

# Модели стохастических объектов (методы анализа данных)

## Практическая работа №2

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Вариант 14.

Признаки cat6\_brandOfSoaps , cat2\_hairColor , num1\_15 , num2\_15 , num3\_15 .

In [296...]

```
from collections import Counter
from scipy.stats import sigmaclip
from sklearn import preprocessing
from sklearn.impute import KNNImputer
from sklearn.preprocessing import OneHotEncoder, MinMaxScaler
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import seaborn as sns
```

In [297...]

```
all_data = pd.read_excel('practice_dataframe.xlsx')

data = pd.DataFrame({
    'cat6_brandOfSoaps': all_data['cat6_brandOfSoaps'],
    'cat2_hairColor': all_data['cat2_hairColor'],
    'num1_15': all_data['num1_15'],
    'num2_15': all_data['num2_15'],
    'num3_15': all_data['num3_15']
})
data
```

Out[297...]

	cat6_brandOfSoaps	cat2_hairColor	num1_15	num2_15	num3_15
0	lux	black	282.228024	NaN	807.976365
1	Nan	Red	1039.089956	268.388947	444.529442
2	Lux	black	500.696085	174.992127	428.723966
3	lux	Red	1346.429995	108.647900	520.238037
4	Dove	black	230.576445	150.113065	896.476167
...	...	...	...	...	...
995	Dove	Red	422.732662	170.810692	586.254084
996	lux	Brown	783.565442	237.086397	826.693312
997	Dove	Red	410.595523	196.686858	600.377791
998	Lux	black	-237.136035	179.262913	371.380055
999	Lux	black	22.006500	230.004737	326.04169

1000 rows × 5 columns

In [298...]

```
data.info()
```

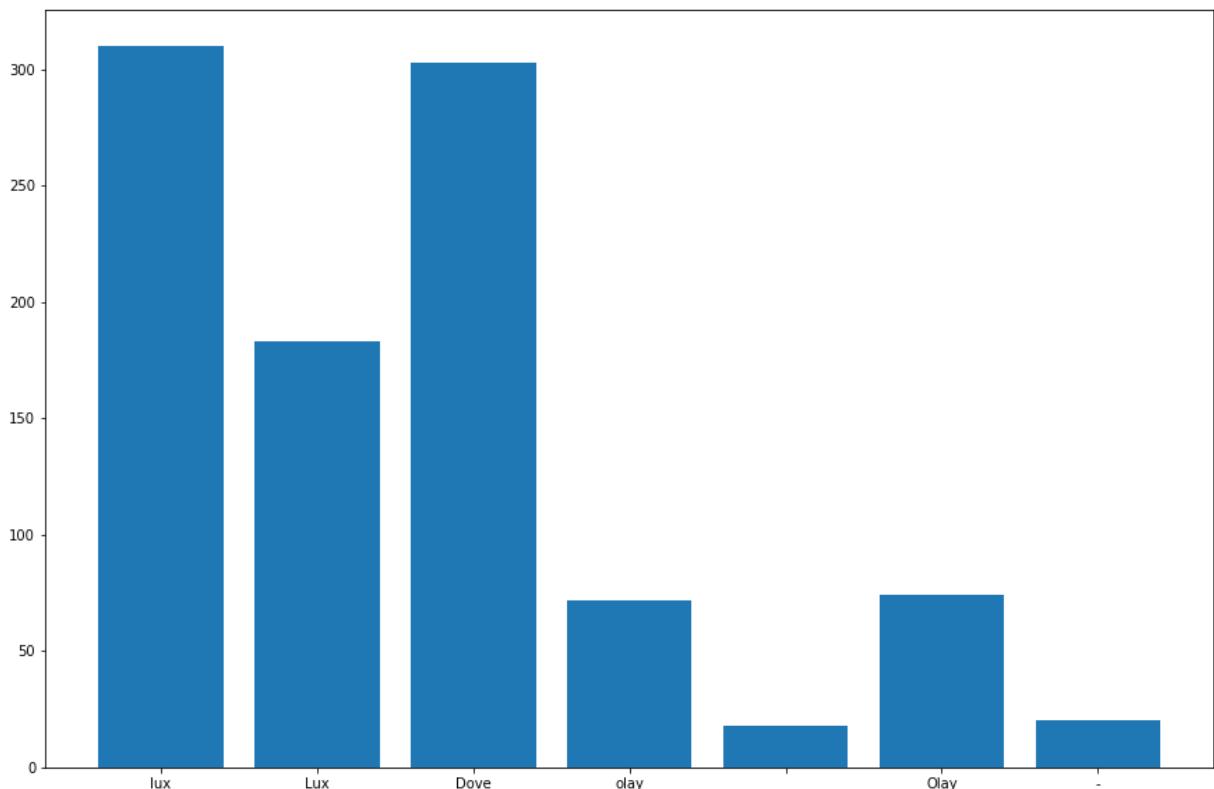
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 5 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   cat6_brandOfSoaps    980 non-null   object  
 1   cat2_hairColor        980 non-null   object  
 2   num1_15              1000 non-null   float64 
 3   num2_15              950 non-null   float64 
 4   num3_15              971 non-null   object  
dtypes: float64(2), object(3)
memory usage: 39.2+ KB
```

## 1. Визуальный анализ данных

Столбчатые для категориальных до форматирования

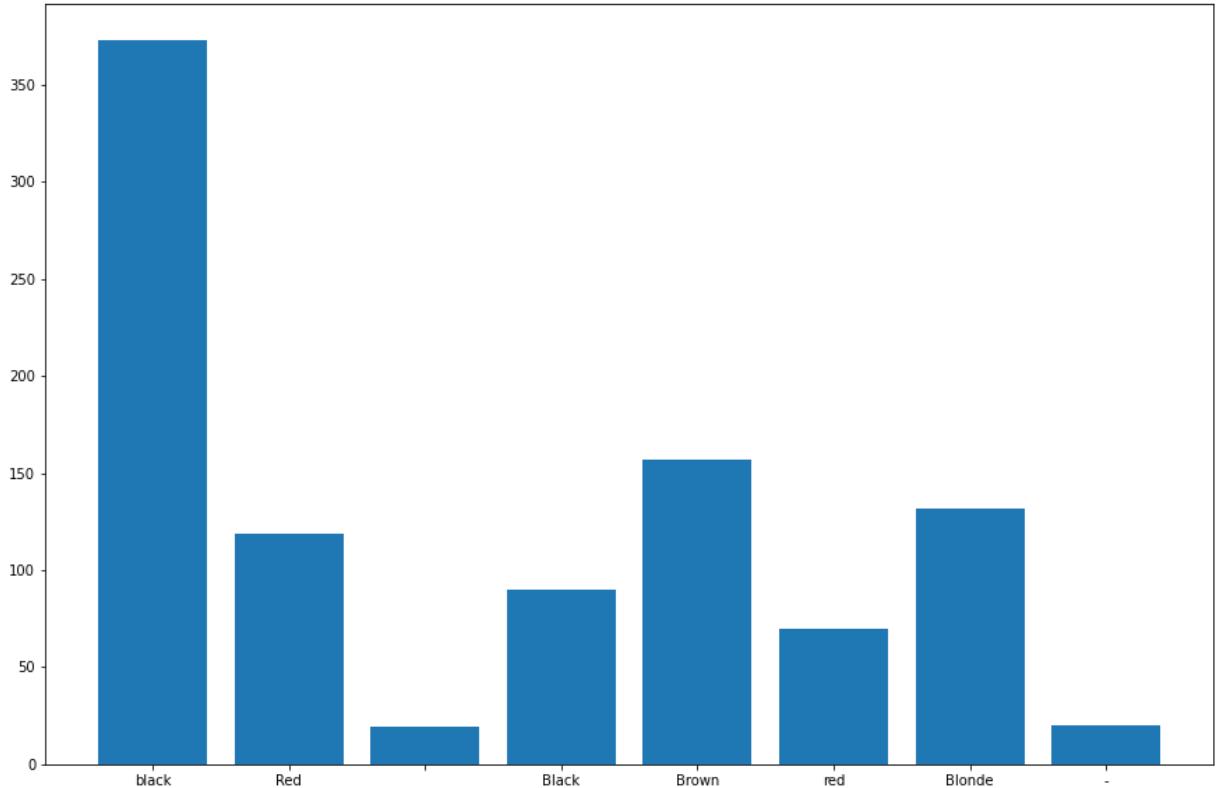
In [299...]

```
# cat6_brandOfSoaps
counts = Counter(data['cat6_brandOfSoaps'].dropna())
plt.figure(figsize=(15, 10))
plt.bar(counts.keys(), counts.values())
plt.show()
```



In [300...]

```
# cat2_hairColor
counts = Counter(data['cat2_hairColor'].dropna())
plt.figure(figsize=(15, 10))
plt.bar(counts.keys(), counts.values())
plt.show()
```



In [301...]

```
def base_format(data_frame: pd.DataFrame):
    return data_frame.replace({-9999: np.nan}).replace(r'^[\s-]*$', np.nan, r'')
def format_str(data_frame: pd.DataFrame):
    return base_format(data_frame).str.capitalize().str.rstrip()

data['cat6_brandOfSoaps'] = format_str(data['cat6_brandOfSoaps'])
data['cat2_hairColor'] = format_str(data['cat2_hairColor'])
data["num1_15"] = base_format(data['num1_15'])
data["num2_15"] = base_format(data['num2_15'])
data["num3_15"] = base_format(data['num3_15'])

data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 5 columns):
 #   Column           Non-Null Count  Dtype  
 ---  -- 
 0   cat6_brandOfSoaps 942 non-null    object 
 1   cat2_hairColor     941 non-null    object 
 2   num1_15            951 non-null    float64
 3   num2_15            950 non-null    float64
 4   num3_15            913 non-null    float64
dtypes: float64(3), object(2)
memory usage: 39.2+ KB
```

In [302...]

```
data.describe()
```

Out[302...]

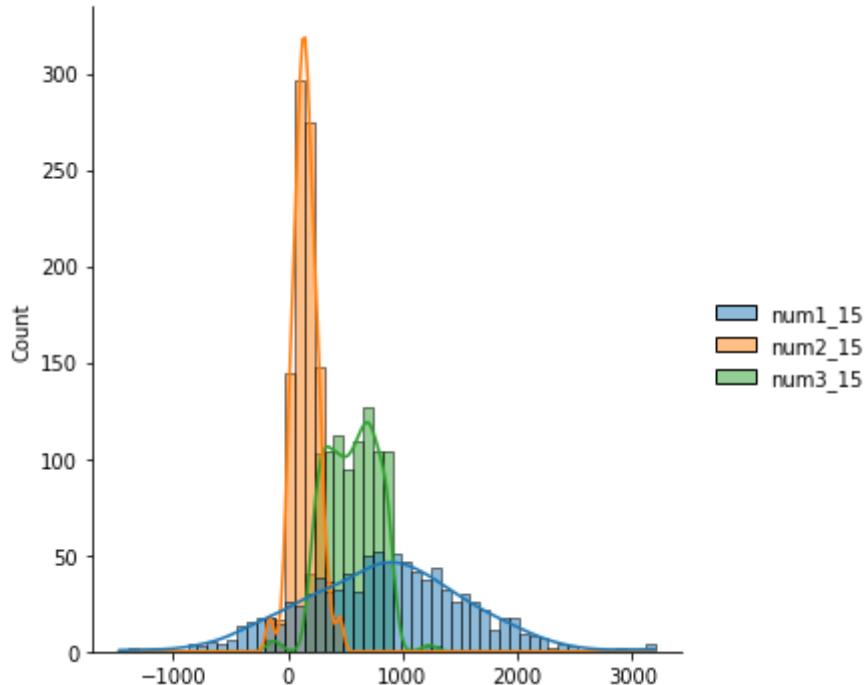
	num1_15	num2_15	num3_15
count	951.000000	950.000000	913.000000
mean	843.822989	148.963833	555.429557
std	716.182943	106.847552	221.358521
min	-1467.940219	-184.922189	-168.839273

	num1_15	num2_15	num3_15
25%	377.718319	82.347973	373.272578
50%	859.934249	145.220453	562.668399
75%	1301.342806	215.636532	731.795739
max	3201.191252	479.162302	1265.142875

In [303...]

```
sns.displot(data=data, kde=True)
```

Out[303...]

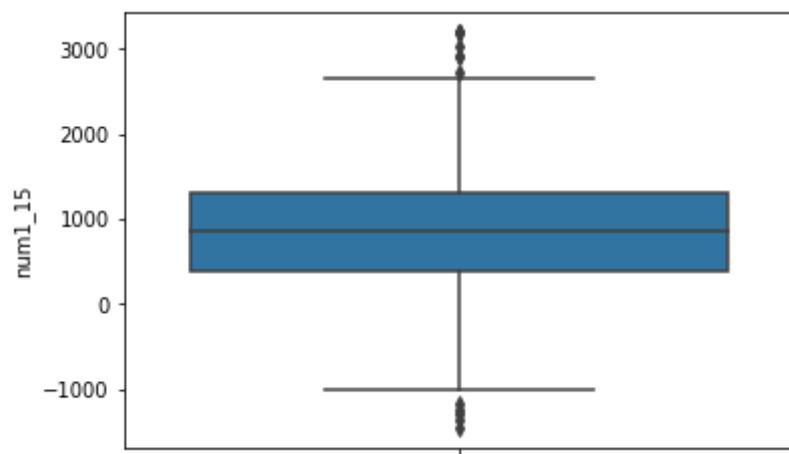


Ящик с усами для числовых

In [304...]

```
# num1_15
sns.boxplot(y=data['num1_15'])
```

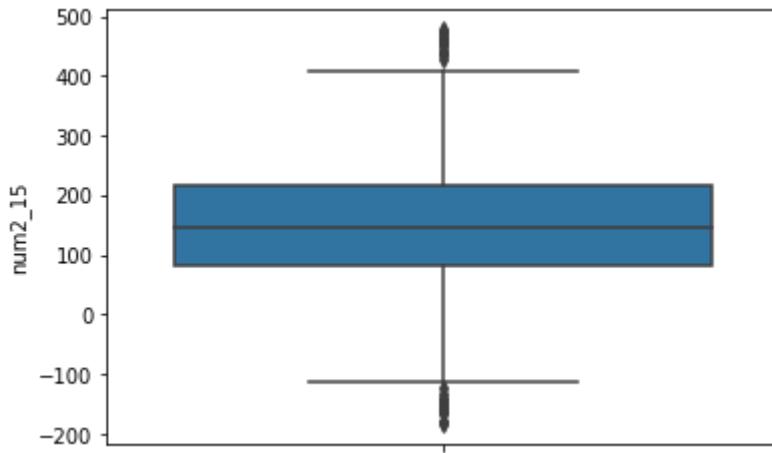
Out[304...]



In [305...]

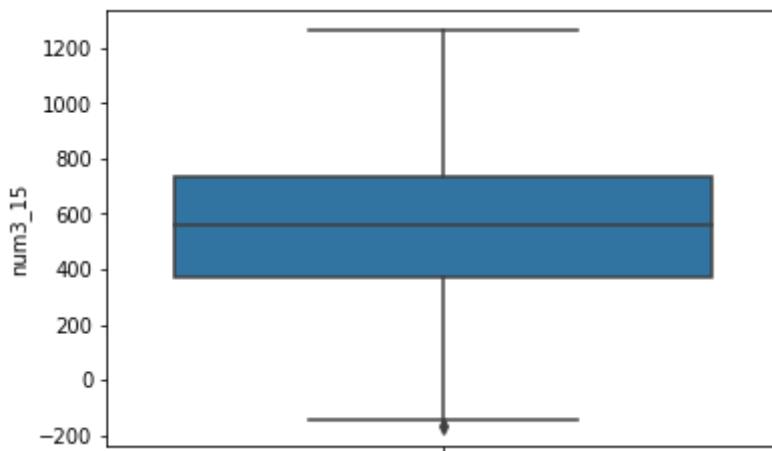
```
# num2_15
sns.boxplot(y=data['num2_15'])
```

```
Out[305... <AxesSubplot:ylabel='num2_15'>
```



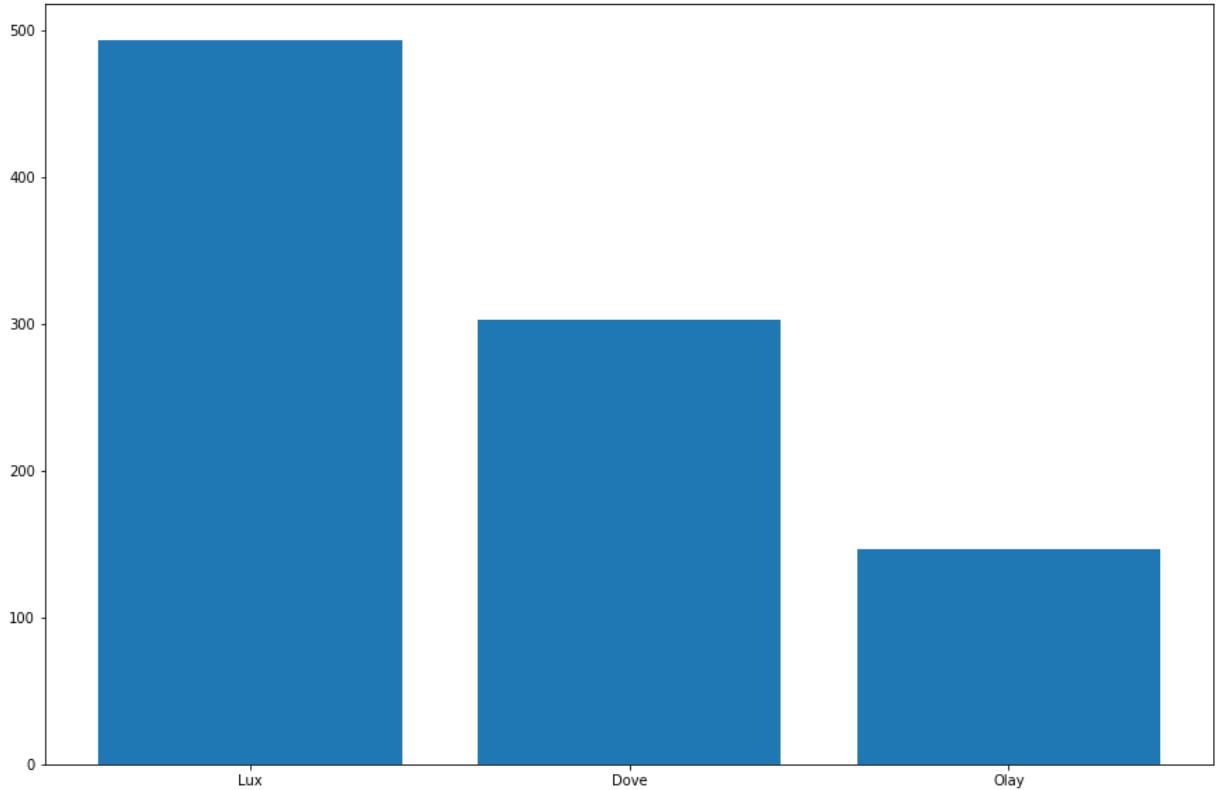
```
In [306... # num3_15  
sns.boxplot(y=data['num3_15'])
```

```
Out[306... <AxesSubplot:ylabel='num3_15'>
```



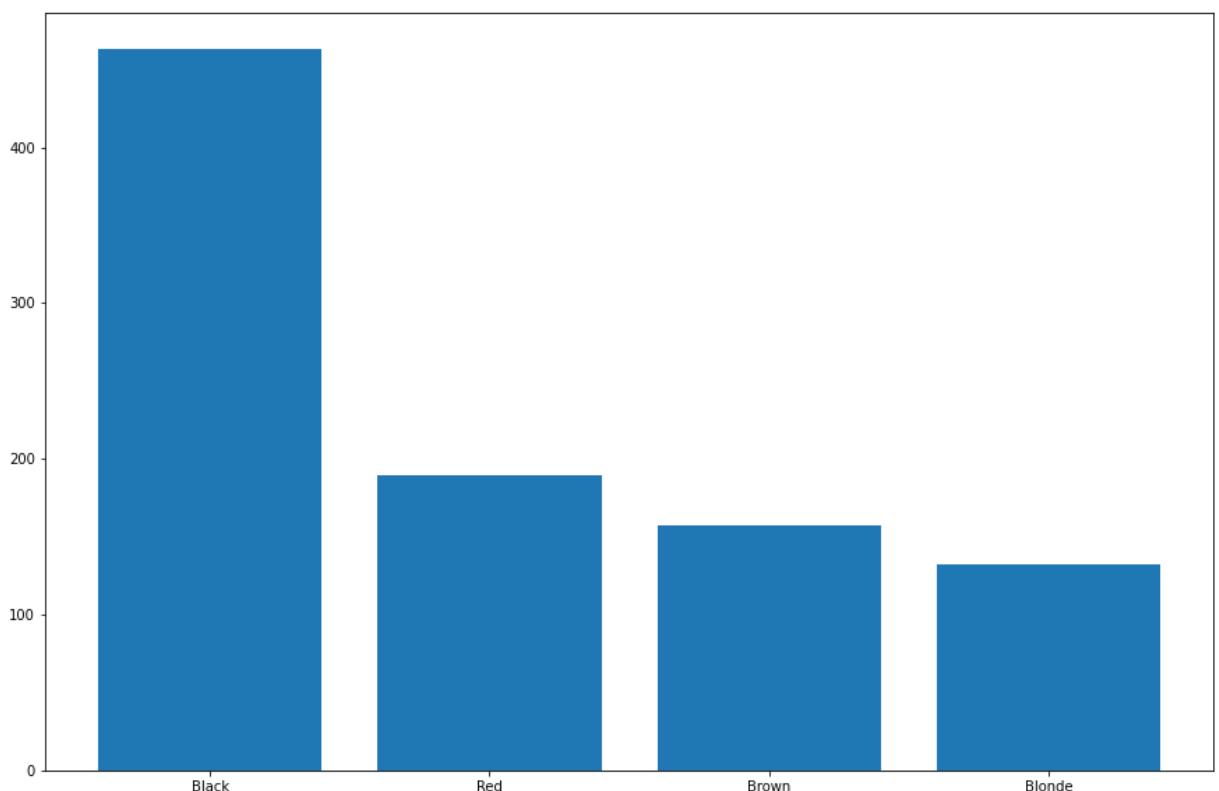
Столбчатые для категориальных

```
In [307... # cat6_brandOfSoaps  
counts = Counter(data['cat6_brandOfSoaps'].dropna())  
plt.figure(figsize=(15, 10))  
plt.bar(counts.keys(), counts.values())  
plt.show()
```



In [308]:

```
# cat2_hairColor
counts = Counter(data['cat2_hairColor'].dropna())
plt.figure(figsize=(15, 10))
plt.bar(counts.keys(), counts.values())
plt.show()
```



## 2. Провести проверку правдоподобности исходных данных

In [309]:

```
data
```

Out[309...]

	cat6_brandOfSoaps	cat2_hairColor	num1_15	num2_15	num3_15
0	Lux	Black	282.228024	NaN	807.976365
1	NaN	Red	1039.089956	268.388947	444.529442
2	Lux	Black	500.696085	174.992127	428.723966
3	Lux	Red	1346.429995	108.647900	520.238037
4	Dove	Black	230.576445	150.113065	896.476167
...	...	...	...	...	...
995	Dove	Red	422.732662	170.810692	586.254084
996	Lux	Brown	783.565442	237.086397	826.693312
997	Dove	Red	410.595523	196.686858	600.377791
998	Lux	Black	-237.136035	179.262913	371.380055
999	Lux	Black	22.006500	230.004737	326.041690

1000 rows × 5 columns

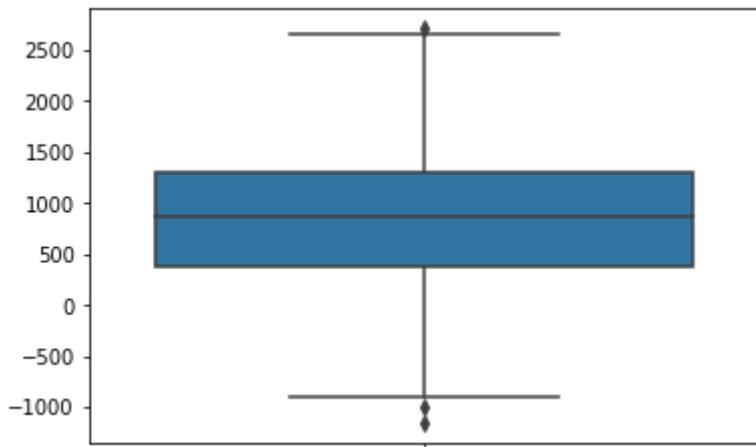
### 3. Поиск аномальных значений

In [310...]

```
c, low, up = sigmaclip(data['num1_15'].dropna(), 3, 3)  
sns.boxplot(y=c)
```

Out[310...]

<AxesSubplot:>

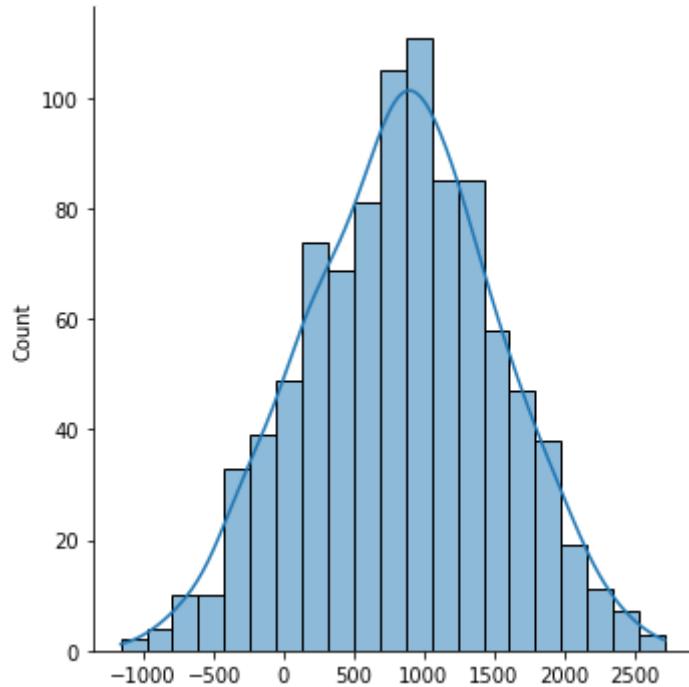


In [311...]

```
sns.displot(c, kde=True)
```

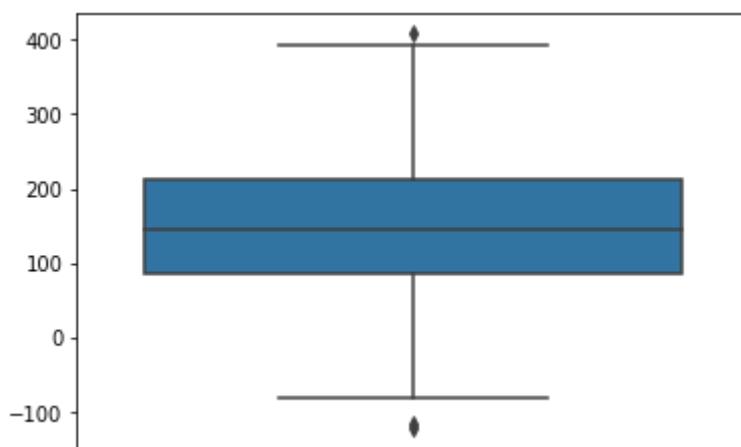
Out[311...]

<seaborn.axisgrid.FacetGrid at 0x7f6b3ebf30a0>



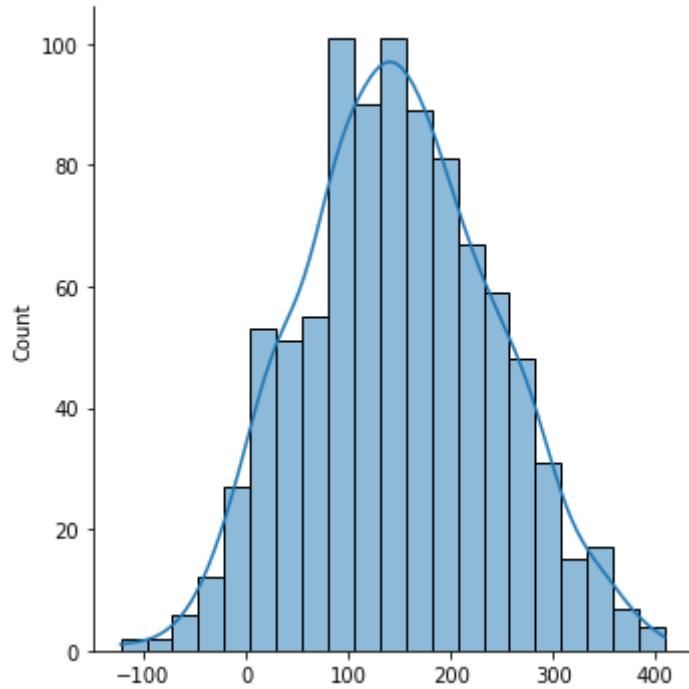
```
In [312...]: c, low, up = sigmaclip(data['num2_15'].dropna(), 3, 3)
sns.boxplot(y=c)
```

```
Out[312...]: <AxesSubplot:>
```



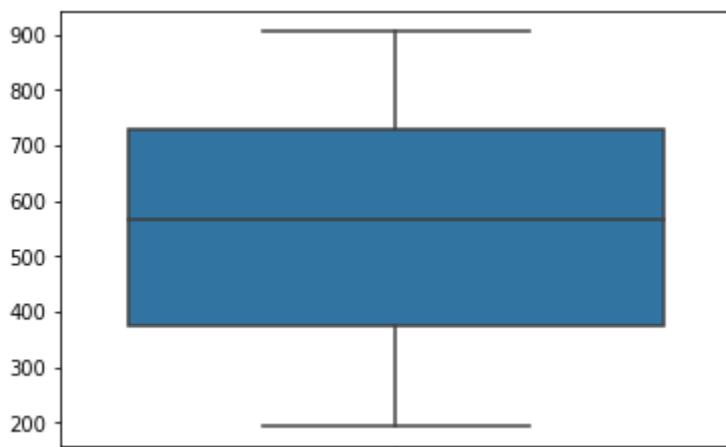
```
In [313...]: sns.displot(c, kde=True)
```

```
Out[313...]: <seaborn.axisgrid.FacetGrid at 0x7f6b3ec029a0>
```



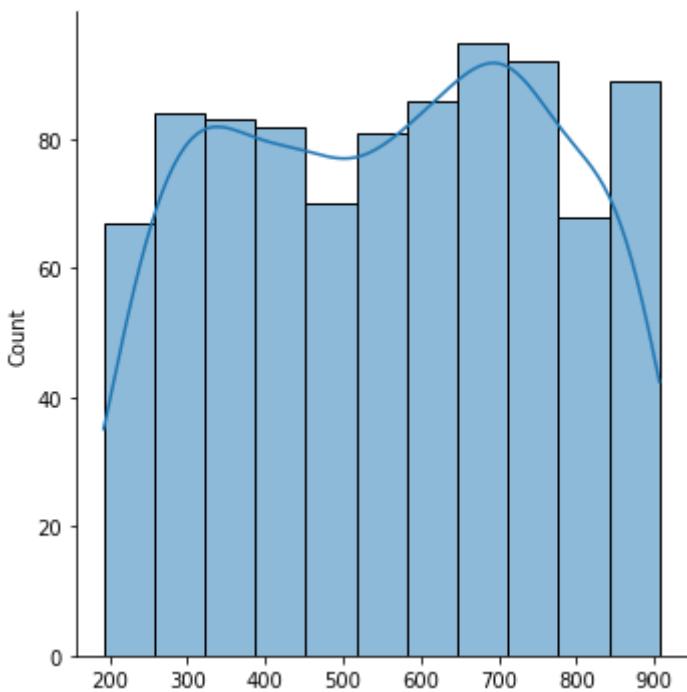
```
In [314...]: c, low, up = sigmaclip(data['num3_15'].dropna(), 3, 3)
sns.boxplot(y=c)
```

```
Out[314...]: <AxesSubplot:>
```



```
In [315...]: sns.displot(c, kde=True)
```

```
Out[315...]: <seaborn.axisgrid.FacetGrid at 0x7f6b3eac6820>
```

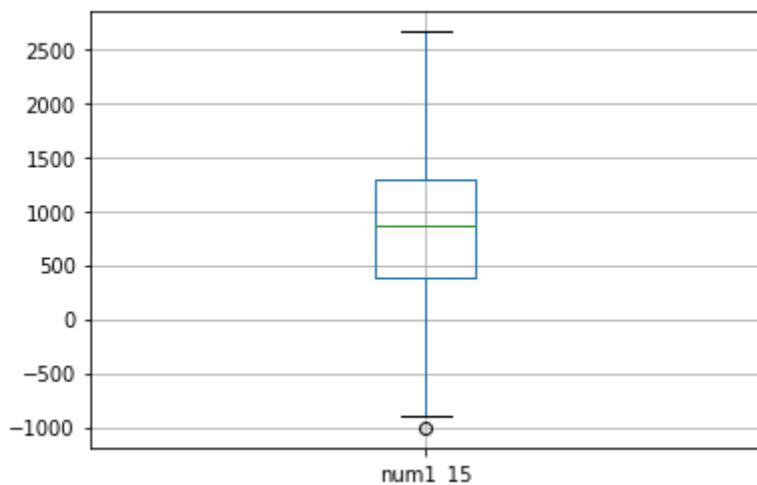


In [316...]

```
# Метод квантилей

q1 = np.nanquantile(data['num1_15'], 0.25)
q3 = np.nanquantile(data['num1_15'], 0.75)
iqr = q3 - q1
data.query("@q1 - 1.5*iqr < num1_15 < @q3 + 1.5*iqr").boxplot(column='num1_15')
```

Out[316...]



In [316...]

#### 4. Поиск и восстановление пропущенных значений

In [317...]

```
knn_imputer = KNNImputer()
enc = OneHotEncoder()
le = preprocessing.LabelEncoder()
```

Метод ближайших соседей для cat6\_brand0fSoaps

In [318...]

```
def formate_inputed_array(array):
    m = max(array)
```

```

    for i in range(len(array)):
        if array[i] == m:
            array[i] = 1
        else:
            array[i] = 0

array = np.append(array, 0)

return array

enc_dataset = pd.DataFrame(enc.fit_transform(data[['cat6_brandOfSoaps']])).toarray()
idx = enc_dataset.groupby([np.nan]).indices[1]
enc_dataset['Lux'] = pd.DataFrame(np.where(enc_dataset['Lux'].index.isin(idx)))
enc_dataset['Dove'] = pd.DataFrame(np.where(enc_dataset['Dove'].index.isin(idx)))
enc_dataset['Olay'] = pd.DataFrame(np.where(enc_dataset['Olay'].index.isin(idx)))
enc_dataset = enc_dataset.drop(columns=[np.nan])
imputed_dataset = knn_imputer.fit_transform(enc_dataset)

imputed_dataset = np.apply_along_axis(formate_inputed_array, 1, imputed_dataset)

restored_data = enc.inverse_transform(imputed_dataset)
data['cat6_brandOfSoaps'] = restored_data
data['cat6_brandOfSoaps']

```

Out[318...]

0	Dove
1	Olay
2	Dove
3	Dove
4	Dove
...	
995	Dove
996	Dove
997	Dove
998	Dove
999	Dove

Name: cat6\_brandOfSoaps, Length: 1000, dtype: object

Самым популярным значением для cat2\_hairColor

In [319...]

```

data1 = Counter(data['cat2_hairColor'].dropna())
mode = data1.most_common(1)[0][0]
data['cat2_hairColor'].fillna(mode, inplace=True)
data['cat2_hairColor']

```

Out[319...]

0	Black
1	Red
2	Black
3	Red
4	Black
...	
995	Red
996	Brown
997	Red
998	Black
999	Black

Name: cat2\_hairColor, Length: 1000, dtype: object

Среднее для num1\_15

In [320...]

```

data['num1_15'].fillna(np.mean(data['num1_15'].dropna()), inplace=True)
data['num1_15']

```

Out[320...]

0	282.228024
---	------------

```
1      1039.089956
2      500.696085
3      1346.429995
4      230.576445
...
995     422.732662
996     783.565442
997     410.595523
998    -237.136035
999     22.006500
Name: num1_15, Length: 1000, dtype: float64
```

Медиана для num2\_15

```
In [321...]: data['num2_15'].fillna(np.median(data['num2_15'].dropna()), inplace=True)
data['num2_15']
```

```
Out[321...]: 0      145.220453
1      268.388947
2      174.992127
3      108.647900
4      150.113065
...
995     170.810692
996     237.086397
997     196.686858
998     179.262913
999     230.004737
Name: num2_15, Length: 1000, dtype: float64
```

Метод ближайших соседей для num3\_15

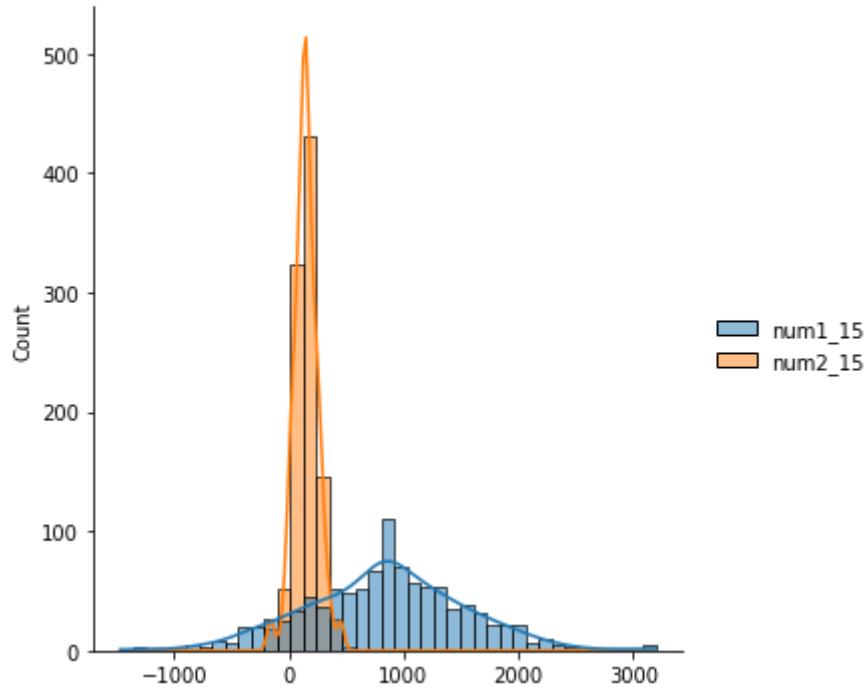
```
In [322...]: scaler = MinMaxScaler()
scaled_values = scaler.fit_transform(data[['num3_15']])
data[['num3_15']] = scaled_values

dataset_copy = data.copy()
inputed_df = knn_imputer.fit_transform(dataset_copy[['num3_15']])
data['num3_15'] = inputed_df[:, 0]
data['num3_15']
```

```
Out[322...]: 0      0.681191
1      0.427738
2      0.416716
3      0.480534
4      0.742907
...
995     0.526571
996     0.694243
997     0.536420
998     0.376727
999     0.345110
Name: num3_15, Length: 1000, dtype: float64
```

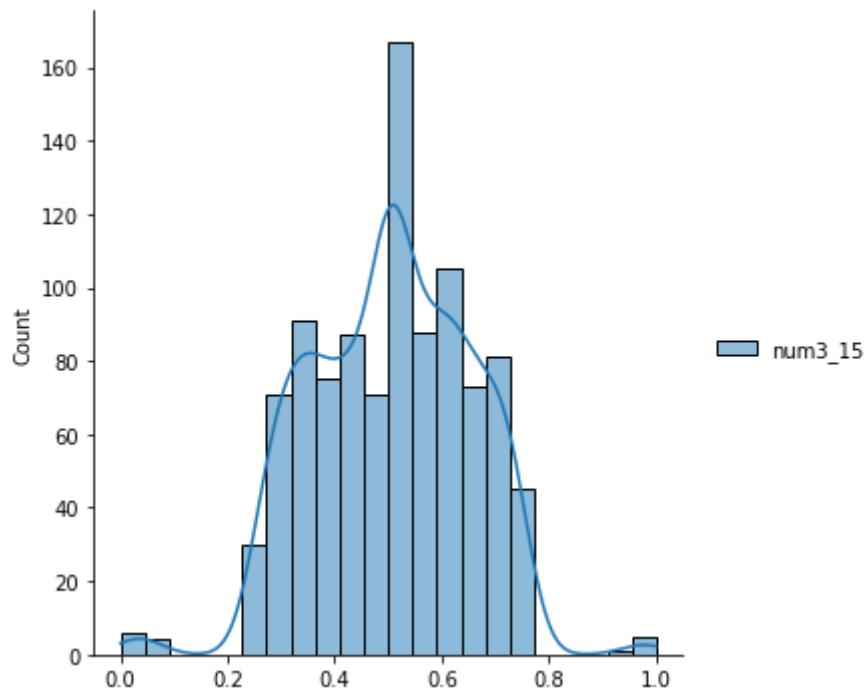
```
In [323...]: sns.displot(data=[data['num1_15'], data['num2_15']], kde=True)
```

```
Out[323...]: <seaborn.axisgrid.FacetGrid at 0x7f6b3ea34e80>
```



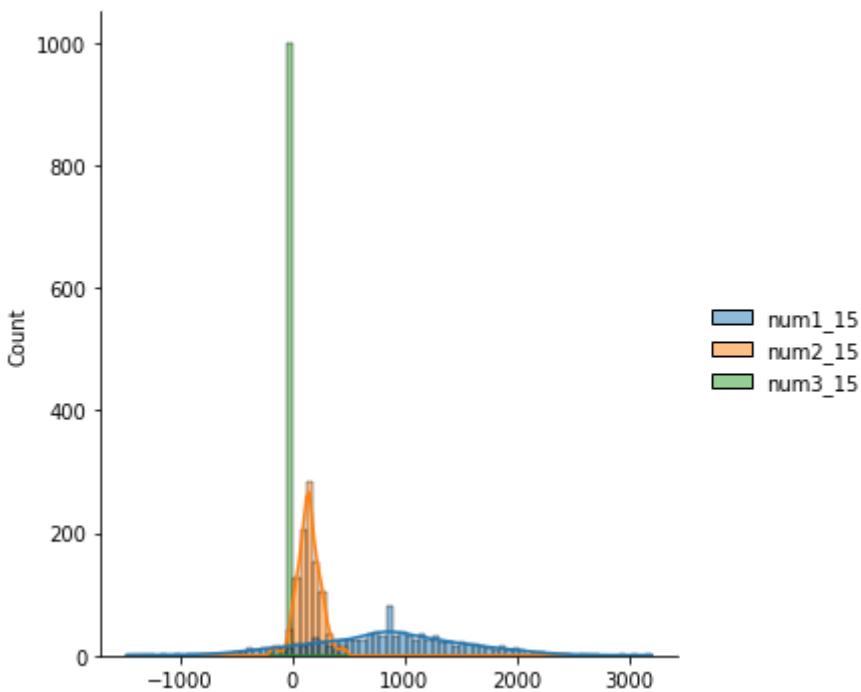
```
In [324...]: sns.displot(data=[data['num3_15']], kde=True)
```

```
Out[324...]: <seaborn.axisgrid.FacetGrid at 0x7f6b3e8a22e0>
```



```
In [325...]: sns.displot(data=data, kde=True)
```

```
Out[325...]: <seaborn.axisgrid.FacetGrid at 0x7f6b3ebf3940>
```



In [326...]

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 5 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   cat6_brandOfSoaps    1000 non-null   object 
 1   cat2_hairColor       1000 non-null   object 
 2   num1_15              1000 non-null   float64
 3   num2_15              1000 non-null   float64
 4   num3_15              1000 non-null   float64
dtypes: float64(3), object(2)
memory usage: 39.2+ KB
```

In [327...]

```
data.describe()
```

Out[327...]

	num1_15	num2_15	num3_15
<b>count</b>	1000.000000	1000.000000	1000.000000
<b>mean</b>	843.822989	148.776664	0.505075
<b>std</b>	698.398072	104.142567	0.147492
<b>min</b>	-1467.940219	-184.922189	0.000000
<b>25%</b>	394.078433	86.660775	0.390650
<b>50%</b>	843.822989	145.220453	0.505075
<b>75%</b>	1284.983740	210.869817	0.617954
<b>max</b>	3201.191252	479.162302	1.000000

## 5. Преобразование данных

In [328...]

```
# One-hot encoding для cat6_brandOfSoaps
one = enc.fit_transform(data[['cat6_brandOfSoaps']]).toarray()
one
```

```
array([[1., 0., 0.],
```

```
Out[328... [0., 0., 1.],  
[1., 0., 0.],  
[...]  
[1., 0., 0.],  
[1., 0., 0.],  
[1., 0., 0.]])
```

```
In [329... # Label encoding для cat2_hairColor
```

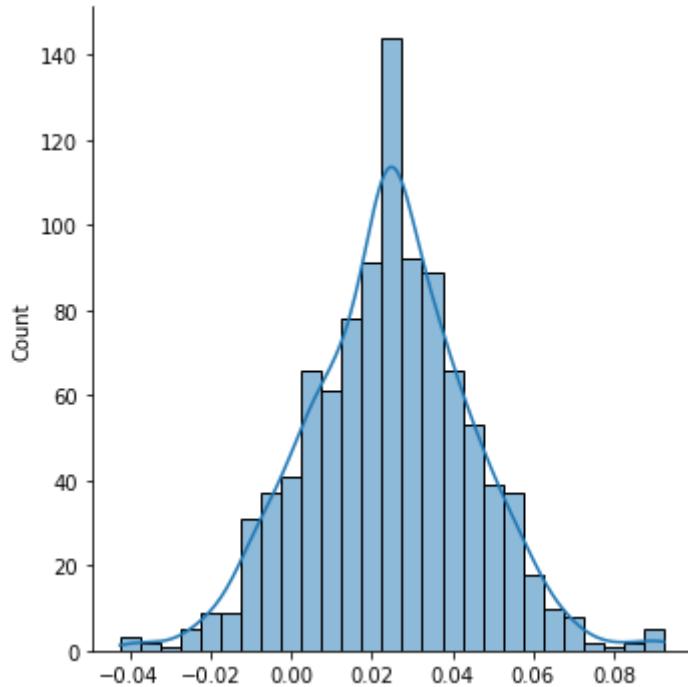
```
le.fit_transform(data['cat2_hairColor'])
```

```
Out[329... array([0, 3, 0, 3, 0, 3, 0, 0, 0, 0, 0, 2, 3, 1, 0, 2, 0, 1, 0, 0, 0, 3,  
0, 0, 2, 3, 2, 0, 2, 0, 2, 2, 0, 0, 3, 0, 0, 0, 0, 0, 2, 0, 2, 1, 0,  
0, 0, 2, 0, 0, 3, 0, 3, 0, 0, 0, 3, 0, 0, 0, 0, 2, 2, 0, 0, 0, 0, 0, 0,  
3, 0, 0, 3, 3, 2, 0, 0, 0, 0, 3, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1,  
0, 1, 0, 3, 0, 0, 0, 1, 0, 3, 3, 1, 0, 0, 0, 0, 0, 3, 2, 0, 2, 1, 0,  
0, 0, 1, 3, 3, 2, 0, 2, 0, 0, 0, 0, 2, 0, 0, 0, 0, 3, 0, 0, 0, 0, 0, 0,  
0, 3, 0, 1, 1, 2, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 3, 2, 0, 0, 0, 0,  
1, 0, 0, 3, 1, 0, 0, 2, 0, 1, 0, 0, 0, 2, 0, 0, 0, 2, 1, 3, 0, 1, 2, 0,  
1, 3, 3, 0, 2, 3, 3, 0, 1, 3, 2, 1, 0, 0, 2, 2, 0, 3, 0, 3, 2, 0, 0,  
3, 0, 1, 1, 0, 2, 0, 0, 0, 2, 0, 0, 3, 2, 2, 3, 0, 2, 1, 0, 0, 0, 0, 0,  
3, 3, 3, 2, 0, 0, 0, 2, 1, 0, 2, 3, 0, 1, 0, 0, 0, 0, 3, 0, 0, 0, 2, 0,  
3, 0, 0, 2, 2, 0, 3, 3, 0, 2, 0, 3, 3, 0, 0, 0, 0, 1, 2, 0, 1, 3, 1,  
0, 3, 1, 3, 0, 0, 0, 0, 1, 0, 0, 0, 0, 3, 2, 2, 0, 2, 0, 2, 2, 0,  
0, 0, 0, 3, 3, 0, 2, 1, 0, 2, 0, 2, 0, 2, 2, 1, 3, 0, 0, 0, 0, 0, 0, 0,  
0, 2, 0, 0, 3, 2, 1, 0, 2, 1, 3, 2, 3, 2, 2, 2, 0, 0, 0, 1, 1, 1, 3,  
3, 1, 0, 0, 3, 0, 0, 0, 1, 1, 3, 2, 1, 0, 0, 3, 0, 0, 0, 0, 1, 2, 0,  
0, 3, 0, 0, 0, 0, 2, 2, 0, 3, 3, 3, 3, 0, 0, 0, 2, 2, 0, 0, 2, 0, 0,  
0, 0, 0, 0, 0, 1, 0, 3, 0, 0, 0, 0, 0, 0, 3, 0, 2, 0, 0, 0, 3, 0, 3,  
0, 1, 1, 0, 2, 0, 1, 0, 0, 0, 1, 0, 1, 3, 0, 0, 0, 1, 3, 1, 0, 0, 3,  
3, 1, 3, 3, 0, 0, 1, 0, 0, 1, 3, 0, 3, 0, 0, 0, 0, 0, 0, 0, 3, 2, 3,  
2, 2, 1, 2, 0, 0, 3, 2, 1, 2, 2, 2, 0, 3, 0, 0, 0, 3, 0, 0, 2, 2, 1,  
0, 2, 2, 0, 2, 1, 2, 1, 2, 0, 3, 1, 1, 2, 3, 0, 3, 0, 2, 2, 3, 1, 0,  
0, 2, 0, 0, 2, 1, 3, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 3, 3, 0,  
0, 0, 0, 0, 2, 0, 3, 2, 1, 0, 0, 0, 0, 2, 1, 1, 3, 0, 0, 1, 3, 3, 0, 0,  
3, 0, 0, 1, 2, 3, 0, 3, 0, 2, 1, 3, 0, 0, 0, 1, 1, 0, 0, 3, 3, 0, 0,  
1, 0, 3, 0, 1, 3, 0, 3, 2, 1, 0, 3, 2, 1, 1, 1, 0, 0, 0, 2, 0, 1,  
0, 0, 1, 0, 0, 2, 0, 0, 2, 2, 0, 1, 2, 0, 1, 3, 0, 0, 0, 0, 0, 0, 0, 0,  
0, 0, 0, 3, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 3, 3, 3, 0, 0, 0, 1, 0, 0, 0,  
0, 1, 0, 2, 2, 0, 2, 0, 0, 2, 2, 0, 0, 1, 1, 3, 1, 1, 0, 0, 3, 3, 2,  
0, 0, 0, 3, 3, 2, 1, 3, 0, 1, 0, 0, 3, 0, 0, 0, 0, 0, 1, 2, 3, 3, 2,  
1, 0, 0, 2, 2, 0, 0, 0, 3, 3, 1, 1, 0, 0, 3, 3, 0, 0, 3, 0, 0, 0, 3, 1,  
3, 1, 0, 0, 2, 0, 3, 0, 1, 0, 3, 0, 0, 0, 0, 1, 3, 2, 2, 0, 2, 0,  
0, 0, 0, 0, 0, 3, 0, 2, 0, 3, 0, 0, 0, 0, 0, 1, 0, 0, 0, 3, 1, 0, 0,  
0, 3, 0, 0, 3, 0, 2, 2, 0, 0, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 1, 1, 3, 3,  
0, 3, 2, 0, 0, 3, 0, 0, 0, 2, 3, 0, 1, 3, 0, 1, 2, 1, 2, 1, 0, 0, 2, 2,  
2, 0, 3, 3, 3, 1, 2, 3, 0, 3, 3, 0, 3, 2, 3, 3, 0, 0, 0, 0, 0, 2, 0,  
0, 1, 0, 3, 0, 0, 0, 0, 0, 0, 0, 2, 2, 0, 2, 0, 0, 0, 0, 0, 0, 3, 2, 0,  
0, 3, 1, 0, 0, 0, 0, 2, 3, 0, 2, 0, 0, 3, 2, 0, 2, 0, 0, 3, 1, 2, 0,  
0, 1, 0, 0, 0, 0, 0, 2, 0, 3, 0, 2, 0, 3, 2, 0, 3, 0, 0, 0, 0, 0, 2, 3,  
1, 0, 0, 3, 0, 3, 0, 2, 0, 0, 0, 0, 0, 0, 0, 2, 0, 2, 0, 0, 0, 0, 3, 2,  
1, 0, 3, 1, 0, 0, 3, 0, 0, 0, 0, 0, 2, 0, 1, 0, 1, 0, 0, 0, 2, 3, 1,  
2, 0, 0, 1, 1, 0, 3, 1, 0, 1, 2, 0, 0, 3, 0, 0, 0, 1, 3, 0, 0, 0, 3, 0,  
2, 1, 3, 0, 3, 1, 0, 3, 2, 3, 3, 2, 0, 3, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0,  
2, 3, 0, 0, 2, 1, 0, 0, 0, 0, 3, 0, 0, 0, 0, 0, 3, 0, 1, 0, 0, 3, 3, 0,  
0, 0, 0, 0, 0, 1, 0, 0, 0, 3, 3, 0, 3, 0, 3, 0, 2, 0, 0, 1, 3, 0, 0,  
2, 0, 2, 0, 0, 3, 2, 3, 0, 0, 0])
```

```
In [330... # Нормализация для num1_15
```

```
n = preprocessing.normalize([data['num1_15']])
sns.displot(n[0], kde=True)
```

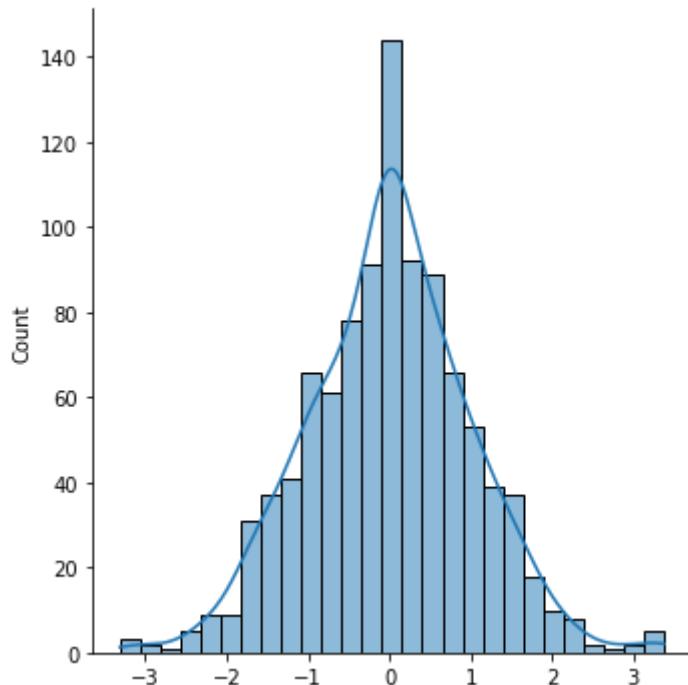
```
Out[330... <seaborn.axisgrid.FacetGrid at 0x7f6b3e7d3a00>
```



In [331...]

```
# Стандартизация для num1_15  
st = preprocessing.scale(data['num1_15'])  
sns.displot(data=st, kde=True)
```

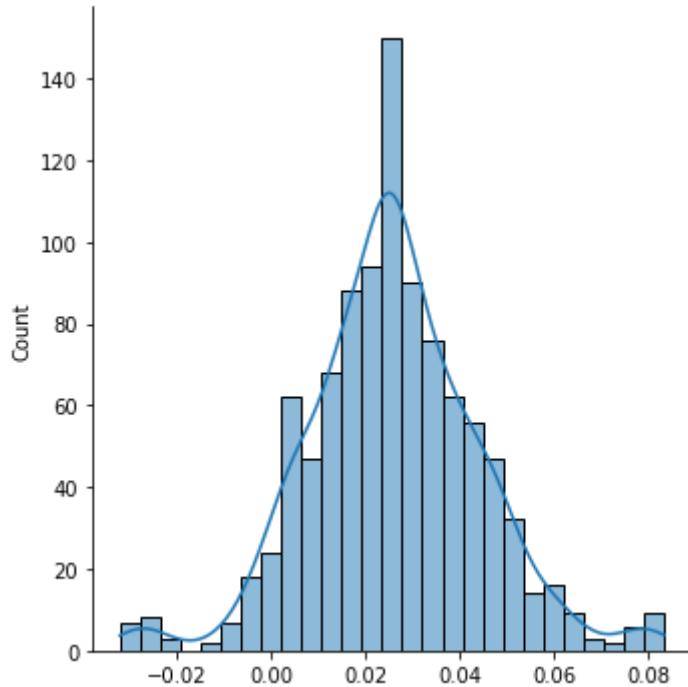
Out[331...]



In [332...]

```
# Нормализация для num2_15  
n = preprocessing.normalize([data['num2_15']])  
sns.displot(n[0], kde=True)
```

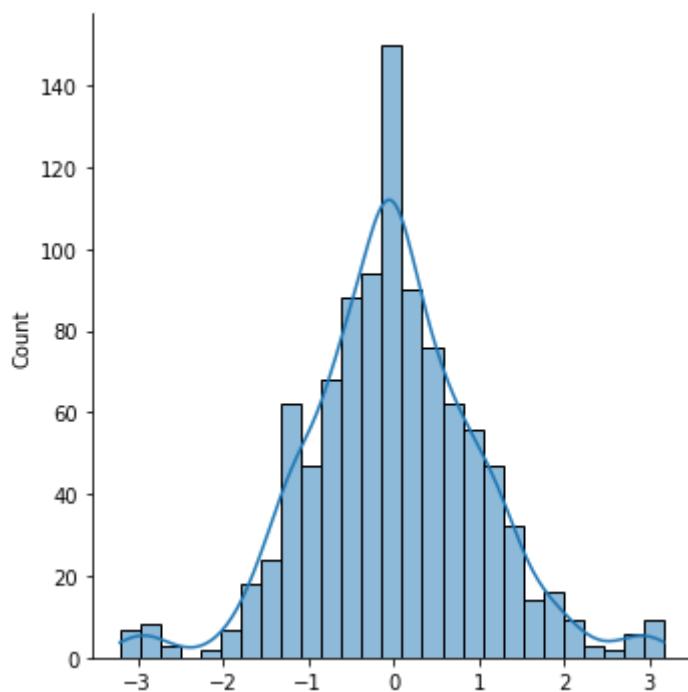
Out[332...]

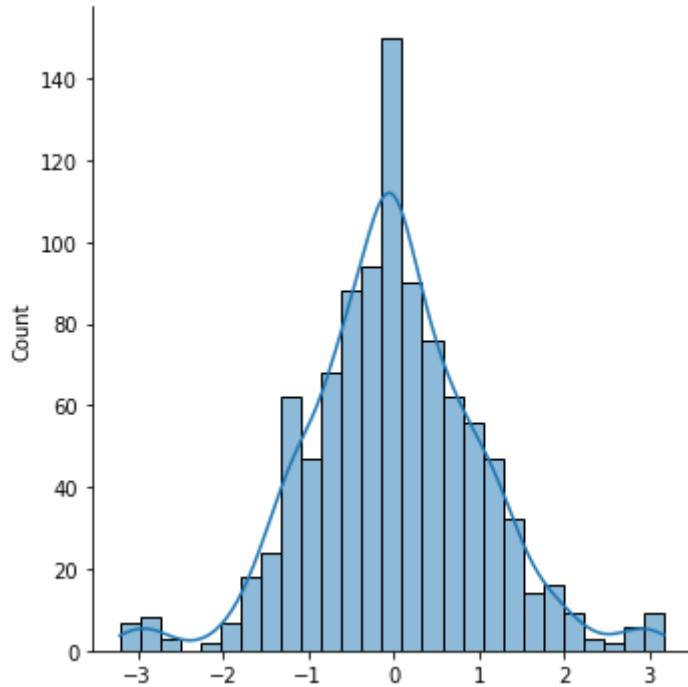


In [333...]

```
# Стандартизация для num2_15  
  
st = preprocessing.scale(data['num2_15'])  
sns.displot(data=st, kde=True)  
sns.displot(data=st, kde=True)
```

Out[333...]

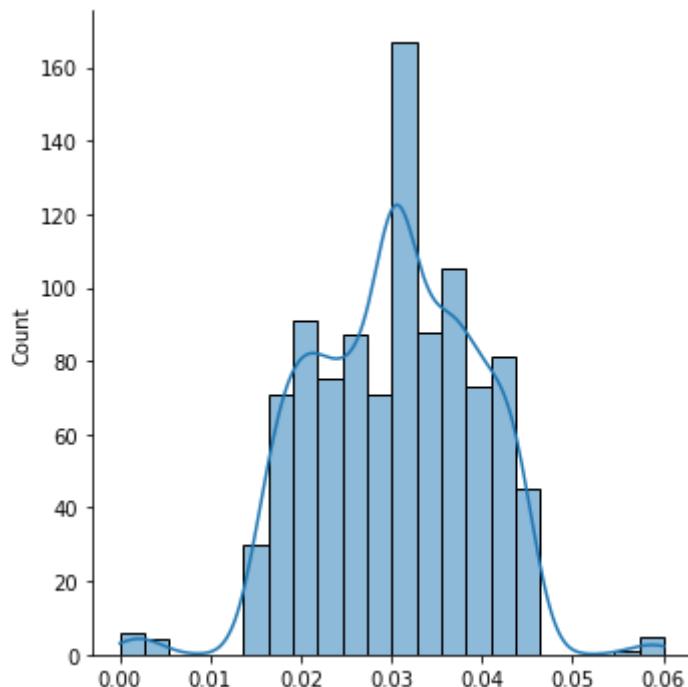




In [334...]

```
# Нормализация для num3_15  
n = preprocessing.normalize([data['num3_15']])  
sns.displot(n[0], kde=True)
```

Out[334...]



In [335...]

```
# Стандартизация для num3_15  
st = preprocessing.scale(data['num3_15'])  
sns.displot(data=st, kde=True)
```

Out[335...]

