## Problem 1

i

```
wsnr1: 34.9608166112122
wsnr2: 54.990959931839626
wsnr3: 63.94678206339311
```

Figure 1 WSNR of window size = 100,500,1000 (filters of order = 5)

```
wsnr1: 80.81119327151103
wsnr2: 98.69810588345615
wsnr3: 100.83009041272322
```

Figure 2 WSNR of window size = 100,500,1000 (filters of order = 15)

```
wsnr1: 60.30885649239928
wsnr2: 83.69093324637845
wsnr3: 93.73876730731774
```

Figure 3 WSNR of window size = 100,500,1000 (filters of order = 30)

	5	15	30
100	34.96	80.81	60.31
500	54.99	98.70	83.69
1000	63.95	100.83	93.74

Figure 4 table about WSNR

From the table, we can see that the WSNR gets bigger as the larger window size. But as the filter orders changing, the WSNR increases first and then decreases.

ii

```
wsnr3 = wsnr(X,15,1000)

wsnr4 = wsnr(X[1000:len(X)],15,1000)

wsnr5 = wsnr(X[2000:len(X)],15,1000)

wsnr6 = wsnr(X[3000:len(X)],15,1000)
```

Figure 5 setting for different windows

wsnr3: 99.54718002386956 wsnr4: 99.54718002386956 wsnr5: 99.54718002386956 wsnr6: 99.54718002386956

Figure 6 WSNR for different windows

From the figures above we can see that after applying the Wiener filter in different windows(0,1000,2000,3000), the result is the same. So only one window suffices.

iii

mse1: 0.09339820384898823 mse2: 0.29297595375303226 mse3: 1.4834507601251334

Figure 7 MSE of different noise(0.1,0.3,1.5)

wsnr1: 97.82953344450773 wsnr2: 82.41917315660089 wsnr3: 74.91767758255003

Figure 8 WSNR of different noise(0.1,0.3,1.5)

From the figures above we can see that with the noise increasing, the results of Wiener filter become worse. Because the noise makes Wiener filter more difficult get right results.

iv

mseLMS: 0.11241284520110223 mseWF: 0.104103135519533

Figure 9 MSE of LMS and Wiener filter

From the figure above we can see that the MSE of LMS is larger than the MSE of Wiener solution. So the performance of Wiener filter is better.

## **Problem 2**

i

```
epower1 = epower(6,100)
epower2 = epower(6,200)
epower3 = epower(6,500)
epower4 = epower(15,100)
epower5 = epower(15,200)
epower6 = epower(15,500)
```

Figure 10 setup for filters

```
epower1: 798.8138363927461
epower2: 863.793700357631
epower3: 1047.2021337197102
epower4: 374.48870576194406
epower5: 430.9110631996131
epower6: 568.3305819084
```

Figure 11 error power for prediction

	6	15
100	798.81	374.49
200	863.79	430.91
500	1047.20	568.33

Figure 12 table about error power

From the table above we can see that the performance of filter order=15 is better than the performance of filter order=6. And the bigger the window size, the worse the performance of the predictor.

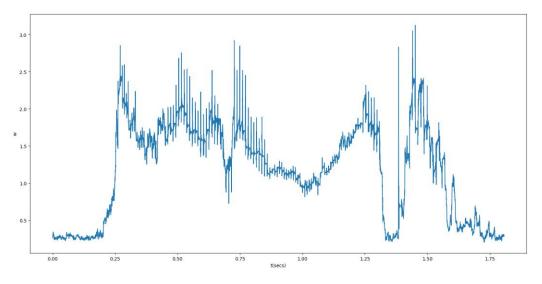


Figure 13 w changes over time

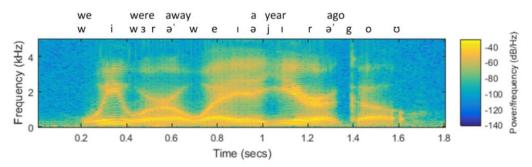


Figure 14 wide-band sound spectrogram

From the figure above we can see that the trend of w changes is the same as the changing of frequency of the sounds.

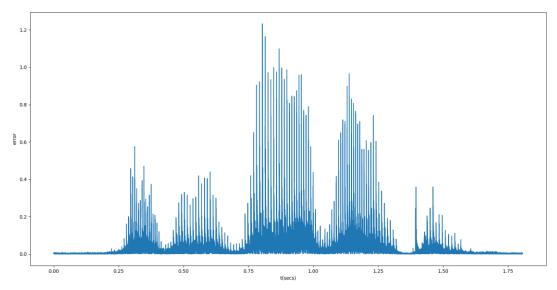


Figure 15 prediction error changes over time

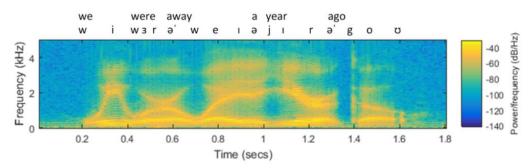


Figure 16 wide-band sound spectrogram

After comparing the two figures above, we can see that when the frequency of the sounds changes a lot, the prediction error will also change a lot. The trends of the sound frequency and the prediction error are the same.

For improving prediction, we can change the window size with sound frequency. If the sound frequency does not change much over time, we can set windows during this time to improve prediction. Also, we can increase the orders of filter.

Convergence rate is important because it can tell us how to set window. If the convergence rate becomes smaller over a period of time, we can make the window size larger. Otherwise, if the convergence rate becomes very big, we should know that there are some big changes about the sounds, so we should make the window size smaller.

iv

epowerLMS: 109.84029750737945 epowerWF: 798.8138363927461

Figure 17 error power of LMS and Wiener filter

From the figure above we can see that the error power of Wiener filter is larger than the error power of LMS. So the performance of LMS is better.