

Amath 482 homework 4: Music Classification

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Abstract

This report demonstrates the idea of Machine Learning on music classification, which is separated into 3 cases: Band Classification with different music genres, Band Classification with the same genre, and Genre Classification. To accomplish such goal, we are going to do spectrogram transform to our data first, then apply Principal Component Analysis (PCA) for dimension reduction and Linear Discriminant Analysis (LDA) to statistically determine threshold values. Finally, the result will be applied to new data to test the algorithm.

1 Introduction

As described in the report from Homework 3, PCA is an useful technique to extract lower-dimension data from the dynamic. Since machine learning commonly deals with a bunch of similar data, applying PCA makes it much more efficient (referring to energy captured by each mode) and provides a matrix U that we can project our data onto.

LDA is based on the idea of projecting two sets of data onto a new bases that separate the inter-class distance to the max extent while making the intra-class data as compact as possible. The corresponding eigenvector of the maximum eigenvalue between the within-class variance S_W and between-class variance S_B will give us a good one to project onto.

In this report, we are going to apply these techniques to several 5-seconds music clips for classification. In order to guarantee the success rate, different amounts of clips might be chosen for different cases, since it would definitely be harder to classify three artists performing the same genre than those performing different genres.

2 Theoretical Background

2.1 Spectrogram transform

The spectrogram is the visual representation of the spectrum of frequencies along the signal verses time. We can extract time-frequency component of the

music clips by simply using the Matlab built-in function:

$$\text{spectrogram}(\text{data}(:, k)) \quad (1)$$

and then rearranging it into a single column vector that represents one specific music clip.

2.2 Principal Component Analysis (PCA)

Since we're dealing with similar spectrogram data that represent the frequency-time components of music clips, it's necessary and feasible to do PCA that helps dimension reduction, and thus making the algorithm much more efficient. PCA is based on the idea of singular value decomposition, as described in the report of **Homework 3**.

2.3 Linear Discriminant Analysis (LDA)

LDA is a statistic method that can classify different groups to the max extent. In this case of music classification, we have three groups of data, thus, we should apply formula of multi-group between-class variance S_B by:

$$S_B = \sum_{j=1}^n m_j (\vec{\mu}_j - \vec{\mu})(\vec{\mu}_j - \vec{\mu})^T \quad (2)$$

, where n denotes the number of groups and $\vec{\mu}$ is the mean of all $\vec{\mu}_1$ to $\vec{\mu}_n$. m_j is the size of each group. In this case, we're going to use same amount of clips from each artist, so we can omit it when calculating S_B .

The within-class variance, S_W can be computed by:

$$S_W = \sum_{j=1}^n \sum_{\vec{x}} (\vec{x} - \vec{\mu}_j)(\vec{x} - \vec{\mu}_j)^T \quad (3)$$

, where \vec{x} represents each column vector of the projection onto principal component matrix for each group of the data.

The projection vector w is defined as:

$$w = \arg \max_w \frac{w^T S_B w}{w^T S_W w} \quad (4)$$

We can interpret it as the vector w that maximizes the relation of S_B over S_W . To find what the projection basis, w exactly is, we can simply find the maximum eigenvalue and the corresponding eigenvector between $S_B w$ and $S_W w$, computed by:

$$S_B \vec{w} = \lambda S_W \vec{w} \quad (5)$$

Finally, we project the data onto vector w and then need to find threshold values that help to distinguish between groups. We are going to choose the boundary values from each group and taking average of the closest two as thresholds. Since we are dealing with three data groups, two threshold values are then needed.

3 Algorithm Implementation and Development

Prepare data

To start with each case, I first load the .wav files as data for machine learning. Since the clips I choose are stereos, I then take the average and make them mono types. I set up a function called **prepareTest** that's able to do the procedures above.

Do Spectrogram transform

Another function called **doSpec** is set up for this part. First, it will spectrogram transform each column vector of the given data, taking the absolute value, and then resize the result from a $m \times n$ matrix into a $m \times n$ by 1 column vector. This part ends up by arranging each column vector into a matrix called **spec**, with the size of $m \times n$ by the amount of .wav files chosen.

Training the given data

First, we are going to choose an appropriate amount of features for dimension reduction by referring to the Energy-Mode plots for each of the three tests. Using the matrix **spec** and given feature number, we then do SVD to **spec** and choose feature amounts of columns of U and feature amounts of rows of the principal component projection matrix $S \cdot V^T$. $1/3$ of the columns of $S \cdot V^T$ are then distributed to corresponding artist. Using the mean-of-rows vector $\vec{\mu}_j$ of each matrix, we are able to perform LDA that generate three sorted projection vectors that are used for finding 2 threshold values. This part is also stored in a function called **trainer**.

Test the algorithm on new data

To test the success rate of the algorithm, I choose one more clip of each used music for machine learning and two new clips that's not used for the algorithm. I then compute the exact labels for each of the music clip (3 being the artist with the largest mean value of the sorted projection vector, and 1 being the one with the least mean). To have the label generated by the algorithm, I first compute the spectrogram and project it onto matrix U and vector w following then. Naming the projection vector **pval**, I then call $label = (pval > threshold1) + (pval > threshold2) + 1$. Subtracting this label vector from correct labels gives us a vector of either 0 or not. If an entry is not 0, that means we fail to classify that clip. Using functions **check** generates the label and **pval**, and **checkRate** gives us the success rate in each test.

4 Computational Results

Test 1 - Band classification with different genres

Three artists, Dan Bodan, Density & Time, Dan Lebowitz are chosen for this test, each performing classical, electronics, and folk musics. 9 clips are chosen from each artist (3 clips per music piece). While for the test data, 5 clips are chosen from each, with a fourth clip from each of the three pieces used as algorithm data and 2 clips from new pieces. 8 features are applied, which results in a success rate of 0.8667, that is, 13/15 being correct. What's wrongly classified are two clips from Dan Bodan's "National Express" and "Kallimachou", and they are mis-classified to be folk music from Dan Lebowitz. The reason might be that, these two pieces are composed with instruments that generate high-frequency sounds. Since I decrease the sampling rate to 11025 Hz, they may then have similar pitches as folk musics that are mostly played by guitars.

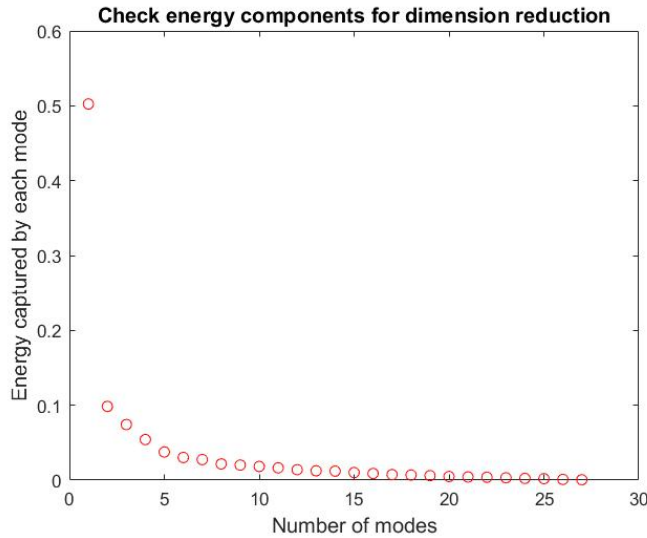


Figure 1: Energy captured by each mode in Test 1

Test 2 - Band classification with the same genre

The jazz & blues music pieces, from Quincas Moreira, Chris Haugen, and Freedom Trail Studio are used for this test. 4 clips from each of 4 pieces are chosen from each artist, which makes a total of 48 algorithm data. 6 clips from each artist (another 4 from the music pieces that are used as algorithm data and 2 from new pieces) are used for testing. In this case, I apply 20 features and the success rate is 0.7222, which means 13 out of 18 are rightly classified. The result tells us that 3 clips from Quincas Moreira and 2 from Freedom trauk

studio are thought to be performed by Chris Haugen. In this case, this makes sense since jazz or blues are composed with similar chords.

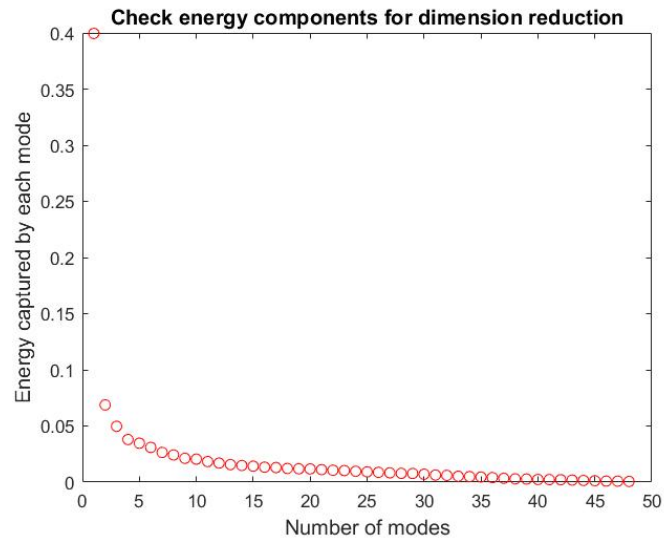


Figure 2: Energy captured by each mode in Test 2

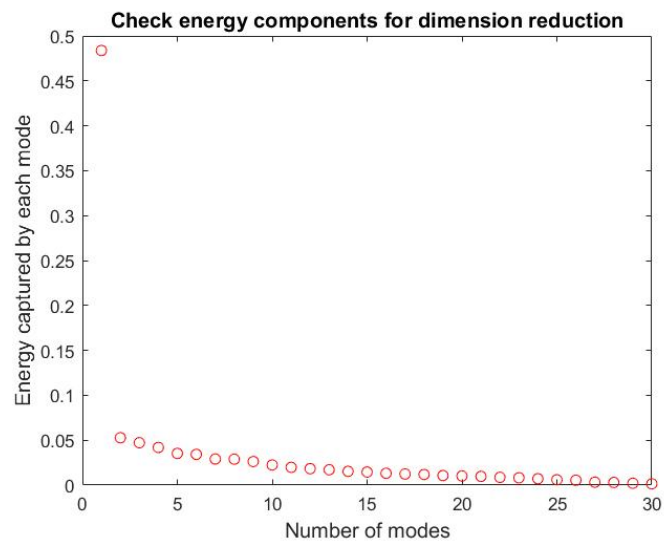


Figure 3: Energy captured by each mode in Test 3

Test 3 - Genre classification

In this case, 10 music clips are randomly chosen in the fields of Jazz & Blues, Electronics, and Punk. The test data are 4 new clips from each of the three genres. 8 features are used for the algorithm, yielding 0.8333 success rate, 10/12 being correct. One of jazz and one of electronics are thought to be punk, these two clips might also be wrongly classified due to the decreased sampling rate.

5 Summary and Conclusion

To sum up, we would see a much higher success rate when dealing with Band Classification (different genres) and Genre Classification than classifying different bands performing same genre of musics. We may also see a increase in all of the tests if we apply more input data and a higher sampling rate, however, this would also make our algorithm much slower, i.e, the tradeoff between efficiency and accuracy should be considered in the field of machine learning.

6 Appendix A

function spec = doSpec(data) returns the spectrogram of data with the right size.

function file = prepareTest(clip) reads stereo audio files and converts them to monos as a column vector.

function sucRate = checkRate(M1, M2, M3, label) checks means of each projection basis onto w and provides us a correct label. It also gives us a success rate determined by correct label and label determined by the algorithm.

function [label, pval] = check(U, w, threshold1, threshold2, spec) takes two threshold values, U and w for projections, and the spectrogram of test data and returns generated label and **pval**.

function [U, S, V, threshold1, threshold2, w, M1, M2, M3] = trainer(sp, feature) is the training function for machine learning by performing PCA and LDA.

S = spectrogram(data) produces the spectrogram by using STFT.

B = sort(A) sorts A in ascending order.

C = setdiff(A,B, 'rows') returns the values in A that are not in B with no repetitions and C will be sorted in rows.

7 Appendix B

```
1 %%
2 % Test 1: Band Classification (different genres)
3 clear; close all; clc;
4
5 % Dan Bodan
6 x1 = audioread('Anton1.wav');
7 x2 = audioread('Anton2.wav');
8 x3 = audioread('Anton4.wav');
9 x4 = audioread('City4.wav');
10 x5 = audioread('City2.wav');
11 x6 = audioread('City3.wav');
12 x7 = audioread('National1.wav');
13 x8 = audioread('National2.wav');
14 x9 = audioread('National3.wav');
15
16 x = [x1, x2, x3, x4, x5, x6, x7, x8, x9];
17 for i = 1 : 9
18     x(:, i) = (x(:, 2*i - 1) + x(:, 2*i)) / 2;
19 end
20 x = x(:, 1 : end/2);
21
22
23
24 % Density & Time
25 y1 = audioread('Clean1.wav');
26 y2 = audioread('Clean2.wav');
27 y3 = audioread('Clean3.wav');
28 y4 = audioread('Dream4.wav');
29 y5 = audioread('Dream2.wav');
30 y6 = audioread('Dream3.wav');
31 y7 = audioread('Day4.wav');
32 y8 = audioread('Day1.wav');
33 y9 = audioread('Day3.wav');
34
35 y = [y1, y2, y3, y4, y5, y6, y7, y8, y9];
36 for i = 1 : 9
37     y(:, i) = (y(:, 2*i - 1) + y(:, 2*i)) / 2;
38 end
39 y = y(:, 1 : end/2);
40
41 % Dan Lebowitz
42 z1 = audioread('Rag3.wav');
43 z2 = audioread('Rag1.wav');
```

```

44 z3 = audioread('Rag4.wav');
45 z4 = audioread('Opening4.wav');
46 z5 = audioread('Opening2.wav');
47 z6 = audioread('Opening1.wav');
48 z7 = audioread('Birds1.wav');
49 z8 = audioread('Birds2.wav');
50 z9 = audioread('Birds3.wav');
51
52 z = [z1, z2, z3, z4, z5, z6, z7, z8, z9];
53 for i = 1 : 9
54     z(:, i) = (z(:, 2*i - 1) + z(:, 2*i)) / 2;
55 end
56 z = z(:, 1 : end/2);
57
58 Test1 = [x, y, z];
59 spec = doSpec(Test1);
60 feature = 8;
61
62 [U, S, V, threshold1, threshold2, w, M1, M2, M3] =
    trainer(spec, feature);
63
64 file(:, 1) = prepareTest('Anton3.wav');
65 file(:, 2) = prepareTest('City1.wav');
66 file(:, 3) = prepareTest('National4.wav');
67 file(:, 4) = prepareTest('Kallimachou1.wav');
68 file(:, 5) = prepareTest('Kallimachou2.wav');
69 file(:, 6) = prepareTest('Clean4.wav');
70 file(:, 7) = prepareTest('Dream1.wav');
71 file(:, 8) = prepareTest('Day2.wav');
72 file(:, 9) = prepareTest('Castlevania1.wav');
73 file(:, 10) = prepareTest('Castlevania2.wav');
74 file(:, 11) = prepareTest('Rag2.wav');
75 file(:, 12) = prepareTest('Opening3.wav');
76 file(:, 13) = prepareTest('Birds4.wav');
77 file(:, 14) = prepareTest('Cliffsides1.wav');
78 file(:, 15) = prepareTest('Cliffsides2.wav');
79
80
81 spec = doSpec(file);
82 [label, pval] = check(U, w, threshold1, threshold2, spec)
    ;
83 sucRate = checkRate(M1, M2, M3, label);
84
85
86 %%
87 % Test 2: Band Classification (same genre, Jazz & Blues)

```



```

88 clear all; close all; clc;
89
90 % Quincas Moreira
91 file(:, 1) = prepareTest('Brooklin1.wav');
92 file(:, 2) = prepareTest('Brooklin2.wav');
93 file(:, 3) = prepareTest('Brooklin3.wav');
94 file(:, 4) = prepareTest('Brooklin4.wav');
95 file(:, 5) = prepareTest('Infusion1.wav');
96 file(:, 6) = prepareTest('Infusion2.wav');
97 file(:, 7) = prepareTest('Infusion3.wav');
98 file(:, 8) = prepareTest('Infusion4.wav');
99 file(:, 9) = prepareTest('Malan1.wav');
100 file(:, 10) = prepareTest('Malan2.wav');
101 file(:, 11) = prepareTest('Malan3.wav');
102 file(:, 12) = prepareTest('Malan4.wav');
103 file(:, 13) = prepareTest('Paradox1.wav');
104 file(:, 14) = prepareTest('Paradox2.wav');
105 file(:, 15) = prepareTest('Paradox3.wav');
106 file(:, 16) = prepareTest('Paradox4.wav');
107
108
109 i = 16;
110 % Chris Haugen
111 file(:, 1+i) = prepareTest('Bleeker1.wav');
112 file(:, 2+i) = prepareTest('Bleeker2.wav');
113 file(:, 3+i) = prepareTest('Bleeker3.wav');
114 file(:, 4+i) = prepareTest('Bleeker4.wav');
115 file(:, 5+i) = prepareTest('Et1.wav');
116 file(:, 6+i) = prepareTest('Et2.wav');
117 file(:, 7+i) = prepareTest('Et3.wav');
118 file(:, 8+i) = prepareTest('Et4.wav');
119 file(:, 9+i) = prepareTest('Old1.wav');
120 file(:, 10+i) = prepareTest('Old2.wav');
121 file(:, 11+i) = prepareTest('Old3.wav');
122 file(:, 12+i) = prepareTest('Old4.wav');
123 file(:, 13+i) = prepareTest('Sunshine1.wav');
124 file(:, 14+i) = prepareTest('Sunshine2.wav');
125 file(:, 15+i) = prepareTest('Sunshine3.wav');
126 file(:, 16+i) = prepareTest('Sunshine4.wav');
127
128
129 % Freedom trauck studio
130 i = 32;
131 file(:, 1+i) = prepareTest('Crazy1.wav');
132 file(:, 2+i) = prepareTest('Crazy2.wav');
133 file(:, 3+i) = prepareTest('Crazy3.wav');

```

```

134 file(:, 4+i) = prepareTest('Crazy4.wav');
135 file(:, 5+i) = prepareTest('Current1.wav');
136 file(:, 6+i) = prepareTest('Current2.wav');
137 file(:, 7+i) = prepareTest('Current3.wav');
138 file(:, 8+i) = prepareTest('Current4.wav');
139 file(:, 9+i) = prepareTest('Elder1.wav');
140 file(:, 10+i) = prepareTest('Elder2.wav');
141 file(:, 11+i) = prepareTest('Elder3.wav');
142 file(:, 12+i) = prepareTest('Elder4.wav');
143 file(:, 13+i) = prepareTest('Mix1.wav');
144 file(:, 14+i) = prepareTest('Mix2.wav');
145 file(:, 15+i) = prepareTest('Mix3.wav');
146 file(:, 16+i) = prepareTest('Mix4.wav');
147
148
149 spec = doSpec(file);
150 feature = 20;
151
152 [U, S, V, threshold1, threshold2, w, M1, M2, M3] =
    trainer(spec, feature);
153
154 file2(:, 1) = prepareTest('Brooklin5.wav');
155 file2(:, 2) = prepareTest('Infusion5.wav');
156 file2(:, 3) = prepareTest('Malan5.wav');
157 file2(:, 4) = prepareTest('Paradox5.wav');
158 file2(:, 5) = prepareTest('Miles1.wav');
159 file2(:, 6) = prepareTest('Miles2.wav');
160
161 file2(:, 7) = prepareTest('Bleeker5.wav');
162 file2(:, 8) = prepareTest('Et5.wav');
163 file2(:, 9) = prepareTest('Front1.wav');
164 file2(:, 10) = prepareTest('Front2.wav');
165 file2(:, 11) = prepareTest('Old5.wav');
166 file2(:, 12) = prepareTest('Sunshine5.wav');
167
168 file2(:, 13) = prepareTest('Crazy5.wav');
169 file2(:, 14) = prepareTest('Current5.wav');
170 file2(:, 15) = prepareTest('Elder5.wav');
171 file2(:, 16) = prepareTest('Mix5.wav');
172 file2(:, 17) = prepareTest('Swing1.wav');
173 file2(:, 18) = prepareTest('Swing2.wav');
174
175 spec2 = doSpec(file2);
176 [label, pval] = check(U, w, threshold1, threshold2, spec2
    );
177 sucRate = checkRate(M1, M2, M3, label);

```

```

178
179
180
181
182
183 %%
184 % Test 3: Genre Classification
185 clear; close all; clc;
186
187 % Jazz & Blues
188 file(:, 1) = prepareTest('Et5.wav');
189 file(:, 2) = prepareTest('Miles2.wav');
190 file(:, 3) = prepareTest('Malan2.wav');
191 file(:, 4) = prepareTest('Malan3.wav');
192 file(:, 5) = prepareTest('Bleeker3.wav');
193 file(:, 6) = prepareTest('Swing2.wav');
194 file(:, 7) = prepareTest('Current3.wav');
195 file(:, 8) = prepareTest('Front2.wav');
196 file(:, 9) = prepareTest('Sunshine1.wav');
197 file(:, 10) = prepareTest('Swing1.wav');
198
199 i = 10;
200
201 % Electronics
202 file(:, 1+i) = prepareTest('Day3.wav');
203 file(:, 2+i) = prepareTest('Dream4.wav');
204 file(:, 3+i) = prepareTest('Clean2.wav');
205 file(:, 4+i) = prepareTest('Macaw1.wav');
206 file(:, 5+i) = prepareTest('Dragonfly1.wav');
207 file(:, 6+i) = prepareTest('Stranger2.wav');
208 file(:, 7+i) = prepareTest('Tea1.wav');
209 file(:, 8+i) = prepareTest('Tea2.wav');
210 file(:, 9+i) = prepareTest('Clean4.wav');
211 file(:, 10+i) = prepareTest('Macaw2.wav');
212
213 i = 20;
214
215 % Punk
216 file(:, 1+i) = prepareTest('Beach.wav');
217 file(:, 2+i) = prepareTest('Fail1.wav');
218 file(:, 3+i) = prepareTest('Fail2.wav');
219 file(:, 4+i) = prepareTest('Girl1.wav');
220 file(:, 5+i) = prepareTest('Girl2.wav');
221 file(:, 6+i) = prepareTest('Runner2.wav');
222 file(:, 7+i) = prepareTest('Pumps1.wav');
223 file(:, 8+i) = prepareTest('Radio1.wav');

```

```

224 file(:, 9+i) = prepareTest('Radio2.wav');
225 file(:, 10+i) = prepareTest('Runner1.wav');
226
227 spec = doSpec(file);
228 feature = 8;
229
230 [U, S, V, threshold1, threshold2, w, M1, M2, M3] =
    trainer(spec, feature);
231
232 file2(:, 1) = prepareTest('Infusion2.wav');
233 file2(:, 2) = prepareTest('Paradox5.wav');
234 file2(:, 3) = prepareTest('Elder3.wav');
235 file2(:, 4) = prepareTest('Old4.wav');
236
237 file2(:, 5) = prepareTest('Firefly1.wav');
238 file2(:, 6) = prepareTest('Cubic.wav');
239 file2(:, 7) = prepareTest('Operatic2.wav');
240 file2(:, 8) = prepareTest('Anniversary2.wav');
241
242 file2(:, 9) = prepareTest('Boy.wav');
243 file2(:, 10) = prepareTest('Give1.wav');
244 file2(:, 11) = prepareTest('Grassy.wav');
245 file2(:, 12) = prepareTest('Outlet1.wav');
246
247
248 spec2 = doSpec(file2);
249 [label, pval] = check(U, w, threshold1, threshold2, spec2
    );
250 sucRate = checkRate(M1, M2, M3, label);
251
252
253
254
255 %%
256 function spec = doSpec(data)
257     for k = 1 : size(data, 2)
258         temp = spectrogram(data(:, k));
259         temp = abs(temp);
260         [m, n] = size(temp);
261         spec(:, k) = reshape(temp, m * n, 1);
262     end
263 end
264
265 function file = prepareTest(clip)
266     temp = audioread(clip);
267     file = (temp(:, 1) + temp(:, 2))/2;

```

```

268 end
269
270
271 function sucRate = checkRate(M1, M2, M3, label)
272     Mean = [M1, M2, M3];
273     Mean = sort(Mean);
274     for i = 1 : 3
275         if Mean(i) == M1
276             ind1 = i;
277         end
278         if Mean(i) == M2
279             ind2 = i;
280         end
281         if Mean(i) == M3
282             ind3 = i;
283         end
284     end
285     correctLabel = [];
286     siz = length(label)/3;
287     for i = 1 : siz
288         correctLabel(i) = ind1;
289         correctLabel(i + siz) = ind2;
290         correctLabel(i + 2*siz) = ind3;
291     end
292
293     fail = 0;
294     result = correctLabel - label;
295     for i = 1 : length(result)
296         if result(i) ~= 0
297             fail = fail + 1;
298         end
299     end
300
301     sucRate = 1 - fail/length(result);
302 end
303
304 function [label, pval] = check(U, w, threshold1,
305     threshold2, spec)
306     TestMat = U' * spec;
307     pval = w' * TestMat;
308
309     label = (pval > threshold1) + (pval > threshold2) +
310         1;
311 end
312
313 function [U, S, V, threshold1, threshold2, w, M1, M2, M3]

```

```

        = trainer(sp, feature)
312 [U, S, V] = svd(sp, 'econ');
313
314 PCA_proj = S * V';
315 U = U(:, 1 : feature);
316
317 plot(diag(S).^2/sum(diag(S).^2), 'ro')
318 xlabel('Number of modes', 'FontSize', 12)
319 ylabel('Energy captured by each mode', 'FontSize', 12)
320 title('Check energy components for dimension reduction',
        'FontSize', 12)
321
322 Art1 = PCA_proj(1 : feature, 1 : end/3);
323 Art2 = PCA_proj(1 : feature, end/3 + 1 : end * 2/3);
324 Art3 = PCA_proj(1 : feature, end * 2/3 + 1 : end);
325
326 m1 = mean(Art1, 2);
327 m2 = mean(Art2, 2);
328 m3 = mean(Art3, 2);
329
330 Sw = 0;
331 for k = 1 : size(Art1, 2)
332     Sw = Sw + (Art1(:, k) - m1) * (Art1(:, k) - m1)';
333 end
334 for k = 1 : size(Art2, 2)
335     Sw = Sw + (Art2(:, k) - m2) * (Art2(:, k) - m2)';
336 end
337 for k = 1 : size(Art3, 2)
338     Sw = Sw + (Art3(:, k) - m3) * (Art3(:, k) - m3)';
339 end
340
341 mu = (m1 + m2 + m3) / 3;
342 Sb = 0;
343 Sb = Sb + (m1 - mu) * (m1 - mu)';
344 Sb = Sb + (m2 - mu) * (m2 - mu)';
345 Sb = Sb + (m3 - mu) * (m3 - mu)';
346
347 [V2, D] = eig(Sb, Sw);
348 [~, ind1] = max(diag(D));
349
350 w = V2(:, ind1);
351 w = w / norm(w, 2);
352
353 vArt1 = w' * Art1;
354 vArt2 = w' * Art2;
355 vArt3 = w' * Art3;

```

```

356
357 M1 = mean(vArt1);
358 M2 = mean(vArt2);
359 M3 = mean(vArt3);
360
361 sortArt1 = sort(vArt1);
362 sortArt2 = sort(vArt2);
363 sortArt3 = sort(vArt3);
364 sortArt = [sortArt1; sortArt2; sortArt3];
365
366 Mean = [M1, M2, M3];
367 Mmax = max([M1, M2, M3]);
368
369 IndMax = Mean == Mmax;
370 sortMax = sortArt(IndMax, :);
371
372 sortArt = setdiff(sortArt, sortMax, 'rows');
373
374
375 sortNext = sortArt(2, :);
376 sortLeast = sortArt(1, :);
377
378
379 sizeMax = length(sortArt1);
380 t1 = sizeMax;
381 t2 = 1;
382 while sortLeast(t1) > sortNext(t2)
383     t1 = t1 - 1;
384     t2 = t2 + 1;
385 end
386 threshold1 = (sortLeast(t1) + sortNext(t2)) / 2;
387
388 t3 = sizeMax;
389 t4 = 1;
390 while sortNext(t3) > sortMax(t4)
391     t3 = t3 - 1;
392     t4 = t4 + 1;
393 end
394 threshold2 = (sortNext(t3) + sortMax(t4)) / 2;
395
396 end

```