Рубежный контроль №2

Группа: ИУ5И-22М

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Задание:Решение задачи классификации текстов.

Необходимо решить задачу классификации текстов, сформировав два варианта векторизации признаков - на основе CountVectorizer и на основе TfidfVectorizer. В качестве классификаторов необходимо использовать два классификатора:

Random Forest Classifier

Complement Naive Bayes

In [1]:

```
import numpy as np
import pandas as pd
from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.metrics import classification_report
from sklearn.ensemble import RandomForestClassifier
from sklearn.naive_bayes import ComplementNB
```

In [2]:

```
df = pd.read_csv('news_articles.csv', usecols=['text', 'hasImage'], nrows=1000)
df
```

Out[2]:

	text	haslmage
0	print they should pay all the back all the mon	1
1	why did attorney general loretta lynch plead t	1
2	red state \nfox news sunday reported this mor	1
3	email kayla mueller was a prisoner and torture	1
4	email healthcare reform to make america great	1
995	freitag november gaagnagna zum babywort des	1
996	freitag november junge in brunnen gefallen	1
997	samstag november bremen ersetzt als erstes b	1
998	morgen in pams steinmeier mit absoluter mehrh	1
999	sonntag november stiftung warentest benotet	1

1000 rows × 2 columns

Предобработка признаков

```
In [3]:
```

```
tfidfv = TfidfVectorizer()
tfidf_ngram_features = tfidfv.fit_transform(df['text'])
tfidf_ngram_features
```

Out[3]:

```
<1000x29611 sparse matrix of type '<class 'numpy.float64'>'
    with 197165 stored elements in Compressed Sparse Row format>
```

In [4]:

```
countvec = CountVectorizer()
countvec_ngram_features = countvec.fit_transform(df['text'])
countvec_ngram_features
```

Out [4]:

```
<1000x29611 sparse matrix of type '<class 'numpy.int64'>'
with 197165 stored elements in Compressed Sparse Row format>
```

Random Forest Classificator

In [5]:

```
# TFIDF + RFC
X_train, X_test, y_train, y_test = train_test_split(tfidf_ngram_features, df['hasImage'], test_size=
model = RandomForestClassifier()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)
print(classification_report(y_test, y_pred, digits=4, target_names=list(map(str, list(y_test.unique)))
```

	precision	recal1	f1-score	support
1 0	0. 9444 0. 8636	0. 4857 0. 9913	0. 6415 0. 9231	70 230
accuracy macro avg weighted avg	0. 9040 0. 8825	0. 7385 0. 8733	0. 8733 0. 7823 0. 8574	300 300 300

In [6]:

```
# CountVec + RFC
X_train, X_test, y_train, y_test = train_test_split(countvec_ngram_features, df['hasImage'], test_si
model = RandomForestClassifier()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)
print(classification_report(y_test, y_pred, digits=4, target_names=list(map(str, list(y_test.unique()))
```

	precision	recall	f1-score	support
1 0	0. 8750 0. 8889	0. 6000 0. 9739	0. 7119 0. 9295	70 230
accuracy macro avg weighted avg	0. 8819 0. 8856	0. 7870 0. 8867	0. 8867 0. 8207 0. 8787	300 300 300

Complement Naive Bayes

In [7]:

```
# TFIDF + CNB
X_train, X_test, y_train, y_test = train_test_split(tfidf_ngram_features, df['hasImage'], test_size=
model = ComplementNB()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)
print(classification_report(y_test, y_pred, digits=4, target_names=list(map(str, list(y_test.unique(str, list(y_test
```

	precision	recal1	f1-score	support
1 0	1. 0000 0. 7718	0. 0286 1. 0000	0. 0556 0. 8712	70 230
accuracy			0.7733	300
macro avg	0.8859	0. 5143	0.4634	300
weighted avg	0.8251	0.7733	0.6809	300

In [8]:

```
# CountVec + CNB
X_train, X_test, y_train, y_test = train_test_split(countvec_ngram_features, df['hasImage'], test_si
model = ComplementNB()
model.fit(X_train, y_train)
y_pred = model.predict(X_test)
print(classification_report(y_test, y_pred, digits=4, target_names=list(map(str, list(y_test.unique)))
```

	precision	recal1	f1-score	support
1 0	0. 8846 0. 8285	0. 3286 0. 9870	0. 4792 0. 9008	70 230
accuracy macro avg weighted avg	0. 8565 0. 8416	0. 6578 0. 8333	0. 8333 0. 6900 0. 8024	300 300 300

Выводы:

- 1. CountVectorizer показал лучший результат в обоих моделях
- 2. Random Forest Classifier показал лучшие, чем Complement Naive Bayes результаты