



**KIT- KalaigmarKarunanidhi Institute of Technology**

(An Autonomous Institution, Affiliated to Anna University Chennai)

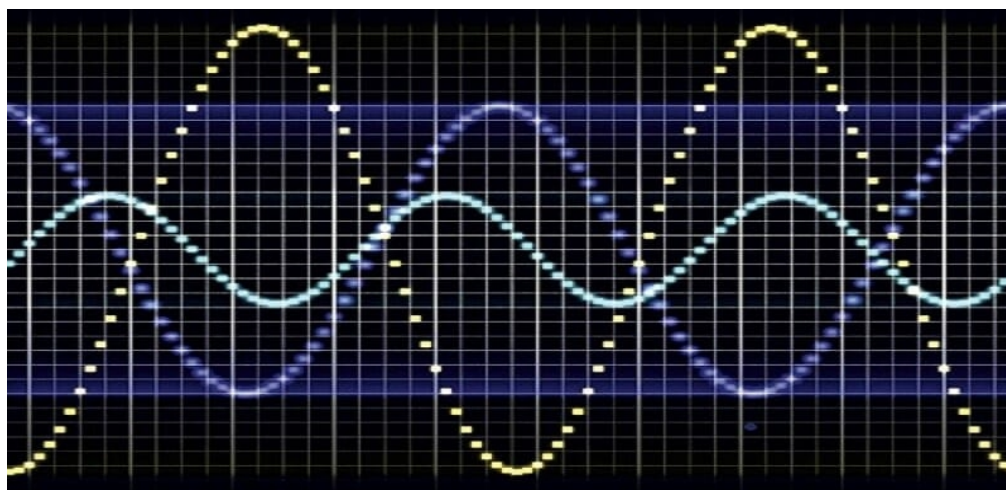
Coimbatore – 641402

**Department of Electronics and Communication Engineering**



**B23ECP303**

**SIGNALS AND SYSTEMS LABORATORY**



## LABORATORY RECORD

**Academic Year 2024-25 (Odd Semester)**

**Student's Name** : .....

**Register Number** : .....

**Year / Semester / Sec** : .....



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Coimbatore – 641402

**Department of Electronics and Communication Engineering**



## **BONAFIDE CERTIFICATE**

Department of

.....

Record Work of .....

Laboratory Certified that this record is the bona fide work done by

Name: .....

Class: ..... Roll No: .....

Branch.....

Place: KIT, CBE

Faculty In-Charge

HoD / ECE

University Register No.....

Submitted for the University Practical Examination held on

.....

**INTERNAL EXAMINER**

**EXTERNAL EXAMINER**



# KIT- Kalaigarkarunanidhi Institute of Technology

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Coimbatore – 641402

Department of Electronics and Communication Engineering



## Syllabus

### LIST OF EXPERIMENTS

1	Build a program to generate elementary , continuous and discrete signal
2	Generate discrete signal and calculate energy/ Power of a signal.
3	Understand the properties and the different representations of LTI systems
4	Understand the concepts of convolution
5	Analyse the effects of sampling in the time and frequency domains
6	Understand the basics of Uniform Quantization
7	Analysis and Synthesis of Signals using Discrete Fourier Transform
8	Analyse the ECG signal to find heart rate of a patient

**Total Instructional hours: 30**

### **Course Outcomes: Students will be able to**

CO1	Examine the mathematical concepts on signals and systems using appropriate tools
CO2	Involve in independent / team work, communicate effectively and engage in lifelong learning

### **List of Equipment Required; Requirements for a Batch of 30 Students**

Sl. No.	Description of the Equipment	Quantity required (Nos.)
1.	Computer	30
2.	Software — Python/ Appropriate tools	-



# KIT- KalaigmarKarunanidhi Institute of Technology

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Coimbatore – 641402

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## Rubrics for the Evaluation of Laboratory

Ex. No.	Date	Name of the Experiment	Page Number	Algorithm(20marks)	Output & Conclusion (25Marks)	Inference (10 marks)	Viva Voce (20 marks)	Total (75 marks)	Signature of the Faculty Member
1		Build a program to generate elementary , continuous and discrete signal							
2		Generate discrete signal and calculate energy/ Power of a signal.							
3		Understand the properties and the different representations of LTI systems							
4		Understand the concepts of convolution							
5		Analyse the effects of sampling in the time and frequency domains							
6		Understand the basics of Uniform Quantization							

Ex. No.	Date	Name of the Experiment	Page Number	Algorithm(20marks)	Output & Conclusion (25Marks)	Inference (10 marks)	Viva Voce (20 marks)	Total (75 marks)	Signature of the Faculty Member
7		Analysis and Synthesis of Signals using Discrete Fourier Transform							
8		Analyse the ECG signal to find heart rate of a patient							

Model Exam Marks (25):\_\_\_\_\_ Total (100):\_\_\_\_\_

Signature of the Faculty Member

EX NO : 01	Build a program to generate elementary , continuous and discrete signal
DATE:	

### AIM:

To Build a program to generate elementary , continuous and discrete signal

### CONTINUOUS TIME - UNIT IMPULSE

### ALGORITHM:

1. Start the program and import the numpy and matplotlib libraries.
2. Define the discrete time range np. arrange
3. Generate a discrete unit impulse signal using signal.unit\_impulse().  
Specify the length of the impulse based on the length of the time array t and set the index to 'mid' to position the impulse at the center.
4. Plot the signal plt. plot() to create a stem plot of continuous signal x(n) and set x- axis and y-axis.
5. Display the plot.
6. Stop the program

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

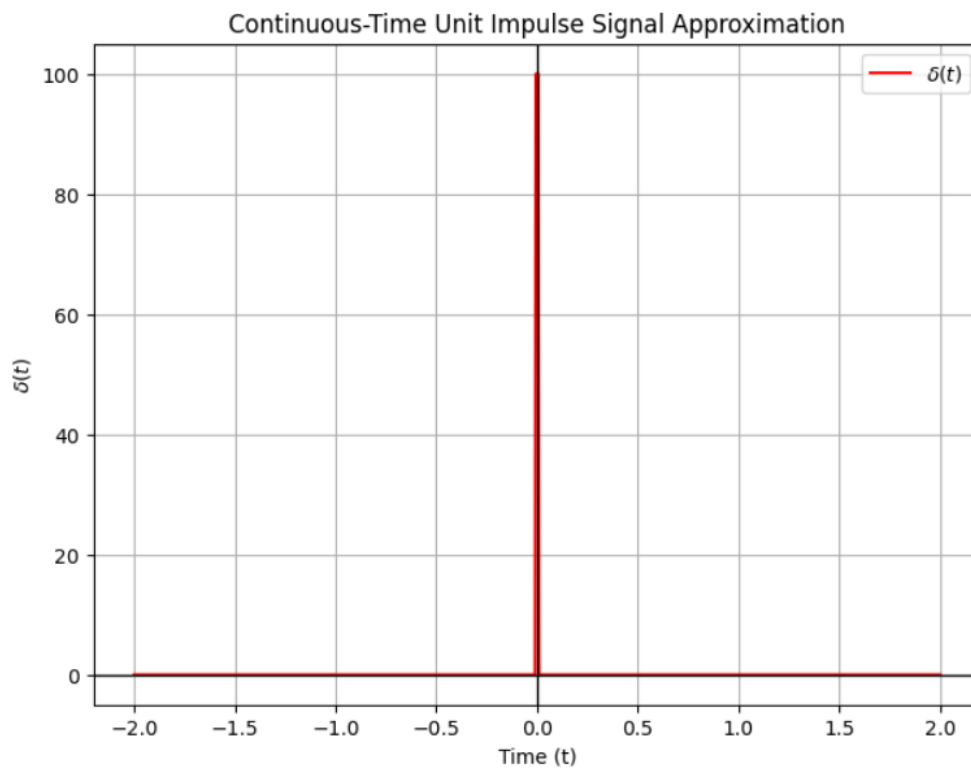
# Time range (continuous time)
t = np.linspace(-2, 2, 1000) # Time from -2 to 2 seconds with 1000 points
epsilon = 0.01 # Small value to simulate the width of the impulse

# Continuous-time unit impulse signal approximation
impulse = np.zeros_like(t) # Start with zeros
impulse[np.abs(t) < epsilon] = 1 / epsilon # Set a very narrow spike at t = 0

# Plot the continuous-time unit impulse signal
plt.figure(figsize=(8, 6))
```

```
plt.plot(t, impulse, label=r'$\delta(t)$', color='red')
plt.axhline(0, color='black', linewidth=1)
plt.axvline(0, color='black', linewidth=1)
plt.title("Continuous-Time Unit Impulse Signal Approximation")
plt.xlabel("Time (t)")
plt.ylabel(r"$\delta(t)$")
plt.grid(True)
plt.legend()
plt.show()
```

OUTPUT:





## DISCRETE TIME - UNIT IMPULSE

### ALGORITHM:

1. Start the program and import the numpy and matplotlib libraries.
2. Define the discrete time range np. arrange
3. Generate a discrete unit impulse signal using signal.unit\_impulse().  
Specify the length of the impulse based on the length of the time array t and set the index to 'mid' to position the impulse at the center.
4. Plot the signal plt. stem() to create a stem plot of Discrete signal  $x(n)$  and set x- axis and y-axis.
5. Display the plot.
5. Stop the program.

### PROGRAM:

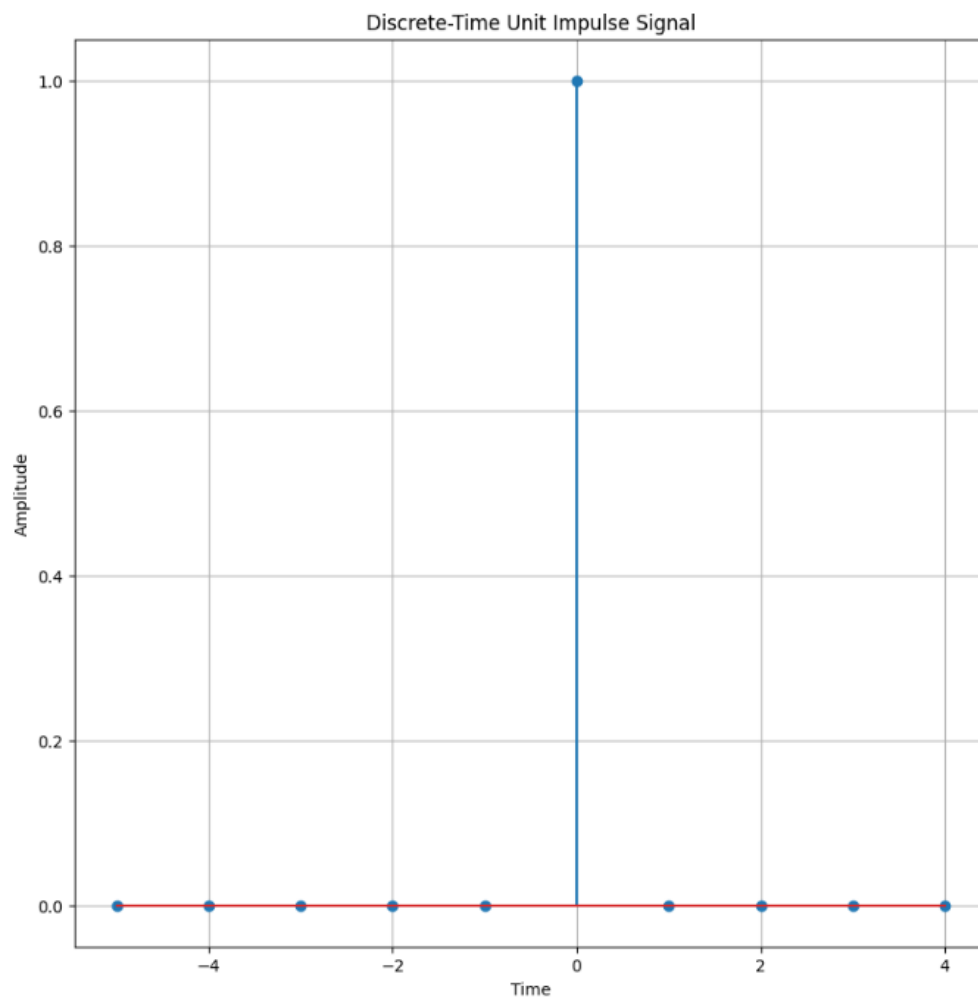
```
import numpy as np
import matplotlib.pyplot as plt
from scipy import signal # Import the signal module from SciPy

# Time values
n = np.arange(-5, 5, 1) # Adjust range and step as needed

# Unit impulse signal
impulse_signal = signal.unit_impulse(len(n), idx='mid')

# Plot the signal
plt.figure(figsize=(10,10))
plt.stem(n, impulse_signal)
plt.xlabel('Time')
plt.ylabel('Amplitude')
plt.title('Discrete-Time Unit Impulse Signal')
plt.grid(True)
plt.show()
```

OUTPUT:



## CONTINUOUS TIME - UNIT PARABOLIC

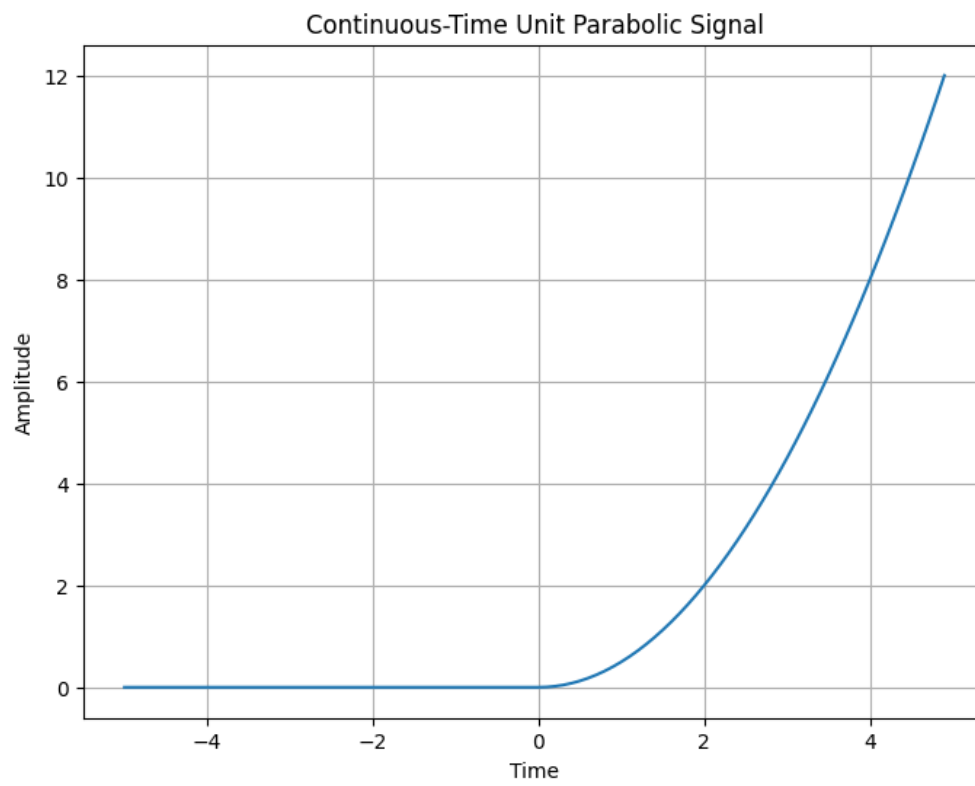
### ALGORITHM:

1. Start the program and import the numpy and matplotlib libraries.
2. Define the discrete time range np. arrange
3. Create an array n that spans from -5 to 5 with a step of 0.1.
- 4.. Generate a continuous unit parabolic signal using the condition for non-negativity.
5. Plot the signal plt. plot() to create a stem plot of continuous signal x(n) and set x- axis and y-axis.
- 6.Display the plot.
7. Stop the program.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt
# Time values
t = np.arange(-5, 5, 0.1)
# Unit parabolic signal
parabolic_signal = 0.5 * t**2 * (t >= 0) # Condition for non-negativity
# Plot
plt.figure(figsize=(8, 6))
plt.plot(t, parabolic_signal)
plt.xlabel('Time')
plt.ylabel('Amplitude')
plt.title('Continuous-Time Unit Parabolic Signal')
plt.grid(True)
plt.show()
```

OUTPUT:



## DISCRETE TIME - UNIT PARABOLIC

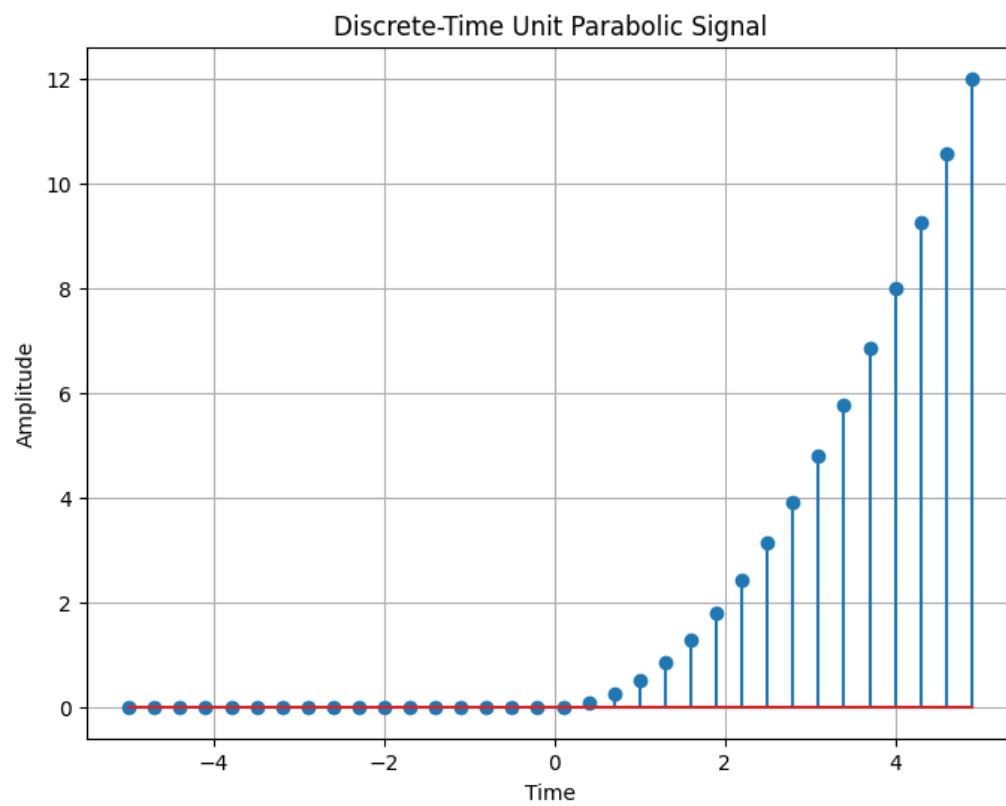
### ALGORITHM:

1. Start the program and import the numpy and matplotlib libraries.
2. Define the continuous time range np. arrange
3. Create an array n that spans from -5 to 5 with a step of 0.3.
- 4.. Generate a continuous unit parabolic signal using the condition for non-negativity.
5. Plot the signal plt. stem() to create a stem plot of discrete signal x(n) and set x- axis and y-axis.
- 6.Display the plot.
7. Stop the program.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt
# Time values
n = np.arange(-5, 5, 0.3)
# Unit parabolic signal
parabolic_signal = 0.5 * n**2 * (n >= 0) # Condition for non-negativity
# Plot
plt.figure(figsize=(8, 6))
plt.stem(n, parabolic_signal)
plt.xlabel('Time')
plt.ylabel('Amplitude')
plt.title('Discrete-Time Unit Parabolic Signal')
plt.grid(True)
plt.show()
```

OUTPUT:



## CONTINUOUS TIME - UNIT RAMP

### ALGORITHM:

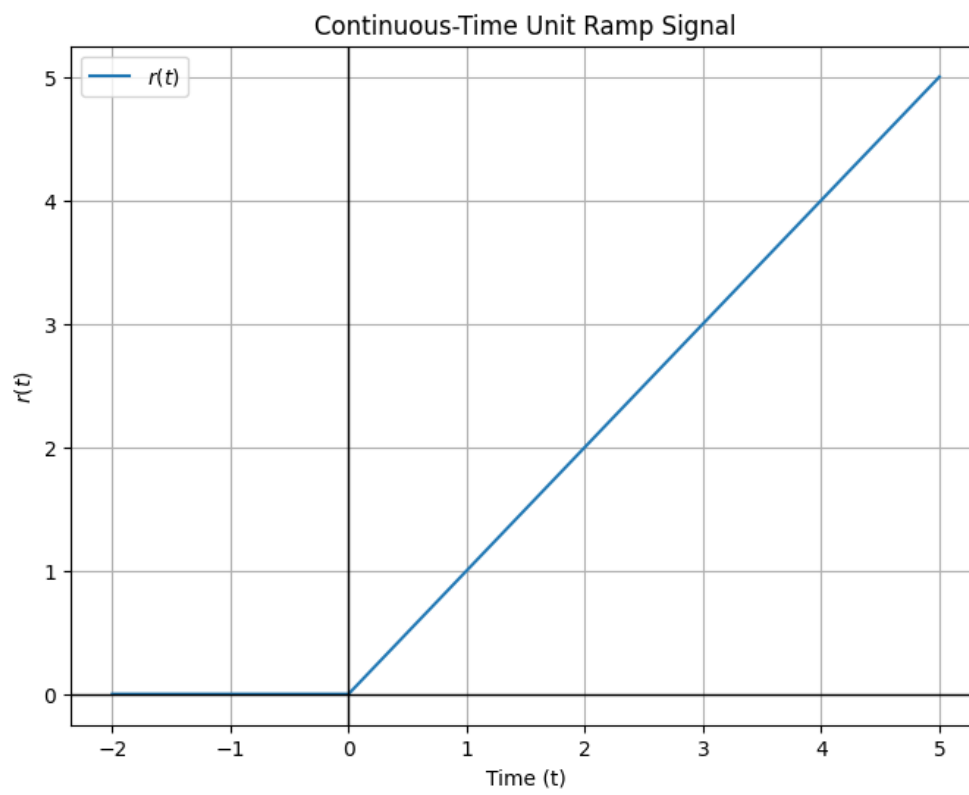
1. Start the program and import the numpy and matplotlib libraries.
2. Define the discrete time range np. arrange
3. Create an array t that spans from -5 to 10 with a step of 0.1.
4. Generate a continuous unit ramp signal using np.where by setting values to 1 when t is greater than or equal to 0, and 0 otherwise.
5. Plot the signal plt. plot() to create a plot of continuous signal x(n) and set x- axis and y-axis.
6. Display the plot.
7. Stop the program.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

# Time range (continuous time)
t = np.linspace(-2, 5, 1000) # Time from -2 to 5 seconds with 1000 points
# Continuous-time unit ramp signal r(t)
r = np.maximum(0, t) # r(t) = t for t >= 0, and 0 otherwise
# Plot the continuous-time unit ramp signal
plt.figure(figsize=(8, 6))
plt.plot(t, r, label=r'$r(t)$')
plt.axhline(0, color='black', linewidth=1)
plt.axvline(0, color='black', linewidth=1)
plt.title("Continuous-Time Unit Ramp Signal")
plt.xlabel("Time (t)")
plt.ylabel(r"$r(t)$")
plt.grid(True)
plt.legend()
plt.show()
```

OUTPUT:





## DISCRETE TIME - UNIT RAMP

### ALGORITHM:

1. Start the program and import the numpy and matplotlib libraries.
2. Define the discrete time range np. arrange
3. Create an array n that spans from -5 to 10.
4. Generate a discrete unit ramp signal using np.where by setting values to 1 when n is greater than or equal to 0, and 0 otherwise.
5. Plot the signal plt. stem() to create a stem plot of digital signal x(n) and set x- axis and y-axis.
6. Display the plot.
7. Stop the program.

### PROGRAM:

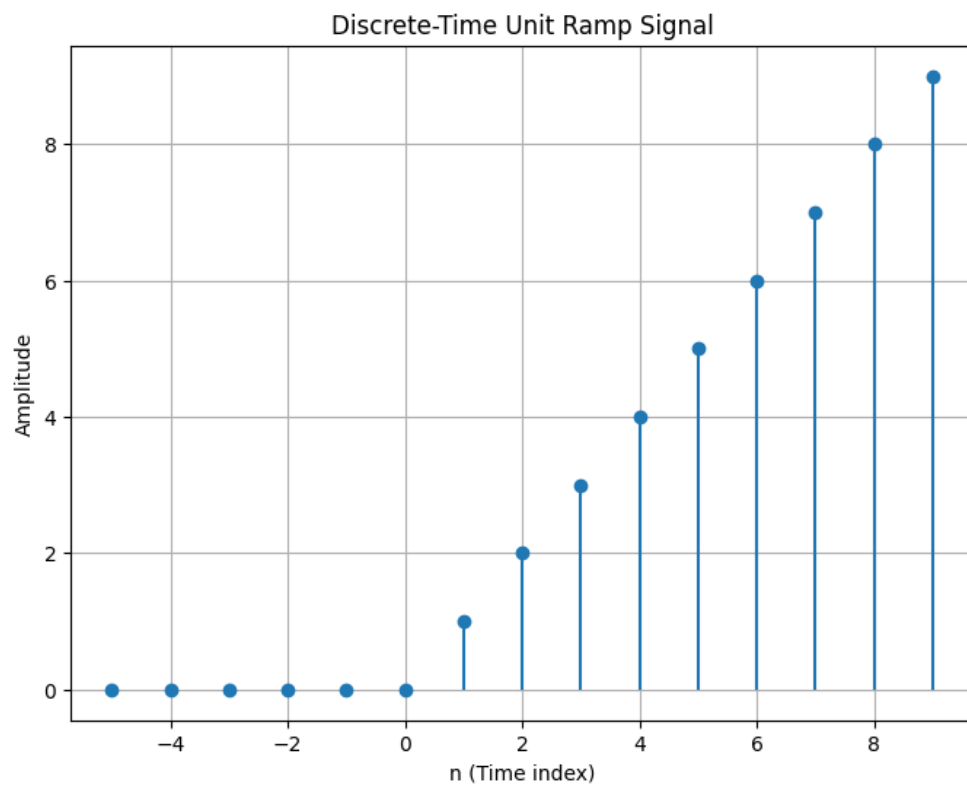
```
import numpy as np
import matplotlib.pyplot as plt

# Time index n (for discrete time signals, it's usually integers)
n = np.arange(-5, 10, 1) # Time range from -5 to 9 (discrete time indices)

# Unit Ramp signal r[n]
r = np.maximum(0, n) # r[n] = n for n >= 0, and 0 otherwise

# Plot the discrete-time unit ramp signal
plt.figure(figsize=(8, 6))
plt.stem(n, r, basefmt=" ", use_line_collection=True)
plt.xlabel('n (Time index)')
plt.ylabel('Amplitude')
plt.title('Discrete-Time Unit Ramp Signal')
plt.grid(True)
plt.show()
```

OUTPUT:



## CONTINUOUS TIME - UNIT STEP

### ALGORITHM:

1. Start the program and import the numpy and matplotlib libraries.
2. Define the discrete time range np. arrange
3. Create an array t that ranges from -5 to 10 with increments of 0.1.
4. Generate a continuous unit step signal using np.where by setting values to 1 when t is greater than or equal to 0, and 0 otherwise.
5. Plot the signal plt. plot() to create a plot of continuous signal x(n) and set x- axis and y-axis.
6. Display the plot.
7. Stop the program

### PROGRAM:

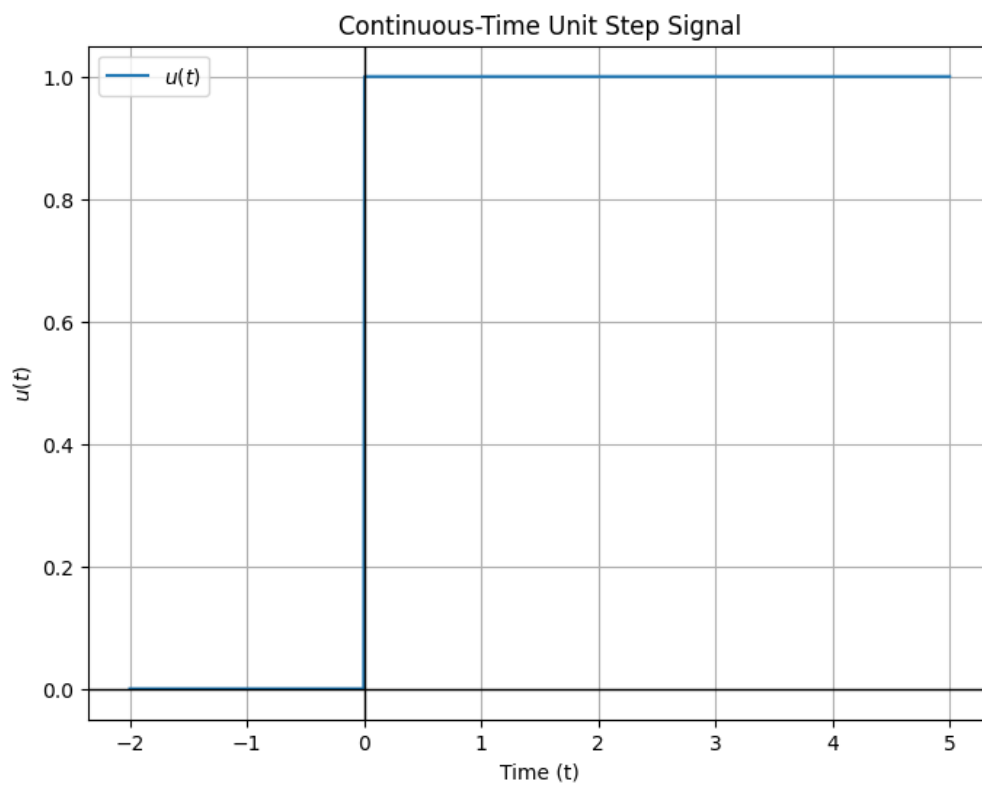
```
import numpy as np
import matplotlib.pyplot as plt

# Time range (continuous time)
t = np.linspace(-2, 5, 1000) # Time from -2 to 5 seconds with 1000 points

# Continuous-time unit step signal u(t)
u = np.heaviside(t, 0) # u(t) = 1 for t >= 0, and 0 for t < 0

# Plot the continuous-time unit step signal
plt.figure(figsize=(8, 6))
plt.plot(t, u, label=r'$u(t)$')
plt.axhline(0, color='black', linewidth=1)
plt.axvline(0, color='black', linewidth=1)
plt.title("Continuous-Time Unit Step Signal")
plt.xlabel("Time (t)")
plt.ylabel(r"$u(t)$")
plt.grid(True)
plt.legend()
plt.show()
```

OUTPUT:



## DISCRETE TIME - UNIT STEP

### ALGORITHM:

1. Start the program and import the numpy and matplotlib libraries.
2. Define the discrete time range np. arrange
3. Create an array n that ranges from -5 to 10 with increments of 1.
4. Generate a discrete unit step signal using np.where by setting values to 1 when n is greater than or equal to 0, and 0 otherwise.
5. Plot the signal plt. stem() to create a stem plot of discrete signal x(n) and set x- axis and y-axis.
6. Display the plot.
7. Stop the program.

### PROGRAM:

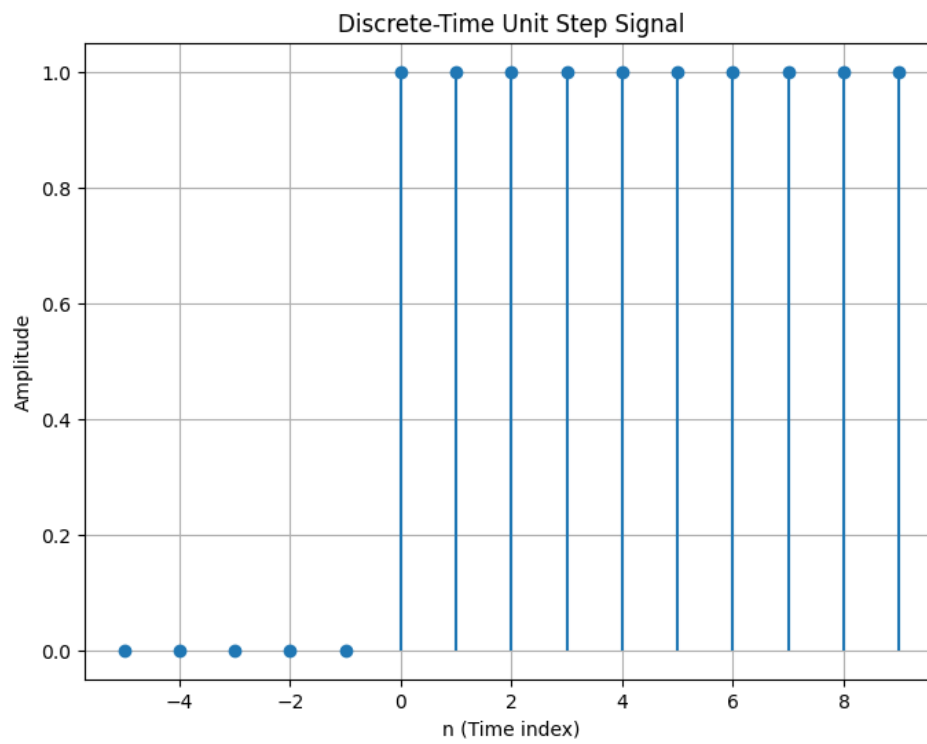
```
import numpy as np
import matplotlib.pyplot as plt

# Time index n (discrete time)
n = np.arange(-5, 10, 1) # Time range from -5 to 9 (discrete indices)

# Discrete-time unit step signal u[n]
u = np.where(n >= 0, 1, 0) # u[n] = 1 for n >= 0, and 0 for n < 0

# Plot the discrete-time unit step signal
plt.figure(figsize=(8, 6))
plt.stem(n, u, basefmt=" ", use_line_collection=True)
plt.xlabel('n (Time index)')
plt.ylabel('Amplitude')
plt.title('Discrete-Time Unit Step Signal')
plt.grid(True)
plt.show()
```

OUTPUT:



INFERENCE:

RESULT:

EX NO : 02	Generate discrete signal and calculate energy/ Power of a signal
DATE:	

AIM:

To Generate discrete signal and calculate energy/ Power of a signal

ALGORITHM:

1. Start the program and import the numpy and matplotlib libraries.
2. Define the discrete time range np. arrange
3. Define the signal as  $n*u(n)$ .
4. Plot the signal plt. stem() to create a stem plot of discrete signal  $x(n)$  and set x- axis and y-axis.
5. Calculate the energy of the signal.
6. Calculate the average power of the signal.
7. Classify the signal,
  - 7.1 if the energy is finite ( $\text{energy} < \text{np.inf}$ ) and the average power is greater than 1, classify it as a power signal.
  - 7.2 if the energy is finite and the avg power is 0, classify it as an energy signal
  - 7.3 otherwise, classify it as neither a power signal nor an energy signal.
8. Print the results
9. Stop the program.

### ENERGY SIGNAL

PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt
```

# Parameters

A = 1        # Amplitude

$\phi = 0$         # Phase shift

n = np.arange(0, 1, 0.01) # Time index (discrete)

# Generate the sinusoidal signal

x\_n = A \* np.sin(2 \* np.pi \* n +  $\phi$ )

# Plotting

plt.figure(figsize=(10, 4))

plt.plot(n, x\_n, label='x(n) = A \* sin( $\omega_0 n + \phi$ )', color='blue')

plt.title('Sinusoidal Power Signal')

plt.xlabel('Time (seconds)')

plt.ylabel('Amplitude')

plt.grid()

plt.legend()

plt.show()

# Calculate the energy of the signal (finite time interval)

energy = np.sum(x\_n\*\*2) # Energy over the finite time interval

# Calculate the average power of the signal

average\_power = np.mean(x\_n\*\*2) # Average power over the time interval

# Classification

if average\_power > 0:

    classification = "The signal is a power signal."

else:

    classification = "The signal is an energy signal."

# Print results

print(f"Energy of the signal: {energy}") # Energy over the finite interval

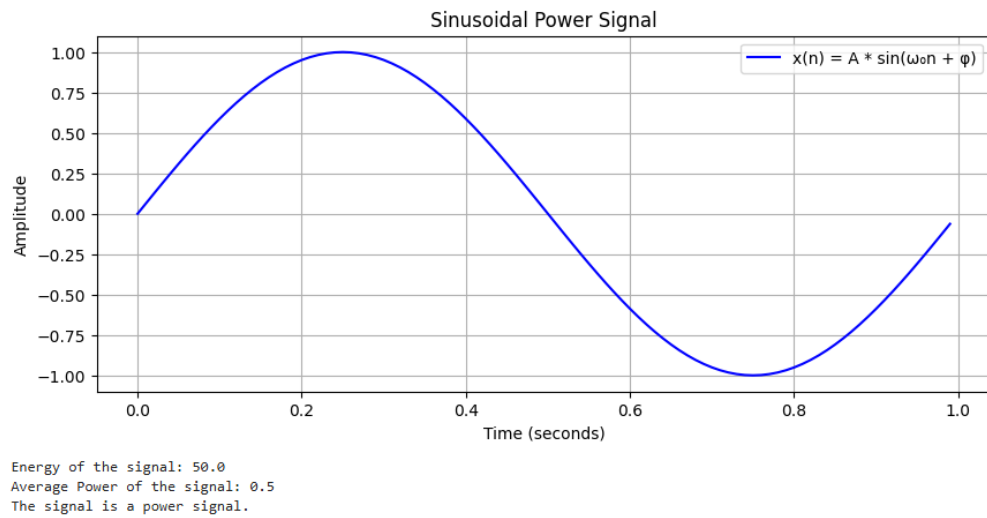
print(f"Average Power of the signal: {round(average\_power, 2)}") # Print

average power

print(classification)



## OUTPUT:



## POWER SIGNAL

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

# Define the time range
t = np.linspace(-5, 5, 1000) # Increased resolution for better visualization

# Define the signal  $x(t) = e^{-t} * u(t)$ 
signal = np.exp(-t) * np.where(t >= 0, 1, 0)

# Plot the signal
plt.plot(t, signal)
plt.xlabel('Time')
plt.ylabel('Amplitude')
plt.title('exp(-t) * u(t)')
plt.grid(True)
plt.show()

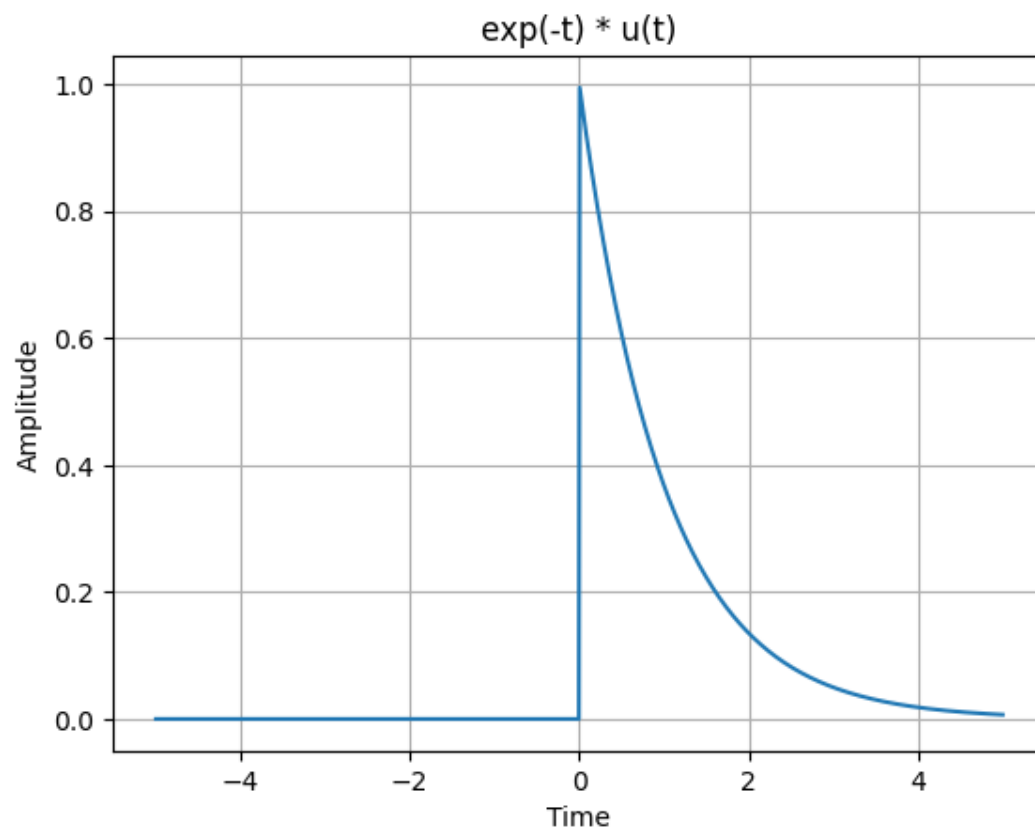
# Calculate the energy of the signal
energy = np.trapz(signal**2, t)
```

```
# Calculate the power of the signal (over a long enough interval)
T = 1000 # Large enough T for power calculation
power = (1 / (2 * T)) * np.trapz(signal**2, t)

# Classification based on energy and average power
if energy < np.inf and power == 0:
    classification = "The signal is a power signal."
elif energy < np.inf and power > 0:
    classification = "The signal is an energy signal."
else:
    classification = "The signal is neither energy nor a power signal."

# Print results
print(f"Energy of the signal: {round(energy, 1)}") # Energy is conceptually
finite
print(f"Average Power of the signal: {round(power, 2)}") # Print average
power
print(classification)
```

OUTPUT:



Energy of the signal: 0.5  
Average Power of the signal: 0.0  
The signal is an energy signal.

INFERENCE:

RESULT:

EX NO : 03	Understand the properties and the different representations of LTI systems
DATE:	

AIM:

To Understand the properties and the different representations of LTI systems

### LINEARITY PROPERTY

ALGORITHM:

- 1.Start the program
- 2.Import NumPy for numerical operations and Matplotlib for plotting.
- 3.Create a function system(x) that returns the square of the input x.
- 4.Initialize two input arrays, x1 and x2.
- 5.Calculate the outputs y1 and y2 by passing x1 and x2 through the system function.
- 6.Compute a new input array by adding x1 and x2, and find its output y3.
- 7.Verify if the system is linear by checking if y3 equals y1 + y2.
- 8.Output the sums of y1 and y2, the value of y3, and whether the system is linear.
- 9.Create a plot to visualize the relationship between inputs and outputs, including labels and a grid.
- 10.Stop the program.

PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

# Define a system (example: squaring the input)
def system(x):
    return x**2

# Input signals
x1 = np.array([-2, -1, 0, 1, 2])
x2 = np.array([1, 0, -1, -2, 0])

# Outputs for individual inputs
```

```
y1 = system(x1)
y2 = system(x2)

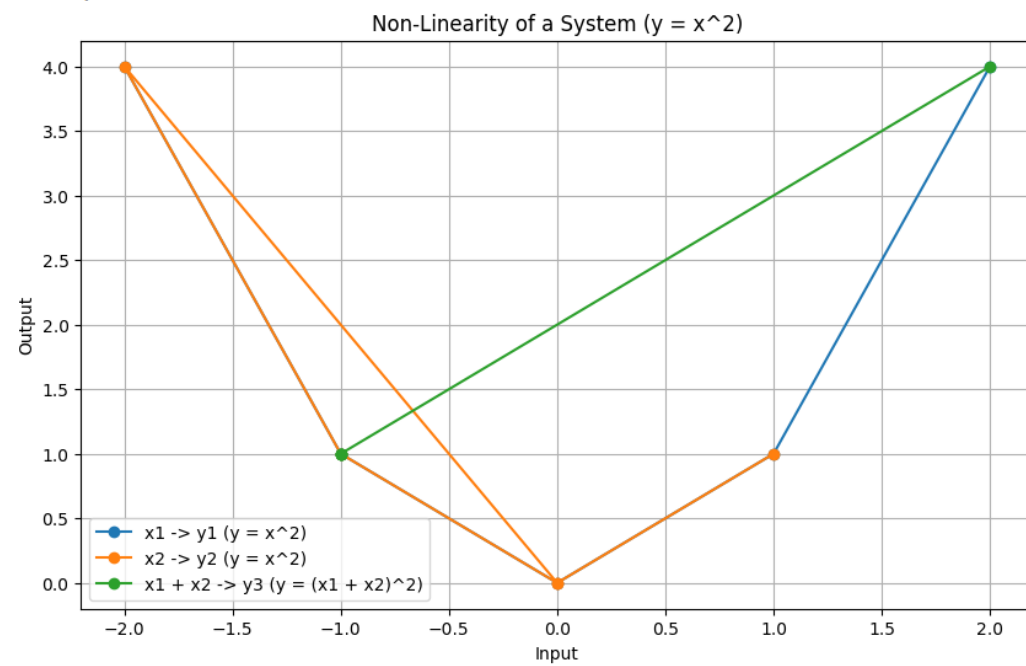
# Combined input and output
y3 = system(x1 + x2) # Applying system to the sum of x1 and x2

# Check for linearity: y3 should be equal to y1 + y2 for linearity
print("y1 + y2:", y1 + y2)
print("y3:", y3)
print("Is the system linear?", np.array_equal(y3, y1 + y2))

# Plotting for visualization
plt.figure(figsize=(10, 6))
plt.plot(x1, y1, 'o-', label='x1 -> y1 (y = x^2)')
plt.plot(x2, y2, 'o-', label='x2 -> y2 (y = x^2)')
plt.plot(x1 + x2, y3, 'o-', label='x1 + x2 -> y3 (y = (x1 + x2)^2)')
plt.xlabel('Input')
plt.ylabel('Output')
plt.title('Non-Linearity of a System (y = x^2)')
plt.legend()
plt.grid(True)
plt.show()
```

## OUTPUT:

```
y1 + y2: [5 1 1 5 4]  
y3: [1 1 1 1 4]  
Is the system linear? False
```



## TIME SHIFTING PROPERTY

### ALGORITHM:

1. Start the program.
2. Import NumPy for numerical operations and Matplotlib for plotting.
3. Create an array  $t$  using `np.linspace` to represent time from 0 to 1 second, divided into 100 points.
4. Create a sine wave signal  $x$  with a frequency of 5 Hz using the formula  $x = \sin(2\pi ft)$ .
5. Set a variable  $t_0$  to define the time delay.
6. Calculate the time-shifted signal  $x_{\text{shifted}}$  by adjusting the time array in the sine function.
7. Initialize a plot with a specified figure size.
8. Create the first subplot for the original sine wave and the second subplot for the time-shifted sine wave.
9. Stop the program.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

# Generate a time domain signal (example: sine wave)
t = np.linspace(0, 1, 100) # Time from 0 to 1 second
x = np.sin(2 * np.pi * 5 * t) # Sine wave with frequency 5 Hz

# Time shift the signal
t0 = 0.10 # Time delay in seconds (0.1 seconds for a visible shift)
t_shifted = t - t0 # Shift time axis by t0
x_shifted = np.sin(2 * np.pi * 5 * t_shifted) # Apply the sine wave with
shifted time

# Plotting
plt.figure(figsize=(10, 6))

# Original signal plot
plt.subplot(2, 1, 1)
plt.plot(t, x, label='Original Signal')
plt.xlabel('Time (s)')
```

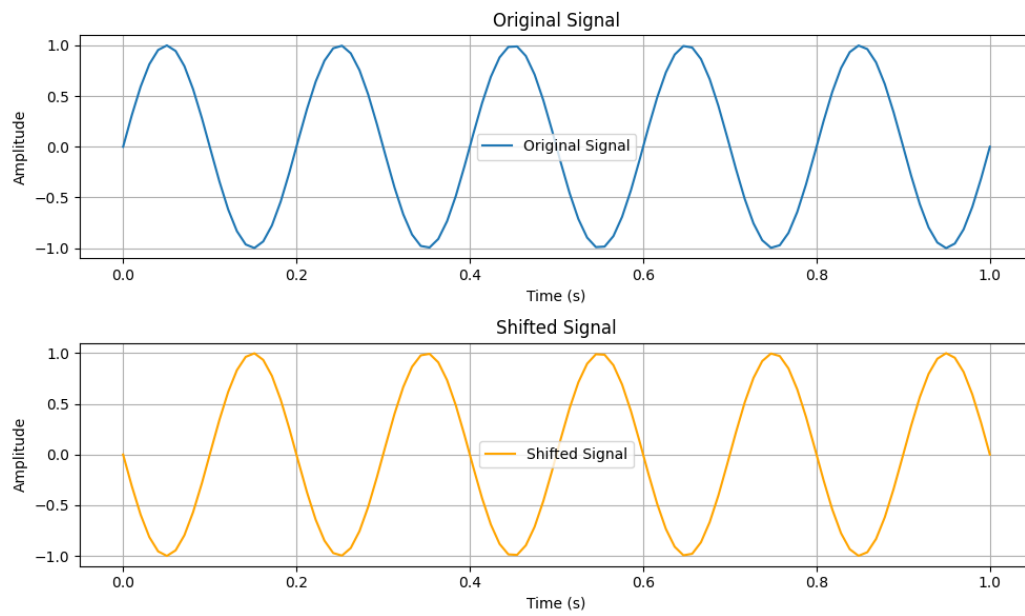
```
plt.ylabel('Amplitude')  
plt.title('Original Signal')  
plt.legend()  
plt.grid(True)
```

```
# Shifted signal plot  
plt.subplot(2, 1, 2)  
plt.plot(t, x_shifted, label='Shifted Signal', color='orange')  
plt.xlabel('Time (s)')  
plt.ylabel('Amplitude')  
plt.title('Shifted Signal')  
plt.legend()  
plt.grid(True)
```

```
plt.tight_layout() # Ensure that the subplots don't overlap  
plt.show()
```



OUTPUT:



## TIME REVERSAL PROPERTY

### ALGORITHM:

1. Start the program
2. Import NumPy for numerical calculations and Matplotlib for plotting.
3. Create an array  $t$  that spans from -5 to 5 with a step of 0.1.
4. Create `ramp_signal` using `np.where`, setting values to  $t$  when  $t$  is greater than or equal to 0, and 0 otherwise.
5. Use `np.flip()` to reverse the `ramp_signal`, storing the result in `time_reversed_signal`.
6. Initialize a plot with a specified figure size.
7. Plot the `ramp_signal` against the time array  $t$  and overlay the `time_reversed_signal` on the same plot.
9. Add labels for the x-axis and y-axis, as well as a title and grid lines.
10. Stop the program.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

# Time values
t = np.arange(-5, 5, 0.1) # Adjust range and step as needed

# Unit ramp signal
ramp_signal = np.where(t >= 0, t, 0)

# Time reversal using np.flip()
time_reversed_signal = np.flip(ramp_signal) # Use np.flip() to reverse the
signal

# Create reversed time vector to match the flipped signal
t_reversed = np.flip(t) # Reverse the time vector

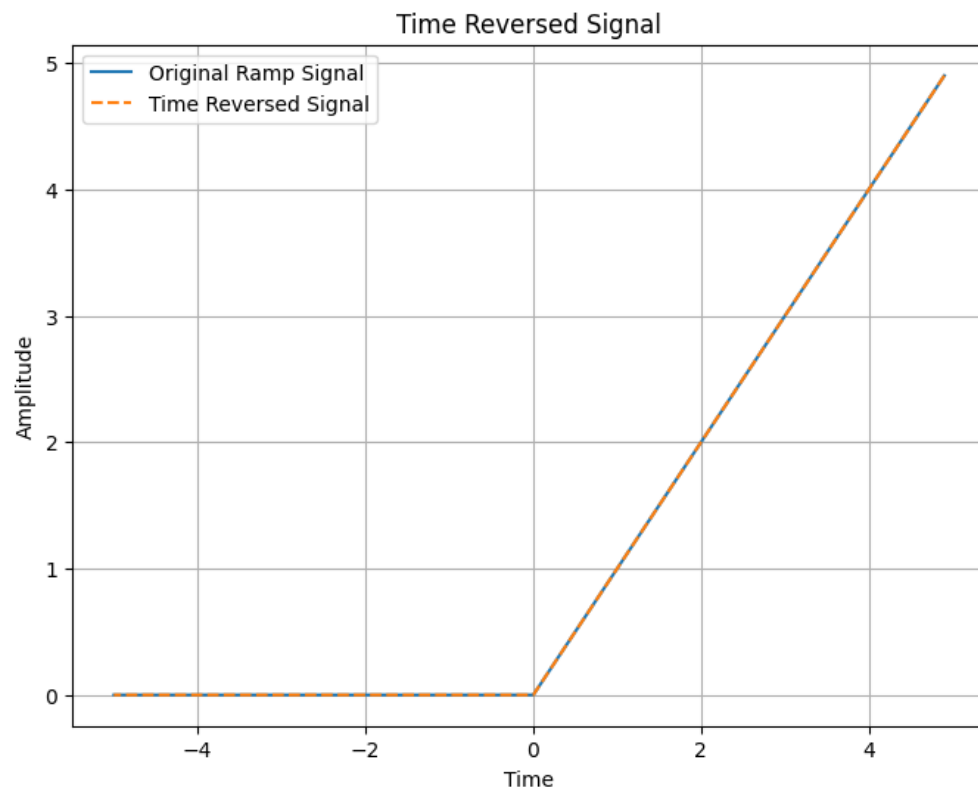
# Plot the signals
plt.figure(figsize=(8, 6))

# Plot original ramp signal
plt.plot(t, ramp_signal, label='Original Ramp Signal')
```

```
# Plot time-reversed signal
plt.plot(t_reversed, time_reversed_signal, label='Time Reversed Signal',
linestyle='--')

plt.xlabel('Time')
plt.ylabel('Amplitude')
plt.title('Time Reversed Signal')
plt.grid(True)
plt.legend()
plt.show()
```

OUTPUT:



## FREQUENCY SHIFTING PROPERTY

### ALGORITHM:

1. Start the program.
2. Import NumPy for numerical operations and Matplotlib for plotting.
3. Create an array  $t$  using `np.linspace` to represent time from 0 to 1 second, divided into 100 points.
4. Create a sine wave signal  $x$  with a frequency of 5 Hz using the formula  $x = \sin(2\pi ft)$ .
5. Set a variable  $f\_shift$  to define the frequency shift.
6. Apply frequency shifting by multiplying the original signal  $x$  with a complex exponential using the formula  $x\_shifted = x * e^{j \sin(2\pi f\_shift t)}$ .
7. Initialize a plot with a specified figure size.
8. Create the first subplot for the original sine wave and Create the second subplot for the frequency-shifted sine wave.
9. Stop the program.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

# Generate a time domain signal (example: sine wave)
t = np.linspace(0, 1, 100)
x = np.sin(2 * np.pi * 5 * t)

# Frequency to shift by
f_shift = 1 # Shift by 1 Hz

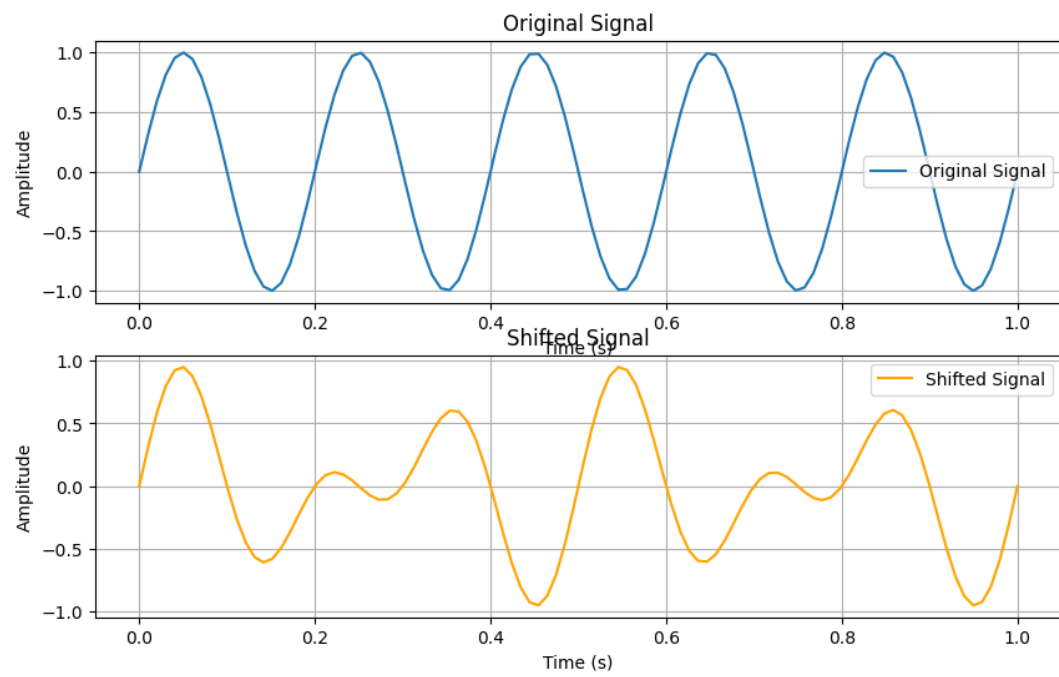
# Frequency shifting using complex exponential multiplication
x_shifted = x * np.exp(2j * np.pi * f_shift * t)

# Plotting
plt.figure(figsize=(10, 6))
```

```
plt.subplot(2, 1, 1)
plt.plot(t, x, label='Original Signal')
plt.xlabel('Time (s)')
plt.ylabel('Amplitude')
plt.title('Original Signal')
plt.legend()
plt.grid(True)
```

```
plt.subplot(2, 1, 2)
plt.plot(t, x_shifted, label='Shifted Signal', color='orange')
plt.xlabel('Time (s)')
plt.ylabel('Amplitude')
plt.title('Shifted Signal')
plt.legend()
plt.grid(True)
plt.show()
```

OUTPUT:



## CONJUGATE PROPERTY

### ALGORITHM:

1. Start the program.
2. Import the NumPy library for numerical operations.
3. Create a complex number  $z$  using a real part and an imaginary part (e.g.,  $3 + 4j$ ).
4. Use `np.conjugate(z)` to calculate the conjugate of the complex number, storing it in `z_conjugate`.
5. Output the original complex number and its conjugate using print statements.
6. Stop the program

### PROGRAM:

```
import numpy as np

# Create a complex number
z = 3 + 4j

# Calculate the conjugate
z_conjugate = np.conjugate(z)

# Print the results
print("Original complex number:", z)
print("Conjugate:", z_conjugate)
```



OUTPUT:

Original complex number: **(3+4j)**

Conjugate: **(3-4j)**

INFERENCE:

RESULT:

EX NO : 04	Understand the concepts of convolution
DATE:	

AIM:

To Understand the concepts of convolution

ALGORITHM:

- 1: Take the two sequences you want to convolve (the input sequence and the filter or impulse response).
- 2: Find out how long each sequence is.
- 3: Calculate the length of the final result by adding the lengths of the two sequences and subtracting 1.
- 4: Create a new matrix where each row will hold a shifted version of the second sequence, and the matrix will have enough rows to accommodate all shifts.
- 5: Fill the matrix by placing values from the second sequence in each row, making sure the values are aligned properly as you move down the rows.
- 6: Convert the first sequence into a column (like a vertical list).
- 7: Multiply the matrix you created by the column from the first sequence to calculate the convolution.
- 8: Take the result and flatten it back into a single list, which will be your final answer.

PROGRAM:

```
import numpy as np

def linear_convolution_matrix_method(x, h):
    # Length of the input sequences
    M = len(x)
    L = len(h)

    # Length of the result (convolution output)
```

$N = M + L - 1$

# Create a Toeplitz matrix for  $h[n]$

$H = \text{np.zeros}((N, M))$

# Fill the Toeplitz matrix with appropriate values from  $h[n]$

for  $i$  in  $\text{range}(N)$ :

    for  $j$  in  $\text{range}(M)$ :

        if  $i - j \geq 0$  and  $i - j < L$ :

$H[i, j] = h[i - j]$

# Convert  $x[n]$  to a column vector

$x\_col = \text{np.reshape}(x, (M, 1))$

# Perform the matrix multiplication  $H * x$  to get the convolution result

$y = \text{np.dot}(H, x\_col)$

# Flatten the result to a 1D array

return  $y.\text{flatten}()$

# Example usage

$x = [1, 2, 3]$  # Input sequence

$h = [4, 5]$  # Impulse response

# Perform convolution

$y = \text{linear\_convolution\_matrix\_method}(x, h)$

# Print the result

$\text{print}(\text{"Convolution Result:"}, y)$

OUTPUT:

Convolution Result: [ 4. 13. 22. 15.]

INFERENCE:

RESULT:

EX NO : 05	Analyse the effects of sampling in the time and frequency domains
DATE:	

### AIM:

To Analyse the effects of sampling in the time and frequency domains

### ALGORITHM:

- 1: Define signal parameters (frequencies, amplitudes, duration).
- 2: Select different sampling rates to analyze.
- 3: For each sampling rate:
  - Generate time vector based on sampling rate.
  - Sum sine waves with given frequencies and amplitudes to form the signal.
- 4: Plot the signal in the time domain for each sampling rate.
- 5: Compute and plot the frequency spectrum using FFT:
  - Take the positive frequencies and plot their magnitudes.
- 6: Adjust plot layout and display the results.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.fft import fft, fftfreq
```

```
# Function to generate a continuous signal (sum of sine waves)
def generate_continuous_signal(frequencies, amplitudes, duration,
sampling_rate=1000):
    """
```

Generates a continuous signal (sum of sine waves) sampled at a given sampling rate.

:param frequencies: List of frequencies of the sine waves (Hz).

```

:param amplitudes: List of amplitudes corresponding to the sine waves.
:param duration: Duration of the signal (seconds).
:param sampling_rate: Sampling rate (samples per second).

:return: time (discrete), signal (discrete)
"""

time = np.linspace(0, duration, int(sampling_rate * duration),
endpoint=False) # Continuous time
signal = np.zeros_like(time) # Initialize signal

# Sum of sine waves with different frequencies and amplitudes
for freq, amp in zip(frequencies, amplitudes):
    signal += amp * np.sin(2 * np.pi * freq * time)

return time, signal

# Function to compute the FFT and plot the frequency spectrum
def plot_frequency_spectrum(signal, sampling_rate):
    """
    Compute and plot the frequency spectrum of the signal using FFT.

    :param signal: The discrete signal.
    :param sampling_rate: Sampling rate (samples per second).
    """

    # Compute FFT
    N = len(signal)
    freqs = fftfreq(N, 1/sampling_rate) # Frequency axis
    fft_values = fft(signal) # FFT of the signal

    # Only take the positive half of the spectrum (real signals are
    symmetric)
    half_N = N // 2
    freqs = freqs[:half_N]
    fft_values = fft_values[:half_N]

    # Plot the magnitude of the frequency spectrum
    plt.plot(freqs, np.abs(fft_values))
    plt.title("Frequency Spectrum")
    plt.xlabel("Frequency [Hz]")
    plt.ylabel("Magnitude")

```

```

plt.grid(True)
plt.xlim(0, sampling_rate / 2) # Nyquist frequency limit

# Function to plot the signal in the time domain
def plot_time_domain(time, signal, title="Signal in Time Domain"):
    """
    Plot the signal in the time domain.

    :param time: Time vector for the signal.
    :param signal: The signal in the time domain.
    :param title: Title for the plot.
    """
    plt.plot(time, signal)
    plt.title(title)
    plt.xlabel("Time [s]")
    plt.ylabel("Amplitude")
    plt.grid(True)

# Main Program
if __name__ == "__main__":
    # Signal parameters
    frequencies = [50, 120, 200] # Frequencies of the sine waves (Hz)
    amplitudes = [1, 0.5, 0.8] # Amplitudes of the sine waves
    duration = 1 # Duration of the signal (seconds)

    # Different sampling rates to demonstrate the effect of sampling
    sampling_rates = [300, 500, 1000, 2000] # Different sampling rates

    # Create the figure for plotting
    plt.figure(figsize=(12, 12))

    # Analyze the effect of different sampling rates
    for i, sampling_rate in enumerate(sampling_rates, start=1):
        # Generate the continuous signal with the given sampling rate
        time, signal = generate_continuous_signal(frequencies, amplitudes,
        duration, sampling_rate)

        # Plot the signal in the time domain
        plt.subplot(4, 2, 2*i-1)

```

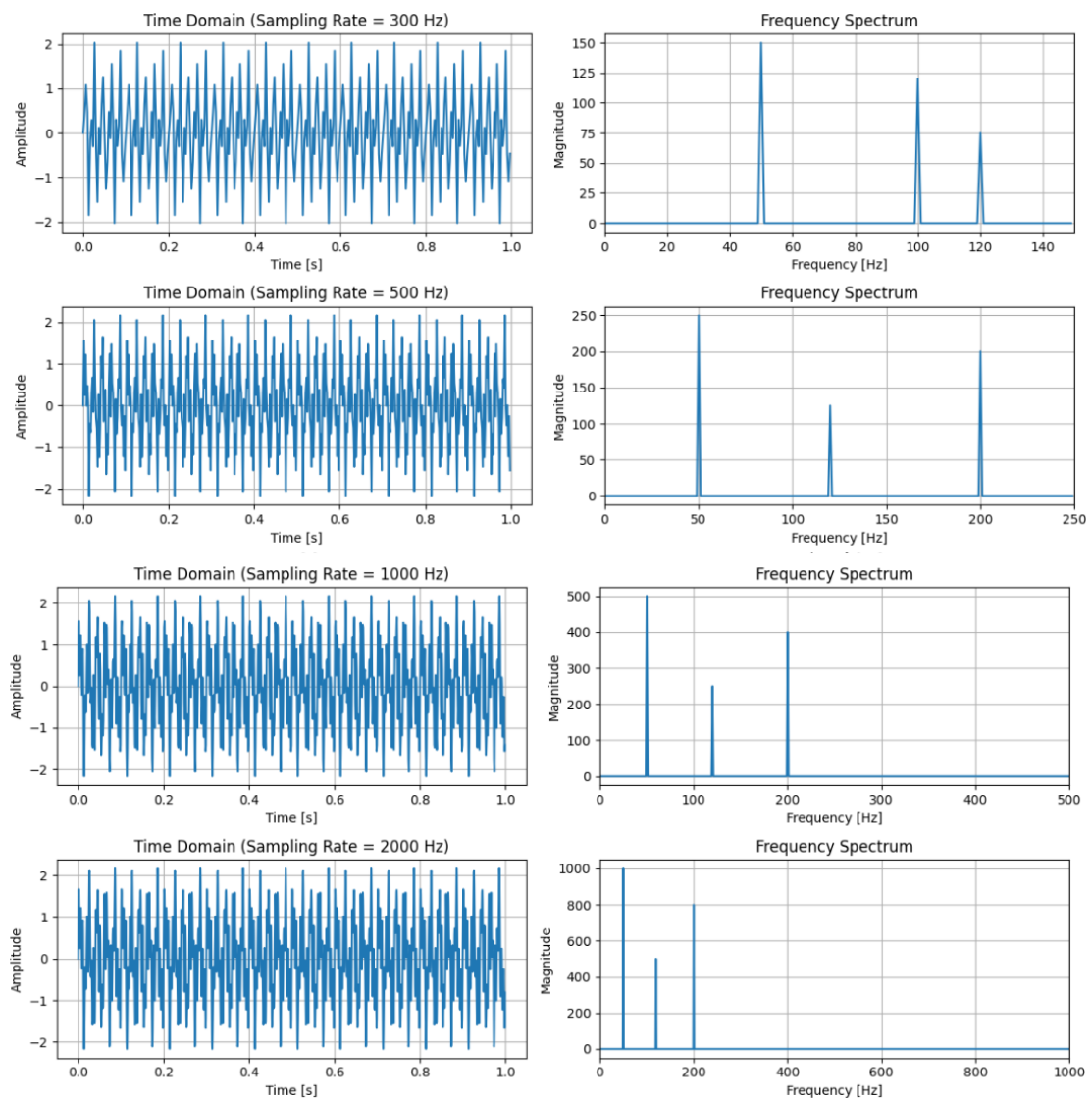
```
    plot_time_domain(time, signal, title=f"Time Domain (Sampling Rate  
= {sampling_rate} Hz)")
```

```
    # Plot the frequency spectrum of the signal  
    plt.subplot(4, 2, 2*i)  
    plot_frequency_spectrum(signal, sampling_rate)
```

```
    # Adjust the layout  
    plt.tight_layout()  
plt.show()
```



OUTPUT:



INFERENCE:

RESULT:

EX NO : 06	Understand the basics of Uniform Quantization
DATE:	

### AIM:

To Understand the basics of Uniform Quantization

### ALGORITHM:

- 1: Define the input signal (e.g., a sine wave) and the number of quantization levels.
- 2: Find the minimum and maximum values of the signal.
- 3: Calculate the quantization step size using the formula:
- 4: Quantize the signal:
  - Subtract the min value from the signal.
  - Divide by the step size and round the result.
  - Multiply by the step size and add the min value back.
- 5: Plot the original signal and the quantized signal.
- 6: Display the quantization step size.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

def uniform_quantization(signal, num_levels):
    """
    Perform uniform quantization on a signal.

    :param signal: The input signal (numpy array).
    :param num_levels: Number of quantization levels.
    :return: Quantized signal (numpy array).
```

```

"""
# Find the min and max values of the signal
min_val = np.min(signal)
max_val = np.max(signal)

# Calculate the quantization step size
step_size = (max_val - min_val) / num_levels

# Quantize the signal
quantized_signal = np.round((signal - min_val) / step_size) * step_size
+ min_val

return quantized_signal, step_size

# Example usage:
if __name__ == "__main__":
    # Generate a continuous signal (e.g., sine wave)
    time = np.linspace(0, 1, 500) # Time vector
    original_signal = np.sin(2 * np.pi * 5 * time) # A sine wave with
frequency of 5 Hz

    # Set the number of quantization levels (e.g., 4 levels, 8 levels, etc.)
    num_levels = 4

    # Perform uniform quantization
    quantized_signal, step_size = uniform_quantization(original_signal,
num_levels)

    # Plot the original and quantized signals
    plt.figure(figsize=(10, 6))

    # Plot the original signal
    plt.subplot(2, 1, 1)
    plt.plot(time, original_signal, label='Original Signal')
    plt.title('Original Signal')
    plt.xlabel('Time [s]')
    plt.ylabel('Amplitude')
    plt.grid(True)

    # Plot the quantized signal

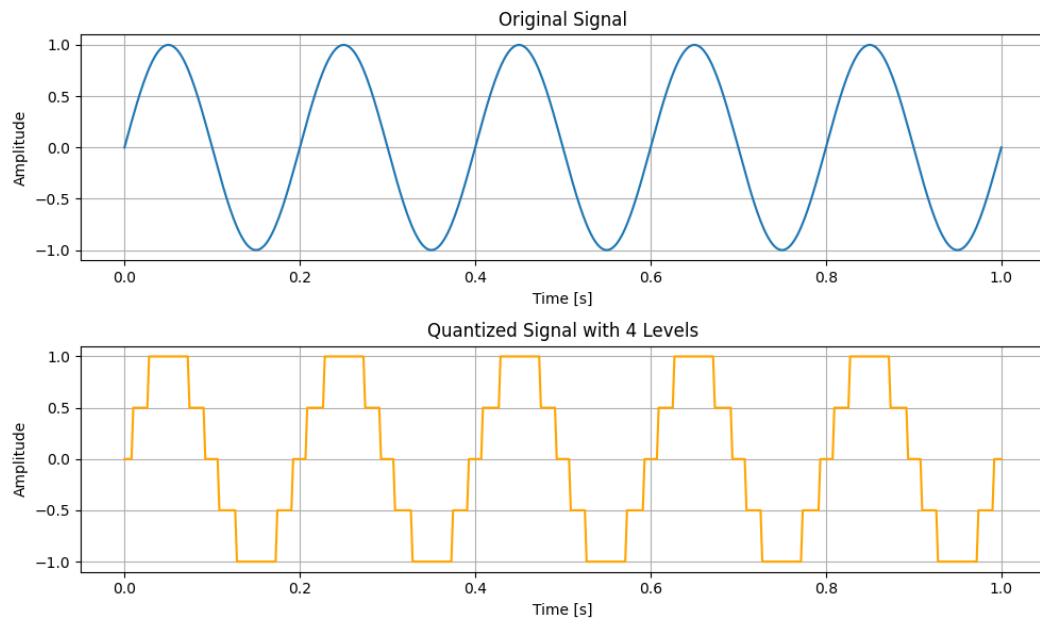
```

```
plt.subplot(2, 1, 2)
plt.plot(time, quantized_signal, label=f'Quantized Signal (Levels =
{num_levels})', color='orange')
plt.title(f'Quantized Signal with {num_levels} Levels')
plt.xlabel('Time [s]')
plt.ylabel('Amplitude')
plt.grid(True)

# Show the plots
plt.tight_layout()
plt.show()

print(f"Quantization step size: {step_size}")
```

OUTPUT:



INFERENCE:

RESULT:

EX NO : 07	Analysis and Synthesis of Signals using Discrete Fourier Transform
DATE:	

### AIM:

To Analysis and Synthesis of Signals using Discrete Fourier Transform

### ALGORITHM:

- 1: Define signal parameters (sampling rate, duration, and frequencies).
- 2: Generate the time vector and create the signal (sum of sine waves).
- 3: Perform DFT on the signal using FFT to get the frequency-domain coefficients.
- 4: Plot the original signal, magnitude spectrum, and phase spectrum.
- 5: Perform IDFT using IFFT to reconstruct the signal from DFT coefficients.
- 6: Plot the original and reconstructed signals for comparison.

### PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt

# Function to perform Discrete Fourier Transform (DFT)
def dft_analysis(signal):
    """
    Perform Discrete Fourier Transform (DFT) on the input signal.

    :param signal: The input signal (numpy array).
    :return: The DFT coefficients (numpy array).
```

```

"""
return np.fft.fft(signal)

# Function to perform Inverse Discrete Fourier Transform (IDFT)
def dft_synthesis(dft_coefficients):
    """
    Perform Inverse Discrete Fourier Transform (IDFT) to reconstruct the
    signal.

    :param dft_coefficients: The DFT coefficients (numpy array).
    :return: The reconstructed signal (numpy array).
    """
    return np.fft.ifft(dft_coefficients)

# Function to plot the signal, its magnitude, and phase spectrum
def plot_spectra(signal, dft_coefficients):
    """
    Plot the time-domain signal, magnitude spectrum, and phase
    spectrum.

    :param signal: The original signal (numpy array).
    :param dft_coefficients: The DFT coefficients (numpy array).
    """
    # Time-domain signal
    time = np.arange(len(signal))

    # Frequency domain (for plotting)
    freq = np.fft.fftfreq(len(signal))

    # Magnitude and Phase of DFT coefficients
    magnitude = np.abs(dft_coefficients)
    phase = np.angle(dft_coefficients)

    plt.figure(figsize=(12, 8))

    # Plot original signal in time domain
    plt.subplot(3, 1, 1)
    plt.plot(time, signal)
    plt.title("Original Signal in Time Domain")
    plt.xlabel("Time [samples]")

```

```

plt.ylabel("Amplitude")

# Plot Magnitude Spectrum
plt.subplot(3, 1, 2)
plt.plot(freq, magnitude)
plt.title("Magnitude Spectrum")
plt.xlabel("Frequency [Hz]")
plt.ylabel("Magnitude")
plt.xlim(-0.5, 0.5)

# Plot Phase Spectrum
plt.subplot(3, 1, 3)
plt.plot(freq, phase)
plt.title("Phase Spectrum")
plt.xlabel("Frequency [Hz]")
plt.ylabel("Phase [radians]")
plt.xlim(-0.5, 0.5)

plt.tight_layout()
plt.show()

# Example usage
if __name__ == "__main__":
    # Parameters
    sampling_rate = 1000 # Sampling rate (samples per second)
    duration = 1         # Duration of the signal in seconds
    freq1 = 50           # Frequency of the first sine wave (Hz)
    freq2 = 150          # Frequency of the second sine wave (Hz)

    # Generate a signal composed of two sine waves
    time = np.linspace(0, duration, int(sampling_rate * duration),
endpoint=False)
    signal = np.sin(2 * np.pi * freq1 * time) + 0.5 * np.sin(2 * np.pi * freq2
* time)

    # Perform DFT Analysis
    dft_coefficients = dft_analysis(signal)

    # Plot the signal and its spectrum
    plot_spectra(signal, dft_coefficients)

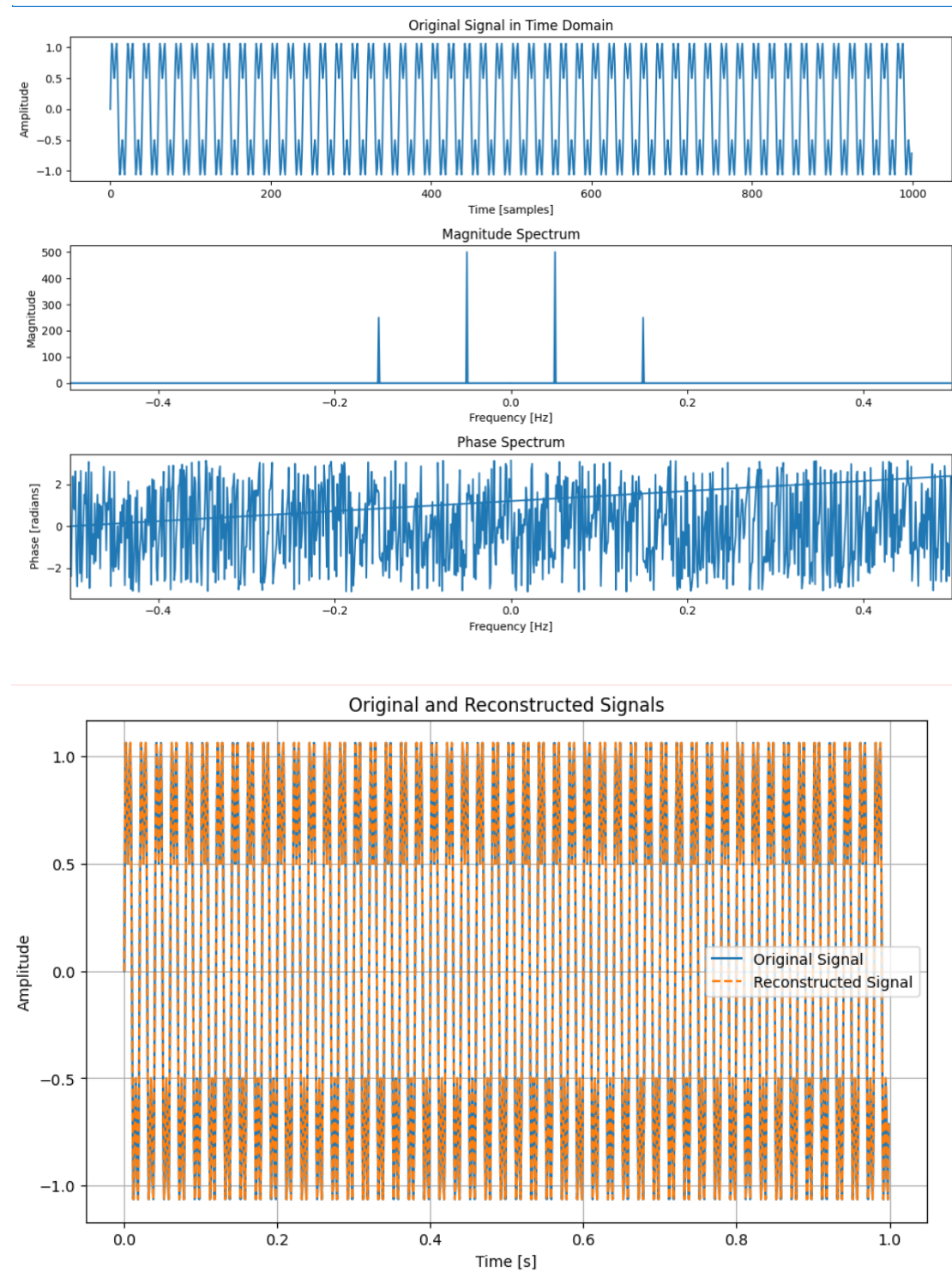
```



```
# Perform DFT Synthesis (Inverse DFT)
reconstructed_signal = dft_synthesis(dft_coefficients)

# Plot the original and reconstructed signals
plt.figure(figsize=(10, 6))
plt.plot(time, signal, label='Original Signal')
plt.plot(time, reconstructed_signal, label='Reconstructed Signal',
linestyle='--')
plt.title("Original and Reconstructed Signals")
plt.xlabel("Time [s]")
plt.ylabel("Amplitude")
plt.legend()
plt.grid(True)
plt.show()
```

## OUTPUT:



INFERENCE:

RESULT:

EX NO : 08	Analyse the ECG signal to find heart rate of a patient
DATE:	

AIM:

To Analyse the ECG signal to find heart rate of a patient

ALGORITHM:

- 1: Define the ECG signal and set the sampling rate.
- 2: Detect R-peaks using peak detection (find\_peaks), with height and distance thresholds.
- 3: Calculate R-R intervals (time differences between consecutive R-peaks).
- 4: Calculate heart rate using the average R-R interval:
- 5: Plot the ECG signal and detected R-peaks, displaying the heart rate.
- 6: Print the detected heart rate.

PROGRAM:

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.signal import find_peaks

# Assuming ecg is a 1D numpy array containing the ECG signal data
ecg = np.random.randn(500) # Replace this with your actual ECG signal

# Sampling rate
fs = 200 # Hz

# Step 1: Identify the R-peaks using a peak detection method
# Use find_peaks from scipy to find the R-peaks (usually the local maxima)
```

```
# We will look for peaks that have a minimum height to avoid noise
peaks, _ = find_peaks(ecg, height=0.5, distance=fs/2.5) # Adjust distance
as per the signal characteristics
```

```
# Step 2: Calculate R-R intervals (time difference between consecutive R-
peaks)
```

```
rr_intervals = np.diff(peaks) / fs # In seconds
```

```
# Step 3: Calculate the heart rate (in beats per minute)
```

```
heart_rate = 60 / np.mean(rr_intervals)
```

```
# Step 4: Plot the ECG signal and the detected R-peaks
```

```
plt.figure(figsize=(10, 6))
```

```
plt.plot(ecg)
```

```
plt.plot(peaks, ecg[peaks], "x", label="R-peaks")
```

```
plt.title(f"Detected R-peaks and Heart Rate: {heart_rate:.2f} bpm")
```

```
plt.xlabel("Sample Index")
```

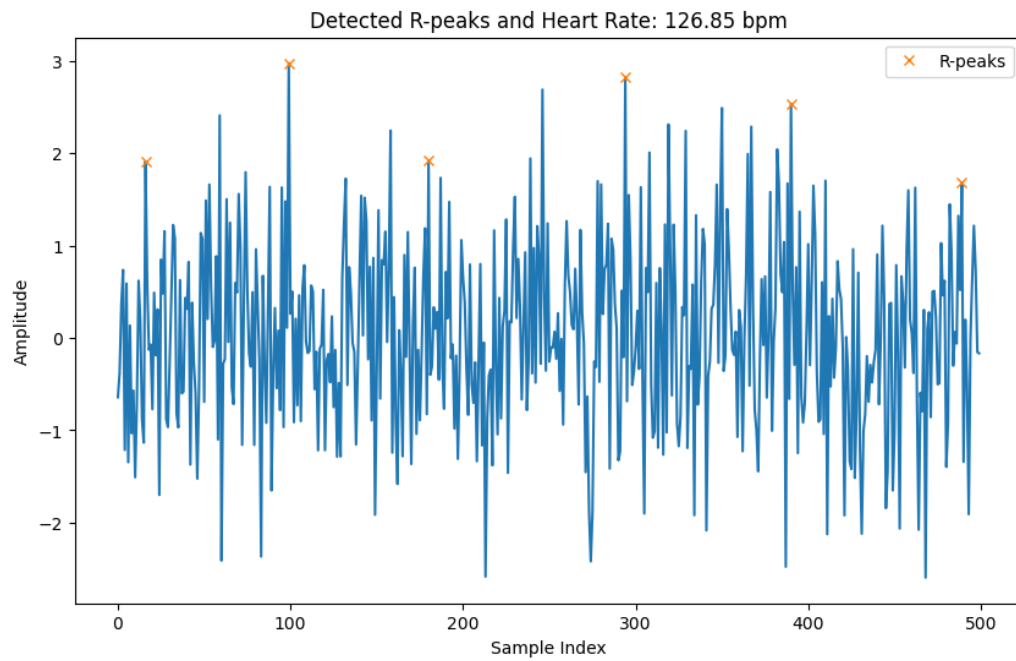
```
plt.ylabel("Amplitude")
```

```
plt.legend()
```

```
plt.show()
```

```
print(f"Detected Heart Rate: {heart_rate:.2f} bpm")
```

OUTPUT:



INFERENCE:

RESULT: