

## Question 5

February 11, 2022

```
[1]: # Importing necessary packages
import numpy as np
from scipy.io import wavfile
import math
import matplotlib.pyplot as plt
```

```
[2]: # Function to plot spectrogram
def plot_spectrogram(frequency, data):
    plt.title('Spectrogram')
    plt.specgram(data, Fs=frequency)
    plt.xlabel('Time')
    plt.ylabel('Frequency')
    plt.show()
```

### 1 Functions to compute average of non-diagonal elements

```
[3]: def compute_non_diag_avg(temp_arr, reduced_log_data):
    #Finding mean of clean data
    temp_mean = np.mean(reduced_log_data, axis = 1)
    temp_mean = np.reshape(temp_mean, (128,1))

    x_double_prime = np.matmul(temp_arr, (reduced_log_data-temp_mean))
    x_double_prime_mean = np.mean(x_double_prime, axis = 1)
    x_double_prime_mean = np.reshape(x_double_prime_mean, (128,1))

    Sx_after_whitening = np.matmul((x_double_prime-x_double_prime_mean), np.
    ↪transpose(x_double_prime-x_double_prime_mean))
    Sx_after_whitening = np.multiply(Sx_after_whitening, 1/298)

    Sx_after_whitening_abs = np.absolute(Sx_after_whitening)

    sum_non_diag = 0
    for i in range(128):
        for j in range(128):
            if(i==j):
                continue
            sum_non_diag += Sx_after_whitening_abs[i][j]
```

```

avg_non_diag = sum_non_diag/(128*128-128)

return avg_non_diag

```

```

[4]: def compute_non_diag_avg_B(temp_arr_B, reduced_log_data_B):
    #Finding mean of clean data
    temp_mean = np.mean(reduced_log_data_B, axis = 1)
    temp_mean = np.reshape(temp_mean, (128,1))

    x_double_prime_B = np.matmul(temp_arr_B, (reduced_log_data_B-temp_mean))
    x_double_prime_mean_B = np.mean(x_double_prime_B, axis = 1)
    x_double_prime_mean_B = np.reshape(x_double_prime_mean_B, (128,1))

    Sx_after_whitening_B = np.
    ↪matmul((x_double_prime_B-x_double_prime_mean_B),np.
    ↪transpose(x_double_prime_B-x_double_prime_mean_B))
    Sx_after_whitening_B = np.multiply(Sx_after_whitening_B, 1/298)

    Sx_after_whitening_abs_B = np.absolute(Sx_after_whitening_B)

    sum_non_diag_B = 0
    for i in range(128):
        for j in range(128):
            if(i==j):
                continue
            sum_non_diag_B += Sx_after_whitening_abs_B[i][j]

    avg_non_diag_B = sum_non_diag_B/(128*128-128)

    return avg_non_diag_B

```

## 2 Reading Files and processing

```

[5]: # Reading files
samplerate_clean, data_clean = wavfile.read(r"C:
    ↪\Users\ujjaw\Desktop\MLSP_Assignments\Ass1\5. Speech_
    ↪Spectrogram\speechFiles\clean.wav")
samplerate_noise, data_noise = wavfile.read(r"C:
    ↪\Users\ujjaw\Desktop\MLSP_Assignments\Ass1\5. Speech_
    ↪Spectrogram\speechFiles\noise.wav")
samplerate_noisy, data_noisy = wavfile.read(r"C:
    ↪\Users\ujjaw\Desktop\MLSP_Assignments\Ass1\5. Speech_
    ↪Spectrogram\speechFiles\noisy.wav")

```

```
[6]: #Finding Feature Vector of Clean, Noise and Noisy wav file for 298 frames
```

```
hamming_window_size = int(25*16000/1000)
shift_length = int(10*16000/1000)
feature_vector_clean = []
feature_vector_noise = []
feature_vector_noisy = []
for i in range(298):
    start_point = i*shift_length
    end_point = i*shift_length + hamming_window_size
    feature_vector_clean.append(data_clean[start_point:end_point])
    feature_vector_noise.append(data_noise[start_point:end_point])
    feature_vector_noisy.append(data_noisy[start_point:end_point])

feature_vector_clean = np.transpose(np.array(feature_vector_clean))
feature_vector_noise = np.transpose(np.array(feature_vector_noise))
feature_vector_noisy = np.transpose(np.array(feature_vector_noisy))
```

```
[7]: # Computing Fast Fourier Transformation for the feature vectors
```

```
fft_clean = np.fft.fft(feature_vector_clean, n = 256, axis = 0)
fft_noise = np.fft.fft(feature_vector_noise, n = 256, axis = 0)
fft_noisy = np.fft.fft(feature_vector_noisy, n = 256, axis = 0)
```

```
[8]: #Taking the first 128 dimension as the other half is symmetric
```

```
reduced_clean = fft_clean[:128]
reduced_noise = fft_noise[:128]
reduced_noisy = fft_noisy[:128]
```

```
[9]: # Extracting the magnitude of the first half of the frequency
```

```
reduced_abs_clean = np.absolute(reduced_clean)
reduced_abs_noise = np.absolute(reduced_noise)
reduced_abs_noisy = np.absolute(reduced_noisy)
```

```
# Taking log of magnitude of the first half of the frequency
```

```
reduced_log_clean = np.log(reduced_abs_clean)
reduced_log_noise = np.log(reduced_abs_noise)
reduced_log_noisy = np.log(reduced_abs_noisy)
```

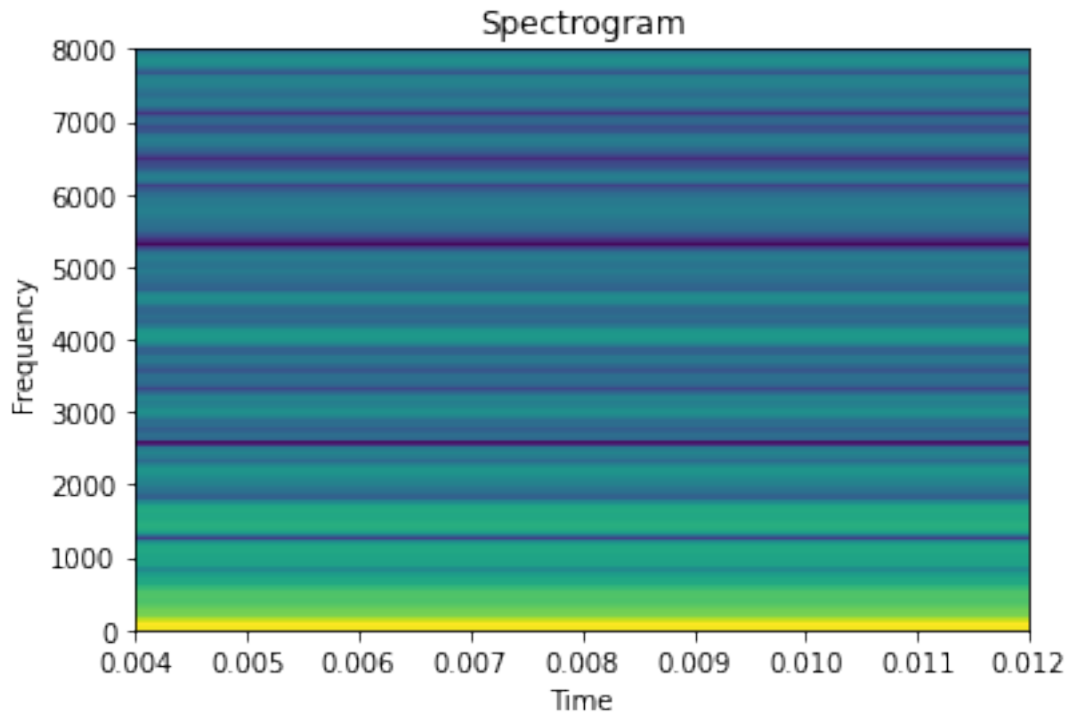
```
[10]: print('Clean File SPECTOGRAM')
```

```
plot_spectrogram(samplerate_clean,reduced_log_clean)
```

Clean File SPECTOGRAM

C:\Users\ujjaw\AppData\Local\Temp\ipykernel\_11724\593759964.py:4: UserWarning:  
Only one segment is calculated since parameter NFFT (=256) >= signal length  
(=128).

```
plt.specgram(data,Fs=frequency)
```

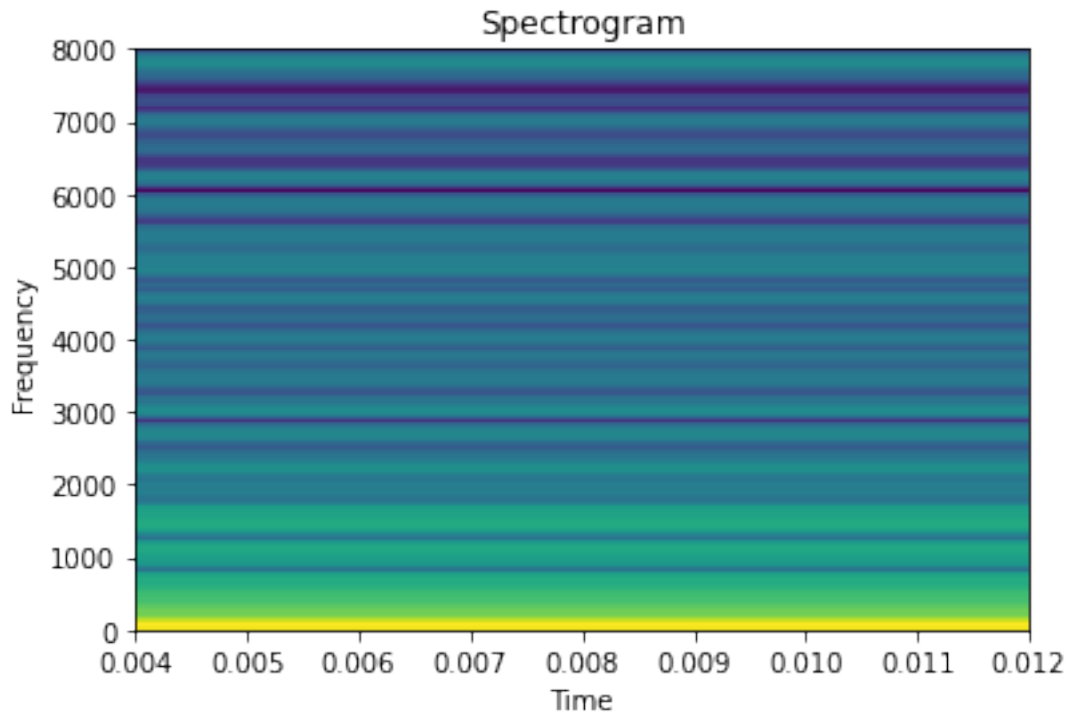


```
[11]: print('Noisy File SPECTOGRAM')
      plot_spectrogram(samplerate_noisy, reduced_log_noisy)
```

Noisy File SPECTOGRAM

C:\Users\ujjaw\AppData\Local\Temp\ipykernel\_11724\593759964.py:4: UserWarning:  
Only one segment is calculated since parameter NFFT (=256) >= signal length  
(=128).

```
plt.specgram(data, Fs=frequency)
```



### 3 Computing Whitening Transformation from clean file

```
[12]: #Finding mean of clean data
clean_mean = np.mean(reduced_log_clean, axis = 1)
clean_mean = np.reshape(clean_mean, (128,1))

# Finding Cov matrix of clean data Sx
Sx_clean = np.matmul((reduced_log_clean-clean_mean),np.
    ↪ transpose(reduced_log_clean-clean_mean))
Sx_clean = np.multiply(Sx_clean,1/298)

# Performing eigen analysis on Sx_clean
e_val, e_vec = np.linalg.eig(Sx_clean)

#Sorting Eigen Values and Corresponding Eigen Vectors
idx = e_val.argsort()[::-1]
e_val = e_val[idx]
e_vec = e_vec[:,idx]

# Finding sigma^(1/2)
sigma_power_half = np.array(np.zeros((128, 128)))
for i in range(128):
```

```

sigma_power_half[i][i] = math.sqrt(e_val[i])

sigma_power_half_inverse = np.linalg.inv(sigma_power_half)

temp_np_array = np.matmul(sigma_power_half_inverse, np.transpose(e_vec))

```

## 4 Applying whitening transformation on clean file

```

[13]: avg_non_diag_clean = compute_non_diag_avg(temp_np_array, reduced_log_clean)
print ("Average of non-diagnoal elements Clean_A = ",avg_non_diag_clean)

```

Average of non-diagnoal elements Clean\_A = 5.624106188325451e-15

## 5 Applying whitening transformation on noisy file

```

[14]: avg_non_diag_noisy = compute_non_diag_avg(temp_np_array, reduced_log_noisy)
print ("Average of non-diagnoal elements Noisy_A =",avg_non_diag_noisy)

```

Average of non-diagnoal elements Noisy\_A = 0.20262565358264523

## 6 Part B

## 7 Computing Whitening Transformation from noisy file

```

[15]: #Finding mean of noisy data
noisy_mean = np.mean(reduced_log_noisy, axis = 1)
noisy_mean = np.reshape(noisy_mean, (128,1))

# Finding Cov matrix of clean data Sx
Sx_noisy = np.matmul((reduced_log_noisy-noisy_mean),np.
    ↪transpose(reduced_log_noisy-noisy_mean))
Sx_noisy = np.multiply(Sx_noisy,1/298)

# Performing eigen analysis on Sx_noisy
e_val_B, e_vec_B = np.linalg.eig(Sx_noisy)

#Sorting Eigen Values and Corresponding Eigen Vectors
idx = e_val_B.argsort()[::-1]
e_val_B = e_val_B[idx]
e_vec_B = e_vec_B[:,idx]

# Finding sigma^(1/2)
sigma_power_half_B = np.array(np.zeros((128, 128)))
for i in range(128):
    sigma_power_half_B[i][i] = math.sqrt(e_val_B[i])

```

```
sigma_power_half_inverse_B = np.linalg.inv(sigma_power_half_B)

temp_np_array_B = np.matmul(sigma_power_half_inverse_B, np.transpose(e_vec_B))
```

## 8 Applying whitening transformation on clean file

```
[16]: avg_non_diag_clean_B = compute_non_diag_avg_B(temp_np_array_B,
    ↪reduced_log_clean)
print ("Average of non-diagnoal elements Clean_B = ",avg_non_diag_clean_B)
```

Average of non-diagnoal elements Clean\_B = 0.12678101775576353

## 9 Applying whitening transformation on noisy file

```
[17]: avg_non_diag_noisy_B = compute_non_diag_avg_B(temp_np_array_B,
    ↪reduced_log_noisy)
print ("Average of non-diagnoal elements Noisy_B = ",avg_non_diag_noisy_B)
```

Average of non-diagnoal elements Noisy\_B = 1.1065210379219143e-15

Observation

Whitening using clean data i. average covariance between the frames is higher in noisy file ii. average covariance between the frames is lower in clean file

Whereas, a complete opposite is observed when Whitening is done using noisy data i. average covariance between the frames is lower in noisy file ii. average covariance between the frames is higher in clean file