```
In [1]: %matplotlib inline
    import matplotlib.pyplot as plt
    import seaborn

import librosa
    import librosa.display

from IPython.display import Audio
    import requests
    import os

import sklearn
    import numpy as np

import pandas as pd
```

```
In [2]: #Function for a genre classifier of two different audios
        def genre classifier(audio1, audio2):
            #Load 120 seconds of the audios
            song1 = librosa.load(audio1, duration = 120)
            song2 = librosa.load(audio2, duration = 120)
            #Prepare scaling the features to have zero mean and unit variance
            scaler = sklearn.preprocessing.StandardScaler()
            #Calculate the MFCC for audio1
            mfcc_song1 = librosa.feature.mfcc(song1[0], song1[1])
            mfcc song1 = mfcc song1.T
            #Scale the MFCC
            mfcc_song1_scaled = scaler.fit_transform(mfcc_song1)
            #Calculate the MFCC for audio2
            mfcc song2 = librosa.feature.mfcc(song2[0], song2[1])
            mfcc song2 = mfcc song2.T
            #Scale the MFCC
            mfcc song2 scaled = scaler.fit transform(mfcc song2)
            #Concatenate all of the scaled feature vectors into one feature table
            features = np.vstack((mfcc song1 scaled, mfcc song2 scaled))
            labels = np.concatenate((np.zeros(len(mfcc song1 scaled)), np.ones(len(mfcc song2 scaled))))
            # Support Vector Machine
            model = sklearn.svm.SVC()
            #Train the classifier
            model.fit(features, labels)
            #Load smaller samples of the audios
            x song1 test, fs song1 = librosa.load(audio1, duration=10, offset=120)
            x_song2_test, fs_song2 = librosa.load(audio2, duration=10, offset=120)
            #Calculate the MFCCs for the two samples
            mfcc song1 test = librosa.feature.mfcc(x song1 test, fs song1)
            mfcc song1 test = mfcc song1 test.T
            mfcc song2 test = librosa.feature.mfcc(x song2 test, fs song2)
            mfcc song2 test = mfcc song2 test.T
            #Scale the MFCCs
```

```
mfcc_song1_test_scaled = scaler.fit_transform(mfcc_song1_test)
mfcc_song2_test_scaled = scaler.fit_transform(mfcc_song2_test)

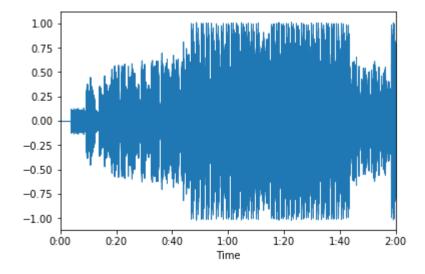
#Concatenate all test features together
test_features = np.vstack((mfcc_song1_test_scaled, mfcc_song2_test_scaled))
#Concatenate all test labels together
test_labels = np.concatenate((np.zeros(len(mfcc_song1_test_scaled)), np.ones(len(mfcc_song2_test_scaled))))

#Compute the accuracy score of the classifier on the test data
score = model.score(test_features, test_labels)
return score
```

```
In [3]: #Load 120 seconds of the audios
    JColeApparently = librosa.load('data/J. Cole - Apparently (Official Music Video).mp3', duration = 120)
    RihannaDiamonds = librosa.load('data/Rihanna - Diamonds.mp3', duration = 120)

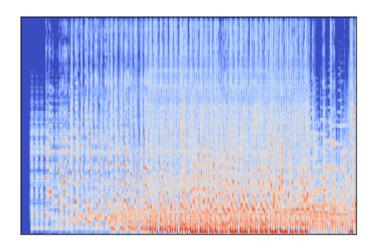
#Plot the time-domain waveform of JColeApparently
    librosa.display.waveplot(JColeApparently[0])
```

Out[3]: <matplotlib.collections.PolyCollection at 0x21586b72278>



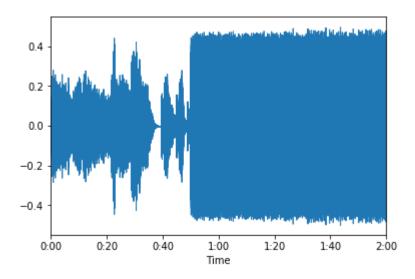
In [4]: #Calculate and display the mel spectogram with a logged scale for JColeApparently
 JColeApparently_mel = librosa.feature.melspectrogram(JColeApparently[0])
 JColeApparently_mel_log = librosa.power_to_db(JColeApparently_mel)
 librosa.display.specshow(JColeApparently_mel_log)

Out[4]: <matplotlib.axes._subplots.AxesSubplot at 0x215878296d8>



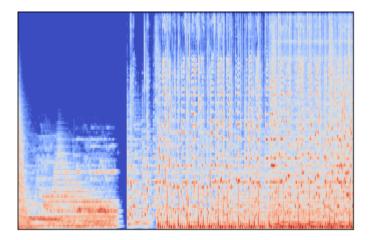
In [5]: #Plot the time-domain waveform of RihannaDiamonds
librosa.display.waveplot(RihannaDiamonds[0])

Out[5]: <matplotlib.collections.PolyCollection at 0x21586867e10>



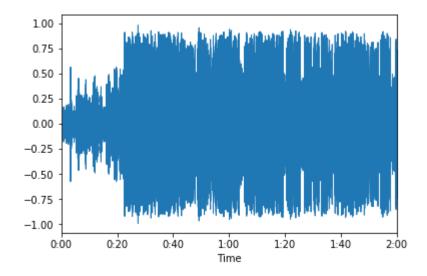
In [6]: #Calculate and display the mel spectogram with a logged scale for RihannaDiamonds RihannaDiamonds_mel = librosa.feature.melspectrogram(RihannaDiamonds[0]) RihannaDiamonds_mel_log = librosa.power_to_db(RihannaDiamonds_mel) librosa.display.specshow(RihannaDiamonds_mel_log)

Out[6]: <matplotlib.axes._subplots.AxesSubplot at 0x215868b2588>

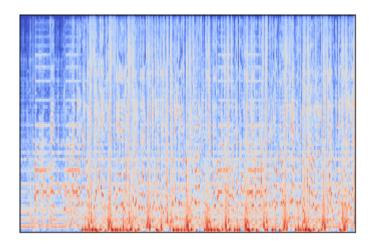


In [7]: #Load 120 seconds of the audios JColeWetDreamz = librosa.load('data/J. Cole - Wet Dreamz (Official Music Video).mp3', duration = 120) RihannaWork = librosa.load('data/Rihanna - Work (Explicit) ft. Drake.mp3', duration = 120) #Plot the time-domain waveform of JColeWetDreamz librosa.display.waveplot(JColeWetDreamz[0])

Out[7]: <matplotlib.collections.PolyCollection at 0x21586835630>

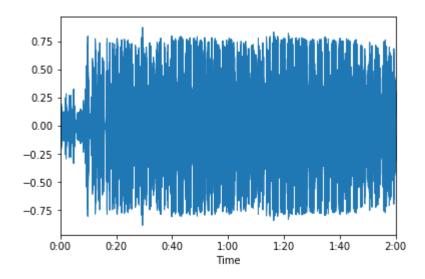


Out[8]: <matplotlib.axes._subplots.AxesSubplot at 0x21586ac3940>



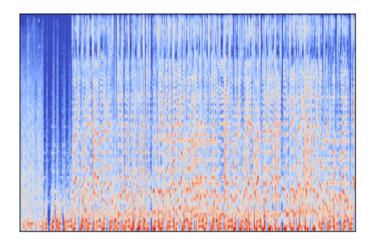
In [9]: #Plot the time-domain waveform of RihannaWork
librosa.display.waveplot(RihannaWork[0])

Out[9]: <matplotlib.collections.PolyCollection at 0x21586adbf28>



In [10]: #Calculate and display the mel spectogram with a logged scale for RihannaWork RihannaWork_mel = librosa.feature.melspectrogram(RihannaWork[0]) RihannaWork_mel_log = librosa.power_to_db(RihannaWork_mel) librosa.display.specshow(RihannaWork_mel_log)

Out[10]: <matplotlib.axes._subplots.AxesSubplot at 0x21586ad0978>



```
In [11]: genre_classifier('data/J. Cole - Apparently (Official Music Video).mp3', 'data/Rihanna - Diamonds.mp3')
Out[11]: genre_classifier('data/J. Cole - Wet Dreamz (Official Music Video).mp3', 'data/Rihanna - Diamonds.mp3')
```

Out[12]: 0.728538283062645

```
In [13]: genre_classifier('data/J. Cole - Wet Dreamz (Official Music Video).mp3', 'data/Rihanna - Work (Explicit) ft. Dra
Out[13]: 0.8793503480278422
In [14]: genre_classifier('data/J. Cole - Apparently (Official Music Video).mp3', 'data/Rihanna - Work (Explicit) ft. Dra
Out[14]: 0.8897911832946636
In [15]: genre_classifier('data/Rihanna - Diamonds.mp3', 'data/Rihanna - Work (Explicit) ft. Drake.mp3')
Out[15]: 0.6264501160092807
In [16]: genre_classifier('data/J. Cole - Apparently (Official Music Video).mp3', 'data/J. Cole - Wet Dreamz (Official Music Video).mp3', 'data/J. Cole
```

Findings:

The two songs from J. Cole (Apparently, Wet Dreamz) are the genre of Hip-Hop/Rap, and the two songs from Rihanna (Diamonds, Work) are the genre of Pop according to Google (Google Search tell you the genre of song when you search the song name). Thus, the songs Apparently and Wet Dreamz should have low scores when compared, and the songs Diamonds and Work should also have low scores.

Apparently, that is seems to be the case as the classifier score between between Diamonds and Work is low (0.62). Expectedly, the score between Rihanna's Work, a pop song, and both J. Cole's Hip-Hop/Rap songs is really high (0.88 and 0.89), showing that they are indeed two different genres. Score between Apparently and Wet Dreamz is lower, but not the best (0.81). Though, the scores between Rihanna's Diamonds and the two of J. Cole's Hip-Hop/Rap songs are 0.75 and 0.73, and lower than the score between J. Cole's own songs Apparently and Wet Dreamz.

Some guesses can be drawn:

- 1) The method of the genre classifier which uses MFCCs to measure similarities between songs' genres may be reliable.
 - Just a note, obviously, Work and Apparently and Wet Dreamz have similar looking plotted waveforms (no t the MFCC) in terms of shapes and therefore they may have high scores.
 - Maybe songs with similar looking plotted waveforms tend to have high score; there may be possible cor relation between Waveforms and MFCCs.

- 2) It is just the song Diamonds that's special; it has a lower score compared against the Hip-Hop songs than the Hip-Hop songs themselves, but not much.
 - Then that means artists has many different genres of songs.
 - Nowadays, so-called "Pop" songs has a mix of different genres and styles inside their arrangemment. M any music styles can be all used in and called "Pop". Because "Pop" music is so diverse, it is hard to use genre classifier and be accurate all the time.

T. F. T	
In :	
[] .	