

# B. Deepak Kumar

AIE21028 

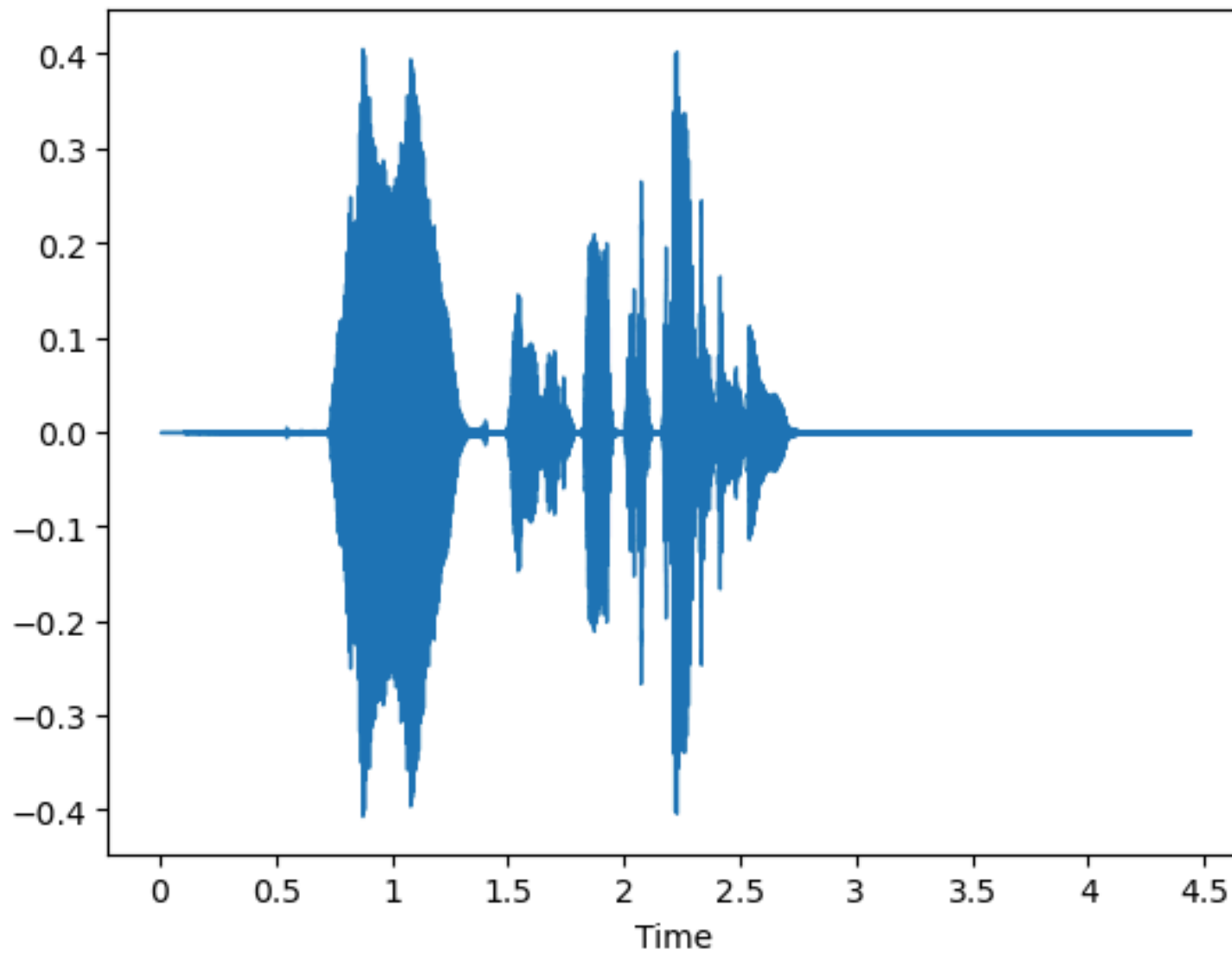
## LAB - 4

**A1. Use `numpy.fft.fft()` to transform the speech signal to its spectral domain. Please plot the amplitude part of the spectral components and observe it.**

```
In [82]: 1 import numpy as np
          2 import librosa
          3 import matplotlib.pyplot as plt
          4 import IPython.display as ipd
          5 import scipy.signal as signal
          6 import scipy.io.wavfile as wavfile
          7 from glob import glob
          8 import seaborn as sns
          9 from scipy.signal import spectrogram
         10
```

```
In [22]: 1 y, sr = librosa.load('AISPS.wav')
          2 librosa.display.waveshow(y)
          3
```

```
Out[22]: <librosa.display.AdaptiveWaveplot at 0x1eb33958c40>
```



In [23]:

```
1 a = glob('AISPS.wav')
2 ipd.Audio(a[0])
```

Out[23]:



In [27]:

```
1 # Use numpy.fft.fft() to transform the speech signal to its spectral domain
2 fft_result = np.fft.fft(y)
3 print("after fft:")
4 ipd.display(ipd.Audio(fft_result, rate=sr))
```

after fft:

C:\Users\said\anaconda3\lib\site-packages\IPython\lib\display.py:159: ComplexWarning: Casting complex values to real discards the imaginary part  
data = np.array(data, dtype=float)



In [29]:

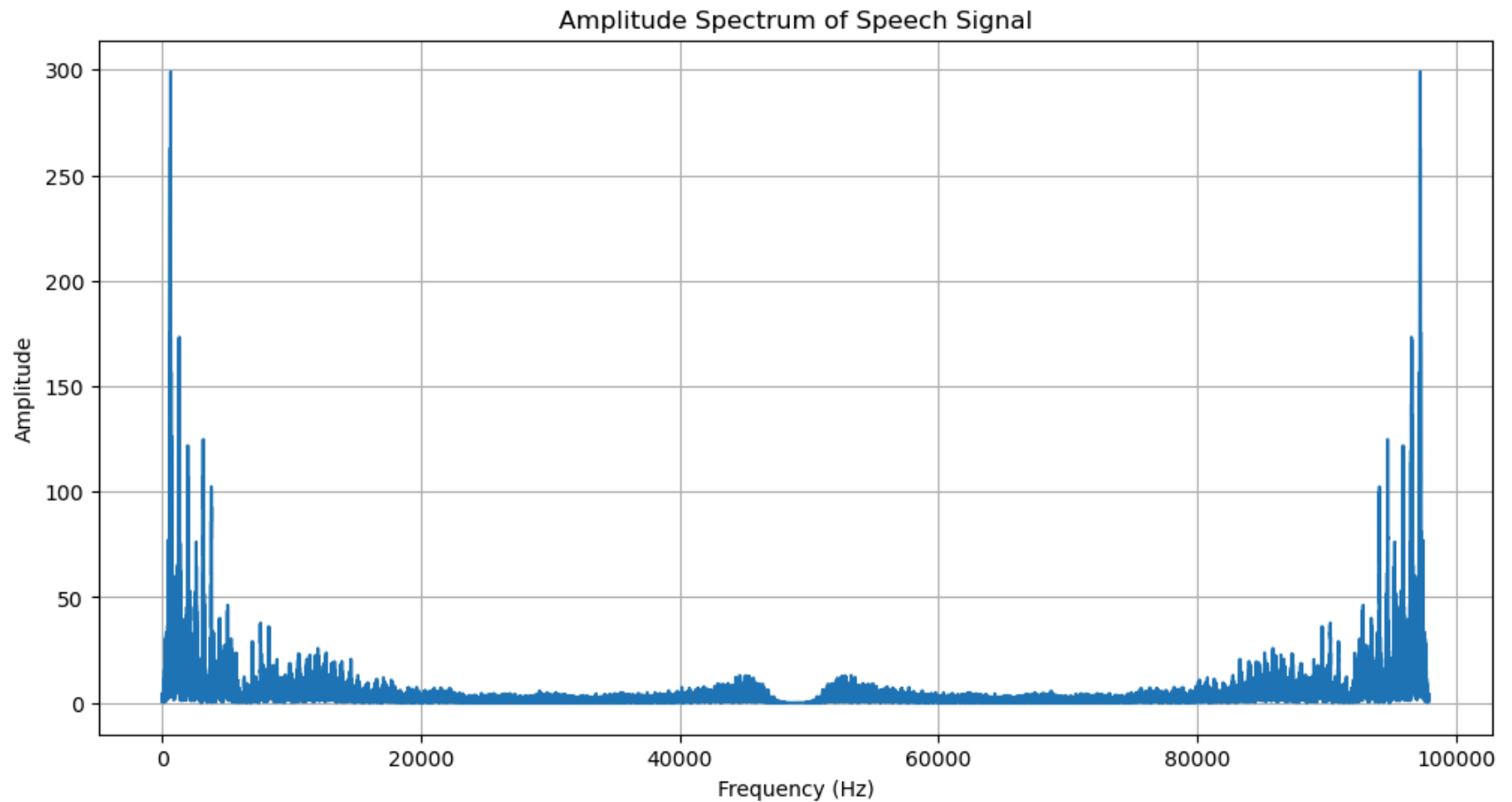
```
1 # Calculate the amplitude spectrum (absolute values of the complex numbers)
2 amplitude_spectrum = np.abs(fft_result)
3 print("amplitude spectrum")
4 ipd.display(ipd.Audio(amplitude_spectrum, rate=sr))
```

amplitude spectrum



In [8]:

```
1 plt.figure(figsize=(12, 6))
2 plt.plot(amplitude_spectrum)
3 plt.title('Amplitude Spectrum of Speech Signal')
4 plt.xlabel('Frequency (Hz)')
5 plt.ylabel('Amplitude')
6 plt.grid(True)
7 plt.show()
```



**A2. Use `numpy.fft.ifft()` to inverse transform the frequency spectrum of the speech signal from frequency domain to time domain. Compare the generated time domain signal with the original signal.**

In [30]:

```
1 # Use numpy.fft.ifft() to transform the speech signal from frequency domain to it
2 ifft_result = np.fft.ifft(fft_result)
3 print("after reconstruction")
4 ipd.display(ipd.Audio(ifft_result, rate=sr))
```

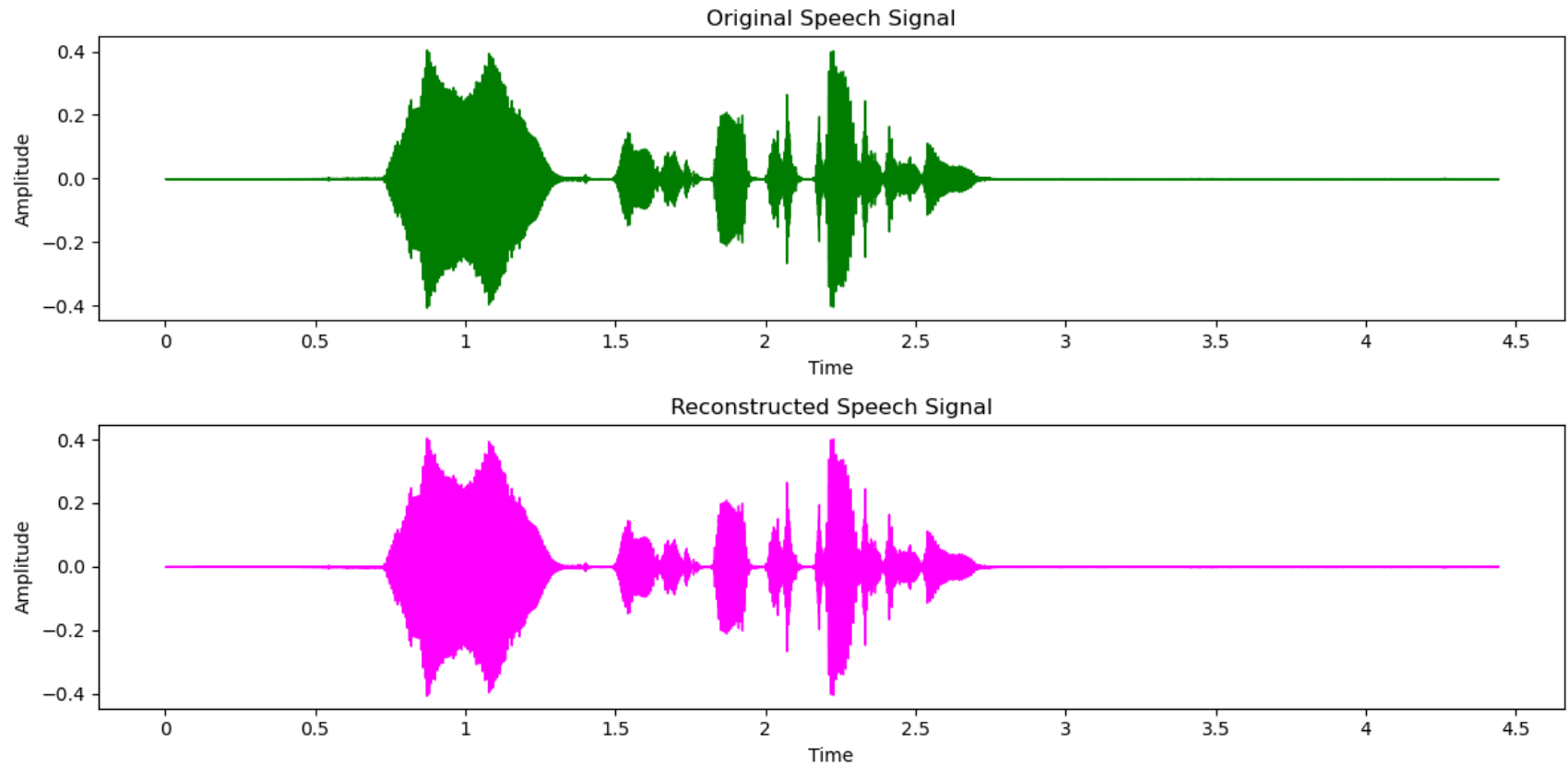
after reconstruction



In [19]:

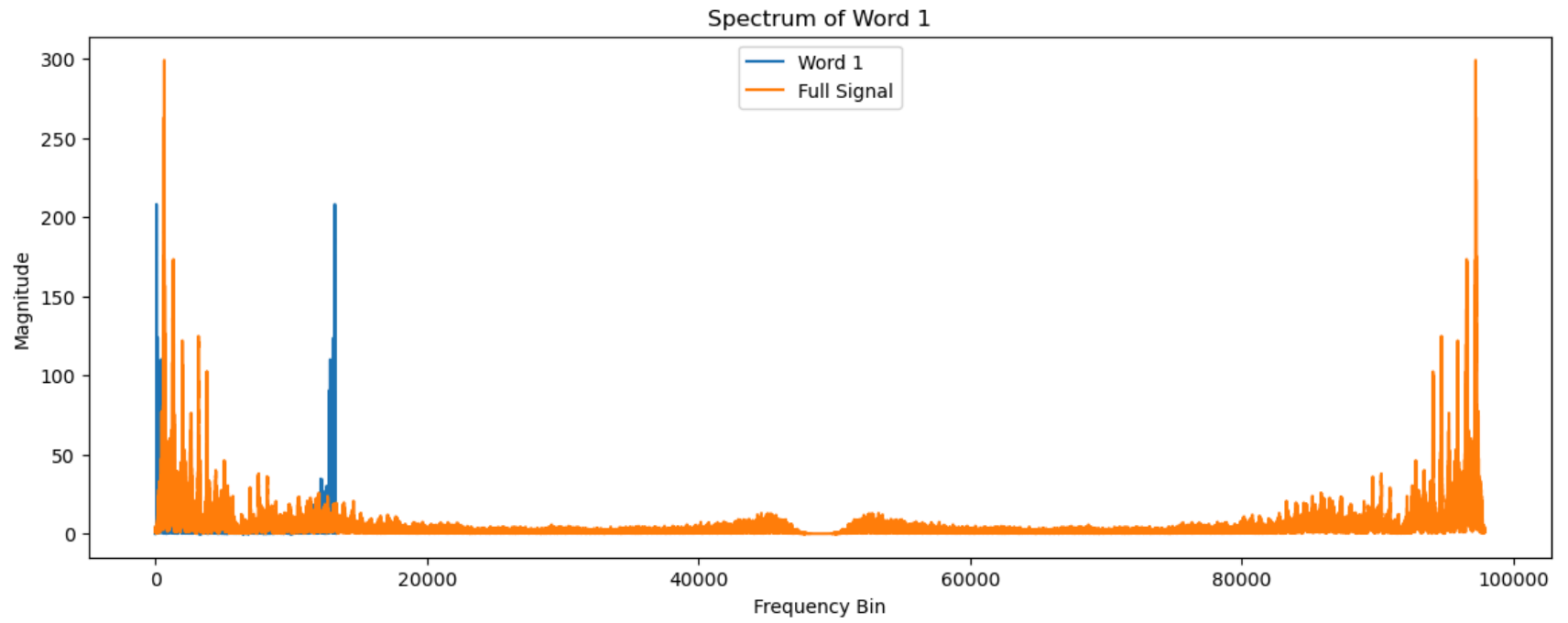
```
1  # Plot the original and reconstructed signals for comparison
2  plt.figure(figsize=(12, 6))
3
4  # Plot the original signal
5  plt.subplot(2, 1, 1)
6  librosa.display.waveshow(y, sr=sr, color='green')
7  plt.title('Original Speech Signal')
8  plt.xlabel('Time')
9  plt.ylabel('Amplitude')
10
11 # Plot the reconstructed signal
12 plt.subplot(2, 1, 2)
13 librosa.display.waveshow(np.real(iff_t_result), sr=sr, color='magenta') # Use np
14 plt.title('Reconstructed Speech Signal')
15 plt.xlabel('Time')
16 plt.ylabel('Amplitude')
17
18 plt.tight_layout()
19 plt.show()
```

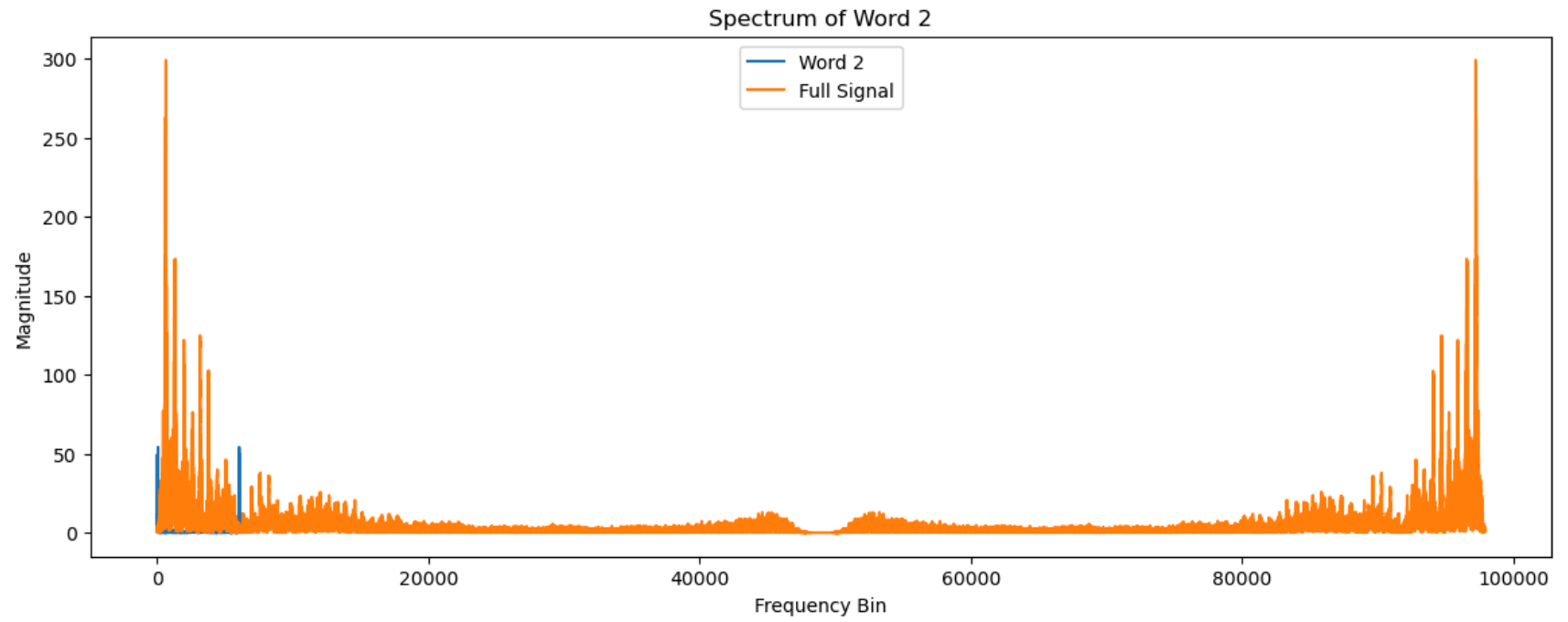


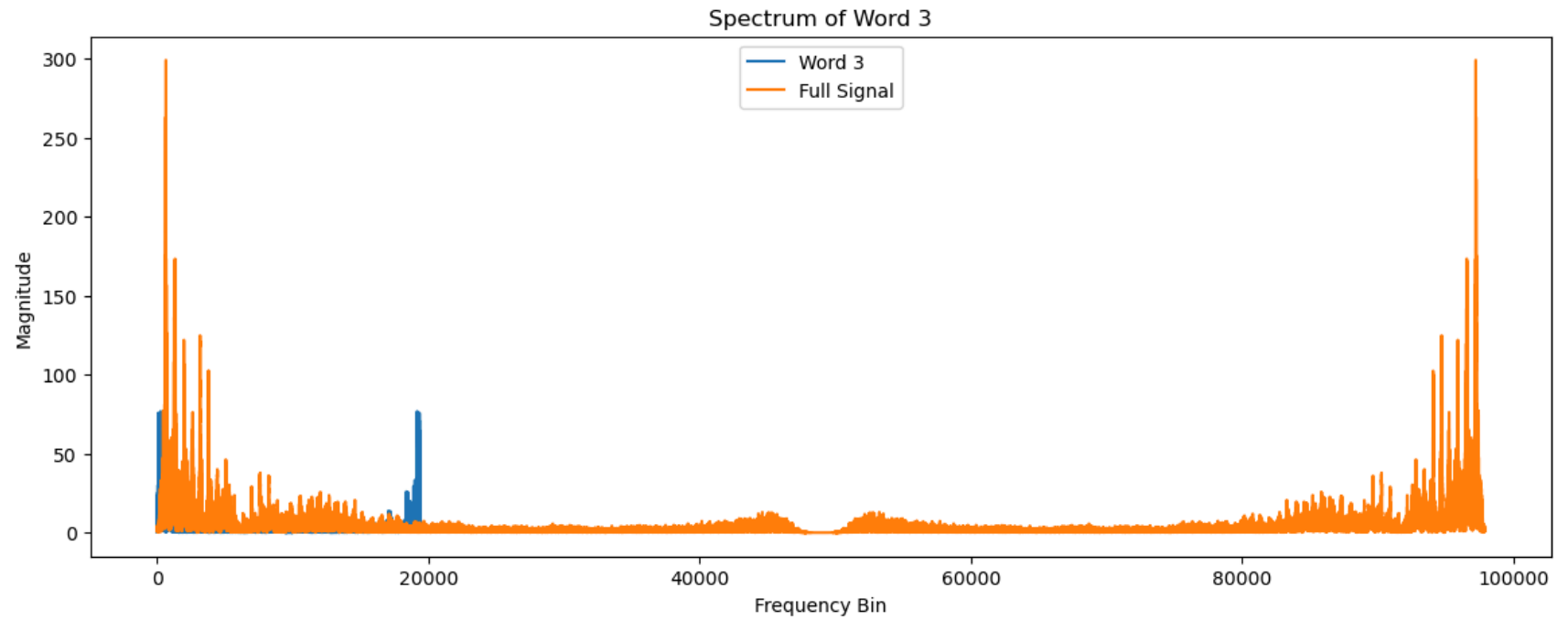


### A3. Perform the spectral analysis of a word present in the recorded speech. Compare the spectrum with the spectrum of the full signal.

```
In [48]: 1 threshold = np.percentile(np.abs(y), 92)
2 segments = librosa.effects.split(y, top_db=-15 * np.log10(threshold))
3 for i, (start, end) in enumerate(segments):
4     word = y[start:end]
5     D_full = np.fft.fft(y)
6     D_word = np.fft.fft(word)
7     plt.figure(figsize=(14, 5))
8     plt.plot(np.abs(D_word), label=f'Word {i+1}')
9     plt.plot(np.abs(D_full), label='Full Signal')
10
11     plt.title(f'Spectrum of Word {i+1}')
12     plt.xlabel('Frequency Bin')
13     plt.ylabel('Magnitude')
14     plt.legend()
15     plt.show()
```







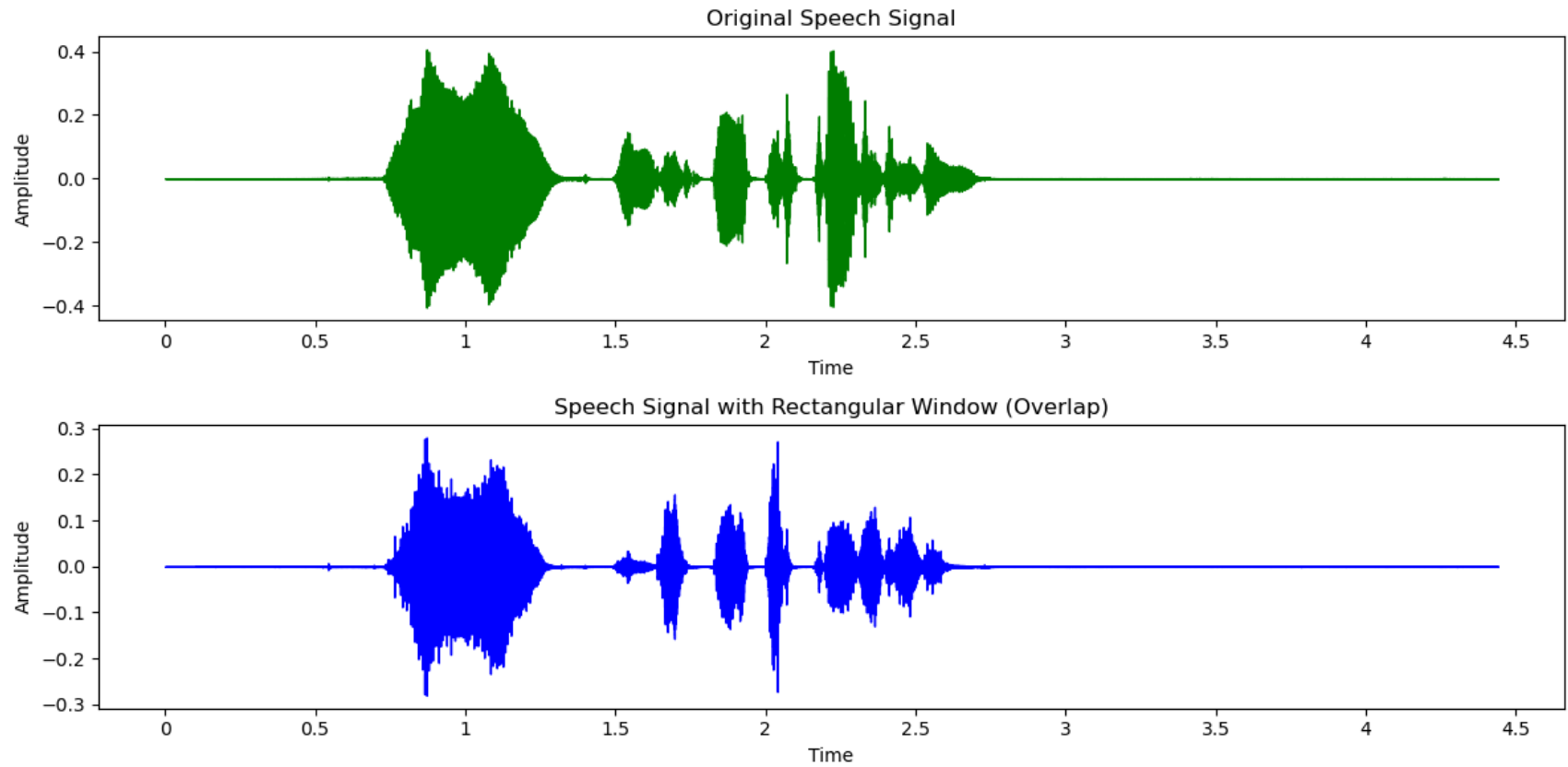
#### A4. Take a rectangular window of 20 mili-second sampled at 22.5 KHz. Using FFT, analyse the spectral components.

In [53]:

```
1 # Define the parameters for the rectangular window
2 window_size = int(0.02 * sr) # 20 milliseconds window size
3 overlap = int(0.01 * sr) # 10 milliseconds overlap
4
5 # Apply the window to the signal with overlap
6 y_windowed = librosa.effects.preemphasis(y)
7 y_frames = librosa.util.frame(y_windowed, frame_length=window_size, hop_length=ov
8
```

In [57]:

```
1  # Display the original and windowed signals
2  plt.figure(figsize=(12, 6))
3
4
5  # Plot the original signal
6  plt.subplot(2, 1, 1)
7  librosa.display.waveshow(y, sr=sr, color='green')
8  plt.title('Original Speech Signal')
9  plt.xlabel('Time')
10 plt.ylabel('Amplitude')
11
12 # Plot the windowed signal (considering overlap)
13 plt.subplot(2, 1, 2)
14 librosa.display.waveshow(y_windowed, sr=sr, color='blue')
15 plt.title('Speech Signal with Rectangular Window (Overlap)')
16 plt.xlabel('Time')
17 plt.ylabel('Amplitude')
18
19 plt.tight_layout()
20 plt.show()
21
22 print("rectangular window")
23 ipd.display(ipd.Audio(y_windowed, rate=sr))
```



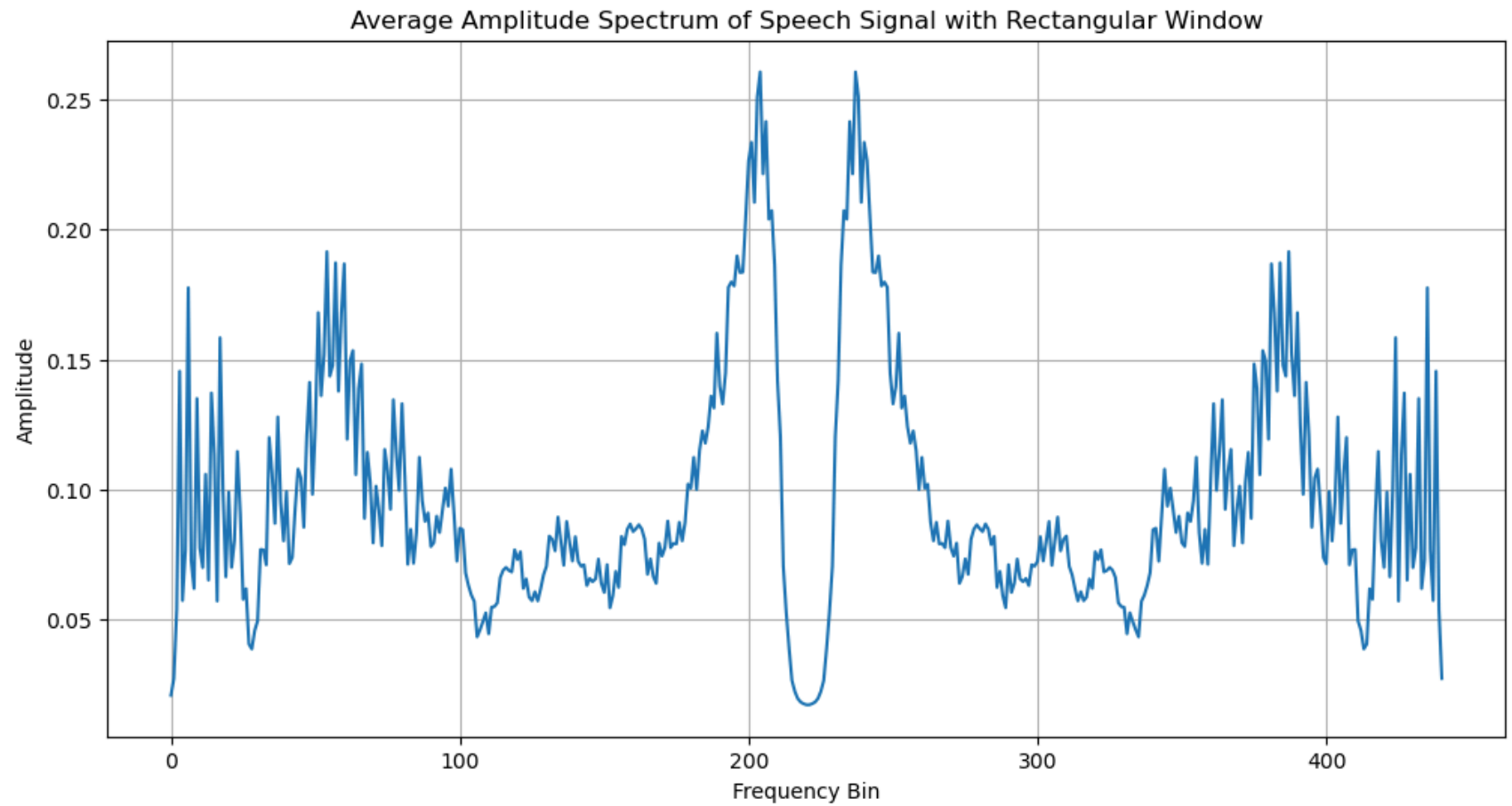
rectangular window





In [61]:

```
1
2 # Apply FFT to each windowed segment
3 fft_results_windowed = np.fft.fft(y_frames, axis=0)
4
5 # Calculate the amplitude spectrum of the windowed signal
6 amplitude_spectrum_windowed = np.abs(fft_results_windowed)
7
8 # Display the amplitude spectrum
9 plt.figure(figsize=(12, 6))
10 plt.plot(np.mean(amplitude_spectrum_windowed, axis=1)) # Plot the average spectrum
11 plt.title('Average Amplitude Spectrum of Speech Signal with Rectangular Window')
12 plt.xlabel('Frequency Bin')
13 plt.ylabel('Amplitude')
14 plt.grid(True)
15 plt.show()
```

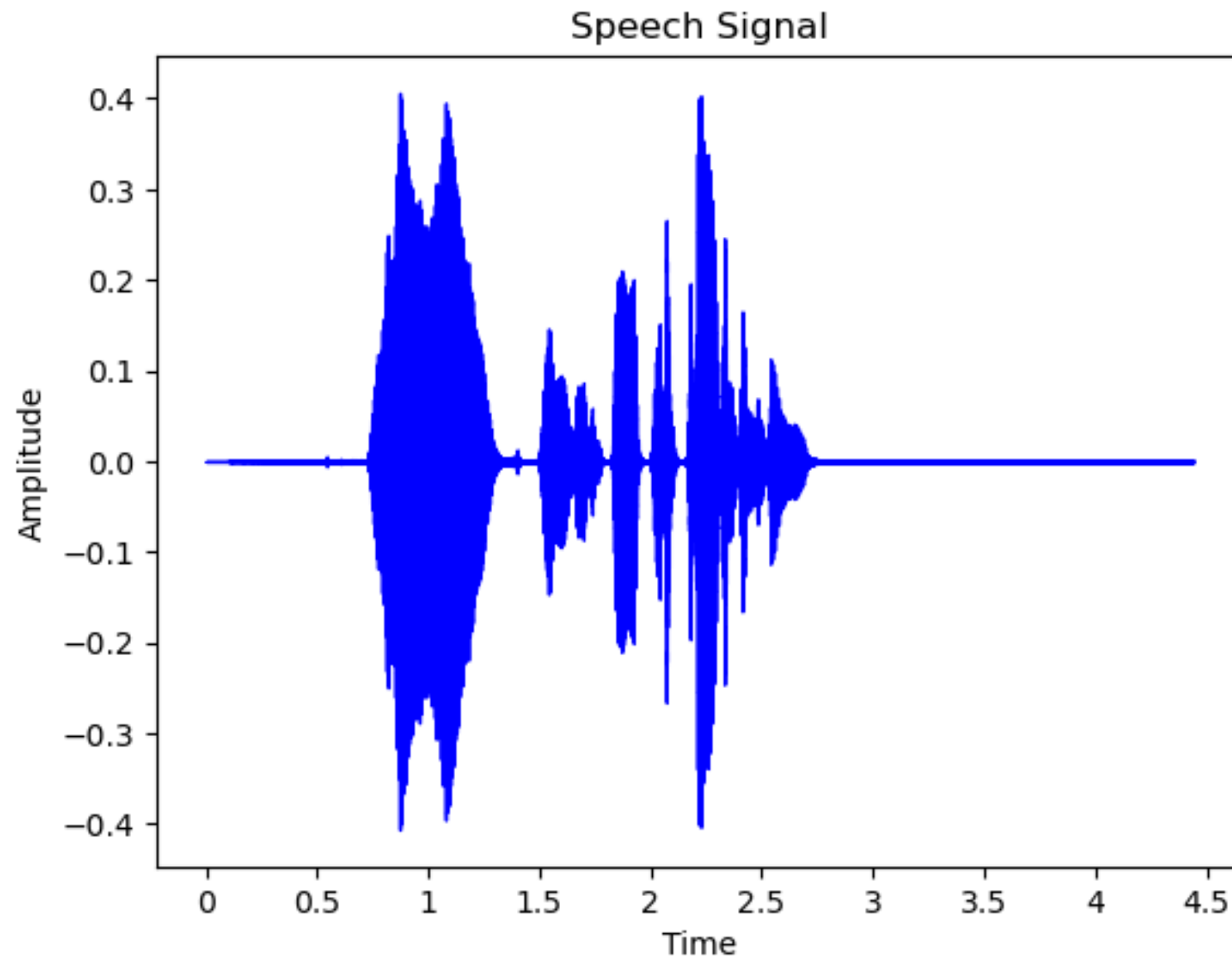


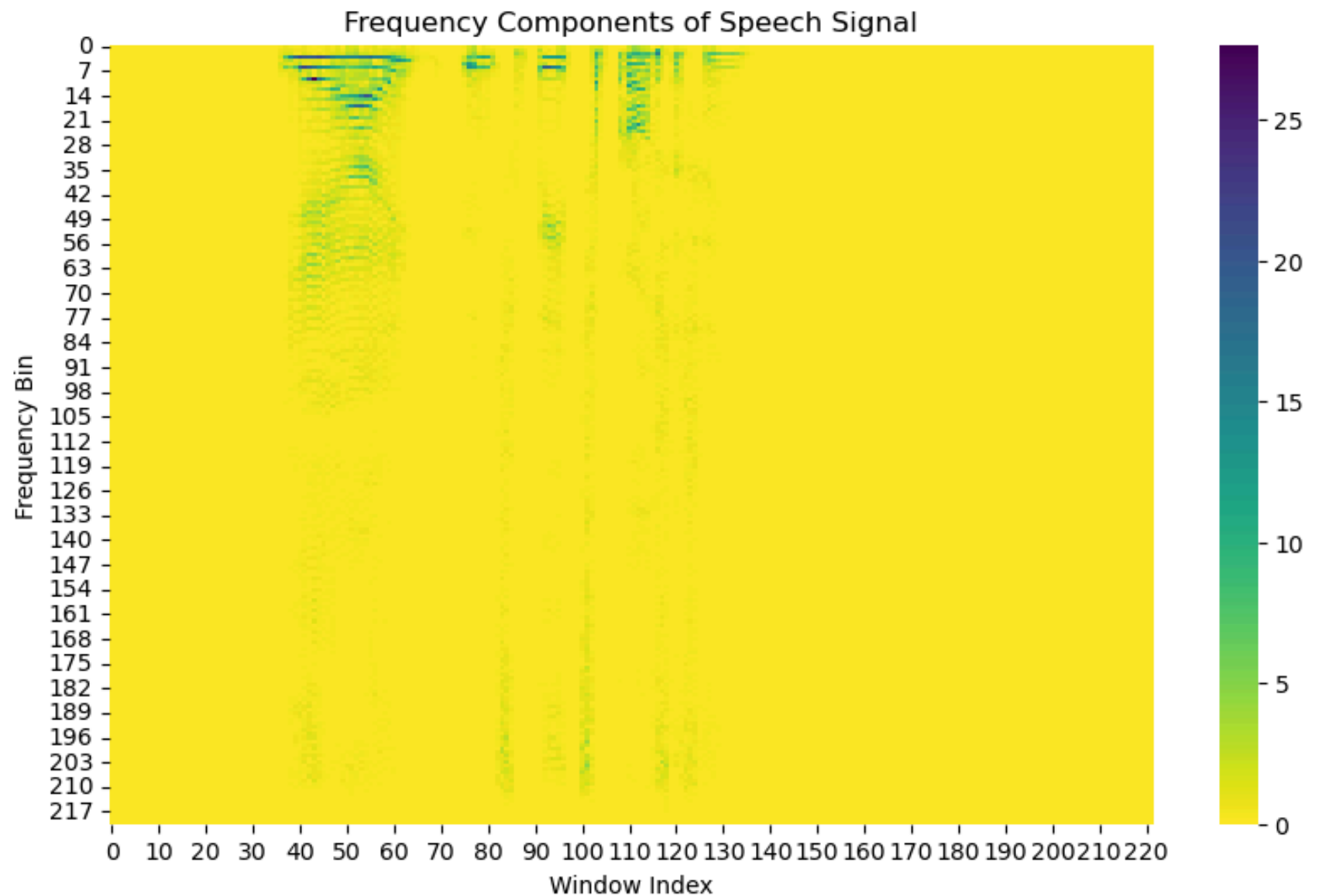
**A5. Break your speech signal into window lengths of 20 mSec intervals. Evaluate the frequency components using `numpy.fft.rfft()`. Stack these frequency components as columns in a matrix. Use heatmap plot to display the matrix. You may use `librosa.stft()` or `scipy.signal.stft()` as well to achieve this.**

```
In [66]: 1 frequencies, times, spectrogram = signal.stft(y, fs=sr, nperseg=window_size, nov
```

In [77]:

```
1 librosa.display.waveshow(y, sr=sr, color = 'blue')
2 plt.title('Speech Signal')
3 plt.xlabel('Time')
4 plt.ylabel('Amplitude')
5 plt.show()
6 window_length_sec = 0.02
7 window_length = int(window_length_sec * sr)
8 num_windows = len(y) // window_length
9 freq_matrix = np.zeros((num_windows, window_length // 2 + 1))
10 for i in range(num_windows):
11     window = y[i * window_length: (i + 1) * window_length]
12     fft_result = np.fft.rfft(window)
13     freq_matrix[i, :] = np.abs(fft_result)
14
15 # Plot the heatmap
16 plt.figure(figsize=(10, 6))
17 sns.heatmap(freq_matrix.T, cmap='viridis_r', xticklabels=10)
18 plt.title('Frequency Components of Speech Signal')
19 plt.xlabel('Window Index')
20 plt.ylabel('Frequency Bin')
21 plt.show()
```

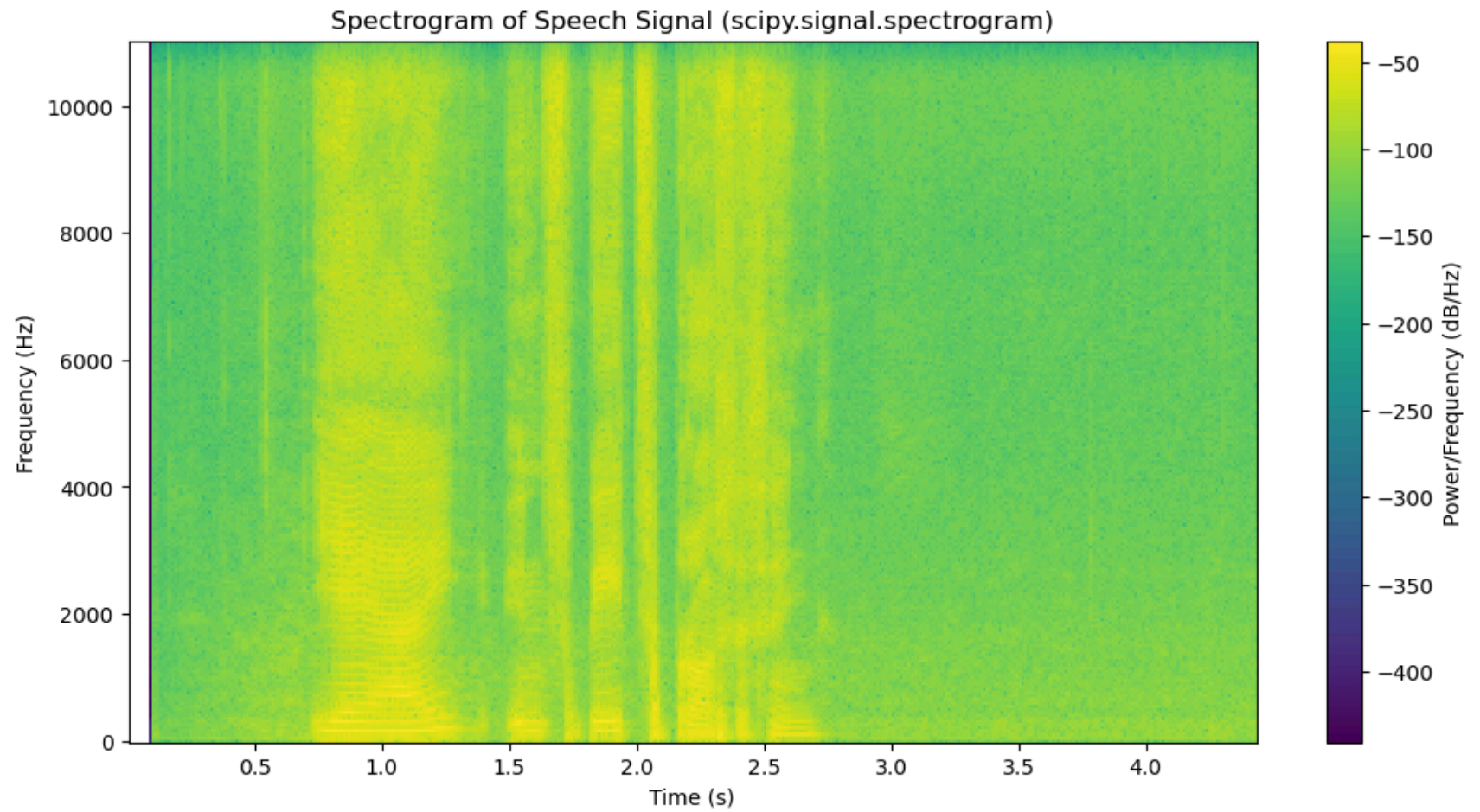




## A6. Use `scipy.signal.spectrogram()` to plot the spectrogram of the speech signal at the same duration. Compare the plots.

```
In [83]: 1 frequencies, times, Sxx = spectrogram(y, fs=sr, nperseg=window_size, noverlap=ov  
2  
3 # Display the spectrogram using matplotlib  
4 plt.figure(figsize=(12, 6))  
5 plt.pcolormesh(times, frequencies, 10 * np.log10(Sxx), shading='auto', cmap='vir  
6  
7 plt.title('Spectrogram of Speech Signal (scipy.signal.spectrogram)')  
8 plt.xlabel('Time (s)')  
9 plt.ylabel('Frequency (Hz)')  
10 plt.colorbar(label='Power/Frequency (dB/Hz)')  
11 plt.show()
```

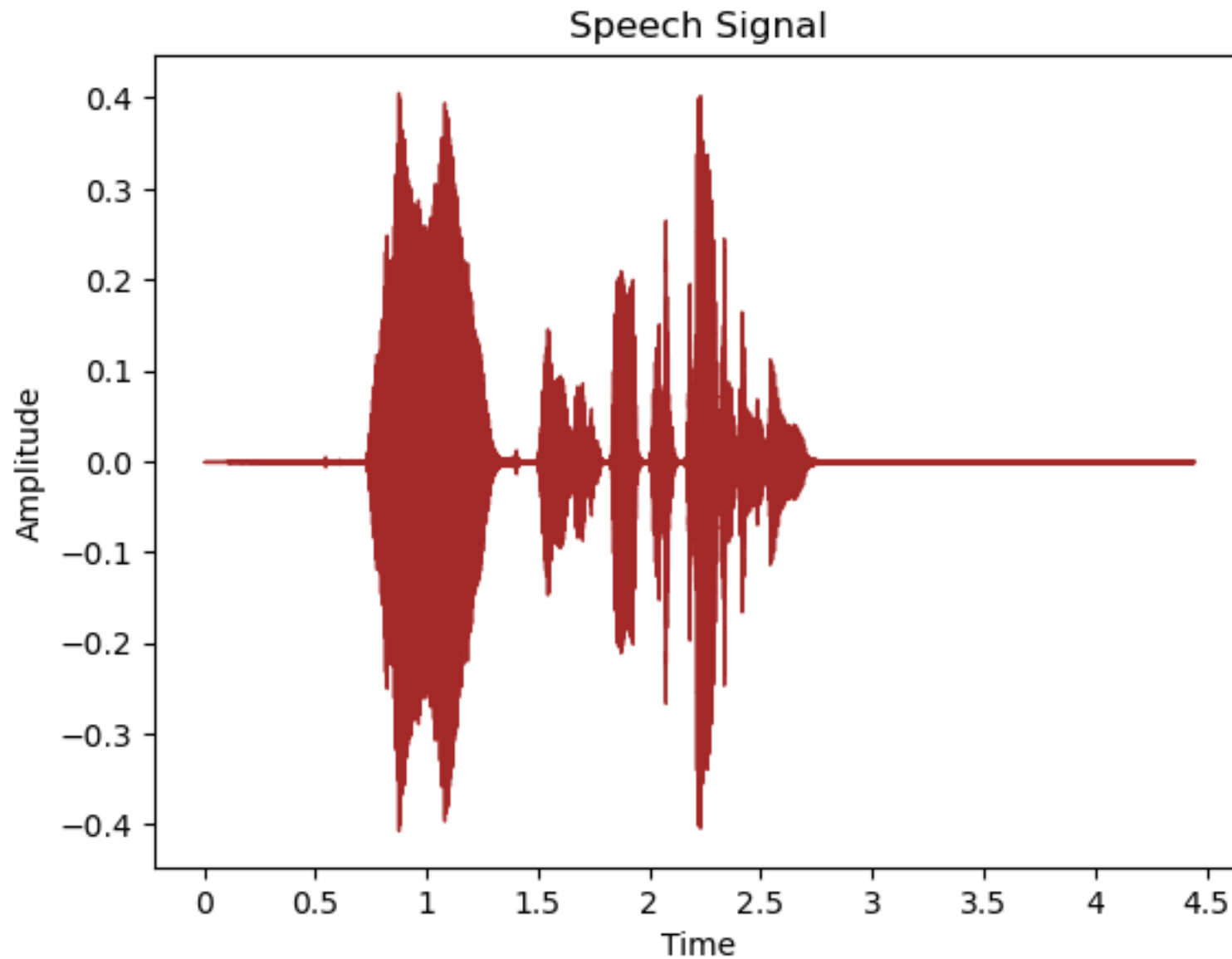
```
C:\Users\saide\AppData\Local\Temp\ipykernel_28520\3548733323.py:5: RuntimeWarning: d  
ivide by zero encountered in log10  
  plt.pcolormesh(times, frequencies, 10 * np.log10(Sxx), shading='auto', cmap='virid  
is')
```



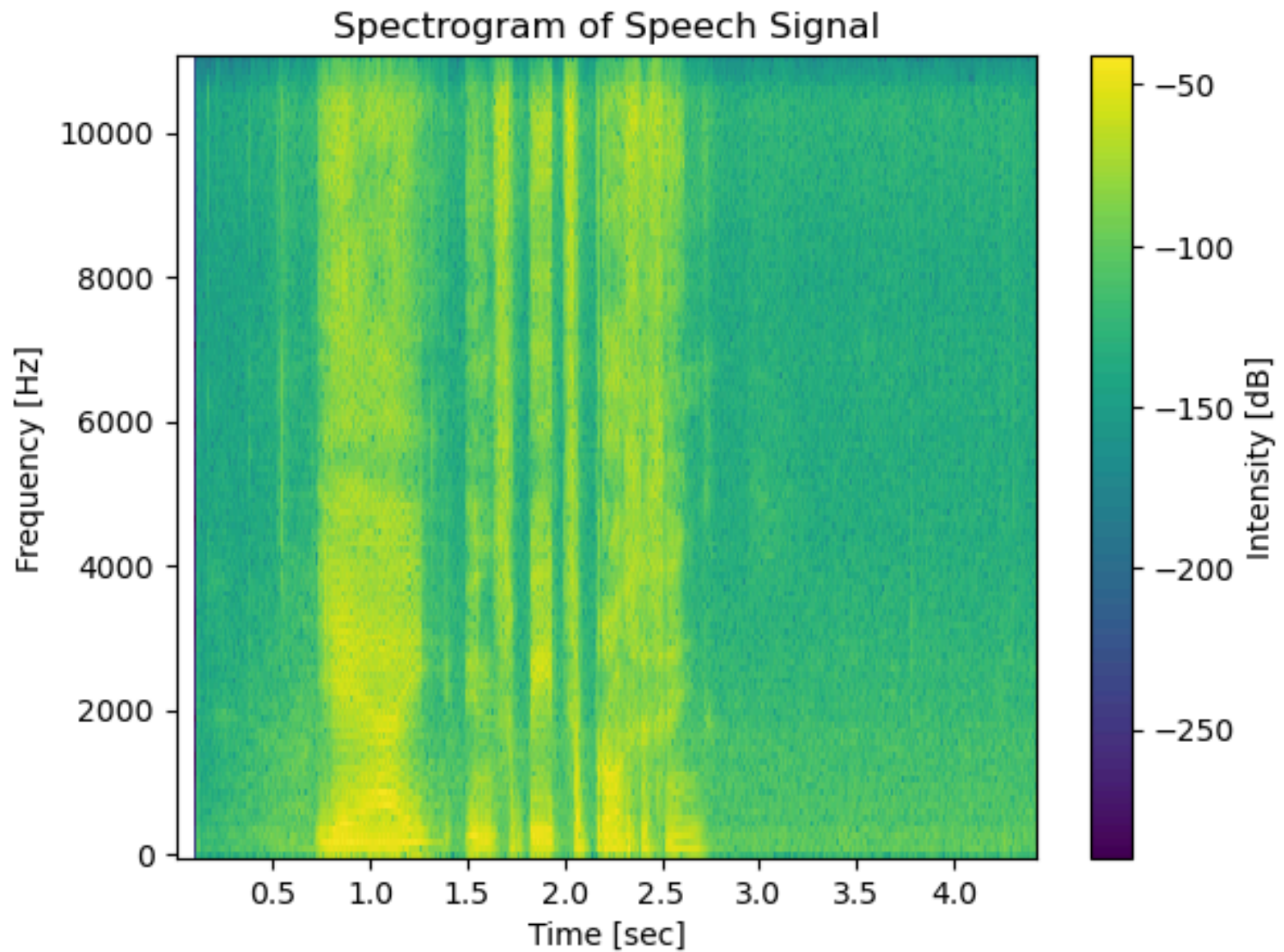


In [80]:

```
1 from scipy.signal import spectrogram
2
3 librosa.display.waveshow(y, color = 'brown')
4 plt.title('Speech Signal')
5 plt.xlabel('Time')
6 plt.ylabel('Amplitude')
7 plt.show()
8
9 f, t, Sxx = spectrogram(y, sr)
10 plt.pcolormesh(t, f, 10 * np.log10(Sxx))
11 plt.ylabel('Frequency [Hz]')
12 plt.xlabel('Time [sec]')
13 plt.title('Spectrogram of Speech Signal')
14 plt.colorbar(label='Intensity [dB]')
15 plt.show()
```



```
C:\Users\said\AppData\Local\Temp\ipykernel_28520\590411667.py:10: RuntimeWarning: d  
ivide by zero encountered in log10  
plt.pcolormesh(t, f, 10 * np.log10(Sxx))
```



In [ ]:

1