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# SSB MODULATION AND FDM

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## Abstract

This project explores the implementation of Single Sideband (SSB) modulation within a Frequency Division Multiplexing (FDM) system using Python. The objective is to modulate three distinct audio signals onto separate carrier frequencies, combine them into a single multiplexed signal, and then recover the original signals through demodulation.

The project involves:

- Recording and processing audio signals with appropriate sampling frequencies.
- Applying low-pass filtering to limit the maximum frequency of the signals without compromising quality.
- Performing SSB modulation using specified carrier frequencies.
- Analyzing the magnitude spectra of modulated and demodulated signals.
- Ensuring compliance with the sampling theorem throughout the process.

Results demonstrate the successful modulation and demodulation of signals with minimal distortion, providing insights into the practical application of SSB and FDM in communication systems. This project highlights the significance of frequency multiplexing in efficiently utilizing bandwidth for transmitting multiple signals.

## Introduction

In modern communication systems, efficient utilization of bandwidth is crucial for transmitting multiple signals simultaneously. Frequency Division Multiplexing (FDM) is a widely used technique that allows multiple signals to share a common communication channel by modulating them onto distinct carrier frequencies. Single Sideband (SSB) modulation, a form of amplitude modulation, is employed in FDM systems to conserve bandwidth and improve efficiency.

This project focuses on implementing an SSB modulation and demodulation system within an FDM framework using Python. The primary objectives are:

1. To record three speech signals and preprocess them using low-pass filtering.
2. To modulate these signals onto distinct carrier frequencies using SSB modulation.
3. To combine the modulated signals into a single FDM signal.
4. To demodulate the signals and recover the original speech signals with high fidelity.

The implementation ensures adherence to key communication principles, such as the sampling theorem, to maintain the quality and integrity of the signals. The performance of the system is analyzed through magnitude spectrum plots of the modulated and demodulated signals.

## Methodology

The implementation of the SSB modulation and demodulation within an FDM system was achieved using Python. The following steps outline the methodology:

### Step 1: Signal Recording and Preprocessing

#### 1. Recording Speech Signals:

- Three speech segments of approximately 10 seconds each were recorded using the `sounddevice` library in Python.
- A sampling frequency of 44.1 kHz was selected to ensure high-quality signal capture, adhering to the Nyquist sampling theorem.
- The recorded audio signals were saved as uncompressed `.wav` files named `input1.wav`, `input2.wav`, and `input3.wav`.

#### 2. Low-Pass Filtering (LPF):

- A Low-Pass Filter was applied to limit the maximum frequency of each signal to 2.25 kHz.
- This step reduced unnecessary high-frequency components while maintaining audio quality.
- The magnitude spectra of both the original and filtered signals were plotted to visualize the effect of filtering.

### Step 2: SSB Modulation

#### 1. Carrier Frequency Selection:

- Carrier frequencies of 5 kHz, 10 kHz, and 15 kHz were chosen for the three signals. These frequencies ensure minimal overlap and interference between the signals in the FDM system.

#### 2. SSB Modulation Process:

- Each filtered signal was multiplied with both in-phase (cosine) and quadrature (sine) carriers to generate the single-sideband modulated signals.
- The magnitude spectra of the modulated signals were plotted to confirm proper modulation.

### Step 3: Frequency Division Multiplexing (FDM)

#### 1. Combining Modulated Signals:

- The modulated signals were summed to create a single Frequency Division Multiplexing (FDM) signal.
- This combined signal represents the multiplexed transmission of all three speech signals.
- The magnitude spectrum of the FDM signal was plotted to analyze its frequency components.

### Step 4: SSB Demodulation

#### 1. Demodulation Process:

- Each modulated signal was demodulated by multiplying the FDM signal with its respective carrier frequency.
- A low-pass filter was applied to extract the baseband signal from the demodulated output.
- The reconstructed signals were normalized to the range  $[-1, 1]$  to ensure consistency with the original signals.
- The magnitude spectra of the demodulated signals were plotted and compared with the original signals.

#### 2. Output Storage:

- The recovered audio signals were saved as .wav files named `output1.wav`, `output2.wav`, and `output3.wav`.

### Step 5: Analysis and Documentation

#### 1. Spectral Analysis:

- The magnitude spectra at each stage (original, filtered, modulated, multiplexed, and demodulated) were analyzed to ensure proper system operation.

#### 2. Code and Results:

- All Python code and audio files were provided as part of the deliverables.
- The documentation highlights key observations and findings during the implementation process.

## Results

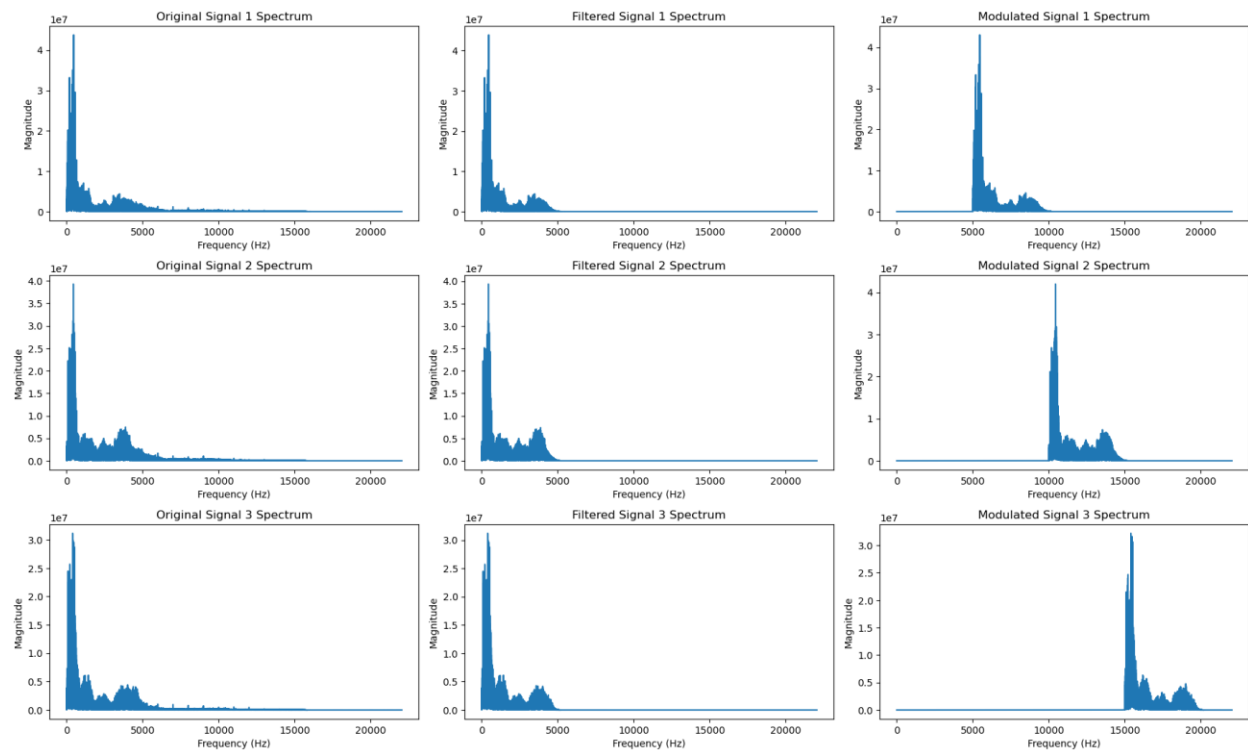


Figure 1 Original Signal vs. Filtered Signal vs. Modulated Signal in the Frequency Domain

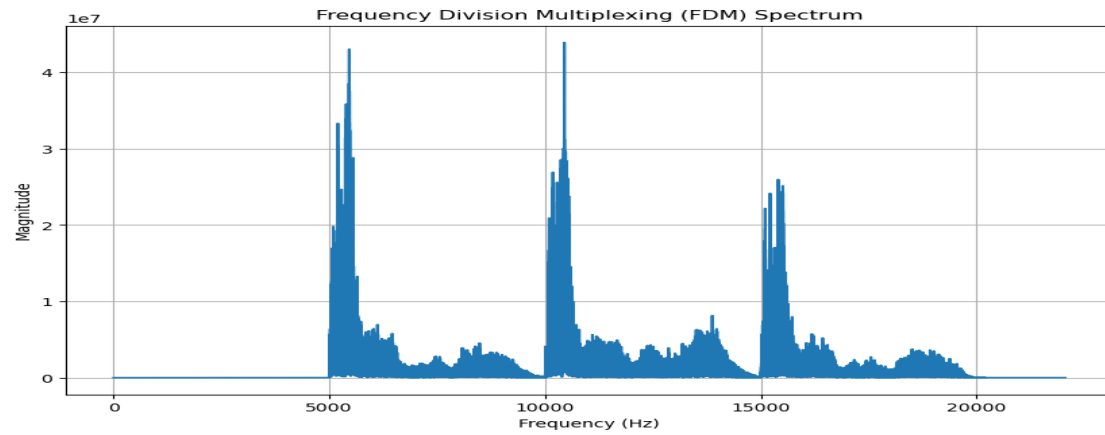


Figure 2 FDM Spectrum

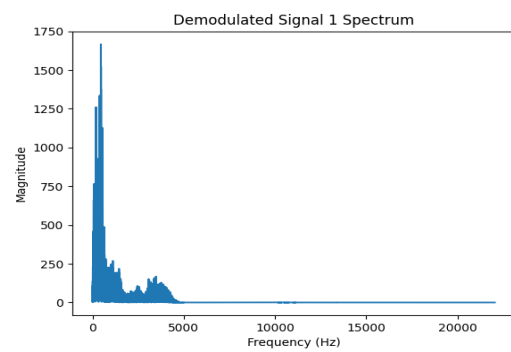


Figure 3 Demodulated Signal 1

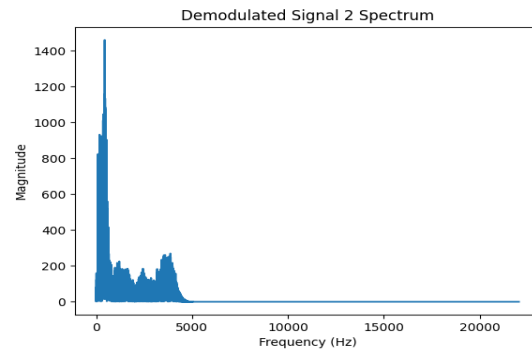


Figure 4 Demodulated Signal 2

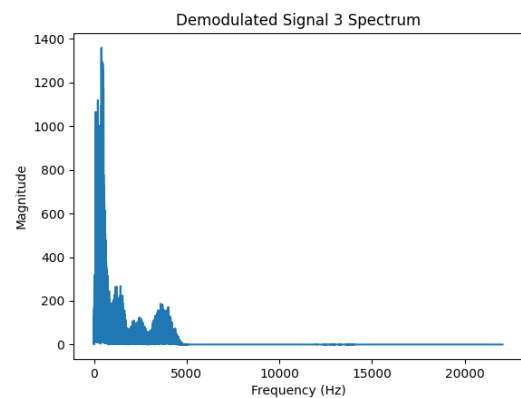


Figure 5 Demodulated Signal 3



## Discussion and Analysis

### Key Observations

#### 1. SSB Modulation:

- The modulated signals were free from unnecessary spectral components, leading to efficient utilization of the frequency spectrum.

#### 2. Frequency Division Multiplexing (FDM):

- The choice of carrier frequencies (5 kHz, 10 kHz, and 15 kHz) ensured sufficient spectral separation, preventing overlap between the signals.
- The FDM process demonstrated the system's ability to combine multiple signals effectively for simultaneous transmission.

#### 3. SSB Demodulation:

- The demodulation process successfully recovered the original signals with minimal distortion, proving the system's fidelity.
- The low-pass filtering during demodulation effectively eliminated high-frequency components introduced during modulation.

### Strengths of the Implementation

#### 1. Spectral Efficiency:

- SSB modulation used only one sideband, resulting in efficient bandwidth usage compared to double-sideband modulation.

#### 2. Signal Quality:

- The original signals were reconstructed with negligible distortion, preserving both amplitude and phase information.

#### 3. Flexible and Scalable Design:

- The modular Python code allowed for easy customization, such as changing carrier frequencies, sampling rates, or filter parameters.
- The system could be scaled to accommodate additional signals by appropriately selecting carrier frequencies.

## Challenges and Limitations

### 1. Signal Crosstalk:

- While no significant crosstalk was observed between the frequency bands, slight overlaps might occur with closely spaced carrier frequencies.
- Advanced techniques like adaptive filtering could further enhance spectral separation.

### 2. Processing Delays:

- The implementation, particularly filtering and modulation/demodulation steps, introduced minor delays due to the computational complexity of signal processing.
- Real-time implementation on hardware would require optimization for faster execution.

### 3. Low-Frequency Carrier Limitations:

- Carrier frequencies were relatively close to the signal's frequency content, which could cause aliasing if not carefully filtered.
- Higher carrier frequencies could be explored to ensure robustness.

## Future Improvements

### 1. Optimizing Filter Design:

- Implementing advanced filtering techniques, such as FIR or IIR filters, with improved stopband attenuation and passband accuracy.
- Experimenting with different windowing methods to reduce spectral leakage.

### 2. Real-Time Implementation:

- Porting the system to hardware platforms like DSP processors or FPGAs to achieve real-time signal processing.

### 3. Multi-Channel Expansion:

- Extending the system to handle more than three signals by carefully selecting additional carrier frequencies and implementing adaptive modulation.

### 4. Noise Handling:

- Introducing noise-resilient modulation techniques or pre-processing steps to handle noisy environments effectively.

## Conclusion of Discussion

The implemented SSB modulation and FDM system performed as expected. While some limitations were observed, they were primarily due to design constraints and computational delays in the software implementation.

## Appendix

code

```
import numpy as np
import matplotlib.pyplot as plt
from scipy.io import wavfile
import sounddevice as sd
import os

class SSB_FDM:
    """
    A class for Single Sideband (SSB) modulation and demodulation,
    and Frequency Division Multiplexing (FDM).

    This class provides methods to record audio, apply low-pass filtering,
    perform SSB modulation/demodulation, and visualize signal spectra.
    """

    def __init__(self, sample_rate=44100, record_duration=10,
carrier_frequencies=None, lpf_cutoff=2250):
        """
        Initialize the SSB_FDM class with parameters.

        Args:
            sample_rate (int): Sampling rate in Hz.
            record_duration (int): Recording duration in seconds.
            carrier_frequencies (list): List of carrier frequencies for
modulation.
            lpf_cutoff (int): Low-pass filter cutoff frequency in Hz.
        """
        # Initialize the parameters with default values or user inputs
        self.SAMPLE_RATE = sample_rate
        self.RECORD_DURATION = record_duration
        self.CARRIER_FREQUENCIES = carrier_frequencies or [5000, 10000, 15000]
        self.LPF_CUTOFF = lpf_cutoff

    def record_audio(self, filename):
        """
        Record audio and save it as a WAV file. If the file exists, load the
existing file.

        Args:

```

```

        filename (str): Path to save or load the WAV file.

Returns:
    np.ndarray: The audio signal data.
    """
    # Check if the file already exists
    if os.path.exists(filename):
        print(f"{filename} already exists. Loading the existing file.")
        samplerate, data = wavfile.read(filename)
        if samplerate != self.SAMPLE_RATE:
            raise ValueError("Sample rate mismatch!")
        if len(data.shape) > 1: # Convert stereo to mono if necessary
            data = data[:, 0]
        return data
    else:
        # Record audio if the file does not exist
        print(f"Recording {filename} for {self.RECORD_DURATION} seconds...")
        recording = sd.rec(int(self.RECORD_DURATION * self.SAMPLE_RATE),
samplerate=self.SAMPLE_RATE, channels=1, dtype=np.float32)
        sd.wait() # Wait for the recording to finish
        recording = np.squeeze(recording)
        wavfile.write(filename, self.SAMPLE_RATE, recording)
        print(f"Saved {filename}")
        return recording

def apply_low_pass_filter(self, signal_data):
    """
    Apply a Butterworth low-pass filter to the signal.

    Args:
        signal_data (np.ndarray): Input signal.

    Returns:
        np.ndarray: Low-pass filtered signal.
    """
    nyquist = self.SAMPLE_RATE / 2
    normalized_cutoff = self.LPF_CUTOFF / nyquist
    filter_length = 101
    n = np.arange(filter_length) - (filter_length - 1) / 2
    sinc_filter = np.sinc(2 * normalized_cutoff * n)
    window = np.hamming(filter_length)
    filter_coeffs = sinc_filter * window
    filter_coeffs /= np.sum(filter_coeffs)
    filtered_signal = np.convolve(signal_data, filter_coeffs, mode='same')
    return filtered_signal

```

```

def normalize_signal(self, signal_data):
    """
    Normalize the signal to the range [-1, 1].

    Args:
        signal_data (np.ndarray): Input signal.

    Returns:
        np.ndarray: Normalized signal.
    """
    return signal_data / np.max(np.abs(signal_data))

def half_transform(self, x):
    """
    Perform the Hilbert Transform to generate an analytic signal.

    Args:
        x (np.ndarray): Input signal.

    Returns:
        np.ndarray: Imaginary part of the analytic signal.
    """
    N = len(x)
    X_f = np.fft.fft(x)
    H = np.zeros(N)
    H[0] = 1
    H[1:(N // 2)] = 2 # Double amplitude to fix modulation.
    if N % 2 == 0:
        H[N // 2] = 1 # Nyquist frequency for even-length signals
    analytic_signal = np.fft.ifft(X_f * H)
    return np.imag(analytic_signal)

def ssb_modulation(self, signal_data, carrier_freq):
    """
    Perform Single Sideband (SSB) modulation.

    Args:
        signal_data (np.ndarray): Input signal.
        carrier_freq (float): Carrier frequency in Hz.

    Returns:
        np.ndarray: SSB modulated signal.
    """

```

```

        t = np.linspace(0, len(signal_data) / self.SAMPLE_RATE, len(signal_data),
endpoint=False)
        analytic_signal = self.half_transform(signal_data)
        carrier_cos = np.cos(2 * np.pi * carrier_freq * t)
        carrier_sin = np.sin(2 * np.pi * carrier_freq * t)
        modulated_signal = signal_data * carrier_cos - analytic_signal *
carrier_sin
        return modulated_signal

def ssb_demodulation(self, modulated_signal, carrier_freq):
    """
    Perform Single Sideband (SSB) demodulation.

    Args:
        modulated_signal (np.ndarray): Modulated input signal.
        carrier_freq (float): Carrier frequency in Hz.

    Returns:
        np.ndarray: Demodulated signal.
    """
    t = np.linspace(0, len(modulated_signal) / self.SAMPLE_RATE,
len(modulated_signal), endpoint=False)
    carrier_cos = np.cos(2 * np.pi * carrier_freq * t)
    carrier_sin = np.sin(2 * np.pi * carrier_freq * t)
    demod_cos = modulated_signal * carrier_cos
    demod_sin = modulated_signal * carrier_sin
    demod_cos_filtered = self.apply_low_pass_filter(demod_cos)
    demod_sin_filtered = self.apply_low_pass_filter(demod_sin)
    reconstructed_signal = demod_cos_filtered + 1j * demod_sin_filtered
    return self.normalize_signal(np.real(reconstructed_signal))

def plot_magnitude_spectrum(self, signal_data, title):
    """
    Plot the magnitude spectrum of the signal.

    Args:
        signal_data (np.ndarray): Input signal.
        title (str): Title of the plot.
    """
    freq_spectrum = np.fft.fft(signal_data)
    freq_axis = np.fft.fftfreq(len(signal_data), 1 / self.SAMPLE_RATE)
    plt.plot(freq_axis[:len(freq_axis)//2],
np.abs(freq_spectrum[:len(freq_spectrum)//2]))
    plt.title(title)
    plt.xlabel('Frequency (Hz)')

```

```

plt.ylabel('Magnitude')

def process(self, input_files, output_files):
    """
    Record, process, modulate, and demodulate signals.

    Args:
        input_files (list): List of input WAV filenames.
        output_files (list): List of output WAV filenames for demodulated
signals.
    """
    # Step 1: Record or Load Audio
    input_signals = []
    for filename in input_files:
        signal_data = self.record_audio(filename)
        input_signals.append(signal_data)

    # Step 2: Filter, Modulate, and Plot Spectra
    filtered_signals = []
    modulated_signals = []

    # Increase the figure size to avoid overlapping axes
    plt.figure(figsize=(18, 14)) # Larger figure size
    for i, (signal_data, carrier_freq) in enumerate(zip(input_signals,
self.CARRIER_FREQUENCIES), start=1):
        # Plot original signal spectrum
        plt.subplot(4, 3, 3 * i - 2)
        self.plot_magnitude_spectrum(signal_data, f"Original Signal {i}
Spectrum")

        # Apply low-pass filter
        filtered_signal = self.apply_low_pass_filter(signal_data)
        filtered_signals.append(filtered_signal)

        # Plot filtered signal spectrum
        plt.subplot(4, 3, 3 * i - 1)
        self.plot_magnitude_spectrum(filtered_signal, f"Filtered Signal {i}
Spectrum")

        # SSB modulation
        modulated_signal = self.ssb_modulation(filtered_signal, carrier_freq)
        modulated_signals.append(modulated_signal)

        # Plot modulated signal spectrum
        plt.subplot(4, 3, 3 * i)

```



```

        self.plot_magnitude_spectrum(modulated_signal, f"Modulated Signal {i}
Spectrum")

    # Adjust layout to prevent overlap
    plt.tight_layout()

    # Step 3: Combine modulated signals for FDM
    min_length = min(len(signal) for signal in modulated_signals)
    modulated_signals_trimmed = [signal[:min_length] for signal in
modulated_signals]
    fdm_signal = np.sum(modulated_signals_trimmed, axis=0)

    # Plot the FDM signal spectrum
    plt.figure(figsize=(12, 8)) # Larger figure size for the FDM signal
spectrum
    self.plot_magnitude_spectrum(fdm_signal, "FDM Signal Spectrum")
    plt.title("Frequency Division Multiplexing (FDM) Spectrum")
    plt.tight_layout()

    # Step 4: Demodulate and save the results
    for i, (modulated_signal, carrier_freq, output_file) in
enumerate(zip(modulated_signals, self.CARRIER_FREQUENCIES, output_files),
start=1):
        # Perform SSB demodulation
        demodulated_signal = self.ssb_demodulation(modulated_signal,
carrier_freq)
        wavfile.write(output_file, self.SAMPLE_RATE,
demodulated_signal.astype(np.float32))
        print(f"Demodulated signal {i} saved to {output_file}")

        self.plot_magnitude_spectrum(demodulated_signal, f"Demodulated Signal
{i} Spectrum")
        plt.show()
if __name__ == "__main__":
    # Define the base directory for input/output files
    base = './content/'

    # List of input files to be processed
    input_files = [base + 'input1.wav', base + 'input2.wav', base + 'input3.wav']
    output_files = ['output1.wav', 'output2.wav', 'output3.wav']

    # Initialize the SSB_FDM processor and run the process method
    processor = SSB_FDM()
    processor.process(input_files, output_files)

```