

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

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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 `K`. 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 кросс-валидации.

```
[1]: import numpy as np
import pandas as pd
import sklearn
import warnings
warnings.filterwarnings('ignore')
```

Для данной задачи выберем датасет с красными винами

```
[2]: data = pd.read_csv('../data/winequality-red.csv')
```

```
[3]: data.head()
```

```
[3]: fixed acidity volatile acidity citric acid residual sugar chlorides \
0      7.4          0.70      0.00      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     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 \
0          11.0          34.0  0.9978 3.51    0.56
1          25.0          67.0  0.9968 3.20    0.68
2          15.0          54.0  0.9970 3.26    0.65
3          17.0          60.0  0.9980 3.16    0.58
4          11.0          34.0  0.9978 3.51    0.56
```

```
alcohol quality
0    9.4      5
1    9.8      5
2    9.8      5
3    9.8      6
4    9.4      5
```

```
[4]: data.dtypes
```

```
[4]: fixed acidity    float64
      volatile acidity float64
      citric acid     float64
      residual sugar  float64
      chlorides       float64
      free sulfur dioxide float64
      total sulfur dioxide float64
      density         float64
      pH              float64
      sulphates       float64
      alcohol         float64
      quality         int64
      dtype: object
```

```
[5]: data.shape
```

```
[5]: (1599, 12)
```

```
[6]: wine_target = data['quality']
      del data['quality']
```

```
[7]: wine_target[:10]
```

```
[7]: 0  5
      1  5
      2  5
      3  6
      4  5
      5  5
      6  5
      7  7
      8  7
      9  5
      Name: quality, dtype: int64
```

```
[8]: data.head()
```

```
[8]: fixed acidity volatile acidity citric acid residual sugar chlorides \
0      7.4         0.70      0.00      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     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 \
0         11.0          34.0  0.9978  3.51    0.56
1         25.0          67.0  0.9968  3.20    0.68
2         15.0          54.0  0.9970  3.26    0.65
3         17.0          60.0  0.9980  3.16    0.58
4         11.0          34.0  0.9978  3.51    0.56
```

```

alcohol
0    9.4
1    9.8
2    9.8
3    9.8
4    9.4

```

Разделяем выборку на обучающую и тестовую

```

[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=0.3, random_state=1)

```

Обучаем модель ближайших соседей для произвольно заданного гиперпараметра k=3
Оценим качество модели с помощью MAE

```

[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)

```

```

[10]: 0.5833333333333333

```

Строим модель и оцениваем ее используя кросс-валидацию

```

[11]: from sklearn.model_selection import cross_val_score
     scores = cross_val_score(KNeighborsClassifier(n_neighbors=3), data, wine_target, cv=5,
     ↪ scoring='mean_absolute_error')
     scores, np.mean(scores)

```

```

[11]: (array([-0.63354037, -0.62305296, -0.63862928, -0.8081761 , -0.66561514]),
     -0.67380277164924562)

```

произведем подбор гиперпараметров используя GridSearch

```

[12]: from sklearn.model_selection import GridSearchCV
     n_range = np.arange(1, 50)
     turned_params = [{'n_neighbors': n_range}]
     turned_params

```

```

[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])}]

```

```

[13]: %%time
     clf_gs = GridSearchCV(KNeighborsClassifier(), turned_params, cv=5,
     ↪ scoring='mean_absolute_error')
     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

```
[14]: clf_gs.cv_results_
```

```
[14]: {'mean_fit_time': array([ 0.00178742, 0.00129266, 0.00114484, 0.00118623,
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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],
mask = [False False False False False False False False
```

False False False

False False False False False False False False False False False
False False False False False False False False False False False
False False False False False False False False False False False
False],

fill_value = ?),

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```

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```

```

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```



```

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```

```

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0.02549575, 0.02636609, 0.01274455, 0.02347102, 0.02479964,
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0.03175331, 0.0237958 , 0.02118148, 0.02857687, 0.0239226 ,
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0.03269609, 0.02563156, 0.04462335, 0.03496263, 0.03721895,
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0.01408809, 0.01343035, 0.01531873, 0.00971561, 0.01467019,
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```

```

[15]: # Лучшая модель
      clf_gs.best_estimator_

```

```

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

```

```

[16]: # Лучшее значение метрики
      clf_gs.best_score_

```

```

[16]: -0.55227882037533516

```

```

[17]: # Лучшее значение параметров
      clf_gs.best_params_

```

```

[17]: {'n_neighbors': 34}

```

```

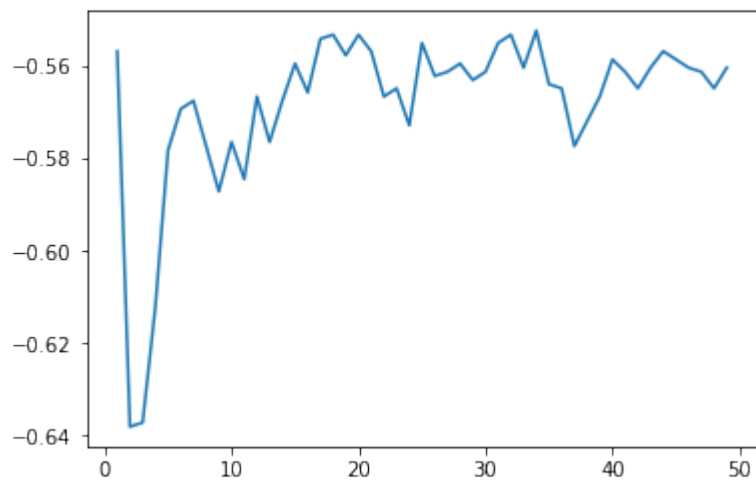
[18]: %matplotlib inline
      import matplotlib.pyplot as plt
      # Изменение качества на тестовой выборке в зависимости от K-соседей
      plt.plot(n_range, clf_gs.cv_results_['mean_test_score'])

```

```

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

```



[]: