Ансамбли моделей машинного обучения

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In [70]:
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import numpy as np
import pandas as pd
from typing import Dict, Tuple
from scipy import stats
from scipy.optimize import fmin tnc
from IPython.display import Image
from sklearn.model selection import train test split
from sklearn.neighbors import KNeighborsRegressor, KNeighborsClassifier
from sklearn.metrics import accuracy_score, balanced_accuracy_score
from sklearn.metrics import precision score, recall score, f1 score, classification repor
from sklearn.metrics import confusion matrix
from sklearn.metrics import mean_absolute_error, mean_squared error, mean squared log err
or, median_absolute_error, r2_score, root_mean_squared_error
from sklearn.metrics import roc curve, roc auc score
from sklearn.linear_model import LinearRegression
from sklearn.linear_model import SGDRegressor
from sklearn.linear_model import SGDClassifier
from sklearn import linear_model
from sklearn.ensemble import BaggingRegressor
from sklearn.ensemble import RandomForestRegressor
from sklearn.ensemble import ExtraTreesRegressor
from sklearn.ensemble import AdaBoostRegressor
from sklearn.ensemble import GradientBoostingRegressor
import xgboost as xgb
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set(style="ticks")
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In [6]:
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mpg = pd.read_csv('C:\\MGTU\\6 semestr\\TMO\\auto-mpg.csv')
```

In [7]:

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mpg.head()
```

Out[7]:

	mpg	cylinders	displacement	horsepower	weight	acceleration	model year	origin	car name
0	18.0	8	307.0	130	3504	12.0	70	1	chevrolet chevelle malibu
1	15.0	8	350.0	165	3693	11.5	70	1	buick skylark 320
2	18.0	8	318.0	150	3436	11.0	70	1	plymouth satellite
3	16.0	8	304.0	150	3433	12.0	70	1	amc rebel sst
4	17.0	8	302.0	140	3449	10.5	70	1	ford torino

In [8]:

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mpg.dtypes
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Out[8]:

mpg	float64
cylinders	int64
displacement	float64
horsepower	object
weight	int64
acceleration	float64
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THEOA
шочет уеат
origin
                 int64
car name
               object
dtype: object
In [9]:
mpg = mpg[mpg['horsepower'] != '?']
mpg['horsepower'] = mpg['horsepower'].astype(float)
In [10]:
mpg.dtypes
Out[10]:
mpg
               float64
                 int64
cylinders
displacement float64
horsepower
              float64
weight
                 int64
             float64
acceleration
model year
                 int64
origin
                 int64
car name
                object
dtype: object
In [11]:
mpg = mpg.drop(columns=['car name'])
In [12]:
X = mpg.drop(columns=['mpg']) # Признаки
у = mpg['mpg'] # Целевая переменная
# Разделение данных на обучающую и тестовую выборки
mpg_X_train, mpg_X_test, mpg_y_train, mpg_y_test = train_test_split(X, y, test_size=0.2,
random state=1)
Bagging
In [53]:
br mpg = BaggingRegressor(n estimators=1000, oob score = False, random state=10)
br mpg.fit(mpg X train, mpg y train )
br mpg predict = br mpg.predict(mpg X test)
In [54]:
root mean squared error(mpg y test, br mpg predict)
Out[54]:
3.1432575417344863
Random Forest
In [55]:
rfr mpg = RandomForestRegressor(n estimators=1000, oob score = True, random state=10)
rfr_mpg.fit(mpg_X_train,mpg_y_train )
rfr mpg predict = rfr mpg.predict(mpg X test)
In [56]:
root mean squared error(mpg y test, rfr mpg predict)
Out[56]:
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3.151838599486819
In [89]:
rfr mpg1 = RandomForestRegressor(n estimators=1000, oob score = True, max features = 2,
min samples leaf = 5, random state=10)
rfr_mpg1.fit(mpg_X_train,mpg_y_train )
rfr mpg predict1 = rfr mpg1.predict(mpg X test)
In [90]:
root_mean_squared_error(mpg_y_test, rfr_mpg_predict1)
Out[90]:
3.55224352114633
Extremely Randomized Trees
In [59]:
etr mpg = ExtraTreesRegressor(n estimators=1000, random state=10)
etr mpg.fit(mpg X train, mpg y train )
etr_mpg_predict = etr_mpg.predict(mpg_X_test)
In [60]:
root mean squared error(mpg y test, etr mpg predict)
Out[60]:
3.312486072538161
AdaBoost
In [64]:
abr mpg = AdaBoostRegressor(n estimators=1000, random state=10)
abr mpg.fit(mpg X train, mpg y train )
abr_mpg_predict = abr_mpg.predict(mpg_X_test)
In [65]:
root mean squared_error(mpg_y_test, abr_mpg_predict)
Out[65]:
3.620154134360523
GradientBoosting
In [67]:
gbr_mpg = GradientBoostingRegressor(n_estimators=1000, random state=10)
gbr_mpg.fit(mpg_X_train,mpg_y_train )
gbr_mpg_predict = gbr_mpg.predict(mpg_X test)
In [68]:
root mean squared error(mpg y test, gbr mpg predict)
Out[68]:
3.1436494559859263
```

XGBoost

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In [77]:

dtrain = xgb.DMatrix(mpg_X_train, label=mpg_y_train)
dtest = xgb.DMatrix(mpg_X_test, label=mpg_y_test)

params = {
    'objective': 'reg:squarederror', # функция потерь для задачи регрессии
    'eval_metric': 'rmse'
}

num_rounds = 1000
model = xgb.train(params, dtrain, num_rounds)

y_pred = model.predict(dtest)

In [78]:
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root_mean_squared_error(mpg_y_test, y_pred)
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Out[78]:

3.2797096825132277