Отчёт по лабораторной работе №5 студентки группы ИУ5-21М Дьяконовой Светланы

```
import numpy as np
import pandas as pd
from typing import Dict, Tuple
from scipy import stats
from IPython.display import Image
from sklearn.datasets import load iris, load boston
from sklearn.feature extraction.text import CountVectorizer,
TfidfVectorizer
from sklearn.model selection import train test split
from sklearn.neighbors import KNeighborsRegressor,
KNeighborsClassifier
from sklearn.linear model import LogisticRegression
from sklearn.model selection import GridSearchCV, RandomizedSearchCV
from sklearn.metrics import accuracy score, balanced accuracy score
from sklearn.metrics import precision score, recall score, f1 score,
classification report
from sklearn.metrics import confusion matrix
from sklearn.model selection import cross val score
from sklearn.pipeline import Pipeline
from sklearn.metrics import mean absolute error, mean squared error,
mean squared log error, 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 seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set(style="ticks")
from spacy.lang.ru import Russian
from spacy.lang.en import English
import spacy
from spacy import displacy
import gensim
from gensim.models import word2vec
from gensim.models import Word2Vec
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 ассиracy 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
df = pd.read_csv('train.csv', usecols=['Description', 'Class Index'],
nrows=10000)
df
      Class Index
                                                          Description
                   Reuters - Short-sellers, Wall Street's dwindli...
0
1
                   Reuters - Private investment firm Carlyle Grou...
2
                   Reuters - Soaring crude prices plus worries\ab...
3
                   Reuters - Authorities have halted oil export\f...
4
                3
                   AFP - Tearaway world oil prices, toppling reco...
9995
                4
                   Users of the music player should watch out for...
                   BMC Software has released a new version of Pat...
9996
                4
9997
                   The chief of Beijing-backed China Aviation Oil...
9998
                   BRUSSELS The European Commission has opened an...
9999
                   Operation Digital Gridlock targets peer-to-pee...
[10000 \text{ rows } \times 2 \text{ columns}]
1
!python -m spacy download en core web sm
Collecting en core web sm==2.2.5
  Downloading
https://github.com/explosion/spacy-models/releases/download/en core we
b sm-2.2.5/en core web sm-2.2.5.tar.gz (12.0 MB)
ent already satisfied: spacy>=2.2.2 in /usr/local/lib/python3.7/dist-
```

```
packages (from en core web sm==2.2.5) (2.2.4)
Requirement already satisfied: plac<1.2.0,>=0.9.6 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (1.1.3)
Requirement already satisfied: murmurhash<1.1.0,>=0.28.0 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (1.0.7)
Requirement already satisfied: numpy>=1.15.0 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (1.21.6)
Requirement already satisfied: preshed<3.1.0,>=3.0.2 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (3.0.6)
Requirement already satisfied: srsly<1.1.0,>=1.0.2 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (1.0.5)
Requirement already satisfied: cymem<2.1.0,>=2.0.2 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (2.0.6)
Requirement already satisfied: setuptools in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (57.4.0)
Requirement already satisfied: tgdm<5.0.0,>=4.38.0 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (4.64.0)
Requirement already satisfied: wasabi<1.1.0,>=0.4.0 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (0.9.1)
Requirement already satisfied: requests<3.0.0,>=2.13.0 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (2.23.0)
Requirement already satisfied: catalogue<1.1.0,>=0.0.7 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (1.0.0)
Requirement already satisfied: thinc==7.4.0 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (7.4.0)
Requirement already satisfied: blis<0.5.0,>=0.4.0 in
/usr/local/lib/python3.7/dist-packages (from spacy>=2.2.2-
>en core web sm==2.2.5) (0.4.1)
Requirement already satisfied: importlib-metadata>=0.20 in
/usr/local/lib/python3.7/dist-packages (from catalogue<1.1.0,>=0.0.7-
>spacy>=2.2.2->en core web sm==2.2.5) (4.11.3)
Requirement already satisfied: typing-extensions>=3.6.4 in
/usr/local/lib/python3.7/dist-packages (from importlib-metadata>=0.20-
>catalogue<1.1.0,>=0.0.7->spacy>=2.2.2->en core_web_sm==2.2.5) (4.2.0)
Requirement already satisfied: zipp>=0.5 in
/usr/local/lib/python3.7/dist-packages (from importlib-metadata>=0.20-
>catalogue<1.1.0,>=0.0.7->spacy>=2.2.2->en core web sm==2.2.5) (3.8.0)
Requirement already satisfied: urllib3!=1.25.0,!=1.25.1,<1.26,>=1.21.1
```

```
in /usr/local/lib/python3.7/dist-packages (from
requests<3.0.0,>=2.13.0->spacy>=2.2.2->en core web sm==2.2.5) (1.24.3)
Requirement already satisfied: certifi>=2017.4.17 in
/usr/local/lib/python3.7/dist-packages (from requests<3.0.0,>=2.13.0-
>spacy>=2.2.2->en core web sm==2.2.5) (2021.10.8)
Requirement already satisfied: chardet<4,>=3.0.2 in
/usr/local/lib/python3.7/dist-packages (from requests<3.0.0,>=2.13.0-
>spacy>=2.2.2->en core web sm==2.2.5) (3.0.4)
Requirement already satisfied: idna<3,>=2.5 in
/usr/local/lib/python3.7/dist-packages (from requests<3.0.0,>=2.13.0-
>spacy>=2.2.2->en core web sm==2.2.5) (2.10)
✓ Download and installation successful
You can now load the model via spacy.load('en core web sm')
# Tokenize
nlp = spacy.load('en core web sm')
spacy text1 = nlp(df['Description'][0])
spacy text1
Reuters - Short-sellers, Wall Street's dwindling\band of ultra-cynics,
are seeing green again.
# POS-tagging
for token in spacy text1:
    print('{} - {} - {}'.format(token.text, token.pos , token.dep ))
Reuters - PROPN - compound
- - PUNCT - punct
Short - PROPN - compound
- - PUNCT - punct
sellers - NOUN - nsubi
, - PUNCT - punct
Wall - PROPN - compound
Street - PROPN - poss
's - PART - case
dwindling\band - NOUN - appos
of - ADP - prep
ultra - ADJ - dep
- - NOUN - dep
cynics - NOUN - pobj
, - PUNCT - punct
are - AUX - aux
seeing - VERB - ROOT
green - ADJ - dobj
again - ADV - advmod
. - PUNCT - punct
# lemmatization
for token in spacy text1:
      print(token, token.lemma, token.lemma )
```

```
Reuters 17690189795809227049 Reuters
- 9153284864653046197 -
Short 7602511986449097345 Short
- 9153284864653046197 -
sellers 15370787710306132986 seller
, 2593208677638477497 ,
Wall 8806908179280924345 Wall
Street 11849903144683346075 Street
's 16428057658620181782 's
dwindling\band 16257729132875505059 dwindling\band
of 886050111519832510 of
ultra 8871178158770945376 ultra
- 9153284864653046197 -
cynics 14078618470999779890 cynic
, 2593208677638477497
are 10382539506755952630 be
seeing 11925638236994514241 see
green 3487151913243070096 green
again 4502205900248518970 again
. 12646065887601541794 .
# NFR
displacy.render(spacy text1, style='ent', jupyter=True)
<IPython.core.display.HTML object>
# dependency
displacy.render(spacy text1, style='dep', jupyter=True)
<IPython.core.display.HTML object>
2
TFIDF
tfidfv = TfidfVectorizer(ngram range=(1,1))
tfidf ngram features = tfidfv.fit transform(df['Description'])
tfidf ngram features
<10000x20257 sparse matrix of type '<class 'numpy.float64'>'
     with 283180 stored elements in Compressed Sparse Row format>
X train, X test, y train, y test =
train test split(tfidf ngram features, df['Class Index'],
test size=0.3, random state=1)
model = LogisticRegression()
model.fit(X train, y train)
y pred = model.predict(X test)
```

/Users/danilafedyukin/opt/anaconda3/lib/python3.8/site-packages/sklearn/linear_model/_logistic.py:763: ConvergenceWarning: lbfgs failed to converge (status=1):

STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:

https://scikit-learn.org/stable/modules/preprocessing.html Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

n_iter_i = _check_optimize_result(

print(classification_report(y_test, y_pred, digits=4,
target names=list(map(str, list(y test.unique())))))

support	f1-score	recall	precision	
741 699 741 819	0.8524 0.9116 0.8330 0.8497	0.8259 0.9299 0.8178 0.8730	0.8806 0.8941 0.8487 0.8275	1 4 3 2
3000 3000 3000	0.8610 0.8617 0.8607	0.8617 0.8610	0.8627 0.8614	accuracy macro avg weighted avg

[nltk data] Downloading package stopwords to

word2vec

[nltk_data]

[nltk data]

```
import re
import pandas as pd
import numpy as np
from typing import Dict, Tuple
from sklearn.metrics import accuracy_score, balanced_accuracy_score
from sklearn.feature_extraction.text import CountVectorizer,
TfidfVectorizer
from sklearn.linear_model import LogisticRegression
from sklearn.pipeline import Pipeline
from nltk import WordPunctTokenizer
from nltk.corpus import stopwords
import nltk
nltk.download('stopwords')
from gensim.test.utils import datapath
```

/Users/danilafedyukin/nltk data...

Unzipping corpora/stopwords.zip.

```
model =
gensim.models.KeyedVectors.load word2vec format(datapath('word2vec pre
_kv_c'), binary=False)
class EmbeddingVectorizer(object):
    Для текста усредним вектора входящих в него слов
    def init (self, model):
        self.model = model
        self.size = model.vector size
    def fit(self, X, y):
        return self
    def transform(self, X):
        return np.array([np.mean(
            [self.model[w] for w in words if w in self.model]
            or [np.zeros(self.size)], axis=0)
            for words in X1)
corpus = []
stop words = stopwords.words('english')
tok = WordPunctTokenizer()
for line in df['Description'].values:
    line1 = line.strip().lower()
    line1 = re.sub("[^a-zA-Z]"," ", line1)
    text tok = tok.tokenize(line1)
    text_tok1 = [w for w in text_tok if not w in stop_words]
    corpus.append(text tok1)
model imdb = word2vec.Word2Vec(corpus, workers=4, min count=10,
window=10, sample=1e-3)
def sentiment(v, c):
    model = Pipeline(
        [("vectorizer", v),
         ("classifier", c)])
    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(np.unique(y test))))))
boundary = 700
X train = corpus[:boundary]
X test = corpus[boundary:]
y train = df['Class Index'].values[:boundary]
y test = df['Class Index'].values[boundary:]
sentiment(EmbeddingVectorizer(model imdb.wv), LogisticRegression())
```

/Users/danilafedyukin/opt/anaconda3/lib/python3.8/site-packages/sklearn/linear_model/_logistic.py:763: ConvergenceWarning: lbfgs failed to converge (status=1):

STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

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https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

n_iter_i = _check_optimize_result(

	precision	recall	f1-score	support
1 2 3 4	0.8913 0.8888 0.8059 0.4275	0.5718 0.7039 0.3318 0.9398	0.6967 0.7856 0.4701 0.5876	2410 2236 2378 2276
accuracy macro avg weighted avg	0.7534 0.7554	0.6368 0.6323	0.6323 0.6350 0.6334	9300 9300 9300