Assignment 2: Linear Predictive Analysis

Pranav Sankhe— 150070009 4/10/2018

1 Question

Synthesized vowel: Consider the synthesized vowel /a/ (formants: 730, 1090, 2440 Hz; bandwidths: 50 Hz) at two fundamental frequencies: 120 Hz, 300 Hz. Sampling rate = 8 kHz. Using a 30 ms Hamming window, implement LP analysis on a single segment using LP orders 2, 4, 6, 8, 10 using the Levinson algorithm. Compute the gain, and plot the LP spectrum magnitude (i.e. the dB magnitude frequency response of the estimated all-pole filter) for each order "p". Superimpose each plot on the original 6-pole spectral envelope with the discrete harmonic components shown with vertical lines. Comment on the characteristics of the spectral approximations of different orders.

1.1 Answer

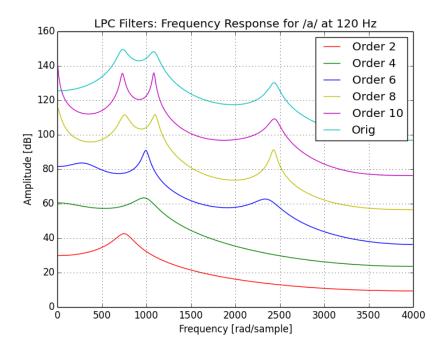


Figure 1: \a\at 120 Hz

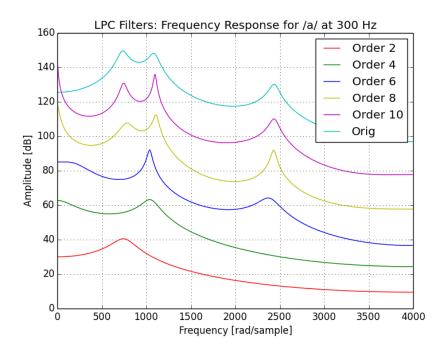


Figure 2: \a\at 300 Hz

It is evident from the graphs that the approximation of the real spectrum keeps on getting better as we increase the order at both the signal frequencies. This is expected since as we increase the order, the LP spectrum is able to capture more and more details of the original spectrum. Also, as we increase the number of poles (order) we observe that there's an increase in the sharpness of the peaks.

File containing all the paramaters:

```
formant_freq = [730, 1090, 2440]
formant_bw = [100, 100, 100]
samp_freq = 8000.0
sig_freq = 300
time_length = 0.5
win_size = 30.0
dft_len = 1024
filter_order= [2, 4, 6, 8, 10]
```

Listing 1: hparams.py

main file:

```
import librosa
import numpy as np
from matplotlib import pyplot as plt
import pylab
from scipy import signal
from scipy.io.wavfile import write
import hparams
import sys
```

```
9 from math import pi
10
11
def autocorr(x):
        result = np.correlate(x, x, mode='full')
13
        return result[int(result.size/2):]
14
15
16
sig_freq = hparams.sig_freq
19 y, samp_freq = librosa.load( str(sig_freq) + `.wav', sr=None)
win_size = hparams.win_size /1000.0
num_samples = int(samp_freq*win_size)
window = y[:num_samples]*np.hamming(num_samples)
fig1 = plt.figure()
plt.title('LPC Filters: Frequency Response for /a/ at '+str(
        sig_freq)+' Hz')
plt.ylabel('Amplitude [dB]')
plt.xlabel('Frequency [rad/sample]')
{\tt 29} \  \, {\tt colors} \, = \, \{2 \colon \ {\tt 'r'} \, , \  \, 4 \colon \ {\tt 'g'} \, , \  \, 6 \colon \ {\tt 'b'} \, , \  \, 8 \colon \ {\tt 'y'} \, , \  \, 10 \colon \ {\tt 'm'} \}
30 orders = hparams.filter_order
31
32 for order in orders:
33
        R = autocorr(window)
34
35
        error = np.zeros(order+1)
        error[0] = R[0]
36
        G = np.zeros(order + 1)
37
38
        coeffs = np.zeros(order+1)
39
        dummy\_coeffs = np.zeros(order + 1)
40
41
42
        for i in range (1, order +1):
43
44
             reflec\_coeffs = 0
             dummy_coeffs[1:len(coeffs)] = coeffs[1:len(coeffs)]
45
46
             \begin{array}{lll} & \text{for } j & \text{in } range\,(1\,,\ i\,): \\ & & \text{reflec\_coeffs} \,=\, reflec\_coeffs \,+\, dummy\_coeffs\,[\,j\,]*R[\,i-j\,] \end{array}
47
48
             reflec_coeffs = (R[i] - reflec_coeffs)/error[i-1]
49
50
             coeffs [i] = reflec_coeffs
52
             for j in range (1, i):
53
                  coeffs[j] = dummy_coeffs[j] - reflec_coeffs*
54
        dummy_coeffs[i-j]
56
             error[i] = (1-np.square(reflec_coeffs))*error[i-1]
57
        coeffs[0] = 1.0
58
        coeffs[1:len(coeffs)] = -coeffs[1:len(coeffs)]
59
        num_coeffs = np.zeros(coeffs.shape)
60
        num\_coeffs[0] = 1
61
        G[i] = np.sqrt(error[i])
62
        w,\ h = signal.freqz(num\_coeffs,\ coeffs)
63
64
        plt.plot\left(samp\_freq*w/(2*np.pi\right),\ 10*order\ +\ 20\ *\ np.log10\left(abs(h)\right)
65
        ), colors [order])
formant_freq = hparams.formant_freq
```

```
formant_bw = hparams.formant_bw
formant_bw = hparams.formant_bw | formant_bw | freq
formant_bw = hparams.formant_bw | freq
formant_bw | freq
formant_freq
formant_
```

Listing 2: q1.py

2 Question 2

Natural speech: Consider the speech signal in machali.wav (male voice), sampled at 8 kHz. Consider the following signal segments in the final word "pani": (1) \a\(first half); (2) \n\; (3) \I\\and (4) \s\\in the word uska.

Use PRAAT to extract the above segments to separate .wav files for further analyses as below. (Note: for $\s\setminus$, 16 kHz sampled audio is better.)

Compute and plot the narrowband spectrum using a Hamming window of duration = 30 ms before and after pre-emphasis.

Using a 30 ms Hamming window centered in the segment of the waveform (pre-emphasised for the voiced sounds):

Compute the autocorrelation coefficients required for LPC calculation at various p=4,6,8,10,12,20. Use the Levinson algorithm to compute the LP coefficients from the autocorrelation coefficients. Show the pole-zero plots of the estimated all-pole filter for p=6,10.

Compute the gain and plot the LPC spectrum magnitude (i.e. the dB magnitude frequency response of the estimated all-pole filter) for each order "p". Superimpose each plot on the narrowband dB magnitude spectrum of part 1 (after pre-emphasis). Comment on the characteristics of the spectra.

Plot error signal energy (i.e. square of gain) vs p.

2.1 Answer

Pre-Emphasis Note that in the plots, the original spectra is in blue and the pre-emphasized spectra is in green.

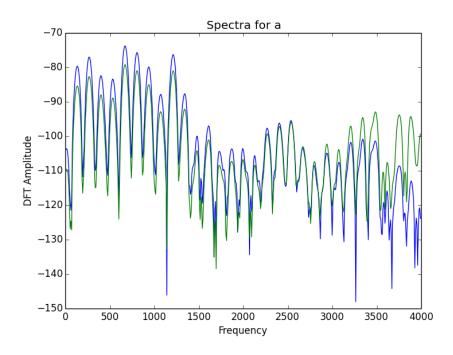


Figure 3: $\arrowvert \arrowvert \arrowvert$

Poles and Zeros

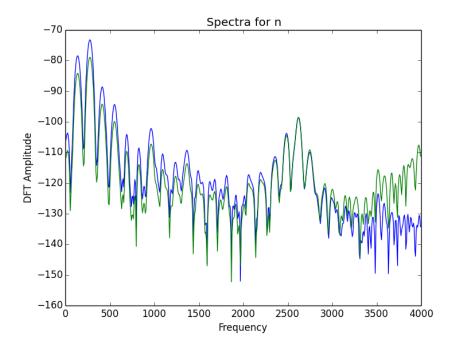


Figure 4: \n from the natural recording of machali

Linear Predictive Coding Spectrum

The gain for $\acksim a$ from the natural recording of machali is:

• Order 2: 15.30

• Order 4: 13.13

• Order 6: 12.84

• Order 8: 12.42

• Order 10: 10.99

• Order 12: 10.47

• Order 20: 10.07

The gain for \n from the natural recording of machali is:

• Order 2: 39.31

• Order 4: 24.83

• Order 6: 17.51

• Order 8: 17.07

• Order 10: 17.00

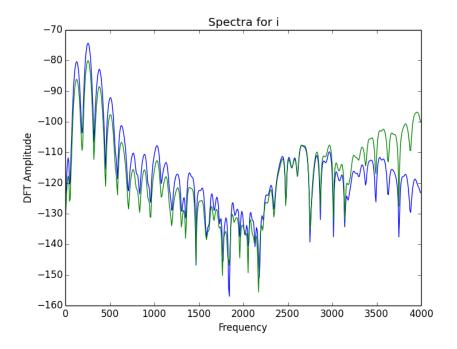


Figure 5: $\I\$ from the natural recording of machali

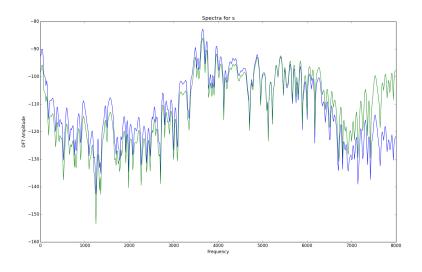


Figure 6: \s\from the natural recording of machali

• Order 12: 16.58

• Order 20: 16.12

The gain for \I\from the natural recording of machali is:

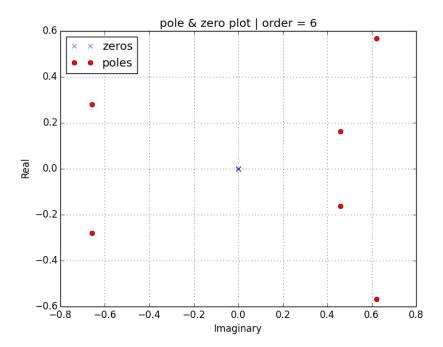


Figure 7: Poles and zeros when order = 6 for \a\from the natural recording of machali

• Order 2: 50.79

• Order 4: 28.53

• Order 6: 25.04

• Order 8: 22.36

• Order 10: 21.19

• Order 12: 21.05

• Order 20: 19.42

The gain for \s\from the natural recording of machali is:

• Order 2: 67.89

 \bullet Order 4: 65.96

 \bullet Order 6: 45.62

• Order 8: 39.59

• Order 10: 39.07

• Order 12: 38.32

• Order 20: 36.29

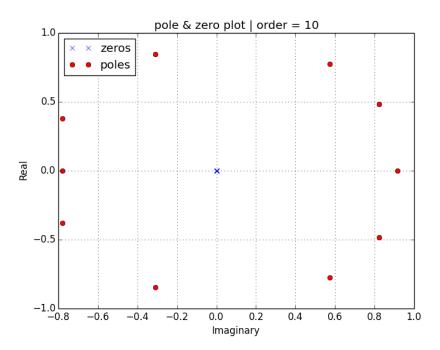


Figure 8: Poles and zeros when order = 10 for \a\from the natural recording of machali

Error signal energy

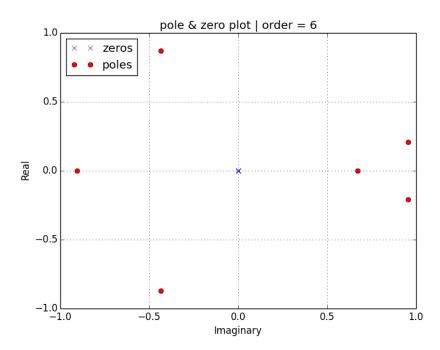


Figure 9: Poles and zeros when order = 6 for \n\from the natural recording of machali

3 Question 3

Based on the 10th-order LPCs, carry out the inverse filtering of the \a\vowel segment and of the unvoiced sound \s\. Obtain the residual error signal in each case. Can you measure the pitch period of the voiced sound from the residual waveform? Use the acf to detect the pitch. Plot the magnitude spectrum of each of the residual signals.

3.1 Answer

From the figure, we can see that the autocorrelation function for \a achieves a maxima (spike) at around 60 samples. Given that the sampling rate is 8 KHz, the pitch of the signal should be 8000/60 = 133.33 Hz.

We could do this since \a\is a voice sound. On the other hand, \s\is an unvoiced sound and hence we can't observe any maxima in it't autocorrelation function. Hence we cannot calculate a pitch for this sound.

Spectrum of residual signals

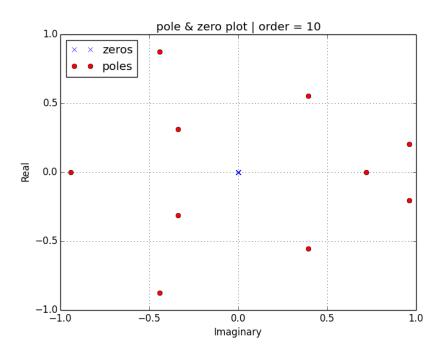


Figure 10: Poles and zeros when order = 10 for \n\from the natural recording of machali

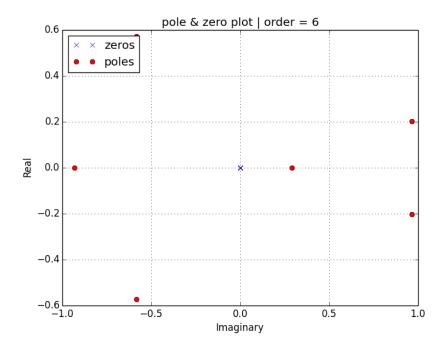


Figure 11: Poles and zeros when order = 6 for \i\from the natural recording of machali

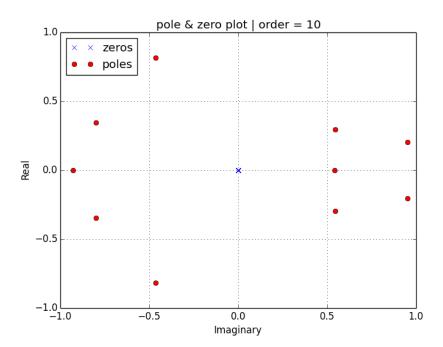


Figure 12: Poles and zeros when order = 10 for \i\from the natural recording of machali

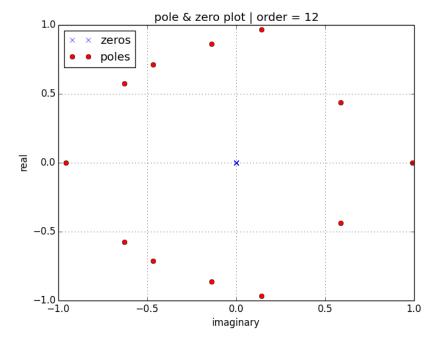


Figure 13: Poles and zeros when order = 12 for \s\from the natural recording of machali

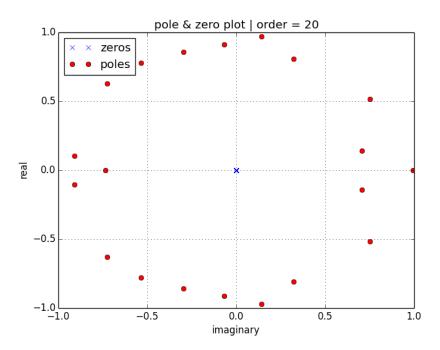


Figure 14: Poles and zeros when order = 20 for \s\from the natural recording of machali

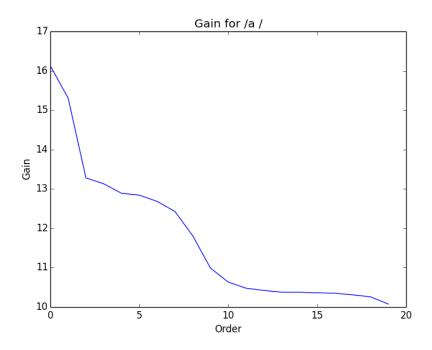


Figure 15: Gain of \a\from the natural recording of machali

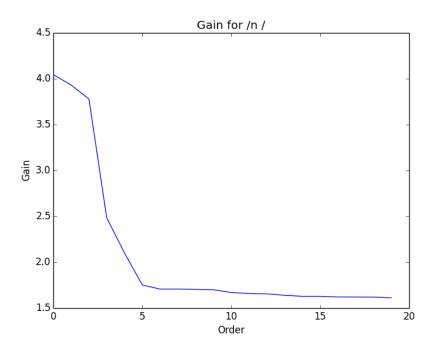


Figure 16: Gain of \n\from the natural recording of machali

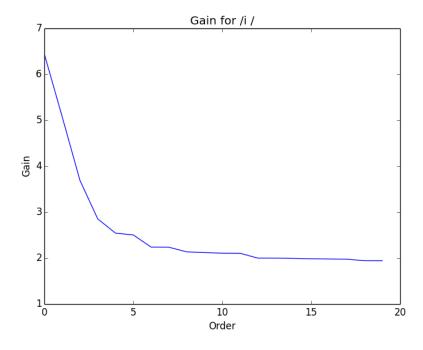


Figure 17: Gain of **\I\from** the natural recording of machali

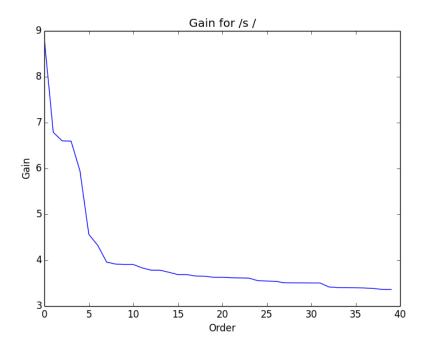


Figure 18: Gain of \s\from the natural recording of machali

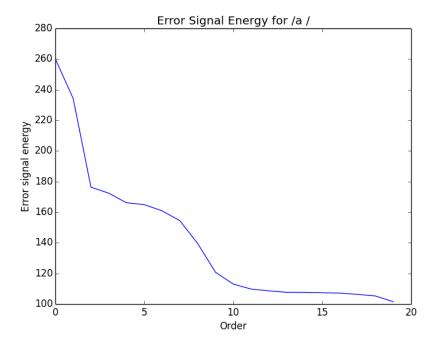


Figure 19: Error Signal Energy for \i\from the natural recording of machali

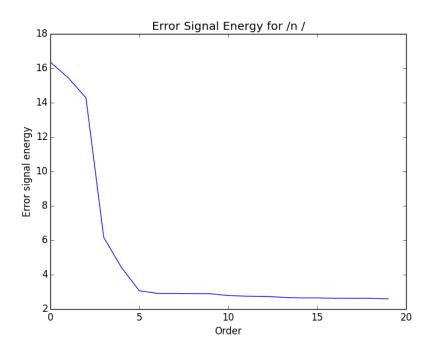


Figure 20: Error Signal Energy for $\in \text{Moreover}$ from the natural recording of machali

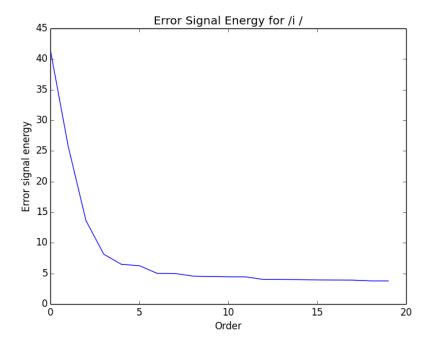


Figure 21: Error Signal Energy for \s\from the natural recording of machali

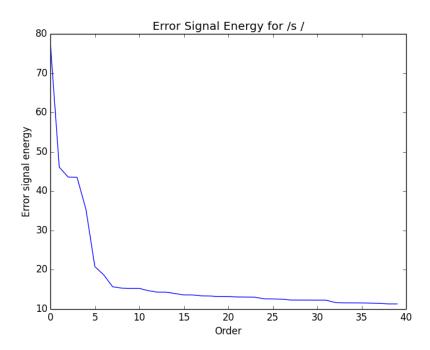


Figure 22: Error Signal Energy for \s\from the natural recording of machali

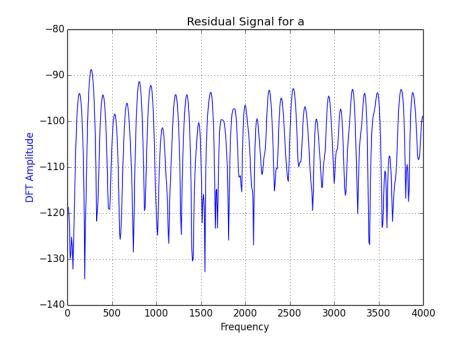


Figure 23: Residual signal for \a\from the natural recording of machali

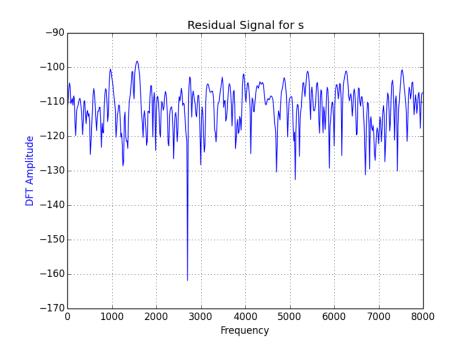


Figure 24: Residual signal for \slash sfrom the natural recording of machali

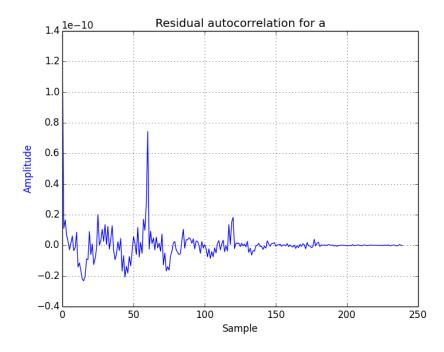


Figure 25: Autocorrelation of the residual signal for \a\from the natural recording of machali

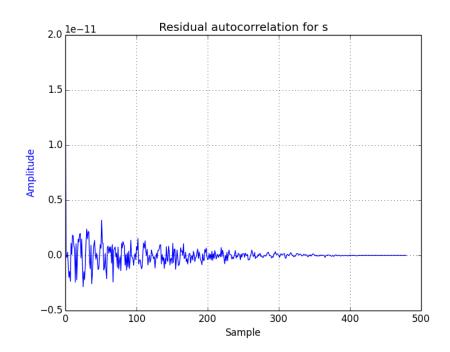


Figure 26: Autocorrelation of the residual signal for \s\from the natural recording of machali

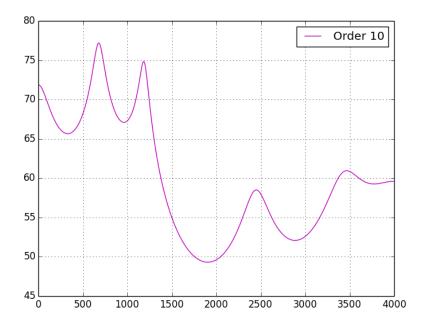


Figure 27: Spectrum of the residual signal for \a\from the natural recording of machali

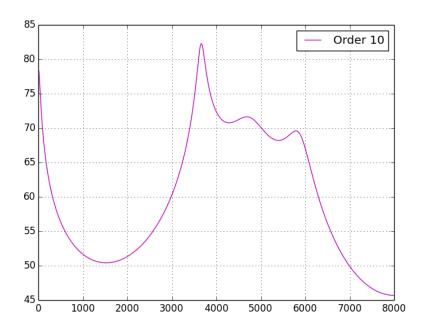


Figure 28: Spectrum of the residual signal for \s\from the natural recording of machali