ML Homework4

Due Date: 2019/05/02 (THU.) 23:59

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Description:

1. Logistic regression

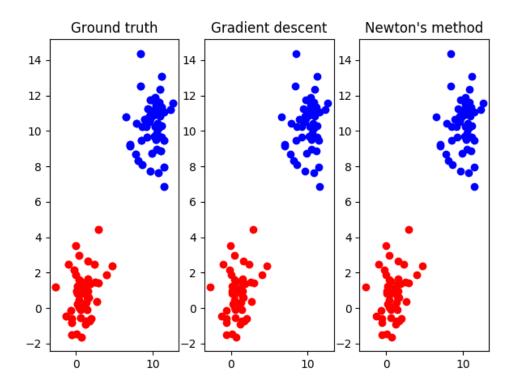
- Input:
 - 1. N (number of data points)
 - 2. $mx_1, vx_1, my_1, vy_1, mx_2, vx_2, my_2, vy_2$ (m: mean, v: variance)
- Function:
 - 1. Generate n data point: $D1=(x_1,y_1),(x_2,y_2),\ldots,(x_n,y_n)$, where x and y are independently sampled from $N(mx_1,vx_1)$ and $N(my_1,vy_1)$ respectively.
 - 2. Generate n data point: $D2=(x_1,y_1),(x_2,y_2),\ldots,(x_n,y_n)$, where x and y are independently sampled from $N(mx_2,vx_2)$ and $N(my_2,vy_2)$ respectively.
 - 3. Use Logistic regression to separate D1 and D2. You should implement both Newton's and steepest gradient descent method during optimization.
 - In other words, when the Hessian is singular, use steepest descent for instead. You should come up with a reasonable rule to determine convergence.(a simple run out of the loop should be used as the ultimatum)
- Output:
 - 1. The confusion matrix and the **sensitivity** and **specificity** of the logistic regression applied to the training data *D*.
 - 2. Visualization
 - Plot the ground truth
 - Plot the predict result
 - Gradient descent
 - Newton's method

Use the Gaussian random number generator in homework 3.

• Sample input & output (for reference only)

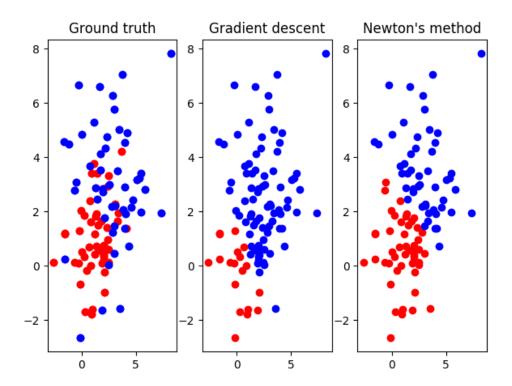
```
\circ \;\; Case 1: N=50, mx_1=my_1=1, mx_2=my_2=10, vx_1=vy_1=vx_2=vy_2=2
```

```
Gradient descent:
1
2
3
   w:
4
   -78.1766393662
5
     6.7233419236
    11.2430677919
6
7
  Confusion Matrix:
8
9
         Predict cluster 1 Predict cluster 2
   Is cluster 1 50
10
                   0
                                   50
   Is cluster 2
11
12
13
   Sensitivity (Successfully predict cluster 1): 1.00000
   Specificity (Successfully predict cluster 2): 1.00000
14
15
16
   _____
17
   Newton's method:
18
19
   w:
   -118.3601516394
20
21
    8.7747332848
22
    10.1954120077
23
24
  Confusion Matrix:
             Predict cluster 1 Predict cluster 2
25
  Is cluster 1 50
26
27
   Is cluster 2
                   0
                                    50
28
  Sensitivity (Successfully predict cluster 1): 1.00000
29
30 Specificity (Successfully predict cluster 2): 1.00000
```



 \circ Case 2: $N=50, mx_1=my_1=1, mx_2=my_2=3, vx_1=vy_1=2, vx_2=vy_2=4$

```
Gradient descent:
 1
 2
 3
   w:
    -71.1902536008
 4
5
     46.0123814025
     54.6803199701
 6
 7
8
    Confusion Matrix:
9
               Predict cluster 1 Predict cluster 2
10
    Is cluster 1
                       16
                                         34
    Is cluster 2
                                         47
11
12
    Sensitivity (Successfully predict cluster 1): 0.32000
13
14
    Specificity (Successfully predict cluster 2): 0.94000
15
16
17
    Newton's method:
18
19
    w:
20
     -1.9045831451
      0.3940876974
21
       0.5695243849
22
23
   Confusion Matrix:
24
25
                Predict cluster 1 Predict cluster 2
26
    Is cluster 1
                                         10
27
    Is cluster 2
                        10
                                          40
```



2. EM algorithm

- Input: MNIST training data and label sets. (Same as HW02)
- Function:
 - 1. Binning the gray level value into **two bins**. Treating all pixels as random variables following Bernoulli distributions. Note that each pixel follows a different Binomial distribution independent to others.
 - 2. Use EM algorithm to cluster each image into ten groups. You should come up with a reasonable rule to determine convergence. (a simple run out of the loop should be used as the ultimatum)
- Output:
 - 1. For each digit, output a confusion matrix and the **sensitivity** and **specificity** of the clustering applied to the training data.
 - 2. Print out the imagination of numbers in your classifier
 - Just like before, about the details please refer to HW02
- Hint: The algorithm is a kind of unsupervised learning, so the labels are not used during training. But you can use these labels to help you to figure out which class belongs to which number.

In other words, you should find a way to assign label to each class which you classified **before you compute the confusion matrix**

• Sample input & output (for reference only)

```
50
51
53
54
55
56
57
58
59
60
61
... all other unlabeled imagination of numbers goes here ...
62
63
class 9:
64
65
66
67
68
70
71
72
73
74
75
76
77
78
79
80
81
82
83
84
85
86
87
88
89
90
91
92
93
No. of Iteration: 1, Difference: 3176.579389514846
94
95
96
97
class 0:
98
```

```
99
100
101
102
103
104
105
106
107
108
109
110
111
112
113
114
115
116
117
118
119
120
121
122
123
124
125
126
127
... all other iterations goes here ...
128
129
class 9:
130
131
132
133
134
135
136
137
138
139
140
141
142
143
144
145
146
147
```

```
148
149
150
151
152
153
154
155
156
157
158
No. of Iteration: 10, Difference: 19.89546432548733
159
160
161
162
163
164
labeled class 0:
165
166
167
168
169
170
171
172
173
174
175
176
177
178
179
180
181
182
183
184
0 0 0 0 0 0 0 1 1 1 1 1 0 0 0 0 0 1 1 1 1 1 1 0 0 0 0 0 0
185
186
187
188
189
190
191
192
193
194
labeled class 1:
195
196
```

```
197
198
199
200
2.01
202
203
204
2.05
206
207
208
2.09
210
211
2.12
213
214
215
2.16
217
218
219
2.2.0
221
222
223
224
... all other labeled imagination of numbers goes here ...
225
226
labeled class 9:
2.2.7
228
229
230
231
232
233
234
235
236
237
238
239
240
2.41
242
243
244
2.45
```

```
246 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 1 1 0 0 0 0 0 0 0 0
  247
248
  249
  250
  251
  252
253
  254
255
256
257
258
  Confusion Matrix 0:
259
            Predict number 0 Predict not number 0
260
  Is number 0
              3023
                            2900
261
   Isn't number 0
             113
                            53964
262
   Sensitivity (Successfully predict number 0) : 0.51038
263
   Specificity (Successfully predict not number 0): 0.99791
264
265
266
267
268
  Confusion Matrix 1:
        Predict number 1 Predict not number 1
269
            5986
270
  Is number 1
                800
271
   Isn't number 1
                            52458
272
   Sensitivity (Successfully predict number 1) : 0.88787
273
   Specificity (Successfully predict not number 1): 0.98498
274
275
276
277
278
   ... all other confusion matrix goes here ...
279
280
   ______
281
282
  Confusion Matrix 9:
            Predict number 9 Predict not number 9
283
284
   Is number 9
               2718
                            3231
285
   Isn't number 9
               5147
                            48904
286
287
   Sensitivity (Successfully predict number 9) : 0.45688
   Specificity (Successfully predict not number 9): 0.90478
288
289
290 Total iteration to converge: 10
291 Total error rate: 0.5081666666666667
```