12/12/2020 nbm.py

```
1 import numpy as np
 2 import pandas as pd
 3 import matplotlib.pyplot as plt
 4 import collections
 5
 6 class NB_model():
7
       def __init__(self):
           self.pi = {} # to store prior probability of each class
 8
 9
           self.Pr_dict = None
10
           self.num_vocab = None
11
           self.num_classes = None
12
       def fit(self, train_data, train_label, vocab, if_use_smooth=True):
13
           # get prior probabilities
14
           self.num vocab = len(vocab['index'].tolist())
15
           self.get_prior_prob(train_label)
16
           # ========= YOUR CODE HERE =================
17
           # Calculate probability of each word based on class
18
19
           # Hint: Store each probability value in matrix or dict:
   self.Pr dict[classID][wordID] or Pr dict[wordID][classID])
20
           # Remember that there are possible NaN or 0 in Pr_dict matrix/dict.
   Use smooth method
           self.classes = collections.defaultdict(int)
21
22
           word count per class = collections.defaultdict(lambda:
   collections.defaultdict(int))
           self.Pr_dict = collections.defaultdict(lambda:
23
   collections.defaultdict(float))
24
           train dict = train data.to dict()
25
           for i in range(len(train_dict['classIdx'])):
26
               self.classes[train_dict['classIdx'][i]] += train_dict['count'][i]
27
               word_count_per_class[train_dict['classIdx'][i]]
28
   [train_dict['wordIdx'][i]] += train_dict['count'][i]
29
30
           for classID in word count per class:
31
               for wordID in word count per class[classID]:
                   self.Pr_dict[classID][wordID] = (word_count_per_class[classID]
32
   [wordID] + 1) /
                                                       (self.classes[classID] +
33
   self.num vocab)
34
           # ====
35
           print("Training completed!")
36
37
       def predict(self, test_data):
38
           test dict = test data.to dict() # change dataframe to dict
39
           new dict = \{\}
40
           prediction = []
41
42
           for idx in range(len(test_dict['docIdx'])):
43
               docIdx = test_dict['docIdx'][idx]
               wordIdx = test_dict['wordIdx'][idx]
44
45
               count = test dict['count'][idx]
46
               try:
                   new_dict[docIdx][wordIdx] = count
47
48
               except:
49
                   new dict[test dict['docIdx'][idx]] = {}
50
                   new dict[docIdx][wordIdx] = count
51
           for docIdx in range(1, len(new_dict)+1):
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localhost:4649/?mode=python 1/2

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12/12/2020
                                            nbm.py
               score_dict = {}
53
54
               max score = 0
55
               #Creating a probability row for each class
               for classIdx in range(1,self.num_classes+1):
56
                   score dict[classIdx] = 0
57
                   # ========= YOUR CODE HERE =================
58
                   ### Implement the score_dict for all classes for each document
59
60
                   ### Remember to use log addtion rather than probability
   multiplication
                   ### Remember to add prior probability, i.e. self.pi
61
                   score_dict[classIdx] += np.log(self.pi[classIdx])
62
63
                   for wordId in new_dict[docIdx]:
64
                        if self.Pr_dict[classIdx][wordIdx] == 0:
                            score_dict[classIdx] += new_dict[docIdx][wordId] *
65
   np.log(1/(self.classes[classIdx] + self.num vocab))
66
                        else:
                            score_dict[classIdx] += new_dict[docIdx][wordId] *
67
   np.log(self.Pr dict[classIdx][wordId])
68
                   # ==============
               max_score = max(score_dict, key=score_dict.get)
69
70
               prediction.append(max_score)
71
           return prediction
72
73
74
       def get_prior_prob(self,train_label, verbose=True):
75
           unique class = list(set(train label))
76
           self.num_classes = len(unique_class)
           total = len(train_label)
77
78
           for c in unique_class:
79
               # ========= YOUR CODE HERE ================
               ### calculate prior probability of each class ####
80
81
               ### Hint: store prior probability of each class in self.pi
82
               counter = 0
               for label in train_label:
83
                   if c is label:
84
85
                       counter += 1
               self.pi[c] = counter / total
86
               # =======
87
88
           if verbose:
               print("Prior Probability of each class:")
89
90
               print("\n".join("{}: {}".format(k, v) for k, v in
   self.pi.items()))
91
```

localhost:4649/?mode=python 2/2