Московский государственный технический университет им. Н.Э. Баумана Кафедра «Системы обработки информации и управления»

Лабораторная работа №4 по дисциплине «Методы машинного обучения» на тему «Подготовка обучающей и тестовой выборки, кросс-валидация и подбор гиперпараметров на примере метода ближайших соседей.»

Выполнил: студент группы ИУ5-61Б Белоусов Е. А.

1. Цель

изучение сложных способов подготовки выборки и подбора гиперпараметров на примере метода ближайших соседей.

2. Задание

- 1. Выберите набор данных (датасет) для решения задачи классификации или регрессии.
- 2. С использованием метода train_test_split разделите выборку на обучающую и тестовую.
- 3. Обучите модель ближайших соседей для произвольно заданного гиперпараметра К. Оцените качество модели с помощью подходящих для задачи метрик.
- 4. Постройте модель и оцените качество модели с использованием кросс-валидации.
- 5. Произведите подбор гиперпараметра K с использованием GridSearchCV и кроссвалидации.

```
In [1]: import numpy as np
     import pandas as pd
     import sklearn
    import warnings
    warnings.filterwarnings('ignore')
   Для данной задачи выберем датасет с красными винами
In [2]: data = pd.read_csv('../data/winequality-red.csv')
In [3]: data.head()
         fixed acidity volatile acidity citric acid residual sugar chlorides \
Out[3]:
     0
             7.4
                        0.70
                                  0.00
                                              1.9
                                                     0.076
     1
             7.8
                        88.0
                                  0.00
                                              2.6
                                                     0.098
     2
             7.8
                        0.76
                                  0.04
                                              2.3
                                                     0.092
     3
            11.2
                         0.28
                                  0.56
                                               1.9
                                                     0.075
     4
             7.4
                        0.70
                                  0.00
                                              1.9
                                                     0.076
      free sulfur dioxide total sulfur dioxide density
                                                      pH sulphates \
                               34.0 0.9978 3.51
     0
                11.0
                                                     0.56
     1
                25.0
                               67.0 0.9968 3.20
                                                      0.68
```

2	15.0	54.0	0.9970 3.2	26 0.65
3	17.0	60.0	0.9980 3.3	16 0.58
4	11.0	34.0	0.9978 3.5	0.56

alcohol quality 0 9.4 5 1 9.8 5

2 9.8 53 9.8 64 9.4 5

In [4]: data.dtypes

```
Out[4]: fixed acidity
                           float64
     volatile acidity
                         float64
    citric acid
                       float64
    residual sugar
                         float64
     chlorides
                       float64
    free sulfur dioxide
                          float64
     total sulfur dioxide float64
                      float64
     density
    pН
                     float64
                       float64
    sulphates
     alcohol
                       float64
     quality
                       int64
     dtype: object
In [5]: data.shape
Out[5]: (1599, 12)
In [6]: wine_target = data['quality']
     del data['quality']
In [7]: wine_target[:10]
Out[7]: 0 5
     1
        5
     2
        5
     3
        6
     4
        5
     5
        5
     6
        5
     7
        7
     8
        7
     9
    Name: quality, dtype: int64
In [8]: data.head()
Out[8]:
         fixed acidity volatile acidity citric acid residual sugar chlorides \
                         0.70
                                   0.00
     0
             7.4
                                                1.9
                                                       0.076
     1
             7.8
                         0.88
                                   0.00
                                                2.6
                                                       0.098
     2
             7.8
                         0.76
                                   0.04
                                                2.3
                                                       0.092
     3
                          0.28
                                    0.56
             11.2
                                                1.9
                                                       0.075
             7.4
     4
                         0.70
                                   0.00
                                                1.9
                                                       0.076
      free sulfur dioxide total sulfur dioxide density
                                                        pH sulphates \
                                34.0 0.9978 3.51
     0
                 11.0
                                                       0.56
                 25.0
                                67.0 0.9968 3.20
     1
                                                       0.68
     2
                 15.0
                                54.0 0.9970 3.26
                                                       0.65
     3
                                60.0 0.9980 3.16
                 17.0
                                                        0.58
     4
                 11.0
                                34.0 0.9978 3.51
                                                       0.56
```

```
9.4
    0
         9.8
    1
    2
         9.8
    3
         9.8
    4
         9.4
   Разделяем выборку на обучающую и тестовую
In [9]: from sklearn.model_selection import train_test_split
    wine_X_train, wine_X_test, wine_y_train, wine_y_test = train_test_split(data, wine_target, test_size=
   Обучаем модель ближайших соседей для произвольно заданного гиперпараметра k=3
Оценим качество модели с помощбю МАЕ
In [10]: from sklearn.neighbors import KNeighborsClassifier
     from sklearn.metrics import mean_absolute_error
     model\_1 = KNeighborsClassifier(n\_neighbors=3)
     model_1.fit(wine_X_train, wine_y_train)
     target = model_1.predict(wine_X_test)
     mean_absolute_error(wine_y_test, target)
Out[10]: 0.583333333333333333
   Строим модель и оценивем ее используя кросс-валидацию
In [11]: from sklearn.model_selection import cross_val_score
     scores = cross_val_score(KNeighborsClassifier(n_neighbors=3), data, wine_target, cv=5, scoring='m
     scores, np.mean(scores)
Out[11]: (array([-0.63354037, -0.62305296, -0.63862928, -0.8081761, -0.66561514]),
      -0.67380277164924562)
   произведем подбор гиперпараметров используя GridSearch
In [12]: from sklearn.model_selection import GridSearchCV
     n_range = np.arange(1, 50)
     turned_parametrs = [{'n_neighbors' : n_range}]
     turned_parametrs
Out[12]: [{'n_neighbors': array([ 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16, 17,
          18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28, 29, 30, 31, 32, 33, 34,
          35, 36, 37, 38, 39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49])}]
In [13]: %%time
     clf_gs = GridSearchCV(KNeighborsClassifier(), turned_parametrs, cv=5, scoring='mean_absolute_er
     clf_gs.fit(wine_X_train, wine_y_train)
CPU times: user 3.11 s, sys: 0 ns, total: 3.11 s
Wall time: 3.1 s
In [14]: clf_gs.cv_results_
```

alcohol

```
Out[14]: {'mean_fit_time': array([ 0.00178742, 0.00129266, 0.00114484, 0.00118623, 0.0011426,
                                     0.00118914, 0.00122261, 0.00116243, 0.00121493, 0.0011888,
                                     0.00119019, 0.00119557, 0.00119338, 0.00124426, 0.00125017,
                                     0.00116644, 0.00122519, 0.00136981, 0.00136666, 0.00161529,
                                     0.00120463, 0.00132117, 0.00120082, 0.00119681, 0.00124373,
                                     0.00121365, 0.00118818, 0.00126786, 0.00120974, 0.0012434,
                                     0.0012044, 0.00121074, 0.00135059, 0.00142961, 0.00154557,
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                                     0.00270681, 0.00267005, 0.00285244, 0.00308123, 0.00323558,
                                     0.00291739, 0.00284786, 0.00288582, 0.00295944, 0.00297804,
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                                   -0.5844504, -0.5665773, -0.57640751, -0.56747096, -0.55942806,
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                                   -0.55495979, -0.55317248, -0.56032172, -0.55227882, -0.56389634,
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                      False 
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```

```
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```

```
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```

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In [15]: # Лучшая модель clf_gs.best_estimator_

Out[15]: KNeighborsClassifier(algorithm='auto', leaf_size=30, metric='minkowski', metric_params=None, n_jobs=1, n_neighbors=34, p=2, weights='uniform')

In [16]: # Лучшее значение метрики clf_gs.best_score_

Out[16]: -0.55227882037533516

In [17]: # Лучшее значение параметров clf_gs.best_params_

Out[17]: {'n_neighbors': 34}

In [18]: %matplotlib inline import matplotlib.pyplot as plt

Изменение качества на тестовой выборке в зависимости от K-соседей plt.plot(n_range, clf_gs.cv_results_['mean_test_score'])

Out[18]: [<matplotlib.lines.Line2D at 0x7f4e39a42c10>]

