IMPORTS

```
import os
import math
import librosa
import librosa.display
import numpy as np
import matplotlib.pyplot as plt
import numpy as np
from sklearn.decomposition import PCA
from sklearn.preprocessing import StandardScaler
from scipy.spatial.distance import cdist
import joblib
import sounddevice as sd
from scipy.signal import butter, lfilter
import soundfile as sf
```

RUTAS Y TIPOS DE FRUTAS

ORIGINAL TESTS

```
In [ ]: original = []
    original.extend([os.path.join(original_path, filename) for filename in os.
```

PROCESSED TESTS DICT

```
In [ ]: processed = []
    processed.extend([os.path.join(processed_path, filename) for filename in o
```

PARAMETROS DEL AUDIO

```
In [ ]: FRAME_SIZE = 512# In the documentation says it's convenient for speech.C
HOP_SIZE = int(FRAME_SIZE/2)
```

FUNCIONES GENERALES DE AUDIO

```
In [ ]: def load_audio(audiofile):
    test_audio, sr = librosa.load(audiofile, sr = None)
    duration = librosa.get_duration(filename=audiofile, sr=sr)
    return test_audio, sr, duration
```

FILTERS

```
In [ ]: def band_pass_filter(signal, sr, low_cutoff, high_cutoff):
    b, a = butter(N=3, Wn = [low_cutoff, high_cutoff], btype='band', fs=sr
    return lfilter(b, a, signal)
```

PROCCESSING OF THE AUDIO FILES FUNCTIONS

```
File naming
In [ ]: def get_name(original:list):
            return os.path.join(original path, "validation" + f"{len(original) + 1}
        Processing
In [ ]: def spectral_flux(signal):
            # Calcular el espectrograma de magnitudes
            spectrogram = np.abs(librosa.stft(signal, n fft = FRAME SIZE, hop leng
            # Calcular el flujo espectral
            spectral flux values = np.sum(np.diff(spectrogram, axis=1)**2, axis=0)
            return spectral flux values
In [ ]: def process(audio_in, audio_out, rms_umbral = 0.043, flux_umbral = 0.096):
            signal, sr, = load audio(audio in)
            rms = librosa.feature.rms(y = signal, frame_length = FRAME_SIZE, hop_l
            rms /= np.max(np.abs(rms))
            trms = librosa.times_like(rms, sr = sr, hop_length = HOP_SIZE, n_fft =
            trms /= trms[-1]
            flux = spectral flux(signal)
            flux /= np.max(np.abs(flux))
            fluxframes = range(len(flux))
            tflux = librosa.frames to time(fluxframes, hop length=HOP SIZE, n fft
            tflux /= tflux[-1]
            left_index = np.argmax(np.abs(flux) > flux_umbral)
            rigth index = len(flux) - 1 - np.argmax(np.abs(np.flip(flux)) > flux u
            tsignal = librosa.times_like(signal, sr = sr, hop_length=HOP_SIZE, n_f
            tsignal /= tsignal[-1]
                  = False
            flag
            pad_left = 0
            pad rigth = 0
            flag left = False
            flag_rigth = False
            while not flag:
```

if rms[0, left index] > rms umbral:

```
if left_index > pad_left + 15:
            rms left = left index - np.argmax(np.flip(np.abs(rms[0, :1
            if rms left <= 0:</pre>
                rms left = left index
            flag left = True
        else:
            pad left += 15
            left index = pad left + np.argmax(np.abs(flux[pad left:])
    else:
            rms_left = left_index
            flag left = True
    if rms[0, rigth index] > rms umbral:
        if rigth_index < (len(flux) - 1 - pad_rigth-15):</pre>
            rms_rigth = rigth_index + np.argmax(np.abs(rms[0, rigth_in
            if rms_rigth >= len(flux):
                rms rigth = rigth index
            flag_rigth = True
        else:
            pad rigth += 15
            rigth_index = len(flux[:-pad_rigth]) - 1 - np.argmax(np.fl
    else:
        rms_rigth = rigth_index
        flag rigth = True
    flag = flag left and flag rigth
left_index = min(left_index, rms_left)
rigth_index = max(rigth_index, rms_rigth)
mask = tsignal >= tflux[left_index]
ttrimed = tsignal[mask]
trimed = signal[mask]
mask = ttrimed <= tflux[rigth index]</pre>
ttrimed = ttrimed[mask]
trimed = trimed[mask]
sf.write(audio out, trimed, sr)
```

KNN

```
In [ ]:
        def knn(training, test, k_n):
                        = np.concatenate([v for v in training.values()], axis = 0)
            Χ
                        = np.concatenate([[k] * v.shape[0] for k, v in training.it
            У
            dist
                       = cdist(test, X)
            sorted_ind = np.argsort(dist, axis = 1)
                       = sorted ind[:, 0:k n]
            sorted k
            predicted = []
            for row in sorted k:
                           = list(y[row])
                prediction = max(set(labels), key = labels.count)
                predicted.append(prediction)
            return predicted
```

PLOTTING

```
In [ ]:
        #3d
        def plot_features3d(features):
            fig = plt.figure()
            ax = fig.add subplot(111, projection='3d')
            colors = dict(zip(fruit_types,['green','yellow','red','orange']))
            for fruit, points in features.items():
                ax.scatter(points[:, 0], points[:, 1], points[:, 2], c=colors[frui
            ax.set xlabel('Eje X')
            ax.set_ylabel('Eje Y')
            ax.set_zlabel('Eje Z')
            plt.show()
In [ ]: #3d
        def plot features3d extra(features):
            fig = plt.figure()
            ax = fig.add_subplot(111, projection='3d')
            colors = dict(zip(features.keys(),['green','yellow','red','orange','cy
            for fruit, points in features.items():
                ax.scatter(points[:, 0], points[:, 1], points[:, 2], c=colors[frui
                if fruit == 'validation':
                    for i, point in enumerate(points):
                         ax.text(point[0], point[1], point[2], f"{i}", color='red',
            ax.set_xlabel('Eje X')
            ax.set_ylabel('Eje Y')
            ax.set_zlabel('Eje Z')
            plt.show()
```

FEATURES EXTRACTION

Features extraction functions

band_energy_ratio = []

calculate power spectrogram

power_spectrogram = np.abs(spectrogram) ** 2

```
In [ ]: def calculate_split_frequency_bin(split_frequency, sample_rate, num_freque
    """Infer the frequency bin associated to a given split frequency."""

    frequency_range = sample_rate / 2
    frequency_delta_per_bin = frequency_range / num_frequency_bins
    split_frequency_bin = math.floor(split_frequency / frequency_delta_per
    return int(split_frequency_bin)
In [ ]: def band_energy_ratio(spectrogram, split_frequency, sample_rate):
    """Calculate band energy ratio with a given split frequency."""
```

split frequency bin = calculate split frequency bin(split frequency, s

```
power_spectrogram = power_spectrogram.T

# calculate BER value for each frame
for frame in power_spectrogram:
    sum_power_low_frequencies = frame[:split_frequency_bin].sum()
    sum_power_high_frequencies = frame[split_frequency_bin:].sum()
    band_energy_ratio_current_frame = sum_power_low_frequencies / (sum band_energy_ratio.append(band_energy_ratio_current_frame)

return np.array(band_energy_ratio)
```

```
In [ ]: def rms(signal, frames, hop):
    return librosa.feature.rms(y=signal, frame_length = frames, hop_length
```

function to get the features

```
In [ ]: def get_features(signal, sr):
            feature = np.empty((1, 0))
            # BER
            spec = librosa.stft(signal, n fft = FRAME SIZE, hop length = HOP SIZE)
            # max
            split frequency = 600
            BER = band_energy_ratio(spec, split_frequency, sr)
            feat = np.max(np.abs(BER))
            feature = np.append(feature, feat)
            # min
            # 1
            split_frequency = 1900
            BER = band energy ratio(spec, split frequency, sr)
            feat = np.min(np.abs(BER))
            feature = np.append(feature, feat)
            # 2
            split_frequency = 5000
            BER = band energy ratio(spec, split frequency, sr)
            feat = np.min(np.abs(BER))
            feature = np.append(feature, feat)
            # 3
            split_frequency = 9000
            BER = band energy ratio(spec, split frequency, sr)
            feat = np.min(np.abs(BER))
            feature = np.append(feature, feat)
            # std
            # 1
            split_frequency = 8000
            BER = band_energy_ratio(spec, split_frequency, sr)
            BER /= np.max(np.abs(BER))
            feat = np.std(BER)/np.mean(np.abs(BER))
            feature = np.append(feature, feat)
            # 2
            split frequency = 1000
            BER = band_energy_ratio(spec, split_frequency, sr)
            BER /= np.max(np.abs(BER))
```

```
feat = np.std(BER)/np.mean(np.abs(BER))
feature = np.append(feature, feat)
#ZCR
cutoff = 5000
cuton = 1000
filtered = band_pass_filter(signal, sr, cuton, cutoff)
zcr = librosa.feature.zero_crossing_rate(filtered, frame_length=FRAME_
zcr /= np.max(np.abs(zcr))
# mean
feat = np.mean(zcr)
feature = np.append(feature, feat)
# maximum
cutoff = 10000
cuton = 10
filtered = band_pass_filter(signal, sr, cuton, cutoff)
zcr = librosa.feature.zero_crossing_rate(filtered, frame_length=FRAME_
feat = np.max(np.abs(zcr))
feature = np.append(feature, feat)
# std
cutoff = 10000
cuton = 20
filtered = band_pass_filter(signal, sr, cuton, cutoff)
zcr = librosa.feature.zero crossing rate(filtered, frame length=FRAME
feat = np.std(zcr)/np.mean(np.abs(zcr))
feature = np.append(feature, feat)
# mean Local
cutoff = 5000
cuton = 1000
filtered = band_pass_filter(signal, sr, cuton, cutoff)
zcr = librosa.feature.zero_crossing_rate(filtered, frame_length=FRAME_
zcr /= np.max(np.abs(zcr))
feat = np.mean(zcr[((len(zcr)*3)//14 - 5) : ((len(zcr)*3)//14 + 5)])
feature = np.append(feature, feat)
# Local max
cutoff = 10000
cuton = 10
filtered = band pass filter(signal, sr, cuton, cutoff)
zcr = librosa.feature.zero_crossing_rate(filtered, frame_length=FRAME_
feat = np.max(zcr[((len(zcr)*3)//4 - 10) : ((len(zcr)*3)//4 + 10)])
feature = np.append(feature, feat)
# Roll off
cuton = 100
cutoff = 8500
filtered = band_pass_filter(signal, sr, cuton, cutoff)
roll_off = librosa.feature.spectral_rolloff(y=filtered, sr=sr, n_fft=F
roll off /= np.max(np.abs(roll off))
# mean
feat = np.mean(np.abs(roll_off))
feature = np.append(feature, feat)
# max
cuton = 100
cutoff = 8500
```

```
filtered = band_pass_filter(signal, sr, cuton, cutoff)
roll_off = librosa.feature.spectral_rolloff(y=filtered, sr=sr, n_fft=F
feat = np.max(np.abs(roll off))
feature = np.append(feature, feat)
cutoff = 8500
cuton = 50
filtered = band pass filter(signal, sr, cuton, cutoff)
roll off = librosa.feature.spectral_rolloff(y=filtered, sr=sr, n_fft=F
roll off /= np.max(np.abs(roll off))
feat = np.std(np.abs(roll_off))/np.mean(np.abs(roll_off))
feature = np.append(feature, feat)
#MFCCS
n \text{ mfcc} = 4
# 1
cuton = 500
cutoff = 5000
filtered = band pass filter(signal, sr, cuton, cutoff)
mfccs = librosa.feature.mfcc(y = signal, sr=sr, n mfcc = n mfcc, n fft
feat = np.max(mfccs, axis = 1)
feature = np.append(feature, feat[3])
# 2
cuton = 10
cutoff = 8000
filtered = band_pass_filter(signal, sr, cuton, cutoff)
mfccs = librosa.feature.mfcc(y = filtered, sr=sr, n_mfcc = n_mfcc, n_f
mfccs /= np.max(np.abs(mfccs), axis = 1, keepdims=True)
feat = np.std(np.abs(mfccs), axis = 1)/np.mean(np.abs(mfccs), axis = 1
feature = np.append(feature, feat[3])
# 3
cuton = 10
cutoff = 8000
filtered = band_pass_filter(signal, sr, cuton, cutoff)
mfccs = librosa.feature.mfcc(y = filtered, sr=sr, n_mfcc = n_mfcc, n_f
mfccs /= np.max(np.abs(mfccs), axis = 1, keepdims=True)
mfccs = mfccs[:, ((mfccs.shape[1]*4) // 5 - 10):((mfccs.shape[1]*4) //
feat = np.std(np.abs(mfccs), axis=1) / np.mean(np.abs(mfccs), axis=1)
feature = np.append(feature, feat[1])
#envelope
env = rms(signal, FRAME_SIZE, HOP_SIZE)
env = env.reshape(-1,)
selected = np.linspace(0, len(env) - 1, 30, dtype=int)
env = env[selected]
feat = env[11]
feature = np.append(feature, feat)
feat = env[12]
feature = np.append(feature, feat)
return feature
```

AUFIO RECORGING

Out[]: 'duration = 3 # Duración de la grabación en segundos\nfs = 480
00 # Frecuencia de muestreo en Hz\n\nprint("Grabando...")\ndata = sd.rec
 (int(duration * fs), samplerate = fs, channels = 1, dtype = \'int16\')\ns
 d.wait()\nprint("Grabación completa.")\n\nfile = get_name(original)\nsf.w
 rite(file, data, fs)\noriginal.append(file)'

PROCESSING

```
In [ ]: already_processed = [os.path.basename(audio) for audio in processed]
for audio in original:
    basename = os.path.basename(audio)
    if basename in already_processed:
        pass
    else:
        audio_out = os.path.join(processed_path, basename)
        process(audio, audio_out)
        processed.append(audio_out)
```

FEATURE EXTRACTION

```
In [ ]: validation_features = None
    for audio in processed:
        signal, sr, _ = load_audio(audio)
        feature = get_features(signal, sr)

if validation_features is not None:
        validation_features = np.vstack([validation_features, feature])
    else:
        validation_features = feature.reshape(1, -1)
```

```
C:\Users\Juan\AppData\Local\Temp\ipykernel_1852\1587896153.py:3: FutureWarn
ing: get_duration() keyword argument 'filename' has been renamed to 'path'
in version 0.10.0.
        This alias will be removed in version 1.0.
        duration = librosa.get_duration(filename=audiofile, sr=sr)
c:\Users\Juan\AppData\Local\Programs\Python\Python311\Lib\site-packages\lib
rosa\feature\spectral.py:2143: UserWarning: Empty filters detected in mel f
requency basis. Some channels will produce empty responses. Try increasing
your sampling rate (and fmax) or reducing n_mels.
    mel_basis = filters.mel(sr=sr, n_fft=n_fft, **kwargs)
```

LOAD THE REDUCED MODEL

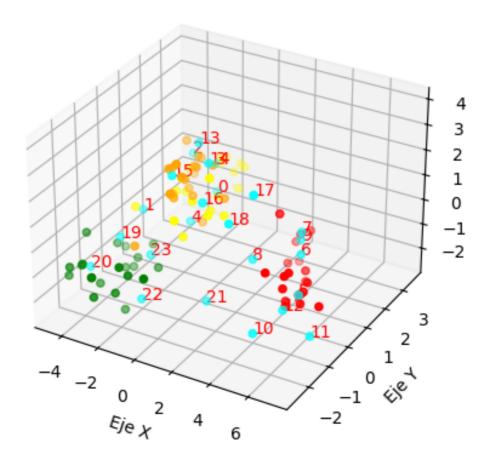
```
In [ ]: model = joblib.load(model_file)
    reduced:dict = model['features']
    pca = model['pca']
    scaler = model['scaler']
```

TRANSFORM

```
In [ ]: scaled_validation_features = scaler.transform(validation_features)
    reduced_validation = pca.transform(scaled_validation_features)
```

EXTENSION OF THE FEATURES DICT

```
In [ ]: extra = reduced.copy()
   extra['validation'] = reduced_validation
   plot_features3d_extra(extra)
```



PREDICTION

```
In [ ]: from prettytable import PrettyTable
        prediction = knn(reduced, reduced_validation, 3)
        aciertos = []
        audios = []
        N_aciertos = 0
        for i, audio in enumerate(processed):
            audios.append(os.path.basename(audio))
            if prediction[i] in os.path.basename(audio):
                aciertos.append('acierto')
                N aciertos += 1
            else:
                aciertos.append('falla')
        precission = N_aciertos*100/len(aciertos)
        table = PrettyTable()
        table.field_names = ['audios'] + audios
        table.add_row(['prediction'] + prediction)
        table.add_row(['results'] + aciertos)
        print(table)
        print(f"Se acertaron: {N_aciertos}/{len(prediction)}")
        print(f"El porcentaje de aciertos es: {precission}")
```

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