

## Ткачева Диана ИУ5-22М РК1

### Вариант №15

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import scipy.stats as stats
import seaborn as sns
from sklearn.tree import DecisionTreeClassifier
from sklearn.datasets import load_iris
```

### Задача 15

```
data = pd.read_csv('online_store_customer_data.csv', sep=",")

data.Amount_spent.unique()

array([2051.36,  544.04, 1572.6 , ..., 2030.07, 1909.77, 1073.15])
```

```
data.head()
```

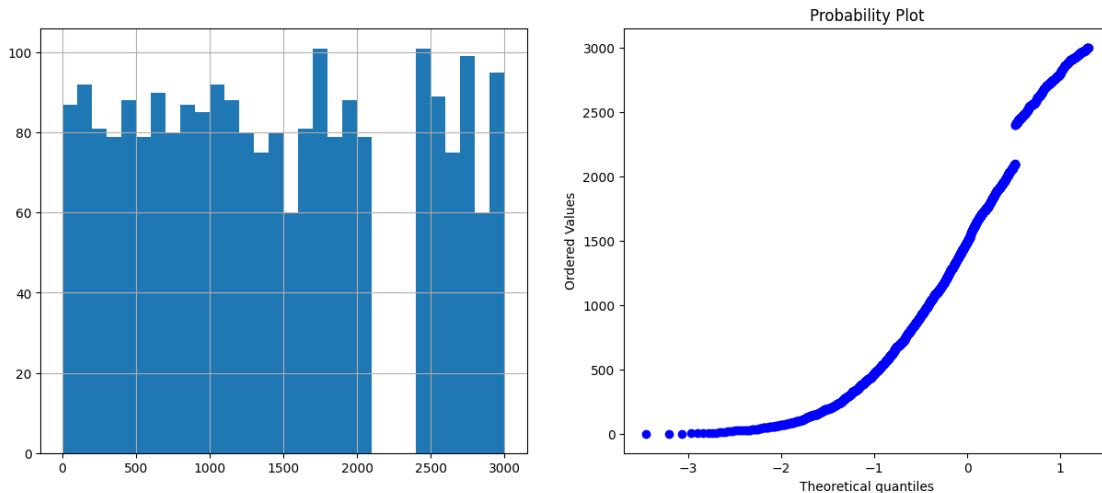
|               | Transaction_date | Transaction_ID | Gender | Age  | Marital_status |     |
|---------------|------------------|----------------|--------|------|----------------|-----|
| State_names \ |                  |                |        |      |                |     |
| 0             | 1/1/2019         | 151200         | Female | 19.0 | Single         |     |
| Kansas        |                  |                |        |      |                |     |
| 1             | 1/1/2019         | 151201         | Male   | 49.0 | Single         |     |
| Illinois      |                  |                |        |      |                |     |
| 2             | 1/1/2019         | 151202         | Male   | 63.0 | Married        | New |
| Mexico        |                  |                |        |      |                |     |
| 3             | 1/1/2019         | 151203         | NaN    | 18.0 | Single         |     |
| Virginia      |                  |                |        |      |                |     |
| 4             | 1/1/2019         | 151204         | Male   | 27.0 | Single         |     |
| Connecticut   |                  |                |        |      |                |     |

|         | Segment  | Employees_status | Payment_method | Referral | Amount_spent |
|---------|----------|------------------|----------------|----------|--------------|
| Age_num |          |                  |                |          |              |
| 0       | Basic    | Unemployment     | Other          | 1.0      | 2051.36      |
| 19.0    |          |                  |                |          |              |
| 1       | Basic    | self-employed    | Card           | 0.0      | 544.04       |
| 49.0    |          |                  |                |          |              |
| 2       | Basic    | workers          | PayPal         | 1.0      | 1572.60      |
| 63.0    |          |                  |                |          |              |
| 3       | Platinum | workers          | Card           | 1.0      | 1199.79      |
| 18.0    |          |                  |                |          |              |
| 4       | Basic    | self-employed    | Card           | 0.0      | NaN          |
| 27.0    |          |                  |                |          |              |

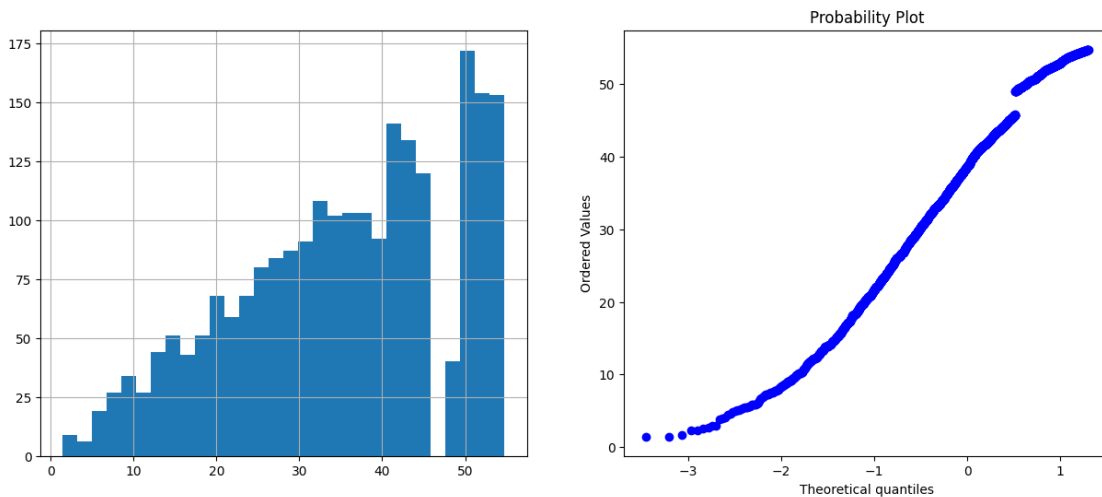
```
def diagnostic_plots(df, variable):
    plt.figure(figsize=(15,6))
    # гистограмма
    plt.subplot(1, 2, 1)
```

```
df[variable].hist(bins=30)
## Q-Q plot
plt.subplot(1, 2, 2)
stats.probplot(df[variable], dist="norm", plot=plt)
plt.show()
```

```
diagnostic_plots(data, "Amount_spent")
```



```
data['Amount_spent_num'] = data['Amount_spent']**(1/2)
diagnostic_plots(data, 'Amount_spent_num')
```



### Задача 35

```
iris = load_iris()
dataX = iris.data
dataY = iris.target

d1 = pd.DataFrame(data=iris['data'], columns=iris['feature_names'])
d2 = pd.DataFrame(data=iris['target'], columns=['class']).apply(lambda
x: iris['target_names'][x])
df = pd.concat([d1,d2],axis=1)
```

```
df.head()

   sepal length (cm)  sepal width (cm)  petal length (cm)  petal width
0      5.1           3.5           1.4
0.2
1      4.9           3.0           1.4
0.2
2      4.7           3.2           1.3
0.2
3      4.6           3.1           1.5
0.2
4      5.0           3.6           1.4
0.2

   class
0  setosa
1  setosa
2  setosa
3  setosa
4  setosa
```

```
dtc1 = DecisionTreeClassifier()
dtc1.fit(dataX, dataY)
```

*# Важность признаков*

```
dtc1.feature_importances_, sum(dtc1.feature_importances_)
(array([0.01333333, 0.          , 0.06405596, 0.92261071]), 1.0)
```

```
from operator import itemgetter
```

```
def draw_feature_importances(tree_model, X_dataset, title,
figsize=(7,4)):
```

```
    """
```

*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*

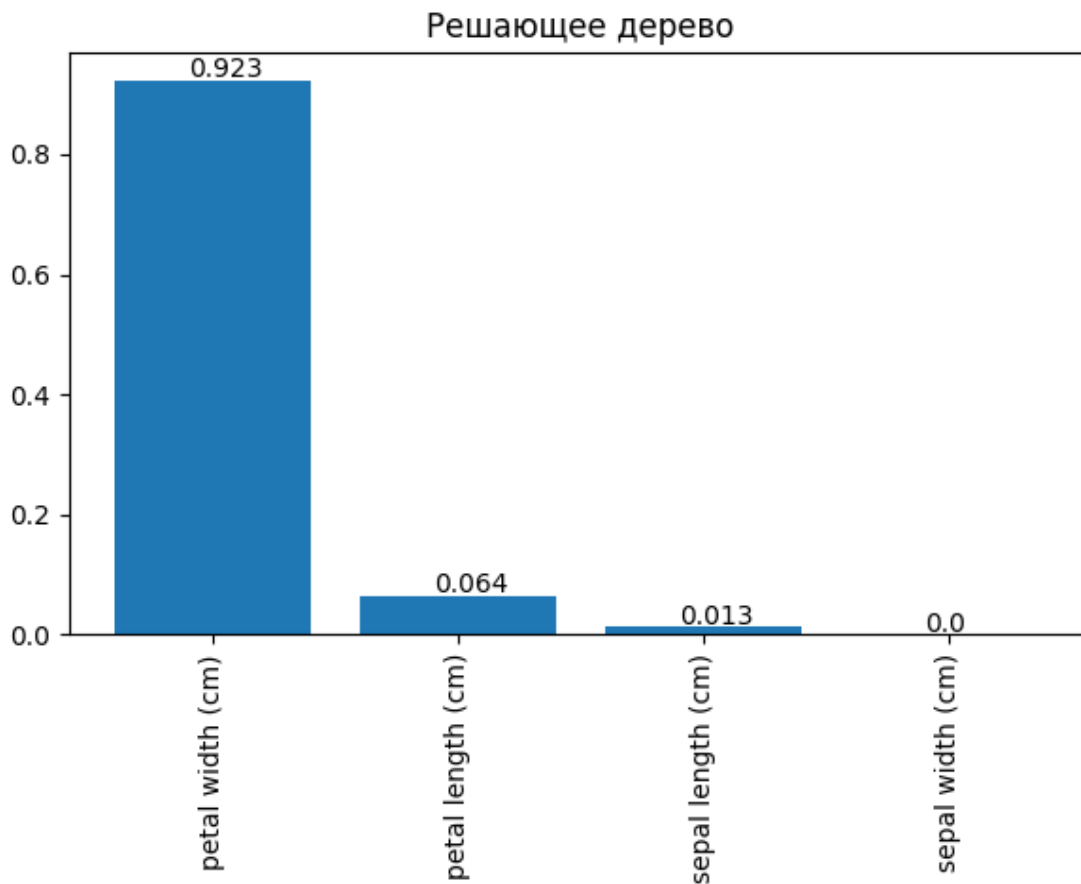
```
    fig, ax = plt.subplots(figsize=figsize)
```

```
    ax.set_title(title)
```

```
    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-0.1, b+0.005, str(round(b,3)))
plt.show()
return labels, data
```

```
_,_=draw_feature_importances(dtc1, d1, 'Решающее дерево')
```



#### Дополнительное задание

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2512 entries, 0 to 2511
Data columns (total 11 columns):
#   Column              Non-Null Count  Dtype
---  -
0   Transaction_date     2512 non-null   object
1   Transaction_ID       2512 non-null   int64
2   Gender               2484 non-null   object
3   Age                  2470 non-null   float64
```

```
4   Marital_status      2512 non-null   object
5   State_names         2512 non-null   object
6   Segment             2512 non-null   object
7   Employees_status    2486 non-null   object
8   Payment_method      2512 non-null   object
9   Referral            2357 non-null   float64
10  Amount_spent         2270 non-null   float64
dtypes: float64(3), int64(1), object(7)
memory usage: 216.0+ KB
```

```
fig, ax = plt.subplots(figsize=(10, 10))
sns.histplot(data, x='Age', binwidth=5, ax=ax)
plt.xticks(range(10, 80, 5), rotation=90)
plt.xlabel('Возраст')
plt.ylabel('Количество')
plt.show()
```

