

Министерство науки и высшего образования Российской Федерации Федеральное государственное бюджетное образовательное учреждение высшего образования «Московский государственный технический университет имени Н.Э. Баумана (национальный исследовательский университет)» (МГТУ им. Н.Э. Баумана)

## Работа №5 по курсу «Технологии машинного обучения»

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## 1 Исходное задание

- 1. Выберите набор данных (датасет) для решения задачи классификации или регрессии.
- 2. В случае необходимости проведите удаление или заполнение пропусков и кодирование категориальных признаков
- 3. С использованием метода train\_test\_split разделите выборку на обучающую и тестовую.
- 4. Обучите две ансамблевые модели. Оцените качество моделей с помощью одной из подходящих для задачи метрик. Сравните качество полученных моделей.

## 2 Код программы

```
[21]: from IPython.display import Image
      import numpy as np
      import pandas as pd
      import graphviz
      import pydotplus
      from sklearn.tree import DecisionTreeClassifier, DecisionTreeRegressor, u
      ⇔export_graphviz
      from sklearn import svm, datasets
      from sklearn.model_selection import train_test_split
      from sklearn.model_selection import cross_val_score, cross_validate
      from sklearn.model_selection import KFold, RepeatedKFold
      from sklearn.metrics import mean_absolute_error, mean_squared_error, __
      →mean_squared_log_error, median_absolute_error, r2_score
      from sklearn.model_selection import GridSearchCV
      import seaborn as sns
      import matplotlib.pyplot as plt
      from sklearn.linear_model import LinearRegression
      from sklearn.linear_model import Ridge,ElasticNet
      from sklearn.preprocessing import OrdinalEncoder, StandardScaler
      from sklearn.tree import DecisionTreeRegressor
      from sklearn.ensemble import RandomForestRegressor
      from sklearn.datasets import load_wine
      from io import StringIO
      %matplotlib inline
      pd.set_option("display.max_rows", None, "display.max_columns", None)
      sns.set(style="ticks")
```

```
[22]: data = pd.read_csv("/home/igor/Downloads/CarPrice_Assignment.xls",sep=',')
      data.shape
[22]: (205, 26)
[23]: cleanup_nums = {"doornumber": {"four": 1, "two": 0},
                      "cylindernumber": {"four": 4, "six": 6, "five": 5, "eight": 8,
                                        "two": 2, "twelve": 12, "three":3 },
                      "aspiration":{"std": 0, "turbo": 1},
                      "fueltype": { "gas": 0, "diesel": 1},
                      "enginelocation":{"front": 0, "rear": 1}}
      data = data.replace(cleanup_nums)
      data=pd.get_dummies(data, columns=["drivewheel"], prefix=["drive"])
      data=pd.get_dummies(data, columns=["carbody"], prefix=["body"])
      data["OHC_Code"] = np.where(data["enginetype"].str.contains("ohc"), 1, 0)
      data.drop(data[(data['aspiration']=='turbo')].index,inplace=True)
      data.drop(data[(data['fueltype'] == 'diesel')].index,inplace=True)
      data.
      →drop(["CarName","enginetype","fuelsystem","symboling","car_ID"],axis=1,inplace=True)
[24]: data_X = data.loc[:,data.columns]
      clnm = StandardScaler()
      data_X = clnm.fit_transform(data_X)
      data_X = pd.DataFrame(data_X,columns=data.columns)
      data_Y = data.loc[:, 'price']
      data_X.drop(['price'],axis=1,inplace=True)
      data_X_train, data_X_test, data_y_train, data_y_test = train_test_split(
          data_X, data_Y,test_size=0.4, random_state=360)
      data_Y.head()
[24]: 0
           13495.0
      1
          16500.0
      2
          16500.0
          13950.0
      3
          17450.0
      Name: price, dtype: float64
[25]: data_X.head()
```

```
0 -0.328798
                    -0.469295
                                -1.130388
                                                 -0.121867
                                                           -1.690772
                                                                       -0.426521
      1 -0.328798
                    -0.469295
                                -1.130388
                                                 -0.121867
                                                            -1.690772
                                                                       -0.426521
                                                -0.121867
      2 -0.328798
                    -0.469295
                                -1.130388
                                                            -0.708596
                                                                      -0.231513
      3 -0.328798
                    -0.469295
                                 0.884652
                                                 -0.121867
                                                                        0.207256
                                                             0.173698
      4 -0.328798
                    -0.469295
                                 0.884652
                                                 -0.121867
                                                             0.107110
                                                                        0.207256
         carwidth
                   carheight
                              curbweight
                                          cylindernumber
                                                           enginesize
                                                                       boreratio
     0 -0.844782
                   -2.020417
                               -0.014566
                                                -0.352887
                                                             0.074449
                                                                        0.519071
      1 -0.844782
                   -2.020417
                               -0.014566
                                                -0.352887
                                                             0.074449
                                                                        0.519071
      2 -0.190566
                   -0.543527
                                0.514882
                                                 1.502032
                                                             0.604046 -2.404880
        0.136542
                    0.235942
                               -0.420797
                                                -0.352887
                                                            -0.431076
                                                                      -0.517266
                    0.235942
        0.230001
                                0.516807
                                                 0.574572
                                                             0.218885
                                                                       -0.517266
           stroke
                   compressionratio
                                    horsepower
                                                  peakrpm
                                                             citympg
                                                                      highwaympg
      0 -1.839377
                          -0.288349
                                       0.174483 -0.262960 -0.646553
                                                                       -0.546059
      1 -1.839377
                                       0.174483 -0.262960 -0.646553
                          -0.288349
                                                                       -0.546059
                                       1.264536 -0.262960 -0.953012
        0.685946
                          -0.288349
                                                                       -0.691627
        0.462183
                          -0.035973
                                                 0.787855 -0.186865
                                      -0.053668
                                                                       -0.109354
      4 0.462183
                          -0.540725
                                       0.275883 0.787855 -1.106241
                                                                       -1.273900
         drive_4wd drive_fwd drive_rwd body_convertible body_hardtop \
      0 -0.214286 -1.188177
                                1.302831
                                                  5.759051
                                                                -0.201517
        -0.214286 -1.188177
                                1.302831
                                                  5.759051
                                                                -0.201517
      2 -0.214286 -1.188177
                                1.302831
                                                  -0.173640
                                                                -0.201517
        -0.214286
                     0.841625
                               -0.767559
                                                  -0.173640
                                                                -0.201517
          4.666667 -1.188177 -0.767559
                                                  -0.173640
                                                                -0.201517
         body_hatchback
                         body_sedan
                                     body_wagon
                                                  OHC_Code
      0
              -0.720082
                          -0.938474
                                      -0.372678
                                                  0.290957
      1
              -0.720082
                          -0.938474
                                      -0.372678
                                                 0.290957
      2
              1.388730
                          -0.938474
                                      -0.372678
                                                  0.290957
                                      -0.372678
      3
              -0.720082
                           1.065559
                                                  0.290957
              -0.720082
                           1.065559
                                      -0.372678
                                                 0.290957
[26]: \# data_X = data_X.to_numpy()
      #data_Y = data_Y.to_numpy()
      data_X_train, data_X_test, data_Y_train, data_Y_test = train_test_split(
          data_X, data_Y,test_size=0.4, random_state=360)
```

enginelocation wheelbase

carlength \

[25]:

fueltype

aspiration doornumber

```
data_X_train.shape
[26]: (123, 27)
     1я модель
[27]: from sklearn.ensemble import BaggingRegressor
      reg1 = BaggingRegressor(base_estimator = ElasticNet(tol=1e-1),n_estimators = __
      →5,oob_score=True, random_state=10)
      reg1.fit(data_X_train,data_Y_train)
     /home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/_bagging.py:1069:
     UserWarning: Some inputs do not have OOB scores. This probably means too few
     estimators were used to compute any reliable oob estimates.
       warn("Some inputs do not have OOB scores. "
[27]: BaggingRegressor(base_estimator=ElasticNet(tol=0.1), n_estimators=5,
                       oob_score=True, random_state=10)
[29]: target1_0=reg1.predict(data_X_test)
      r2_score(data_Y_test, target1_0)
[29]: 0.8312979740120918
[30]: #
      def get_png_tree(tree_model_param, feature_names_param):
          dot data = StringIO()
          export_graphviz(tree_model_param, out_file=dot_data,__
       →feature_names=feature_names_param,
                          filled=True, rounded=True, special_characters=True)
          graph = pydotplus.graph_from_dot_data(dot_data.getvalue())
          return graph.create_png()
[31]: scores = cross_val_score(reg1, data_X, data_Y, cv=RepeatedKFold(n_splits=3,_u
       →n_repeats=3))
      print("%0.2f r^2 with a standard deviation of %0.2f" % (scores.mean(), scores.
       →std()))
     0.81 r<sup>2</sup> with a standard deviation of 0.03
     /home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/_bagging.py:1069:
     UserWarning: Some inputs do not have OOB scores. This probably means too few
     estimators were used to compute any reliable oob estimates.
       warn("Some inputs do not have OOB scores. "
```

/home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/\_bagging.py:1069: UserWarning: Some inputs do not have OOB scores. This probably means too few estimators were used to compute any reliable oob estimates.

warn("Some inputs do not have OOB scores. "

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warn("Some inputs do not have OOB scores. "

/home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/\_bagging.py:1069: UserWarning: Some inputs do not have OOB scores. This probably means too few estimators were used to compute any reliable oob estimates.

warn("Some inputs do not have OOB scores. "

2я модель

```
[32]: reg2 = RandomForestRegressor(n_estimators=6, oob_score=True, □ 

→random_state=10, max_depth=4)

reg2.fit(data_X_train, data_y_train)
```

 $/home/igor/.local/lib/python 3.9/site-packages/sklearn/ensemble/\_forest.py: 833: \\$ 

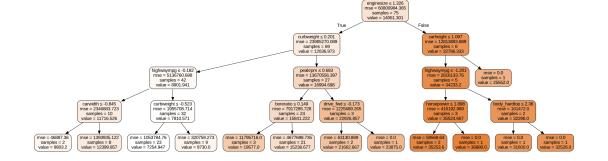
UserWarning: Some inputs do not have OOB scores. This probably means too few trees were used to compute any reliable oob estimates.

warn("Some inputs do not have OOB scores. "

[32]: RandomForestRegressor(max\_depth=4, n\_estimators=6, oob\_score=True, random\_state=10)

[33]: Image(get\_png\_tree(reg2.estimators\_[5], data\_X.columns), width="500")

[33]:



[34]: reg2.oob\_score\_ , 1-reg2.oob\_score\_

[34]: (0.4245449499224554, 0.5754550500775446)

[35]: scores = cross\_val\_score(reg2, data\_X, data\_Y, cv=RepeatedKFold(n\_splits=3,\_u

→n\_repeats=3))

print("%0.2f r^2 with a standard deviation of %0.2f" % (scores.mean(), scores.

→std()))

0.86 r^2 with a standard deviation of 0.06

/home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/\_forest.py:833:
UserWarning: Some inputs do not have OOB scores. This probably means too few trees were used to compute any reliable oob estimates.

warn("Some inputs do not have OOB scores. "

/home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/\_forest.py:833:
UserWarning: Some inputs do not have OOB scores. This probably means too few trees were used to compute any reliable oob estimates.

warn("Some inputs do not have OOB scores. "

/home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/\_forest.py:833:
UserWarning: Some inputs do not have OOB scores. This probably means too few trees were used to compute any reliable oob estimates.

warn("Some inputs do not have OOB scores. "

/home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/\_forest.py:833: UserWarning: Some inputs do not have OOB scores. This probably means too few trees were used to compute any reliable oob estimates.

warn("Some inputs do not have OOB scores. "

/home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/\_forest.py:833: UserWarning: Some inputs do not have OOB scores. This probably means too few

warn("Some inputs do not have OOB scores. "

trees were used to compute any reliable oob estimates.

/home/igor/.local/lib/python3.9/site-packages/sklearn/ensemble/\_forest.py:833: UserWarning: Some inputs do not have OOB scores. This probably means too few trees were used to compute any reliable oob estimates.

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