

# Рубежный контроль №2

Коновалов Максим ИУ5-63Б

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

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

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

	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

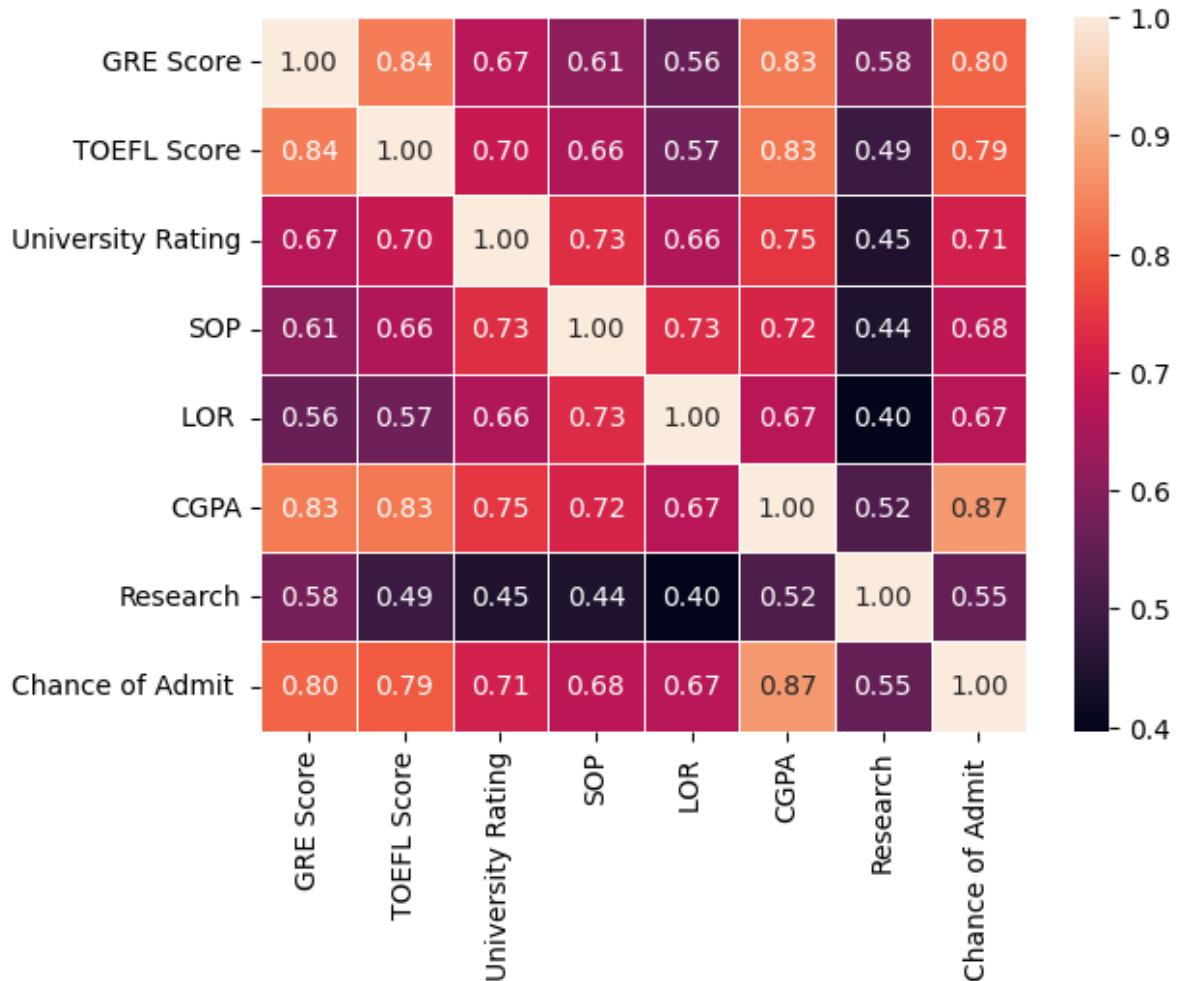
```
In [4]: df.info()
```

```
<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 влияют 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.

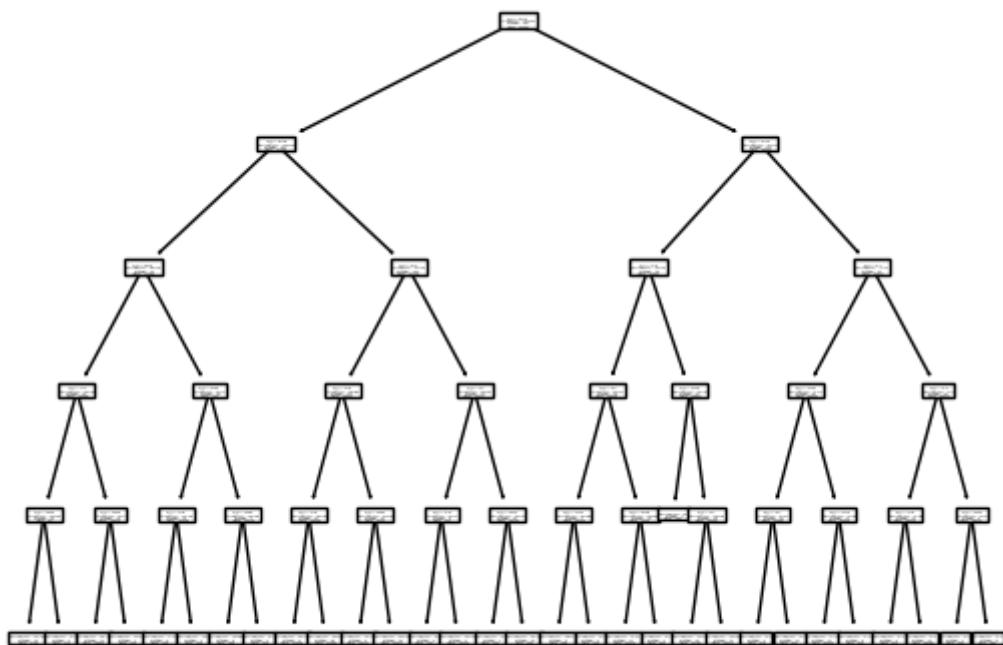
```
In [35]: from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, accuracy_score
from sklearn.svm import SVC, NuSVC, LinearSVC
from sklearn.tree import DecisionTreeClassifier, DecisionTreeRegressor, export_graphviz
from sklearn.model_selection import GridSearchCV
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
```

```

|   |   |   |   |   |   |   |--- value: [34.00]
|   |   |   |   |   |--- CGPA >  8.14
|   |   |   |   |--- value: [33.00]
|--- CGPA >  8.19
|--- GRE Score <= 305.50
|   |--- value: [16.00]
|--- GRE Score >  305.50
|   |--- SOP <= 3.25
|   |   |--- value: [11.00]
|--- SOP >  3.25
|   |   |--- value: [9.00]
|--- CGPA >  8.20
|--- GRE Score <= 322.50
|--- University Rating <= 3.50
|   |--- SOP <= 3.75
|   |   |--- Research <= 0.50
|   |   |   |--- value: [29.84]
|   |   |--- Research >  0.50
|   |   |   |--- value: [33.00]
|--- SOP >  3.75
|   |   |--- value: [18.00]
|--- University Rating >  3.50
|--- CGPA <= 8.54
|   |--- CGPA <= 8.47
|   |   |--- value: [19.00]
|   |--- CGPA >  8.47
|   |   |--- value: [23.00]
|--- CGPA >  8.54
|   |   |--- value: [32.00]
|--- GRE Score >  322.50
|   |--- value: [7.00]
|--- LOR >  3.25
|--- GRE Score <= 306.50
|--- CGPA <= 8.58
|--- TOEFL Score <= 105.50
|--- GRE Score <= 301.50
|   |--- CGPA <= 8.33
|   |   |--- value: [30.00]
|   |--- CGPA >  8.33
|   |   |--- value: [31.00]
|--- GRE Score >  301.50
|   |--- LOR <= 3.75
|   |   |--- value: [26.00]
|   |--- LOR >  3.75
|   |   |--- LOR <= 4.50
|   |   |   |--- value: [28.00]
|   |   |--- LOR >  4.50
|   |   |   |--- value: [27.00]
|--- TOEFL Score >  105.50
|--- Research <= 0.50
|--- GRE Score <= 302.50
|   |--- value: [26.00]
|--- GRE Score >  302.50
|   |--- value: [25.00]
|--- Research >  0.50
|   |--- value: [24.00]
|--- CGPA >  8.58

```

```
| | | | |--- 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
```



```
| | | | | --- 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
```



```

|   |   |   |   |--- 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

```

```

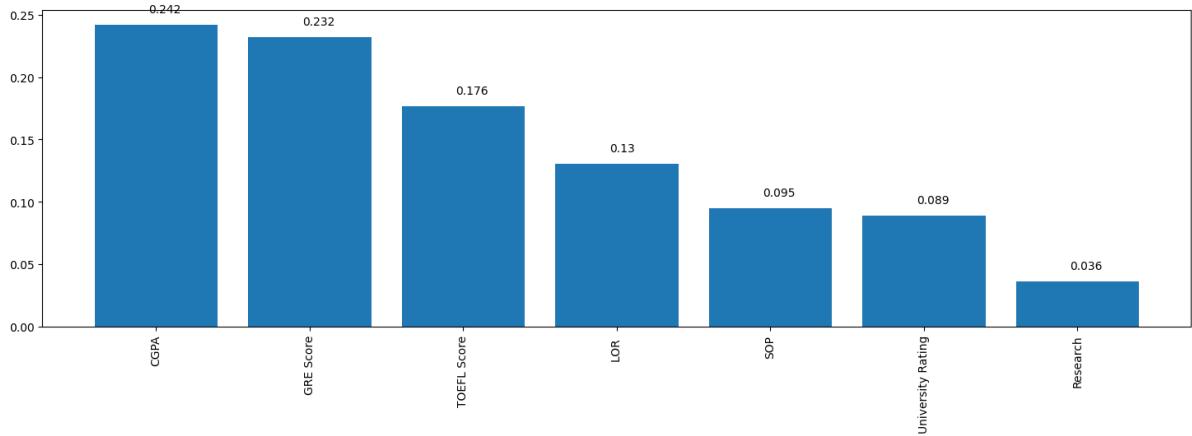
|   |   |   |   |   |   | --- CGPA <= 9.66
|   |   |   |   |   |   |   | --- value: [54.00]
|   |   |   |   |   |   | --- CGPA >  9.66
|   |   |   |   |   |   |   | --- value: [55.00]
|   |   |   |   |   | --- SOP >  4.75
|   |   |   |   |   |   | --- value: [56.00]
|   |   |   |   | --- TOEFL Score > 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_importances_))
    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_leaf = 3),
                ['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 обучаются 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: {}, grid_search.best_params_")
print("Оценка точности на кросс-валидации: {:.2f}%".format(grid_search.best_score_*100))
print(accuracy_RandomForestClassifier)

```

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