

## Зенгер ИУ5-24М Вариант 5

<https://www.kaggle.com/iabhishekofficial/mobile-price-classification>  
(<https://www.kaggle.com/iabhishekofficial/mobile-price-classification>).

### Задача 1 - №5

Для набора данных проведите кодирование одного (произвольного) категориального признака с использованием метода "one-hot encoding".

```
In [1]: import pandas as pd
        from sklearn.preprocessing import OneHotEncoder
```

```
In [2]: data = pd.read_csv('train.csv')
```

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

```
Out[3]:
```

	battery_power	blue	clock_speed	dual_sim	fc	four_g	int_memory	m_dep	mobile_wt	n_cc
0	842	0	2.2	0	1	0	7	0.6	188	
1	1021	1	0.5	1	0	1	53	0.7	136	
2	563	1	0.5	1	2	1	41	0.9	145	
3	615	1	2.5	0	0	0	10	0.8	131	
4	1821	1	1.2	0	13	1	44	0.6	141	

5 rows × 21 columns

```
In [5]: # This is the target variable with value of 0(low cost), 1(medium cost),
        2(high cost) and 3(very high cost).
        data['price_range'].unique()
```

```
Out[5]: array([1, 2, 3, 0])
```

### sklearn

```
In [7]: ohe = OneHotEncoder()
        price_ohe = ohe.fit_transform(data[['price_range']])
```

```
In [18]: data['price_range'].head(10)
```

```
Out[18]: 0    1
          1    2
          2    2
          3    2
          4    1
          5    1
          6    3
          7    0
          8    0
          9    0
          Name: price_range, dtype: int64
```

```
In [16]: price_ohe.todense()[0:10]
```

```
Out[16]: matrix([[0., 1., 0., 0.],
                  [0., 0., 1., 0.],
                  [0., 0., 1., 0.],
                  [0., 0., 1., 0.],
                  [0., 1., 0., 0.],
                  [0., 1., 0., 0.],
                  [0., 0., 0., 1.],
                  [1., 0., 0., 0.],
                  [1., 0., 0., 0.],
                  [1., 0., 0., 0.]])
```

## pandas

```
In [20]: pd.get_dummies(data['price_range']).head(10)
```

```
Out[20]:
```

	0	1	2	3
0	0	1	0	0
1	0	0	1	0
2	0	0	1	0
3	0	0	1	0
4	0	1	0	0
5	0	1	0	0
6	0	0	0	1
7	1	0	0	0
8	1	0	0	0
9	1	0	0	0

## Задача 2 - №25

Для набора данных для одного (произвольного) числового признака проведите обнаружение и удаление выбросов на основе межквартильного размаха.

<https://www.kaggle.com/rashikrahmanpritom/heart-attack-analysis-prediction-dataset>

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```
In [29]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import scipy.stats as stats
```

```
In [67]: data = pd.read_csv('heart.csv')
```

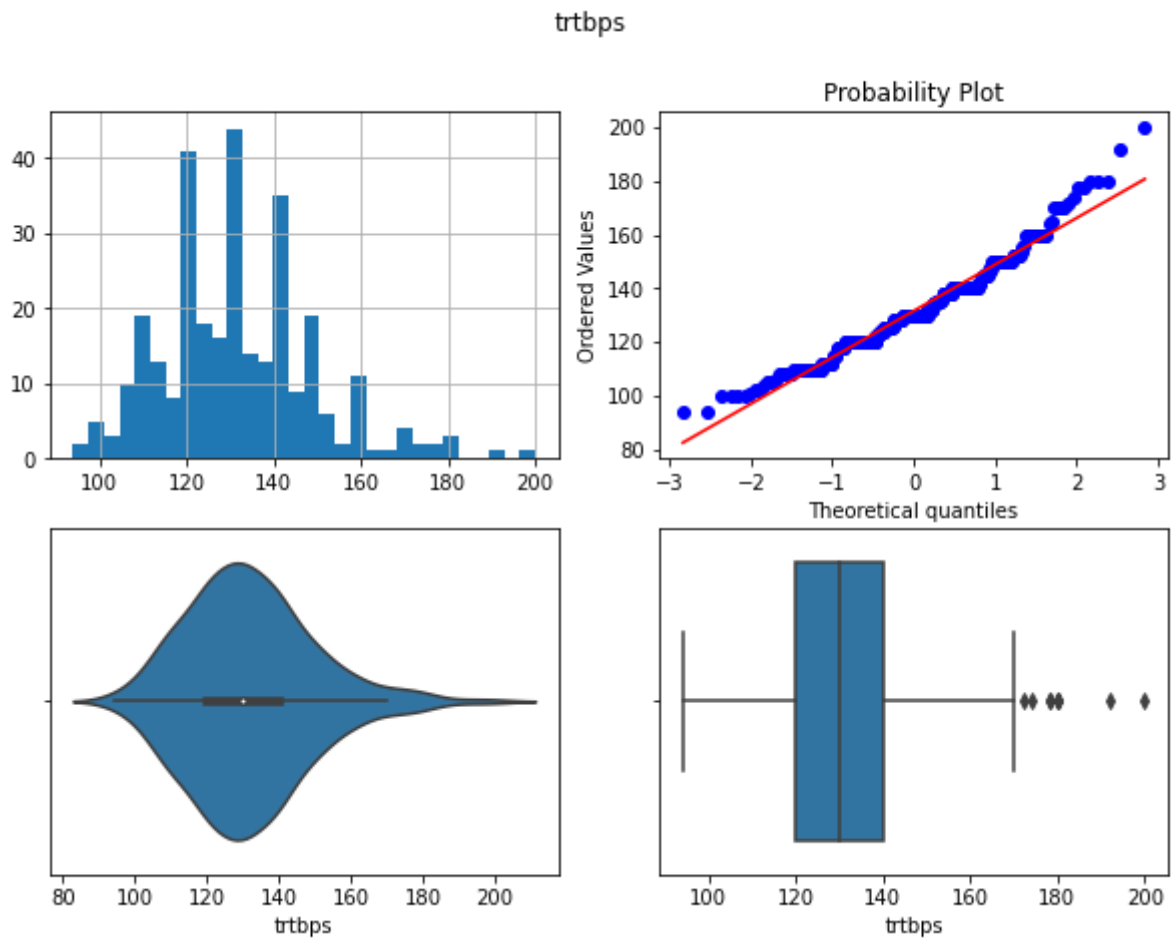
```
In [68]: data.describe()
```

Out[68]:

	age	sex	cp	trtbps	chol	fbs	restecg	th
<b>count</b>	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.000000	303.
<b>mean</b>	54.366337	0.683168	0.966997	131.623762	246.264026	0.148515	0.528053	149.
<b>std</b>	9.082101	0.466011	1.032052	17.538143	51.830751	0.356198	0.525860	22.
<b>min</b>	29.000000	0.000000	0.000000	94.000000	126.000000	0.000000	0.000000	71.
<b>25%</b>	47.500000	0.000000	0.000000	120.000000	211.000000	0.000000	0.000000	133.
<b>50%</b>	55.000000	1.000000	1.000000	130.000000	240.000000	0.000000	1.000000	153.
<b>75%</b>	61.000000	1.000000	2.000000	140.000000	274.500000	0.000000	1.000000	166.
<b>max</b>	77.000000	1.000000	3.000000	200.000000	564.000000	1.000000	2.000000	202.

```
In [69]: def diagnostic_plots(df, variable, title):
fig, ax = plt.subplots(figsize=(10,7))
# гистограмма
plt.subplot(2, 2, 1)
df[variable].hist(bins=30)
## Q-Q plot
plt.subplot(2, 2, 2)
stats.probplot(df[variable], dist="norm", plot=plt)
# ящик с усами
plt.subplot(2, 2, 3)
sns.violinplot(x=df[variable])
# ящик с усами
plt.subplot(2, 2, 4)
sns.boxplot(x=df[variable])
fig.suptitle(title)
plt.show()
```

```
In [70]: diagnostic_plots(data, 'trtbps', 'trtbps')
```



```
In [71]: def get_outlier_boundaries(df, col):
    K = 1.5
    quant_75 = df[col].quantile(0.75)
    quant_25 = df[col].quantile(0.25)
    IQR = quant_75 - quant_25
    print('quantile 0.75: ', quant_75)
    print('quantile 0.25: ', quant_25)
    print('IQR: ', IQR)
    lower_boundary = quant_25 - (K * IQR)
    upper_boundary = quant_75 + (K * IQR)
    return lower_boundary, upper_boundary
```

```
In [72]: get_outlier_boundaries(data, 'trtbps')
```

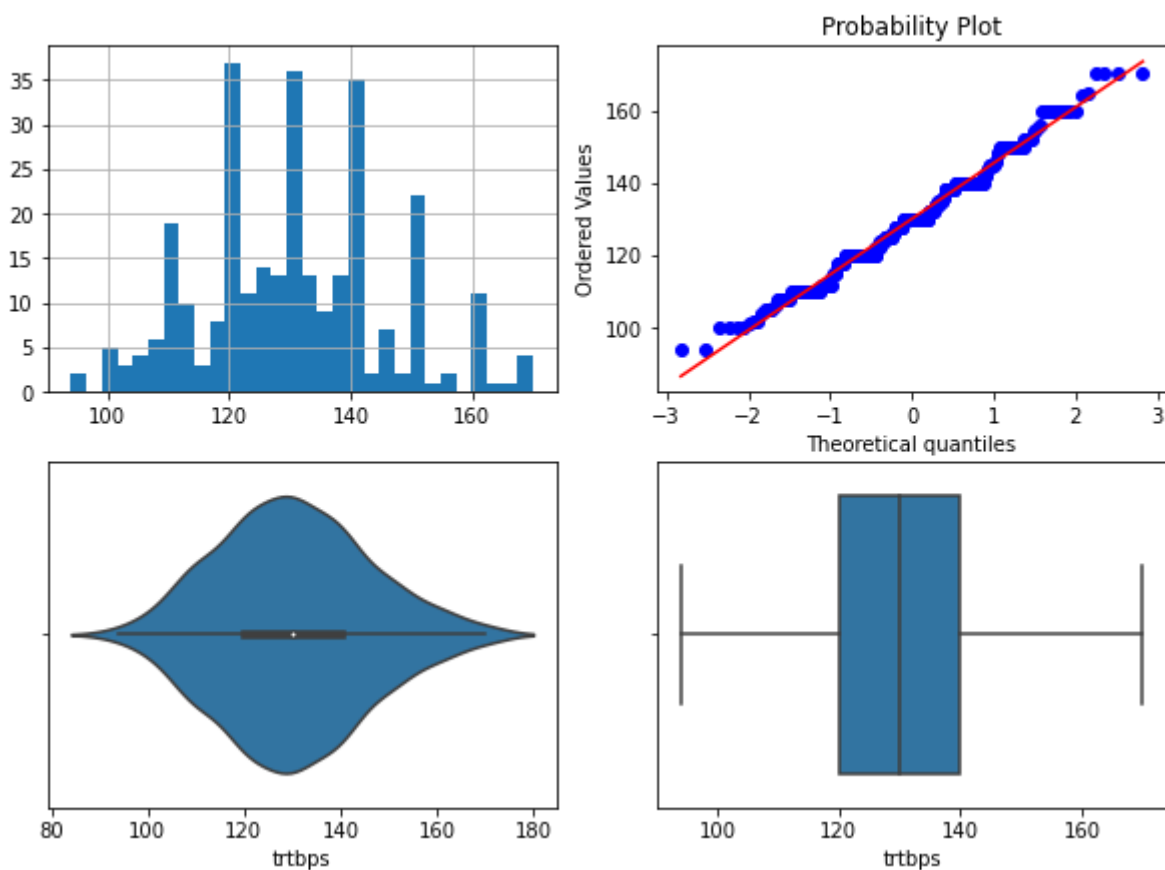
```
quantile 0.75: 140.0
quantile 0.25: 120.0
IQR: 20.0
```

```
Out[72]: (90.0, 170.0)
```

```
In [74]: # Удаление выбросов
lower_boundary, upper_boundary = get_outlier_boundaries(data, 'trtbps')
# Флаги для удаления выбросов
outliers_temp = np.where(data['trtbps'] > upper_boundary, True,
                           np.where(data['trtbps'] < lower_boundary, True,
                                     False))
# Удаление данных на основе флага
data_trimmed = data.loc[~(outliers_temp), ]
title = 'Поле-{}, строка-{}'.format('trtbps', data_trimmed.shape[0])
diagnostic_plots(data_trimmed, 'trtbps', title)

quantile 0.75: 140.0
quantile 0.25: 120.0
IQR: 20.0
```

Поле-trtbps, строка-294



In [ ]:

In [ ]: