E11 Expected Maximize Algorithm

Suixin Ou

School of Computer Science Sun Yat-sen University

December 21, 2021





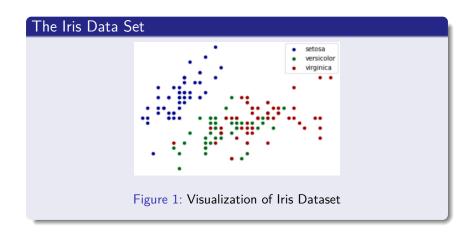
Background

The Iris Data Set

- The UCI dataset (http://archive.ics.uci.edu/ml/index.php) is the most widely used dataset for machine learning. If you are interested in other datasets in other areas, you can refer to https:// www.zhihu.com/question/63383992/answer/222718972.
- It is perhaps the best known database to be found in the pattern recognition literature. The data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.



Background





Task

Description

Dataset statistics

Data Set Characteristics:	Multivariate	Number of Instances:	150	Area:	Life
Attribute Characteristics:	Real	Number of Attributes:	4	Date Donated	1988-07-01
Associated Tasks:	Classification	Missing Values?	No	Number of Web Hits:	4403737

- Domain information Attribute Information:
 - 1. sepal length in cm
 - 2. sepal width in cm
 - 3. petal length in cm
 - 4. petal width in cm
 - 5. class:
 - -- Iris Setosa
 - -- Iris Versicolour
 - -- Iris Virginica





Read the file "iris.data"

```
127
     def loadData(filename):
         """从文件中读取数据
128
129
130
         :param filename : the path of file
         :return : the dataset
131
132
         :return type : list
133
         .....
134
135
         dataSet = []
136
         with open(filename) as fr:
137
             for i, line in enumerate(fr.readlines()):
138
                 curLine = line.strip().split(",")
139
                 fltLine = list(map(float, curLine[:-1]))
140
                 dataSet.append(fltLine)
141
         return dataSet
```

Initialize parameters

```
95
     def init params(shape, K):
         """initialize the parameters : mu, gamma, pi
 96
 97
 98
         :param shape: the row and column of data
 99
         :param K: the number of model
         :return : the initial parameters
100
101
         11 11 11
102
103
         N, D = shape
         mu = np.random.rand(K, D)
194
         Sigma = np.array([np.eye(D)] * K)
105
         pi = np.array([1.0 / K] * K)
106
107
         return mu, Sigma, pi
```





Expected Maximize algorithm framework

```
def GMM EM(Y, K, times):
110
111
         """GMM EM
112
113
         :param Y :dataset
114
         :param K :the number of model (3)
115
         :param times : the iteration times
116
         :return : the parameters of three models - mu, gamma , pi
117
         H H H
118
119
         Y = scale data(Y)
120
         mu, Sigma, pi = init params(Y.shape, K)
121
         for i in range(times):
122
             gamma = getExpectation(Y, mu, Sigma, pi)
123
             mu, Sigma, pi = maximize(Y, gamma)
124
         return mu, Sigma, pi, gamma
```





Please Finish the getExpectation function.

```
19
   def getExpectation(Y, mu, Sigma, pi):
       """E step
20
22
       :param Y : data matrix
23
       :param mu: the mean of each characterristic of each sample ; mu is a 3*4 matrix
24
       :param Sigma :three-covariance-matrix list
25
       :param pi: the responsibilities array
       :return : the new responsibilities matrix(gamma)
26
       :return type : matrix
28
29
       # 样本数
30
       N = Y.shape[0]
31
       # 模型数
32
       K = pi.shape[0]
33
34
       # 响应度矩阵. 行对应样本. 列对应响应度
35
       gamma = np.mat(np.zeros((N, K)))
36
37
       # 计算各模型中所有样本出现的概率。 行对应样本, 列对应模型
       prob = np.zeros((N, K))
38
       for k in range(K):
           prob[:, k] = phi(Y, mu[k], Sigma[k])
40
41
       prob = np.mat(prob)
42
43
       # 计算每个模型对每个样本的响应度
44
       # TODO
       return gamma
45
```





Please Finish the maximize function.

```
def maximize(Y, gamma):
48
        """M step
49
50
51
        :param Y: data matrix
52
        :param gamma : the responsibilities matrix
53
        :return : the parameters : mu, gamma, pi
54
55
        11 11 11
56
        # 样本数和特征数
57
        N, D = Y.shape
58
        # 模型数
59
        K = gamma.shape[1]
60
61
        # 初始化参数值
        mu = np.zeros((K, D))
62
63
        Sigma = []
64
        pi = np.zeros(K)
65
```



December 21, 2021

Submission

Submission

pack your report E11_YourNumber.pdf and source code into zip file E11_YourNumber.zip, then send it to ai_course2021@163.com.



The End





December 21, 2021