

# REPORT

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# Signal Synthesis Using Inverse Discrete Fourier Transform (IDFT)

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## 1 Problem Statement

The task involves synthesizing discrete-time signals from their DFT coefficients using the Inverse Discrete Fourier Transform (IDFT) in matrix notation. The goal is to reconstruct time-domain signals from frequency-domain sequences  $\mathbf{X}_\mu$  and visualize the results, including real and imaginary parts.

## 2 Input Data

The input data for this task is a sequence of DFT coefficients  $\mathbf{X}_\mu$ , defined as follows:

$$\mathbf{X}_\mu = [6, 2, 4, 4, 4, 5, 0, 0, 0, 0]$$

The corresponding time-domain signal is computed using the IDFT. The task also involves plotting the synthesized signal for analysis.

## 3 Python Code

The following Python code was used to synthesize signals from the input DFT coefficients using the IDFT:

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

def idft_matrix(N):
    W_N = np.exp(-2j * np.pi * np.outer(np.arange(N), np.arange(N))
                ) / N
    return W_N

def synthesize_signal(X_mu):
    N = len(X_mu)
    W_N = idft_matrix(N)
```

```

x_mu = np.dot(W_N.conj().T, X_mu) / N
return x_mu

X_mu = np.array([6, 2, 4, 4, 4, 5, 0, 0, 0], dtype=complex)

x_mu_synthesized = synthesize_signal(X_mu)

plt.figure(figsize=(10, 6))
plt.stem(np.real(x_mu_synthesized), linefmt='b-', markerfmt='bo',
         baselfmt='r-')
plt.title('Synthesized_Signal_(Real_Part)')
plt.xlabel('Sample_Index')
plt.ylabel('Amplitude')
plt.grid(True)
plt.show()

plt.figure(figsize=(10, 6))
plt.stem(np.imag(x_mu_synthesized), linefmt='g-', markerfmt='go',
         baselfmt='r-')
plt.title('Synthesized_Signal_(Imaginary_Part)')
plt.xlabel('Sample_Index')
plt.ylabel('Amplitude')
plt.grid(True)
plt.show()

plt.figure(figsize=(12, 6))
plt.plot(np.real(x_mu_synthesized), label='Real_Part', marker='o',
         linestyle='-', color='b')
plt.plot(np.imag(x_mu_synthesized), label='Imaginary_Part', marker=
         'x', linestyle='—', color='g')
plt.title('Signal_Synthesis:_Real_and_Imaginary_Parts_Combined')
plt.xlabel('Sample_Index')
plt.ylabel('Amplitude')
plt.axhline(0, color='red', linewidth=0.8, linestyle='—', label='
         Zero_Line')
plt.legend()
plt.grid(True)
plt.show()

```

## 4 Outcomes

The synthesized signals were analyzed and visualized. The results include the real part, the imaginary part, and the combined representation of both.

### 4.1 Real Part of the Signal

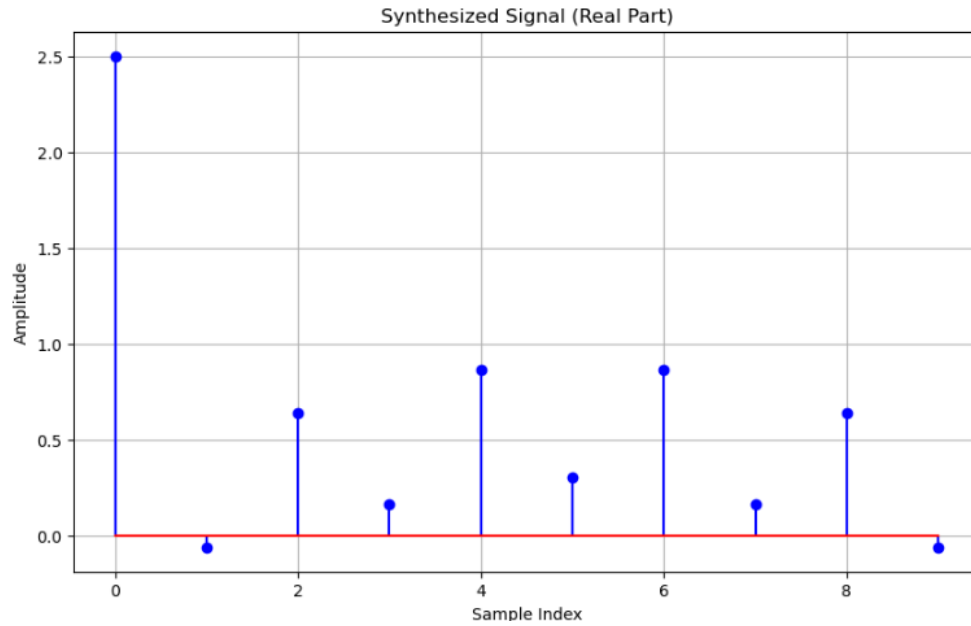


Figure 1: Synthesized Signal (Real Part)

## 4.2 Imaginary Part of the Signal

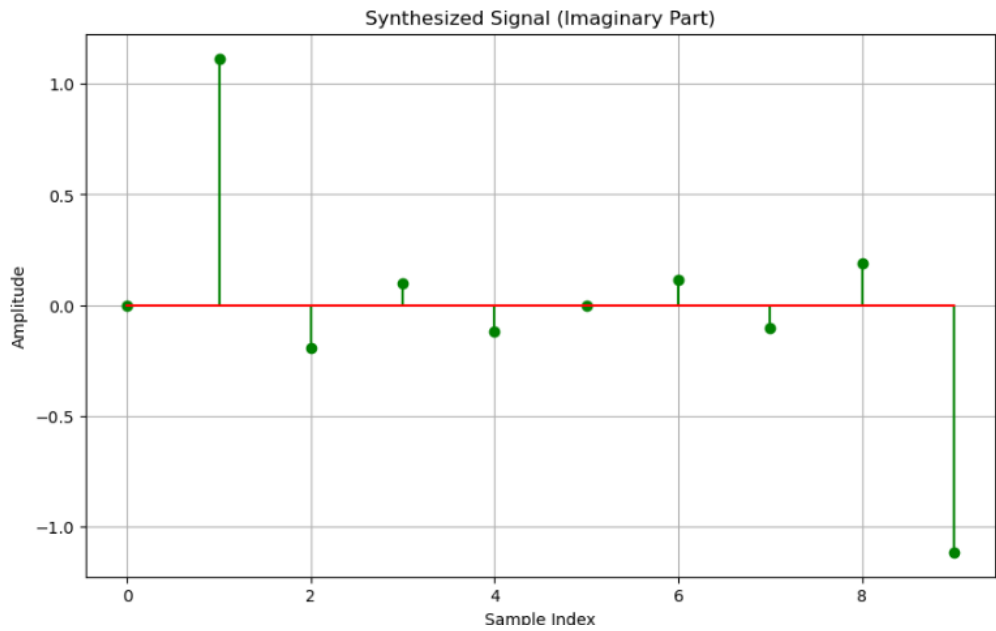


Figure 2: Synthesized Signal (Imaginary Part)

### 4.3 Combined Signal (Real and Imaginary)

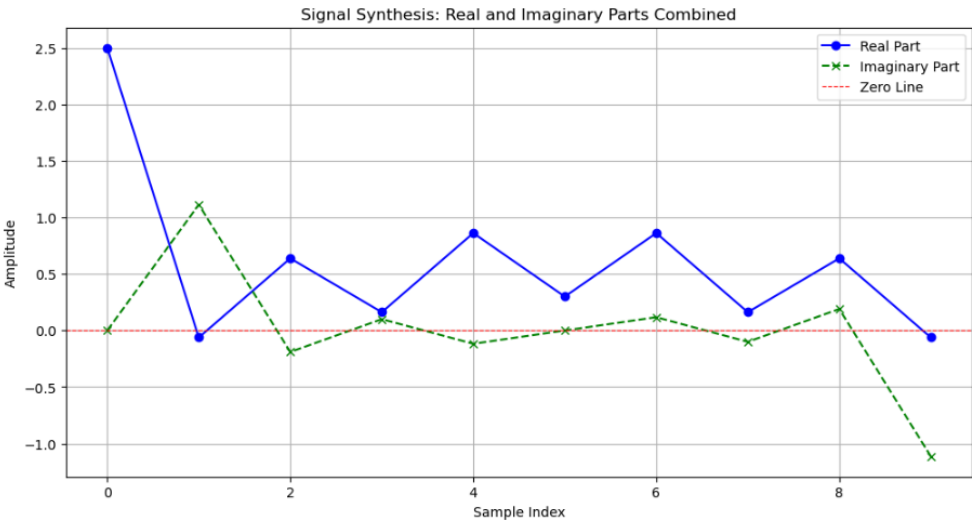


Figure 3: Signal Synthesis: Real and Imaginary Parts Combined

## 5 Conclusions

- The comparison between the real and imaginary parts of the signal shows how these two components are interconnected in the signal synthesis process.
- The real part of the signal is more intuitive and easier to understand, as it reflects the signal's amplitude, while the imaginary part provides information about its phase.
- Both parts are essential to obtain a complete and accurate time-domain reconstruction of the signal.
- The zero line plays a crucial role in visualizing the signal's behavior, providing a reference point for identifying the positive and negative oscillations.
- The real and imaginary components fluctuate around this line, and it helps to better interpret the amplitude and phase changes, ensuring a more accurate understanding of the signal's characteristics.