資訊檢索與文字探勘導論 作業三

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- 1. 執行環境 Anaconda Spyder
- 2. 程式語言
 Python 3.6.8
- 3. 執行方式
- (1) 需要 import nltk(preprocessor.py 檔需要)、numpy、pandas 及 math 套件

使用 pip install nltk、pip install numpy、pip install pandas

- (2) preprocessor.py 放入 hw1 裡面文字前處理的 function, stopwords.txt 裡面放入要被讀進來的 stopwords, training_index 裡面放入每一個 class 的訓練集文章的編號, IRTM 資料夾裡放入所有文章的 txt 檔
 - (3) 執行 204 到 211 行即可得到結果

```
training_index_txt, training_index_all = get_training_index()

df_list, tf_list = calculate_df_tf(training_index_txt)

df_all, tf_all = calculuate_df_tf_all(df_list, tf_list)

total_training_doc_num = calculate_num_doc(training_index_txt)

features = chi_square_feature_selection(df_list, df_all, total_training_doc_num)

conditional_prob, prior = train_NB_clf(tf_list, tf_all, features)

result_class = test_NB_clf(training_index_all, prior, features, conditional_prob)

save_result(result_class)
```

- 4. 程式處理邏輯
 - (0) import 需要套件

```
#import necessary modules from hw1
from Preprocessor import text_preprocessing, get_text

#import necessary module
import numpy as np
import pandas as pd
import math
```

(1) 建立兩個 list

training_index_txt (list of list): 外層的 list 代表每一個 class,內層的 list 放入所有 training document 的編號並且在結尾加上.txt

training_index_all (list): 將所有的 training document 的編號放入一個 list 內

```
def get_training_index():
    '''

training_index_txt : list of list, each element consists of document number in each class with .txt at the end
training_index_all : list, each element consists of document number in each class

'''

training_index_txt = []
training_index_all = []
f = open('training_index.txt')
lines = f.readlines()
for line in lines:
    line = line.split(' ')
    line = line[1:16]
temp = []
for docNum in line:
    training_index_all.append(int(docNum))
    temp.append(docNum + '.txt')
training_index_txt.append(temp)
return training_index_txt, training_index_all
```

(2) 計算每個 class 內每個 term 的 term frequency 和 document frequency 並且存入 df_list 及 tf_list 內。兩個 list 中的每一個元素都是一個 dictionary(總共13 個), 放入每個 term 的 df 及 tf

```
def calculate_df_tf(training_index_txt):
       df\_list:13 dictionary in a list, each ditionary contains word and its document frequency tf\_list:13 dictionary in a list, each ditionary contains word and its term frequency
      df_list = []
tf_list = []
for index_by_class in training_index_txt:
    temp_list_df = [] #contains dictionary with keys = words, values = 1
    temp_dict_tf = dict()
    temp_dict_df = dict()
             #iterate each document in class
for index in index_by_class:
    raw_text = get_text(index)
                    text = text_preprocessing(raw_text)
                   #count df by combining dictionaries
temp_list_df.append(dict((word, 1) for word in set(text)))
                    #iterate each word in document
                     for word in text:
                            if word not in temp_dict_tf.keys():
                                 temp_dict_tf[word] = 1
                                  temp_dict_tf[word] += 1
             for my_dict in temp_list_df:
                     img_are in my_dict.items():
    temp_dict_df.setdefault(key, 0)
    temp_dict_df[key] += value
             tf_list.append(temp_dict_tf)
df_list.append(temp_dict_df)
       #do not record word:''
for my_dict in tf_list:
    my_dict.pop('')
       for my_dict in df_list:
    my_dict.pop('')
        return df_list, tf_list
```

(3) 對所有 term 計算所有 class 合併的 df 及 tf 留著備用

```
#calculate df and tf for each term for all training documents

def calculuate_df_tf_all(df_list, tf_list):
    df_all = {}

for my_dict in df_list:
    for key, value in my_dict.items():
        df_all.setdefault(key, 0)
        df_all[key] += value

tf_all = {}

for my_dict in tf_list:
    for key, value in my_dict.items():
        tf_all.setdefault(key, 0)
        tf_all.setdefault(key, 0)

tf_all[key] += value

return df_all, tf_all
```

(4) 計算每一個 class document 的數量及所有 class 加起來 document 的數量留著備用

```
#calculate number of training document

def calculate_num_doc(training_index_txt):

doc_num = []

for index in training_index_txt:

doc_num.append(len(index))

total_training_doc_num = sum(doc_num)

return total_training_doc_num
```

-----feature selection------

(5) chi-square

對所有 class,去計算裡面每一個字的 chi-square,並且挑出每一個 class 內 chi-square 值最高的 500 個存到 set 內,再將這 13 個 set 放入名為 features 的 list 中。

```
v def chi_square_feature_selection(df_list, df_all, total_training_doc_num):
        features = []
        for my_dict in df_list:
             word_list = []
             score_list = []
             for word in my_dict.keys():
                   word_list.append(word)
                   presentNon : number of document that this word appears in class c
absentNon : number of document that this word doesn't appear in class c
                   presentNoff : number of document that this word appears in class other than c
                   absentNoff : number of document that this word doesn't appear in class other than c
                  on_topic = 15
                  off_topic = total_training_doc_num - on_topic
present = df_all[word]
                   absent = total_training_doc_num - present
                  presentNon = my_dict[word]
                  absentNon = on_topic - presentNon
presentNoff = present - presentNon
                   absentNoff = absent - absentNon
                  E_00 = absent * off_topic / total_training_doc_num
E_01 = absent * on_topic / total_training_doc_num
E_10 = present * off_topic / total_training_doc_num
E_11 = present * on_topic / total_training_doc_num
                  chi_00 = ((absentNoff - E_00) ** 2) / E_00
chi_01 = ((absentNon - E_01) ** 2) / E_01
chi_10 = ((presentNoff - E_10) ** 2) / E_10
chi_11 = ((presentNon - E_11) ** 2) / E_11
                   score = chi_00 + chi_01 + chi_10 + chi_11
                  score_list.append(score)
              score_arr = np.asarray(score_list)
              index = (-score_arr).argsort()[:k]
              feature_set = set([word_list[i] for i in index])
              features.append(feature_set)
        return features
```

-----NB model-----

(6) training Multinomial Naïve Bass Classifier

對所有的字,計算在每一個 class 內的條件機率,若有出現在挑出來的 500 個 feature 內才存入最後的字典中。最後,對每一個 class 都蒐集這 500 字的條件機率,並且將 13 個字典存入一個名為 conditional prob 的 list。

```
def train_NB_clf(tf_list, tf_all, features):
     conditional_prob = []
     count = 0
     prior = []
     for class_dict in tf_list:
         conditional_prob_dict = dict()
prior.append(15 / 195)
         word_num = len(tf_all.keys()) ##500 * 13 or 全部不重複的字
         sum_tf = sum(class_dict.values())
         for word in tf_all.keys():
             if word in features[count]:
                  #add-one smoothing
                  cond_prob = (class_dict[word] + 1) / (sum_tf + word_num)
                  conditional_prob_dict[word] = cond_prob
         conditional_prob.append(conditional_prob_dict)
         count += 1
     return conditional_prob, prior
```

(7) predicting class by Multinomial Naïve Bass Classifier

對每個 document 裡面的 term,去對每一個 class 裡面的字典搜尋條件機率,若有找到則取自然對數後加到最後的 score 中。最後每一個 term 都會有13 個 score,將 class 最高的 score 取出並且存入 result_class 這個 dictionary 內。(key: document index, value: predicting class)

```
def test_NB_clf(training_index_all, prior, features, conditional_prob):
     result_class = dict()
     for index in range(1,1096,1):
         if index not in training_index_all:
             raw_text = get_text(str(index) + '.txt')
             text = text_preprocessing(raw_text)
             score_list = []
             for i in range(13):
                 score = 0
                  score += math.log(prior[i])
                 for word in text:
                     if word in features[i]:
                          score += math.log(conditional_prob[i][word])
                 score_list.append(score)
             score_arr = np.asarray(score_list)
             best_class = (-score_arr).argsort()[:1] + 1
             result_class[index] = best_class[0]
     return result_class
```

(8) save result

將結果轉成 dataframe 後存入'hw3_result'.csv 並上傳 kaggle