# EEC 201 Final Project

#### Bishane Sagal and Karmvir Thind

#### March 2024

#### 1 Introduction

The goal of this project is to develop a speaker recognition program. Using 11 given speech samples and 19 speech samples from our colleagues as training data we aim to write a program that will match input test data to the correct student.

## 2 Approach

Our approach to the problem can be summarized into three steps:

- 1. Perform feature extraction on the audio data
- 2. Develop/Choose an algorithm to learn from the features and fit a model
- 3. Classify new audio data using the developed model

#### 2.1 Feature Extraction

For feature extraction we made use of the Mel Frequency Cepstrum Coefficients (MFCCs). We followed a the mechanical procedure for producing the MFCCs: Compute the Short Time Fourier Transform of the input data (STFT), convert the magnitude of the STFT (to eliminate complex numbers) to mel-scale using appropriately spaced filter banks, take the log of the mel-scale STFTs, and perform the Discrete Cosine Transform (DCT) on these values to produce the MFCCs.

### 2.2 Algorithm Selection

We tested various models available through Python's ScikitLearn library and also developed our own implementation of the recommended Linde-Buzo-Gray (LBG) algorithm. Ultimately we landed on the [INSERT FINAL MODEL HERE]. We tested various models using for loops to sweep through various parameters and hyperparameters. Using the classification output we tuned the model to achieve our results of [insert accuracy here]

#### 2.3 Model Prediction

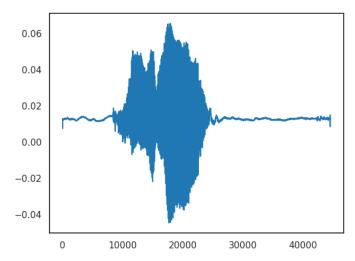
Assessment of our model was done using features of the pre-built models from Python's Scikitlearn or in the case of our LBG algorithm, we displayed a confusion matrix to make our results easy to determine.

### 3 Tests

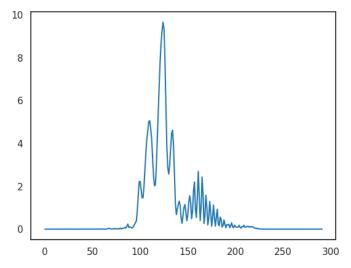
#### 3.1 Test 1

Auditory Performance rate was 7 out of 11 students guessed correctly.

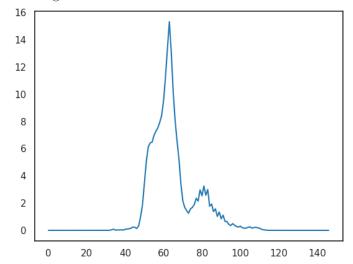
#### 3.2 Test 2



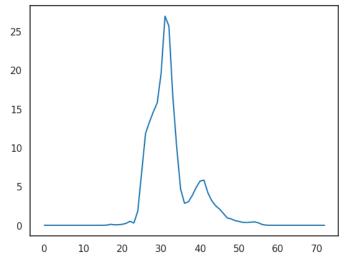
The sampling rate of our load function is 22050, giving us 256 samples in 11 milliseconds.



Periodogram with N=128

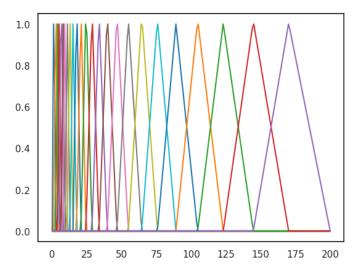


Periodogram with N=256

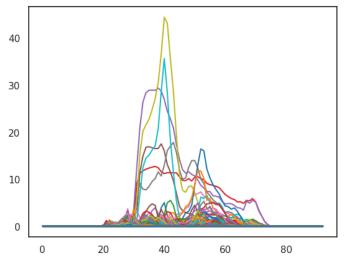


Periodogram with N=512

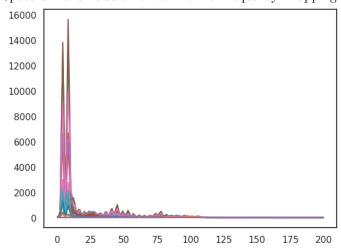
## 3.3 Test 3



Ideal Mel Spaced Filter Banks



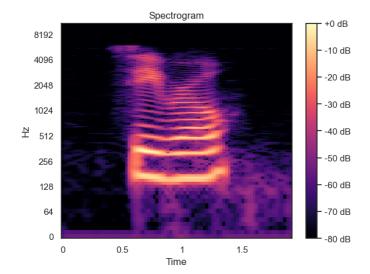
Spectrum of an audiofile with no mel frequency wrapping



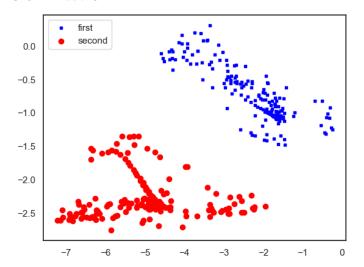
Spectrum of audio file with mel frequency wrapping

### 3.4 Identifying Points of High Energy

As seen below, we have high intensity points in the range of 0.5 to 1 seconds. The highest energy points are with the 128-512 frequency band. The intensity plotting is in accordance with our raw audio file coefficients because our largest amplitudes lie around the 20000 sample which corresponds to approximately 1 second.



## 3.5 Test 5



Training Data MFCC vectors plotted in a 2-D plane