## Final Lab: Speaker Detection

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#### **Abstract**

This lab focuses on the creation of a voice detection algorithm that uses filtering methods and comparisons ("profiles") to determine certain voices from others. After vocal samples of intended "nominal" voices and unauthorized "intruder" voices are recorded, they were be passed through the program, which would output whether or not the vocal sample is authorized or not. The conclusion of the DSP pipeline was that the algorithm achieved a 95% for recognizing nominal voices, and 93% for recognizing intruders.

#### Introduction

The purpose of this lab is to create a program that takes audio input from a multitude of different voices and makes a distinction between them by cleaning up each individual file by removing back ground noise via FIR filter, running it through additional filters designed around an FFT function, and detecting a specific threshold of amplitudes to distinguish one voice from another.

It should be noted that all recordings were manually loaded in code via the audioread() function. Usage of the console for inputs was only required when finding out if a post-processed recording was authorized or not. Batch processing was used in the forms of loops for all audio recordings.

### Methodology

#### **Data Collection**

From a total of 15 different people, a total of 65 vocal samples were recorded, with everyone speaking the same phrase: "The quick brown fox jumps over the lazy dog". 10 individuals were chosen as the "nominal" vocal sample group, and the other 5 were labeled as "intruders". Each nominal was asked to record 5 samples each, while each intruder was asked to record only 3, as the intruders were the voices that the algorithm would be trained to reject, or pinpoint as intruders.

Additionally, the background noise of the recording room was taken in an effort to remove as much noise from the original voice recording as possible.

#### **Input Processing**

Once all samples were recorded, they were all run through a number of specific filters to ensure consistency. All audio was converted from stereo to mono. This is due to the matrices being an L  $\times$  2 measurement initially. Processing a linear row (L  $\times$  1) is computationally easier to navigate.

```
%%%%%%%% batch process the matrixes from stereo to mono. all the matrices
%%%%%%%% are in a stereo format and mono closely represents how we
%%%%%%% physically speak. ie, we are not speaking from two mouths
vars = whos;
for i = 1:length(vars)
   name = vars(i).name;
   % all initial matrixes basically will be grabbed. that's why we went
   % with the naming convention
    if startsWith(name, 'x ')
        data = eval(name);
        if ismatrix(data) && size(data, 2) == 2
            mono = mean(data, 2);
            assignin('base', name, mono);
        end
    end
end
```

Figure 1; code section detailing the conversion from stereo sound to mono.

Additionally, utilizing the recorded silent room noise (which was also converted from stereo to mono), each audio file ensured that any background frequencies were filtered out using an FIR filter for smoother processing and recognition.

Figure 2; code section detailing the construction of the background frequency FIR filter.

Each audio file is processed and given a new naming convention to distinguish it as a "clean" version, rather than the original unprocessed recording. The golden copies were not modified other than transforming them into a mono version.

```
%%%%%%% clean everything now with FIR using another loop similar to
%%%%%%%% earlier one.
vars = whos:
for i = 1:length(vars)
   name = vars(i).name;
    if startsWith(name, 'x_') && ~contains(name, '_clean')
       x = eval(name);
        if size(x, 2) == 2
           x = mean(x, 2);
       x = x(:);
       x = x / max(abs(x));
        x_{clean} = filter(b, 1, x);
        x_clean = x_clean / max(abs(x_clean));
        clean_name = [name '_clean'];
        assignin('base', clean_name, x_clean);
    end
```

Figure 3; code section detailing the loop for filtering and renaming each audio file.

Then the magnitude of each sound file is isolated. This was selected as the best aspect of each signal that can be quantified and compared to create the required distinction between each voice. A lot of time was spent attempting to find fundamental frequencies for each speaker, but this wasn't found to be constructive.

```
%%%%%%%%%%% grabs magnitude. we can't do voice recognition on
%%%%%%%% frequency alone unfortunately. I spent a significant
%%%%%%%%%%%% amount of time concluding this.
features = struct():
vars = whos;
for i = 1:length(vars)
   name = vars(i).name;
   if startsWith(name, 'x_') && endsWith(name, '_clean') %%%%% grabs the cleaned up ones and doesn't mess up golden copies
       x = eval(name);
       N = 2^nextpow2(length(x));
       X = abs(fft(x, N));
       X = X(1:floor(N/2));
       X = X / max(X);
       features.(name) = X;
   end
end
```

Figure 4; code section for finding average magnitudes of each cleaned file.

#### Algorithm

The algorithm takes a struct comprised of each cleaned nominal audio files (named "profiles" for authorized speakers) and begins to parse it in two nested loops. The outer loop determines which speaker is being averaged, and the inner loop averages all amplitudes of all five recordings of that one speaker. After the average is found for each nominal, this is when the main zero padding begins, ensuring that each file is the same length.

```
%%%%%%% so next we want to try averaging the cleaned recordings together.
%%%%%%%% This took a bit longer specifically on finding a way to traverse
%%%%%%%% everything.
speakers = {'gabe', 'gracelyn', 'laura', 'lydia', 'mark', 'micah', 'sana',...
'ben', 'blessing', 'dj'};
speaker_profiles = struct();
for i = 1:length(speakers)
    spk = speakers{i};
   fft_list = [];
   max_len = 0;
   temp_ffts = {};
    for j = 1:5
        key = sprintf('x_%s_%d_clean', spk, j);
        if isfield(features, key)
          temp = features.(key);
           max_len = max(max_len, length(temp));
           temp_ffts{end+1} = temp;
    \ensuremath{\%\%\%\%\%\%\%\%} had to pad with zeros and the reason for this is that
    %%%%%%%%% all the audio lengths are different from one another
    for k = 1:length(temp_ffts)
        vec = temp_ffts\{k\}(:)';
                                  %%%%%%%% matrices werent matching
       padded = [vec, zeros(1, max_len - length(vec))];
       fft_list(end+1, :) = padded; % Now dimensions match
    speaker_profiles.(spk) = mean(fft_list, 1);
```

Figure 5; code snapshot of the main sound file averaging loop as well as the zero padding.

From this point on in the code, the program waits for input in the console for any specific audio file to be named to be processed. Once a specific audio file is typed in correctly, the distance between the maximum and minimum amplitude in the audio file is determined and compared to the averaged data. Using a carefully calculated threshold of **8.5**, the voice is identified and matched as either a nominal voice or an intruder. The threshold was refined and adjusted based on detection of profiles with the test console output later listed in this report.

```
test_fft = features.(test_name);
    % Pad test_fft if needed with zeros
    test_fft = [test_fft(:)', zeros(1, max_len - length(test_fft))];

min_dist = inf;
    best_match = '';
    speaker_names = fieldnames(speaker_profiles);]

for i = 1:length(speaker_names)
    spk = speaker_nemes{1};
    profile = speaker_profiles.(spk);

    ** Padding with zeros again
    profile = [profile(:)', zeros(1, max_len - length(profile))];

%%%%%%%%%%%% this is very important we are using this value to calculate amplitude distance. Very important.
    dist = norm(test_fft - profile);
    if dist < min_dist
        min_dist = dist;
        best_match = spk;
    end
end

%* Threshold check
threshold = 8.5;

if min_dist > threshold
    fprintf('DeBUG - Closest Match: %s\n', best_match);
    % fprintf('DeBUG - Distance to Match: %.4f\n', min_dist);

else
    fprintf('Authorized. You may enter \n\n');
    % fprintf('DEBUG - Distance to Match: %.4f\n', min_dist);
    end
end
end
```

Figure 6; code snapshot of the main comparison section of the algorithm.

The only other addition to the code is a debug feature that does the same action as the section described but applies it to every single sound file within the directory, laying out a large spread of results for every sound file. This was extremely important in debugging and specifically refining an appropriate threshold value.

#### Results

Utilizing the algorithm's capability to display all sound files and their detection as authorized or rejected, the following data spread is displayed:

Intended intruder voices are highlighted in **orange**, and incorrect deductions are highlighted in **red**.

| Cleaned Samples vs Speaker Profiles |                                        |                                                                      |                    |                         |                              |
|-------------------------------------|----------------------------------------|----------------------------------------------------------------------|--------------------|-------------------------|------------------------------|
| x_ben_1_clean                       | Closest Match: ben                     | amplitude: 7.2184 AUTHORIZED                                         | x_laura_1_clean    | Closest Match: laura    | amplitude: 8.2477 AUTHORIZED |
| x_ben_2_clean                       | Closest Match: ben                     | amplitude: 9.4178 INTRUDER                                           | x_laura_2_clean    | Closest Match: laura    | amplitude: 6.8980 AUTHORIZED |
| x_ben_3_clean                       | Closest Match: ben                     | amplitude: 7.0945 AUTHORIZED                                         | x_laura_3_clean    | Closest Match: laura    | amplitude: 5.6323 AUTHORIZED |
| x_ben_4_clean                       | Closest Match: ben                     | amplitude: 8.1722 AUTHORIZED                                         | x_laura_4_clean    | Closest Match: laura    | amplitude: 6.4889 AUTHORIZED |
| x_ben_5_clean                       | Closest Match: ben                     | amplitude: 7.1511 AUTHORIZED                                         | x laura 5 clean    | Closest Match: laura    | amplitude: 6.6911 AUTHORIZED |
| x_blessing_1_clean                  | Closest Match: blessing                | amplitude: 6.2776 AUTHORIZED                                         | x lydia 1 clean    | Closest Match: lydia    | amplitude: 6.7995 AUTHORIZED |
| x_blessing_2_clean                  | Closest Match: blessing                | amplitude: 5.3341 AUTHORIZED                                         | x lydia 2 clean    | Closest Match: lydia    | amplitude: 6.4435 AUTHORIZED |
| x_blessing_3_clean                  | Closest Match: blessing                | amplitude: 4.7100 AUTHORIZED                                         |                    | Closest Match: lydia    | amplitude: 5.3993 AUTHORIZED |
| x_blessing_4_clean                  | Closest Match: blessing                | amplitude: 5.5218 AUTHORIZED                                         | x_lydia_3_clean    |                         |                              |
| x_blessing_5_clean                  | Closest Match: blessing                | amplitude: 7.3058 AUTHORIZED                                         | x_lydia_4_clean    | Closest Match: blessing | amplitude: 6.7833 AUTHORIZED |
| x_clean                             | Closest Match: sana                    | amplitude: 12.7774 INTRUDER                                          | x_lydia_5_clean    | Closest Match: lydia    | amplitude: 6.5367 AUTHORIZED |
| x_crawley_1_clean                   | Closest Match: gabe                    | amplitude: 12.4495 INTRUDER                                          | x_mark_1_clean     | Closest Match: mark     | amplitude: 7.3618 AUTHORIZED |
| x_crawley_2_clean                   | Closest Match: ben                     | amplitude: 11.5170 INTRUDER                                          | x_mark_2_clean     | Closest Match: mark     | amplitude: 5.9912 AUTHORIZED |
| x_crawley_3_clean                   | Closest Match: gracelyn                | amplitude: 10.9750 INTRUDER                                          | x_mark_3_clean     | Closest Match: mark     | amplitude: 5.2363 AUTHORIZED |
| x_dj_1_clean                        | Closest Match: dj                      | amplitude: 6.5266 AUTHORIZED                                         | x mark 4 clean     | Closest Match: mark     | amplitude: 6.0018 AUTHORIZED |
| x_dj_2_clean                        | Closest Match: dj                      | amplitude: 6.7710 AUTHORIZED                                         | x_mark_5_clean     | Closest Match: mark     | amplitude: 5.8236 AUTHORIZED |
| x_dj_3_clean<br>x di 4 clean        | Closest Match: dj<br>Closest Match: di | <pre>amplitude: 7.3641 AUTHORIZED amplitude: 8.4760 AUTHORIZED</pre> | x micah 1 clean    | Closest Match: micah    | amplitude: 9.0790 INTRUDER   |
| x_dj_4_clean<br>x di 5 clean        | Closest Match: dj                      | amplitude: 6.7251 AUTHORIZED                                         | x micah 2 clean    | Closest Match: micah    | amplitude: 6.3475 AUTHORIZED |
| x_dj_5_clean<br>x_gabe_1_clean      | Closest Match: dj                      | amplitude: 7.7639 AUTHORIZED                                         |                    | Closest Match: micah    | amplitude: 7.9582 AUTHORIZED |
| x_gabe_i_clean<br>x gabe 2 clean    | Closest Match: gabe                    | amplitude: 7.7639 AUTHORIZED                                         | x_micah_3_clean    |                         |                              |
| x gabe 3 clean                      | Closest Match: gabe                    | amplitude: 6.4707 AUTHORIZED                                         | x_micah_4_clean    | Closest Match: micah    | amplitude: 7.8924 AUTHORIZED |
| x_gabe_4_clean                      | Closest Match: gabe                    | amplitude: 7.3344 AUTHORIZED                                         | x_micah_5_clean    | Closest Match: micah    | amplitude: 8.8611 INTRUDER   |
| x gabe 5 clean                      | Closest Match: gabe                    | amplitude: 7.9451 AUTHORIZED                                         | x_sana_1_clean     | Closest Match: sana     | amplitude: 5.0143 AUTHORIZED |
| x gracelyn 1 clean                  | Closest Match: gracelyn                | amplitude: 4.0856 AUTHORIZED                                         | x_sana_2_clean     | Closest Match: sana     | amplitude: 4.2468 AUTHORIZED |
| x gracelyn 2 clean                  | Closest Match: gracelyn                | amplitude: 5.7407 AUTHORIZED                                         | x_sana_3_clean     | Closest Match: sana     | amplitude: 4.9676 AUTHORIZED |
| x gracelyn 3 clean                  | Closest Match: gracelyn                | amplitude: 4.0672 AUTHORIZED                                         | x sana 4 clean     | Closest Match: sana     | amplitude: 4.0118 AUTHORIZED |
| x_gracelyn_4_clean                  | Closest Match: gracelyn                | amplitude: 3.8747 AUTHORIZED                                         |                    | Closest Match: sana     | amplitude: 5.6802 AUTHORIZED |
| x gracelyn 5 clean                  | Closest Match: gracelyn                | amplitude: 5.2683 AUTHORIZED                                         | x silvario 1 clean | Closest Match: blessing | amplitude: 7.6978 AUTHORIZED |
| x holden 1 clean                    | Closest Match: ben                     | amplitude: 21.3099 INTRUDER                                          | x silvario 2 clean | Closest Match: mark     | amplitude: 10.8000 INTRUDER  |
| x_holden_2_clean                    | Closest Match: ben                     | amplitude: 23.2154 INTRUDER                                          | x silvario 3 clean | Closest Match: micah    | amplitude: 8.6486 INTRUDER   |
| x_holden_3_clean                    | Closest Match: ben                     | amplitude: 18.9845 INTRUDER                                          |                    |                         | amplitude: 20.0724 INTRUDER  |
| x_kris_1_clean                      | Closest Match: dj                      | amplitude: 11.9215 INTRUDER                                          | x_tim_1_clean      | Closest Match: dj       |                              |
| x_kris_2_clean                      | Closest Match: dj                      | amplitude: 10.7851 INTRUDER                                          | x_tim_2_clean      | Closest Match: dj       | amplitude: 20.6512 INTRUDER  |
| x kris 3 clean                      | Closest Match: di                      | amplitude: 12.5576 INTRUDER                                          | x_tim_3_clean      | Closest Match: sana     | amplitude: 12.7774 INTRUDER  |

Figure 7; direct console output of the algorithm, with color coded annotations.

With all 65 samples recorded, cleaned, and processed, the total percentage of success for the chosen comparison threshold is 95% for recognizing nominal voices, and 93% for recognizing intruders.

In retrospect as of the writing of this report, it would have been better to heed the initial directive which was to **first use 3 out of 5 of the voice recordings for the profile, and then use 2 additional recordings to determine if people were authorized or not.** This would have reinforced the concept of a 'learning' based model and not a 'profile' based model that was achieved. This could have further reduced the false negatives and false positives that were encountered with data. This would have drastically affected the validation process. Given additional time, this would have been the one significant change to implement that would have likely increased the accuracy of detection for both nominal voices and intruders.

### Conclusion

The lab served as a culmination of the foundation of signals lab over the past four months. The application of an FIR filter was significant choice in removing background noise as the majority of other groups after presentation day didn't even factor this into their pipeline. Running FFTs across the different recordings helped in determine profiles for authorized speakers. Achieving a high but imperfect recognition rate of both nominal and intruders was a very surprising outcome for this lab.

# Appendix final.m

%%%%%%%%%% " the quick brown fox jumps over the lazy dog %%%%%%%%%%%%%% so far we have 95% accuracy rating %%%%%%% first load all audio [x\_gabe\_1, fs] = audioread('gabe\_1.wav'); [x\_gabe\_2, ~] = audioread('gabe\_2.wav'); [x\_gabe\_3, ~] = audioread('gabe\_3.wav'); [x\_gabe\_4, ~] = audioread('gabe\_4.wav'); [x\_gabe\_5, ~] = audioread('gabe\_5.wav'); [x gracelyn 1, ~] = audioread('gracelyn 1.wav'); [x gracelyn 2, ~] = audioread('gracelyn 2.wav'); [x\_gracelyn\_3, ~] = audioread('gracelyn\_3.wav'); [x\_gracelyn\_4, ~] = audioread('gracelyn\_4.wav'); [x\_gracelyn\_5, ~] = audioread('gracelyn\_5.wav'); [x laura 1, ~] = audioread('laura 1.wav'); [x\_laura\_2, ~] = audioread('laura\_2.wav'); [x\_laura\_3, ~] = audioread('laura\_3.wav'); [x laura 4, ~] = audioread('laura 4.wav'); [x laura 5, ~] = audioread('laura 5.wav'); [x\_lydia\_1, ~] = audioread('lydia\_1.wav'); [x\_lydia\_2, ~] = audioread('lydia\_2.wav'); [x lydia 3, ~] = audioread('lydia 3.wav'); [x\_lydia\_4, ~] = audioread('lydia\_4.wav'); [x lydia 5, ~] = audioread('lydia 5.wav'); [x\_mark\_1, ~] = audioread('mark\_1.wav'); [x mark 2, ~] = audioread('mark 2.wav'); [x\_mark\_3, ~] = audioread('mark\_3.wav'); [x\_mark\_4, ~] = audioread('mark\_4.wav'); [x mark 5, ~] = audioread('mark 5.wav'); [x\_micah\_1, ~] = audioread('micah 1.wav'); [x\_micah\_2, ~] = audioread('micah\_2.wav'); [x\_micah\_3, ~] = audioread('micah\_3.wav'); [x\_micah\_4, ~] = audioread('micah\_4.wav'); [x micah 5, ~] = audioread('micah 5.wav'); [x sana 1, ~] = audioread('sana 1.wav'); [x\_sana\_2, ~] = audioread('sana\_2.wav'); [x\_sana\_3, ~] = audioread('sana 3.wav');

[x\_sana\_4, ~] = audioread('sana\_4.wav'); [x\_sana\_5, ~] = audioread('sana\_5.wav');

[x\_ben\_1, ~] = audioread('ben\_1.wav'); [x\_ben\_2, ~] = audioread('ben\_2.wav'); [x ben 3, ~] = audioread('ben 3.wav');

```
[x_ben_4, ~] = audioread('ben_4.wav');
[x_ben_5, ~] = audioread('ben 5.wav');
[x blessing 1, ~] = audioread('blessing 1.wav');
[x_blessing_2, ~] = audioread('blessing_2.wav');
[x_blessing_3, ~] = audioread('blessing_3.wav');
[x_blessing_4, ~] = audioread('blessing_4.wav');
[x_blessing_5, ~] = audioread('blessing_5.wav');
[x dj 1, ~] = audioread('dj 1.wav');
[x_dj_2, \sim] = audioread('dj_2.wav');
[x_dj_3, \sim] = audioread('dj_3.wav');
[x_dj_4, \sim] = audioread('dj_4.wav');
[x_dj_5, \sim] = audioread('dj_5.wav');
%%%%%%%% intruder audio. we need to reject these.
[x_silvario_1, ~] = audioread('silvario 1.wav');
[x_silvario_2, ~] = audioread('silvario_2.wav');
[x_silvario_3, ~] = audioread('silvario_3.wav');
[x_crawley_1, ~] = audioread('crawley_1.wav');
[x_crawley_2, ~] = audioread('crawley_2.wav');
[x crawley 3, ~] = audioread('crawley 3.wav');
[x holden 1, ~] = audioread('holden 1.wav');
[x_holden_2, ~] = audioread('holden_2.wav');
[x_holden_3, ~] = audioread('holden_3.wav');
[x_tim_1, ~] = audioread('tim_1.wav');
[x_tim_2, ~] = audioread('tim_2.wav');
[x tim 3, ~] = audioread('tim 3.wav');
[x kris 1, ~] = audioread('kris 1.wav');
[x_kris_2, ~] = audioread('kris_2.wav');
[x kris 3, ~] = audioread('kris 3.wav');
[noise_background, ~] = audioread('empty_room.wav');
%%%%%%%% batch process the matrixes from stereo to mono. all the matrices
%%%%%%%% are in a stereo format and mono closely represents how we
%%%%%%% physically speak. ie, we are not speaking from two mouths
vars = whos;
for i = 1:length(vars)
    name = vars(i).name;
    % all initial matrixes basically will be grabbed. that's why we went
    % with the naming convention
    if startsWith(name, 'x_')
        data = eval(name);
        if ismatrix(data) && size(data, 2) == 2
            mono = mean(data, 2);
```

```
assignin('base', name, mono);
        end
   end
end
%%%%%%%% background noise to mono for our first filter
if size(noise_background, 2) == 2
    noise_background = mean(noise_background, 2);
noise background = noise background / max(abs(noise background));
%%%%%%%% filter background noise.
N = length(noise_background);
Y = abs(fft(noise_background));
Y = Y(1:floor(N/2));
f = linspace(0, fs/2, length(Y));
[\sim, idx] = max(Y);
                           % we have 28.37hz
f_{noise} = f(idx);
%%%%%%%%% attempting to run an fir filter with the matlab function
%%%%%%%% fir1.
%%%%%%%%% https://www.mathworks.com/help/signal/ref/fir1.html
bandwidth = 3; %%%%%%%% parameter for a tolerance
f_low = (f_noise - bandwidth/2) / (fs/2);
f_high = (f_noise + bandwidth/2) / (fs/2);
f_{low} = max(f_{low}, 0);
f_high = min(f_high, 1);
filter order = 128;
                                 %%%%%%%% amount of coefficients is actually
filter_order+1. so 129
b = fir1(filter_order, [f_low f_high], 'stop');
%%%%%%%%% clean everything now with FIR using another loop similar to
%%%%%%%% earlier one.
vars = whos;
for i = 1:length(vars)
   name = vars(i).name;
    if startsWith(name, 'x_') && ~contains(name, '_clean')
       x = eval(name);
        if size(x, 2) == 2
            x = mean(x, 2);
        end
        x = x(:);
        x = x / max(abs(x));
```

```
x_clean = filter(b, 1, x);
       x_clean = x_clean / max(abs(x_clean));
       clean_name = [name '_clean'];
       assignin('base', clean_name, x_clean);
   end
end
%%%%%%%%%%%%%% grabs magnitude. we can't do voice recognition on
%%%%%%%%%%%%% frequency alone unfortunately. I spent a significant
%%%%%%%%%%%% amount of time concluding this.
features = struct();
vars = whos;
for i = 1:length(vars)
   name = vars(i).name;
   if startsWith(name, 'x_') && endsWith(name, '_clean') %%%%% grabs the cleaned up
ones and doesn't mess up golden copies
       x = eval(name);
       N = 2^nextpow2(length(x));
       X = abs(fft(x, N));
       X = X(1:floor(N/2));
       X = X / max(X);
       features.(name) = X;
   end
end
%%%%%%%% so next we want to try averaging the cleaned recordings together.
%%%%%%%%% This took a bit longer specifically on finding a way to traverse
%%%%%%% everything.
speakers = {'gabe', 'gracelyn', 'laura', 'lydia', 'mark', 'micah', 'sana',...
    'ben', 'blessing', 'dj'};
speaker_profiles = struct();
for i = 1:length(speakers)
   spk = speakers{i};
   fft_list = [];
   max_len = 0;
   temp_ffts = {};
   for j = 1:5
       key = sprintf('x_%s_%d_clean', spk, j);
       if isfield(features, key)
           temp = features.(key);
           max_len = max(max_len, length(temp));
           temp_ffts{end+1} = temp;
       end
   end
```

```
%%%%%%%%% had to pad with zeros and the reason for this is that
   %%%%%%%% all the audio lengths are different from one another
   for k = 1:length(temp ffts)
       vec = temp_ffts{k}(:)';
                                  %%%%%% matrices werent matching
       padded = [vec, zeros(1, max_len - length(vec))];
       fft_list(end+1, :) = padded; % Now dimensions match
   end
   speaker profiles.(spk) = mean(fft list, 1);
end
fprintf('\n Password required! \n');
while true
   test name = input('Enter cleaned test sample name: ', 's');
   if strcmpi(test_name, 'exit')
       disp('exiting');
       break;
   end
   if ~isfield(features, test name)
       fprintf(' audio recording "%s" not found. \n\n', test name);
       continue;
   end
   test_fft = features.(test_name);
   % Pad test fft if needed with zeros
   test_fft = [test_fft(:)', zeros(1, max_len - length(test_fft))];
   min_dist = inf;
   best_match = '';
   speaker_names = fieldnames(speaker_profiles);
   for i = 1:length(speaker names)
       spk = speaker names{i};
       profile = speaker profiles.(spk);
       % Padding with zeros again
       profile = [profile(:)', zeros(1, max_len - length(profile))];
%%%%%%%% this is very important we are using this value to calculate amplitude
distance. Very important.
       dist = norm(test fft - profile);
       if dist < min_dist</pre>
           min_dist = dist;
           best_match = spk;
       end
   end
```

```
% Threshold check
   threshold = 8.5;
   if min_dist > threshold
       fprintf(' Not Authorized \n\n');
        fprintf('DEBUG - Closest Match: %s\n', best_match);
   %
        fprintf('DEBUG - Distance to Match: %.4f\n', min_dist);
   else
       fprintf(' Authorized. You may enter \n\n');
        fprintf('DEBUG - Closest Match: %s\n', best_match);
        fprintf('DEBUG - Distance to Match: %.4f\n', min dist);
   end
end
%%%%%%%%%%%%%%%%%%%%%%%%%%%%%%% this was used for debugging and not
fprintf('\n Cleaned Samples vs Speaker Profiles\n');
sample names = fieldnames(features);
profile_names = fieldnames(speaker_profiles);
%max_len = max([structfun(@length, speaker_profiles), structfun(@length, features)]);
all_lengths = [structfun(@length, speaker_profiles); structfun(@length, features)];
max_len = max(all_lengths);
%max_len = max(structfun(@length, speaker_profiles));
threshold = 8.5; %%%%%%%%%% this was tweaked until we have our confidence range.
for i = 1:length(sample names)
   sample = sample_names{i};
   test fft = features.(sample);
   test_fft = [test_fft(:)', zeros(1, max_len - length(test_fft))];
   min dist = inf;
   best match = '';
   for j = 1:length(profile names)
       profile = speaker profiles.(profile names{j});
       profile = [profile(:)', zeros(1, max_len - length(profile))];
       dist = norm(test_fft - profile);
       if dist < min dist</pre>
           min dist = dist;
           best_match = profile_names{j};
       end
   end
   if min_dist > threshold
       result = 'INTRUDER';
   else
```

```
result = 'AUTHORIZED';
end

fprintf('%-25s Closest Match: %-10s amplitude: %.4f %s\n', sample, best_match,
min_dist, result);
end
```