# **Homework 4**

## **Import libraries**

```
In [1]: import numpy as np import librosa import matplotlib.pyplot as plt from spectrum import * from spectrum.linear_prediction import * from scipy import signal
```

## Let matplotlib open a figure window

```
In [2]: %matplotlib inline
```

## **IpcFrames function**

```
In [3]: | # create a function to compute lpcs per frame
        def lpcFrames(x, P = 10, fs = 8000, doPlot = 1):
            x = sound waveform vector
            P = order \ of \ the \ LPC \ analysis \ (default \ 10)
            fs = the sampling rate in Hz (default 8000Hz)
            doPlot = flag for turning plots off (0) or on (1 = default)
            # set the frame size
            fSize = 512
            #the current frame number
            fNum = 0
            # the start index of the current frame
            fStart = 0
            # the end index of the current frame
            fEnd = fSize - 1
            # the frame hop size
            fHop = fSize / 2
            # length of input sound vector
            L = len(x)
            \# calculated number of frames in x
            nFrames = int(np.floor(L/fHop) - 1)
            # create empty output array
            A = np.zeros((P+1, nFrames))
            # hanning window
            win = np.hanning(fSize)
            # frequency axis vector for plots
            f = (fs/2) * np.arange(fHop) / fHop
            if doPlot:
                fig, axes = plt.subplots(2,1,figsize=(18,12))
            while (fEnd < L - 1):
                # current frame of data
                frame = x[int(fStart):int(fEnd + 1)]
                # windowed frame
                wFrame = np.multiply(frame, win)
                # lpc polynomial a(z): H(z) = g/a(z)
                a, e = lpc(wFrame, P)
                # add one to beginning of a
                a = np.concatenate(([1],a),0)
                # collect lpc coefficients into output matrix
                A[:,fNum] = a.T
                # generate plots
                if doPlot:
                     #plot windowed frame
                     axes[0].plot(wFrame)
                     plt.cla()
                    # plot stft and lpc spectral envelope
                     S = np.fft.fft(wFrame)
                     w, h = signal.freqz(np.abs(e * 1024), a, int(fSize/2))
                     hdh = 20 * nn.log10(nn.ahs(h))
```

### Load audio file

### **Compute LPC frames**

```
In [6]: A = {}
for fname in filenames:
    A[fname] = lpcFrames(X[fname][0], doPlot=0)
```

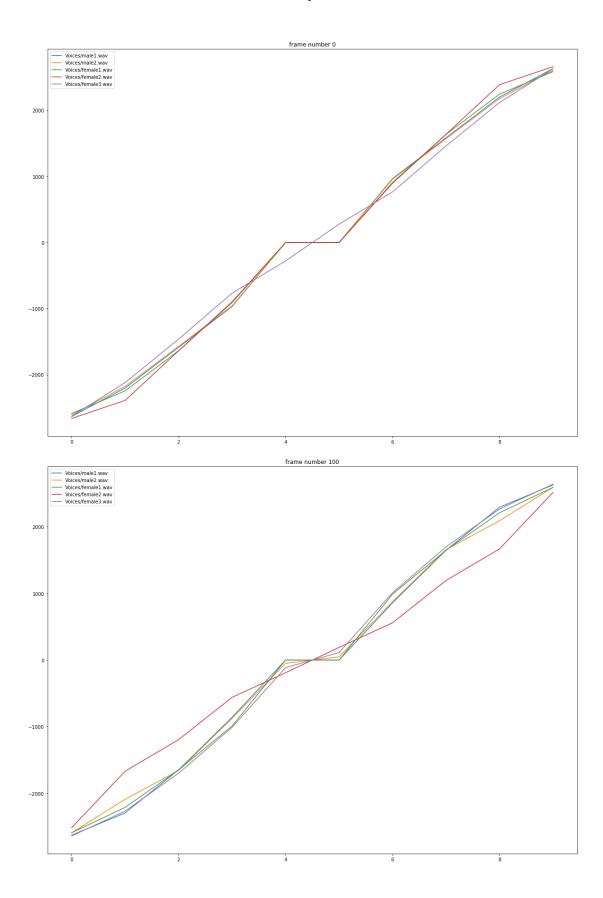
#### Your answers

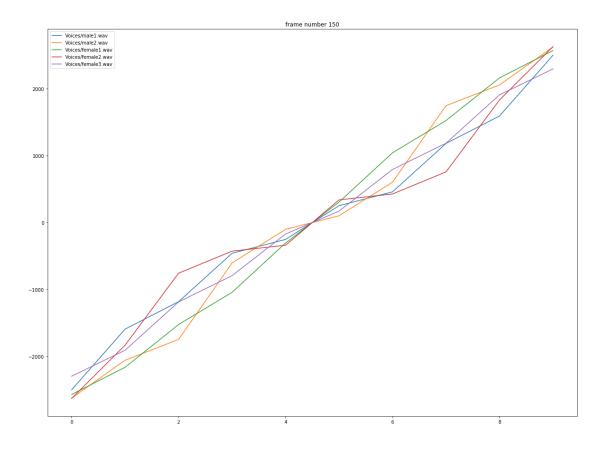
(a)

```
In [7]:

def get_pole_angles(A, fs=8000):
    pole_angles = np.zeros((A.shape[0]-1, A.shape[1]))
    for i in range(0, A.shape[1]):
        pole_angles[:,i] = sorted(np.angle(np.roots(A[:,i])))
    pole_angles = pole_angles/np.pi*(fs/2)
    return pole_angles

frames = [0, 100, 150];
for fr in frames:
    plt.figure(figsize=(20,15))
    for fname in filenames:
        pole_angles = get_pole_angles(A[fname])
        plt.plot(pole_angles[:,fr], label=fname)
    plt.legend(loc='upper left')
    plt.title("frame number {}}".format(fr))
    plt.pause(0.05)
```





(b)

Since the roots of the columns of A come in complex conjugate pairs, the extracted angles are symmetric at y=0. Hence, it would be better to filter out the negative frequencies. However, this will cause the number of frequencies per frame to be divided in half (5 instead of 10), which will cause some information to be lost. Also, it is seen from the graph that some voices' pole angles' patterns are very different from all of the other voices. Hence, it can be difficult to know which pattern should be picked.

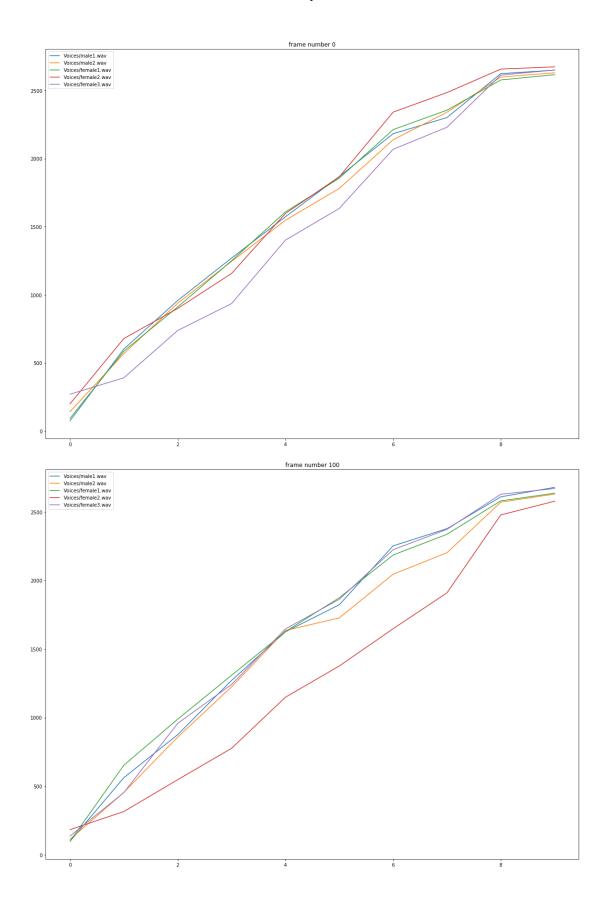
(c)

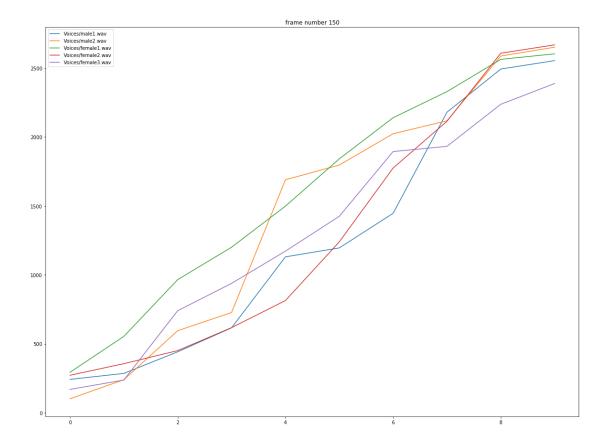
```
In [8]:

def get_lsf(A, fs=8000):
    lsf = np.zeros((A.shape[0]-1, A.shape[1]))
    for i in range(0, A.shape[1]):
        lsf[:,i] = poly2lsf(A[:,i])
    lsf = lsf/np.pi*(fs/2)
    return lsf

frames = [0, 100, 150];
for fr in frames:
    plt.figure(figsize=(20,15))
    for fname in filenames:
        lsf = get_lsf(A[fname])
        plt.plot(lsf[:,fr], label=fname)
    plt.legend(loc='upper left')
    plt.title("frame number {}}".format(fr))
    plt.pause(0.05)
```

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(d)

The LSF is better suited for identifying phonemems than the pole angles because LSF only contains positive frequencies, which means that there is no need to cut the freq vector in half, hence, higher resolution. Also, LSF is less susceptible to noise, which makes identifying phonemems patterns easier. From the graph, there are still frames where the voice's pattern don't exactly match. However, their curve shapes do match better than that in part (a).