

Delta Modulation and Demodulation using Python

Objective:

- Understand and implement the process of Delta Modulation and Demodulation using Python.
- Visualize the encoded signal and reconstructed signal.
- Apply a Low-Pass Filter to improve the quality of the reconstructed signal.
- Compare the reconstructed signals with the original signal.

Introduction: A delta modulation (DM or Δ -modulation) is an analog-to-digital and digital-to-analog signal conversion technique used for transmission of voice information where quality is not of primary importance. DM is the simplest form of differential pulse-code modulation (DPCM) where the difference between successive samples are encoded into n-bit data streams. In delta modulation, the transmitted data are reduced to a 1-bit data stream. Its main features are:

- The analog signal $x(t)$ is approximated with a series of segments.
- Each segment of the approximated signal is compared to the preceding bits and the successive bits are determined by this comparison.
- Only the change of information is sent, that is, only an increase or decrease of the signal amplitude from the previous sample is sent whereas a no-change condition causes the modulated signal to remain at the same 0 or 1 state of the previous sample.

To achieve high signal-to-noise ratio, delta modulation must use oversampling techniques, that is, the analog signal is sampled at a rate several times higher than the Nyquist rate.

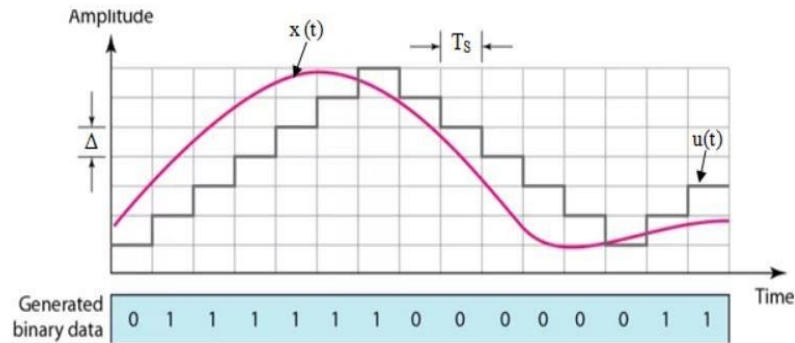


Figure 1: Delta modulation

$x(t)$ represents the analog signal and $x_q(t)$ represents the staircase approximation. Following discrete relations explain the construction of the staircase waveform which forms the basis of delta modulation.

$$e(nT_s) = x_q(nT_s) - x_q(nT_s - T_s)$$

$$e_q(nT_s) = \delta \operatorname{sgn}[e_q(nT_s)]$$

$$x_q(nT_s) = x_q(nT_s - T_s) + e_q(nT_s)$$

Where T_s is the sampling instant, $e(nT_s)$ is the error signal and $e_q(nT_s)$ is the quantized version of error signal.

Block Diagram:

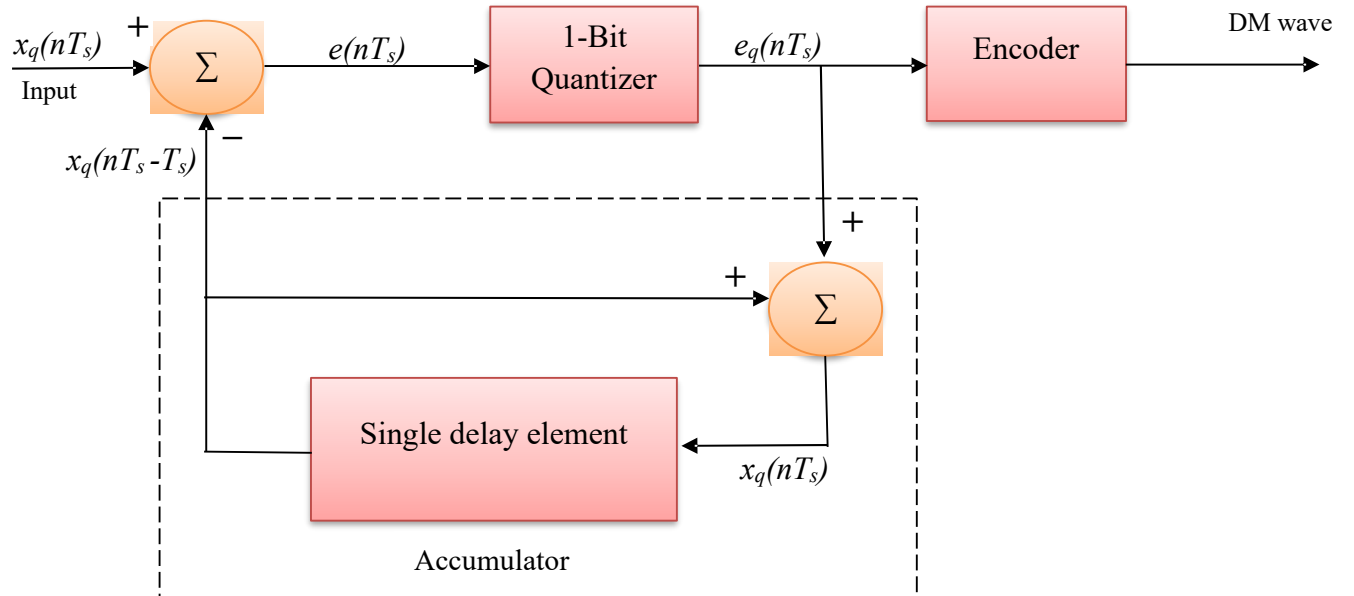


Figure 2: Block diagram of the DM transmitter

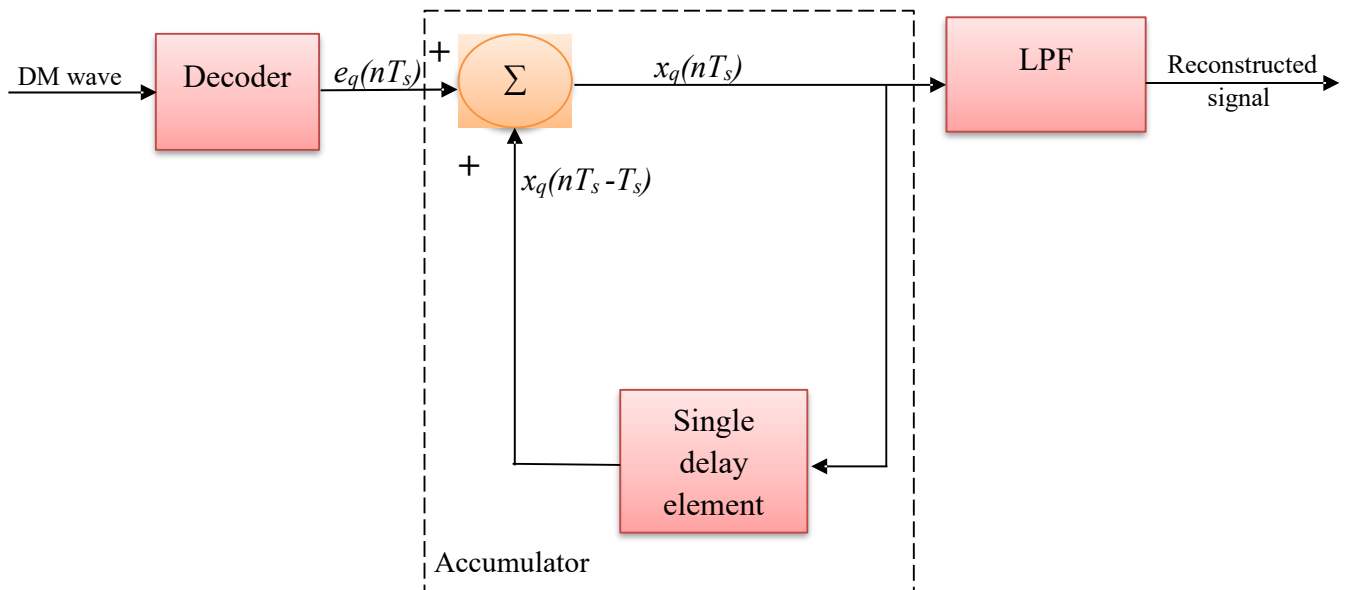


Figure 3: Block diagram of the DM receiver

Implementation Using Python

Steps

1. **Generate the Input Signal:**
 - A sine wave is used as the analog signal for demonstration.
2. **Delta Modulation:**
 - The difference between the current sample and the quantized signal is compared.
 - If the sample is greater, a 1 is appended to the binary sequence; otherwise, a 0.
3. **Delta Demodulation:**
 - The binary sequence is interpreted to reconstruct the signal using the fixed step size.
 - A cumulative sum of the step size (positive or negative) generates the reconstructed signal.
4. **Apply Low-Pass Filter:**
 - A Butterworth filter is designed using `scipy` to smooth the reconstructed signal.
5. **Visualization:**
 - Plot the original signal, delta-modulated signal, reconstructed signal (before and after filtering), and overlay the original signal for comparison.

Python Code

```
1 import numpy as np
2 import matplotlib.pyplot as plt
3 from scipy.signal import butter, filtfilt
4
5 # Parameters
6 sampling_rate = 100 # Sampling rate (Hz)
7 duration = 1 # Duration of the signal (seconds)
8 step_size = 0.3 # Step size for Delta Modulation
9
10 # Generate a sine wave signal
11 time = np.arange(0, duration, 1 / sampling_rate)
12 frequency = 5 # Frequency of the sine wave (Hz)
13 input_signal = np.sin(2 * np.pi * frequency * time)
14
15 # Delta Modulation
16 delta_modulated_signal = []
17 quantized_signal = [0] # Initialize quantized signal with zero
18
```

```

19 for sample in input_signal:
20     # Determine the step (1 or -1)
21     if sample > quantized_signal[-1]:
22         delta_modulated_signal.append(1)
23         new_value = quantized_signal[-1] + step_size
24     else:
25         delta_modulated_signal.append(0)
26         new_value = quantized_signal[-1] - step_size
27     quantized_signal.append(new_value)
28
29 # Remove the initial zero from quantized_signal for plotting
30 quantized_signal = quantized_signal[1:]
31
32 # Delta Demodulation
33 reconstructed_signal = [0] # Initialize the reconstructed
34 signal
35 for bit in delta_modulated_signal:
36     if bit == 1:
37         new_value = reconstructed_signal[-1] + step_size
38     else:
39         new_value = reconstructed_signal[-1] - step_size
40     reconstructed_signal.append(new_value)
41
42 # Remove the initial zero from reconstructed_signal for
43 plotting
44 reconstructed_signal = reconstructed_signal[1:]
45
46 # Apply Low-Pass Filter to the Reconstructed Signal
47 def low_pass_filter(data, cutoff_freq, fs, order=4):
48     nyquist = fs / 2.0
49     normal_cutoff = cutoff_freq / nyquist
50     b, a = butter(order, normal_cutoff, btype='low',
51 analog=False)
52     filtered_signal = filtfilt(b, a, data)
53     return filtered_signal
54

```

```

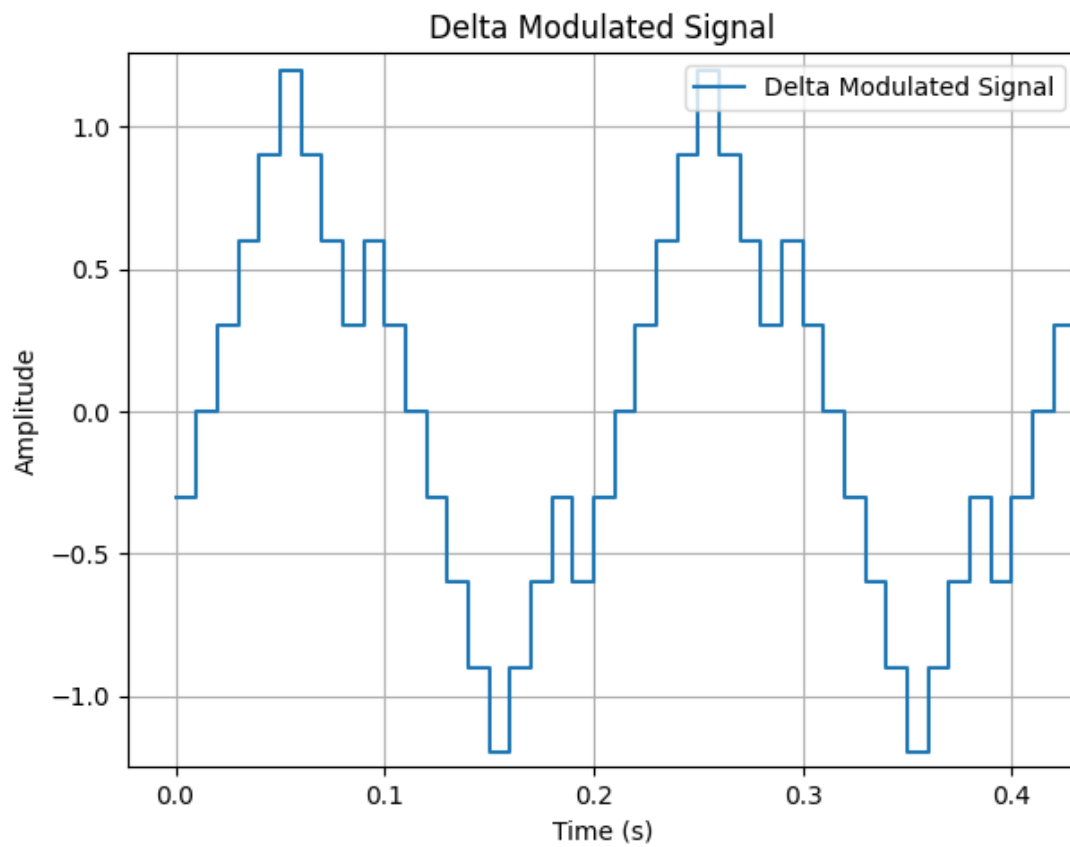
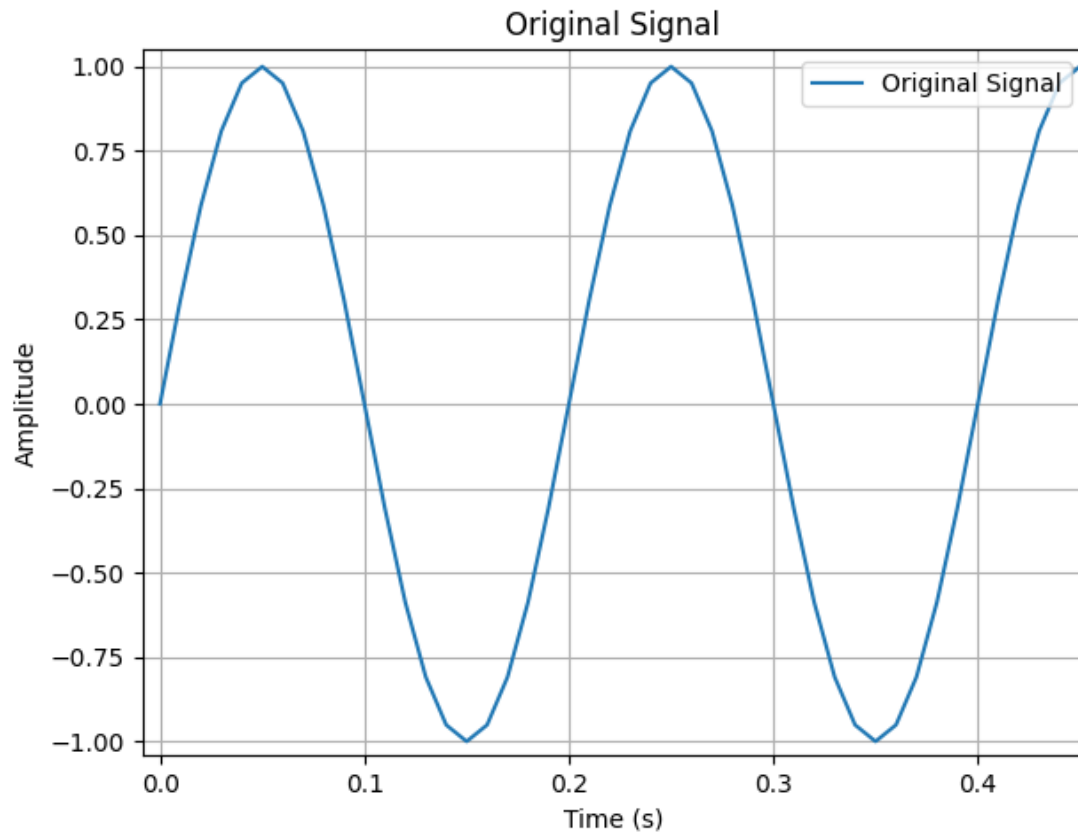
55  cutoff_frequency = 10 # Cutoff frequency for the LPF (Hz)
56  reconstructed_signal_lpf =
57  low_pass_filter(reconstructed_signal, cutoff_frequency,
58  sampling_rate)
59
60  # Plotting
61  fig, axs = plt.subplots(2, 3, figsize=(16, 10))
62
63  # Plot 1: Original Signal
64  axs[0, 0].plot(time, input_signal, label="Original Signal")
65  axs[0, 0].set_title("Original Signal")
66  axs[0, 0].set_xlabel("Time (s)")
67  axs[0, 0].set_ylabel("Amplitude")
68  axs[0, 0].grid()
69  axs[0, 0].legend(loc="upper right")
70
71  # Plot 2: Delta Modulated Signal
72  axs[0, 1].step(time, quantized_signal, label="Delta Modulated
73  Signal", where="post")
74  axs[0, 1].set_title("Delta Modulated Signal")
75  axs[0, 1].set_xlabel("Time (s)")
76  axs[0, 1].set_ylabel("Amplitude")
77  axs[0, 1].grid()
78  axs[0, 1].legend(loc="upper right")
79
80  # Plot 3: Original Signal on Top of Delta Modulated Signal
81  axs[0, 2].plot(time, input_signal, label="Original Signal")
82  axs[0, 2].step(time, quantized_signal, label="Delta Modulated
83  Signal", where="post", alpha=0.7)
84  axs[0, 2].set_title("Original Signal and Delta Modulated
85  Signal")
86  axs[0, 2].set_xlabel("Time (s)")
87  axs[0, 2].set_ylabel("Amplitude")
88  axs[0, 2].grid()
89  axs[0, 2].legend(loc="upper right")
90

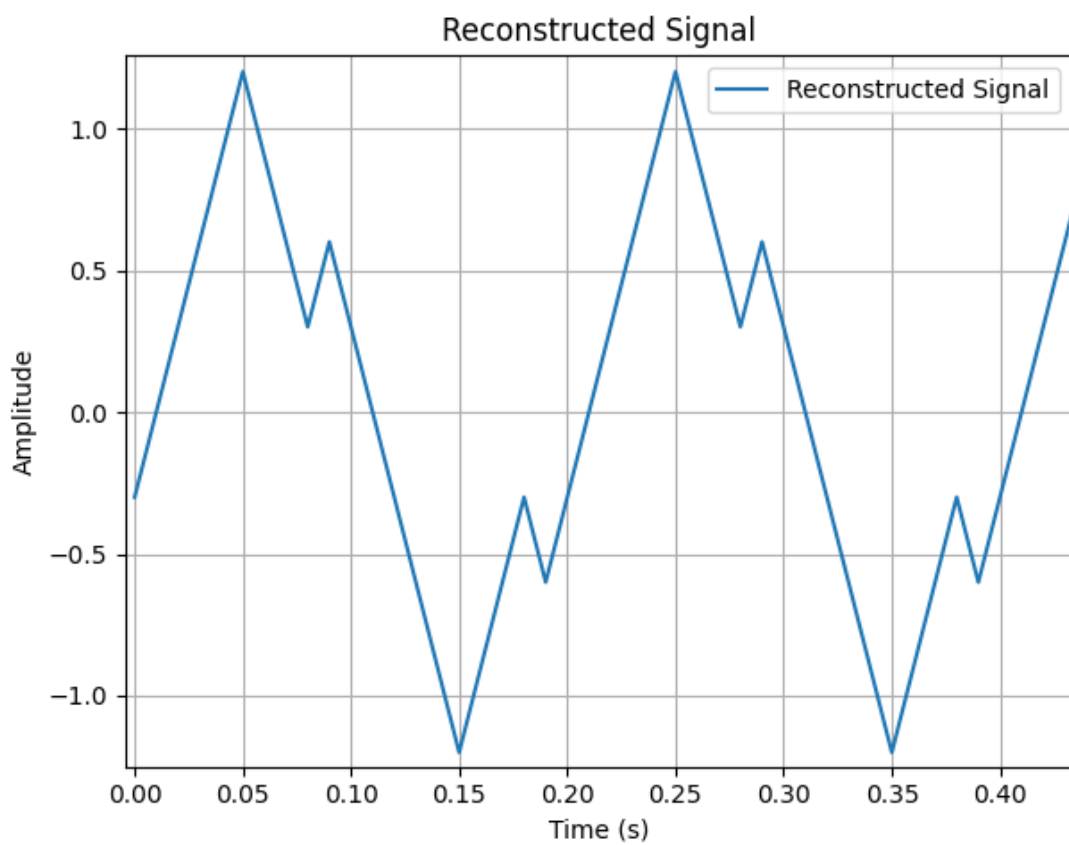
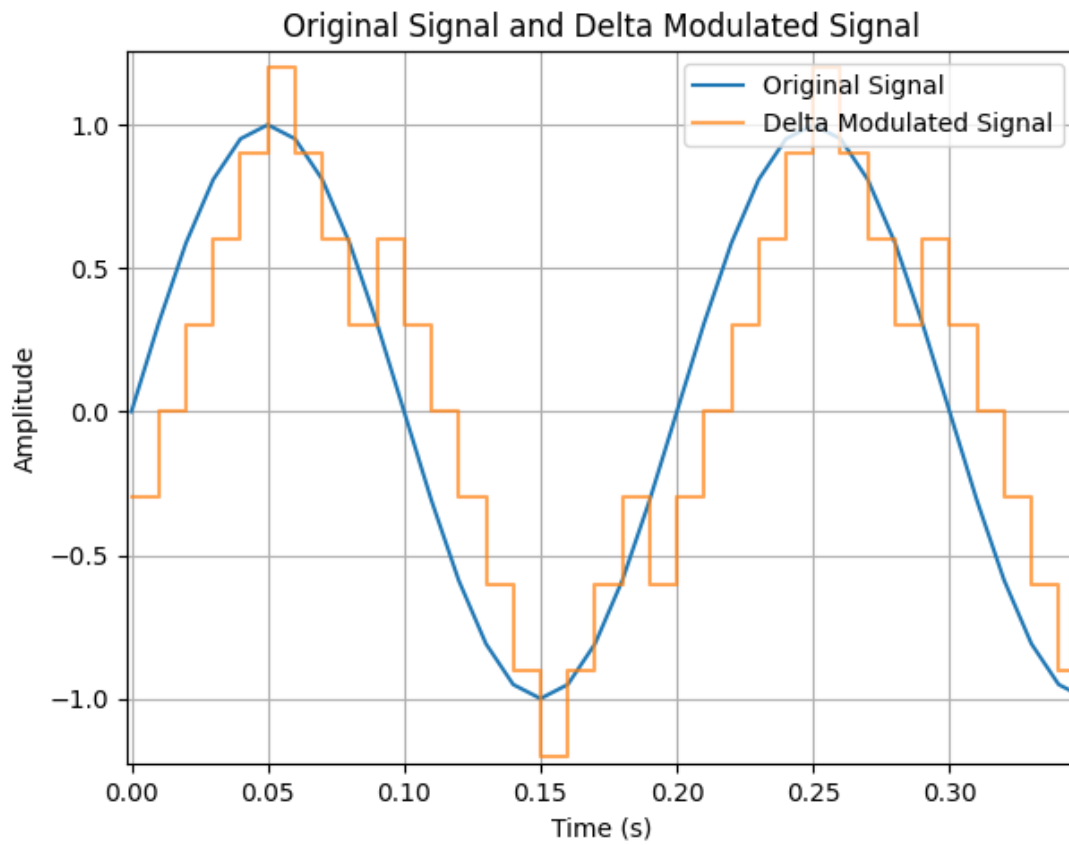
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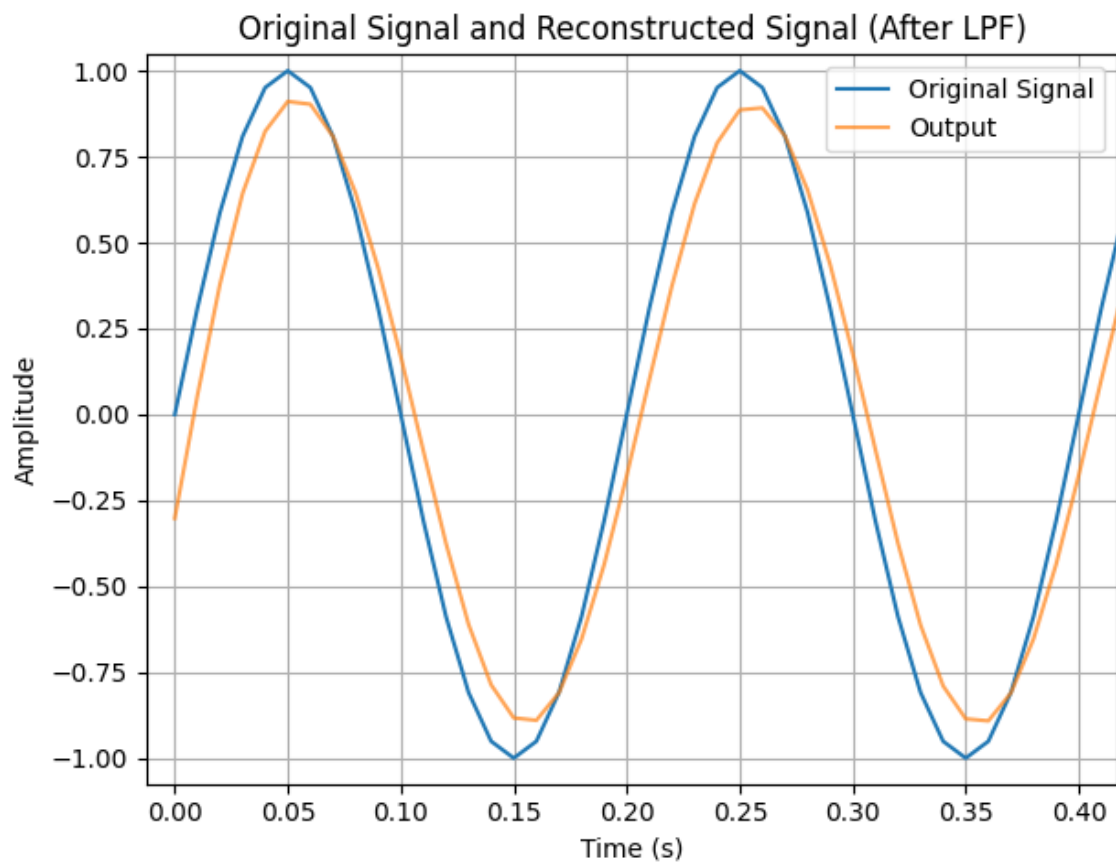
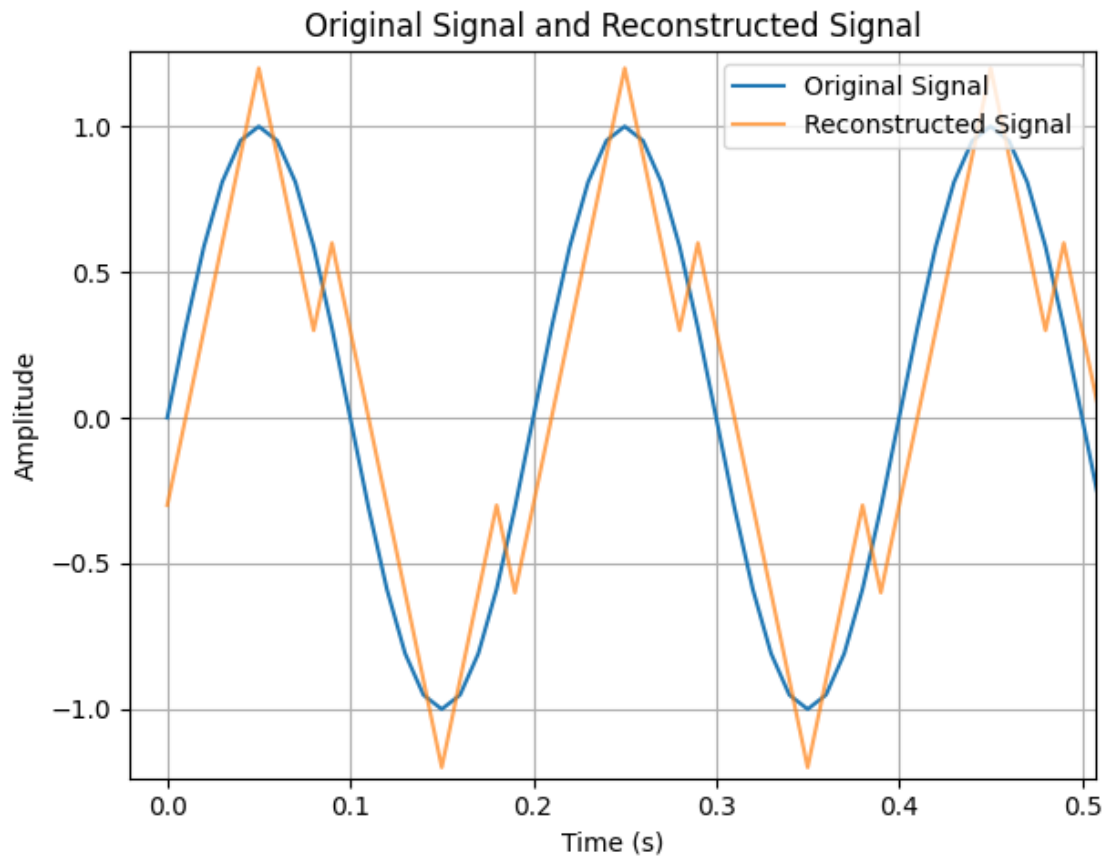
```

91 # Plot 4: Reconstructed Signal
92 axs[1, 0].plot(time, reconstructed_signal, label="Reconstructed
93 Signal")
94 axs[1, 0].set_title("Reconstructed Signal")
95 axs[1, 0].set_xlabel("Time (s)")
96 axs[1, 0].set_ylabel("Amplitude")
97 axs[1, 0].grid()
98 axs[1, 0].legend(loc="upper right")
99
100 # Plot 5: Reconstructed Signal on Top of Original Signal
101 axs[1, 1].plot(time, input_signal, label="Original Signal")
102 axs[1, 1].plot(time, reconstructed_signal, label="Reconstructed
103 Signal", alpha=0.7)
104 axs[1, 1].set_title("Original Signal and Reconstructed Signal")
105 axs[1, 1].set_xlabel("Time (s)")
106 axs[1, 1].set_ylabel("Amplitude")
107 axs[1, 1].grid()
108 axs[1, 1].legend(loc="upper right")
109
110 # Plot 6: Reconstructed Signal after LPF on Top of Original
111 Signal
112 axs[1, 2].plot(time, input_signal, label="Original Signal")
113 axs[1, 2].plot(time, reconstructed_signal_lpf, label="Output",
114 alpha=0.7)
115 axs[1, 2].set_title("Original Signal and Reconstructed Signal
116 (After LPF)")
117 axs[1, 2].set_xlabel("Time (s)")
118 axs[1, 2].set_ylabel("Amplitude")
119 axs[1, 2].grid()
120 axs[1, 2].legend(loc="upper right")
121
122 # Adjust Layout
123 plt.tight_layout()
124 plt.show()

```







Results

1. **Original Signal:** A sine wave with a frequency of 5 Hz is generated as the input.
2. **Delta Modulated Signal:** The staircase approximation of the input signal shows quantization.
3. **Reconstructed Signal:** The staircase output is the cumulative sum of step sizes from the binary sequence.
4. **Reconstructed Signal after LPF:** The Low-Pass Filter smooths the staircase into a continuous signal resembling the original.

Discussion

The implementation demonstrates the core principles of Delta Modulation and Demodulation.

- The **step size** plays a crucial role in determining the fidelity of the modulation and reconstruction process. Too large a step size increases granular noise, while a small step size may cause slope overload.
- The **Low-Pass Filter** effectively mitigates the staircase artifact, providing a closer match to the original signal.

Conclusion

Delta Modulation and Demodulation were successfully implemented and visualized in Python. By incorporating a Low-Pass Filter, the reconstructed signal closely approximates the original, highlighting the utility of post-processing in communication systems.