



AVIPULSE

PROJECT REPORT

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Web Bird Detector

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1 Task

The task of this project was to create a working web-based tool which can input a bird call and implement the developed sound processing algorithm to identify the bird species with probability.

2 Methodology and Plan

The original code that trains a Naive Bayes learner and calculates bird probability for an unknown sample after deriving parameters from the sound based using Short Time Fourier Transform was implemented in MATLAB. To develop a web application, we would need to invoke MATLAB CGI which is not known for being efficient or easy to implement. Rather, we could try to convert the MATLAB code to Python.

2.1 Pros and Cons

- The conversion requires extra effort. This is justifiable given better compatibility of Python with web technologies and more developer support for libraries, especially in the field of computational sciences.
- Python is infamous for its poor ability to handle floating points. This difficulty is tackled by using numpy, a mathematical package for Python, and the fact that we maintain variables upto 4 variables only in the original algorithm keeps the accuracy up to the mark.
- Python is a lot slower than MATLAB for numerical computations.
- The whole application will have heavy library dependencies. If we use a compiled Python script, this is not really a problem.

2.2 Keynotes

- Use Gaussian Naive Bayes class in Scilabs package to create machine learner
- Use scikits.audiolabs and numpy to process audio file for feature extraction
- Use xlrd to read excel files

3 The GNB Learner

3.1 Code

This section describes the code of Gaussian Naive Bayes learner and its usage. The code is supposed to be self-explanatory.

```
1  #!/usr/bin/env/python
2
3  #Project: AviPulse
4  #Author: Riddhish et. al. (matlab)
5  #Translation: Kumar Ayush (python)
6  #Methodology: You need an excel file with all input parameters. The pr
7
8  #to read excel files
9  import xlrd
10
11 #numpy
12 import numpy as np
13
14 #Naive Bayes Learner Class
15 from sklearn.naive_bayes import GaussianNB as GNB
16
17 #to save model state into a file
18 from sklearn.externals import joblib
19
20 #open workbook and load a specific sheet by name
21 workbook = xlrd.open_workbook('data2.xlsx') #open excel file
22 datasheet = workbook.sheet_by_name('data_without_outliers_manual')
23
24 tmp2 = []
25 tmp3 = []
26
27 #load values in arrays
28 for i in range(1,datasheet.nrows):
29     tmp2.append(datasheet.cell_value(i,6))
30     tmp3.append([datasheet.cell_value(i,j) for j in range(2,6)])
31
32 #numpyfy arrays, prepare for processing
33 data = np.array(tmp3)
34 target = np.array(tmp2)
35
```

```

36 #the test data
37 testsheet = workbook.sheet_by_name('test_data')
38 tmp = []
39 for i in range(1, testsheet.nrows):
40     tmp.append([testsheet.cell_value(i, j) for j in range(2, 6)])
41
42 testdata = np.array(tmp)
43
44 #create a learner class
45 gnb = GNB()
46
47 #fit it
48 y_fit = gnb.fit(data, target)
49
50 #get prediction for testdata
51 y_pred = y_fit.predict(testdata)
52
53 #get model score
54 score = y_fit.score(data, target)
55
56 #Print output
57 print "Prediction is Bird Number %d and score is %f" %(y_pred[0], score)
58
59 #save model state onto file
60 joblib.dump(y_fit, 'model.pkl')

```

3.2 Usage

data2.xlsx is attached with this document. You can witness the format in which data is to be fed. The model is trained and the state of the model is stored in a pickle file called *model.pkl*. Generally, you would no longer need the database, but the model state only relates bird-ids and not their names. So, you would need *data2.xlsx* for bird identification to produce names from matched bird id. To train the model, just run the file on using

`python preproc.py`

IMP : Make sure that *data2.xlsx* is in the same folder as the python script

4 The Identifier

4.1 Code

This code is more complicated owing to the mathematical algorithms it uses. It takes multipart form data, extracts features from it, predicts a bird label and outputs the bird name. As in the previous case, the code is supposed to be self-explanatory, except during the STFT computation where textual knowledge of signal processing algorithms is expected.

```
1  #!/usr/bin/env python
2
3  #Project: AviPulse
4  #Author: Riddhish et. al. (matlab)
5  #Translation: Kumar Ayush (python)
6  #Methodology: You need a web form to post a syllable file in .wav form
7
8  #import CGI libraries
9  import cgi
10 import cgitb
11
12 #numpy
13 import numpy as np
14
15 #to read a wave
16 from scikits.audiolab import wavread
17
18 #Python math library
19 import math
20
21 #To read an excel file
22 import xlrd
23
24 #The Gaussian Naive Bayes Machine Learner Class
25 from sklearn.naive_bayes import GaussianNB as GNB
26
27 #To load the saved model state
28 from sklearn.externals import joblib
29
30 #Enable errors
31 #use log directory for logdir, set display=0 if don't want error display
32 cgitb.enable(display=1, logdir="/home/cheeku/log")
```

```

33
34 #Define the cgi form object
35 form = cgi.FieldStorage()
36
37 #HTML headers
38 print "Content-type:_text/html\n\n"
39
40 #Write the multipart sound data to a temporary file
41 tmpf = open("tmp.wav","wb")
42 tmpf.write(form['wav'].value)
43 tmpf.close()
44
45 #Variable initializations
46 scmed = 0          #Spectral Centroid Median
47 scmean = 0         #Spectral Centroid Mean
48 scmax = 0          #Spectral Centroid Max
49 scmin = 0          #Spectral Centroid Min
50 sfluxmed = 0       #Spectral Flux Median
51 sfluxmean = 0      #Spectral Flux Mean
52 intmode = 0
53 pitmean = 0        #Pitch Mean
54 pitmed = 0         #Pitch Median
55 flatmean = 0       #Spectral Flatness Mean
56 flatmed = 0        #Spectral Flatness Median
57
58 #Read the sound, stored in the temporary file
59 #'s' is data,'enc' is encoding
60 s,fs,enc = wavread("tmp.wav")
61 s = np.array(s)
62 s = s/max(abs(s))   #normalization , to imitate matlab's wavread
63
64 Frame_size=20.0     #Parameter of computation
65 Frame_shift = 10.0  #Parameter of computation
66
67 #Variable recasts and normalizers , prep for computation
68 y = s
69 reconsine = y
70 Frame_size = Frame_size/1000.0
71 Frame_shift = Frame_shift/1000.0
72
73 #Empty lists

```

```

74 secondpeak = []
75 scarr = []           #Spectral Centroid Array
76 spcroll = []
77 spflux = []         #for Spectral Flux calculaion
78 spcrest = []
79 spflatness = []     #for Spectral Flatness calculation
80 spspread = []
81 amplitude = []      #for Amplitude calculation
82 pitch = []
83 pitch1 = []         #for Pitch calculations
84 dip_bin = []
85 dip_amp = []
86 max_amp_bin = []
87 max_amp = []
88 peaks = []
89 peakpeak = []
90 max2_amp = []
91 max2_bin = []
92 atimbre = []
93
94 #Re-normalization
95 max_value = max(abs(y))
96 y = y/max_value
97
98 #Scale Frame size and length
99 Frame_length = Frame_size*fs
100 sample_shift = Frame_shift*fs
101
102 w = np.hamming(Frame_length)    #create a hamming window of given length
103
104 dftylast = 0    #empty variable to store dft result for previous computation
105 for i in range(int(math.floor(len(y)/sample_shift)-math.ceil(Frame_length/sample_shift)),int(len(y)/sample_shift)):
106     #Fourier Transform with data scaled using a sliding hamming window
107     k,jj = 0,0
108     yy = []
109     yyy = []
110     for j in range(int(i*sample_shift),int(i*sample_shift+Frame_length)):
111         yy.insert(k,y[j]*w[jj])
112         yyy.insert(k,y[j])
113         jj,k = jj+1,k+1
114     dfty = abs(np.fft.fft(yy))

```



```

115 yy = np.array(yy)
116 dftyp = []
117 for it in range(len(yy)):
118     dftyp.append(math.atan2(yy[it].imag,yy[it].real))
119
120 #computation , computation , computation
121 scn ,scd ,add ,sf ,ismax = 0,0,0,0,0
122 q,sctimbre ,geo ,jj = 0,0,1,0
123 M = len(dftyp)/2
124 for p in range(M):
125     scn = scn + (p+1)*dftyp[p]*dftyp[p]
126     scd = scd + dftyp[p]*dftyp[p]
127     add = add + dftyp[p]
128     geo = geo*dftyp[p]
129     if dftyp[p]>ismax:
130         ismax = dftyp[p]
131     else:
132         ismax = ismax
133     if p>0 and p<M-1:
134         if dftyp[p]>dftyp[p-1] and dftyp[p]>dftyp[p+1]:
135             peaks.insert(q,[])
136             peaks[q].insert(1,p)
137             peaks[q].insert(0,dftyp[p])
138             peaks[q].insert(2,dftyp[p])
139             sctimbre = sctimbre + dftyp[p]
140             q = q+1
141     if i>0:
142         sf = sf + (dftyp[p]-dftyp[p-1])*(dftyp[p]-dftyp[p+1])
143     else:
144         sf = 0
145
146 #convolutions and computations
147 s = 0
148 length = len(yyy)
149 acf_clip = np.correlate(yyy[:len(yyy)/2],yyy[:len(yyy)/2],"full")
150 lenacf_clip = len(acf_clip)
151 max_acfclip ,max_acfclip_bin = max(acf_clip),np.argmax(acf_clip)
152 for acfclip_bin in range(max_acfclip_bin ,lenacf_clip-1):
153     if acf_clip[acfclip_bin-1]>=acf_clip[acfclip_bin] and
154         dip_bin = np.append(dip_bin , acfclip_bin)
155         dip_amp = np.append(dip_amp , acf_clip[acfclip_bin])

```

```

156
157     dip_bin = np.array(dip_bin)[np.newaxis].T           #row to column
158     dip_amp = np.array(dip_amp)[np.newaxis].T           #row to column
159     dipMin_amp, Min_bin = min(dip_amp), np.argmin(dip_amp)
160     dipMin_bin = dip_bin[Min_bin]
161     max2_amp.insert(i, max(acf_clip[int(dipMin_bin):lenacf_clip]))
162     max2_bin.insert(i, np.argmax(acf_clip[int(dipMin_bin):lenacf_clip]))
163     max2_bin[i] = max2_bin[i] + dipMin_bin - 1 - max_acfclip_bin
164     if i == 0:
165         max2_bin[i] = max2_bin[i]
166     else:
167         max2_bin[i] = (max2_bin[i-1] + max2_bin[i]) / 2.0
168     pitch1 = np.append(pitch1, fs * (1.0 / (max2_bin[i] + 1)))
169
170     for r in range(1, q-1):
171         if peaks[r][0] > peaks[r-1][0] and peaks[r][0] > peaks[r+1][0]:
172             peakpeak.insert(s, [])
173             peakpeak[s].insert(1, peaks[r][1])
174             peakpeak[s].insert(0, peaks[r][0])
175             peakpeak[s].insert(2, peaks[r][2])
176             s = s+1
177
178     np.sort(peakpeak, axis=0)           #sort ascending
179     peakpeaksorted = np.flipud(peakpeak)   #reverse, effectively
180     np.sort(peaks, axis=0)
181     peakssorted = np.flipud(peaks)
182     energy = float(add)/M
183     sc = float(scn)/scd
184     ssn = 0
185     sctimbre = sctimbre - q*peaks[1][1]
186     for p in range(M):
187         ssn = ssn + (p+1-sc)*(p+1-sc)*dfty[p]
188     ss = float(ssn)/scd
189     ss = ss**0.5
190     geo = geo**((1.0/M))
191
192     sfl = float(geo)/energy
193     spc = float(ismax)/energy
194     add = 0.85*add
195     scc = 0
196     for p in range(M):

```

```

197         scc = scc + dfty[p]
198         if scc >= add:
199             break
200
201     dftylast = dfty #current dfty to dftylast assignment
202
203     #populate result arrays
204     amplitude.insert(i, 20 * math.log10(add / 0.85))
205     pitch.insert(i, peakpeaksorted[0][1] / Frame_size)
206     secondpeak.insert(i, peakpeaksorted[1][1] / Frame_size)
207     spcroll.insert(i, scc)
208     scarr.insert(i, sc)
209     spflux.insert(i, sf)
210     spcrest.insert(i, spc)
211     spflatness.insert(i, sfl)
212     spspread.insert(i, ss)
213     atimbre.insert(i, sctimbre)
214
215     #Since we have no good way of seperating syllables in a sound file
216     #the following code gets the length of the file in seconds
217     #assuming it is a single syllable
218     import wave
219     import contextlib
220     fname = 'tmp.wav'
221     with contextlib.closing(wave.open(fname, 'r')) as f:
222         frames = f.getnframes()
223         rate = f.getframerate()
224         duration = frames / float(rate)
225
226     tmp = [duration, np.median(pitch1), np.median(scarr), np.median(spflux)]
227     testdata = np.array(tmp) #prepare testdata
228     y_fit = joblib.load('model.pkl') #load model from file
229     y_pred = y_fit.predict(testdata) #get predicted label
230
231     #Print predicted label
232     print "Prediction:_"
233     #print y_pred[0]
234
235     #open workbook and load relevant sheet
236     workbook = xlrd.open_workbook('data2.xlsx')
237     datasheet = workbook.sheet_by_name('data_without_outliers_manual')

```

```

238
239 #print name of bird based on id
240 i = 0
241 while datasheet.cell_value(i,6) != int(y_pred[0]):
242     i = i+1
243
244 print datasheet.cell_value(i,0)
245
246 #close the output
247 print "</html>"

```

4.2 Usage

Create a form that POSTS multipart form data in wave file format to the above python script. Make sure *data2.xlsx* is in the same folder as this file.

5 Dependencies

- **xlrd** - <https://pypi.python.org/pypi/xlrd>
- **scikits** - <http://scikit-learn.org/stable/>
- **numpy** - <http://www.numpy.org/>
- **scilabs** - <http://www.scipy.org/topical-software.html>

6 Limitations

The application currently supports single syllable sound files in .wav format. This is mainly because of failure in search for an algorithm that separates syllables efficiently and easily.

7 Learnings

- MATLAB indices start from 1, and Python indices start from 0, much like most of scripts and languages. This was responsible for most of the bugs in the program
- Gained a better understanding of Fourier Transforms and correlation algorithms.

- Naive Bayes is a simple algorithm. I was not initially convinced of its performance. The algorithm you use depends on the situation, and not on the complexity.
- Learnt basics of MATLAB. Being based out of an F77 compiler, it's not really friendly to the thought processes of a regular modern programmer, but it is useful for scientific purposes. It really excels most of other scripts when numerous numerical computations are a necessity.

8 Extending Remarks

Due to insufficient server permissions, it was not possible to install dependencies on the server. This problem has a possible solution. The python code can be compiled into an .exe file, which runs without any dependencies. This will be explored soon.