MAE 298 – Homework 1 Computation of Sound Pressure Level and Octave Band Spectrum

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1 Background

This assignment details the process of analyzing a recorded sound signal and computing the Sound Pressure Level (SPL) in Decibels (dB) across the narrow, 1/3 Octave, and Octave-bands. The source signal is a recording of a sonic boom contained in the included data file 'Boom_F1B2_6.way'.

All computations and plotting for this project were performed using Python, and the source code is attached in the Appendix. All of the primary data processing code is contained in the file 'hw1_00_processing.py' and the plotting code is contained in the file 'hw1_01_plotting.py', which is suplimented by the custom plotting package 'lplot.py'.

2 Problem 1 – Signal Processing

The pressure signal input file 'Boom_F1B2_6.wav' was read using the 'PySoundFile' audio library for Python, which can be found at: https://pypi.python.org/pypi/SoundFile/. This library was chosen over other Python audio libraries (e.g. 'scipy.io.wavfile') because it normalized the .wav file data between -1.0 and 1.0 identically to MAT-LAB's 'audioread' function. (Note: Though other libraries did not read in the same values, all libraries returned the same results when normalized by the maximum value of the data).

2.1 Problem 1.1 – Sonic Boom Pressure Signal

After reading the raw data from the .wav file, signal values were converted from units of Volts (V) to Pascals (Pa) using the conversion ratio obtained from Eqn 1:

$$-116Pa = 1V \tag{1}$$

The resulting pressure history of the sonic boom is displayed in Fig 1, which demonstrates the classic "N" shape

of the sonic boom, where there is an initial positive pressure discontinuity caused by the supersonic front of the aircraft, followed by an almost linear decrease in pressure over the surface of the aircraft until an abrupt discontinuity back to freestream pressure occurs. This results in the observer hearing a "double-boom" for low-flying supersonic aircraft (at higher altitudes, the two shocks can merge by the time they reach the observer). After the aircraft, the pressure does not immediately return to steady-state freestream flow, but experiences a slight oscillation in pressure as the remaining wake dissipates.

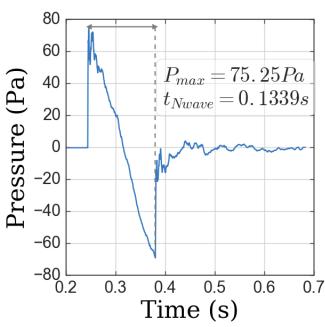


Fig. 1: Recorded sonic boom shockwave pressure time history with peak pressure and N-wave time duration values (Zero-pressure from recording start to initial shock)

Also contained in Fig 1 are the values for the peak sonic boom pressure perturbation:

$$P_{max} = \max(|P|) = \boxed{75.25Pa}$$

and sonic boom N-wave duration:

$$t_{Nwave} = \boxed{133.9ms}$$

3 Problem 1.2 – Power Spectral Density Decomposition

In order to analyze the frequency-dependent nature of the recorded sonic boom, it is nessesary to break the signial into its component frequencies, for which we employ the technique of Fast-Fourer Transform (FFT). This descrete Fourier transform algorithm transforms data from the time domain to the frequency domain, and can be called in Python from 'numpy.fft'.

The FFT is performed in Python according to the following pseudocode:

$$fft = \text{np.fft.fft}(pressure) \cdot dt$$

where *pressure* is the time-domain pressure signal and $dt = \frac{1}{f_s}$ is the time step of the discrete frequency domain, which is equal to the inverse of the sampling frequency f_s . Next, the double-sided power spectrum S_{rr} is obtained ac-

Next, the double-sided power spectrum S_{xx} is obtained according to Eqn 2:

$$S_{xx} = \frac{|fft|^2}{T} \tag{2}$$

where all operations are performed element-wise on the data series fft and $T = \text{len}(fft) \cdot dt$ is the total time interval of the data series.

From the double-sided power spectrum, the singe-sided power spectrum G_{xx} can be acquired as twice of the first half of S_{xx} (Eqn 3):

$$G_{xx} = 2S_{xx}[idx] \tag{3}$$

where all operations are performed element-wise on S_{xx} and idx is the first half of all of the indices of S_{xx} .

Now the the power spectral density G_{xx} has been computed, the corresponding single-sided frequency spectrum can be calculated with the built in numpy function 'fftfreq':

$$freqs = \text{np.fft.fftfreq}(pressure.size, dt)$$

 $freqs = freqs[idx]$

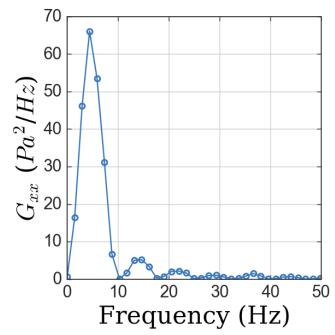


Fig. 2: Shockwave signal power spectral density as a function of frequency (All frequencies above 50Hz very low power)

The resulting power spectrum G_{xx} vs freq is plotted for reference in Fig 2. From the figure, it can be seen that the most dominate frequencies occur between 1 and 50 Hz, with the maximum power spectral density of $66Pa^2/Hz$ corresponding to 4.49Hz, and the other significant peaks (of significantly lower magnitude) occuring around 15, 23, and 37 Hz, respectively.

3.1 Problem 1.3 – Sound Pressure Level

With the power spectrum density function calculated, it is meaninful to compute the Sound Pressure Level (SPL) in dB. The formula for descrete SPL as a function of frequency is given in Eqn 4, which demonstrates that SPL is a logarithmic ratio of the signal's pressure distrubance and a reference pressure $P_{ref} = 20\mu Pa$.

$$SPL(f) = 10\log_{10}\left(\frac{G_{xx}/T}{P_{ref}^2}\right) \tag{4}$$

SPL of the narrow-band frequencies is plotted with the octave-band frequencies in Fig 3 in Section 4

4 Problem 2 – Octave-Band Spectra

In aeroacoustic analysis, it is common for recorded data sets to contain massive amounts of data that can be redundant and combersome to process. To ease this burden, originally sampled narrow-band frequency spectra can be binned into fewer, pre-selected, descrete sets of frequencies called the Third (1/3) Octave and Octave Bands. This binning process is accomplished by summing the narrow-band frequencies over intervals defined by specific center frequencies f_c of the octave-band. Center frequencies can be chosen from a pre-selected "preferred" set of "nicer" numbers, or they can be calculated according to Eqn 5, which does not produce the "nice" numbers.

$$f_{c,m} = f_{c,30} \cdot 2^{-10 + \frac{m}{3}} \tag{5}$$

where $f_{c,30} = 1000Hz$ is the center frequency corresponding to m = 30, and m = 1,2,3,... for 1/3 octave-band and m = 3,6,9,... for octave-band.

In the data processing script for this analysis, center frequencies are calculated specifically for the narrow-band input. The function 'OctaveCenterFreqs' will loop through m in intervals appropriate to the given octave-band choice, but only center frequencies whose lower f_l and upper f_u band limits are contained within the original data set. This ensures that no sum will take place over an incomplete band. Upper and lower band limits for a given center frequencies are calculated according to Eqn 6.

$$f_u = 2^{\frac{octv}{2}} \cdot f_c$$

$$f_l = 2^{-\frac{octv}{2}} \cdot f_c$$
(6)

where $octv = \frac{1}{3}$ for the 1/3 octave-band and octv = 1 for the octave-band.

results now, discuss/derive below

4.1 Problem 2.1 – 1/3 Octave Bands specifics on 1/3

4.2 Problem 2.2 – Octave Bands specifics on octave

4.3 Problem 2.3 – Overall Sound Pressure Level overall

5 Conclusion conclude

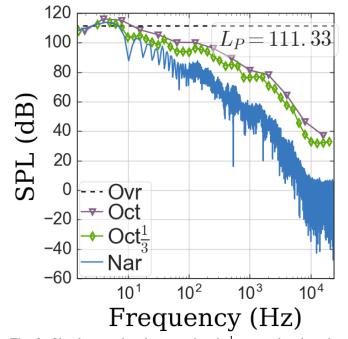


Fig. 3: Shockwave signal narrow-band, $\frac{1}{3}$ octave-band, and octave-band, with overall Sound Pressure Level reported in upper right

Appendix A: Data Processing Script

```
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MAE 298 AEROACOUSTICS
3 HOMEWORK 1 - SIGNAL PROCESSING
 CREATED: 04 OCT 2016
5 MODIFIY: 17 OCT 2016
7 DESCRIPTION: Read sound file of sonic boom and convert signal to
 Narrow-band in Pa.
 Compute Single-side power spectral density (FFT).
10 1/3 octave and octave band
2 NOTE: use 'soundfile' module to read audio data. This normalizes data from
  .....
14
15
16 #IMPORT GLOBAL VARIABLES
from hw1_98_globalVars import *
18
import numpy as np
20 import pandas as pd
21
def ReadWavNorm(filename):
      identically to MATLAB's 'audioread' function
26
27
28
     data, samplerate = sf.read(filename)
      return samplerate, data
def ReadWav(filename):
      """NOTE: NOT USED IN THIS CODE, DOES NOT NORMALIZE LIKE MATLAB
33
34
35
37
      from scipy.io import wavfile
38
      sampFreq, snd = wavfile.read(filename)
39
40
      return sampFreq, snd
41
42
  def Normalize(data):
      """NOTE: NOT USED IN THIS CODE, TRIED BUT FAILED TO NORMALIZE LIKE MATLAB
45
46
47
      return ( 2*(data - min(data)) / (max(data) - min(data)) - 1)
  def SPLt(P, Pref=20e-6):
51
           --> pressure signal (Pa)
52
54
     PrmsSq = 0.5 * P ** 2 #RMS pressure squared
55
      return 10 * np.log10(PrmsSq / Pref ** 2)
58 def SPLf(Gxx, T, Pref=20e-6):
```

```
--> Total time interval of pressure signal
62
63
      return 10 * np.log10( (Gxx / T) / Pref ** 2)
64
65
  def OctaveCenterFreqsGen(dx=3, n=39):
67
68
71
      fc30 = 1000 #Preferred center freq for m=30 is 1000Hz
72
      m = np.arange(1, n+1) * dx #for n center freqs, multiply 1-->n by dx
      freqs = fc30 * 2 ** (-10 + m/3) #Formula for center freqs
74
75
  def OctaveBounds(fc, octv=1):
      """Get upper/lower frequency bounds for given octave band.
77
      fc --> current center frequency
78
79
80
      upper = 2 ** ( octv / 2) * fc
81
82
      return upper, lower
85 def OctaveCenterFreqs(narrow, octv=1):
87
88
      upper band limit are within the original data set.
89
      narrow --> original narrow-band frequencies (provides bounds for octave)
90
91
92
93
      freqs = []
94
95
      for i in range(len(narrow)):
           fc = fc30 * 2 ** (-10 + m/3) #Formula for center freq
99
100
101
102
               freqs.append(fc) #if current fc is in original range, save
103
      return freqs
104
  def OctaveLp(Lp):
106
107
      perform the appropriate log-sum to determine the octave SPL
108
      Lp\_octv = 10 * np.log10 ( np.sum ( 10 ** (Lp / 10) ) )
      return Lp_octv
114
  def GetOctaveBand(df, octv=1):
115
      """Get SPL ( Lp(fc,m) ) for octave-band center frequencies.
116
118
      octv --> octave-band type (octave-->1, 1/3 octave-->1/3)
120
121
```

```
OctaveCenterFreqs(df['freq'], octv)
       Lp_octv = np.zeros(len(fcs))
124
126
           fcu, fcl = OctaveBounds(fc, octv)
128
           band = df[df['freq'] >= fcl]
129
130
131
           Lp = np.array(band['SPL'])
           Lp_octv[i] = OctaveLp(Lp)
136
       return fcs, Lp_octv
138
139
140
  def main(source):
142
143
144
147
148
149
150
       df = pd.DataFrame() #Stores signal data
151
153
       fs, df['V'] = ReadWavNorm( '{}/{}'.format(datadir, source) ) #Like matlab
156
157
       df['Pa'] = df['V'] * volt2pasc
158
159
160
161
162
163
165
166
167
       df['time'] = np.arange(N) * dt #individual sample times
168
       idx = range(int(N/2)) #Indices of single-sided power spectrum (first half)
169
170
       fft = np.fft.fft(df['Pa']) * dt #Fast-Fourier Transform
       Sxx = np.abs(fft) ** 2 / T #Two-sided power spectrum
174
176
177
       freqs = np.fft.fftfreq(df['Pa'].size, dt) #Frequencies
178
179
       freqs = freqs[idx] #single-sided frequencies
180
181
182
       powspec = pd.DataFrame({'freq': freqs, 'Gxx': Gxx})
183
```

```
powspec[powspec['Gxx'] == max(powspec['Gxx'])]
185
186
187
188
189
190
191
192
193
       df['SPL'] = SPLt(df['Pa'])
195
       powspec['SPL'] = SPLf(Gxx, T)
196
198
199
200
201
202
       Pmax = max(abs(df['Pa']))
203
204
205
       shocki = df[df['Pa'] == max(df['Pa'])] #Shock start
       ti = float(shocki['time']) #start time
208
       Pi = float(shocki['Pa']) #start (max) pressure
209
       shockf = df[df['Pa'] == min(df['Pa'])] #Shock end
210
213
       dt_Nwave = tf - ti
214
216
219
220
       octv3rd = pd.DataFrame()
       octv3rd['freq'], octv3rd['SPL'] = GetOctaveBand(powspec, octv=1/3)
224
       octv = pd.DataFrame()
225
       octv['freq'], octv['SPL'] = GetOctaveBand(powspec, octv=1)
226
228
229
230
       Lp_overall = OctaveLp(octv[octv['freq'] >= 10.0]['SPL'])
232
235
236
238
239
       df.to_csv( '{}/timespec.dat'.format(datadir), sep=' ', index=False ) #save
240
241
242
243
       powspec.to_csv( '{}/freqspec.dat'.format(datadir), sep=' ', index=False )
244
245
       octv3rd.to_csv( '{}/octv3rd.dat'.format(datadir), sep=' ', index=False)
```

```
octv.to_csv( '{}/octv.dat'.format(datadir), sep=' ', index=False)
247
248
249
       params = pd.DataFrame()
params = params.append(pd.Series())
250
251
252
             'Pmax' : Pmax, 'tNwave' : dt_Nwave,
253
254
255
            ), ignore_index=True)
       params.to_csv( '{}/params.dat'.format(datadir), sep=' ', index=False)
256
257
259 if __name__ == "__main__":
260
261
       Source = 'Boom_F1B2_6.wav'
262
263
       main(Source)
264
```

Listing 1: $hw1_00_process.py$ - Performs all primary data processing such as pressure signal input, power spectral density decomposition, and octave-band conversion and saves data to text files