Introduction

這個作業主要是利用關鍵字和MultinomialNB訓練一分析文章的model,並用此model分析其它文章的內容並預測其分類

Objective

主要目標

- 1. 取199筆文章並做關鍵字分析和training
- 2. 取50筆文章測試training model並給出F-score, Precision score 和 Recall score

Actions

我們選用 MultinomialNB訓練我們從文章所得之關鍵字,關鍵字之截取因程式碼過長所以不列出

```
vectorizer=CountVectorizer()
transformer=TfidfTransformer()
tfidf=transformer.fit_transform(vectorizer.fit_transform(finalkey)
)
y = np.array([1,2,3,4,5,6,7])#1:htc 2:samsung 3:apple 4:xiaomi
5:sony 6:lg 7:asus
clf = MultinomialNB().fit(tfidf, y)
```

預測test data之內容並印出score

```
y_pred=np.array([])
y_true=np.array([])
brand_dict={"htc":1,"sam":2,"apple":3,"xm":4,"sony":5,"lg":6,"asus
":7}
with open('test_data.json',encoding = 'utf8') as content_file:
    for line in content_file:
    item=json.loads(line)
    testcontent=""

testcontent=testcontent+''.join(item['title'])+''.join(item['content'])
    testkeyword=convert_doc_to_wordlist(testcontent,False)
    testkey=[]
    testkey.append(' '.join(testkeyword))
```

```
testset=transformer.fit_transform(vectorizer.transform(testkey))
    y_pred=np.append(y_pred,clf.predict(testset))
    y_true=np.append(y_true,np.array(brand_dict[item['brand']]))

print("y_prediction:",y_pred,"y_true:",y_true)
print("F1 score:",f1_score(y_true, y_pred, average="macro"))
print("Precision score:",precision_score(y_true, y_pred, average="macro"))
print("Recall score:",recall_score(y_true, y_pred, average="macro"))
```

最後結果

```
y_prediction: [ 2. 7. 1. 2. 5. 6. 3. 2. 1. 1. 2.
                                                            5.
                                                               3.
5. 2. 7. 3. 4.
     1. 1.
             5. 6. 7. 3. 5. 2. 7. 7. 3. 2. 7. 3. 4.
 3.
7. 1.
  1.
     7.
          5.
              1.
                  1.
                      4.
                          3.
                              4.
                                  2.
                                      5.
                                          1.
                                              1.
                                                  5.
                                                      5.] y_true:
[ 3.
     7.
              2.
                  5.
                      6.
                          6.
                              2.
                                  7.
                                      1.
                                          2.
                                              5.
                                                  3.
                                                          2.
          1.
                                                      1.
                                                             7.
3. 4.
     1.
          7.
              5.
                 6.
                      2.
                          3.
                              5.
                                  2.
                                      1.
                                          2.
                                              3.
                                                  2.
                                                      2.
                                                          3.
  3.
                                                             4.
7. 1.
                  2.
                              2.
                                 2.
     4.
          5.
              1.
                      4.
                          3.
                                      2.
                                         1.
                                             1. 2.
                                                      5.]
F1 score: 0.726400979972
Precision score: 0.752705627706
Recall score: 0.741666666667
```