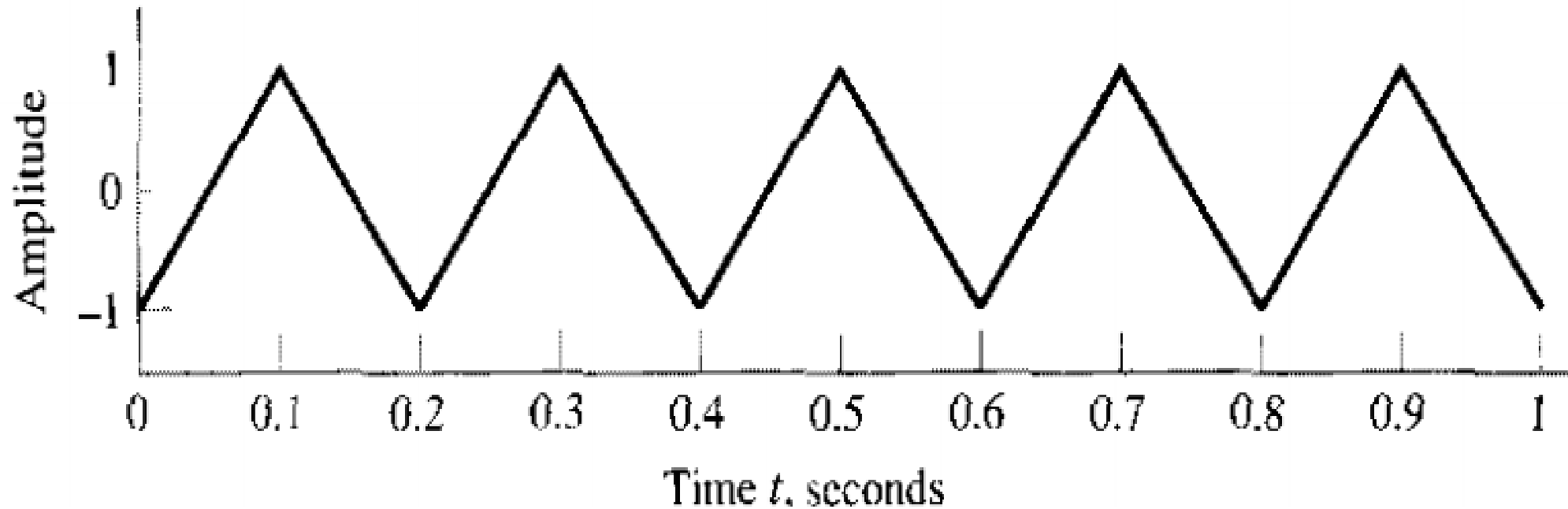


Figure 5: Triangular Wavealtering between - 1 and + 1 with fundamental period of 0.2 second

- We next consider the case of *discrete-time signals*.
- A discrete time signal $x[n]$ is said to be periodic if it satisfies the condition:

$$x[n] = x[n + N] \quad \text{for all integers } n$$

Where N ; is a positive integer.

- The smallest value of *integer N* for which the above equation is satisfied is called fundamental period of *the discrete-time signal $x[n]$* .
- The fundamental angular frequency, or simply fundamental frequency of $x[n]$ is defined by:

$$\Omega = \frac{2\pi}{N} \text{ (in radians)}$$

Cont ...

- The differences between the defining equations for continuous-time and discrete-time periodic signals should be carefully noted.
 - The earlier equation applies to a periodic continuous-time signal whose fundamental period ***T has any positive value.***
 - On the other hand, the later equation applies to ***a periodic discrete-time signal*** whose fundamental period **N can only assume a positive integer value.**
- Two examples of **discrete-time signals** are shown in ***Figure 6 (a)*** and **(b)** below. The signal in **Fig. 6(a)** is periodic where as that of **Fig. 6(b)** is aperiodic.

Example 5: What is the fundamental frequency of the discrete-time square wave shown in Fig. 6(a)

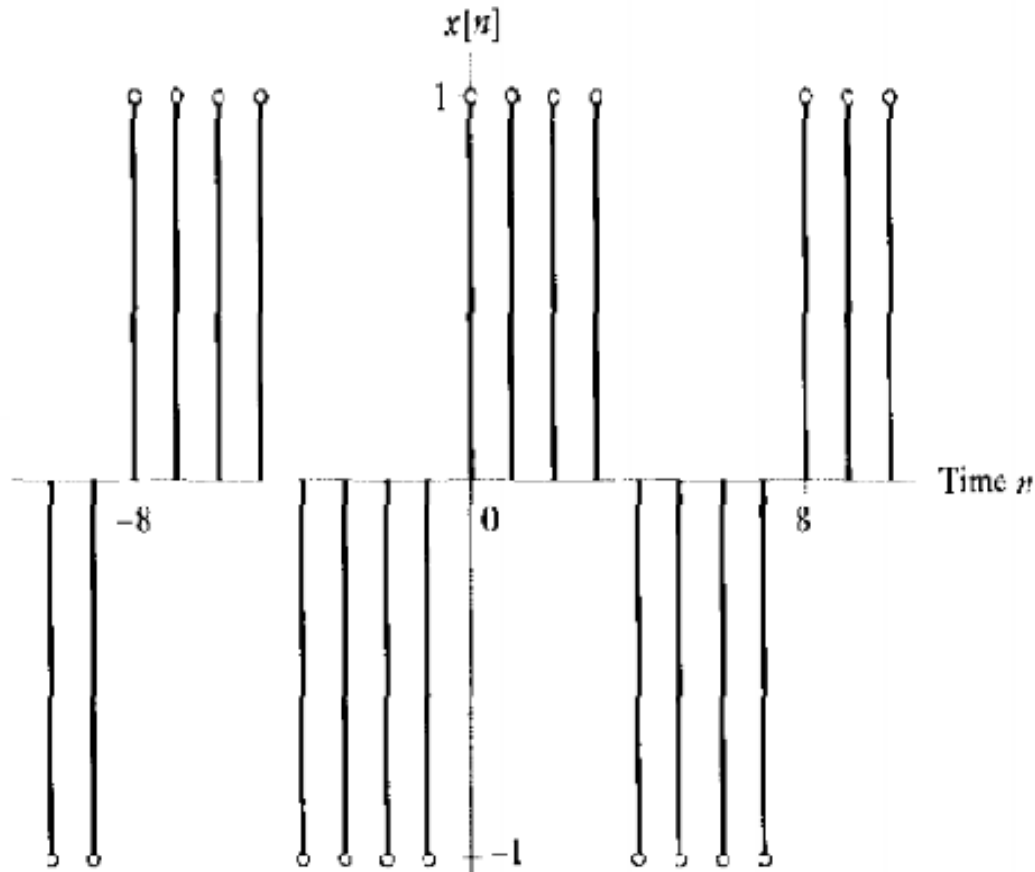


Figure 6(a): Discrete-Time Square Wave alternating between - 1 and + 1

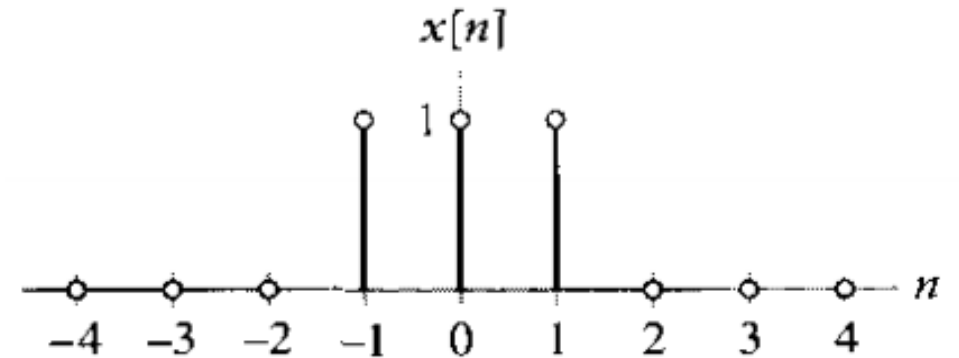


Figure 6(b): Aperiodic Discrete-Time Signal consisting of three non-zero samples

4. Deterministic Signals and Random Signals

- **A deterministic signal** is a signal about which there is **no uncertainty** with respect to its value at any time.
- Accordingly, we find that deterministic signals may be *modeled as completely specified function of time.*
 - The square wave shown in *figure 4(a)* and the rectangular pulse shown in *figure 4(b)* are deterministic signals, and so are the signals shown in *figure 6(a) and (b)*.
- **A random signal** is a signal about which there is **uncertainty** before its actual occurrence.

Cont ...

- Such a signal may be viewed as belonging to an **ensemble(collective) or group** of signals, with each signal in the ensemble having a different wave form.
- **Moreover**, each signal within the ensemble has certain probability of occurrence.
- The ensemble of such signals is referred to as *a random process*.
- The **“noise”** generated in the amplifier of *a radio or television receiver* is an example of *a random signal*. Its amplitude fluctuates between positive and negative values in a completely random fashion.

5. Energy Signals and Power Signals

- From examples provided so far, we see that signals may represent a broad variety of phenomena.
- In many, but not all, applications, the signals we consider are directly related to physical quantities capturing *power and energy* in a physical system.
- For example, if $v(t)$ and $i(t)$ are, respectively, the voltage and current across a resistor with resistance R , then the *instantaneous power* dissipated in this resistor is defined by:

$$p(t) = v(t) * i(t) = \frac{1}{R} v^2(t) = R i^2(t)$$

Cont ...

- In both cases, the instantaneous **power** $p(t)$ is proportional to the squared amplitude of **the signal**. Further more, for a resistance **R of 1Ω** ,
- the **two equations** take on the same mathematical form.
- The **total energy** expended over the time **interval** $t_1 \leq t \leq t_2$ is:

$$E(t) = \int_{t_1}^{t_2} p(t) dt = \int_{t_1}^{t_2} \frac{1}{R} v^2(t) dt$$

- The average power over this time interval is:

$$p_{av} = \frac{1}{t_2 - t_1} \int_{t_1}^{t_2} p(t) dt = \frac{1}{t_2 - t_1} \int_{t_1}^{t_2} \frac{1}{R} v^2(t) dt$$

Cont ...

- In signal analysis, it is customary to define power in terms of **a 1Ω resistor**, so that, regardless of whether a given **signal $x(t)$** represents **a voltage or a current**, we may express the instantaneous power of the signal as:

$$p(t) = x^2(t)$$

- Based on this convention, we define the **total energy of the continuous-time signal $x(t)$** as:

$$E = \lim_{T \rightarrow \infty} \int_{-T/2}^{T/2} x^2(t) dt = \int_{-\infty}^{\infty} x^2(t) dt$$

Cont ...

- And its **average power** as:

$$P_{av} = \lim_{T \rightarrow \infty} \frac{1}{T} \int_{-T/2}^{T/2} x^2(t) dt$$

- From the last equation, we readily see that the *average power of a periodic signal is $x(t)$* of fundamental period T is given by:

$$P_{av} = \frac{1}{T} \int_{-T/2}^{T/2} x^2(t) dt$$

Cont ...

- The square root of the **average power** P_{av} is called the **root mean-square (rms) value of the signal** $x(t)$.
- In the case of discrete-time signal $x[n]$, the integrals are replaced by corresponding sums. Thus the total energy of $x[n]$ is defined by:

$$E = \sum_{n=-\infty}^{\infty} x^2[n]$$

And its average power is defined by:

$$P_{av} = \lim_{N \rightarrow \infty} \frac{1}{2N} \sum_{n=-N}^N x^2[n]$$

Cont ...

- Here again, the average power in *a periodic signal* $x[n]$ with fundamental period N is given by:

$$P_{av} = \frac{1}{N} \sum_{n=0}^{N-1} x^2[n]$$

- ❖ A signal is referred to as **an energy signal**, if and only if the total energy of the signal satisfies the condition:

$$0 < E < \infty$$

- On the other hand, it is referred to **as a power signal**, if and only if the average power of the signal satisfies the condition:

$$0 < P_{av} < \infty$$

Cont ...

- The **energy and power classifications** of signals are **mutually exclusive**.
- **In particular:**
 - **An energy** signal has **zero average power**, whereas **a power signal** has **infinite energy**.
 - **Periodic signals and Random signals** are usually viewed as **power signals**, whereas signals that are both **deterministic and non-periodic** are **energy signals**.

Example 6:

- What is the total energy of:
 - a) the rectangular pulse shown in figure 7(a)?
 - b) the discrete-time signal shown in figure 7(b)?

- What is the average power of:
 - c) the square wave shown in figure 7(c)?
 - d) the triangular wave shown in figure 7(d)?
 - e) The periodic discrete-time signal shown in figure 7(e)?

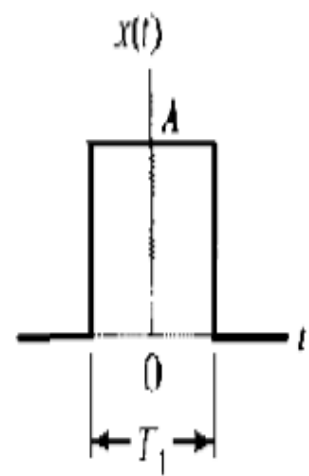


Figure 7(a) Rectangular Pulse

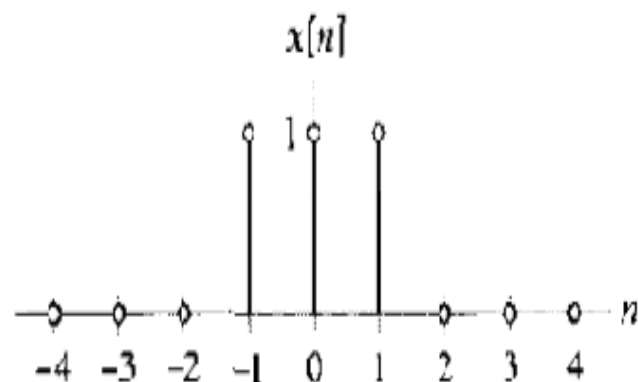


Figure 7(b) Aperiodic Discrete-Time Signal

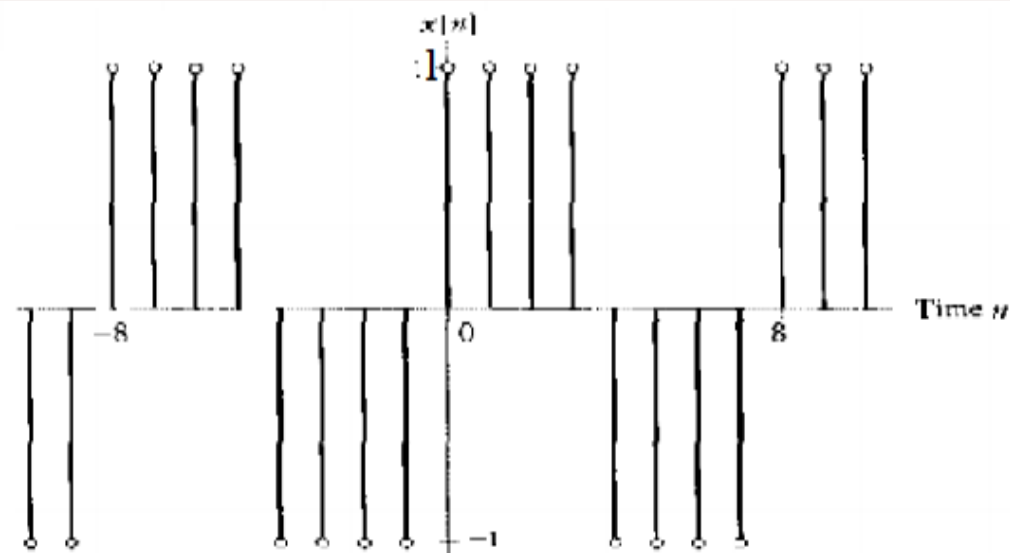


Figure 7(e) Discrete-Time Square Wave

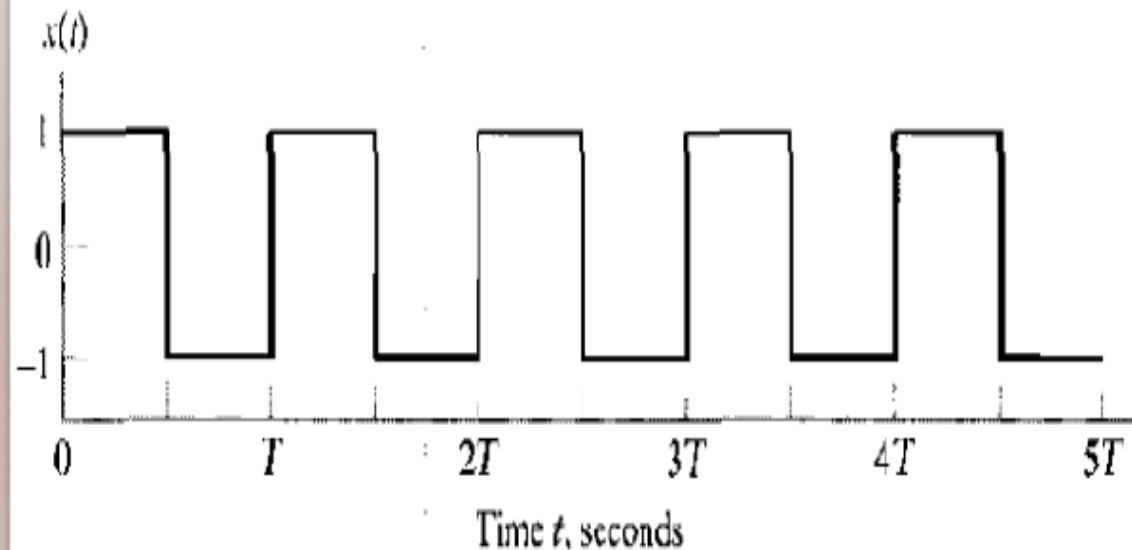


Figure 7(c) Continuous-Time Square Wave

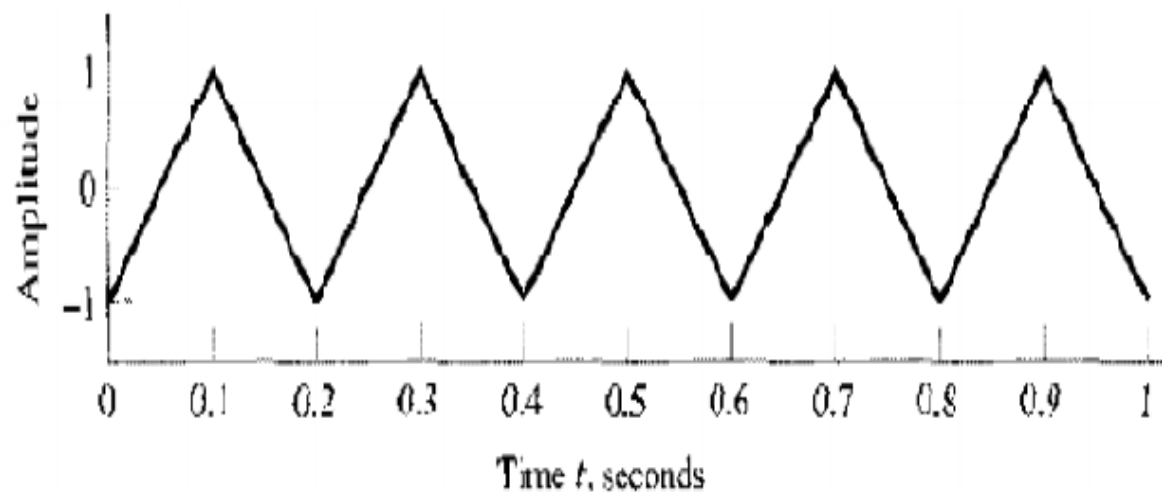


Figure 7(d) Triangular Wave

1.2 Basic operations on Signals

- An issue of fundamental importance in the study of **signals and systems** is the use of *systems to process or manipulate signals*.
- This issue usually involves a combination of *some basic operations*.
- In particular, we may identify a two classes of operations; described as:
 - 1) **Operations performed on dependent variables:**
 - Amplitude scaling, Addition/Subtraction, Multiplication, Differentiation, and Integration
 - 2) **Operations performed on independent variables:**
 - Time scaling, Reflection, and Time shifting

1. Operations performed on dependent variables

Amplitude Scaling:

- Let $x(t)$ denotes a continuous time signal.
- The signal $y(t)$ resulting from amplitude scaling applied to $x(t)$ is defined by:

$$y(t) = cx(t)$$

Where: c is the scaling factor.

- According to the above equation, the value of $y(t)$ is obtained by multiplying the corresponding values of $x(t)$ by the **scalar c** .

Cont ...

- A physical example of a device that **performs amplitude scaling is an electronic Amplifier.**
- A resistor also performs amplitude scaling when $x(t)$ is a current, c is the resistance and $y(t)$ is the output voltage.
- In a similar manner, for discrete-time signals we write:

$$y[n] = cx[n]$$

Addition signals:

- Let $x_1(t)$ and $x_2(t)$ is defined by:

$$y(t) = x_1(t) + x_2(t)$$

Cont ...

- A physical example of a device that adds signals is **an audio mixer** which *combines music and voice signals*.
- In a similar manner, for discrete time signals we write:

$$y[n] = x_1[n] + x_2[n]$$

Multiplication of signals:

- Let $x_1(t)$ and $x_2(t)$ denote a pair of *continuous-time signals*.
- The *signal* $y(t)$ resulting from the multiplication of $x_1(t)$ by $x_2(t)$ is defined by:

$$y(t) = x_1(t)x_2(t)$$

Cont ...

- That is, for each prescribed time 't' the value of $y(t)$ is given by the product of the corresponding values of $x_1(t)$ and $x_2(t)$.
- A physical example of $y(t)$ is an **AM radio signal**, in which $x_1(t)$ consists of **an audio signal plus a DC component**, and $x_2(t)$ *consists of a sinusoidal signal called a carrier wave*.
- In a similar manner, for discrete-time signals we write:

$$y[n] = x_1[n]x_2[n]$$

Cont ...

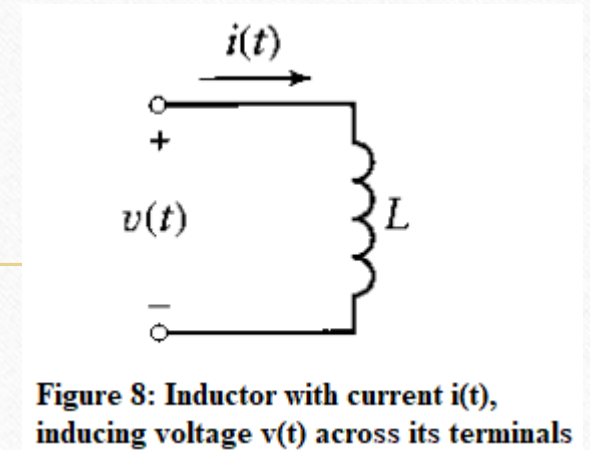
Differentiation of signal:

- Let $x(t)$ denote a continuous-time signal.
- The derivation of $x(t)$ with respect to time is defined by:

$$y(t) = \frac{d}{dt} x(t)$$

- For example, **an inductor performs differentiation.**
- Let $i(t)$ denote the current flowing through an inductor of inductance L , as shown in figure 8.
- The **voltage $v(t)$** developed across the inductor is defined by:

$$v(t) = L \frac{d}{dt} i(t)$$



Cont ...

Integration of signal:

- Let $x(t)$ denote a continuous-time signal.
- The integral of $x(t)$ with respect to time 't' is defined by:

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

Where: τ is the integration variable.

- For example, a capacitor performs integration. Let $i(t)$ denote the current flowing through a capacitor of capacitance C , as shown in **figure 9**.
- The voltage $v(t)$ developed across the capacitor is defined by;

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

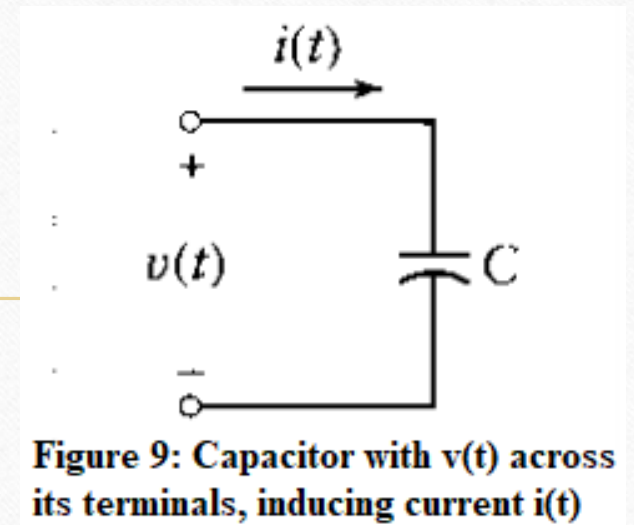


Figure 9: Capacitor with $v(t)$ across its terminals, inducing current $i(t)$

2. Signal operations performed on **independent variables**

Time Scaling:

- Let $x(t)$ denote a continuous time signal.
- The signal $y(t)$ obtained by scaling the independent variable, **time t** , by ***a*** ***factor 'a'*** is defined by:

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

- If $a > 1$, the **signal $y(t)$** is *a compressed version of $x(t)$* .
- If $0 < a < 1$, the **signal $y(t)$** is *a expanded (stretched) version of $x(t)$* .
- The two operations are illustrated in *figure 10 below*.

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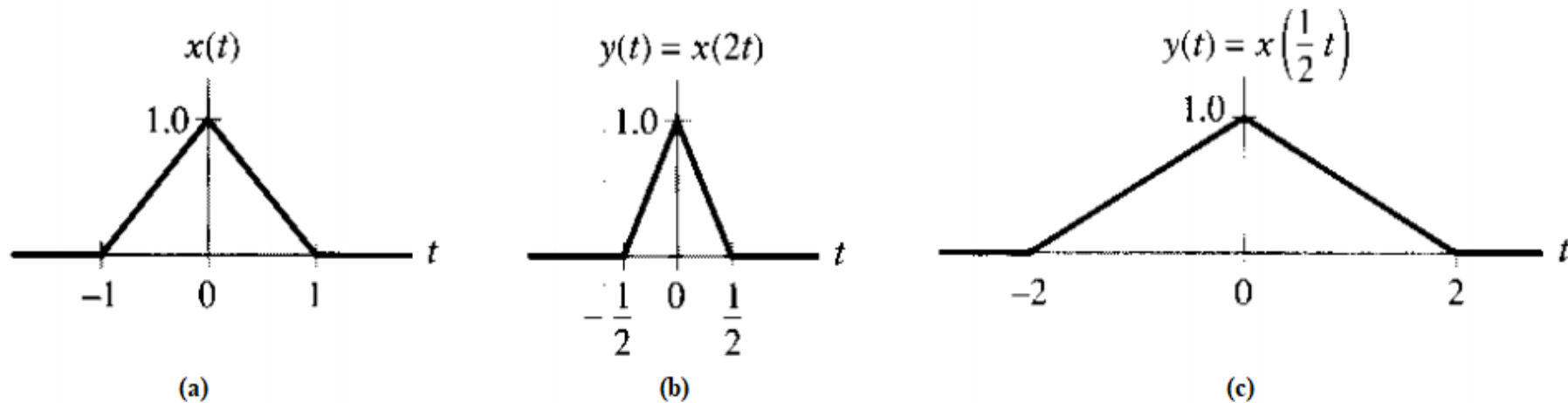


Figure 10: Time Scaling operation: (a) Continuous-Time Signal $x(t)$, (b) compressed version of $x(t)$ by a factor of 2, (c) expanded version of $x(t)$ by a factor of 2.

- In discrete-time case, we write:

$$y[n] = x[kn], \quad k > 0$$

Which is defined only for *integer values of k* .

- If $k > 1$, then some values of the discrete-time signal $y[n]$ are lost, as illustrated in figure 11 for $k = 2$.

Cont ...

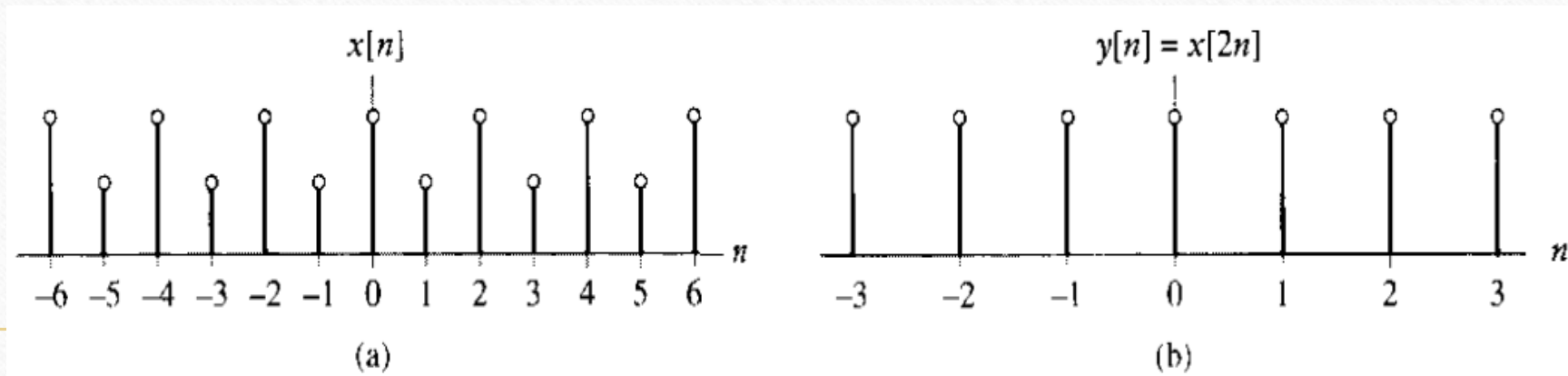


Figure 10: Effect of time scaling on a discrete-time signal: (a) discrete-time signal $x[n]$, and (b) compressed version of $x[n]$ by a factor of 2, with some values of the original $x[n]$ lost as a result of the compression.

Reflection:

- Let $x(t)$ represents a continuous-time signal.
- Let $y(t)$ denote the signal obtained by replacing time ' t ' by ' $-t$ ', as shown by:
$$y(t) = x(-t)$$
- The signal $y(t)$ represents a reflected version of $x(t)$ about the amplitude axis.

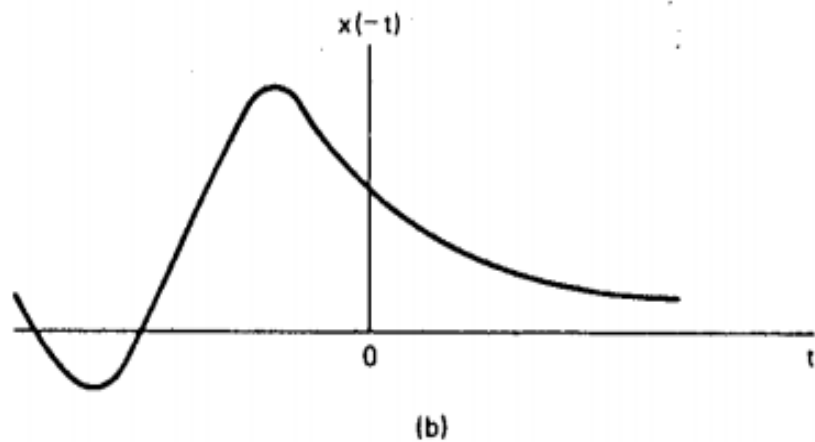
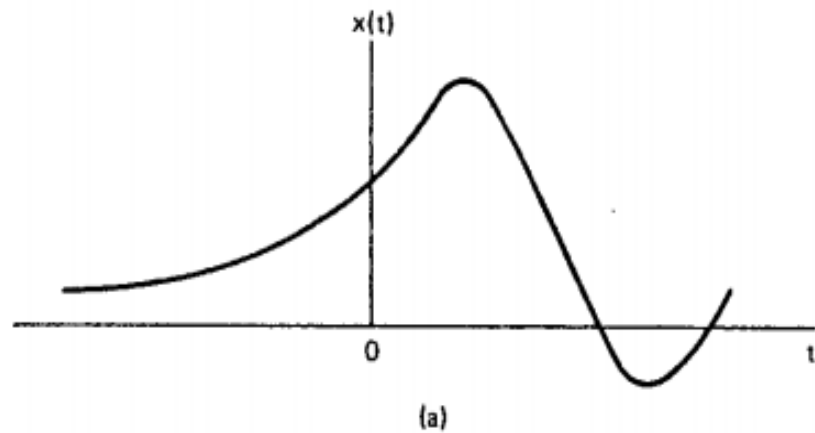


Figure 11 (a) A continuous-time signal $x(t)$; (b) its reflection, $x(-t)$, about $t = 0$.

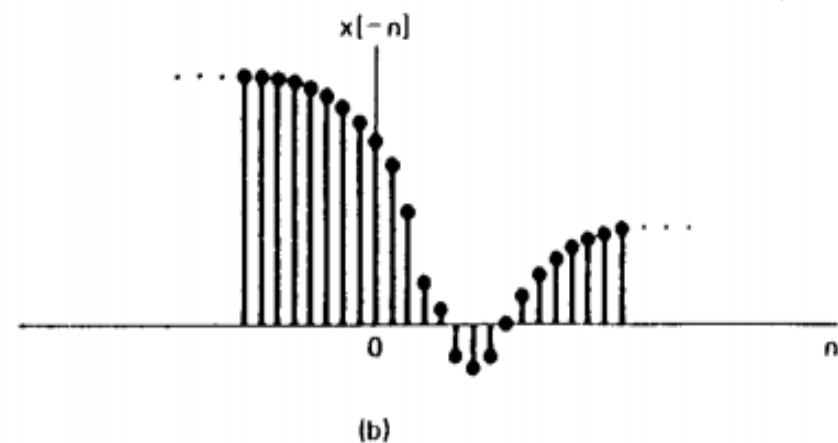
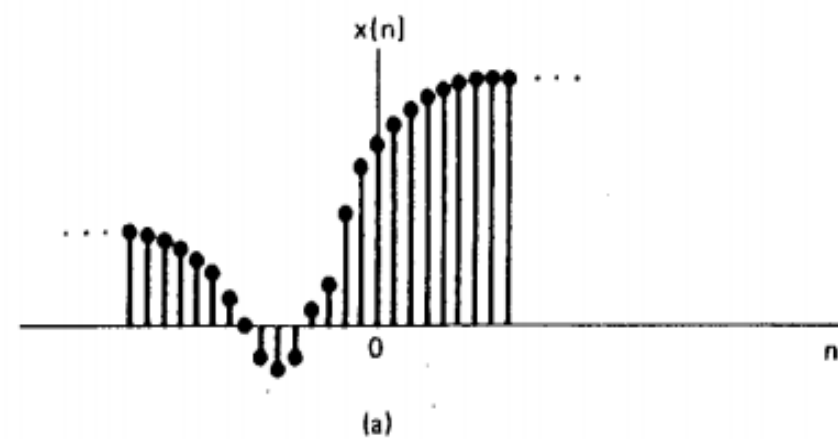


Figure 12 (a) A discrete-time signal $x[n]$; (b) its reflection, $x[-n]$, about $n = 0$.

Cont ...

- The following cases are of special interest.

➤ **Even Signals:** for which we have $x(-t) = x(t)$ for all 't'; that is, *an even signal is the same as its reflected version.*

➤ **Odd Signal:** for which we have $x(-t) = -x(t)$ for all 't'; that is, *an odd signal is the negative of its reflected version.*

- *Similar observation apply to discrete-time(DT) signals*

Example 7:

- Consider the triangular pulse $x(t)$ shown in figure 13(a).

Find the reflected version of $x(t)$ about the amplitude axis.

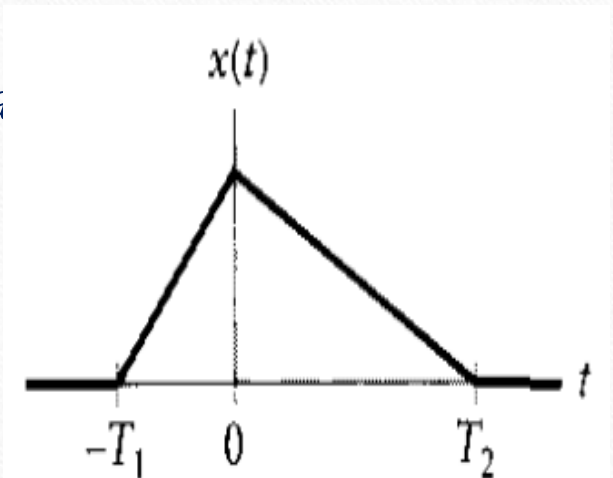


Figure 13(a) Continuous-Time Signal $x(t)$

Cont ...

Solution:

- Replacing the independent variable 't' in $x(t)$, with ' $-t$ ', we get the result $y(t) = x(-t)$ shown in figure 13(b).

Note that: for this example, we have:

$$x(t) = 0 \text{ for } t < -T_1 \text{ and } t > T_2$$

Correspondingly, we find that:

$$y(t) = 0 \text{ for } t > T_1 \text{ and } t < -T_2$$

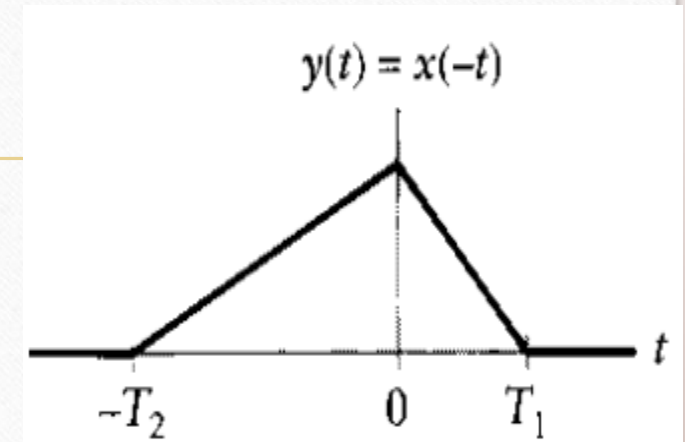


Figure 13(b) Reflected Version of $x(t)$ in figure 13(a) about Amplitude Axis

Example 8:

- Find the composite signal $y[n] = x[n] + x[-n]$ for the discrete-time signal $x[n]$ defined by:

$$i. \quad x[n] = \begin{cases} 1 & n = 1 \\ -1 & n = -1 \\ 0 & n = 0 \text{ and } |n| > 1 \end{cases}$$

$$ii. \quad x[n] = \begin{cases} 1 & n = -1 \text{ and } n = 1 \\ 0 & n = 0 \text{ and } |n| > 1 \end{cases}$$

Answer: for i: $y[n] = 0$ for all integer values of n .

$$\text{for ii: } y[n] = \begin{cases} 2 & n = -1 \text{ and } n = 1 \\ 0 & n = 0 \text{ and } |n| > 1 \end{cases}$$

Cont ...

Time Shifting:

- Let $x(t)$ denote a continuous-time signal.
- The time-shifted version of $x(t)$ is defined by:

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

Where t_o is the time shift.

- If $t_o > 0$, the waveform representing $x(t)$ is **shifted intact/whole to the right**, relative to the time axis.
- If $t_o < 0$, it is ***shifted to the left***.

Example 9:

- Figure 14(a) shows a rectangular *pulse* $x(t)$ of unit amplitude and unit duration. Find $y(t) = x(t - 2)$.

Solution:

In this example, the **time shift** t_o equals 2 time units we get the rectangular **pulse** $y(t)$ shown in *figure 14(b)*.

The pulse $y(t)$ has exactly the same shape as the *original pulse* $x(t)$; it is merely shifted along the same axis.

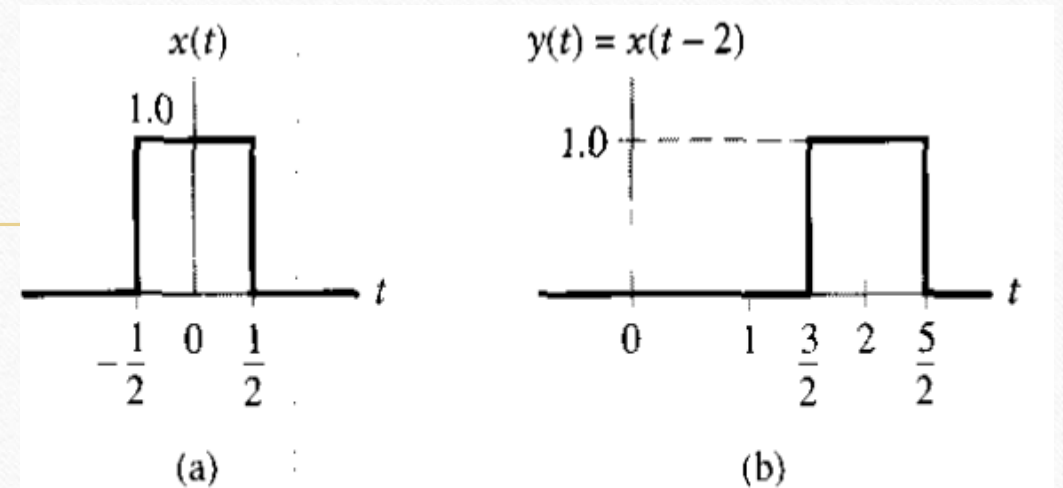


Figure 14: Time-Shifting operation.

- (a) Continuous-Time Signal in the form of a Rectangular Pulse of Amplitude 1.0 and Duration 1.0, Symmetric about the origin;
- (b) Time-Shifted Version of $x(t)$ by 2 time units.

Cont ...

- In the case of discrete-time signal $x[n]$, we define its time-shifted version as follows:

$$y[n] = x[n - m]$$

Where the shift m must be an integer; it can be positive or negative.

Example 10:

- Find the time-shifted signal $y[n] = x[n + 3]$ for the discrete-time signal $x[n]$ defined by:

$$x[n] = \begin{cases} 1 & n = 1, 2 \\ -1 & n = -1, -2 \\ 0 & n = 0, \text{ and } |n| > 2 \end{cases}$$

$$\text{Answer: } y[n] = \begin{cases} 1, & n = -1, -2 \\ -1, & n = -4, -5 \\ 0, & n = -3, n < -5, \text{ and } n > -1 \end{cases}$$

Precedence/Priority Rule For Time Shifting And Time Scaling

- Let $y(t)$ denote a continuous-time signal that is derived from another continuous-time signal $x(t)$ through a combination of time-shifting and time-scaling, as described here:

$$y(t) = x(at - b)$$

- The relationship between $y(t)$ and $x(t)$ satisfies the following conditions:

$$y(0) = x(-b) \quad \text{and} \quad y\left(\frac{b}{a}\right) = x(0)$$

Which provide useful checks on $y(t)$ in terms of corresponding values of $x(t)$.

- To correctly obtain $y(t)$ from $x(t)$, the time-shifting and time-scaling operations must be performed in the correct order.
- The proper order is based on the fact that the scaling operation always replaces “t” by “at”, while the time-shifting operation always replaces “t” by “t-b”.
- Hence the time-shifting operation is performed first on $x(t)$, resulting in an intermediate signal $v(t)$ defined by:

$$v(t) = x(t - b)$$

- The time-shift has replaced “t” in $x(t)$ by $t - b$.
- Next, the time scaling operation is performed on $v(t)$.
- This replaces “t” by “at”, resulting in the desired output:

$$y(t) = v(at) = x(at - b)$$

Example 11:

- Consider the rectangular pulse $x(t)$ of unit amplitude and duration of 2 time units depicted in figure 15. Find $y(t) = x(2t + 3)$.

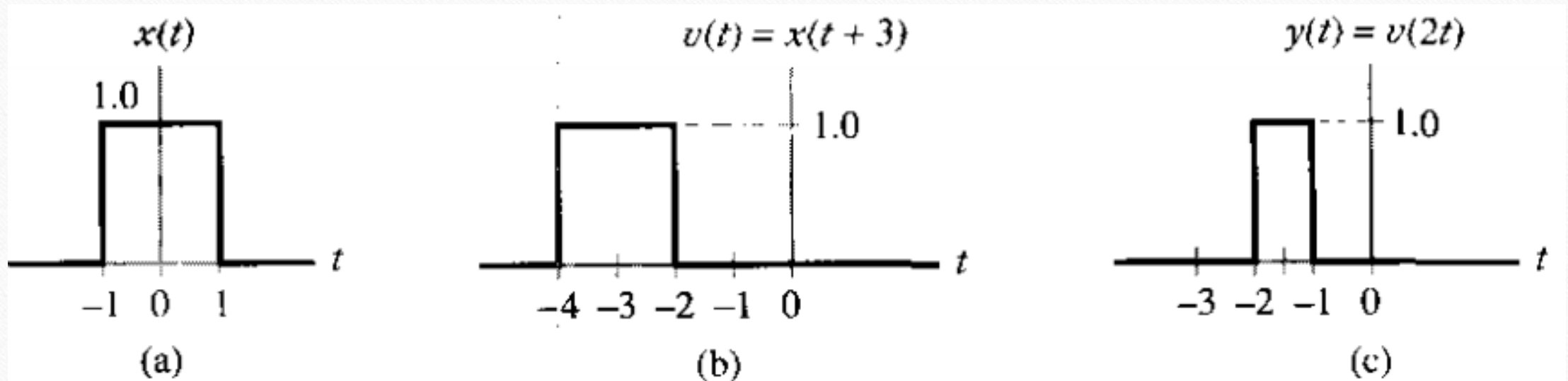


Figure 15: the proper order in which the operations of time scaling and time shifting should be applied for the case of a continuous-time signal. (a) Rectangular pulse $x(t)$ of amplitude 1.0 and duration 2.0, symmetric about the origin
(b) Intermediate pulse $v(t)$, representing time-shifted version of $x(t)$
(c) Desired signal $y(t)$, resulting from the compression of $v(t)$ by a factor of 2

Cont ...

- Suppose next that we purposely do not follow the precedence rule; that is, we first apply time scaling, followed by time shifting.
- For the given signal $x(t)$, shown in figure 16(a), the wave forms resulting from the application of these two operations are shown in figure 16(b) and (c) respectively.
- The signal $y(t)$ so obtained fails to satisfy the condition of the equation:

$$y(b/a) = x(0)$$

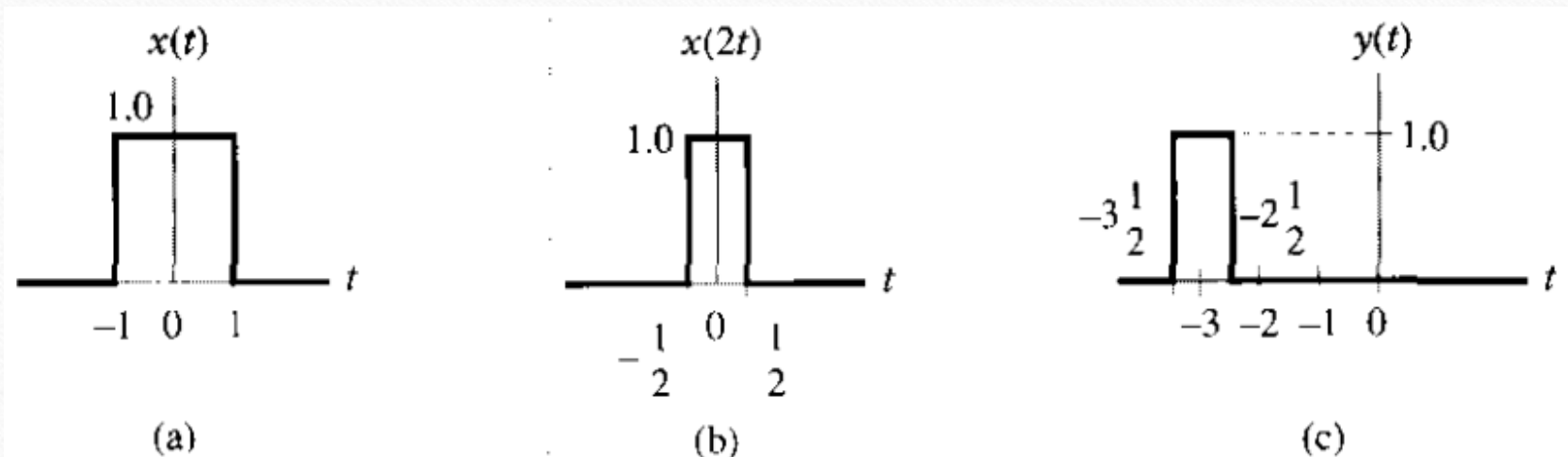


Figure 16 The incorrect way of applying the precedence rule. (a) Signal $x(t)$. (b) Time-Scaled signal $x(2t)$ (c) Signal $y(t)$ obtained by shifting $x(2t)$ by 3 time units.

Example 12:

- Find $y[n] = x[2n + 3]$ for a discrete-time signal $x[n]$ defined by:

$$x[n] = \begin{cases} 1 & n = 1, 2 \\ -1 & n = -1, -2 \\ 0 & n = 0, \text{ and } |n| > 2 \end{cases}$$

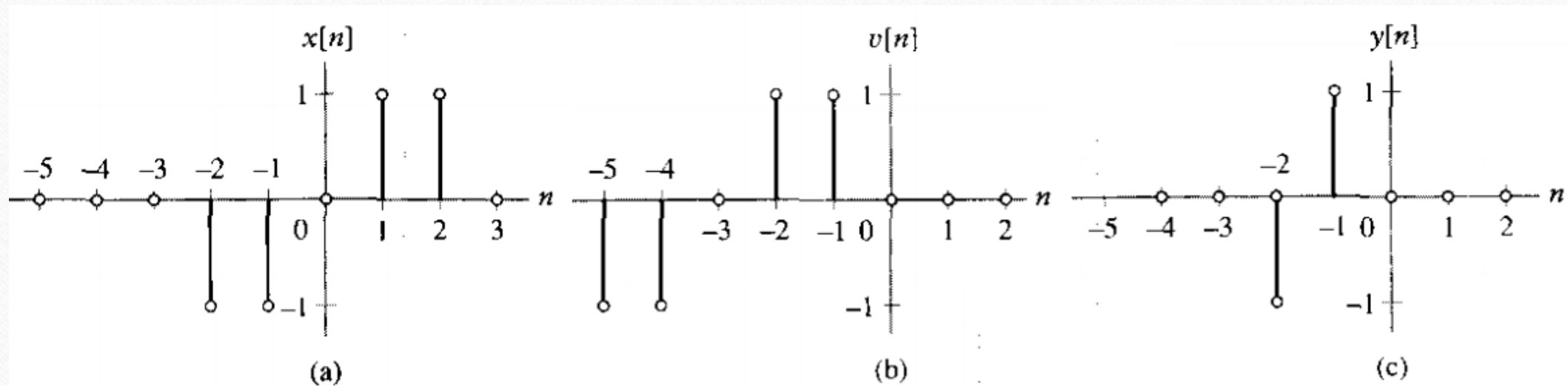


Figure for Example 12

1.3 Elementary Signals

- There are several elementary signals that feature prominently in the study of signals and system.
- The list of elementary signals includes:
 - **Exponential and Sinusoidal Signals**
 - **Step Function**
 - **Impulse Function and**
 - **Ramp Function**
- These elementary signals serve as building blocks for the construction of more complex signals.
- They are also important in their own right, in that they may be used to model physical signals that occur in nature.

Exponential Signals

- A real exponential signal, in its most general form, is written as:

$$x(t) = Be^{at}$$

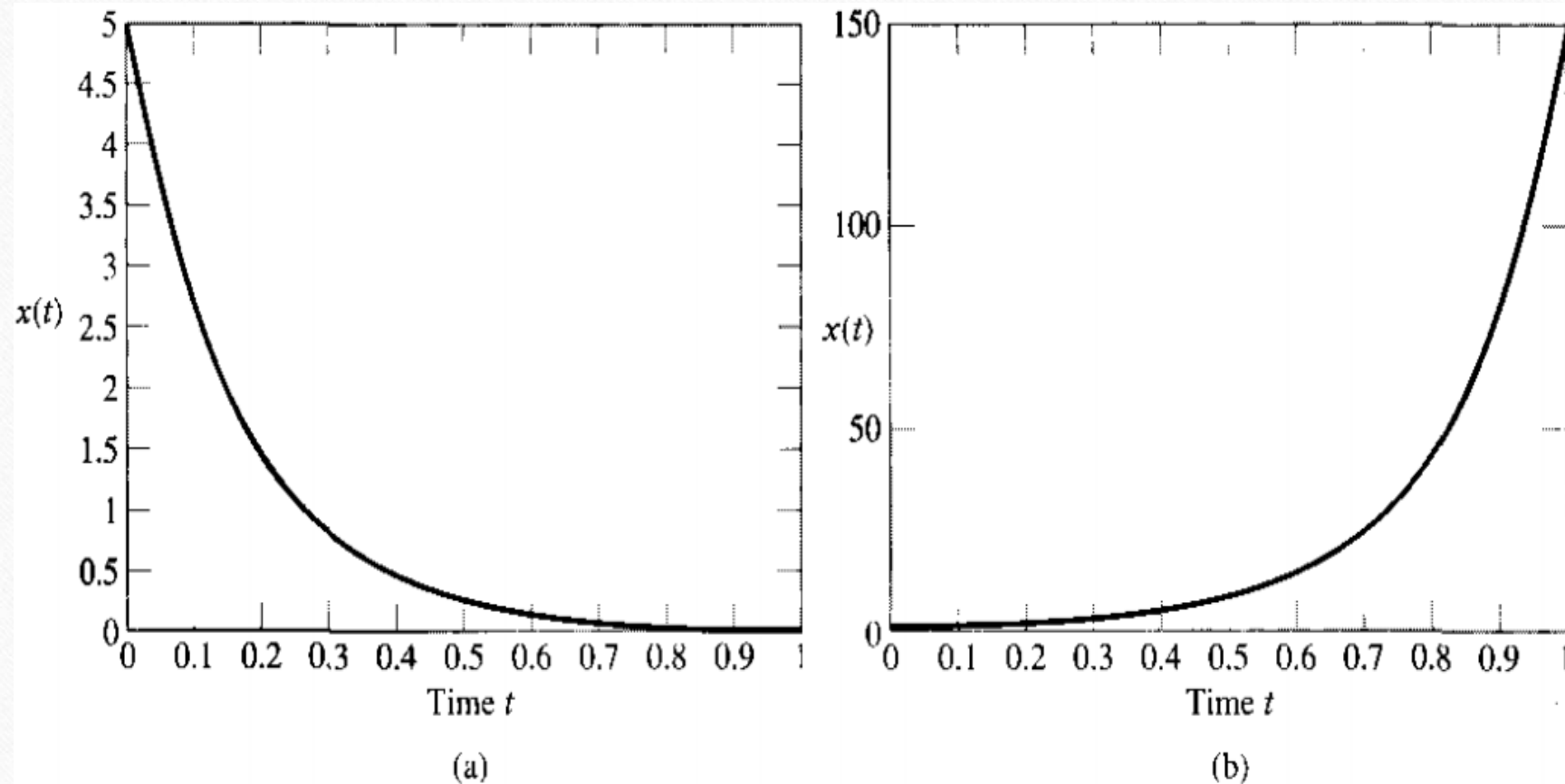
Where both “B” and “a” are real parameters.

- The parameter “B” is the amplitude of the exponential signal measured at $t = 0$.
- Depending on whether the other parameter “a” is positive or negative, we may identify two special cases:
 - Decaying Exponential, for which $a < 0$
 - Growing Exponential, for which $a > 0$

Cont ...

- These two forms of an exponential signal are illustrated in figure 18.
- If $a = 0$, the signal $x(t)$ reduces to a DC signal equal to the constant “B”.
- For a physical example of an exponential signal, consider a “lossy” capacitor, as depicted in figure 19.
 - The capacitor has capacitance C , and the loss is represented by shunt resistance R .
 - The capacitor is charged by connecting a battery across it, and then the battery is removed at $t = 0$.
 - Let V_0 denote the initial value of the voltage developed across the capacitor.

Cont ...



**Figure 18: (a) Decaying Exponential form of Continuous-Time Signal using $a = -6$ and $B = 5$.
(b) Growing Exponential form of Continuous-Time Signal using $a = 5$ and $B = 1$.**

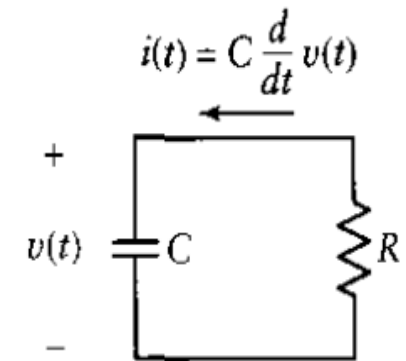


Figure 19: Loosy Capacitor with the loose represented by shunt resistance R

Cont ...

- From figure 19 we readily see that the operation of the capacitor for $t \geq 0$ is described by:

$$RC \frac{d}{dt} v(t) + v(t) = 0$$

Where $v(t)$ is the voltage measured across the capacitor at time t .

- The above equation is a differential equation of order one.
- Its solution is given by:

$$v(t) = V_o e^{-t/RC}$$

Where the product term RC plays the role of a time-constant.

Cont ...

- These equation shows that the voltage across the capacitor decays exponentially with time at a rate determined by the time constant RC .
 - The larger the resistance R (i.e. the less lossy the capacitor), the slower will be the rate of decay of $v(t)$ with time.

- In discrete-time it is common practice to write a real exponential signal as:

$$x[n] = Br^n$$

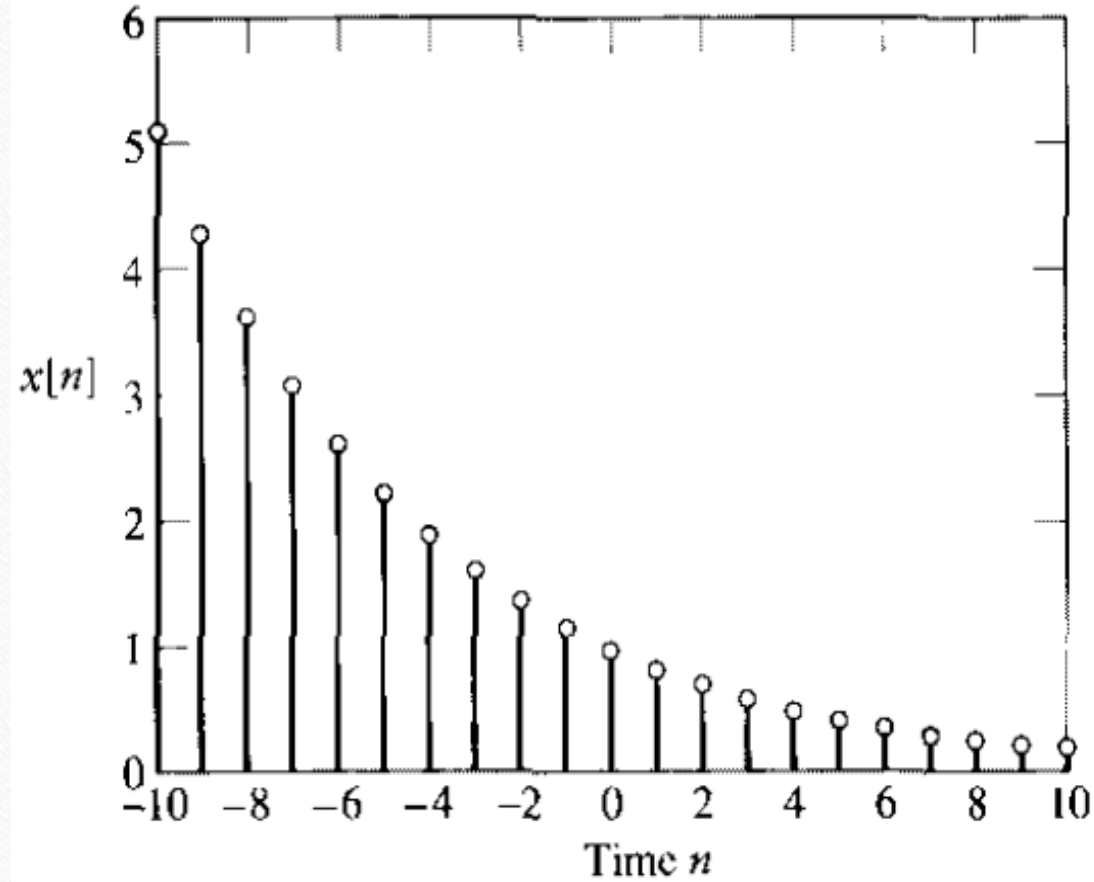
- The exponential nature of this signal is readily confirmed by defining:

$$r = e^\alpha$$

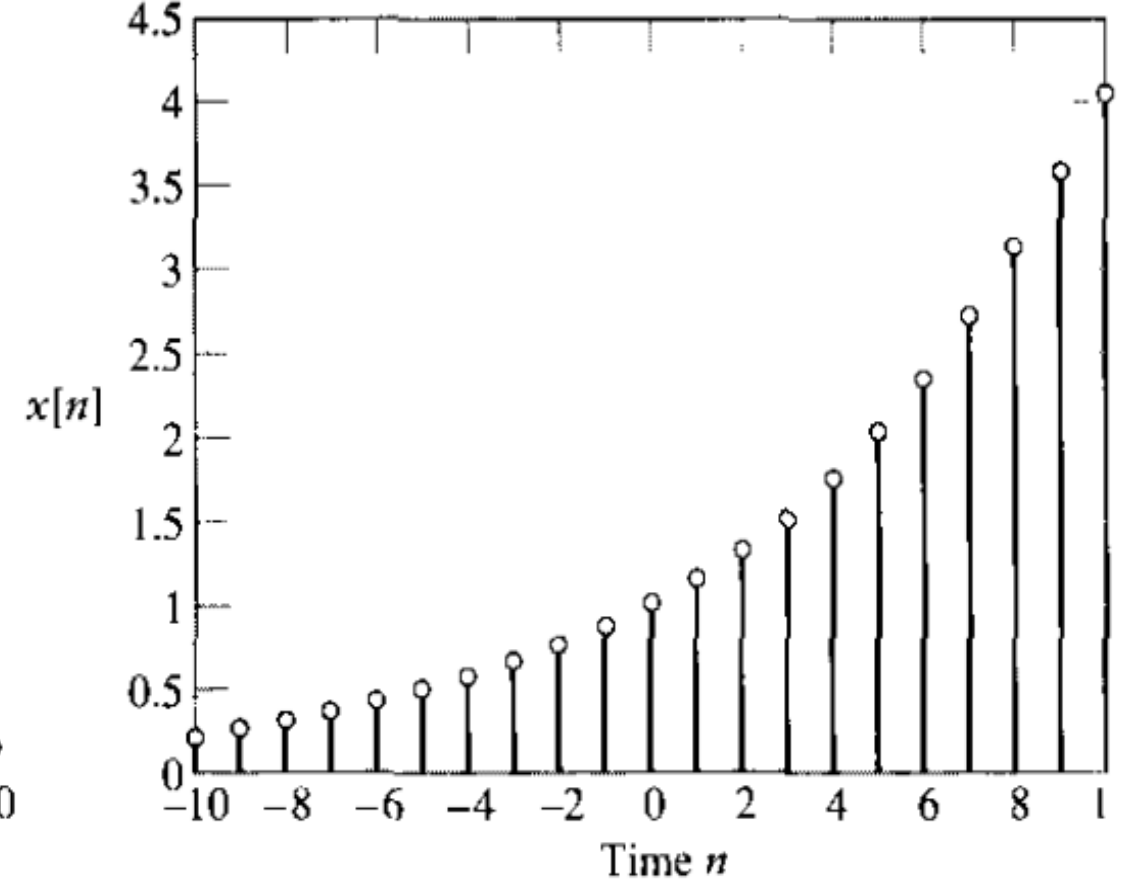
For some α .

Cont ...

- Figure 20 illustrates the decaying and growing forms of a discrete-time exponential signal corresponding to $0 < r < 1$ and $r > 1$, respectively.
- This is where the case of discrete-time exponential signals is distinctly different from continuous-time exponential signals.
- Note that: when $r < 0$, a discrete-time exponential signal assume alternating signs.
- The exponential signals shown in figures 18 and 20 are real valued
- It is possible for an exponential signal to be complex-valued.



(a)



(b)

**Figure 20: (a) Decaying Exponential form of Discrete-Time Signal
(b) Growing Exponential form of Discrete-Time Signal**

Cont ...

- The mathematical forms of complex exponential signals are the same as those shown in equations:

$$x(t) = Be^{at} \text{ and } x[n] = Br^n$$

With some differences explained here.

- In the continuous-time case, the parameter “B” or parameter “a” or both assume complex values.
- Similarly, in discrete-time case, the parameter “B” or parameter “r” or both assume complex values.
- Two commonly encountered examples of complex exponential signals are $e^{j\omega t}$ and $e^{j\Omega n}$

SINUSOIDAL SIGNAL

- The continuous-time version of a sinusoidal signal, in its most general form, may be written as:

$$x(t) = A * \cos(\omega t + \Phi)$$

Where:

- A is the amplitude
- ω is the frequency in radians per second and
- Φ is the phase angle in radians
- Figure 21(a) presents the wave form of a sinusoidal signal for $A = 4$ and $\Phi = +\pi/6$
- A sinusoidal signal is an example of a periodic signal, the period of which is:

$$T = \frac{2\pi}{\omega}$$

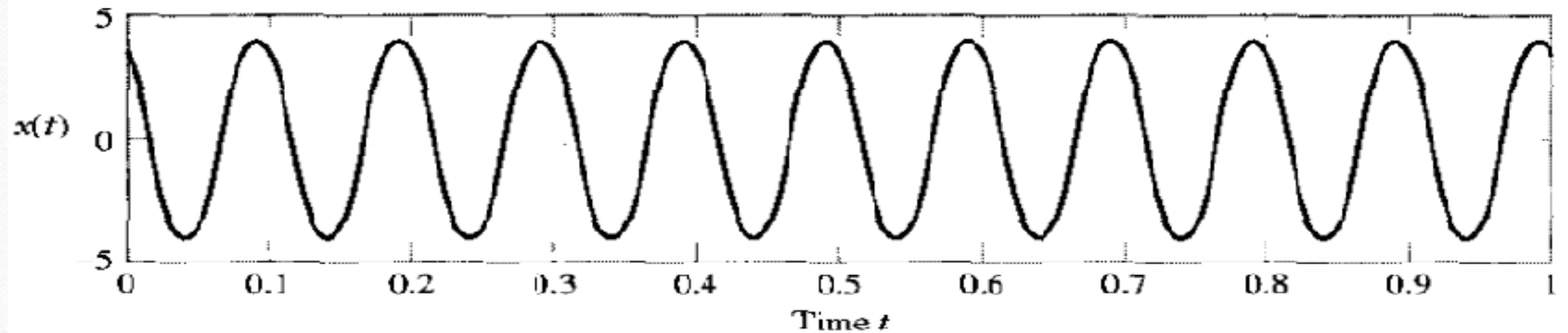


Figure 21(a): Sinusoidal Signal $A \cdot \cos(\omega t + \Phi)$ with phase $\Phi = +\pi/6$ radians

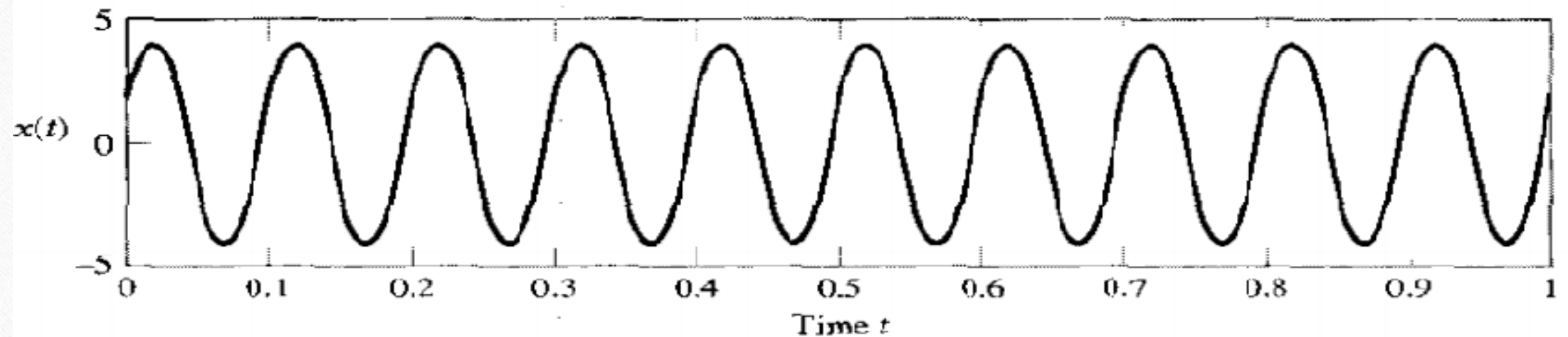


Figure 21(b): Sinusoidal Signal $A \cdot \sin(\omega t + \Phi)$ with phase $\Phi = +\pi/6$ radians

Cont ...

- We may readily prove the periodicity property of a sinusoidal signal by using the above equation to write:

$$\begin{aligned}x(t + T) &= A * \cos(\omega(t + T) + \Phi) \\&= A * \cos(\omega t + \omega T + \Phi) \\&= A * \cos(\omega t + 2\pi + \Phi) \\&= A * \cos(\omega t + \Phi) \\&= x(t)\end{aligned}$$

Which satisfies the defining condition of a periodic signal.

Cont ...

- To illustrate the generation of a sinusoidal signal, consider the circuit of figure 22 consisting of an inductor and capacitor connected in parallel.
- It is assumed that the losses in both components of the circuit are small enough for them to be considered “ideal”.
- The voltage developed across the capacitor at time $t = 0$ is equal to V_o .
- The operation of the circuit in figure 22 for $t \geq 0$ is described by:

$$LC \frac{d^2}{dt^2} v(t) + v(t) = 0$$

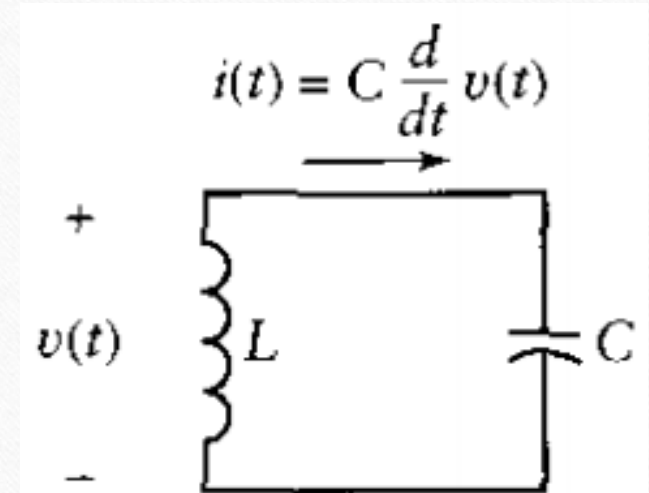


Figure 22: Parallel LC circuit

Where:

- $v(t)$ is the voltage across the capacitor at time t .
 - C is the capacitance of the capacitor,
 - L is the inductance of the inductor
-
- The above equation is a differential equation of order two.
 - Its solution is given by:

$$v(t) = V_o \cos(\omega_o t), \quad t \geq 0$$

Where ω_o is the natural angular frequency of oscillation of the circuit:

$$\omega_o = \frac{1}{\sqrt{LC}}$$

Cont ...

- The equation for instantaneous voltage, $v(t) = V_o \cos(\omega_o t)$, describes a sinusoidal signal of amplitude $A = V_o$, frequency $\omega = \omega_o$ and phase angle $\Phi = 0$.
-

- Consider next the discrete-time version of a sinusoidal signal, written as:

$$x[n] = A * \cos(\Omega n + \Phi)$$

- The period of a periodic discrete-time signal is measured in samples.
- Thus for $x[n]$ to be periodic with a period of N samples, say, it must satisfy the condition of $x[n] = x[n + N]$ for all integer “n” and some integer “N”.

Cont ...

- Substituting $n + N$ for n in the above equation yields:

$$x[n + N] = A * \cos(\Omega n + \Omega N + \Phi)$$

For the condition of periodic signal to be satisfied, in general, we require that:

$$\boxed{\Omega N = 2\pi m \quad \text{radians}}$$

$$\text{or} \quad = \frac{2\pi m}{N} \left(\frac{\text{radians}}{\text{cycle}} \right) \quad \text{integer } m, N$$

- The important point to note here is that, unlike continuous-time sinusoidal signals, not all discrete-time sinusoidal systems with arbitrary values of Ω are periodic.

- Specifically, for the discrete-time sinusoidal signal described in:

$$x[n] = A * \cos(\Omega n + \Phi)$$

to be periodic, the angular frequency Ω must be a rational multiple of 2π as indicated in:

$$\Omega = \frac{2\pi m}{N}$$

Figure 23 illustrates a discrete-time sinusoidal signal for $A = 1$, $\Phi = 0$ and $N = 12$.

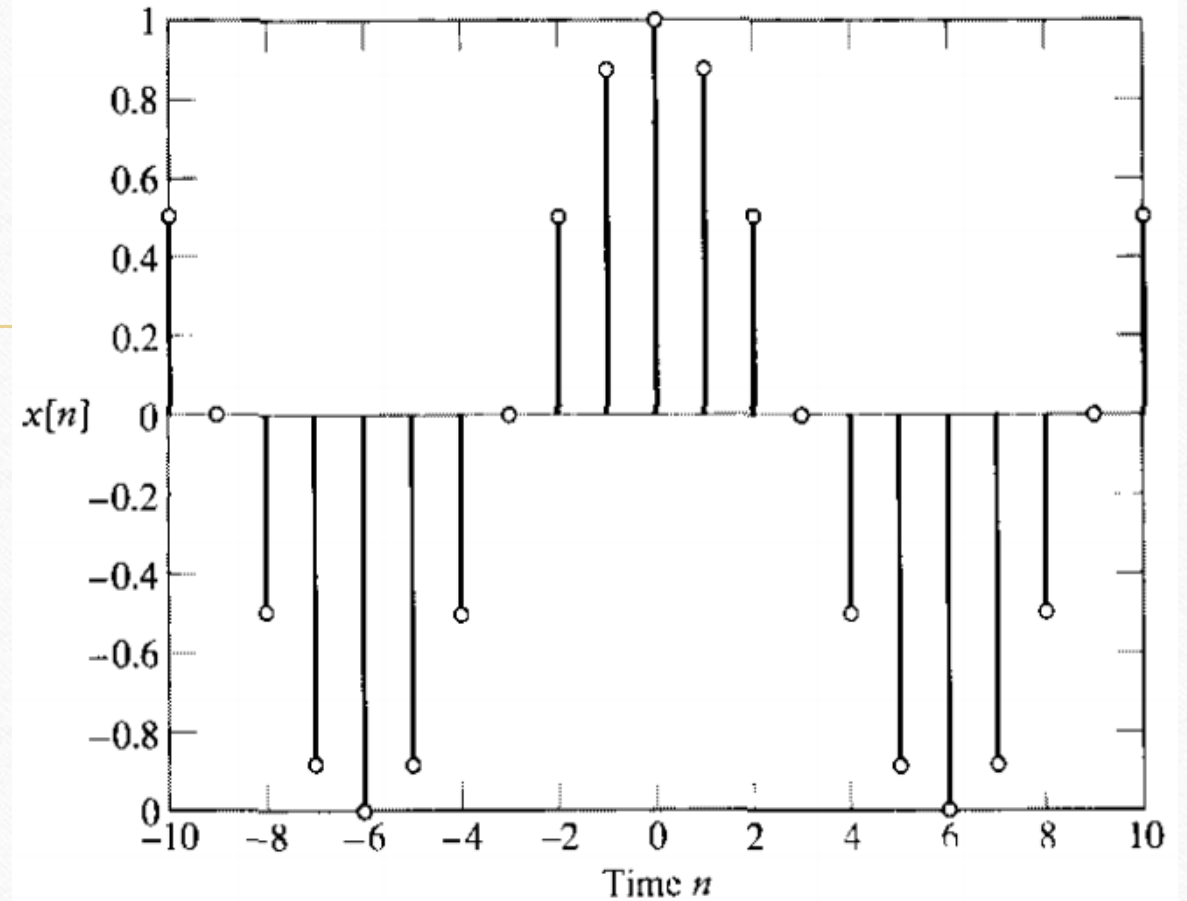


Figure 23: Discrete-Time Sinusoidal Signal

Example 13:

- A pair of sinusoidal signals with a common angular frequency is defined by:

$$x_1[n] = \sin[5\pi n] \quad \text{and} \quad x_2[n] = \sqrt{3}\cos[5\pi n]$$

- a) Specify the condition which the period N of both $x_1[n]$ and $x_2[n]$ must satisfy for them to be periodic.
- b) Evaluate the amplitude and phase angle of the composite sinusoidal signal:

$$y[n] = x_1[n] + x_2[n]$$

Solution:

- a) The angular frequency of both $x_1[n]$ and $x_2[n]$ is:

$$\Omega = 5\pi \text{ radians/cycle}$$

Cont ...

- Solving for the period N:

$$N = \frac{2\pi m}{\Omega} = \frac{2\pi m}{5\pi} = \frac{2m}{5}$$

- For $x_1[n]$ and $x_2[n]$ to be periodic, their period N must be an integer.
- This can only be satisfied for $m = 5, 10, 15, \dots$ which results $N = 2, 4, 6, \dots$

- b) We wish to express $y[n]$ in the form:

$$y[n] = A * \cos(\Omega n + \Phi)$$

Recall the trigonometric identity:

$$A * \cos(\Omega n + \Phi) = A * \cos(\Omega n) * \cos(\Phi) - A * \sin(\Omega n) * \sin(\Phi)$$

Cont ...

- Identifying $\Omega = 5\pi$, we see that the right-hand side of this identity is of the same form as $x_1[n] + x_2[n]$.
- We may therefore write:

$$A \sin(\Phi) = -1 \quad \text{and} \quad A \cos(\Phi) = \sqrt{3}$$

Hence:

$$\tan(\Phi) = \frac{\sin(\Phi)}{\cos(\Phi)} = \frac{\text{Amplitude of } x_1[n]}{\text{Amplitude of } x_2[n]} = \frac{-1}{\sqrt{3}}$$

For which we find that $\Phi = -\pi/6$ radians.

Cont ...

- Similarly, the amplitude A is given by:

$$A = \sqrt{(\text{Amplitude of } x_1[n])^2 + (\text{Amplitude of } x_2[n])^2} = \sqrt{1 + 3} = 2$$

Accordingly, we may express $y[n]$ as: $y[n] = 2\cos(5\pi n - \pi/6)$

Example 14: Consider the following sinusoidal signals. Determine whether each $x[n]$ is periodic, and if it is, find its fundamental period.

a) $x[n] = 5 \sin[2n]$

b) $x[n] = 5\cos[0.2\pi n]$

c) $x[n] = 5 \cos[6\pi n]$

d) $x[n] = 5\sin[6\pi n/35]$

Cont ...

Answer:

(a) Non-periodic (b) Periodic, fundamental period = 10. (c) Periodic,
fundamental period = 1. (d) Periodic, fundamental period = 35.

RELATION BETWEEN SINUSOIDAL AND COMPLEX EXPONENTIAL SIGNALS

- Consider the complex exponential $e^{j\theta}$.
- Using Euler's identity, we may expand this term as:

$$e^{j\theta} = \cos\theta + j\sin\theta$$

- This result indicates that, we may express the continuous-time sinusoidal signal of:

$$x(t) = A * \cos(\omega t + \Phi)$$

as the real part of the complex exponential signal:

$$B e^{j\omega t}$$

Cont ...

Where B is itself a complex quantity defined by:

$$B = Ae^{j\Phi}$$

- That is, we may write:

$$A \cos(\omega t + \Phi) = \text{Re}\{Be^{j\omega t}\}$$

Where $\text{Re}\{ \}$ denotes the real part of the complex quantity enclosed inside the braces.

- We may readily prove this relation by noting that:

$$\begin{aligned} Be^{j\omega t} &= Ae^{j\Phi} e^{j\omega t} \\ &= Ae^{j(\omega t + \Phi)} \\ &= A \cos(\omega t + \Phi) + jA \sin(\omega t + \Phi) \end{aligned}$$

Cont ...

- Previously the sinusoidal signal is defined in terms of a cosine function.
- Of course, we may also define a continuous time sinusoidal signal in terms of a sine function, as shown by:

$$x(t) = A \sin(\omega t + \Phi)$$

- Which is represented by the imaginary part of the complex exponential signal $Be^{j\omega t}$.
- That is, we may write:

$$A \sin(\omega t + \Phi) = \text{Im}\{Be^{j\omega t}\}$$

Where $\text{Im}\{ \}$ denotes the imaginary part of the complex quantity enclosed inside the braces.

Cont ...

- The sinusoidal signal $A \sin(\omega t + \Phi)$ differs from that of $A \cos(\omega t + \Phi)$ by a phase shift of 90° .
- That is, the sinusoidal signal $A \cos(\omega t + \Phi)$ leads the sinusoidal signal $A \sin(\omega t + \Phi)$, as illustrated in figure 21 for $\Phi = \pi/6$.
- Similarly, in the discrete-time case we may write:

$$A \cos(\Omega n + \Phi) = \text{Re} \{ B e^{j\Omega n} \} \quad \text{and} \quad A \sin(\Omega n + \Phi) = \text{Im} \{ B e^{j\Omega n} \}$$

Where B is defined in terms of A and Φ by $B = A e^{j\Phi}$.

Cont ...

- Figure 24 shows the two dimensional representation of complex exponential $e^{j\Omega n}$ for $\Omega = \pi/4$ and $n = 0, 1, \dots, 7$.
- The projection of each value on the real axis is $\cos(\Omega n)$, while the projection on the imaginary axis is $\sin(\Omega n)$.

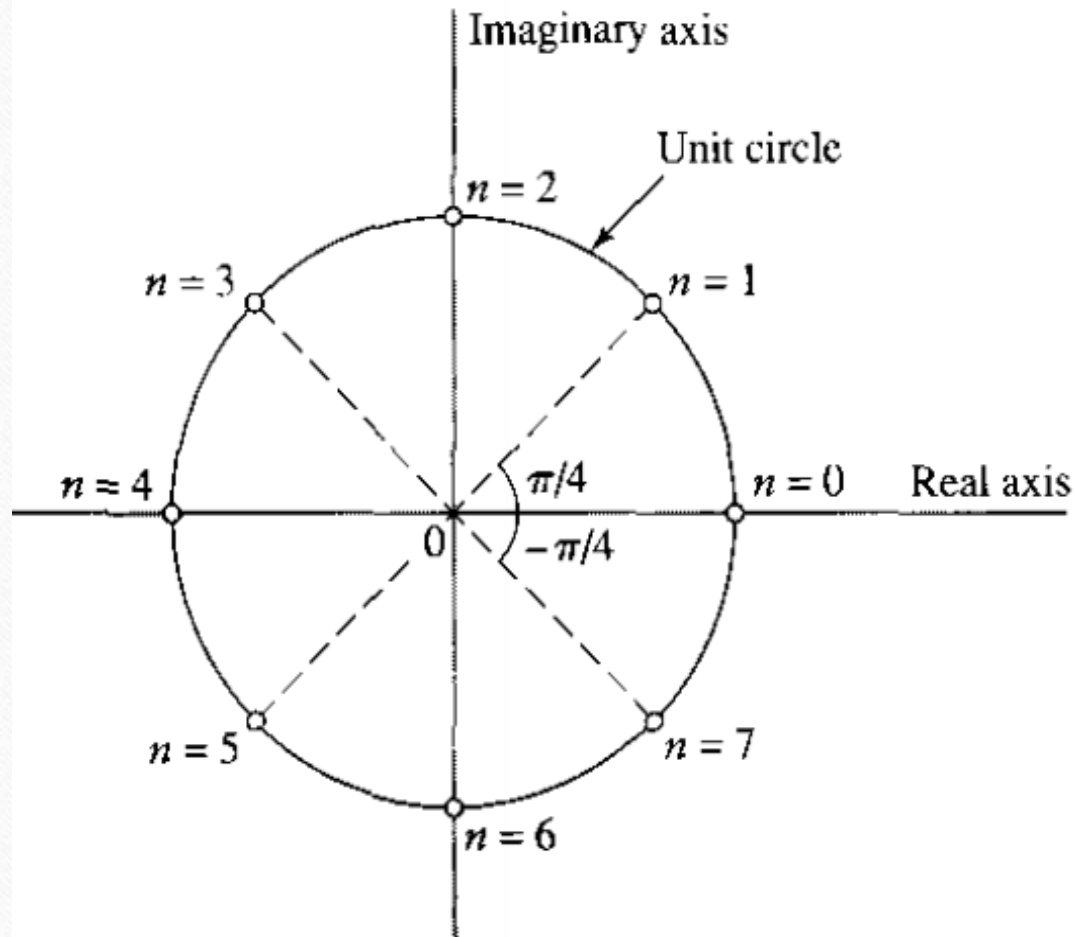


Figure 24: Complex plane, showing eight points uniformly distributed on the unit circle

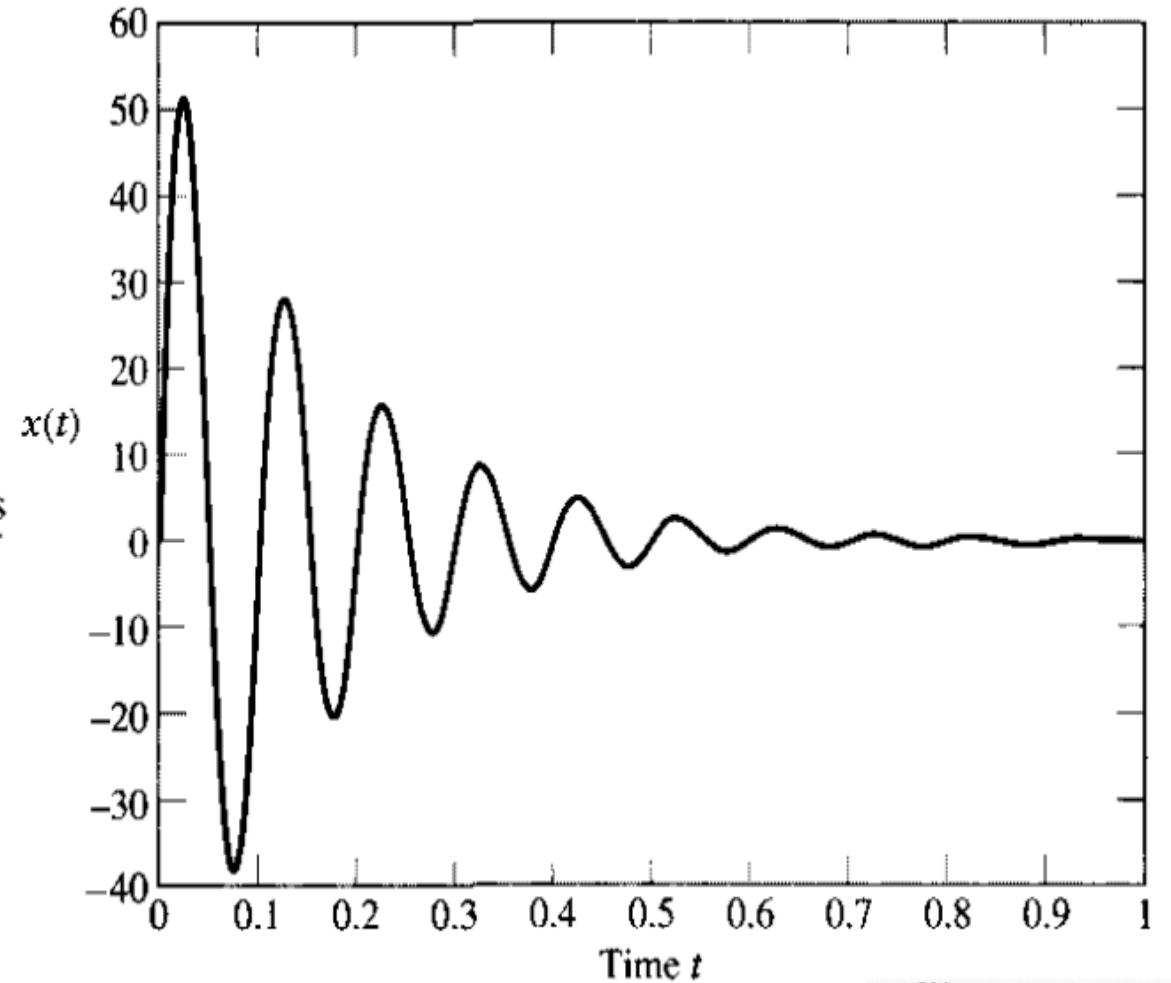


Figure 25: Exponentially damped sinusoidal signal $e^{-\alpha t} \sin(\omega t + \Phi)$ with $\alpha > 0$

EXPONENTIALLY DAMPED SINUSOIDAL SIGNAL

- The multiplication of a sinusoidal signal by a real-valued decaying exponential signal results in a new signal referred to as an exponentially damped sinusoidal signal.
- Specifically, multiplying the continuous-time sinusoidal signal $A\sin(\omega t + \Phi)$ by the exponential $e^{-\alpha t}$ results in the exponentially damped sinusoidal signal:

$$x(t) = Ae^{-\alpha t}\sin(\omega t + \Phi), \quad \alpha > 0$$

- Figure 25 shows the wave form of this signal for $A = 60, \alpha = 6$ and $\Phi = 0$.
- For increasing time “t”, the amplitude of the sinusoidal oscillations decreases in an exponential fashion, approaching zero for infinite time.

Cont ...

- To illustrate the generation of an exponentially damped sinusoidal signal, consider the parallel circuit of figure 26, consisting of a capacitor of capacitance C , an inductor of inductance L , and a resistor of resistance R .
- The resistance R represents the combined effect of losses associated with the inductor and the capacitor.
- Let V_0 denote the voltage developed across the capacitor at time $t = 0$.
- The operation of the circuit in figure 26 is described by:

$$C \frac{d}{dt} v(t) + \frac{1}{R} v(t) + \frac{1}{L} \int_{-\infty}^t v(\tau) d\tau = 0$$

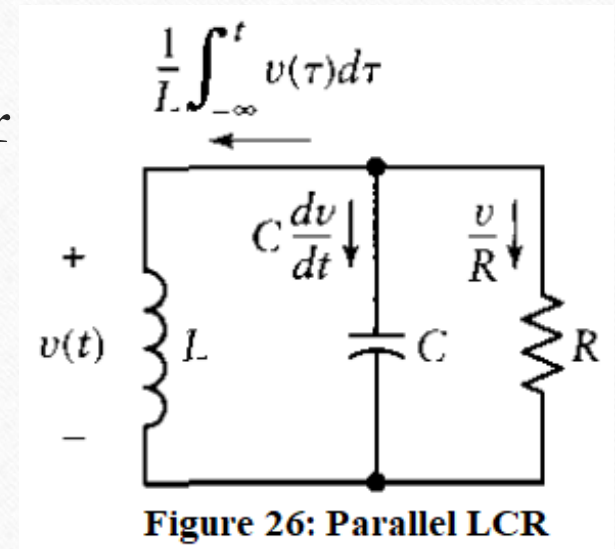


Figure 26: Parallel LCR

Cont ...

- Where $v(t)$ is the voltage across the capacitor at time $t \geq 0$.
- The above equation is an integro-differential equation.
- Its solution is given by:

$$v(t) = V_o e^{-t/2RC} \cos(\omega_o t)$$

Where:

$$\omega_o = \sqrt{\frac{1}{LC} - \frac{1}{4C^2R^2}}$$

In the last equation it is assumed that $4CR^2 > L$

Cont ...

- Comparing $x(t) = Ae^{-\alpha t} \sin(\omega t + \Phi)$ and $v(t) = V_o e^{-t/2RC} \cos(\omega_o t)$; we have:

$$A = V_o, \quad \alpha = 1/2RC, \quad \omega = \omega_o \quad \text{and} \quad \Phi = \pi/2.$$

- Returning to the subject matter at hand, the discrete-time version of the exponentially damped sinusoidal signal is described by:

$$x[n] = Br^n \sin[\Omega n + \Phi]$$

For the signal to decay exponentially with time, the parameter r must lie in the range $0 < r < 1$.