UCS749

Speech Recognition Report
Lab Eval - 1

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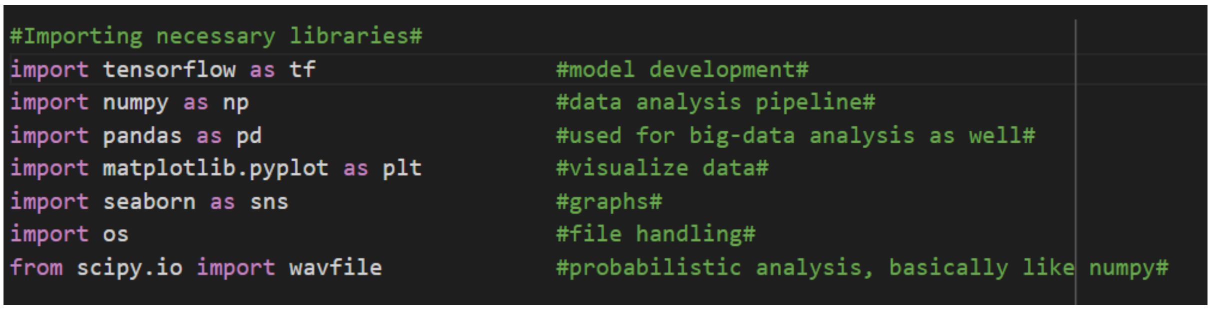
1. Read and summarise the paper in about 50 words.

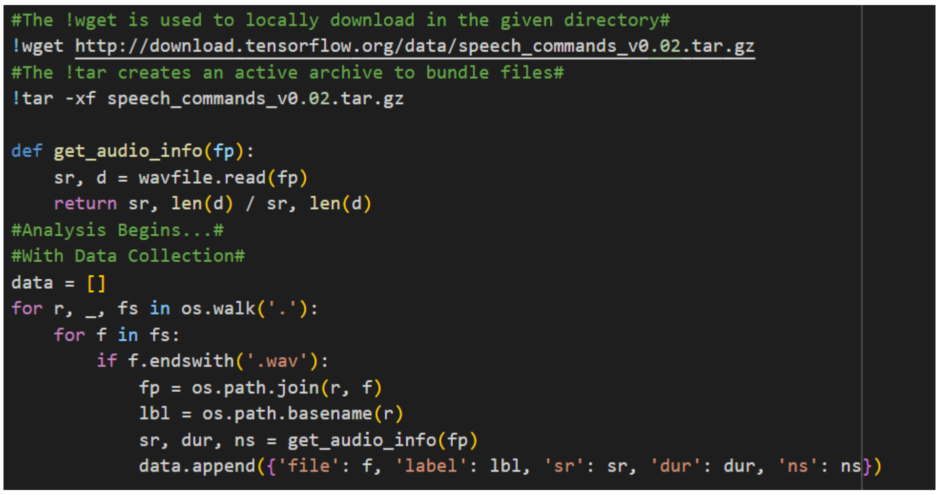
   The paper describes the Speech Commands dataset, designed for training and evaluating keyword spotting systems. It contains over 100,000 utterances of 35 words from 2,618 speakers. The dataset aims to enable comparable accuracy metrics and reproducible results for on-device speech recognition models, particularly for trigger word detection.

2. Download the dataset in the paper, statistically analyse and describe it, so that it may be useful for posterity. (Include code snippets in your . ipynb file to evidence your analysis.)

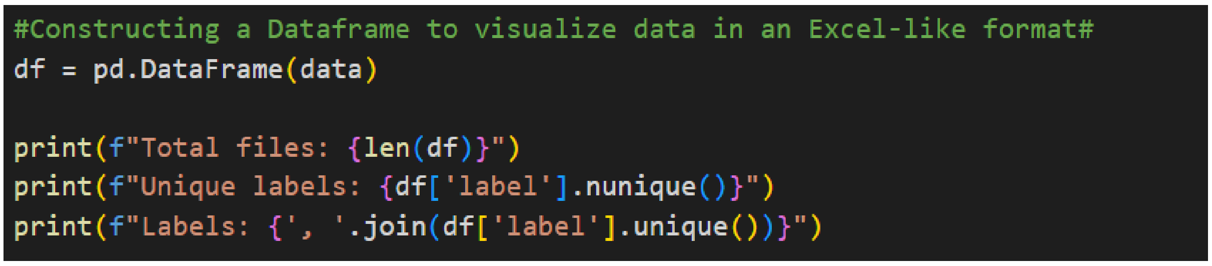
The snippets are provided here:



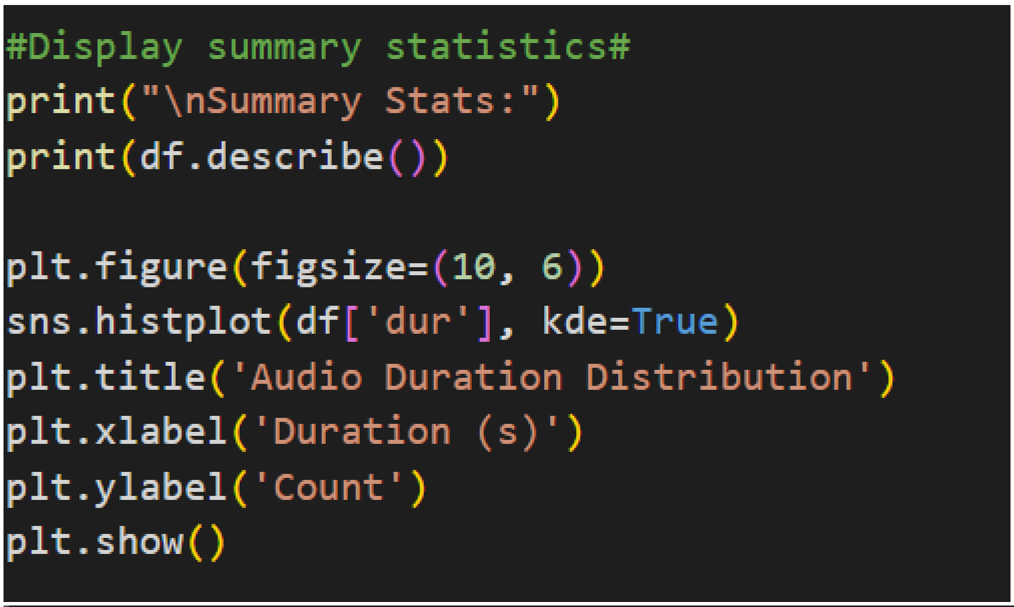
```
#Importing necessary libraries#
import tensorflow as tf                   #model development#
import numpy as np                        #data analysis pipeline#
import pandas as pd                       #used for big-data analysis as well#
import matplotlib.pyplot as plt           #visualize data#
import seaborn as sns                     #graphs#
import os                                 #file handling#
from scipy.io import wavfile              #probabilistic analysis, basically like numpy#
```



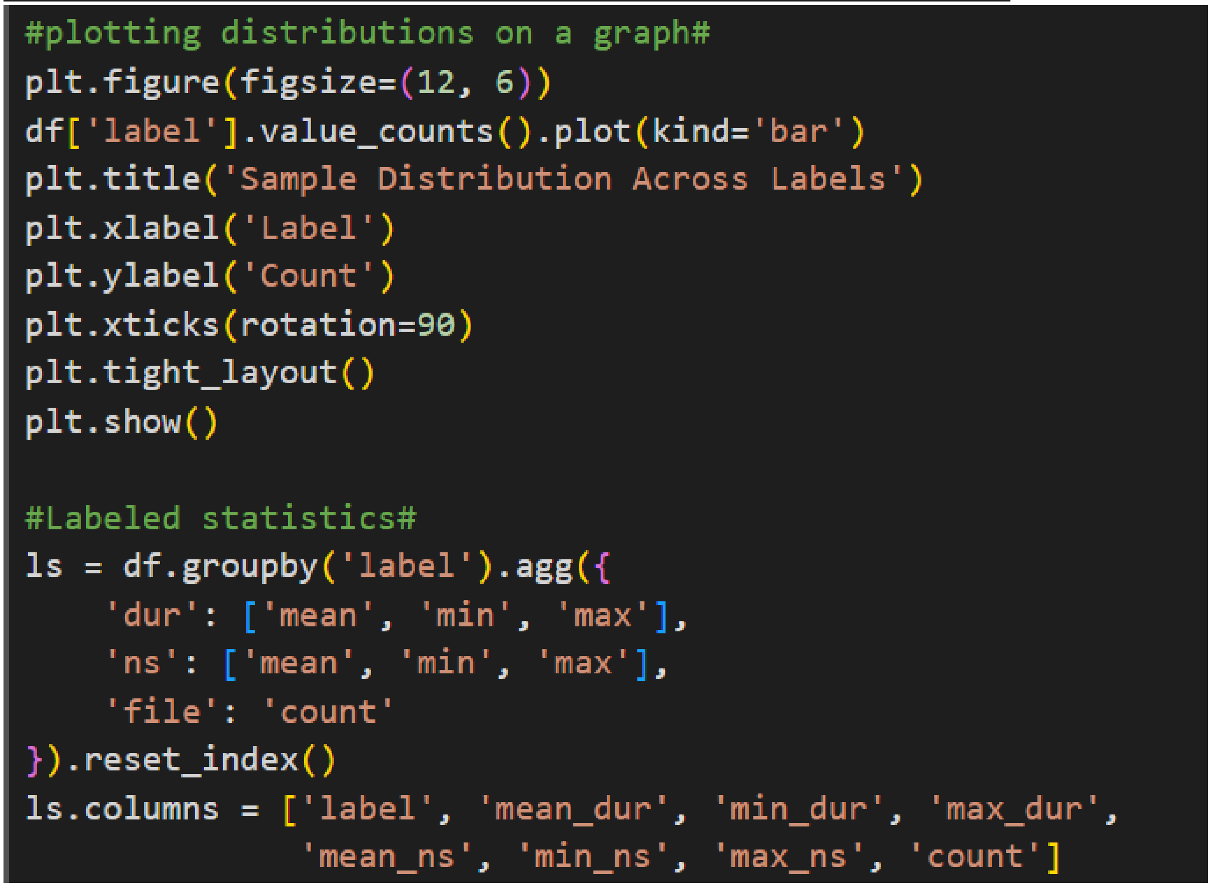
```
#The !wget is used to locally download in the given directory#
!wget http://download.tensorflow.org/data/speech_commands_v0.02.tar.gz
#The !tar creates an active archive to bundle files#
!tar -xf speech_commands_v0.02.tar.gz

def get_audio_info(fp):
    sr, d = wavfile.read(fp)
    return sr, len(d) / sr, len(d)
#Analysis Begins...#
#With Data Collection#
data = []
for r, _, fs in os.walk('.'):
    for f in fs:
        if f.endswith('.wav'):
            fp = os.path.join(r, f)
            lbl = os.path.basename(r)
            sr, dur, ns = get_audio_info(fp)
            data.append({'file': f, 'label': lbl, 'sr': sr, 'dur': dur, 'ns': ns})
```



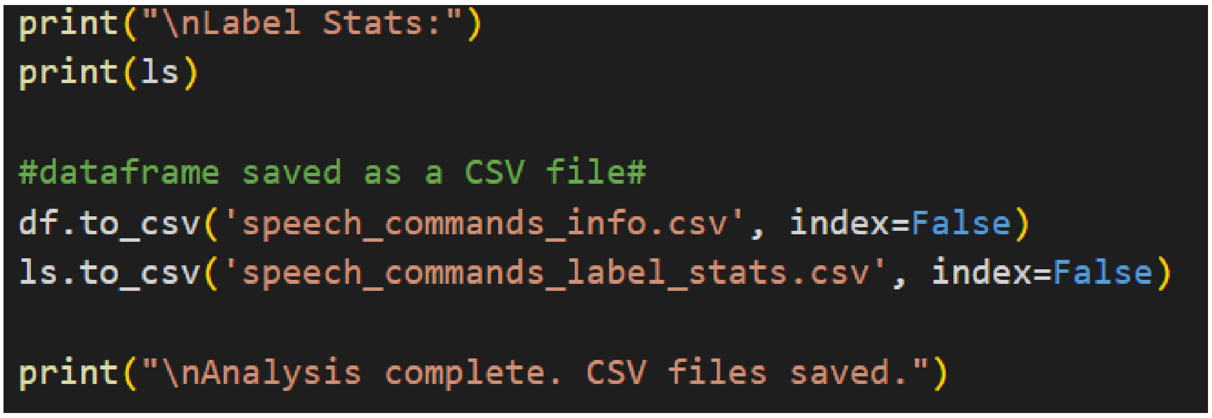
```python
#Constructing a Dataframe to visualize data in an Excel-like format#
df = pd.DataFrame(data)

print(f"Total files: {len(df)}")
print(f"Unique labels: {df['label'].nunique()}")
print(f"Labels: {', '.join(df['label'].unique())}")
```



```python
#Display summary statistics#
print("\nSummary Stats:")
print(df.describe())

plt.figure(figsize=(10, 6))
sns.histplot(df['dur'], kde=True)
plt.title('Audio Duration Distribution')
plt.xlabel('Duration (s)')
plt.ylabel('Count')
plt.show()
```

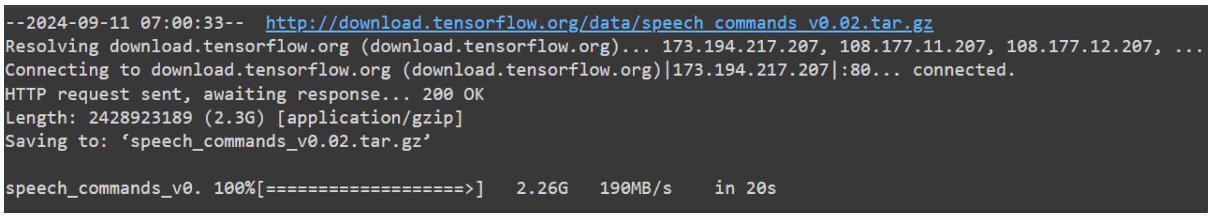


```python
#plotting distributions on a graph#
plt.figure(figsize=(12, 6))
df['label'].value_counts().plot(kind='bar')
plt.title('Sample Distribution Across Labels')
plt.xlabel('Label')
plt.ylabel('Count')
plt.xticks(rotation=90)
plt.tight_layout()
plt.show()

#Labeled statistics#
ls = df.groupby('label').agg({
    'dur': ['mean', 'min', 'max'],
    'ns': ['mean', 'min', 'max'],
    'file': 'count'
}).reset_index()
ls.columns = ['label', 'mean_dur', 'min_dur', 'max_dur',
              'mean_ns', 'min_ns', 'max_ns', 'count']
```

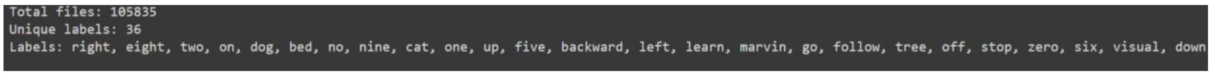


```python
print("\nLabel Stats:")
print(ls)

#dataframe saved as a CSV file#
df.to_csv('speech_commands_info.csv', index=False)
ls.to_csv('speech_commands_label_stats.csv', index=False)

print("\nAnalysis complete. CSV files saved.")
```

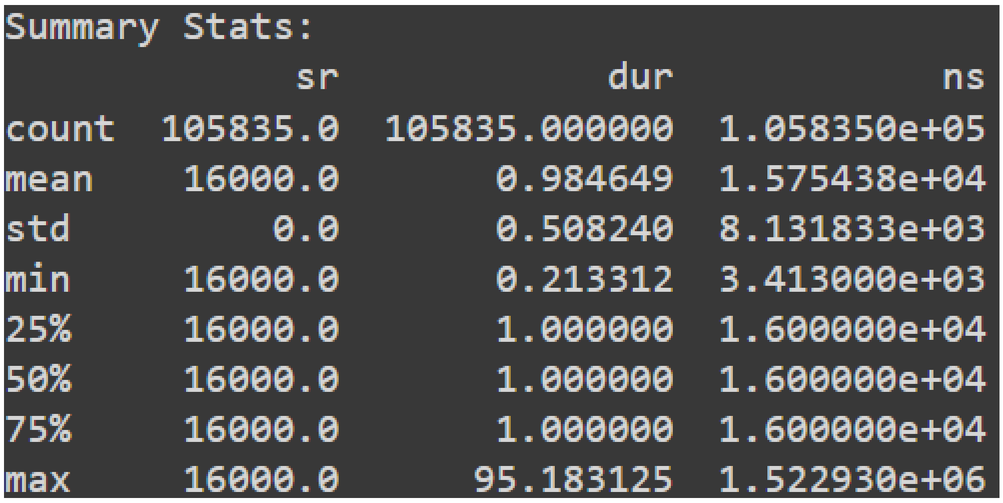
Output:



```
--2024-09-11 07:00:33--  http://download.tensorflow.org/data/speech_commands_v0.02.tar.gz
Resolving download.tensorflow.org (download.tensorflow.org)... 173.194.217.207, 108.177.11.207, 108.177.12.207, ...
Connecting to download.tensorflow.org (download.tensorflow.org)|173.194.217.207|:80... connected.
HTTP request sent, awaiting response... 200 OK
Length: 2428923189 (2.3G) [application/gzip]
Saving to: ‘speech_commands_v0.02.tar.gz’

speech_commands_v0. 100%[===================>]   2.26G   190MB/s    in 20s
```

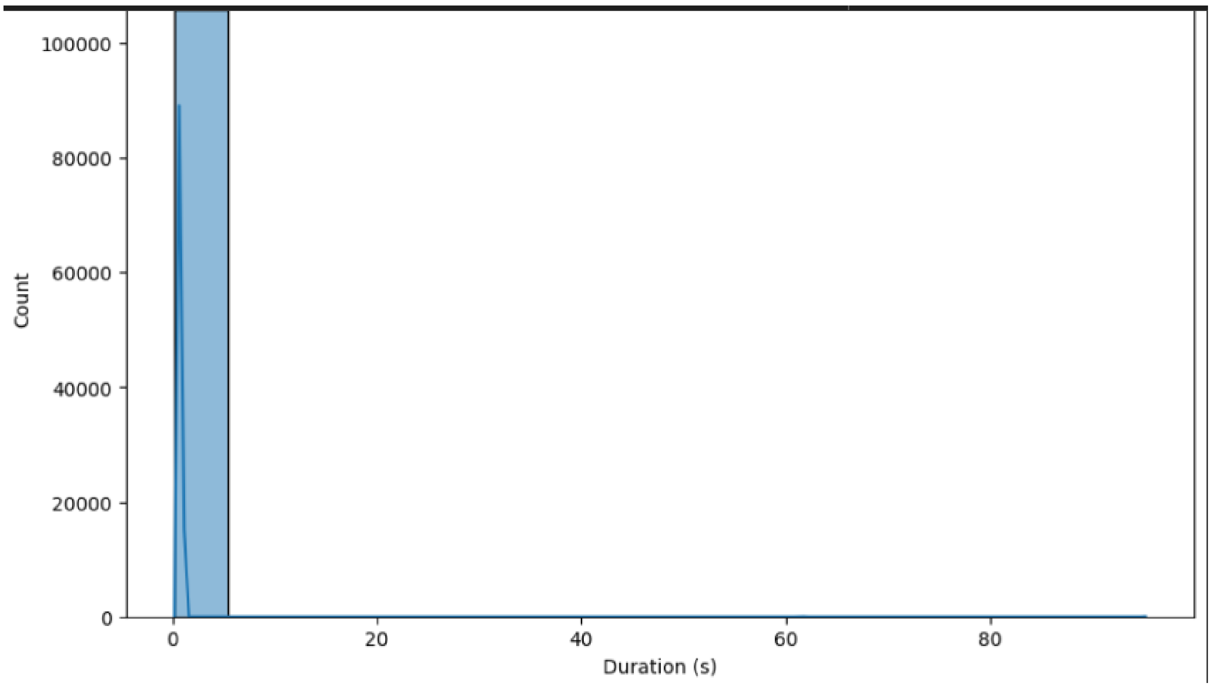


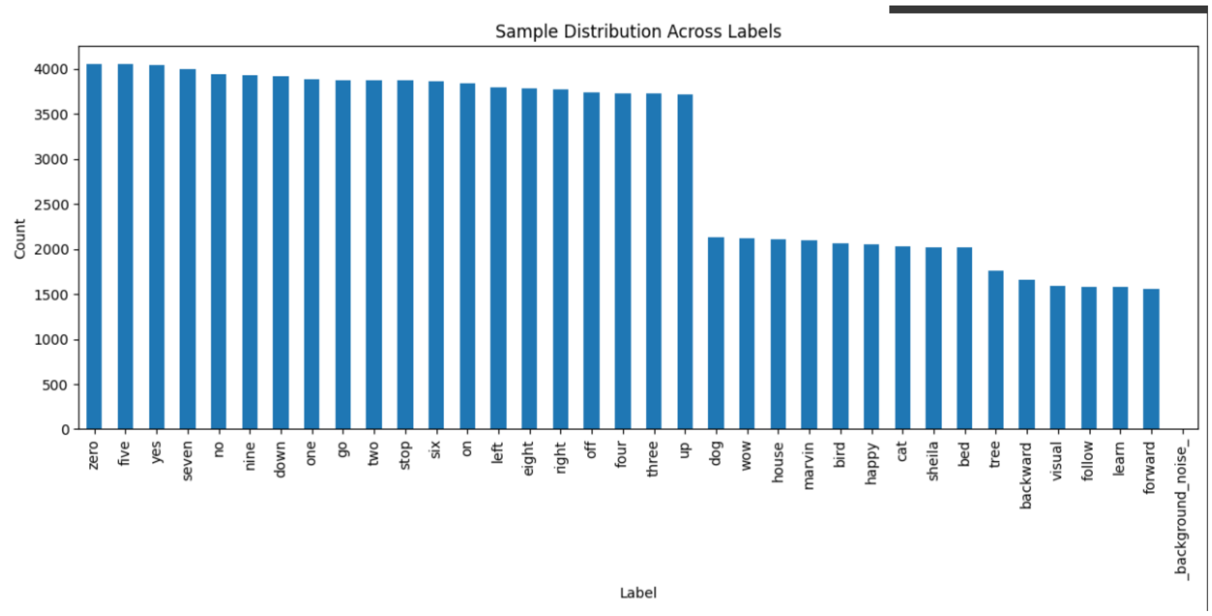
```
Total files: 105835
Unique labels: 36
Labels: right, eight, two, on, dog, bed, no, nine, cat, one, up, five, backward, left, learn, marvin, go, follow, tree, off, stop, zero, six, visual, down
```

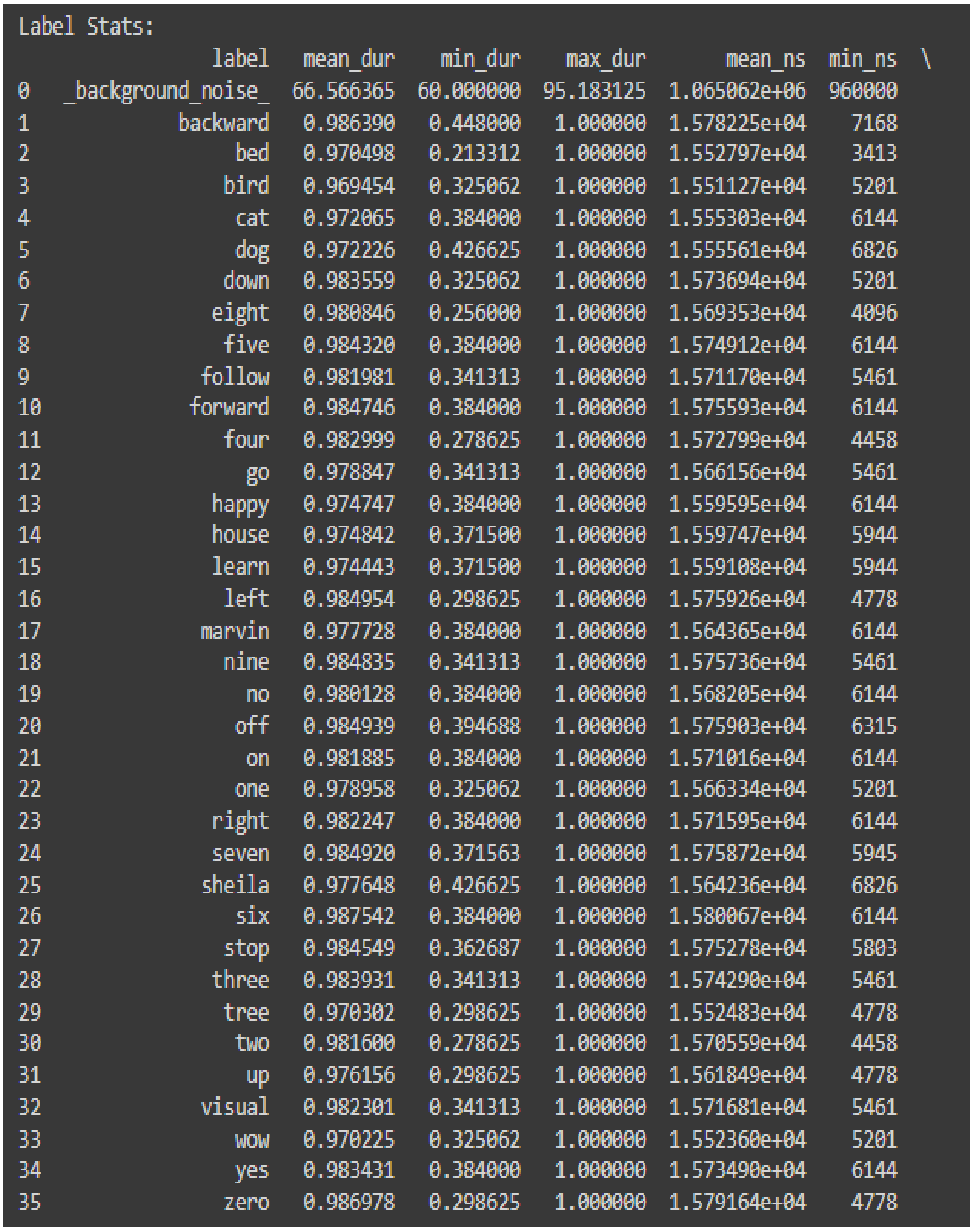


```
Summary Stats:
             sr            dur            ns
count  105835.0  105835.000000  1.058350e+05
mean    16000.0       0.984649  1.575438e+04
std         0.0       0.508240  8.131833e+03
min     16000.0       0.213312  3.413000e+03
25%     16000.0       1.000000  1.600000e+04
50%     16000.0       1.000000  1.600000e+04
75%     16000.0       1.000000  1.600000e+04
max     16000.0      95.183125  1.522930e+06
```

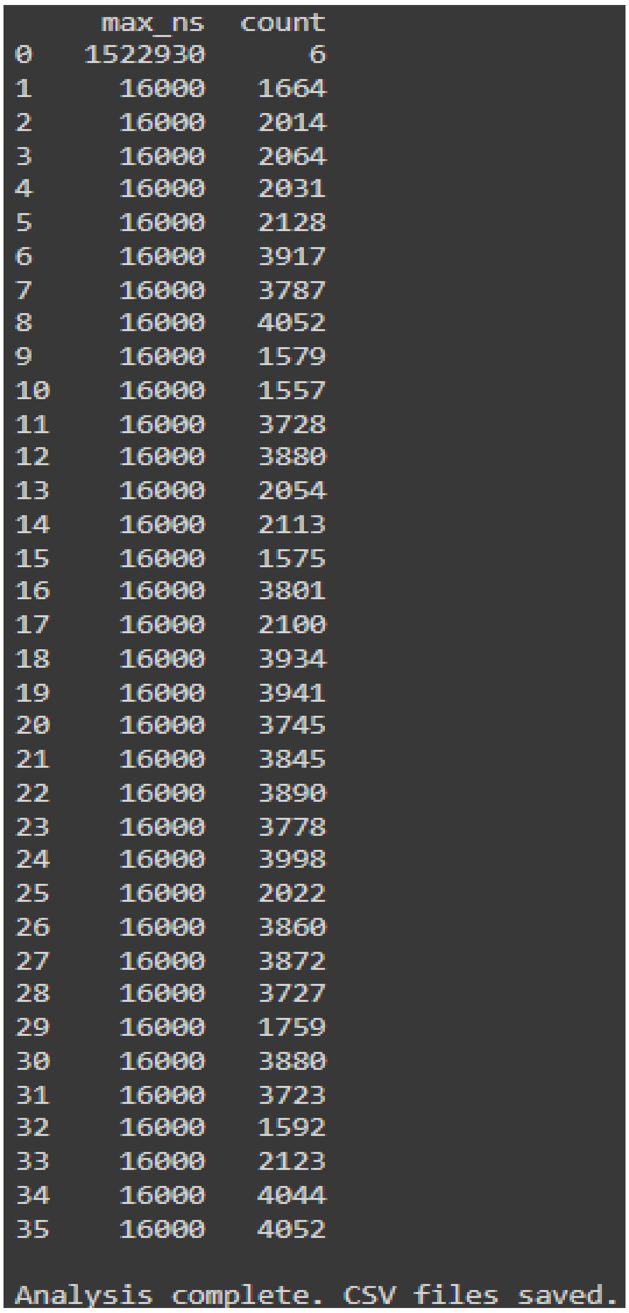
Data Visualization:





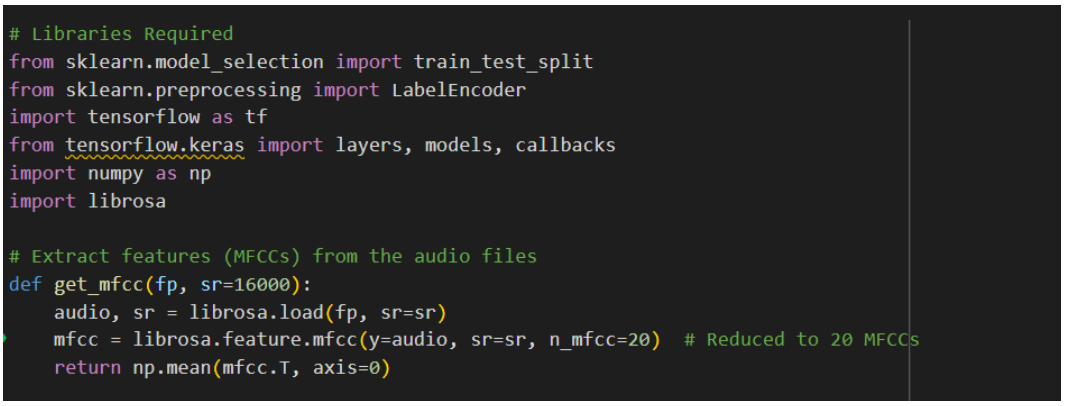


```
Label Stats:
               label   mean_dur    min_dur    max_dur       mean_ns  min_ns  \
0   _background_noise_  66.566365  60.000000  95.183125  1.065062e+06  960000
1             backward   0.986390   0.448000   1.000000  1.578225e+04    7168
2                  bed   0.970498   0.213312   1.000000  1.552797e+04    3413
3                 bird   0.969454   0.325062   1.000000  1.551127e+04    5201
4                  cat   0.972065   0.384000   1.000000  1.555303e+04    6144
5                  dog   0.972226   0.426625   1.000000  1.555561e+04    6826
6                 down   0.983559   0.325062   1.000000  1.573694e+04    5201
7                eight   0.980846   0.256000   1.000000  1.569353e+04    4096
8                 five   0.984320   0.384000   1.000000  1.574912e+04    6144
9               follow   0.981981   0.341313   1.000000  1.571170e+04    5461
10             forward   0.984746   0.384000   1.000000  1.575593e+04    6144
11                four   0.982999   0.278625   1.000000  1.572799e+04    4458
12                  go   0.978847   0.341313   1.000000  1.566156e+04    5461
13               happy   0.974747   0.384000   1.000000  1.559595e+04    6144
14               house   0.974842   0.371500   1.000000  1.559747e+04    5944
15               learn   0.974443   0.371500   1.000000  1.559108e+04    5944
16                left   0.984954   0.298625   1.000000  1.575926e+04    4778
17              marvin   0.977728   0.384000   1.000000  1.564365e+04    6144
18                nine   0.984835   0.341313   1.000000  1.575736e+04    5461
19                  no   0.980128   0.384000   1.000000  1.568205e+04    6144
20                 off   0.984939   0.394688   1.000000  1.575903e+04    6315
21                  on   0.981885   0.384000   1.000000  1.571016e+04    6144
22                 one   0.978958   0.325062   1.000000  1.566334e+04    5201
23               right   0.982247   0.384000   1.000000  1.571595e+04    6144
24               seven   0.984920   0.371563   1.000000  1.575872e+04    5945
25              sheila   0.977648   0.426625   1.000000  1.564236e+04    6826
26                 six   0.987542   0.384000   1.000000  1.580067e+04    6144
27                stop   0.984549   0.362687   1.000000  1.575278e+04    5803
28               three   0.983931   0.341313   1.000000  1.574290e+04    5461
29                tree   0.970302   0.298625   1.000000  1.552483e+04    4778
30                 two   0.981600   0.278625   1.000000  1.570559e+04    4458
31                  up   0.976156   0.298625   1.000000  1.561849e+04    4778
32              visual   0.982301   0.341313   1.000000  1.571681e+04    5461
33                 wow   0.970225   0.325062   1.000000  1.552360e+04    5201
34                 yes   0.983431   0.384000   1.000000  1.573490e+04    6144
35                zero   0.986978   0.298625   1.000000  1.579164e+04    4778
```

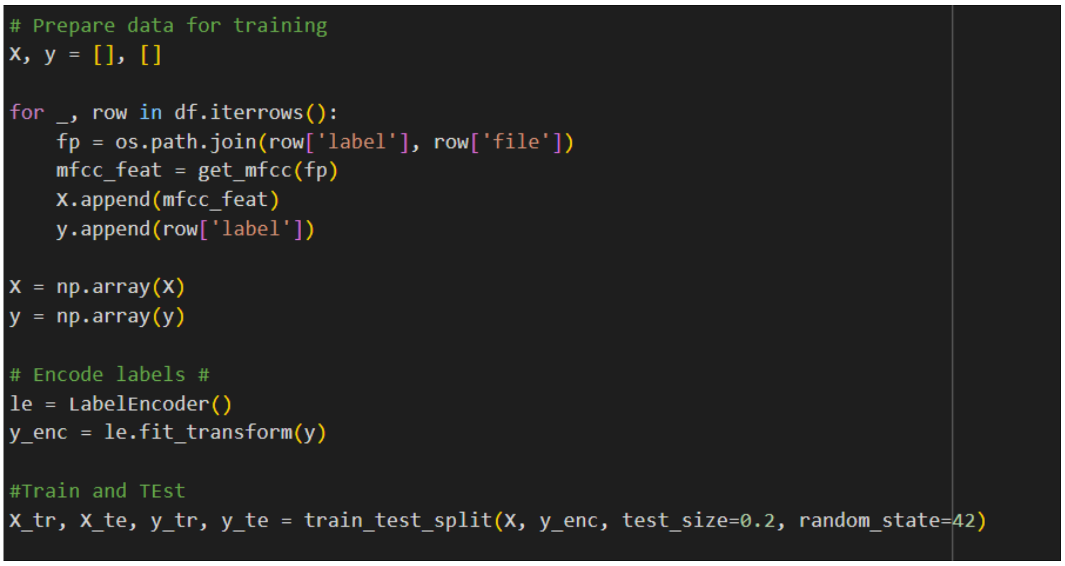


```
     max_ns  count
0   1522930      6
1     16000   1664
2     16000   2014
3     16000   2064
4     16000   2031
5     16000   2128
6     16000   3917
7     16000   3787
8     16000   4052
9     16000   1579
10    16000   1557
11    16000   3728
12    16000   3880
13    16000   2054
14    16000   2113
15    16000   1575
16    16000   3801
17    16000   2100
18    16000   3934
19    16000   3941
20    16000   3745
21    16000   3845
22    16000   3890
23    16000   3778
24    16000   3998
25    16000   2022
26    16000   3860
27    16000   3872
28    16000   3727
29    16000   1759
30    16000   3880
31    16000   3723
32    16000   1592
33    16000   2123
34    16000   4044
35    16000   4052

Analysis complete. CSV files saved.
```

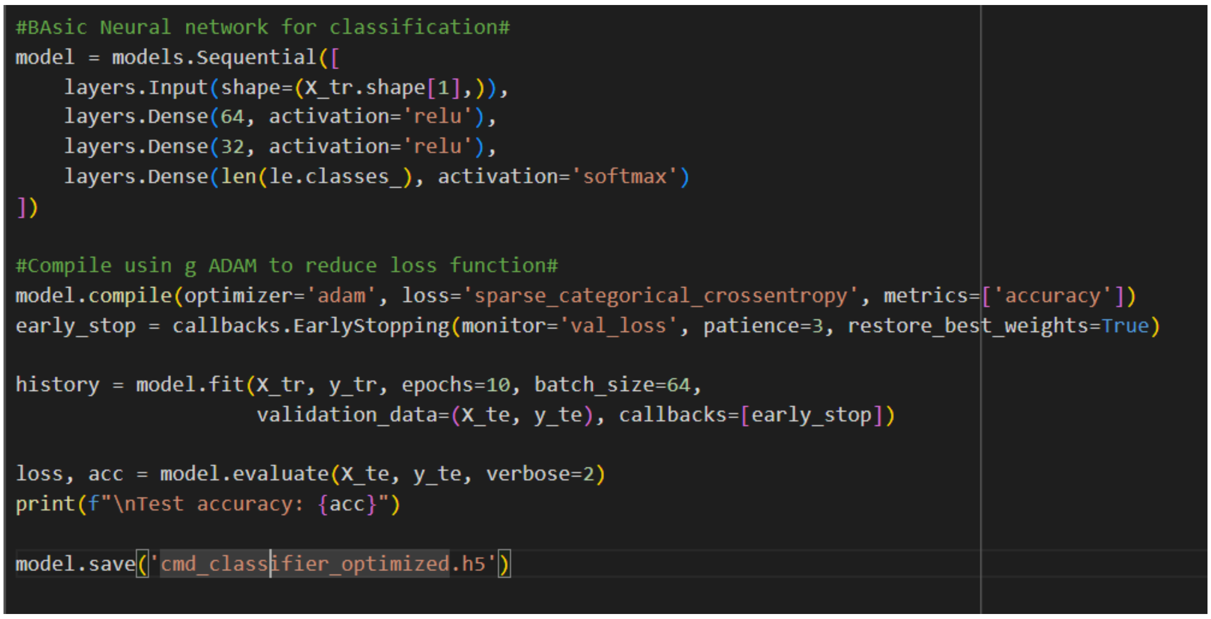
2. Train a classifier so that you are able to distinguish the commands in the dataset.



```
# Libraries Required
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
import tensorflow as tf
from tensorflow.keras import layers, models, callbacks
import numpy as np
import librosa

# Extract features (MFCCs) from the audio files
def get_mfcc(fp, sr=16000):
    audio, sr = librosa.load(fp, sr=sr)
    mfcc = librosa.feature.mfcc(y=audio, sr=sr, n_mfcc=20)  # Reduced to 20 MFCCs
    return np.mean(mfcc.T, axis=0)
```



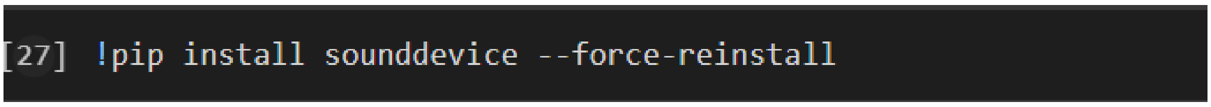
```
# Prepare data for training
X, y = [], []

for _, row in df.iterrows():
    fp = os.path.join(row['label'], row['file'])
    mfcc_feat = get_mfcc(fp)
    X.append(mfcc_feat)
    y.append(row['label'])

X = np.array(X)
y = np.array(y)

# Encode labels #
le = LabelEncoder()
y_enc = le.fit_transform(y)

#Train and TEst
X_tr, X_te, y_tr, y_te = train_test_split(X, y_enc, test_size=0.2, random_state=42)
```



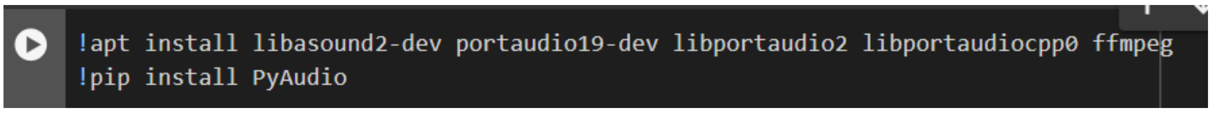
```
#BAsic Neural network for classification#
model = models.Sequential([
    layers.Input(shape=(X_tr.shape[1],)),
    layers.Dense(64, activation='relu'),
    layers.Dense(32, activation='relu'),
    layers.Dense(len(le.classes_), activation='softmax')
])

#Compile usin g ADAM to reduce loss function#
model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])
early_stop = callbacks.EarlyStopping(monitor='val_loss', patience=3, restore_best_weights=True)

history = model.fit(X_tr, y_tr, epochs=10, batch_size=64,
                    validation_data=(X_te, y_te), callbacks=[early_stop])

loss, acc = model.evaluate(X_te, y_te, verbose=2)
print(f"\nTest accuracy: {acc}")

model.save('cmd_classifier_optimized.h5')
```

3. Record about 30 samples of each command in your voice and create a new dataset (including a new user id for yourself).You may use a timer on your computer to synchronise.

Steps to achieve that:

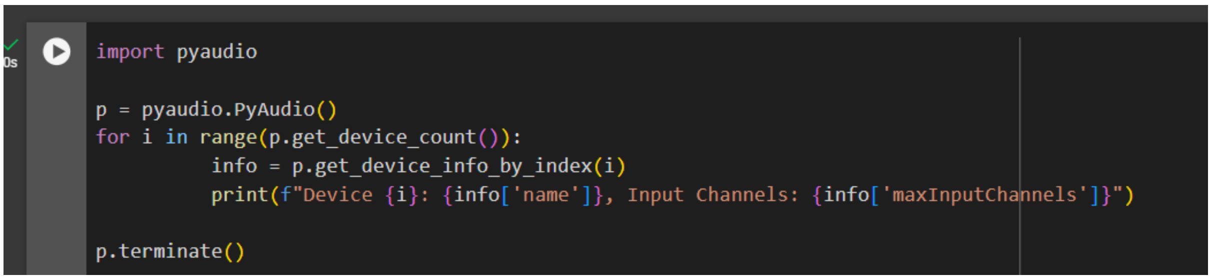


```
[27] !pip install sounddevice --force-reinstall
```



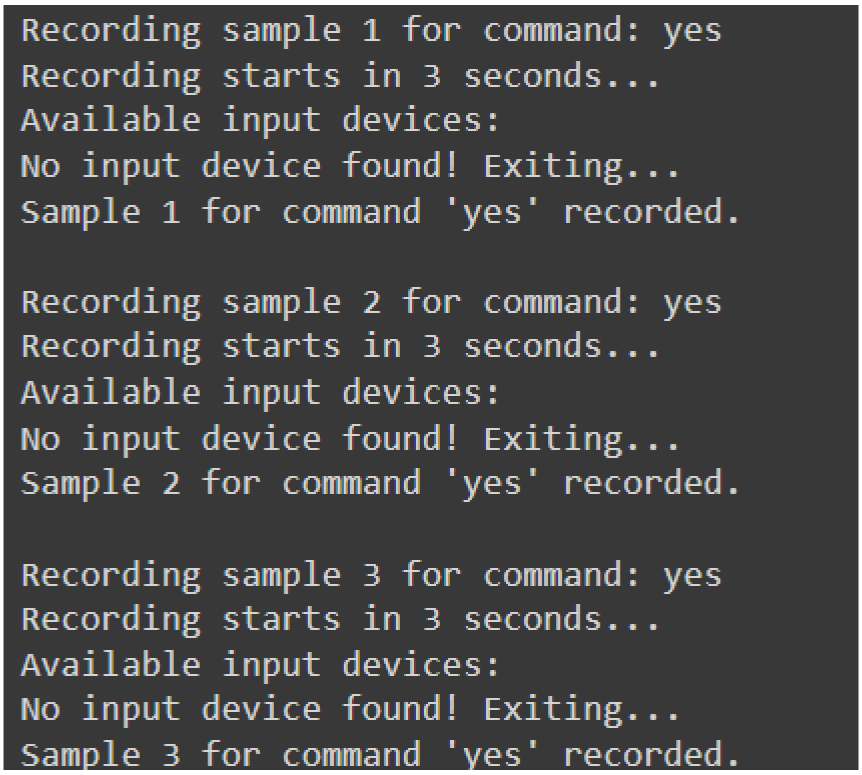
```
!apt install libasound2-dev portaudio19-dev libportaudio2 libportaudiocpp0 ffmpeg
!pip install PyAudio
```

For recording audio directly in colab environment



```
import pyaudio

p = pyaudio.PyAudio()
for i in range(p.get_device_count()):
        info = p.get_device_info_by_index(i)
        print(f"Device {i}: {info['name']}, Input Channels: {info['maxInputChannels']}")

p.terminate()
```

I have encountered a problem where the colab does not detect my microphone input. Ive tried many methods like using PyAudio/PortAudio and Sounddevice libraries but the error still remains



```
Recording sample 1 for command: yes
Recording starts in 3 seconds...
Available input devices:
No input device found! Exiting...
Sample 1 for command 'yes' recorded.

Recording sample 2 for command: yes
Recording starts in 3 seconds...
Available input devices:
No input device found! Exiting...
Sample 2 for command 'yes' recorded.

Recording sample 3 for command: yes
Recording starts in 3 seconds...
Available input devices:
No input device found! Exiting...
Sample 3 for command 'yes' recorded.
```

I did ultimately resort to record audio files from Audacity instead, which is a DAW used primarily make music and record voice. Although assuming its 30 samples of 8 commands its quite a daunting task at least for the given timeframe. It is not my intent to whine and complain and I understand your perspective, hence I apologize for not being apt enough to find a better solution to this.

Additionally, being honest, help from outer resources was indeed taken however I refrained from using LLM's.
Thank you.