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Факультет Информатика и системы управления

Кафедра ИУ-5

«Методы машинного обучения»

Рубежный контроль №2

По дисциплине

Выполнили студенты группы ИУ-5 24М

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Тема: Методы обработки текстов

Решение задачи классификации текстов

Необходимо сформировать два варианта векторизации признаков - на основе CountVectorizer и на основе TfidfVectorizer.

Группа	Классификатор №1	Классификатор №2
ИУ5И-24М	KNeighborsClassifier	Complement Naive Bayes - CNB

```
import numpy as np
import pandas as pd
from typing import Dict, Tuple
from \ \texttt{scipy} \ import \ \texttt{stats}
import torch
from IPython. display import Image
from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
from sklearn.model_selection import train_test_split
from \ {\tt sklearn.neighbors} \ import \ {\tt KNeighborsClassifier}
from sklearn.naive_bayes import ComplementNB
from \ {\tt sklearn.\,metrics} \ import \ {\tt accuracy\_score}, \ balanced\_accuracy\_score
from sklearn.metrics import precision_score, recall_score, fl_score, classification_report
from sklearn.metrics import confusion_matrix
{\bf from} \ {\tt sklearn.} \ {\tt model\_selection} \ {\bf import} \ {\tt cross\_val\_score}
\mathbf{from} \text{ sklearn.pipeline } \mathbf{import} \text{ Pipeline}
\textbf{from} \ \text{sklearn.metrics} \ \textbf{import} \ \text{mean\_absolute\_error}, \ \text{mean\_squared\_error}, \ \text{mean\_squared\_log\_error}, \ \text{median\_absolute\_error}, \ r2\_score
from sklearn.metrics import roc_curve, roc_auc_score
from sklearn.svm import SVC, NuSVC, LinearSVC, OneClassSVM, SVR, NuSVR, LinearSVR
import \ {\tt seaborn} \ as \ {\tt sns}
import \ \mathtt{matplotlib}. \ \mathtt{pyplot} \ \mathbf{as} \ \mathtt{plt}
%matplotlib inline
sns. set(style="ticks")
```

```
def accuracy_score_for_classes(
  y_true: np. ndarray,
  y_pred: np.ndarray) -> Dict[int, float]:
  Вычисление метрики accuracy для каждого класса
  y_true - истинные значения классов
  y_pred - предсказанные значения классов
  Возвращает словарь: ключ - метка класса,
  значение - Accuracy для данного класса
  # Для удобства фильтрации сформируем Pandas DataFrame
  d = {'t': y_true, 'p': y_pred}
  df = pd. DataFrame (data=d)
  # Метки классов
  classes = np.unique(y_true)
  #Результирующий словарь
  res = dict()
  #Перебор меток классов
  for c in classes:
     # отфильтруем данные, которые соответствуют
     # текущей метке класса в истинных значениях
     temp_data_flt = df[df['t']==c]
     # расчет accuracy для заданной метки класса
     temp_acc = accuracy_score(
       temp_data_flt['t'].values,
       temp_data_flt['p'].values)
     # сохранение результата в словарь
     res[c] = temp_acc
  return res
```

```
def print_accuracy_score_for_classes(
   y_true: np.ndarray,
   y_pred: np.ndarray):
   Bывод метрики accuracy для каждого класса
   """
   accs = accuracy_score_for_classes(y_true, y_pred)
   if len(accs)>0:
        print(' Meтка \t Accuracy')
   for i in accs:
        print(' {} \t {}'.format(i, accs[i]))
```

```
# Загрузка данных
df = pd. read_csv("Test. csv")
df. head()
```

text label 1 always wrote this series off as being a comp... 1 1st watched 12/7/2002 - 3 out of 10(Dir-Steve ... 2 This movie was so poorly written and directed ... 3 The most interesting thing about Miryang (Secr... 4 when i first read about "berlin am meer" i did... 0

```
# Сформируем общий словарь для обучения моделей на обучающей и тестовой выборки vocab_list = df['text']. tolist()
vocab_list[1:10]
```

["Ist watched 12/1/2002 - 3 out of 10(Dir-Steve Purcell): Typical Mary Kate & Ashley fare with a few more kisses. It looks to me like the girls are getting pretty tire dof this stuff and it will be interesting what happens to them if they ever decide to split up and go there own ways. In this episode of their adventures they are interns in Rome for a Tashion' designer who puts them right into the mailroom to learn what working hard is all about(I guess.). Besides the typical flirtations with boys there is nothing much else except the Rome scenario until about % way into the movie when it's finally revealed why they are getting fired, then re-hired, then fired again. This is definetly made by people who don't understand the corporate world and it shows in their interpretation of it. Maybe the real world will be their next adventure(if there is one.). Even my kids didn't seem to care for this boring 'adventure' in the make-believe. Let's see they probably only have a couple of years till their legal adults. We'll see what happens then.".

'This movie was so poorly written and directed I fell asleep 30 minutes through the movie. The jokes in the movie are corny and even though the plot is interesting a t some angles, it is too far fetched and at some points- ridiculous. If you are ll or older you will overlook the writing in the movie and be disappointed, but if you are ll or younger this is a film that will capture your attention and be amazed with all the stunts (which I might add are poorly done) and wish you were some warrior to. The casting in this movie wasn't very good, and the music was very disappointing because it was like they were trying to build up the tension but it didn't fit at all. On a scale of 1-10 (10 being excellent, 1 being horrible) the acting in this movie is a 4. Brenda Song is talented in comedy, but with this kind of movie, in some of the more serious scenes, her acting was laughable. When she made some of her "fighting" poses, I started laughing out loud. I think the worst thing about t

(5000, 2)

```
# C \phi ормируем общий vocab_list = df['text'].tolist()
                     бщий словарь для обучения моделей из обучающей и тестовой выборки
vocab_list[1:10]
```

["1st watched 12/7/2002 - 3 out of 10(Dir-Steve Purcell): Typical Mary Kate & Ashley fare with a few more kisses. It looks to me like the girls are getting pretty tire dof this stuff and it will be interesting what happens to them if they ever decide to split up and go there own ways. In this episode of their adventures they are interns in Rome for a fashion' designer who puts them right into the mailroom to learn what working hard is all about(I guess.). Besides the typical flirtations with boys there is nothing much else except the Rome scenario until about % way into the movie when it's finally revealed why they are getting fired, then re-hired again, then re-hired again. This is definetly made by people who don't understand the corporate world and it shows in their interpretation of it. Maybe the real world will be their next adventure (if there is one.). Even my kids didn't seem to care for this boring adventure' in the make-believe. Let's see they probably only h

```
vocabVect = CountVectorizer()
vocabVect.fit(vocab_list)
corpusVocab = vocabVect.vocabulary_
print('Количество сформированных признаков - {}'.format(len(corpusVocab)))
```

Количество сформированных признаков - 39126

```
for i in list(corpusVocab)[1:10]:
   print('{}={}'.format(i, corpusVocab[i]))
```

wrote=38700 this=34932 series=30826 off=24255 as=2274 being=3467 complete = 7225stink=33134 fest=12941

```
test_features = vocabVect.transform(vocab_list)
 <5000x39126 sparse matrix of type '<class 'numpy.int64' '</p>
with 685304 stored elements in Compressed Sparse Row format
 test features. todense()
matrix([[0, 0, 0, ..., 0, 0, 0], [0, 0, 0, ..., 0, 0, 0], [0, 0, 0, ..., 0, 0, 0],
```

len(test_features.todense()[0].getAl())

```
vocabVect.get_feature_names()[100:120]
/usr/local/lib/python3.7/dist=packages/sklearn/utils/deprecation.py:87: FutureWarning: Function get_feature_names is deprecated; get_feature_names is deprecated in 1. 0 and will be removed in 1. 2. Please use get_feature_names_out instead.
warnings. warn(msg, category=PutureWarning)
[165]
[165]
[168]
  '16a',
'16b',
'16ieme'
'16mm',
'16s',
'16th',
  '16ème',
'17',
'170',
'1701',
'1710',
  '17th',
'18',
'180',
'1800']
  \mathbf{def} \ \ \mathsf{VectorizeAndClassify} (vectorizers\_list, \ \ \mathsf{classifiers\_list}):
      vectorizeAndClassify(vectorizers_list, classifiers_list):
for v in vectorizers_list:
    for c in classifiers_list:
    pipelinel = Pipeline(["vectorizer", v), ("classifier", c)])
    score = cross_val_score(pipelinel, df['text'], df['label'], scoring='accuracy', cv=3).mean()
    print(" B e κ τ ο ρ н а а н н н - {}'.format(v))
    print(" M ο д е л ь д л я кла с с н ф н к а н н н - {}'.format(c))
    print("Accuracy = {}'.format(sore))
    print("accuracy = {}'.format(sore))
    print("========"")
 \label{eq:continuous} vectorizers\_list = [CountVectorizer(vocabulary = corpusVocab), \ TfidfVectorizer(vocabulary = corpusVocab)] \\ classifiers\_list = [ComplementNB(), KNeighborsClassifier()] \\ VectorizeAndClassify(vectorizers\_list, classifiers\_list) \\ \\
 Векторизация - CountVectorizer(vocabulary={'00': 0, '000': 1, '00015': 2, '001': 3,
                                                             '003830': 4, '0069': 5, '007': 6, '0079': 7, '0083': 8, '00am': 9, '00s3': 10, '01': 11, '02': 12, '0230': 13, '04': 14, '041': 15, '05': 16, '06': 17, '07': 18, '07b': 19, '09': 20, '0s': 21, '10': 22,
                                                              '100': 23, '1000': 24, '1001': 25, '100b': 26,
                                                              '100k': 27, '101': 28, '102': 29, ...})
Модель для классификации - ComplementNB()
Accuracy = 0.813400105093027
Векторизация - CountVectorizer (vocabulary={'00': 0, '000': 1, '00015': 2, '001': 3,
                                                              '003830': 4, '0069': 5, '007': 6, '0079': 7, '0083': 8, '00am': 9, '00s': 10, '01': 11, '02': 12, '0230': 13, '04': 14, '041': 15, '05': 16, '06': 17,
                                                              '07': 18, '07b': 19, '09': 20, '0s': 21, '10': 22,
                                                              '100': 23, '1000': 24, '1001': 25, '100b': 26, '100k': 27, '101': 28, '102': 29, ...})
Модель для классификации - KNeighborsClassifier()
Accuracy = 0.6017986078654617
```

```
Векторизация - TfidfVectorizer(vocabulary={'00': 0, '000': 1, '00015': 2, '001': 3, '003830': 4, '0069': 5, '007': 6, '0079': 7, '0083': 8, '00am': 9, '00s': 10, '01': 11, '02': 12, '0230': 13, '04': 14, '041': 15, '05': 16, '06': 17, '07': 18, '07b': 19, '09': 20, '0s': 21, '10': 22, '100': 23, '1000': 24, '1001': 25, '100b': 26, '100k': 27, '101': 28, '102': 29, ...})

Модель для классификации - ComplementNB()

Ассигасу = 0. 8214019861093808

Векторизация - TfidfVectorizer(vocabulary={'00': 0, '000': 1, '00015': 2, '001': 3, '003830': 4, '0069': 5, '007': 6, '0079': 7, '0083': 8, '00am': 9, '00s': 10, '01': 11, '02': 12, '0230': 13, '04': 14, '041': 15, '05': 16, '06': 17, '07': 18, '07b': 19, '09': 20, '0s': 21, '10': 22, '100': 23, '1000': 24, '1001': 25, '100b': 26, '100k': 27, '101': 28, '102': 29, ...})

Модель для классификации - KNeighborsClassifier()

Ассигасу = 0. 6558028130268304
```

Вывод: вариант векторизации признаков TfidfVectorizer в паре с классификатором Complement Naive Bayes показал наилучшее качество.