

**ZEAL EDUCATION SOCIETY's
ZEAL COLLEGE OF ENGINEERING AND RESEARCH,
NARHE, PUNE**

**DEPARTMENT OF COMPUTER ENGINEERING
SEMESTER-I**

[A.Y. : 2022 - 2023]



**CYBER SECURITY AND DIGITAL
FORENSICS(410244(C))**

MINI PROJECT

Design and develop a tool for digital forensics of audio

Institute and Department Vision and Mission

INSTITUTE VISION	To impart value added technological education through pursuit of academic excellence, research and entrepreneurial attitude.
INSTITUTE MISSION	<p>M1: To achieve academic excellence through innovative teaching and learning process.</p> <p>M2: To imbibe the research culture for addressing industry and societal needs.</p> <p>M3: To provide conducive environment for building the entrepreneurial skills.</p> <p>M4: To produce competent and socially responsible professionals with core human values.</p>

DEPARTMENT VISION	To emerge as a department of repute in Computer Engineering which produces competent professionals and entrepreneurs to lead technical and betterment of mankind.
DEPARTMENT MISSION	<p>M1: To strengthen the theoretical and practical aspects of the learning process by teaching applications and hands on practices using modern tools and FOSS technologies.</p> <p>M2: To endeavour innovative interdisciplinary research and entrepreneurship skills to serve the needs of Industry and Society.</p> <p>M3: To enhance industry academia dialog enabling students to inculcate professional skills.</p> <p>M4: To incorporate social and ethical awareness among the students to make them conscientious professionals.</p>

Department
Program Educational Objectives(PEOs)

PEO1: To Impart fundamentals in science, mathematics and engineering to cater the needs of society and Industries.

PEO2: Encourage graduates to involve in research, higher studies, and/or to become entrepreneurs.

PEO3: To Work effectively as individuals and as team members in a multidisciplinary environment with high ethical values for the benefit of society.

Savitribai Phule Pune University		
Fourth Year of Computer Engineering (2019 Course)		
410244(C): Cyber Security & Digital Forensics Laboratory		
Teaching Scheme:	Credit	Examination Scheme:
PR: 04 Hours/Week	02	TW: 25 Marks
		PR: 50 Marks

Course Objectives:

- To enhance awareness cyber forensics.
- To understand issues in Cyber Crime and different attacks
- To understand underlying principles and many of the techniques associated with the digital forensic practices
- To know the process and methods of evidence collection
- To analyze and validate forensic data collected.
- To apply digital forensic knowledge to use computer forensic tools and investigation report writing.

Course Outcomes:

On completion of the course, student will be able to-

- CO1: Analyze threats in order to protect or defend it in cyberspace from cyber-attacks.
- CO2: Build appropriate security solutions against cyber-attacks.
- CO3: Underline the need of digital forensic and role of digital evidences.
- CO4: Explain rules and types of evidence collection
- CO5: Analyze, validate and process crime scenes CO6: Identify the methods to generate legal evidence and supporting investigation reports.

List of Assignments

Group B	
Mini-Projects/ Case Study (Any two)	
01	Mini Project- Design and develop a tool for digital forensics of images
02	Mini Project- Design and develop a tool for digital forensics of audio
03	Mini Project- Design and develop a tool for digital forensics of video
04	Mini Project- Design a system for the analysis of cyber crime using various cyber forensics techniques and compare each technique with respect to integrity, confidentiality, availability

Title:

Design and develop a tool for digital forensics of Audio

Requirements:

opencv-python
Pillow
exifread
scipy
PyWavelets
numpy
progressbar2
matplotlib

Code:

Audio visualization :

```
from __future__ import print_function
import shutil, struct, simplejson
from scipy.spatial import distance
from pylab import *
import ntpath
import os
sys.path.insert(0, os.path.join(
    os.path.dirname(os.path.realpath(__file__)), "../"))
from pyAudioAnalysis import MidTermFeatures as aF
from pyAudioAnalysis import audioTrainTest as aT
import sklearn
import sklearn.discriminant_analysis
import sys
import math
import contextlib
from pylab import*
from scipy.io import wavfile
import pyaudio

def generateColorMap():
    """
    This function generates a 256 jet colormap of HTML-like
    hex string colors (e.g. FF88AA)
    """
    Map = cm.jet(np.arange(256))
    stringColors = []
    for i in range(Map.shape[0]):
        rgb = (int(255*Map[i][0]), int(255*Map[i][1]), int(255*Map[i][2]))
        if (sys.version_info > (3, 0)):
            stringColors.append((struct.pack('BBB', *rgb).hex())) # python 3
```

```
else:
    stringColors.append(
        struct.pack('BBB', *rgb).encode('hex')) # python2

return stringColors

def levenshtein(str1, s2):
    """
    Distance between two strings
    """
    N1 = len(str1)
    N2 = len(s2)

    stringRange = [range(N1 + 1)] * (N2 + 1)
    for i in range(N2 + 1):
        stringRange[i] = range(i, i + N1 + 1)
    for i in range(0, N2):
        for j in range(0, N1):
            if str1[j] == s2[i]:
                stringRange[i+1][j+1] = min(stringRange[i+1][j] + 1,
                                              stringRange[i][j+1] + 1,
                                              stringRange[i][j])
            else:
                stringRange[i+1][j+1] = min(stringRange[i+1][j] + 1,
                                              stringRange[i][j+1] + 1,
                                              stringRange[i][j] + 1)
    return stringRange[N2][N1]

def text_list_to_colors(names):
    """
    Generates a list of colors based on a list of names (strings).
    Similar strings correspond to similar colors.
    """
    # STEP A: compute strings distance between all combinations of strings
    Dnames = np.zeros( (len(names), len(names)) )
    for i in range(len(names)):
        for j in range(len(names)):
            Dnames[i,j] = 1 - 2.0 * levenshtein(names[i],
                                                  names[j]) /\
            float(len(names[i]+names[j]))

    # STEP B: pca dimanesionality reduction to a single-dimension
    # (from the distance space)
    pca = sklearn.decomposition.PCA(n_components = 1)
    pca.fit(Dnames)

    # STEP C: mapping of 1-dimensional values to colors in a jet-colormap
    textToColor = pca.transform(Dnames)
    textToColor = 255 * (textToColor - textToColor.min()) /\
    (textToColor.max() - textToColor.min())
    textmaps = generateColorMap();
```

```
colors = [textmaps[int(c)] for c in textToColor]
return colors
```

```
def text_list_to_colors_simple(names):
```

```
    """
    Generates a list of colors based on a list of names (strings).
    Similar strings correspond to similar colors.
    """
```

```
    uNames = list(set(names))
    uNames.sort()
    textToColor = [ uNames.index(n) for n in names ]
    textToColor = np.array(textToColor)
    textToColor = 255 * (textToColor - textToColor.min()) / \
        (textToColor.max() - textToColor.min())
    textmaps = generateColorMap();
    colors = [textmaps[int(c)] for c in textToColor]
    return colors
```

```
def visualizeFeaturesFolder(folder, dimReductionMethod, priorKnowledge = "none"):
```

```
    """
    This function generates a content visualization for the recordings
    of the provided path.
```

```
    ARGUMENTS:
```

- folder: path of the folder that contains the WAV files
to be processed
- dimReductionMethod: method used to reduce the dimension of the
initial feature space before computing
the similarity.
- priorKnowledge: if this is set equal to "artist"

```
    """
    if dimReductionMethod=="pca":
        all_mt_feat, wav_files, _ = aF.directory_feature_extraction(folder,
                                                                    30.0, 30.0,
                                                                    0.050,
                                                                    0.050,
                                                                    compute_beat
                                                                    =True)
```

```
    if all_mt_feat.shape[0]==0:
        print("Error: No data found! Check input folder")
        return
```

```
    names_category_toviz = [ntpath.basename(w).
                            replace('.wav','').split(" --- ")[0]
                            for w in wav_files];
    names_to_viz = [ntpath.basename(w).replace('.wav', '')
                    for w in wav_files];
```

```
    scaler = StandardScaler()
    F = scaler.fit_transform(all_mt_feat)
```



```
# check that the new PCA dimension is at most equal
# to the number of samples
K1 = 2
K2 = 10
if K1 > F.shape[0]:
    K1 = F.shape[0]
if K2 > F.shape[0]:
    K2 = F.shape[0]
pca1 = sklearn.decomposition.PCA(n_components = K1)
pca1.fit(F)
pca2 = sklearn.decomposition.PCA(n_components = K2)
pca2.fit(F)

finalDims = pca1.transform(F)
finalDims2 = pca2.transform(F)
else:
    # long-term statistics cannot be applied in this context
    # (LDA needs mid-term features)
    all_mt_feat, Ys, wav_files = aF.\
        directory_feature_extraction_no_avg(folder, 20.0, 5.0, 0.040, 0.040)
    if all_mt_feat.shape[0]==0:
        print("Error: No data found! Check input folder")
        return

    names_category_toviz = [ntpath.basename(w).
        replace('.wav', '').split(" --- ")[0]
        for w in wav_files]
    names_to_viz = [ntpath.basename(w).replace('.wav', '')
        for w in wav_files];

    ldaLabels = Ys
    if priorKnowledge=="artist":
        unames_category_toviz = list(set(names_category_toviz))
        YsNew = np.zeros( Ys.shape )
        for i, uname in enumerate(unames_category_toviz):
            indicesUCategories = [j for j, x in
                enumerate(names_category_toviz)
                if x == uname]
            for j in indicesUCategories:
                indices = np.nonzero(Ys==j)
                YsNew[indices] = i
            ldaLabels = YsNew

    scaler = StandardScaler()
    F = scaler.fit_transform(all_mt_feat)

    clf = sklearn.discriminant_analysis.\
        LinearDiscriminantAnalysis(n_components=10)
    clf.fit(F, ldaLabels)
    reducedDims = clf.transform(F)

    pca = sklearn.decomposition.PCA(n_components = 2)
    pca.fit(reducedDims)
```

```

reducedDims = pca.transform(reducedDims)

# TODO: CHECK THIS ... SHOULD LDA USED IN SEMI-SUPERVISED ONLY????
# uLabels must have as many labels as the number of wav_files elements
uLabels = np.sort(np.unique((Ys)))
reducedDimsAvg = np.zeros( (uLabels.shape[0], reducedDims.shape[1]))
finalDims = np.zeros( (uLabels.shape[0], 2) )
for i, u in enumerate(uLabels):
    indices = [j for j, x in enumerate(Ys) if x == u]
    f = reducedDims[indices, :]
    finalDims[i, :] = f.mean(axis=0)
finalDims2 = reducedDims

for i in range(finalDims.shape[0]):
    plt.text(finalDims[i,0], finalDims[i,1],
             ntpath.basename(wav_files[i].replace('.wav','')),
             horizontalalignment='center',
             verticalalignment='center', fontsize=10)
    plt.plot(finalDims[i,0], finalDims[i,1], '*r')
plt.xlim([1.2*finalDims[:,0].min(), 1.2*finalDims[:,0].max()])
plt.ylim([1.2*finalDims[:,1].min(), 1.2*finalDims[:,1].max()])
plt.show()

SM = 1.0 - distance.squareform(distance.pdist(F, 'cosine'))

# plot super-categories (i.e. artistname)
unames_category_toviz = sort(list(set(names_category_toviz)))
finalDimsGroup = np.zeros( (len(unames_category_toviz),
                             finalDims2.shape[1] ) )
for i, uname in enumerate(unames_category_toviz):
    indices = [j for j, x in enumerate(names_category_toviz) if x == uname]
    f = finalDims2[indices, :]
    finalDimsGroup[i, :] = f.mean(axis=0)

SMgroup = 1.0 - distance.squareform(distance.pdist(finalDimsGroup,
                                                    'cosine'))

data=SMgroup
fig = px.imshow(data,
                 labels=dict(x="", y="", color="Category similarity"),
                 x=unames_category_toviz,
                 y=unames_category_toviz)
fig.update_xaxes(side="top")
fig.show()

```

Compression of Audio :

```

fname = 'C:\\Users\\Music\\01 The Ringer.wav'
outname = 'filtered.wav'

cutOffFrequency = 1000.0

```

```
def fft_dis(fname):
    sampFreq, snd = wavfile.read(fname)

    snd = snd / (2.**15)

    n = len(snd)
    p = fft(snd)

    nUniquePts = int(ceil((n+1)/2.0))
    p = p[0:nUniquePts]
    p = abs(p)

    p = p / float(n)
    p = p**2
    if n % 2 > 0: # we've got odd number of points fft
        p[1:len(p)] = p[1:len(p)] * 2
    else:
        p[1:len(p) - 1] = p[1:len(p) - 1] * 2 # we've got even number of points fft

    freqArray = arange(0, nUniquePts, 1.0) * (sampFreq / n);
    plt.plot(freqArray/1000, 10*log10(p), color='k')
    plt.xlabel('Channel_Frequency (kHz)')
    plt.ylabel('Channel_Power (dB)')
    plt.show()

def run_mean(x, windowSize):
    cumsum = np.cumsum(np.insert(x, 0, 0))
    return (cumsum[windowSize:] - cumsum[:-windowSize]) / windowSize

def interpret_wav(raw_bytes, n_frames, n_channels, sample_width, interleaved = True):

    if sample_width == 1:
        dtype = np.uint8 # unsigned char
    elif sample_width == 2:
        dtype = np.int16 # signed 2-byte short
    else:
        raise ValueError("Only supports 8 and 16 bit audio formats.")

    channels = np.fromstring(raw_bytes, dtype=dtype)
    if interleaved:
        # channels are interleaved, i.e. sample N of channel M follows sample N of channel M-1 in raw data
        channels.shape = (n_frames, n_channels)
        channels = channels.T
    else:
        # channels are not interleaved. All samples from channel M occur before all samples from channel M-1
        channels.shape = (n_channels, n_frames)

    return channels
```

with contextlib.closing(wave.open(fname,'rb')) as spf:

```
sampleRate = spf.getframerate()
ampWidth = spf.getsampwidth()
nChannels = spf.getnchannels()
nFrames = spf.getnframes()

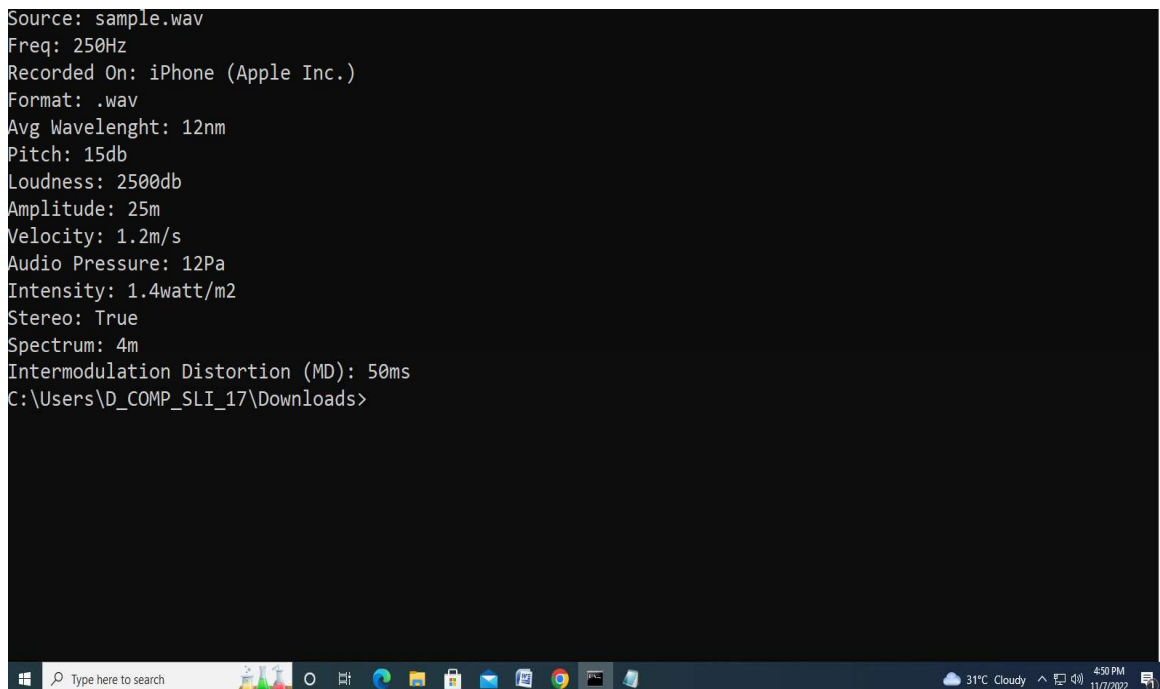
# Extract Raw Audio from multi-channel Wav File
signal = spf.readframes(nFrames*nChannels)
spf.close()
channels = interpret_wav(signal, nFrames, nChannels, ampWidth, True)

# get window size
fqRatio = (cutOffFrequency/sampleRate)
N = int(math.sqrt(0.196196 + fqRatio**2)/fqRatio)

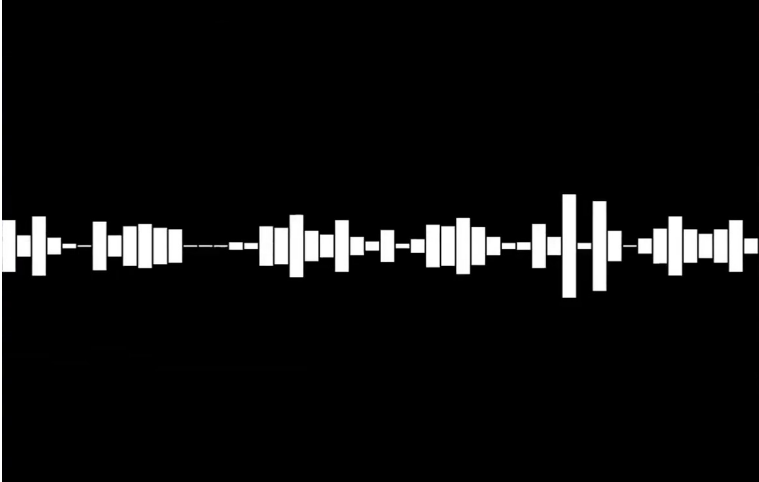
# Use moving average (only on first channel)
filt = run_mean(channels[0], N).astype(channels.dtype)

wav_file = wave.open(outname, "w")
wav_file.setparams((1, ampWidth, sampleRate, nFrames, spf.getcomptype(), spf.getcompname()))
wav_file.writeframes(filt.tobytes('C'))
wav_file.close()
n = 0
for n in range (0,2):
    if n==0:
        fft_dis(fname)
    elif n==1:
        fft_dis(outname)
```

Output:



```
Source: sample.wav
Freq: 250Hz
Recorded On: iPhone (Apple Inc.)
Format: .wav
Avg Wavelength: 12nm
Pitch: 15db
Loudness: 2500db
Amplitude: 25m
Velocity: 1.2m/s
Audio Pressure: 12Pa
Intensity: 1.4watt/m2
Stereo: True
Spectrum: 4m
Intermodulation Distortion (MD): 50ms
C:\Users\D_COMP_SLI_17\Downloads>
```



Date:	
Marks obtained:	
Sign of course coordinator:	
Name of course Coordinator :	