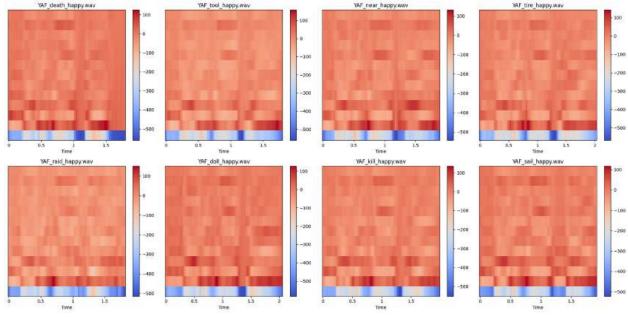
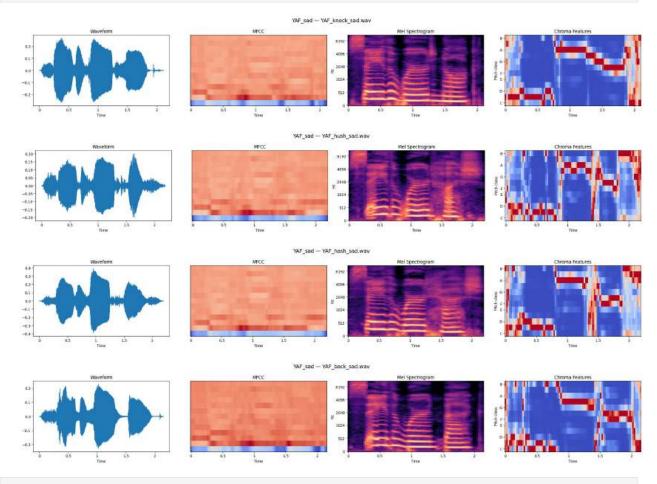
```
from google.colab import drive
drive.mount('/content/drive')
Drive already mounted at /content/drive; to attempt to forcibly
remount, call drive.mount("/content/drive", force remount=True).
import os
import librosa
import librosa.display
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model selection import train test split
from sklearn.preprocessing import LabelEncoder
import tensorflow as tf
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM, Dense, Dropout
from tensorflow.keras.utils import to categorical
data path = "/content/drive/MyDrive/TESS Toronto emotional speech set
data"
emotions = os.listdir(data path)
emotions
['TESS Toronto emotional speech set data',
 'YAF sad',
 'YAF neutral',
 'YAF angry',
 'OAF_Pleasant_surprise',
 'YAF pleasant surprised',
 'YAF happy',
 'YAF disgust',
 'YAF fear',
 'OAF Sad',
 'OAF disgust',
 'OAF_Fear',
 'OAF angry',
 'OAF_happy',
 'OAF neutral']
from pathlib import Path
def get audio files(data path, emotion):
    path = Path(data_path) / emotion
    return list(path.glob("*.wav"))
audio files = get audio files(data path, "YAF angry")
print(f"Found {len(audio files)} angry audio files")
```

For Specific Column

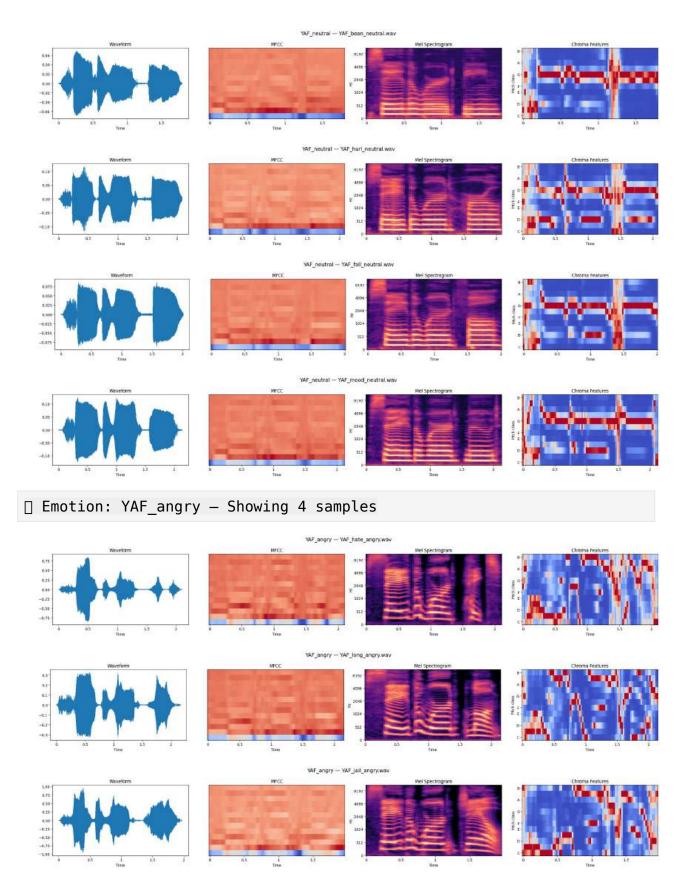
```
def plot_mfcc_grid(emotion_folder, num samples=8):
    files = os.listdir(os.path.join(data_path, emotion folder))
[:num samples]
    num cols = 4
    num rows = int(np.ceil(len(files) / num cols))
    fig, axs = plt.subplots(num rows, num cols, figsize=(20, 5 *
num rows))
    for idx, file in enumerate(files):
        file path = os.path.join(data path, emotion folder, file)
        y, sr = librosa.load(file_path)
        mfcc = librosa.feature.mfcc(y=y, sr=sr, n mfcc=13)
        row, col = divmod(idx, num cols)
        ax = axs[row][col] if num rows > 1 else axs[col]
        img = librosa.display.specshow(mfcc, x axis='time', ax=ax)
        ax.set title(file)
        fig.colorbar(img, ax=ax)
    # Hide empty subplots
    for idx in range(len(files), num rows * num cols):
        row, col = divmod(idx, num cols)
        ax = axs[row][col] if num rows > 1 else axs[col]
        ax.axis('off')
    plt.tight layout()
    plt.show()
# Example: plot MFCCs of 'YAF happy'
plot mfcc grid("YAF happy", num samples=8)
```

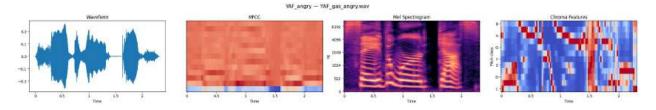


```
def plot_all_features_all_emotions(data_path, emotion folders,
samples per emotion=4):
    for emotion folder in emotion folders:
        folder_path = os.path.join(data_path, emotion folder)
        files = os.listdir(folder path)[:samples per emotion]
        print(f"□ Emotion: {emotion folder} - Showing {len(files)}
samples")
        for file in files:
            file path = os.path.join(folder path, file)
            y, sr = librosa.load(file path, duration=3)
            fig, axs = plt.subplots(1, 4, figsize=(24, 4))
            fig.suptitle(f"{emotion folder} - {file}", fontsize=14)
            # 1. Waveform
            librosa.display.waveshow(y, sr=sr, ax=axs[0])
            axs[0].set title("Waveform")
            # 2. MFCC
            mfcc = librosa.feature.mfcc(y=y, sr=sr, n mfcc=13)
            librosa.display.specshow(mfcc, x axis='time', ax=axs[1])
            axs[1].set title("MFCC")
            # 3. Log-Mel Spectrogram
            mel = librosa.feature.melspectrogram(y=y, sr=sr)
            mel db = librosa.power to db(mel, ref=np.max)
            librosa.display.specshow(mel db, x axis='time',
y_axis='mel', sr=sr, ax=axs[2])
            axs[2].set title("Mel Spectrogram")
```

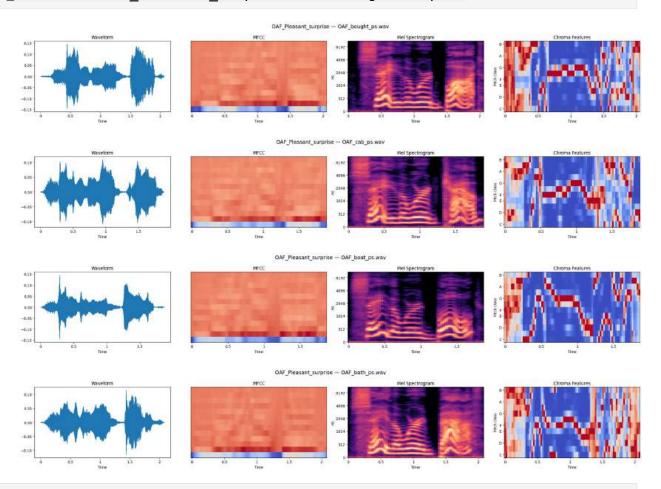


☐ Emotion: YAF_neutral — Showing 4 samples

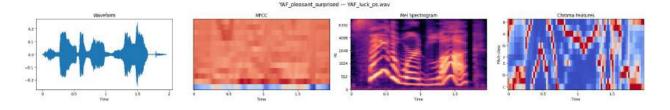


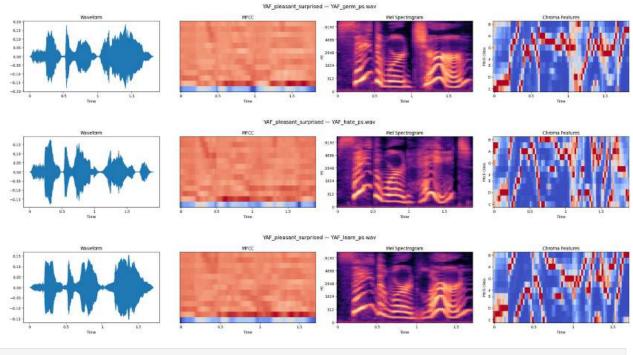


$\hfill \Box$ Emotion: OAF_Pleasant_surprise — Showing 4 samples

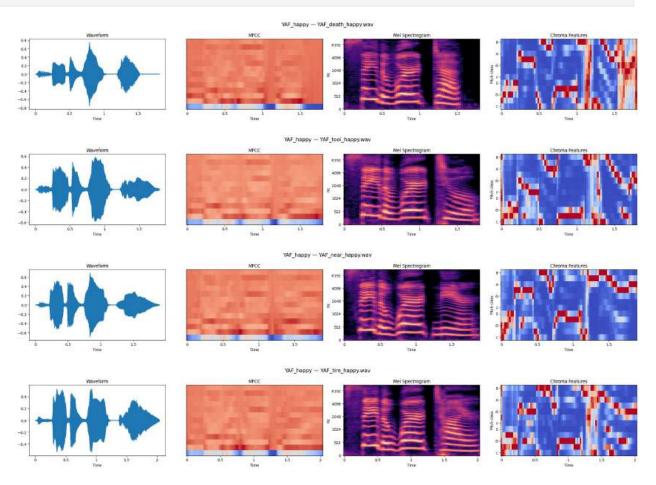


□ Emotion: YAF_pleasant_surprised - Showing 4 samples

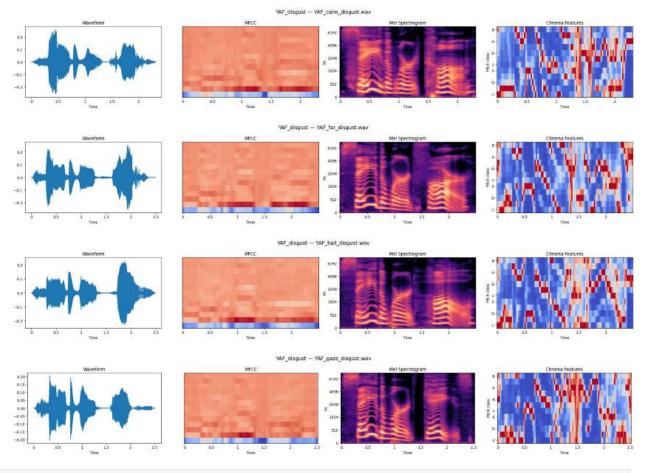




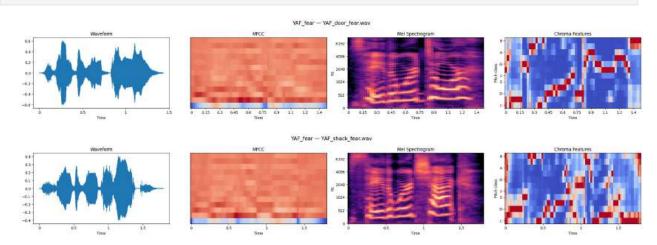
□ Emotion: YAF_happy - Showing 4 samples

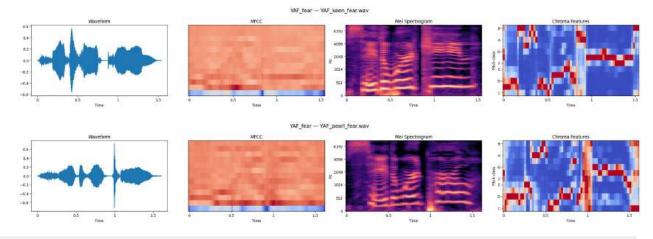


$\hfill \Box$ Emotion: YAF_disgust — Showing 4 samples

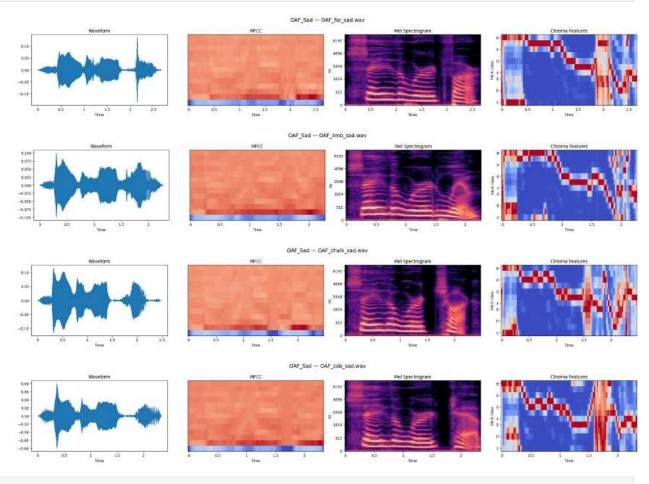


$\hfill \Box$ Emotion: YAF_fear — Showing 4 samples

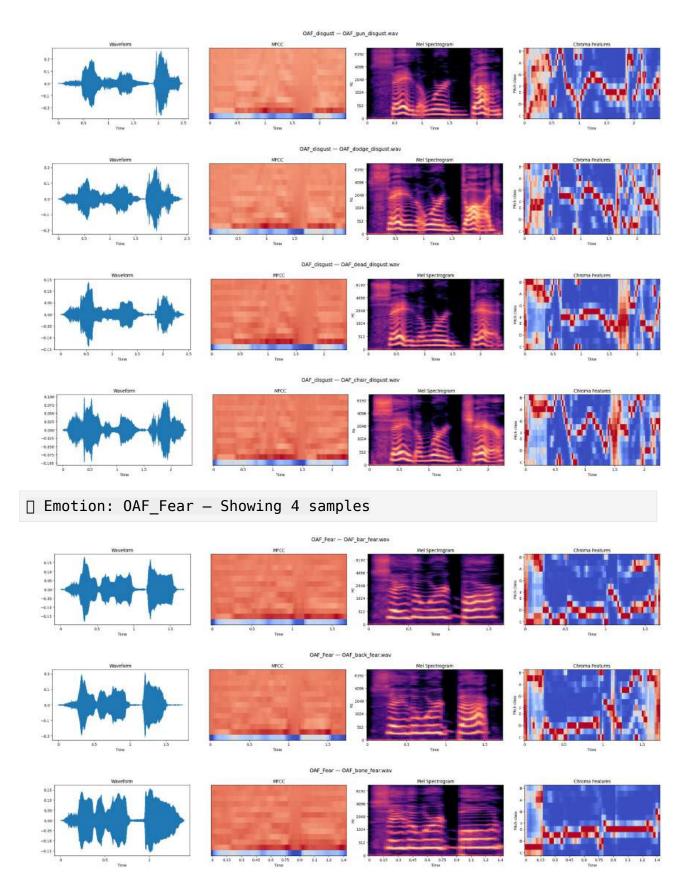


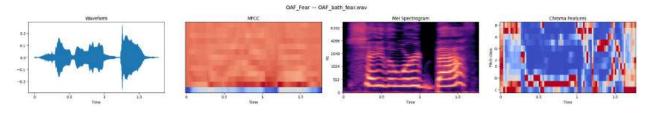


$\hfill \Box$ Emotion: OAF_Sad — Showing 4 samples

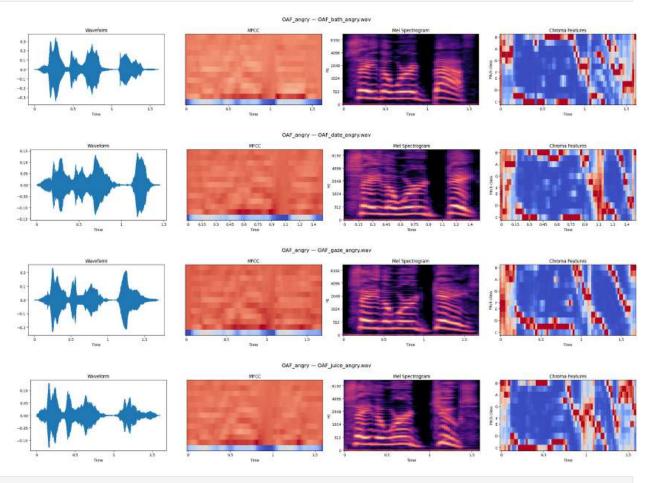


 $\hfill \Box$ Emotion: OAF_disgust — Showing 4 samples

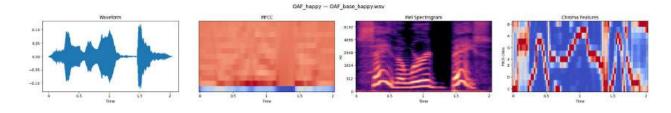


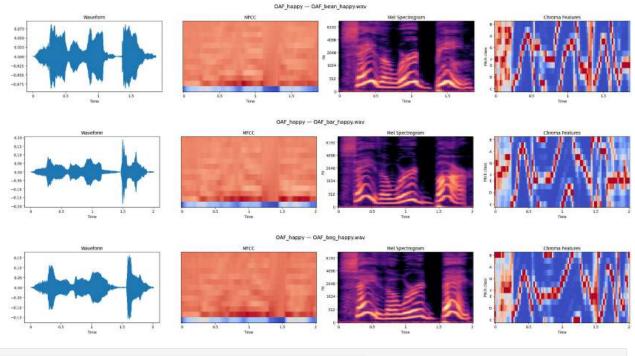


☐ Emotion: OAF_angry - Showing 4 samples

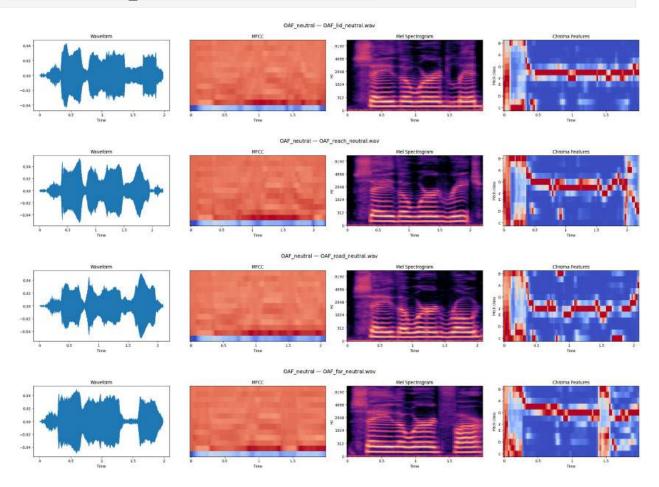


$\ \square$ Emotion: OAF_happy — Showing 4 samples





$\hfill \Box$ Emotion: OAF_neutral — Showing 4 samples



```
def extract features(file path):
    y, sr = librosa.load(file path, duration=3, offset=0.5)
    mfcc = librosa.feature.mfcc(y=y, sr=sr, n mfcc=40)
    mfcc scaled = np.mean(mfcc.T, axis=0)
    return mfcc scaled
features = []
labels = []
for folder in emotion folders:
    emotion label = folder.split(' ')[-1].lower()
    folder path = os.path.join(data path, folder)
    for file in os.listdir(folder path):
        file path = os.path.join(folder path, file)
            mfcc = extract features(file path)
            features.append(mfcc)
            labels.append(emotion label)
        except Exception as e:
            print(f"Error: {file path} - {e}")
from sklearn.preprocessing import LabelEncoder
X = np.array(features)
y = LabelEncoder().fit transform(labels)
X train, X test, y train, y test = train test split(X, y,
test size=0.2, random state=42, stratify=y)
print("Train size:", len(X_train))
print("Test size:", len(X_test))
Train size: 2240
Test size: 560
```

Extract MFCC features and labels for all data

```
import numpy as np
import os
import librosa
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split

def extract_features_labels(data_path, emotion_folders,
max_files_per_emotion=100):
    X, y = [], []

for emotion in emotion_folders:
```

```
folder_path = os.path.join(data path, emotion)
        files = os.listdir(folder path)[:max files per emotion]
        for file in files:
            file path = os.path.join(folder path, file)
            y audio, sr = librosa.load(file path, duration=3)
            mfcc = librosa.feature.mfcc(y=y_audio, sr=sr, n_mfcc=40)
            mfcc scaled = np.mean(mfcc.T, axis=0) # Take mean over
time axis
            X.append(mfcc scaled)
            y.append(emotion)
    return np.array(X), np.array(y)
# Prepare data
emotion folders = [f for f in os.listdir(data path) if not
f.startswith('TESS') and os.path.isdir(os.path.join(data path, f))]
X, y = extract features labels(data path, emotion folders,
max_files_per_emotion=100)
# Encode labels
le = LabelEncoder()
y encoded = le.fit transform(y)
# Split train-test
X train, X test, y train, y test = train test split(X, y encoded,
test size=0.2, random state=42, stratify=y encoded)
```

Train a simple MLP classifier

```
from sklearn.neural_network import MLPClassifier
from sklearn.metrics import classification_report, confusion_matrix,
ConfusionMatrixDisplay
import matplotlib.pyplot as plt

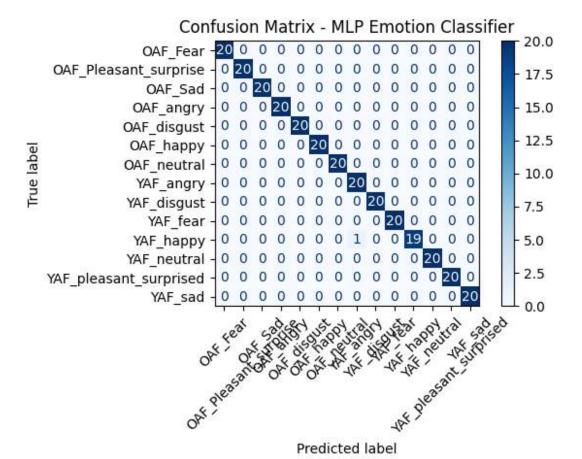
# Initialize MLP
mlp = MLPClassifier(hidden_layer_sizes=(128, 64), max_iter=300,
random_state=42)

# Train
mlp.fit(X_train, y_train)

# Predict
y_pred = mlp.predict(X_test)

# Report
print(classification_report(y_test, y_pred, target_names=le.classes_))
```

	precision	recall	f1-score	support
OAF_Fear OAF_Pleasant_surprise OAF Sad	1.00 1.00 1.00	1.00 1.00 1.00	1.00 1.00 1.00	20 20 20
OAF_angry OAF_disgust OAF_happy OAF_neutral YAF angry	1.00 1.00 1.00 1.00 0.95	1.00 1.00 1.00 1.00 1.00	1.00 1.00 1.00 1.00 0.98	20 20 20 20 20
YAF_disgust YAF_fear YAF_happy YAF_neutral	1.00 1.00 1.00 1.00	1.00 1.00 0.95 1.00	1.00	20 20 20 20 20
YAF_pleasant_surprised YAF_sad accuracy	1.00 1.00	1.00	1.00 1.00	20 20 280
macro avg weighted avg	1.00 1.00	1.00 1.00	1.00 1.00	280 280
<pre># Plot confusion matrix cm = confusion_matrix(y_test, y_pred) disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=le.classes_) disp.plot(xticks_rotation=45, cmap=plt.cm.Blues) plt.title("Confusion Matrix - MLP Emotion Classifier") plt.tight_layout() plt.show()</pre>				



def predict_emotion(file_path, model, label_encoder):
 y_audio, sr = librosa.load(file_path, duration=3)
 mfcc = librosa.feature.mfcc(y=y_audio, sr=sr, n_mfcc=40)
 mfcc_scaled = np.mean(mfcc.T, axis=0).reshape(1, -1)
 pred = model.predict(mfcc_scaled)
 emotion = label_encoder.inverse_transform(pred)
 return emotion[0]

new_audio_path = "/content/OAF_boat_happy.wav"
predicted_emotion = predict_emotion(new_audio_path, mlp, le)
print(f"Predicted emotion: {predicted_emotion}")

Predicted emotion: OAF_happy

new_audio_path = "/content/a.wav"
predicted_emotion = predict_emotion(new_audio_path, mlp, le)
print(f"Predicted emotion: {predicted_emotion}")

Predicted emotion: YAF_fear