

Audio Processing

Exercise 1

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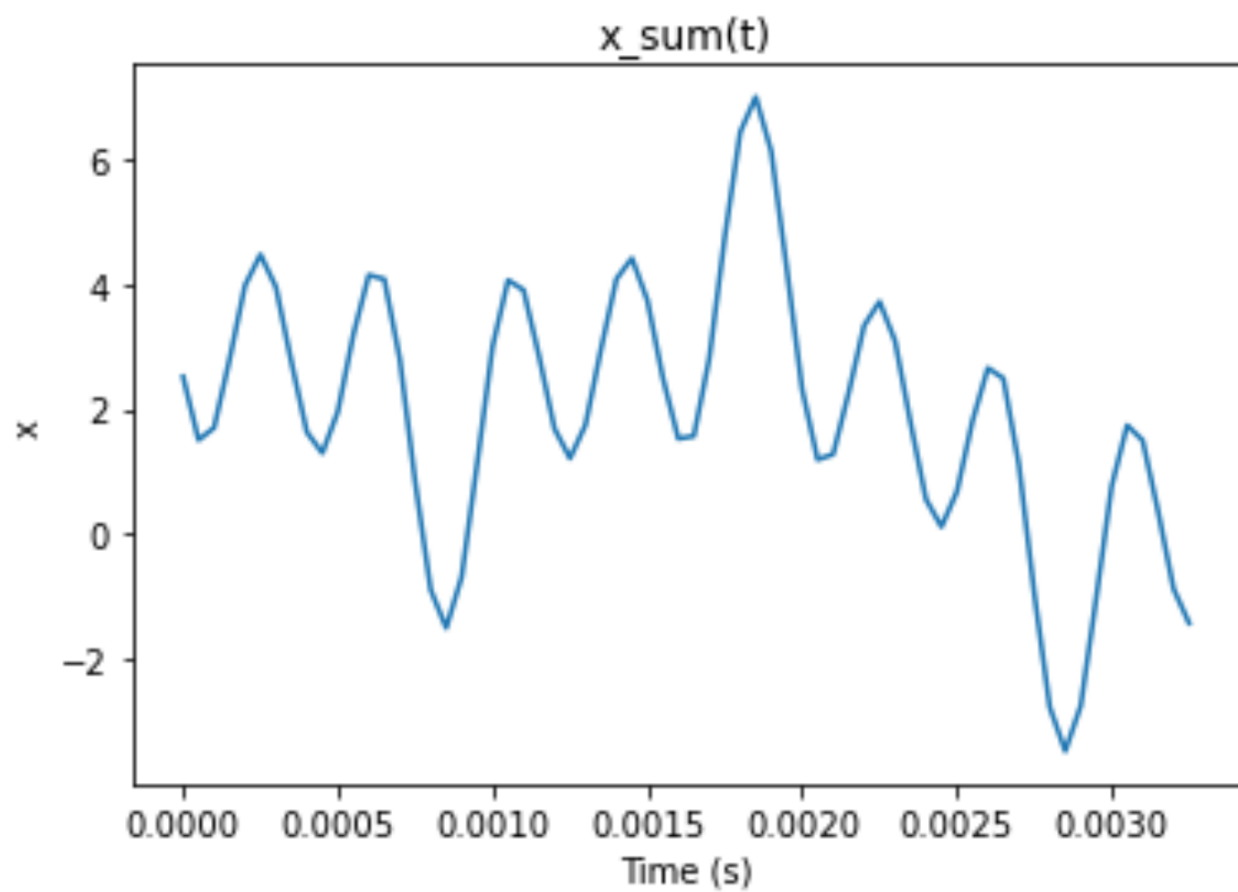
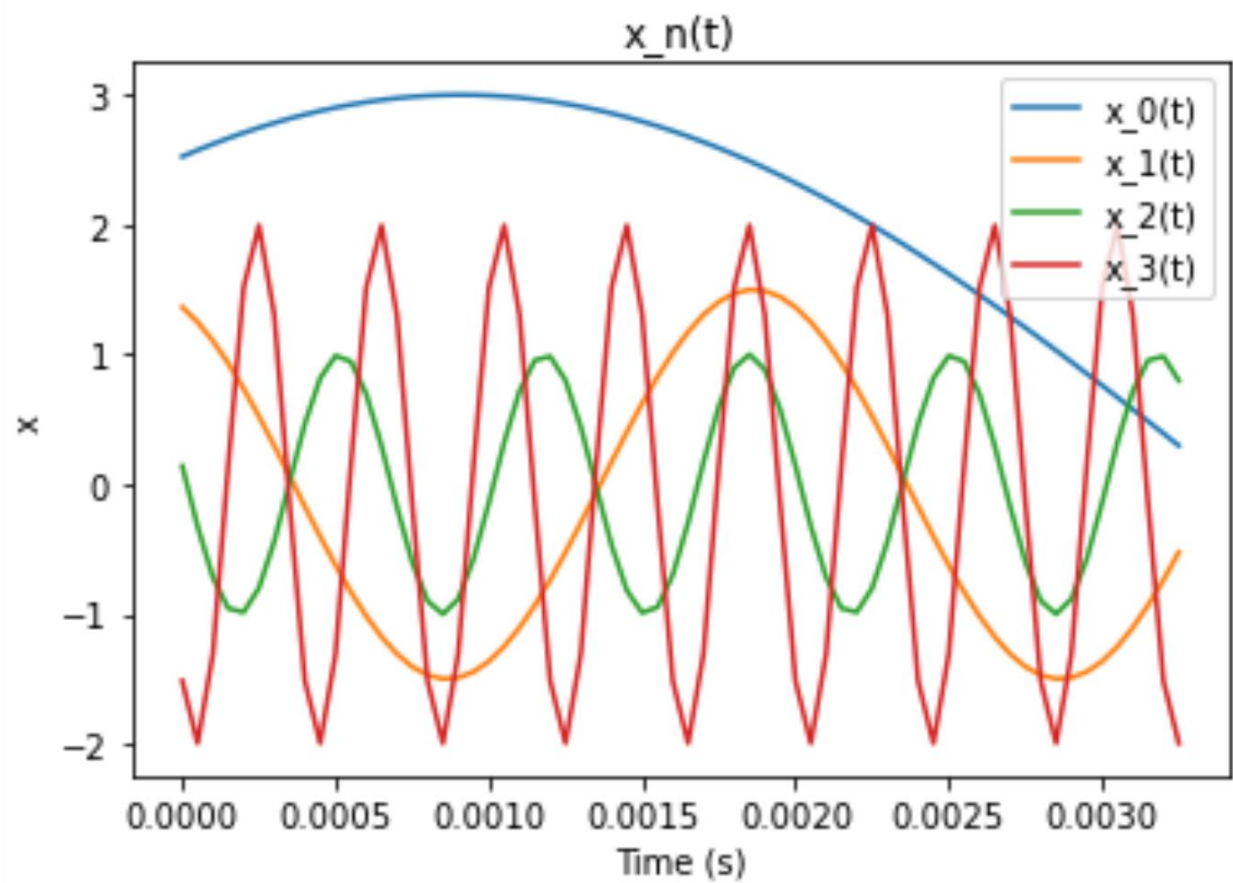
Problem 1:

```
1  #!/usr/bin/python
2  import matplotlib.pyplot as plt
3  import numpy as np
4  from scipy import signal
5  from scipy.io import wavfile
6  from scipy.fftpack import fft
7
8
9  def main():
10     # - Define sampling parameters
11     # - Highest frequency in this exercise is 2500 so
12     # Fs must be at least 5000 (Nyquist Theorem), I chose 20000
13     # to preserve waveform
14     Fs = 20000
15     SimulationTime = 3
16     NumOfSample = int(SimulationTime*Fs)
17
18     # Define 4 sine signal with given frequency
19     # and different chosen amplitudes and phases
20     t = np.linspace(0, SimulationTime, NumOfSample)
21     x = [ 3*np.sin(2 * np.pi * 100 * t +1),
22          1.5*np.sin(2 * np.pi * 500 * t +2),
23          np.sin(2 * np.pi * 1500 * t +3),
24          2*np.sin(2 * np.pi * 2500 * t +4)]
25
26     # Define plot range to have a better observation of waveform
27     plot_range = int(Fs/300)
28     plt.figure(1)
29     for i in range(0,len(x)):
30         plt.plot(t[0:plot_range],x[i][0:plot_range], label = 'x_' + str(i) + '(t)')
31     plt.title('x_n(t)')
32     plt.xlabel('Time (s)')
33     plt.ylabel('x')
34     plt.legend()
35
36
37     # Calculate sum of 4 signals
38     x_sum = 0
39     for i in range(0,len(x)):
40         x_sum += x[i]
41
42     # Plot sum signal
43     plt.figure(2)
44     plt.plot(t[0:plot_range],x_sum[0:plot_range])
45     plt.title('x_sum(t)')
46     plt.xlabel('Time (s)')
47     plt.ylabel('x')
48
49     # Write sum signal to file
50     wavfile.write('Sine.wav', Fs, x_sum)
51
52
53     # Define DFT-point
54     N_point = 512
```

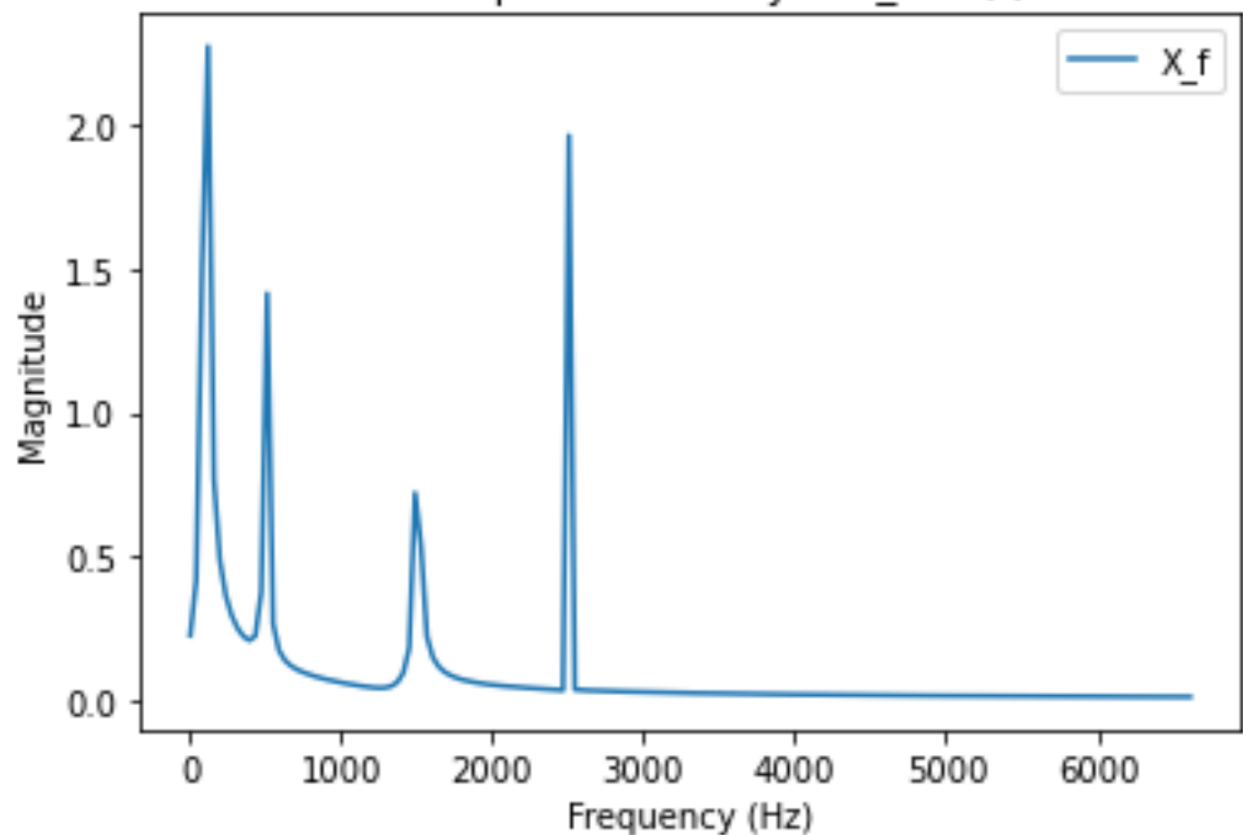
```

55
56 # Fast Fourier Transform summed signal to get Spectral Density of x_sum
57 # fftshift to plot both negative and positive part
58 ff = np.linspace(0, int(Fs), N_point)
59 xf = fft(x_sum, N_point)
60
61 # Plot DFT in suitable range for better observation
62
63 plt.figure(3)
64 plt.plot(ff[0: int(N_point/3)], 2/N_point * np.abs(xf)[0: int(N_point/3)],
65          label='X_f')
66 plt.title('DFT - Spectral density of x_sum(t)')
67 plt.xlabel('Frequency (Hz)')
68 plt.ylabel('Magnitude')
69 plt.legend()
70
71
72 #####
73 # Resample x with 2 times less samples
74 resampled_x_sum = signal.resample(x_sum, int(NumOfSample/2))
75
76 # Fast Fourier Transform resampled signal to get Spectral Density
77 # fftshift to plot both negative and positive part
78 # Because number of samples is reduced by 2 times, so is Fs
79 ff = np.linspace(0, int(Fs/2), N_point)
80 xf = fft(resampled_x_sum, N_point)
81
82 # Plot DFT in suitable range for better observation
83
84 plt.figure(4)
85 plt.plot(ff[0: int(N_point/3)], 2/N_point * np.abs(xf)[0: int(N_point/3)],
86          label='X_f')
87 plt.title('DFT - Spectral density of resampled x_sum(t)')
88 plt.xlabel('Frequency (Hz)')
89 plt.ylabel('Magnitude')
90 plt.legend()
91
92
93
94 plt.show()
95
96
97
98
99 if __name__ == "__main__":
100     main()
101

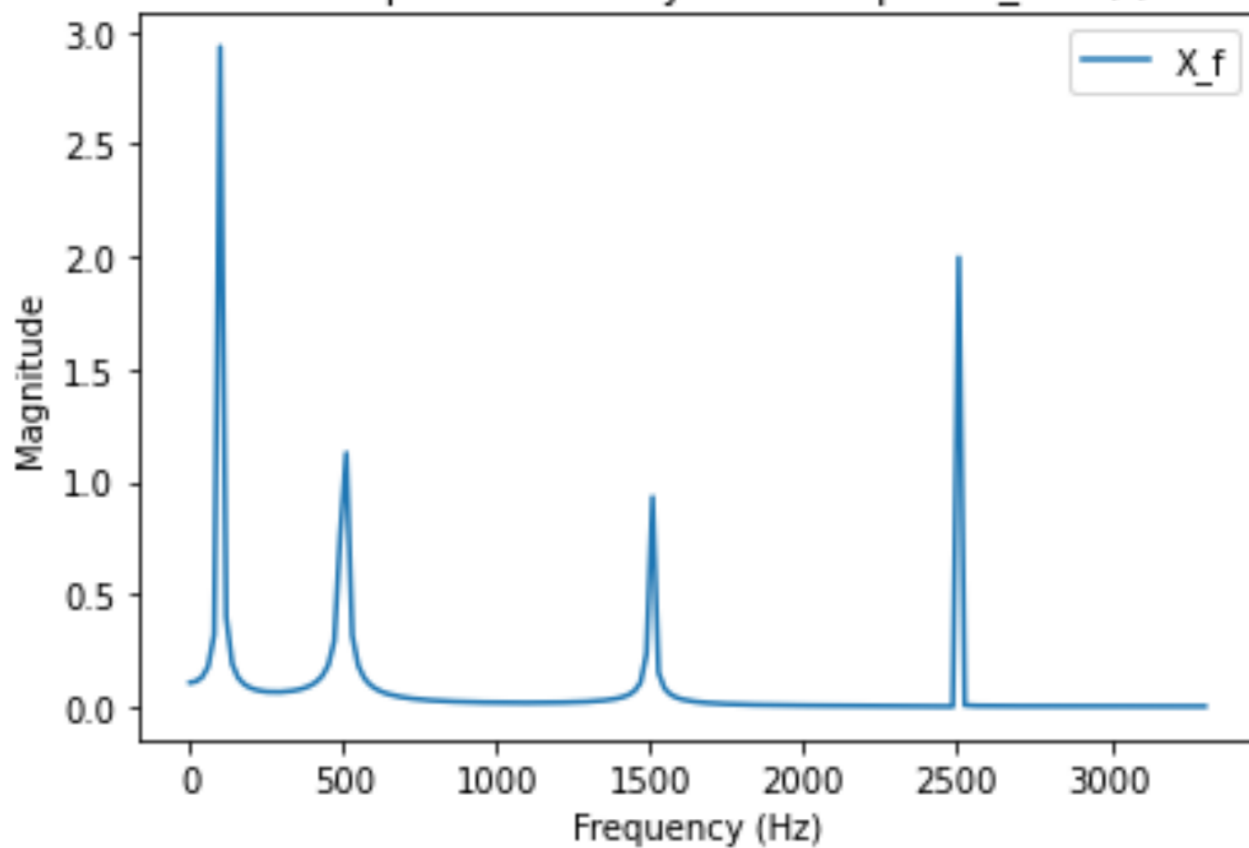
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DFT - Spectral density of $x_{\text{sum}}(t)$



DFT - Spectral density of resampled $x_{\text{sum}}(t)$



Comment:

The highest frequency in DFT plot is the sampling frequency F_s and the lowest frequency is F_s/N . So with a fixed number of N , higher F_s results in lower frequency resolution.

In resampled version, DFT plot is same as original one with sampled with $F_s' = F_s/2$. As can be seen from 2 plots, the resampled one are more precise, especially in range about 0-1000Hz. It can be because of above reason, the resampled one has lower F_s so higher frequency resolution.

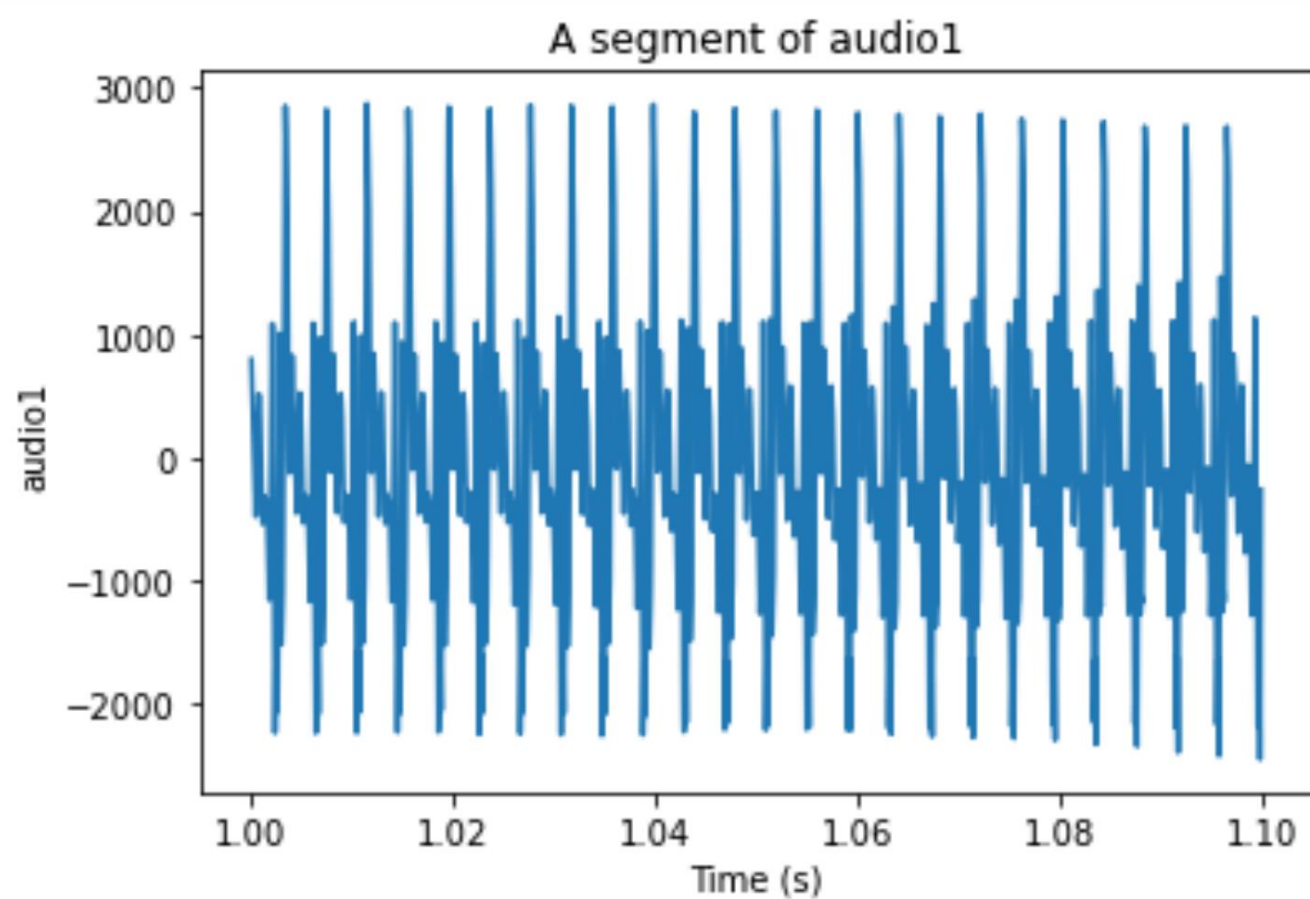
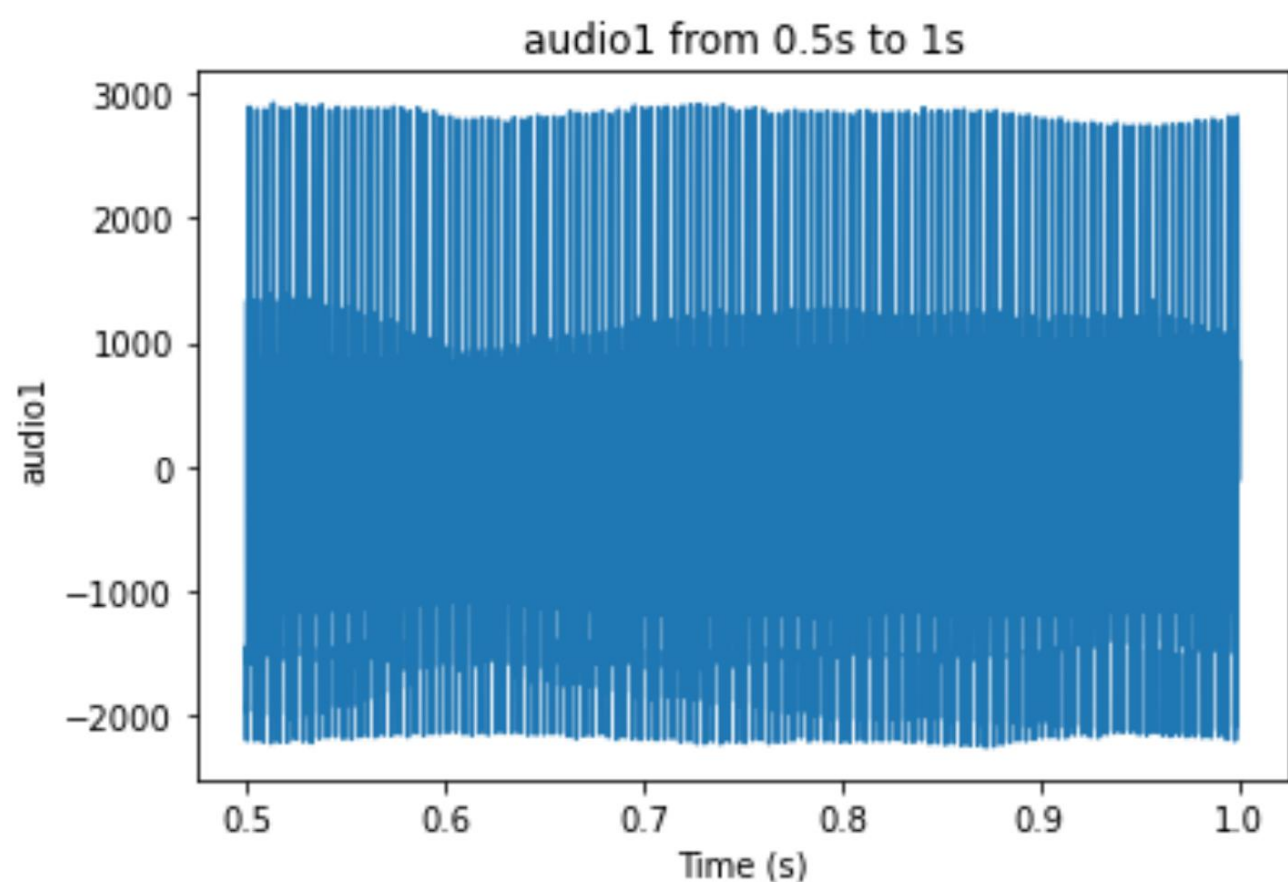
Problem 2:

```
1  #!/usr/bin/python
2  import matplotlib.pyplot as plt
3  import numpy as np
4  from scipy.io import wavfile
5  from scipy.fftpack import fft
6
7  # This number is used to count plot windows
8  figure_index = 0
9
10 # Define DFT-point
11 N_point = 512
12
13 # Define segment period (in second)
14 period = 0.1
15
16 def DFTandPlot(Fs, audio, audio_index):
17     global figure_index
18     global N_point
19     global period
20     NumOfSample = len(audio)
21     # Loop from 1s to the end, each iteration last "period (s)"
22     # NumOfSample/Fs1 return the time in seconds;
23     # subtract 1 as it is the first second.
24     # Div by "period" to transform into multiple of "period"
25     for i in range(0, int((NumOfSample/Fs - 1)/period)):
26         # Define a time step which last period start from second 1
27         # Extract a segment from audio
28         step_time = np.linspace(1+(i*period), 1+(i+1)*period, int(Fs*period))
29         step_amp = audio[(int(i*period*Fs) + Fs): (int((i+1)*period*Fs) + Fs)]
30
31         if (i < 1):
32             # Plot the segment
33             figure_index+=1
34             plt.figure(figure_index)
35             plt.plot(step_time, step_amp)
36             plt.title('A segment of audio' + audio_index)
37             plt.xlabel('Time (s)')
38             plt.ylabel('audiol')
39
40             # Calculate FFT with the segment
41             audio_f = np.linspace(0, int(Fs), N_point)
42             AUDIOinF = fft(step_amp, N_point)
43             # Plot FFT
44             figure_index+=1
45             plt.figure(figure_index)
46             plt.plot(audio_f[0: int(N_point/3)], 2/N_point * np.abs(AUDIOinF)[0: int(N_point/3)],
47                     label='Magnitude audio')
48             plt.title('DFT - Spectral density of audio' + audio_index + '(t)')
49             plt.xlabel('Frequency (Hz)')
50             plt.ylabel('Magnitude')
51             plt.legend()
```

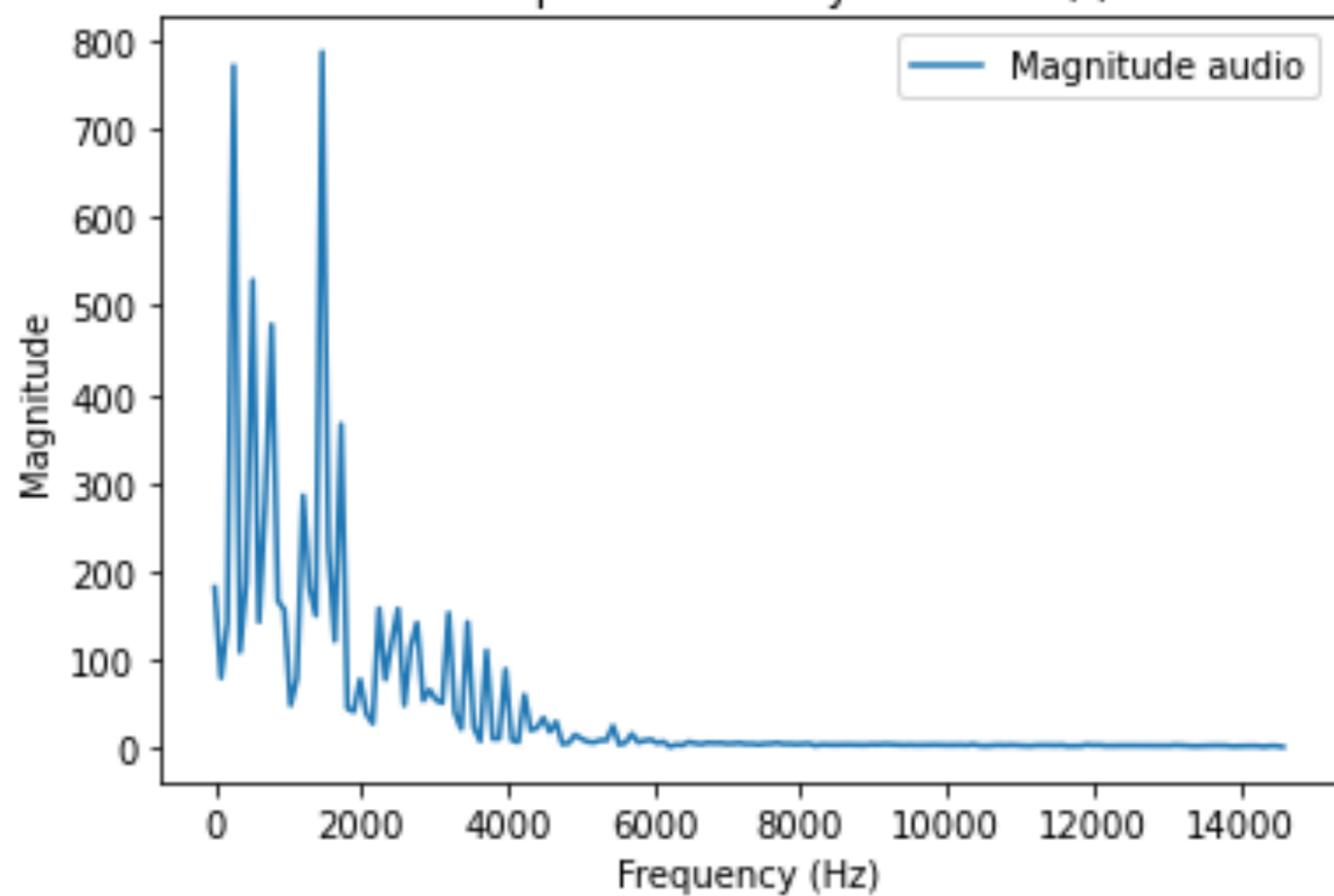
```

52
53 def main():
54     global figure_index
55
56
57     # Load audiol.wav with sample rate and signal
58     [Fs1, audiol] = wavfile.read('audiol.wav')
59
60     # Define time range from 0.5s to 1s
61     # There are "Fs" samples in 1s, so for 0.5s, number of samples is Fs/2
62     audiol_t = np.linspace(1/2, 1, int(Fs1/2))
63     audiol_x = audiol[int(Fs1/2): int(Fs1)]
64
65     # Plot the audiol in range of 0.5s to 1s
66     figure_index+=1
67     plt.figure(figure_index)
68     plt.plot(audiol_t,audiol_x)
69     plt.title('audiol from 0.5s to 1s')
70     plt.xlabel('Time (s)')
71     plt.ylabel('audiol')
72     # Calculate DFT for each segment in audiol
73     # and plot the 1st segment
74     DFTandPlot(Fs1, audiol,'1')
75
76     # Load audio2.wav with sample rate and signal
77     [Fs2, audio2] = wavfile.read('audio2.wav')
78
79     # Define time range from 0.5s to 1s
80     # There are "Fs" samples in 1s, so for 0.5s, number of samples is Fs/2
81     audio2_t = np.linspace(1/2, 1, int(Fs2/2))
82     audio2_x = audiol[int(Fs2/2): int(Fs2)]
83
84     # Plot the audio2 in range of 0.5s to 1s
85     figure_index+=1
86     plt.figure(figure_index)
87     plt.plot(audio2_t,audio2_x)
88     plt.title('audio2 from 0.5s to 1s')
89     plt.xlabel('Time (s)')
90     plt.ylabel('audio2')
91     # Calculate DFT for each segment in audio2
92     # and plot the 1st segment
93     DFTandPlot(Fs2, audio2,'2')
94
95     plt.show()
96
97
98
99 if __name__ == "__main__":
100     main()
101

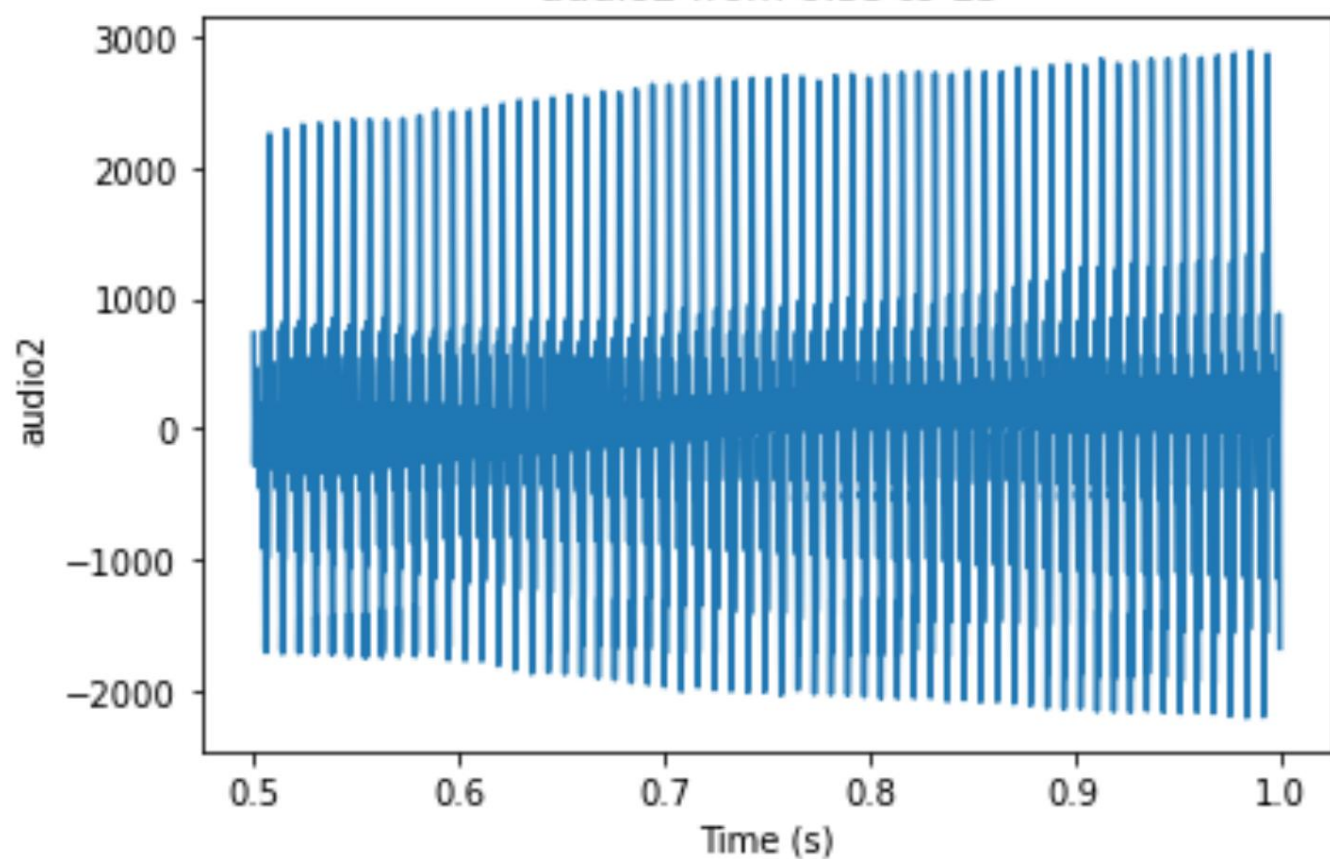
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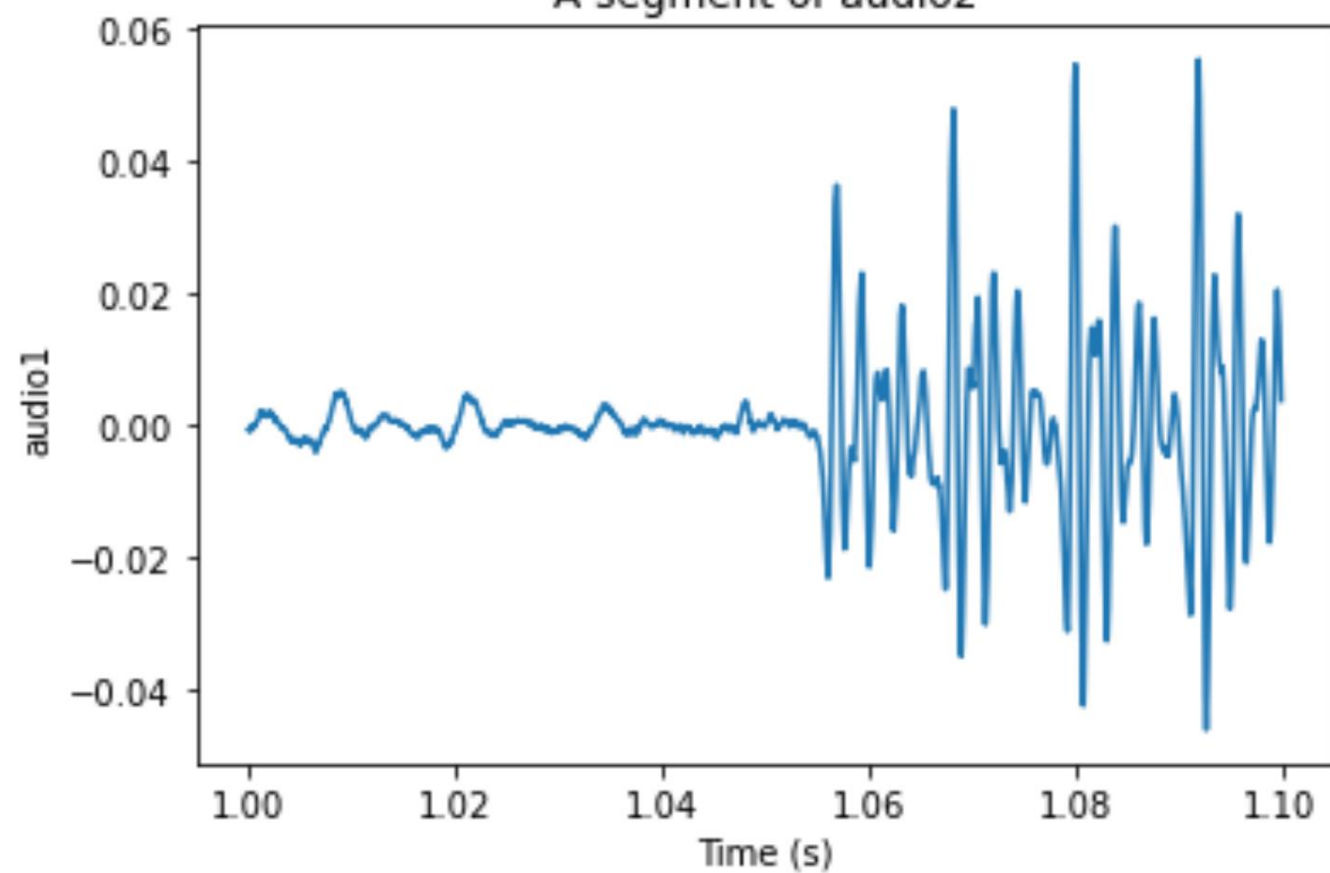
DFT - Spectral density of audio1(t)

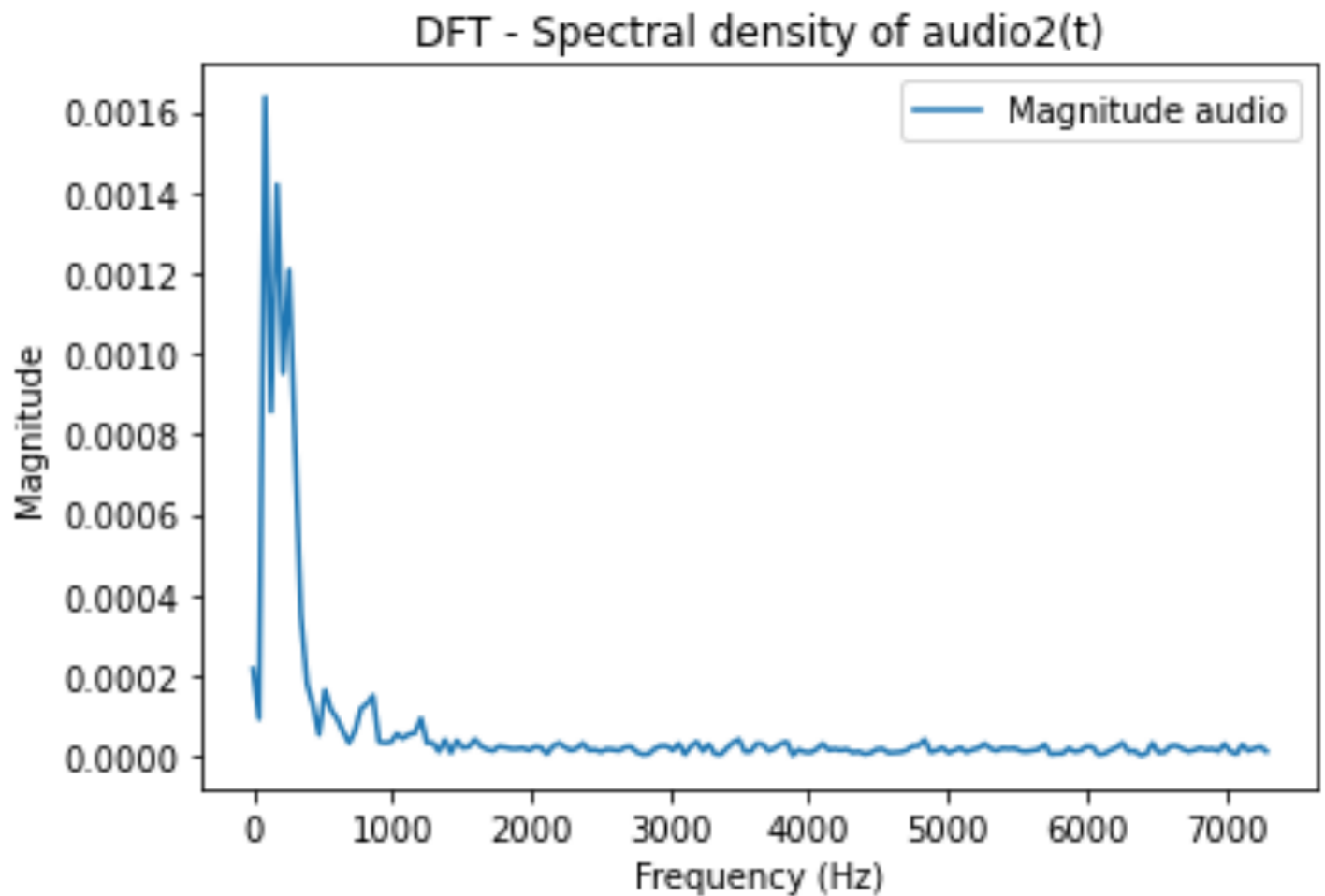


audio2 from 0.5s to 1s



A segment of audio2





Comment:

The DFT plots of both signals indicate that the 2 audios consist of multiple sinusoid waves with different frequencies within. This is described by the fluctuation of DFT plot throughout frequency range, despite the highest around 0 – 500 Hz. On the other hand, sinusoids' DFT just shows that there are only 4 signals clearly seen in the sine.wav