

# Рубежный контроль №2

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## Вариант 6

```
In [1]: import pandas as pd
import warnings
warnings.filterwarnings("ignore")
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np

from sklearn.metrics import mean_absolute_error, mean_squared_error, median
from sklearn.preprocessing import import MinMaxScaler
from sklearn.neighbors import import KNeighborsRegressor, KNeighborsClassifier
from sklearn.model_selection import import train_test_split, GridSearchCV
from sklearn.impute import import SimpleImputer, MissingIndicator
from sklearn.preprocessing import import LabelEncoder, OneHotEncoder, MinMaxScaler,
from sklearn.model_selection import import StratifiedKFold
from sklearn.model_selection import import cross_val_score
```

```
In [2]: df = pd.read_csv('Admission_Predict.csv')
```

```
In [3]: df.head()
```

```
Out[3]:
```

	Serial No.	GRE Score	TOEFL Score	University Rating	SOP	LOR	CGPA	Research	Chance of Admit
0	1	337	118	4	4.5	4.5	9.65	1	0.92
1	2	324	107	4	4.0	4.5	8.87	1	0.76
2	3	316	104	3	3.0	3.5	8.00	1	0.72
3	4	322	110	3	3.5	2.5	8.67	1	0.80
4	5	314	103	2	2.0	3.0	8.21	0	0.65

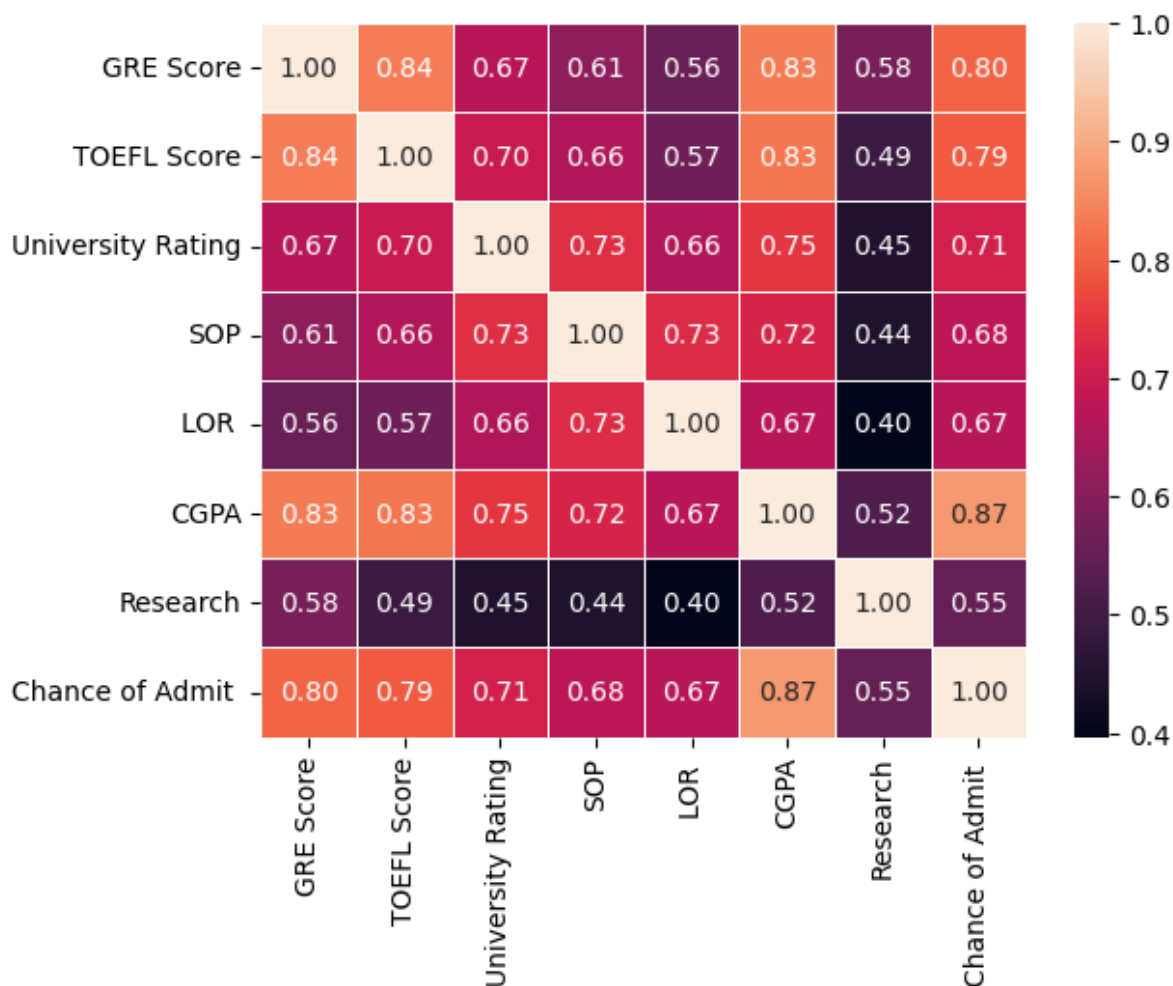
```
In [4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 9 columns):
#   Column                Non-Null Count  Dtype
---  -
0   Serial No.            400 non-null   int64
1   GRE Score              400 non-null   int64
2   TOEFL Score            400 non-null   int64
3   University Rating      400 non-null   int64
4   SOP                    400 non-null   float64
5   LOR                    400 non-null   float64
6   CGPA                   400 non-null   float64
7   Research                400 non-null   int64
8   Chance of Admit        400 non-null   float64
dtypes: float64(4), int64(5)
memory usage: 28.2 KB
```

```
In [5]: df = df.drop(['Serial No.'], axis=1)
df.isnull().sum()
```

```
Out[5]: GRE Score      0
TOEFL Score    0
University Rating 0
SOP            0
LOR            0
CGPA           0
Research       0
Chance of Admit 0
dtype: int64
```

```
In [6]: corr = df.corr()
sns.heatmap(corr, linewidths=.5, annot=True, fmt=".2f")
plt.show()
```



1) С целевым признаком "Chance of Admit" наиболее коррелируют признаки "CGPA", "GRE Score", "TOEFL Score". При построении модели машинного обучения перечисленные признаки будут наиболее информативными.

2) Стоит отметить корреляцию признаков "SOP" и "University Rating".

3) Можно построить модель машинного обучения на основе признаков "CGPA", "GRE Score", "TOEFL Score", "LOR", "Research". Первые 3 признака наиболее сильно повлияют на результат ввиду их высокой корреляции. Обученные модели позволят бакалаврам оценить свои возможности для поступления на магистратуру.

## Разделение данных

Разделим данные на целевой столбец и признаки. При построении предсказательных моделей исходные данные обычно разбиваются на обучающую ("training set") и контрольную ("test set") выборки. **Обучающая выборка** используется для построения математических отношений между некоторой переменной-откликом и предикторами, тогда как **контрольная (= "проверочная")** выборка служит для получения оценки прогнозных свойств модели на новых данных, т.е. данных, которые не были использованы для обучения модели.

```
In [7]: x = df.drop(['Chance of Admit '], axis=1) #Наименования признаков
        y = df['Chance of Admit '] # Значения признаков
```

```
In [8]: # кодируем категориальные данные из строк в числа
        le = LabelEncoder()
        y = le.fit_transform(y)
```

```
In [9]: x_train, x_test, y_train, y_test = train_test_split(x,y,test_size = 0.20, sh
```

```
In [10]: # Размер обучающей выборки
          x_train.shape, y_train.shape
```

```
Out[10]: ((320, 7), (320,))
```

```
In [11]: # Размер тестовой выборки
          x_test.shape, y_test.shape
```

```
Out[11]: ((80, 7), (80,))
```

## дерево решений.

Оцените качество моделей с помощью трех подходящих для задачи метрик. Сравните качество полученных моделей.

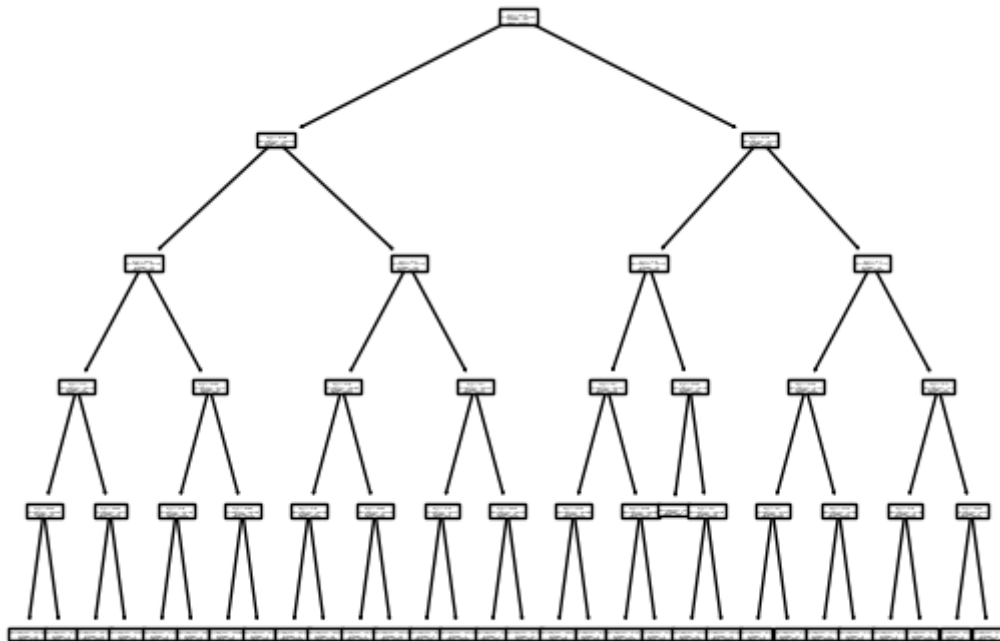
```
In [35]: from sklearn.linear_model import LinearRegression
          from sklearn.metrics import mean_squared_error, accuracy_score
          from sklearn.svm import SVC, NuSVC, LinearSVC
          from sklearn.tree import DecisionTreeClassifier, DecisionTreeRegressor, exp
          from sklearn.model_selection import GridSearchCV
          from sklearn import tree
          from IPython.core.display import HTML
          from sklearn.tree import export_text
          from operator import itemgetter
```

Дерево решений: Допустим, у нас есть задача классификации и мы хотим обучить дерево решений на данных. Одним из гиперпараметров этой модели является максимальная глубина дерева. Мы можем использовать GridSearchCV для подбора оптимальной глубины дерева с помощью кросс-валидации.

```
In [36]: dt_none = DecisionTreeRegressor(max_depth=5)
          dt_none.fit(x_train, y_train)
```

Out[36]: DecisionTreeRegressor(max\_depth=5)

In [37]: `tree.plot_tree(dt_none);`



In [15]: `clf = DecisionTreeClassifier(random_state=1)`  
`clf.fit(X_train, y_train)`

Out[15]: DecisionTreeClassifier(random\_state=1)

In [16]: `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)))`

In [17]: `test_model(dt_none)`

mean\_absolute\_error: 6.960165565428724  
 median\_absolute\_error: 5.19047619047619  
 r2\_score: 0.5696142563467339

In [18]: `tree_rules = export_text(dt_none, feature_names=list(X.columns))`  
`HTML('<pre>' + tree_rules + '</pre>')`

```

Out[18]: |--- CGPA <= 8.85
          |
          |--- CGPA <= 8.03
          |
          |--- GRE Score <= 307.50
          |
          |--- CGPA <= 7.67
          |
          |--- GRE Score <= 303.50
          |
          |--- GRE Score <= 298.50
          |
          |--- CGPA <= 7.64
          |
          |--- LOR <= 2.75
          |
          |--- LOR <= 1.75
          |
          |--- value: [11.00]
          |
          |--- LOR > 1.75
          |
          |--- Research <= 0.50
          |
          |--- value: [7.75]
          |
          |--- Research > 0.50
          |
          |--- value: [6.00]
          |
          |--- LOR > 2.75
          |
          |--- CGPA <= 7.37
          |
          |--- value: [9.00]
          |
          |--- CGPA > 7.37
          |
          |--- value: [15.00]
          |
          |--- CGPA > 7.64
          |
          |--- value: [19.00]
          |
          |--- GRE Score > 298.50
          |
          |--- TOEFL Score <= 98.00
          |
          |--- value: [2.00]
          |
          |--- TOEFL Score > 98.00
          |
          |--- value: [1.00]
          |
          |--- GRE Score > 303.50
          |
          |--- CGPA <= 7.58
          |
          |--- value: [14.00]
          |
          |--- CGPA > 7.58
          |
          |--- value: [20.00]
          |
          |--- CGPA > 7.67
          |
          |--- GRE Score <= 300.50
          |
          |--- SOP <= 3.50
          |
          |--- SOP <= 1.75
          |
          |--- GRE Score <= 298.50
          |
          |--- GRE Score <= 297.50
          |
          |--- value: [14.00]
          |
          |--- GRE Score > 297.50
          |
          |--- value: [16.00]
          |
          |--- GRE Score > 298.50
          |
          |--- University Rating <= 1.50
          |
          |--- SOP <= 1.25
          |
          |--- value: [20.00]
          |
          |--- SOP > 1.25
          |
          |--- value: [21.00]
          |
          |--- University Rating > 1.50
          |
          |--- CGPA <= 7.87
          |
          |--- value: [16.00]
          |
          |--- CGPA > 7.87
          |
          |--- value: [18.00]
          |
          |--- SOP > 1.75
          |
          |--- LOR <= 3.25
          |
          |--- LOR <= 2.75
          |
          |--- value: [23.00]
          |
          |--- LOR > 2.75

```

```
| | | | | | | | | |--- value: [22.00]
| | | | | | | | | |--- LOR > 3.25
| | | | | | | | | |--- GRE Score <= 296.00
| | | | | | | | | |--- value: [26.00]
| | | | | | | | | |--- GRE Score > 296.00
| | | | | | | | | |--- value: [30.00]
| | | | | | | | | |--- SOP > 3.50
| | | | | | | | | |--- value: [0.00]
| | | | | | | | | |--- GRE Score > 300.50
| | | | | | | | | |--- TOEFL Score <= 108.50
| | | | | | | | | |--- SOP <= 1.25
| | | | | | | | | |--- value: [18.00]
| | | | | | | | | |--- SOP > 1.25
| | | | | | | | | |--- SOP <= 1.75
| | | | | | | | | |--- value: [4.00]
| | | | | | | | | |--- SOP > 1.75
| | | | | | | | | |--- TOEFL Score <= 98.50
| | | | | | | | | |--- value: [6.00]
| | | | | | | | | |--- TOEFL Score > 98.50
| | | | | | | | | |--- University Rating <= 1.50
| | | | | | | | | |--- value: [12.00]
| | | | | | | | | |--- University Rating > 1.50
| | | | | | | | | |--- value: [10.00]
| | | | | | | | | |--- TOEFL Score > 108.50
| | | | | | | | | |--- value: [24.00]
| | | | | | | | | |--- GRE Score > 307.50
| | | | | | | | | |--- GRE Score <= 315.50
| | | | | | | | | |--- SOP <= 2.25
| | | | | | | | | |--- CGPA <= 7.98
| | | | | | | | | |--- GRE Score <= 311.00
| | | | | | | | | |--- value: [16.00]
| | | | | | | | | |--- GRE Score > 311.00
| | | | | | | | | |--- LOR <= 3.00
| | | | | | | | | |--- TOEFL Score <= 101.50
| | | | | | | | | |--- value: [20.00]
| | | | | | | | | |--- TOEFL Score > 101.50
| | | | | | | | | |--- value: [19.00]
| | | | | | | | | |--- LOR > 3.00
| | | | | | | | | |--- value: [18.00]
| | | | | | | | | |--- CGPA > 7.98
| | | | | | | | | |--- University Rating <= 2.00
| | | | | | | | | |--- value: [14.00]
| | | | | | | | | |--- University Rating > 2.00
| | | | | | | | | |--- value: [12.00]
| | | | | | | | | |--- SOP > 2.25
| | | | | | | | | |--- TOEFL Score <= 108.50
| | | | | | | | | |--- LOR <= 3.50
| | | | | | | | | |--- GRE Score <= 310.00
| | | | | | | | | |--- value: [24.00]
| | | | | | | | | |--- GRE Score > 310.00
| | | | | | | | | |--- CGPA <= 7.89
| | | | | | | | | |--- value: [25.00]
| | | | | | | | | |--- CGPA > 7.89
| | | | | | | | | |--- TOEFL Score <= 103.00
| | | | | | | | | |--- value: [27.00]
| | | | | | | | | |--- TOEFL Score > 103.00
| | | | | | | | | |--- value: [26.00]
```

```

|--- LOR > 3.50
|   |--- SOP <= 3.25
|       |--- value: [30.00]
|   |--- SOP > 3.25
|       |--- value: [28.00]
|   --- TOEFL Score > 108.50
|       |--- value: [8.00]
|--- GRE Score > 315.50
|   |--- University Rating <= 3.50
|       |--- CGPA <= 7.96
|           |--- value: [26.00]
|       |--- CGPA > 7.96
|           |--- SOP <= 3.50
|               |--- GRE Score <= 319.00
|                   |--- value: [34.00]
|               |--- GRE Score > 319.00
|                   |--- value: [32.00]
|           |--- SOP > 3.50
|               |--- value: [27.00]
|       |--- University Rating > 3.50
|           |--- value: [40.00]
|--- CGPA > 8.03
|   |--- CGPA <= 8.63
|       |--- LOR <= 3.25
|           |--- LOR <= 2.25
|               |--- SOP <= 2.25
|                   |--- value: [4.00]
|               |--- SOP > 2.25
|                   |--- CGPA <= 8.47
|                       |--- GRE Score <= 321.00
|                           |--- Research <= 0.50
|                               |--- CGPA <= 8.33
|                                   |--- value: [27.33]
|                                       |--- CGPA > 8.33
|                                           |--- value: [24.00]
|                                               |--- Research > 0.50
|                                                   |--- value: [23.00]
|                                                       |--- GRE Score > 321.00
|                                                           |--- value: [32.00]
|               |--- CGPA > 8.47
|                   |--- value: [15.00]
|           |--- LOR > 2.25
|               |--- CGPA <= 8.20
|                   |--- CGPA <= 8.19
|                       |--- University Rating <= 2.50
|                           |--- SOP <= 1.75
|                               |--- value: [33.00]
|                                   |--- SOP > 1.75
|                                       |--- GRE Score <= 308.50
|                                           |--- value: [25.75]
|                                               |--- GRE Score > 308.50
|                                                   |--- value: [20.67]
|               |--- University Rating > 2.50
|                   |--- CGPA <= 8.07
|                       |--- value: [29.00]
|                           |--- CGPA > 8.07
|                               |--- CGPA <= 8.14

```

localhost:8889/nbconvert/html/KonovalovMaximRK2.ipynb?download=false



```
| | | | |--- value: [18.00]
| | | | |--- GRE Score > 306.50
| | | | |--- University Rating <= 3.50
| | | | |--- TOEFL Score <= 99.50
| | | | |--- value: [38.00]
| | | | |--- TOEFL Score > 99.50
| | | | |--- TOEFL Score <= 108.50
| | | | |--- GRE Score <= 307.50
| | | | |--- value: [40.00]
| | | | |--- GRE Score > 307.50
| | | | |--- TOEFL Score <= 107.50
| | | | |--- value: [30.81]
| | | | |--- TOEFL Score > 107.50
| | | | |--- value: [24.50]
| | | | |--- TOEFL Score > 108.50
| | | | |--- University Rating <= 2.50
| | | | |--- TOEFL Score <= 109.50
| | | | |--- value: [33.00]
| | | | |--- TOEFL Score > 109.50
| | | | |--- value: [34.00]
| | | | |--- University Rating > 2.50
| | | | |--- value: [37.00]
| | | | |--- University Rating > 3.50
| | | | |--- CGPA <= 8.48
| | | | |--- CGPA <= 8.36
| | | | |--- value: [39.00]
| | | | |--- CGPA > 8.36
| | | | |--- SOP <= 3.50
| | | | |--- value: [40.00]
| | | | |--- SOP > 3.50
| | | | |--- value: [41.00]
| | | | |--- CGPA > 8.48
| | | | |--- value: [33.00]
|--- CGPA > 8.63
| |--- University Rating <= 2.50
| | |--- LOR <= 2.75
| | |--- GRE Score <= 322.50
| | |--- value: [43.00]
| | |--- GRE Score > 322.50
| | |--- value: [37.00]
| |--- LOR > 2.75
| | |--- Research <= 0.50
| | |--- LOR <= 3.50
| | |--- value: [28.00]
| | |--- LOR > 3.50
| | |--- value: [29.00]
| | |--- Research > 0.50
| | |--- value: [19.00]
|--- University Rating > 2.50
| |--- SOP <= 2.25
| |--- value: [26.00]
| |--- SOP > 2.25
| | |--- Research <= 0.50
| | |--- GRE Score <= 324.50
| | |--- GRE Score <= 318.50
| | |--- GRE Score <= 316.50
| | |--- CGPA <= 8.64
```

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```
| | | |--- value: [17.00]
| | | |--- Research > 0.50
| | | |--- TOEFL Score <= 109.50
| | | |   |--- value: [30.00]
| | | |--- TOEFL Score > 109.50
| | | |   |--- CGPA <= 9.08
| | | |     |--- CGPA <= 9.03
| | | |       |--- CGPA <= 8.97
| | | |         |--- CGPA <= 8.86
| | | |           |--- value: [44.00]
| | | |             |--- CGPA > 8.86
| | | |               |--- CGPA <= 8.93
| | | |                 |--- value: [42.00]
| | | |                   |--- CGPA > 8.93
| | | |                     |--- value: [41.00]
| | | |                       |--- CGPA > 8.97
| | | |                         |--- CGPA <= 8.99
| | | |                           |--- value: [46.00]
| | | |                             |--- CGPA > 8.99
| | | |                               |--- value: [44.00]
| | | |                                 |--- CGPA > 9.03
| | | |                                   |--- CGPA <= 9.05
| | | |                                     |--- value: [38.00]
| | | |                                       |--- CGPA > 9.05
| | | |                                         |--- value: [36.00]
| | | |                                           |--- CGPA > 9.08
| | | |                                             |--- SOP <= 3.25
| | | |                                               |--- value: [45.00]
| | | |                                                 |--- SOP > 3.25
| | | |                                                   |--- value: [46.00]
| | | |----- SOP > 3.75
| | | |   |--- GRE Score <= 310.50
| | | |     |--- value: [28.00]
| | | |   |--- GRE Score > 310.50
| | | |     |--- University Rating <= 4.50
| | | |       |--- TOEFL Score <= 116.50
| | | |         |--- CGPA <= 9.11
| | | |           |--- GRE Score <= 322.00
| | | |             |--- GRE Score <= 313.50
| | | |               |--- LOR <= 4.00
| | | |                 |--- value: [41.00]
| | | |                   |--- LOR > 4.00
| | | |                     |--- value: [40.00]
| | | |                       |--- GRE Score > 313.50
| | | |                         |--- TOEFL Score <= 109.50
| | | |                           |--- value: [45.00]
| | | |                             |--- TOEFL Score > 109.50
| | | |                               |--- value: [48.00]
| | | |                                 |--- GRE Score > 322.00
| | | |                                   |--- CGPA <= 8.88
| | | |                                     |--- value: [38.00]
| | | |                                       |--- CGPA > 8.88
| | | |                                         |--- CGPA <= 9.04
| | | |                                           |--- value: [43.00]
| | | |                                             |--- CGPA > 9.04
| | | |                                               |--- value: [41.00]
| | | |----- CGPA > 9.11
```

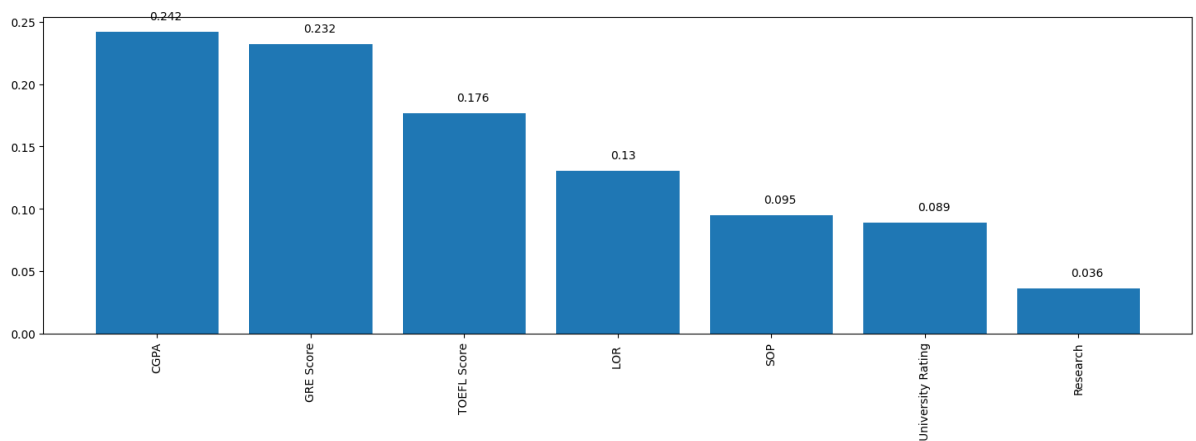
----- LOR > 4.75

```
| | | | | | |--- value: [48.00]
| | | | |---- CGPA > 9.31
| | | | |    |-- University Rating <= 4.50
| | | | |      |-- value: [55.00]
| | | | |    |-- University Rating > 4.50
| | | | |      |-- LOR <= 4.75
| | | | |        |-- LOR <= 4.25
| | | | |          |-- value: [53.00]
| | | | |            |-- LOR > 4.25
| | | | |              |-- value: [51.00]
| | | | |                |-- LOR > 4.75
| | | | |                  |-- value: [55.00]
|-- TOEFL Score > 114.50
|   |-- TOEFL Score <= 117.50
|     |-- LOR <= 3.25
|       |-- value: [52.00]
|     |-- LOR > 3.25
|       |-- SOP <= 4.25
|         |-- SOP <= 3.75
|           |-- value: [53.00]
|             |-- SOP > 3.75
|               |-- TOEFL Score <= 115.50
|                 |-- value: [54.00]
|                   |-- TOEFL Score > 115.50
|                     |-- value: [55.00]
|                       |-- SOP > 4.25
|                         |-- CGPA <= 9.40
|                           |-- CGPA <= 9.32
|                             |-- value: [56.00]
|                               |-- CGPA > 9.32
|                                 |-- value: [55.00]
|                                   |-- CGPA > 9.40
|                                     |-- value: [56.00]
|                                       |-- TOEFL Score > 117.50
|                                         |-- University Rating <= 4.50
|                                           |-- CGPA <= 9.38
|                                             |-- value: [52.00]
|                                               |-- CGPA > 9.38
|                                                 |-- value: [53.00]
|                                                   |-- University Rating > 4.50
|                                                     |-- SOP <= 4.75
|                                                       |-- value: [55.00]
|                                                         |-- SOP > 4.75
|                                                           |-- value: [54.00]
|-- CGPA > 9.47
|   |-- CGPA <= 9.73
|     |-- LOR <= 4.25
|       |-- SOP <= 4.25
|         |-- value: [52.00]
|           |-- SOP > 4.25
|             |-- value: [51.00]
|               |-- LOR > 4.25
|                 |-- TOEFL Score <= 118.50
|                   |-- SOP <= 4.75
|                     |-- CGPA <= 9.59
|                       |-- value: [56.00]
|                         |-- CGPA > 9.59
```

```
| | | | | | | | |--- CGPA <= 9.66  
| | | | | | | | |--- value: [54.00]  
| | | | | | | | |--- CGPA > 9.66  
| | | | | | | | |--- value: [55.00]  
| | | | | | | | |--- SOP > 4.75  
| | | | | | | | |--- value: [56.00]  
| | | | | | | | |--- TOEFL Score > 118.50  
| | | | | | | | |--- SOP <= 4.75  
| | | | | | | | |--- value: [56.00]  
| | | | | | | | |--- SOP > 4.75  
| | | | | | | | |--- CGPA <= 9.58  
| | | | | | | | |--- value: [58.00]  
| | | | | | | | |--- CGPA > 9.58  
| | | | | | | | |--- value: [57.00]  
| | | | | | | | |--- CGPA > 9.73  
| | | | | | | | |--- GRE Score <= 332.00  
| | | | | | | | |--- value: [56.00]  
| | | | | | | | |--- GRE Score > 332.00  
| | | | | | | | |--- TOEFL Score <= 117.50  
| | | | | | | | |--- value: [58.00]  
| | | | | | | | |--- TOEFL Score > 117.50  
| | | | | | | | |--- GRE Score <= 333.50  
| | | | | | | | |--- value: [58.00]  
| | | | | | | | |--- GRE Score > 333.50  
| | | | | | | | |--- value: [59.00]
```

```
In [19]: def draw_feature_importances(tree_model, X_dataset, figsize=(18,5)):
        """
        Вывод важности признаков в виде графика
        """
        # Сортировка значений важности признаков по убыванию
        list_to_sort = list(zip(X_dataset.columns.values, tree_model.feature_imp
sorted_list = sorted(list_to_sort, key=itemgetter(1), reverse = True)
        # Названия признаков
        labels = [x for x,_ in sorted_list]
        # Важности признаков
        data = [x for _,x in sorted_list]
        # Вывод графика
        fig, ax = plt.subplots(figsize=figsize)
        ind = np.arange(len(labels))
        plt.bar(ind, data)
        plt.xticks(ind, labels, rotation='vertical')
        # Вывод значений
        for a,b in zip(ind, data):
            plt.text(a-0.05, b+0.01, str(round(b,3)))
        plt.show()
        return labels, data
```

```
In [20]: dt fl, dt fd = draw feature importances(clf, X train)
```



Произведите для каждой модели подбор одного гиперпараметра с использованием GridSearchCV и кросс-валидации.

```
In [21]: tree = DecisionTreeClassifier()

param_grid = {'max_depth': [2, 4, 6, 8, 10],
              'min_samples_split': [2, 4, 6, 8, 10],
              'min_samples_leaf': [1, 2, 3, 4, 5]}

grid_search = GridSearchCV(tree, param_grid=param_grid, cv=5)

grid_search.fit(X_train, y_train)

accuracy_tree = grid_search.best_estimator_.score(X_test, y_test)

print("Наилучшие параметры: {} ".format(grid_search.best_params_))
print("Оценка точности на кросс-валидации: {:.2f}".format(grid_search.best_score_))
print(accuracy_tree)
```

Наилучшие параметры: {} {'max\_depth': 6, 'min\_samples\_leaf': 3, 'min\_samples\_split': 4}

Оценка точности на кросс-валидации: 0.11

0.0875

Сравните качество полученных моделей с качеством моделей

```
In [22]: models = [['DecisionTree : ', DecisionTreeRegressor()]]
```

```
In [23]: print('Вывод 1')
for name, model in models:
    model = model
    model.fit(X_train, y_train)
    predictions = model.predict(X_test)
    print(name, (np.sqrt(mean_squared_error(y_test, predictions))))
```

Вывод 1

DecisionTree : 9.419394885023134

```
In [24]: models = [['DecisionTree : ', DecisionTreeRegressor(max_depth = 6, min_samples
    ['Linear Regression : ', LinearRegression(normalize = True)],
    ['SVC : ', SVC(C = 1, kernel = 'linear')]]
```

```
In [25]: print('Вывод 2')
for name, model in models:
```

```

model = model
model.fit(X_train, y_train)
predictions = model.predict(X_test)
print(name, (np.sqrt(mean_squared_error(y_test, predictions))))

```

Вывод 2

```

DecisionTree : 7.718895862285397
Linear Regression : 6.252859728579746
SVC : 7.565546906866681

```

**Бэггинг.** В этом случае однородные модели обучают на разных наборах данных и объединяют. Получают прогноз путём усреднения. Если использовать в качестве слабого ученика деревья решений, то получится случайный лес RandomForestClassifier / RandomForestRegressor.

```

In [26]: from sklearn.ensemble import RandomForestClassifier, GradientBoostingClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import cross_val_predict
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import GradientBoostingClassifier

```

**Модель бэггинга - случайный лес (Random Forest):**

```

In [27]: # Создаем модель случайного леса с 100 деревьями
rf_model = RandomForestClassifier(n_estimators=100)

# Обучаем модель на тренировочных данных
rf_model.fit(X_train, y_train)

# Оцениваем качество модели на тестовых данных
accuracy = rf_model.score(X_test, y_test)
print('Accuracy: {:.2f}%'.format(accuracy*100))

```

Accuracy: 11.25%

Произведите для каждой модели подбор значений одного гиперпараметра. В зависимости от используемой библиотеки можно применять функцию GridSearchCV, использовать перебор параметров в цикле, или использовать другие методы.

```

In [28]: model = RandomForestClassifier()

param_grid = {
    'n_estimators': [200, 700],
    'max_features': ['auto', 'sqrt', 'log2']
}

grid_search = GridSearchCV(model, param_grid=param_grid, cv=5)

grid_search.fit(X_train, y_train)

accuracy_RandomForestClassifier = grid_search.best_estimator_.score(X_test, y_test)

print("Наилучшие параметры: {} ".format(grid_search.best_params_))
print("Оценка точности на кросс-валидации: {:.2f}%".format(grid_search.best_score_))
print(accuracy_tree)

```



Наилучшие параметры: {} {'max\_features': 'sqrt', 'n\_estimators': 200}  
Оценка точности на кросс-валидации: 0.12  
0.0875

```
In [29]: models = [  
        ['RandomForestClassifier ', RandomForestClassifier()]  
        ]
```

```
In [30]: for name,model in models:  
        model = model  
        model.fit(X_train, y_train)  
        predictions = model.predict(X_test)  
        print(name, (np.sqrt(mean_squared_error(y_test, predictions))))
```

RandomForestClassifier : 7.440598094239467

```
In [31]: models = [['RandomForestClassifier ', RandomForestClassifier(max_features =  
        ]
```

```
In [32]: for name,model in models:  
        model = model  
        model.fit(X_train, y_train)  
        predictions = model.predict(X_test)  
        print(name, (np.sqrt(mean_squared_error(y_test, predictions))))
```

RandomForestClassifier : 7.402702209328699