Московский государственный технический университет им. Н.Э. Баумана
Факультет «Информатика и системы управления»

Кафедра «Системы обработки информации и управления»



Лабораторная работа №4 по дисциплине

«Методы машинного обучения»

«Подготовка обучающей и тестовой выборки, кросс-валидация и подбор гиперпараметров на примере метода ближайших соседей»

исполнитель:

		Макаров Д. А.
		Группа ИУ5-21М
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```
In [3]:
from datetime import datetime
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
from sklearn.metrics import mean_absolute_error
from sklearn.metrics import median_absolute_error, r2_score
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import KFold, RepeatedKFold, ShuffleSplit
from sklearn.model_selection import cross_val_score, train_test_split
from sklearn.model_selection import learning_curve, validation_curve
from sklearn.neighbors import KNeighborsRegressor
from sklearn.preprocessing import StandardScaler
# Enable inline plots
%matplotlib inline
Выбор датасета и его исследование
In [5]:
```

```
data = pd.read csv('heart.csv', sep=',')
data.head(10)
```

Out[5]:

	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	са	thal	target
0	63	1	3	145	233	1	0	150	0	2.3	0	0	1	1
1	37	1	2	130	250	0	1	187	0	3.5	0	0	2	1
2	41	0	1	130	204	0	0	172	0	1.4	2	0	2	1
3	56	1	1	120	236	0	1	178	0	0.8	2	0	2	1
4	57	0	0	120	354	0	1	163	1	0.6	2	0	2	1
5	57	1	0	140	192	0	1	148	0	0.4	1	0	1	1
6	56	0	1	140	294	0	0	153	0	1.3	1	0	2	1
7	44	1	1	120	263	0	1	173	0	0.0	2	0	3	1
8	52	1	2	172	199	1	1	162	0	0.5	2	0	3	1
9	57	1	2	150	168	0	1	174	0	1.6	2	0	2	1

```
In [6]:
```

```
data.shape
```

Out[6]:

(303, 14)

In [7]:

```
data.dtypes
```

Out[7]:

```
int64
age
           int64
sex
           int64
trestbps
          int64
chol
           int64
           int64
fbs
            int64
restecg
           int64
thalach
           int64
exang
oldpeak
        float64
```

```
slope
           int64
            int64
ca
thal int64 target int64
dtype: object
In [8]:
data.isnull().sum()
Out[8]:
sex
cp v trestbps 0
fbs
restecg 0
thalach 0
exang
          0
        0
oldpeak
slope
thal 0 target 0
dtype: int64
```

3. Разделение выборки на обучающую и тестовую.

```
In [9]:
```

```
# Перейдем к разделению выборки на обучающую и тестовую.

X = data.drop('target',axis = 1).values

y = data['target'].values
```

```
In [10]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.35, random_state=1)
```

In [11]:

```
print(X_train.shape)
print(X_test.shape)
print(y_train.shape)
print(y_test.shape)

(196, 13)
(107, 13)
(196,)
(196,)
```

4.Обучите модель ближайших соседей для произвольно заданного гиперпараметра К. Оцените качество модели с помощью трех подходящих для задачи метрик.

```
In [12]:
```

```
def test_model(model):
    print("mean_absolute_error:",
    mean_absolute_error(y_test, model.predict(X_test)))
    print("median_absolute_error:",
    median_absolute_error(y_test, model.predict(X_test)))
    print("r2_score:",
    r2_score(y_test, model.predict(X_test)))
```

5. Постройте модель и оцените качество модели с использованием кроссвалидации. Проведите эксперименты с тремя различными стратегиями кросс-валидации.

K-Fold

Repeated K-Fold

```
In [16]:

scores = cross_val_score(KNeighborsRegressor(n_neighbors=5), X, y, cv=RepeatedKFold(n_splits=5, n_repeats=2),
scoring="r2")
print(scores)
print(scores.mean(), "±", scores.std())

[-0.17802151 -0.10305011 0.01238095 0.3352614 0.05714286 -0.03857759 0.02131868 0.08034858 -0.008 0.06650999]
0.024531326401118825 ± 0.1284515912142582
```

Shuffle Split

```
In [17]:
```

6. Произведите подбор гиперпараметра K с использованием GridSearchCV и кросс-валидации.

```
In [18]:
n range = np.array(range(1, 14, 2))
tuned_parameters = [{'n_neighbors': n_range}]
Out[18]:
array([ 1, 3, 5, 7, 9, 11, 13])
In [19]:
gs = GridSearchCV(KNeighborsRegressor(), tuned parameters,
cv=ShuffleSplit(n_splits=10), scoring="r2",
return_train_score=True, n_jobs=-1)
gs.fit(X, y)
gs.best_params_
Out[19]:
{'n neighbors': 13}
In [20]:
plt.plot(n_range, gs.cv_results_["mean_train_score"])
Out[20]:
[<matplotlib.lines.Line2D at 0x7f9a51cd15f8>]
1.0
 0.9
 0.8
 0.7
 0.6
 0.5
 0.4
 0.3
 0.2
                                 10
In [21]:
plt.plot(n_range, gs.cv_results_["mean_test_score"])
Out[21]:
[<matplotlib.lines.Line2D at 0x7f9a889fa940>]
  0.0
 -0.2
 -0.4
```

-0.6

7. Повторите пункт 4 для найденного оптимального значения гиперпараметра К. Сравните качество полученной модели с качеством модели, полученной в пункте 4.

In [22]:

```
reg = KNeighborsRegressor(**gs.best_params_)
reg.fit(X_train, y_train)
test_model(reg)
```

mean_absolute_error: 0.43278217109992817
median_absolute_error: 0.38461538461538464

r2_score: 0.05096231703519161

8. Постройте кривые обучения и валидации.

In [23]:

```
def plot_learning_curve(estimator, title, X, y, ylim=None, cv=None):
    train sizes=np.linspace(.1, 1.0, 5)
    plt.figure()
    plt.title(title)
    if ylim is not None:
       plt.ylim(*ylim)
    plt.xlabel("Training examples")
    plt.ylabel("Score")
    train_sizes, train_scores, test_scores = learning_curve(
    estimator, X, y, cv=cv, n jobs=-1, train sizes=train sizes)
    train_scores_mean = np.mean(train_scores, axis=1)
    train_scores_std = np.std(train_scores, axis=1)
    test scores mean = np.mean(test scores, axis=1)
    test_scores_std = np.std(test_scores, axis=1)
    plt.grid()
    plt.fill between (train sizes, train scores mean - train scores std,
    train_scores_mean + train_scores_std, alpha=0.1, color="r")
    plt.fill_between(train_sizes, test_scores_mean - test_scores_std,
    test_scores_mean + test_scores_std, alpha=0.1, color="g")
    plt.plot(train_sizes, train_scores_mean, 'o-', color="r", label="Training score")
    plt.plot(train_sizes, test_scores_mean, 'o-', color="g", label="Cross-validation score")
    plt.legend(loc="best")
    return plt
```

In [24]:

```
plot_learning_curve(reg, str(gs.best_params_), X, y,
cv=ShuffleSplit(n_splits=10))
```

Out[24]:

 $\label{limits} $$\operatorname{module 'matplotlib.pyplot' from '/home/denis/mllabs/env/lib/python3.6/site-packages/matplotlib/pyplot.py'>$$$



```
-0.05 50 100 150 200 250 Training examples
```

In [25]:

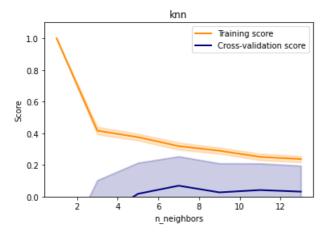
```
def plot_validation_curve(estimator, title, X, y, param_name, param_range, cv, scoring="accuracy"):
   train_scores, test_scores = validation_curve(
    estimator, X, y, param_name=param_name,
    param_range=param_range,
    cv=cv, scoring=scoring, n_jobs=-1)
    train scores mean = np.mean(train scores, axis=1)
    train_scores_std = np.std(train_scores, axis=1)
    test_scores_mean = np.mean(test_scores, axis=1)
    test scores std = np.std(test scores, axis=1)
    plt.title(title)
    plt.xlabel(param name)
    plt.ylabel("Score")
    plt.ylim(0.0, 1.1)
    plt.plot(param_range, train_scores_mean, label="Training score",
    color="darkorange", lw=lw)
    plt.fill between (param range, train scores mean - train scores std,
    train_scores_mean + train_scores_std, alpha=0.2,
    color="darkorange", lw=lw)
    plt.plot(param_range, test_scores_mean,
    label="Cross-validation score",
    color="navy", lw=lw)
    plt.fill_between(param_range, test_scores_mean - test_scores_std,
    test_scores_mean + test_scores_std, alpha=0.2,
    color="navy", lw=lw)
    plt.legend(loc="best")
    return plt
```

In [26]:

```
plot_validation_curve(KNeighborsRegressor(), "knn", X, y,
param_name="n_neighbors", param_range=n_range,
cv=ShuffleSplit(n_splits=10), scoring="r2")
```

Out[26]:

<module 'matplotlib.pyplot' from '/home/denis/mllabs/env/lib/python3.6/site-packages/matplotlib/pyplot.py'>



In []: