

Импорт библиотек

```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from pandas.plotting import scatter_matrix
import warnings
warnings.filterwarnings('ignore')
sns.set(style="ticks")
%matplotlib inline
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.svm import SVC, LinearSVC
from sklearn.datasets.samples_generator import make_blobs
from sklearn.svm import SVR
from sklearn.model_selection import GridSearchCV
from matplotlib import pyplot as plt
from sklearn.metrics import mean_absolute_error, mean_squared_error, median_absolute_error, r2_score
```

```
In [2]: data = pd.read_csv('toy.csv', sep = ';')
data.head()
```

```
Out[2]:
```

	Number	City	Gender	Age	Income	Illness	Unnamed: 6
0	1	Dallas	Male	41	40367.0	No	NaN
1	2	Dallas	Male	54	45084.0	No	NaN
2	3	Dallas	Male	42	52483.0	No	NaN
3	4	Dallas	Male	40	40941.0	No	NaN
4	5	Dallas	Male	46	50289.0	No	NaN

```
In [3]: data.dtypes
```

```
Out[3]: Number      int64
City      object
Gender     object
Age      int64
Income    float64
Illness    object
Unnamed: 6  float64
dtype: object
```

```
In [4]: data['Gender'].value_counts()
```

```
Out[4]: Male      145
Female    119
Name: Gender, dtype: int64
```

```
In [5]: data['Illness'].value_counts()
```

```
Out[5]: No      240
Yes      24
Name: Illness, dtype: int64
```

```
In [6]: data['IsGender']=data.Gender.replace({'Female':0,'Male':1})
data.drop('Gender', axis = 1, inplace = True)
data['IsIllness']=data.Illness.replace({'No':0,'Yes':1})
data.drop('Illness', axis = 1, inplace = True)
```

```
In [7]: data.isnull().sum()
# проверим есть ли пропущенные значения
```

```
Out[7]: Number      0
City      0
Age      0
Income    0
Unnamed: 6    264
IsGender    0
IsIllness    0
dtype: int64
```

```
In [8]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 264 entries, 0 to 263
Data columns (total 7 columns):
#   Column      Non-Null Count  Dtype
---  -
0   Number      264 non-null    int64
1   City        264 non-null    object
2   Age         264 non-null    int64
3   Income      264 non-null    float64
4   Unnamed: 6   0 non-null      float64
5   IsGender     264 non-null    int64
6   IsIllness    264 non-null    int64
dtypes: float64(2), int64(4), object(1)
memory usage: 14.6+ KB
```

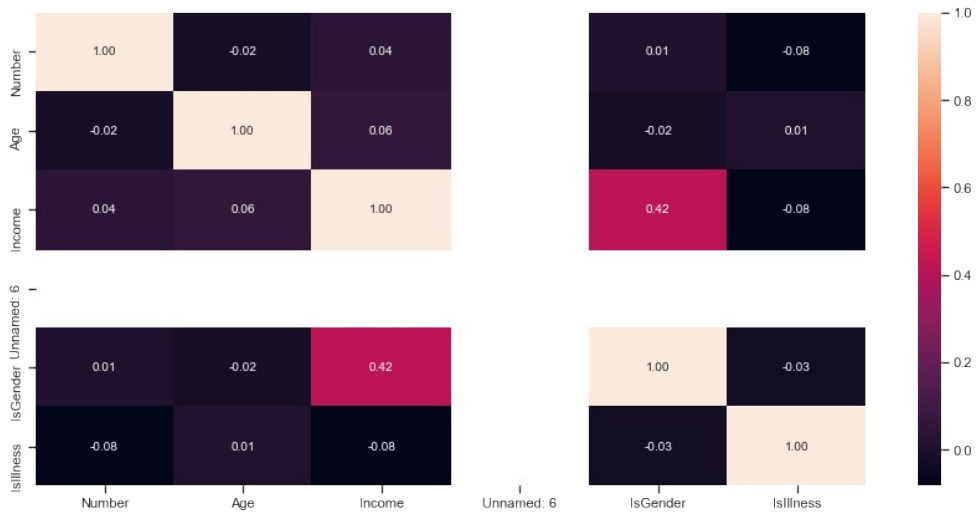
```
In [9]: data.head()
```

Out[9]:

	Number	City	Age	Income	Unnamed: 6	IsGender	IsIllness
0	1	Dallas	41	40367.0	NaN	1	0
1	2	Dallas	54	45084.0	NaN	1	0
2	3	Dallas	42	52483.0	NaN	1	0
3	4	Dallas	40	40941.0	NaN	1	0
4	5	Dallas	46	50289.0	NaN	1	0

```
In [10]: #Построим корреляционную матрицу
fig, ax = plt.subplots(figsize=(15,7))
sns.heatmap(data.corr(method='pearson'), ax=ax, annot=True, fmt='.2f')
```

Out[10]: <AxesSubplot:>



```
In [11]: X = data[["Number", "IsGender"]]  
         Y = data.Income  
         print('Входные данные:\n\n', X.head(), '\n\nВыходные данные:\n\n', Y.head())
```

Входные данные:

	Number	IsGender
0	1	1
1	2	1
2	3	1
3	4	1
4	5	1

Выходные данные:

```
0    40367.0
1    45084.0
2    52483.0
3    40941.0
4    50289.0
Name: Income, dtype: float64
```

```
In [12]: X_train, X_test, Y_train, Y_test = train_test_split(X, Y, random_state = 0, test_size = 0.1)
print('Входные параметры обучающей выборки:\n\n', X_train.head(), \
      '\n\nВходные параметры тестовой выборки:\n\n', X_test.head(), \
      '\n\nВыходные параметры обучающей выборки:\n\n', Y_train.head(), \
      '\n\nВыходные параметры тестовой выборки:\n\n', Y_test.head())
```

Входные параметры обучающей выборки:

	Number	IsGender
45	46	0
241	242	1
74	75	0
201	202	1
258	259	1

Входные параметры тестовой выборки:

	Number	IsGender
136	137	1
101	102	1
240	241	1
8	9	1
181	182	1

Выходные параметры обучающей выборки:

45	40661.0
241	61320.0
74	27897.0
201	32404.0
258	51490.0

Name: Income, dtype: float64

Выходные параметры тестовой выборки:

136	43573.0
101	48433.0
240	23579.0
8	68667.0
181	48899.0

Name: Income, dtype: float64

```
In [13]: from sklearn.ensemble import RandomForestRegressor
```

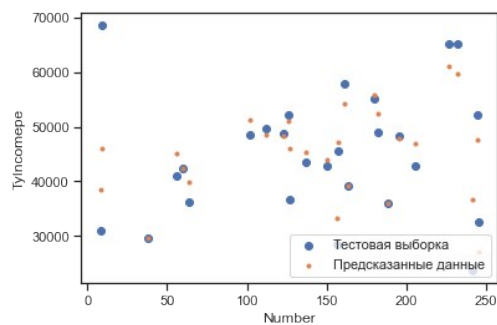
```
In [14]: forest_1 = RandomForestRegressor(n_estimators=5, oob_score=True, random_state=10)
forest_1.fit(X, Y)
```

```
Out[14]: RandomForestRegressor(n_estimators=5, oob_score=True, random_state=10)
```

```
In [15]: Y_predict = forest_1.predict(X_test)
print('Средняя абсолютная ошибка:', mean_absolute_error(Y_test, Y_predict))
print('Средняя квадратичная ошибка:', mean_squared_error(Y_test, Y_predict))
print('Median absolute error:', median_absolute_error(Y_test, Y_predict))
print('Коэффициент детерминации:', r2_score(Y_test, Y_predict))
```

Средняя абсолютная ошибка: 4009.0148148148146
Средняя квадратичная ошибка: 38785407.00888889
Median absolute error: 3479.1999999999997
Коэффициент детерминации: 0.6951589334459994

```
In [16]: plt.scatter(X_test.Number, Y_test, marker='o', label='Тестовая выборка')
plt.scatter(X_test.Number, Y_predict, marker='.', label='Предсказанные данные')
plt.legend(loc='lower right')
plt.xlabel('Number')
plt.ylabel('TyIncomepe')
plt.show()
```



```
In [17]: from sklearn.svm import SVC, LinearSVC
from sklearn.pipeline import make_pipeline
from matplotlib import pyplot as plt
from sklearn.preprocessing import StandardScaler
```

```
In [18]: svc = clf = make_pipeline(StandardScaler(), SVC(gamma='auto'))
svc.fit(X_train, Y_train)
```

```
Out[18]: Pipeline(steps=[('standardscaler', StandardScaler()),
                          ('svc', SVC(gamma='auto'))])
```

```
In [19]: pred_y = svc.predict(X_test)
```

```
In [20]: plt.scatter(X_test.Number, Y_test, marker='s', label='Тестовая выборка')
plt.scatter(X_test.Number, pred_y, marker='.', label='Предсказанные данные')
plt.legend(loc='lower right')
plt.xlabel('Number')
plt.ylabel('Income')
plt.show()
```

