Assignment 1: Time Domain Features Extraction

CS 4347: Sound and Music Computing

due Wednesday 8 February 2017, 11:59 pm

- 0. This assignment will make use of the "Music & Speech" dataset of Marsyas:
 - http://opihi.cs.uvic.ca/sound/music_speech.tar.gz
 - This dataset has two copies of each song; delete the music/ and speech/ directories and use the files in music-wav/ and speech-wav/ directories. There are 64 music and 64 speech files; each file is 30 seconds of audio stored as 16-bit signed integers at 22050 Hz.
 - Ground truth data for this dataset:

http://www.comp.nus.edu.sg/~duanzy/music_speech.mf

Format of the file is filename \t (tab) label \n (newline), one song per line:

filename1\tlabel1\n

filename2\tlabel2\n

. . .

filename128\tlabel128\n

The label will be music or speech.

- 1. Write a python program that will:
 - Read the ground-truth music_speech.mf file
 - Load each wav file and convert the data to floats by dividing the samples by 32768.0. Hint: use scipy.io.wavfile.read()
 - Calculate 5 features for each file according to the given formulae. Use only one vector per file (don't use multiple buffers for each file). Given $X = \{x_0, x_1, x_2, \dots x_{N-1}\}$,
 - (a) Root-mean-squared (RMS)

$$X_{\text{RMS}} = \sqrt{\frac{1}{N} \sum_{i=0}^{N-1} x_i^2}$$

(b) Peak-to-average-ratio (PAR)

$$X_{\text{PAR}} = \frac{\max|x_i|}{X_{\text{PMS}}}$$

(c) Zero crossings (ZCR)

$$X_{\text{ZCR}} = \frac{1}{N-1} \sum_{i=1}^{N-1} \begin{cases} 1 & \text{if } (x_i \cdot x_{i-1}) < 0 \\ 0 & \text{else} \end{cases}$$

(d) Median absolute deviation (MAD)

$$X_{\text{MAD}} = \underset{i}{\text{median}} \left(\left| x_i - \underset{j}{\text{median}}(x_j) \right| \right)$$

Hint: numpy has a built-in numpy.median() function!

(e) Mean Absolute Deviation (MEAN-AD):

$$X_{\text{MEAN_AD}} = \frac{1}{N} \sum_{i=0}^{N-1} \left(\left| x_i - \frac{1}{N} \sum_{j=0}^{N-1} x_j \right| \right)$$

• Output the data to a comma separated value (CSV) text file in the format:

. . .

 $filename 128, RMS 128, PAR 128, AC 128, MAD 128, MEAN_AD 128 \label{eq:mean_add_action} \\$

Concretely, the beginning and ending of the file should be:

music_wav/bagpipe.wav,0.063492,8.149929,0.191660,0.031769,0.044588 music_wav/ballad.wav,0.029699,7.320233,0.039395,0.012695,0.018975

speech_wav/voice.wav,0.070688,4.163124,0.082435,0.031982,0.049338

To pass our automated grading system, the format of your file must match this exactly. The order of filenames must match the order in the music_speech.mf file.

2. Upload your CSV file, and source code to **Assignment 1a** at:

http://cs4347.smcnus.org

This will automatically grade the values you calculated. If any mistake is found, please check your program and resubmit – you are welcome to submit as many versions as you wish before the submission deadline.

Submit a zip file containing your program's source code files (as a .py file), the CSV file, and an optional README.txt file to the same website.

- 3. Make a copy of the previous program. Modify it such that it will:
 - Read the ground-truth music_speech.mf file
 - Load each way file (divide by 32768.0) and split the data into buffers of length 1024 with 50% overlap (or a hopsize of 512). Only include complete buffers; if the final buffer has 1020 samples, omit that buffer.

Hint: the starting and ending indices for the first few buffers are:

Buffer number	start index	end index
		(not included in array)
0	0	1024
1	512	1536
2	1024	2048

I recommend that you use the numpy "array slice" feature:

for i in range(num_buffers):
 start = ...
 end = ...

buffer_data = whole_file_data[start:end]

This should give you 1290 buffers of length 1024. If you are comfortable with linear algebra and numpy, the program will run significantly faster if you store this as a single 1290x1024 matrix and avoid any loops during the feature calculations.

• Calculate 5 features for each buffer: RMS, PAR, ZCR, MAD, MEAN_AD. This means that N=1024 in the equations, instead of the previous N=661500. You should end up with 1290 feature vectors of length 5, or a single 1290x5 matrix.

- After calculating the features for each buffer, calculate the mean and uncorrected sample standard deviation for each feature over all buffers for each file. You should end up with a single feature vector of length 10, or a 1x10 matrix for each file.
- For a later assignment we will be using a tool which reads ARFF files. Output the data to a new ARFF file with the header:

```
ORELATION music_speech
OATTRIBUTE RMS_MEAN NUMERIC
OATTRIBUTE PAR_MEAN NUMERIC
OATTRIBUTE ZCR_MEAN NUMERIC
OATTRIBUTE MAD_MEAN NUMERIC
OATTRIBUTE MEAN_AD_MEAN NUMERIC
OATTRIBUTE MEAN_AD_MEAN NUMERIC
OATTRIBUTE RMS_STD NUMERIC
OATTRIBUTE PAR_STD NUMERIC
OATTRIBUTE ZCR_STD NUMERIC
OATTRIBUTE MAD_STD NUMERIC
OATTRIBUTE MEAN_AD_STD NUMERIC
OATTRIBUTE MEAN_AD_STD NUMERIC
OATTRIBUTE class {music, speech}

The data lines should be:

ODATA
ORDATA
```

Concretely, the QDATA section should be:

```
@DATA 0.057447,3.234547,0.191595,0.037308,0.044607,0.027113,0.397827,0.036597,0.018317,0.021898,music 0.026583,2.590328,0.039416,0.016488,0.018977,0.013284,0.384978,0.015575,0.011143,0.012273,music ... 0.062831,2.822102,0.082504,0.042271,0.049270,0.032323,0.799688,0.070962,0.027540,0.029103,speech
```

4. Upload your ARFF files, and source code to **Assignment 1b** at:

```
http://cs4347.smcnus.org
```

This will automatically grade the values you calculated. If any mistake is found, please check your program and resubmit – you are welcome to submit as many versions as you wish before the submission deadline.

Submit a zip file containing your program's source code files (as a .py file), the ARFF file, and an optional README.txt file to the same website.

• You may use anything in the python standard library, numpy (including pylab / matplotlib), and scipy libraries. No other libraries are permitted.

If you are familiar with python and understood the lecture, this should take about 2 hours. Grading scheme:

- 2/6 marks: correct CSV file (automatically graded by computer).
- 2/6 marks: correct ARFF file (automatically graded by computer).
- 2/6 marks: readable source code (good variable names, clean functions, comments when needed).