

Цель лабораторной работы:

изучение способов предварительной обработки данных для дальнейшего формирования моделей.

Задание:

Выбрать набор данных (датасет), содержащий категориальные признаки и пропуски в данных. Для выполнения следующих пунктов можно использовать несколько различных наборов данных (один для обработки пропусков, другой для категориальных признаков и т.д.) Для выбранного датасета (датасетов) на основе материалов лекции решить следующие задачи: обработку пропусков в данных; кодирование категориальных признаков; масштабирование данных.

```
In [331]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set(style="ticks")
```

```
In [332]: # Набор 1
data = pd.read_csv('quakes.csv', sep=";")
# Набор 2
data1 = pd.read_csv('esoph.csv', sep=";")
```

```
In [333]: # Набор 1
data.shape
```

```
Out[333]: (1000, 6)
```

```
In [334]: # Набор 2
data1.shape
```

```
Out[334]: (88, 6)
```

```
In [335]: # Набор 1
data.dtypes
```

```
Out[335]: Unnamed: 0      int64
lat          float64
long         float64
depth        float64
mag          object
stations     int64
dtype: object
```

```
In [336]: # Набор 2
data1.dtypes
```

```
Out[336]: Unnamed: 0      int64
agegp      float64
alcgp      object
tobgp      object
ncases      int64
ncontrols   float64
dtype: object
```

```
In [337]: # пропуски в наборе 1
data.isnull().sum()
```

```
Out[337]: Unnamed: 0      0
lat        0
long       12
depth      6
mag        0
stations   0
dtype: int64
```

```
In [338]: # пропуски в наборе 2
data1.isnull().sum()
```

```
Out[338]: Unnamed: 0      0
agegp      7
alcgp      0
tobgp      0
ncases      0
ncontrols   6
dtype: int64
```

```
In [339]: # Набор 1
data.head()
```

```
Out[339]:
```

	Unnamed: 0	lat	long	depth	mag	stations
0	1	-20.42	181.62	562.0	04.Aug	41
1	2	-20.62	181.03	650.0	04.Feb	15
2	3	-26.00	184.10	42.0	05.Apr	43
3	4	-17.97	181.66	626.0	04.Jan	19
4	5	-20.42	181.96	649.0	4	11

```
In [340]: # Набор 2
data1.head()
```

```
Out[340]:
```

	Unnamed: 0	agegp	alcgp	tobgp	ncases	ncontrols
0	1	2534.0	0-39g/day	0-9g/day	0	40.0
1	2	2534.0	0-39g/day	Oct.19	0	10.0
2	3	2534.0	0-39g/day	20-29	0	6.0
3	4	2534.0	0-39g/day	30+	0	5.0
4	5	2534.0	40-79	0-9g/day	0	27.0

```
In [341]: total_count = data.shape[0]
print('Строки в наборе 1: {}'.format(total_count))
```

Строки в наборе 1: 1000

```
In [342]: total_count1 = data1.shape[0]
print('Строки в наборе 2: {}'.format(total_count1))
```

Строки в наборе 2: 88

Обработка пропусков

```
In [343]: # Удаление колонок, содержащих пустые значения в наборе 1
data_new_1 = data.dropna(axis=1, how='any')
(data.shape, data_new_1.shape)
```

```
Out[343]: ((1000, 6), (1000, 4))
```

```
In [344]: # Удаление колонок, содержащих пустые значения в наборе 2
data_new_11 = data1.dropna(axis=1, how='any')
(data1.shape, data_new_11.shape)
```

```
Out[344]: ((88, 6), (88, 4))
```

```
In [345]: # Удаление строк, содержащих пустые значения в наборе 1
data_new_2 = data.dropna(axis=0, how='any')
(data.shape, data_new_2.shape)
```

```
Out[345]: ((1000, 6), (982, 6))
```

In [346]: `data.head()`

Out[346]:

	Unnamed: 0	lat	long	depth	mag	stations
0	1	-20.42	181.62	562.0	04.Aug	41
1	2	-20.62	181.03	650.0	04.Feb	15
2	3	-26.00	184.10	42.0	05.Apr	43
3	4	-17.97	181.66	626.0	04.Jan	19
4	5	-20.42	181.96	649.0	4	11

In [347]: *# Удаление строк, содержащих пустые значения в наборе 2*
`data_new_21 = data1.dropna(axis=0, how='any')`
`(data1.shape, data_new_21.shape)`

Out[347]: `((88, 6), (76, 6))`

In [348]: `data1.head()`

Out[348]:

	Unnamed: 0	agegp	alcgp	tobgp	ncases	ncontrols
0	1	2534.0	0-39g/day	0-9g/day	0	40.0
1	2	2534.0	0-39g/day	Oct.19	0	10.0
2	3	2534.0	0-39g/day	20-29	0	6.0
3	4	2534.0	0-39g/day	30+	0	5.0
4	5	2534.0	40-79	0-9g/day	0	27.0

In [349]: *# Заполнение всех пропущенных значений нулями в наборе 1*
`data_new_3 = data.fillna(0)`
`data_new_3.head()`

Out[349]:

	Unnamed: 0	lat	long	depth	mag	stations
0	1	-20.42	181.62	562.0	04.Aug	41
1	2	-20.62	181.03	650.0	04.Feb	15
2	3	-26.00	184.10	42.0	05.Apr	43
3	4	-17.97	181.66	626.0	04.Jan	19
4	5	-20.42	181.96	649.0	4	11

```
In [350]: # Заполнение всех пропущенных значений нулями в наборе 1
data_new_31 = data1.fillna(0)
data_new_31.head()
```

Out[350]:

	Unnamed: 0	agegp	alcgp	tobgp	ncases	ncontrols
0	1	2534.0	0-39g/day	0-9g/day	0	40.0
1	2	2534.0	0-39g/day	Oct.19	0	10.0
2	3	2534.0	0-39g/day	20-29	0	6.0
3	4	2534.0	0-39g/day	30+	0	5.0
4	5	2534.0	40-79	0-9g/day	0	27.0

Импьютация

Числовые данные

```
In [351]: # Выберем числовые колонки с пропущенными значениями
# Цикл по колонкам датасета набора 1
num_cols = []
for col in data.columns:
    # Количество пустых значений
    temp_null_count = data[data[col].isnull()].shape[0]
    dt = str(data[col].dtype)
    if temp_null_count>0 and (dt=='float64' or dt=='int64'):
        num_cols.append(col)
        temp_perc = round((temp_null_count / total_count) * 100.0,
2)
        print('Колонка {}. Тип данных {}. Количество пустых значений {}, {}'.format(col, dt, temp_null_count, temp_perc))
```

Колонка long. Тип данных float64. Количество пустых значений 12, 1.2%.

Колонка depth. Тип данных float64. Количество пустых значений 6, 0.6%.

```
In [352]: # Выберем числовые колонки с пропущенными значениями
# Цикл по колонкам датасета набора 2
num_cols1 = []
for col in data1.columns:
    # Количество пустых значений
    temp_null_count1 = data1[data1[col].isnull()].shape[0]
    dt1 = str(data1[col].dtype)
    if temp_null_count1>0 and (dt1=='float64' or dt1=='int64'):
        num_cols1.append(col)
        temp_perc1 = round((temp_null_count1 / total_count1) * 100.
0, 2)
        print('Колонка {}. Тип данных {}. Количество пустых значени
й {}, {}%.'.format(col, dt1, temp_null_count1, temp_perc1))
```

Колонка agegrp. Тип данных float64. Количество пустых значений 7, 7.95%.

Колонка ncontrols. Тип данных float64. Количество пустых значений 6, 6.82%.

```
In [353]: # Фильтр по колонкам с пропущенными значениями набора 1
data_num = data[num_cols]
data_num
```

Out[353]:

	long	depth
0	181.62	562.0
1	181.03	650.0
2	184.10	42.0
3	181.66	626.0
4	181.96	649.0
...
995	179.54	470.0
996	167.06	248.0
997	184.20	244.0
998	187.80	40.0
999	170.56	165.0

1000 rows × 2 columns

```
In [354]: # Фильтр по колонкам с пропущенными значениями набора 2
data_num1 = data1[num_cols1]
data_num1
```

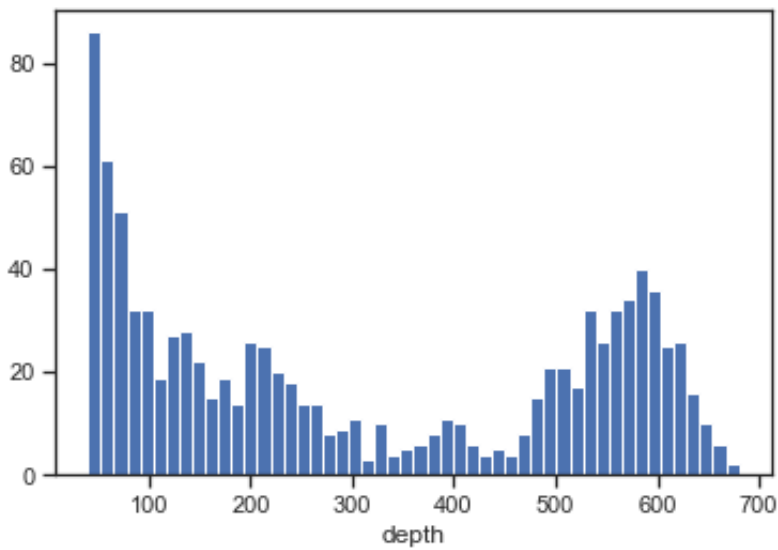
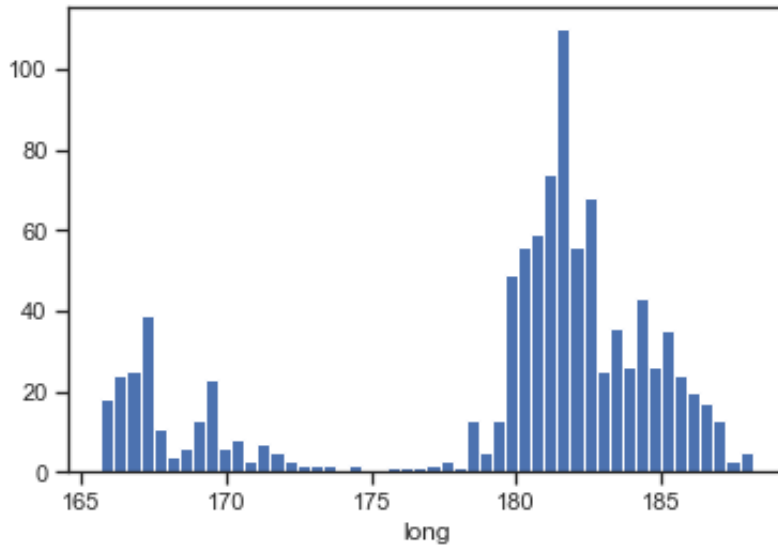
Out[354]:

	agegp	ncontrols
0	2534.0	40.0
1	2534.0	10.0
2	2534.0	6.0
3	2534.0	5.0
4	2534.0	27.0
...
83	75.0	1.0
84	75.0	1.0
85	75.0	1.0
86	75.0	NaN
87	75.0	1.0

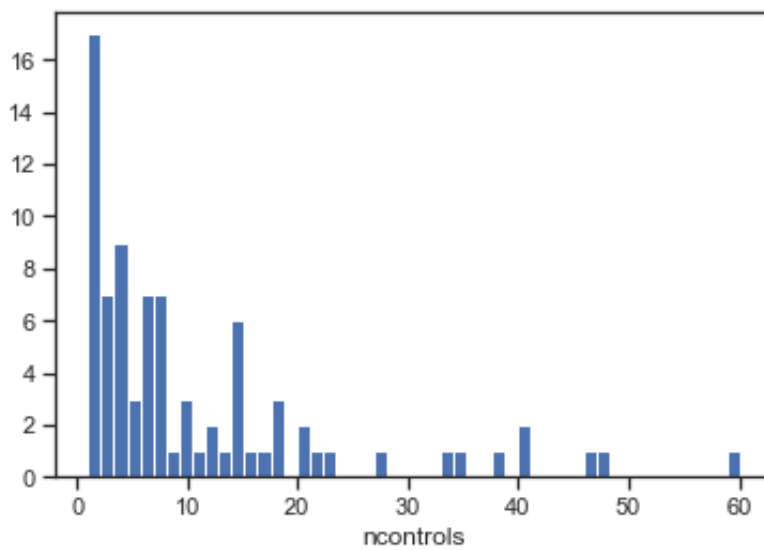
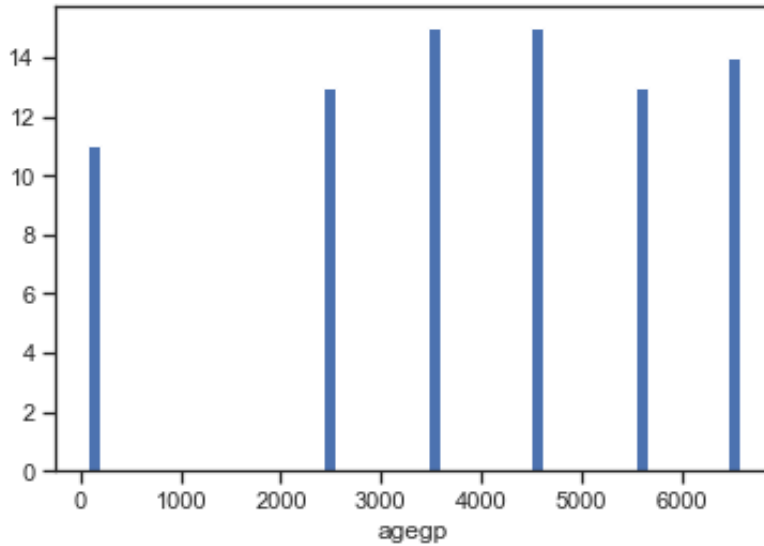
88 rows × 2 columns

```
In [355]: # Гистограмма по признакам набора 1 - Rating
for col in data_num:
    plt.hist(data[col], 50)
    plt.xlabel(col)
    plt.show()
```

```
/usr/local/lib/python3.7/site-packages/numpy/lib/histograms.py:839
: RuntimeWarning: invalid value encountered in greater_equal
  keep = (tmp_a >= first_edge)
/usr/local/lib/python3.7/site-packages/numpy/lib/histograms.py:840
: RuntimeWarning: invalid value encountered in less_equal
  keep &= (tmp_a <= last_edge)
```




```
In [356]: # Гистограмма по признакам набора 1: Sentiment_Polarity, Sentiment
          _subjectivity
          for col in data_num1:
              plt.hist(data1[col], 50)
              plt.xlabel(col)
              plt.show()
```



```
In [357]: # Фильтр по пустым значениям поля Rating
data[data['depth'].isnull()]
```

Out[357]:

	Unnamed: 0	lat	long	depth	mag	stations
81	82	-23.84	180.99	NaN	04.May	27
170	171	-17.82	181.83	NaN	04.Mar	24
188	189	-24.27	179.88	NaN	04.Jun	24
189	190	-15.85	185.13	NaN	04.Jun	29
345	346	-27.71	182.47	NaN	04.Mar	11
361	362	-16.90	185.72	NaN	4	22

```
In [358]: # Фильтр по пустым значениям поля Sentiment_Polarity
data1[data1['ncontrols'].isnull()]
```

Out[358]:

	Unnamed: 0	agegp	alcgp	tobgp	ncases	ncontrols
11	12	2534.0	120+	0-9g/day	0	NaN
22	23	3544.0	40-79	30+	0	NaN
46	47	NaN	0-39g/day	0-9g/day	2	NaN
67	68	6574.0	40-79	Oct.19	3	NaN
81	82	75.0	40-79	Oct.19	1	NaN
86	87	75.0	120+	0-9g/day	2	NaN

```
In [359]: # Запоминаем индексы строк с пустыми значениями поля Rating
flt_index = data[data['depth'].isnull()].index
flt_index
```

Out[359]: Int64Index([81, 170, 188, 189, 345, 361], dtype='int64')

```
In [360]: # Запоминаем индексы строк с пустыми значениями поля Sentiment_Polarity
flt_index1 = data1[data1['ncontrols'].isnull()].index
flt_index1
```

Out[360]: Int64Index([11, 22, 46, 67, 81, 86], dtype='int64')

```
In [361]: # Запоминаем индексы строк с пустыми значениями поля Sentiment_Subjectivity
flt_index11 = data1[data1['ncontrols'].isnull()].index
flt_index11
```

Out[361]: Int64Index([11, 22, 46, 67, 81, 86], dtype='int64')

In [362]: *# Проверяем что выводятся нужные строки Rating*
`data[data.index.isin(flt_index)]`

Out[362]:

	Unnamed: 0	lat	long	depth	mag	stations
81	82	-23.84	180.99	NaN	04.May	27
170	171	-17.82	181.83	NaN	04.Mar	24
188	189	-24.27	179.88	NaN	04.Jun	24
189	190	-15.85	185.13	NaN	04.Jun	29
345	346	-27.71	182.47	NaN	04.Mar	11
361	362	-16.90	185.72	NaN	4	22

In [363]: *# Проверяем что выводятся нужные строки Sentiment_Polarity*
`data1[data1.index.isin(flt_index1)]`

Out[363]:

	Unnamed: 0	agegp	alcgp	tobgp	ncases	ncontrols
11	12	2534.0	120+	0-9g/day	0	NaN
22	23	3544.0	40-79	30+	0	NaN
46	47	NaN	0-39g/day	0-9g/day	2	NaN
67	68	6574.0	40-79	Oct.19	3	NaN
81	82	75.0	40-79	Oct.19	1	NaN
86	87	75.0	120+	0-9g/day	2	NaN

In [364]: *# Проверяем что выводятся нужные строки Sentiment_Subjectivity*
`data1[data1.index.isin(flt_index11)]`

Out[364]:

	Unnamed: 0	agegp	alcgp	tobgp	ncases	ncontrols
11	12	2534.0	120+	0-9g/day	0	NaN
22	23	3544.0	40-79	30+	0	NaN
46	47	NaN	0-39g/day	0-9g/day	2	NaN
67	68	6574.0	40-79	Oct.19	3	NaN
81	82	75.0	40-79	Oct.19	1	NaN
86	87	75.0	120+	0-9g/day	2	NaN

```
In [365]: # фильтр по колонке Rating
data_num[data_num.index.isin(flt_index)][ 'depth' ]
```

```
Out[365]: 81      NaN
          170      NaN
          188      NaN
          189      NaN
          345      NaN
          361      NaN
          Name: depth, dtype: float64
```

```
In [366]: # фильтр по колонке Sentiment_Polarity
data_num1[data_num1.index.isin(flt_index1)][ 'ncontrols' ]
```

```
Out[366]: 11      NaN
          22      NaN
          46      NaN
          67      NaN
          81      NaN
          86      NaN
          Name: ncontrols, dtype: float64
```

```
In [367]: data_num_Rating = data_num[ 'depth' ]
data_num_Rating.head()
```

```
Out[367]:
```

	depth
0	562.0
1	650.0
2	42.0
3	626.0
4	649.0

```
In [368]: data_num_SPol = data_num1[ 'ncontrols' ]
data_num_SPol.head()
```

```
Out[368]:
```

	ncontrols
0	40.0
1	10.0
2	6.0
3	5.0
4	27.0

```
In [369]: from sklearn.impute import SimpleImputer
          from sklearn.impute import MissingIndicator
```

```
In [370]: # Фильтр для проверки заполнения пустых значений
indicator = MissingIndicator()
mask_missing_values_only = indicator.fit_transform(data_num_Rating)
mask_missing_values_only
```

[illegible]

Page 14 of 107

Page 15 of 107

Page 16 of 107

Page 17 of 107

Page 18 of 107

Page 19 of 107

Page 20 of 107

Page 21 of 107

Page 22 of 107

Page 23 of 107

Page 24 of 107

Page 25 of 107

Page 26 of 107

Page 27 of 107

Page 28 of 107

Page 29 of 107

Page 30 of 107

Page 31 of 107

```
In [371]: # Фильтр для проверки заполнения пустых значений
indicator = MissingIndicator()
mask_missing_values_only1 = indicator.fit_transform(data_num_SPol)
mask_missing_values_only1
```

file:///Users/17713067/Downloads/LAB3.html


```
[False],  
[ True],  
[False],  
[False],  
[False],  
[False],  
[False],  
[False],  
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[False],  
[False],  
[False],  
[ True],  
[False],  
[False],  
[False],  
[ True],  
[False]])
```

```
In [372]: # Фильтр для проверки заполнения пустых значений  
indicator = MissingIndicator()  
mask_missing_values_only11 = indicator.fit_transform(data_num_SPol)  
mask_missing_values_only11
```

```
Out[372]: array([[False],  
[False],  
[False],  
[False],
```

```
[False],
[False],
[False],
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[False],
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[False],
[False],
[False],
[False],
[ True],
[False]])
```

```
In [373]: strategies=['mean', 'median', 'most_frequent']
```

```
In [374]: # Rating
def test_num_impute(strategy_param):
    imp_num = SimpleImputer(strategy=strategy_param)
    data_num_imp = imp_num.fit_transform(data_num_Rating)
    return data_num_imp[mask_missing_values_only]
```

```
In [375]: # Sentiment_Polarity
def test_num_impute1(strategy_param):
    imp_num = SimpleImputer(strategy=strategy_param)
    data_num_imp = imp_num.fit_transform(data_num_SPol)
    return data_num_imp[mask_missing_values_only1]
```

```
In [376]: # Sentiment_Subjectivity
def test_num_impute11(strategy_param):
    imp_num = SimpleImputer(strategy=strategy_param)
    data_num_imp = imp_num.fit_transform(data_num_SPol)
    return data_num_imp[mask_missing_values_only11]
```

```
In [377]: strategies[0], test_num_impute(strategies[0])
```

```
Out[377]: ('mean', array([311.18008048, 311.18008048, 311.18008048, 311.1800
      8048,
      311.18008048, 311.18008048]))
```

```
In [378]: strategies[0], test_num_impute1(strategies[0])
```

```
Out[378]: ('mean', array([11., 11., 11., 11., 11., 11.]))
```

```
In [379]: # Sentiment_Subjectivity
strategies[0], test_num_impute11(strategies[0])
```

```
Out[379]: ('mean', array([11., 11., 11., 11., 11., 11.]))
```

```
In [380]: strategies[1], test_num_impute(strategies[1])
```

```
Out[380]: ('median', array([246., 246., 246., 246., 246., 246.]))
```

```
In [381]: strategies[1], test_num_impute1(strategies[1])
```

```
Out[381]: ('median', array([6., 6., 6., 6., 6., 6.]))
```

```
In [382]: # Sentiment_Subjectivity
strategies[1], test_num_impute11(strategies[1])
```

```
Out[382]: ('median', array([6., 6., 6., 6., 6., 6.]))
```

```
In [383]: strategies[2], test_num_impute(strategies[2])
```

```
Out[383]: ('most_frequent', array([40., 40., 40., 40., 40., 40.]))
```

```
In [384]: strategies[2], test_num_impute1(strategies[2])
```

```
Out[384]: ('most_frequent', array([1., 1., 1., 1., 1., 1.]))
```

```
In [385]: # Sentiment_Subjectivity
strategies[2], test_num_impute11(strategies[2])
```

```
Out[385]: ('most_frequent', array([1., 1., 1., 1., 1., 1.]))
```

```
In [386]: # Более сложная функция, которая позволяет задавать колонку и вид и  
мпьютации  
def test_num_impute_col(dataset, column, strategy_param):  
    temp_data = dataset[[column]]  
  
    indicator = MissingIndicator()  
    mask_missing_values_only = indicator.fit_transform(temp_data)  
  
    imp_num = SimpleImputer(strategy=strategy_param)  
    data_num_imp = imp_num.fit_transform(temp_data)  
  
    filled_data = data_num_imp[mask_missing_values_only]  
  
    return column, strategy_param, filled_data.size, filled_data[0]  
    , filled_data[filled_data.size-1]
```

```
In [387]: # Более сложная функция, которая позволяет задавать колонку и вид и  
мпьютации  
def test_num_impute_col1(dataset, column, strategy_param):  
    temp_data = dataset[[column]]  
  
    indicator = MissingIndicator()  
    mask_missing_values_only = indicator.fit_transform(temp_data)  
  
    imp_num = SimpleImputer(strategy=strategy_param)  
    data_num_imp = imp_num.fit_transform(temp_data)  
  
    filled_data = data_num_imp[mask_missing_values_only1]  
  
    return column, strategy_param, filled_data.size, filled_data[0]  
    , filled_data[filled_data.size-1]
```

```
In [388]: # Sentiment_Subjectivity  
# Более сложная функция, которая позволяет задавать колонку и вид и  
мпьютации  
def test_num_impute_col11(dataset, column, strategy_param):  
    temp_data = dataset[[column]]  
  
    indicator = MissingIndicator()  
    mask_missing_values_only = indicator.fit_transform(temp_data)  
  
    imp_num = SimpleImputer(strategy=strategy_param)  
    data_num_imp = imp_num.fit_transform(temp_data)  
  
    filled_data = data_num_imp[mask_missing_values_only11]  
  
    return column, strategy_param, filled_data.size, filled_data[0]  
    , filled_data[filled_data.size-1]
```

```
In [389]: data[['depth']].describe()
```

```
Out[389]:
```

	depth
count	994.000000
mean	311.180080
std	215.647223
min	40.000000
25%	99.000000
50%	246.000000
75%	543.000000
max	680.000000

```
In [390]: data1[['ncontrols']].describe()
```

```
Out[390]:
```

	ncontrols
count	82.000000
mean	11.000000
std	12.364825
min	1.000000
25%	3.000000
50%	6.000000
75%	14.000000
max	60.000000

```
In [391]: test_num_impute_col(data, 'depth', strategies[0])
```

```
Out[391]: ('depth', 'mean', 6, 311.1800804828974, 311.1800804828974)
```

```
In [392]: test_num_impute_col1(data1, 'ncontrols', strategies[0])
```

```
Out[392]: ('ncontrols', 'mean', 6, 11.0, 11.0)
```

```
In [393]: test_num_impute_col(data, 'depth', strategies[1])
```

```
Out[393]: ('depth', 'median', 6, 246.0, 246.0)
```

```
In [394]: test_num_impute_col1(data1, 'ncontrols', strategies[1])
```

```
Out[394]: ('ncontrols', 'median', 6, 6.0, 6.0)
```

```
In [395]: test_num_impute_col(data, 'depth', strategies[2])
```

```
Out[395]: ('depth', 'most_frequent', 6, 40.0, 40.0)
```

```
In [396]: test_num_impute_col1(data1, 'ncontrols', strategies[2])
```

```
Out[396]: ('ncontrols', 'most_frequent', 6, 1.0, 1.0)
```

Обработка пропусков в категориальных данных

```
In [397]: # Выберем категориальные колонки с пропущенными значениями  
# Цикл по колонкам датасета  
  
cat_cols1 = []  
for col in data1.columns:  
    # Количество пустых значений  
    temp_null_count1 = data1[data1[col].isnull()].shape[0]  
    dt1 = str(data1[col].dtype)  
    if temp_null_count1>0 and (dt1=='object'):  
        cat_cols1.append(col)  
        temp_perc1 = round((temp_null_count1 / total_count1) * 100.  
0, 2)  
        print('Колонка {}. Тип данных {}. Количество пустых значени  
й {}, {}%.'.format(col, dt1, temp_null_count1, temp_perc1))
```

```
In [401]: cat_temp_data = data[['depth']]  
cat_temp_data.head()
```

```
Out[401]:
```

	depth
0	562.0
1	650.0
2	42.0
3	626.0
4	649.0

```
In [402]: cat_temp_data1 = data1[['ncontrols']]
cat_temp_data1.head()
```

Out[402]:

	ncontrols
0	40.0
1	10.0
2	6.0
3	5.0
4	27.0

```
In [403]: cat_temp_data['depth'].unique()
```

Out[403]:

```
array([562., 650., 42., 626., 649., 195., 82., 194., 211., 622.,
583.,
249., 554., 600., 139., 306., 50., 590., 570., 598., 576.,
512.,
125., 431., 537., 155., 498., 582., 328., 553., 292., 349.,
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564., 197., 265., 323., 304., 75., nan, 579., 284., 450.,
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117., 538., 123., 69., 128., 236., 497., 271., 224., 375.,
365.,
484., 108., 608., 72., 636., 293., 100., 146., 280., 388.,
477.,
617., 606., 609., 64., 178., 248., 81., 571., 49., 517.,
307.,
189., 527., 63., 510., 624., 53., 199., 149., 210., 658.,
220.,
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597.,
452., 93., 103., 504., 202., 59., 244., 239., 434., 99.,
399.,
216., 544., 542., 339., 640., 67., 161., 534., 45., 309.,
234.,
569., 605., 422., 637., 204., 175., 595., 360., 367., 190.,
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261., 603., 508., 350., 533., 411., 338., 226., 618., 242.,
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90., 130., 65., 397., 505., 71., 207., 154., 232., 106.,
664.,
57., 525., 74., 44., 470., 298., 148., 107., 218., 619.,
150.,
```



```

180., 179., 680., 254., 521., 526., 270., 548., 158., 300.,
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364., 228., 225., 334., 326., 121., 432., 580., 581., 513.,
77.,
315., 567., 560., 266., 231., 262., 331., 558., 268., 193.,
172.,
217., 251., 291.] )

```

```
In [404]: cat_temp_data1['ncontrols'].unique()
```

```
Out[404]: array([40., 10., 6., 5., 27., 7., 4., 2., 1., nan, 60., 14.,
8.,
35., 23., 11., 3., 46., 18., 38., 21., 15., 16., 22., 12.,
17.,
48., 34., 9., 13.] )
```

```
In [405]: cat_temp_data[cat_temp_data['depth'].isnull()].shape
```

```
Out[405]: (6, 1)
```

```
In [406]: # Импутация наиболее частыми значениями
imp2 = SimpleImputer(missing_values=np.nan, strategy='most_frequent')
data_imp2 = imp2.fit_transform(cat_temp_data)
data_imp2
```

```
Out[406]: array([[562.],
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[244.],  
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```

```
In [407]: # Импутация наиболее частыми значениями  
imp2 = SimpleImputer(missing_values=np.nan, strategy='most_frequent'  
)  
data_imp21 = imp2.fit_transform(cat_temp_data1)  
data_imp21
```

```
Out[407]: array([[40.],  
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```

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[ 1.]])
```

```
In [408]: # Импутация наиболее частыми значениями
imp2 = SimpleImputer(missing_values=np.nan, strategy='most_frequent')
data_imp211 = imp2.fit_transform(cat_temp_data11)
data_imp211
```

```
Out[408]: array([[2534.],
```

file:///Users/17713067/Downloads/LAB3.html

```
In [409]: # Пустые значения отсутствуют
np.unique(data_imp2)
```

Page 64 of 107

139.,
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153.,
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641.,
        642., 644., 646., 649., 650., 651., 654., 655., 658., 663.,
664.,
        671., 680.])

```

```

In [410]: # Пустые значения отсутствуют
np.unique(data_imp21)

```

```

Out[410]: array([ 1.,  2.,  3.,  4.,  5.,  6.,  7.,  8.,  9., 10., 11., 12.,
13.,
        14., 15., 16., 17., 18., 21., 22., 23., 27., 34., 35., 38.,
40.,
        46., 48., 60.])

```

```

In [411]: # Пустые значения отсутствуют
np.unique(data_imp211)

```

```

Out[411]: array([ 75., 2534., 3544., 4554., 5564., 6574.])

```

```

In [412]: # Импьютация константой
imp3 = SimpleImputer(missing_values=np.nan, strategy='constant', fi
ll_value=1)
data_imp3 = imp3.fit_transform(cat_temp_data)
data_imp3

```

```

Out[412]: array([[562.],
        [650.],
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        [626.],
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[559.],
[524.],
[69.],
[594.],
[262.],
[538.],
[331.],
[48.],
[47.],
[558.],
[524.],
[545.],
[477.],
[129.],
[268.],
[117.],
[541.],
[112.],
[162.],
[609.],
[76.],
[61.],
[202.],
[90.],
[133.],
[589.],
[190.],
[138.],
[598.],
[600.],

[162.],
[626.],
[137.],
[57.],
[201.],
[69.],
[219.],
[553.],
[524.],
[51.],
[107.],
[44.],
[574.],
[128.],
[568.],
[583.],
[622.],
[193.],
[544.],
[118.],
[51.],
[63.],
[442.],
[87.],
[61.],
[60.],
[561.],
[138.],
[174.],
[543.],
[530.],
[497.],
[63.],
[82.],
[605.],
[234.],
[41.],
[40.],
[137.],
[223.],
[109.],
[595.],
[512.],
[613.],
[60.],
[43.],
[172.],
[54.],
[68.],
[217.],
[102.],
[178.],
[251.],

```
[ 42.],  
[575.],  
[ 43.],  
[577.],  
[ 42.],  
[ 75.],  
[ 71.],  
[ 60.],  
[291.],  
[125.],  
[ 69.],  
[614.],  
[108.],  
[575.],  
[409.],  
[243.],  
[642.],  
[ 45.],  
[470.],  
[248.],  
[244.],  
[ 40.],  
[165.]])
```

```
In [413]: # Импутация константой  
imp3 = SimpleImputer(missing_values=np.nan, strategy='constant', fi  
ll_value=11)  
data_imp31 = imp3.fit_transform(cat_temp_data1)  
data_imp31
```

```
Out[413]: array([[40.],  
[10.],  
[ 6.],  
[ 5.],  
[27.],  
[ 7.],  
[ 4.],  
[ 7.],  
[ 2.],  
[ 1.],  
[ 2.],  
[11.],  
[ 1.],  
[ 1.],  
[ 2.],  
[60.],  
[14.],  
[ 7.],  
[ 8.],  
[35.],  
[23.],  
[14.],  
[11.]])
```

[11.],
[6.],
[2.],
[1.],
[3.],
[3.],
[4.],
[46.],
[18.],
[10.],
[4.],
[38.],
[21.],
[15.],
[7.],
[16.],
[14.],
[5.],
[4.],
[4.],
[4.],
[3.],
[4.],
[11.],
[22.],
[12.],
[6.],
[40.],
[21.],
[17.],
[6.],
[18.],
[15.],
[6.],
[4.],
[10.],
[7.],
[3.],
[6.],
[48.],
[14.],
[7.],
[2.],
[34.],
[11.],
[9.],
[13.],
[12.],
[3.],
[1.],
[4.],
[2.],
[1.],

```
[ 1.],
[18.],
[ 6.],
[ 3.],
[ 5.],
[11.],
[ 3.],
[ 1.],
[ 1.],
[ 1.],
[11.],
[ 1.]])
```

```
In [414]: # ИМПЬЮТАЦИЯ КОНСТАНТОЙ
imp3 = SimpleImputer(missing_values=np.nan, strategy='constant', fi
ll_value=3)
data_imp311 = imp3.fit_transform(cat_temp_data11)
data_imp311
```

[illegible]

Page 88 of 107

[7.500e+01]])

In [415]: np.unique(data_imp3)

```
Out[415]: array([ 1., 40., 41., 42., 43., 44., 45., 46., 47., 48.,
49.,
50., 51., 52., 53., 54., 55., 56., 57., 58., 59.,
60.,
61., 62., 63., 64., 65., 66., 67., 68., 69., 70.,
71.,
72., 74., 75., 76., 77., 78., 79., 80., 81., 82.,
83.,
84., 85., 86., 87., 89., 90., 93., 94., 95., 96.,
97.,
98., 99., 100., 102., 103., 104., 105., 106., 107., 108.,
109.,
111., 112., 116., 117., 118., 119., 121., 123., 124., 125.,
126.,
127., 128., 129., 130., 132., 133., 134., 135., 136., 137.,
138.,
139., 140., 142., 143., 144., 146., 148., 149., 150., 151.,
152.,
153., 154., 155., 156., 158., 161., 162., 164., 165., 166.,
168.,
169., 170., 172., 174., 175., 176., 178., 179., 180., 182.,
183.,
184., 186., 188., 189., 190., 193., 194., 195., 197., 199.,
200.,
201., 202., 203., 204., 205., 206., 207., 208., 209., 210.,
211.,
213., 215., 216., 217., 218., 219., 220., 221., 223., 224.,
225.,
226., 228., 229., 230., 231., 232., 234., 236., 237., 239.,
242.,
243., 244., 246., 248., 249., 251., 254., 255., 257., 259.,
260.,
261., 262., 263., 264., 265., 266., 268., 269., 270., 271.,
273.,
275., 278., 280., 284., 286., 287., 291., 292., 293., 294.,
296.,
297., 298., 299., 300., 302., 304., 306., 307., 309., 315.,
323.,
325., 326., 328., 329., 331., 332., 334., 338., 339., 342.,
343.,
348., 349., 350., 356., 360., 361., 364., 365., 367., 375.,
376.,
377., 383., 384., 385., 388., 390., 391., 393., 397., 399.,
401.,
402., 403., 405., 406., 409., 411., 413., 417., 420., 422.,
423.,
431., 432., 434., 440., 442., 445., 450., 452., 460., 462.,
464.,
467., 470., 474., 475., 476., 477., 479., 480., 481., 482.,
```

```

483.,
    484., 485., 487., 488., 489., 490., 491., 492., 493., 497.,
498.,
    499., 500., 501., 502., 504., 505., 506., 507., 508., 510.,
511.,
    512., 513., 515., 517., 518., 520., 521., 522., 524., 525.,
526.,
    527., 528., 529., 530., 532., 533., 534., 535., 536., 537.,
538.,
    539., 541., 542., 543., 544., 545., 546., 548., 549., 550.,
553.,
    554., 555., 556., 557., 558., 559., 560., 561., 562., 563.,
564.,
    565., 566., 567., 568., 569., 570., 571., 572., 573., 574.,
575.,
    576., 577., 578., 579., 580., 581., 582., 583., 584., 585.,
586.,
    587., 589., 590., 591., 592., 593., 594., 595., 597., 598.,
599.,
    600., 601., 602., 603., 604., 605., 606., 607., 608., 609.,
611.,
    613., 614., 615., 616., 617., 618., 619., 622., 624., 625.,
626.,
    627., 628., 629., 630., 631., 632., 636., 637., 638., 639.,
640.,
    641., 642., 644., 646., 649., 650., 651., 654., 655., 658.,
663.,
    664., 671., 680.])

```

```
In [416]: np.unique(data_imp311)
```

```
Out[416]: array([3.000e+00, 7.500e+01, 2.534e+03, 3.544e+03, 4.554e+03, 5.56
4e+03,
        6.574e+03])
```

```
In [417]: data_imp3[data_imp3==1].size
```

```
Out[417]: 6
```

```
In [418]: data_imp31[data_imp31==2].size
```

```
Out[418]: 6
```

```
In [419]: data_imp311[data_imp311==3].size
```

```
Out[419]: 7
```

```
In [420]: data.shape
```

```
Out[420]: (1000, 6)
```

```
In [421]: data1.shape
```

```
Out[421]: (88, 6)
```

Преобразование категориальных признаков в числовые

```
In [422]: cat_enc = pd.DataFrame({'c1':data_imp2.T[0]})  
cat_enc
```

```
Out[422]:
```

	c1
0	562.0
1	650.0
2	42.0
3	626.0
4	649.0
...	...
995	470.0
996	248.0
997	244.0
998	40.0
999	165.0

1000 rows × 1 columns

```
In [423]: cat_enc1 = pd.DataFrame({'c1':data_imp21.T[0]})  
cat_enc1
```

Out[423]:

	c1
0	40.0
1	10.0
2	6.0
3	5.0
4	27.0
...	...
83	1.0
84	1.0
85	1.0
86	1.0
87	1.0

88 rows × 1 columns

```
In [424]: cat_enc11 = pd.DataFrame({'c1':data_imp211.T[0]})  
cat_enc11
```

Out[424]:

	c1
0	2534.0
1	2534.0
2	2534.0
3	2534.0
4	2534.0
...	...
83	75.0
84	75.0
85	75.0
86	75.0
87	75.0

88 rows × 1 columns

Кодирование категорий целочисленными значениями

```
In [425]: from sklearn.preprocessing import LabelEncoder, OneHotEncoder
```

```
In [426]: le = LabelEncoder()
cat_enc_le = le.fit_transform(cat_enc['c1'])
```

```
In [427]: le1 = LabelEncoder()
cat_enc_le1 = le1.fit_transform(cat_enc1['c1'])
```

```
In [428]: le11 = LabelEncoder()
cat_enc_le11 = le11.fit_transform(cat_enc11['c1'])
```

```
In [429]: cat_enc['c1'].unique()
```

```
Out[429]: array([562., 650., 42., 626., 649., 195., 82., 194., 211., 622.,
583.,
249., 554., 600., 139., 306., 50., 590., 570., 598., 576.,
512.,
125., 431., 537., 155., 498., 582., 328., 553., 292., 349.,
48.,
206., 574., 585., 230., 263., 96., 511., 94., 246., 56.,
329.,
70., 493., 129., 223., 46., 593., 489., 445., 584., 535.,
530.,
260., 613., 84., 286., 587., 627., 40., 152., 201., 506.,
546.,
564., 197., 265., 323., 304., 75., 579., 284., 450., 170.,
117.,
538., 123., 69., 128., 236., 497., 271., 224., 375., 365.,
484.,
108., 608., 72., 636., 293., 100., 146., 280., 388., 477.,
617.,
606., 609., 64., 178., 248., 81., 571., 49., 517., 307.,
189.,
527., 63., 510., 624., 53., 199., 149., 210., 658., 220.,
205.,
614., 186., 97., 462., 573., 127., 229., 112., 140., 597.,
452.,
93., 103., 504., 202., 59., 244., 239., 434., 99., 399.,
216.,
544., 542., 339., 640., 67., 161., 534., 45., 309., 234.,
569.,
605., 422., 637., 204., 175., 595., 360., 367., 190., 629.,
261.,
603., 508., 350., 533., 411., 338., 226., 618., 242., 342.,
90.,
130., 65., 397., 505., 71., 207., 154., 232., 106., 664.,
57.,
```

```

525., 74., 44., 470., 298., 148., 107., 218., 619., 150.,
180.,
179., 680., 254., 521., 526., 270., 548., 158., 300., 482.,
607.,
105., 577., 529., 528., 492., 561., 413., 565., 138., 383.,
522.,
671., 572., 641., 507., 601., 654., 126., 555., 500., 515.,
501.,
55., 644., 442., 464., 200., 479., 325., 575., 483., 118.,
83.,
61., 219., 68., 43., 80., 51., 54., 403., 60., 406.,
221.,
502., 423., 536., 630., 153., 188., 124., 401., 102., 556.,
417.,
591., 646., 52., 41., 109., 475., 66., 481., 151., 47.,
119.,
176., 602., 488., 343., 563., 259., 476., 499., 257., 165.,
136.,
524., 467., 184., 237., 162., 604., 639., 628., 632., 215.,
135.,
297., 568., 168., 269., 143., 95., 142., 104., 169., 474.,
294.,
594., 638., 520., 384., 62., 203., 132., 543., 589., 485.,
58.,
541., 144., 460., 137., 586., 213., 393., 296., 549., 85.,
98.,
89., 76., 273., 264., 174., 420., 559., 405., 599., 480.,
566.,
611., 409., 209., 134., 243., 615., 377., 278., 550., 518.,
116.,
491., 376., 332., 182., 79., 164., 651., 642., 390., 539.,
631.,
299., 255., 616., 655., 356., 385., 208., 545., 487., 183.,
166.,
133., 86., 287., 348., 578., 361., 275., 78., 302., 440.,
156.,
391., 592., 87., 490., 663., 625., 557., 402., 532., 111.,
364.,
228., 225., 334., 326., 121., 432., 580., 581., 513., 77.,
315.,
567., 560., 266., 231., 262., 331., 558., 268., 193., 172.,
217.,
251., 291.])

```

```
In [430]: cat_encl['c1'].unique()
```

```
Out[430]: array([40., 10., 6., 5., 27., 7., 4., 2., 1., 60., 14., 8.,
35.,
23., 11., 3., 46., 18., 38., 21., 15., 16., 22., 12., 17.,
48.,
34., 9., 13.])
```

```
In [431]: cat_enc11['c1'].unique()
```

```
Out[431]: array([2534., 3544., 4554., 5564., 6574., 75.])
```

```
In [432]: np.unique(cat_enc_le)
```

```
Out[432]: array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12,
                13, 14, 15, 16, 17, 18, 19, 20, 21, 22, 23, 24,
                25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35, 36, 37,
                38, 39, 40, 41, 42, 43, 44, 45, 46, 47, 48, 49, 50,
                51, 52, 53, 54, 55, 56, 57, 58, 59, 60, 61, 62, 63,
                64, 65, 66, 67, 68, 69, 70, 71, 72, 73, 74, 75, 76,
                77, 78, 79, 80, 81, 82, 83, 84, 85, 86, 87, 88, 89,
                90, 91, 92, 93, 94, 95, 96, 97, 98, 99, 100, 101, 102,
                103, 104, 105, 106, 107, 108, 109, 110, 111, 112, 113, 114, 115,
                116, 117, 118, 119, 120, 121, 122, 123, 124, 125, 126, 127, 128,
                129, 130, 131, 132, 133, 134, 135, 136, 137, 138, 139, 140, 141,
                142, 143, 144, 145, 146, 147, 148, 149, 150, 151, 152, 153, 154,
                155, 156, 157, 158, 159, 160, 161, 162, 163, 164, 165, 166, 167,
                168, 169, 170, 171, 172, 173, 174, 175, 176, 177, 178, 179, 180,
                181, 182, 183, 184, 185, 186, 187, 188, 189, 190, 191, 192, 193,
                194, 195, 196, 197, 198, 199, 200, 201, 202, 203, 204, 205, 206,
                207, 208, 209, 210, 211, 212, 213, 214, 215, 216, 217, 218, 219,
                220, 221, 222, 223, 224, 225, 226, 227, 228, 229, 230, 231, 232,
                233, 234, 235, 236, 237, 238, 239, 240, 241, 242, 243, 244, 245,
                246, 247, 248, 249, 250, 251, 252, 253, 254, 255, 256, 257, 258,
                259, 260, 261, 262, 263, 264, 265, 266, 267, 268, 269, 270, 271,
                272, 273, 274, 275, 276, 277, 278, 279, 280, 281, 282, 283, 284,
                285, 286, 287, 288, 289, 290, 291, 292, 293, 294, 295, 296, 297,
                298, 299, 300, 301, 302, 303, 304, 305, 306, 307, 308, 309, 310,
```

```

311,
    312, 313, 314, 315, 316, 317, 318, 319, 320, 321, 322, 323,
324,
    325, 326, 327, 328, 329, 330, 331, 332, 333, 334, 335, 336,
337,
    338, 339, 340, 341, 342, 343, 344, 345, 346, 347, 348, 349,
350,
    351, 352, 353, 354, 355, 356, 357, 358, 359, 360, 361, 362,
363,
    364, 365, 366, 367, 368, 369, 370, 371, 372, 373, 374, 375,
376,
    377, 378, 379, 380, 381, 382, 383, 384, 385, 386, 387, 388,
389,
    390, 391, 392, 393, 394, 395, 396, 397, 398, 399, 400, 401,
402,
    403, 404, 405, 406, 407, 408, 409, 410, 411, 412, 413, 414,
415,
    416, 417, 418, 419])

```

```
In [433]: np.unique(cat_enc_le1)
```

```
Out[433]: array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14,
                15, 16,
                17, 18, 19, 20, 21, 22, 23, 24, 25, 26, 27, 28])
```

```
In [434]: np.unique(cat_enc_le11)
```

```
Out[434]: array([0, 1, 2, 3, 4, 5])
```

```
In [435]: le.inverse_transform([0, 1, 2])
```

```
Out[435]: array([40., 41., 42.])
```

```
In [436]: le1.inverse_transform([0, 1, 2])
```

```
Out[436]: array([1., 2., 3.])
```

```
In [437]: le11.inverse_transform([0, 1, 2])
```

```
Out[437]: array([ 75., 2534., 3544.])
```

Кодирование категорий наборами бинарных значений

```
In [438]: ohe = OneHotEncoder()
          cat_enc_ohe = ohe.fit_transform(cat_enc[['c1']])
```

```
In [439]: ohe1 = OneHotEncoder()
          cat_enc_ohe1 = ohe1.fit_transform(cat_enc1[['c1']])
```



```
In [440]: ohell = OneHotEncoder()  
cat_enc_ohell = ohell.fit_transform(cat_enc11[['c1']])
```

```
In [441]: cat_enc.shape
```

```
Out[441]: (1000, 1)
```

```
In [442]: cat_enc1.shape
```

```
Out[442]: (88, 1)
```

```
In [443]: cat_enc11.shape
```

```
Out[443]: (88, 1)
```

```
In [444]: cat_enc_ohel.shape
```

```
Out[444]: (1000, 420)
```

```
In [445]: cat_enc_ohel1.shape
```

```
Out[445]: (88, 29)
```

```
In [446]: cat_enc_ohell1.shape
```

```
Out[446]: (88, 6)
```

```
In [447]: cat_enc_ohel
```

```
Out[447]: <1000x420 sparse matrix of type '<class 'numpy.float64'>'  
          with 1000 stored elements in Compressed Sparse Row format>
```

```
In [448]: cat_enc_ohel1
```

```
Out[448]: <88x29 sparse matrix of type '<class 'numpy.float64'>'  
          with 88 stored elements in Compressed Sparse Row format>
```

```
In [449]: cat_enc_ohel
```

```
Out[449]: <88x29 sparse matrix of type '<class 'numpy.float64'>'  
          with 88 stored elements in Compressed Sparse Row format>
```

```
cat_enc_ohe.todense()[0:10]
```

```
matrix([ [0., 0., 0., ..., 0., 0., 0.],
         [0., 0., 0., ..., 0., 0., 0.],
         [0., 0., 1., ..., 0., 0., 0.],
         ...,
         [0., 0., 0., ..., 0., 0., 0.],
         [0., 0., 0., ..., 0., 0., 0.],
         [0., 0., 0., ..., 0., 0., 0.] ])
```

```
cat_enc_ohe1[:45000].todense()[0:10]
```

[illegible]

```
In [452]: cat_enc_ohe11[:45000].todense()[0:10]
```

```
Out[452]: matrix([[0., 1., 0., 0., 0., 0.],
                  [0., 1., 0., 0., 0., 0.],
                  [0., 1., 0., 0., 0., 0.],
                  [0., 1., 0., 0., 0., 0.],
                  [0., 1., 0., 0., 0., 0.],
                  [0., 1., 0., 0., 0., 0.],
                  [0., 0., 1., 0., 0., 0.],
                  [0., 0., 1., 0., 0., 0.],
                  [0., 1., 0., 0., 0., 0.],
                  [0., 1., 0., 0., 0., 0.]])
```

```
In [453]: cat_enc.head(10)
```

```
Out[453]:
```

	c1
0	562.0
1	650.0
2	42.0
3	626.0
4	649.0
5	195.0
6	82.0
7	194.0
8	211.0
9	622.0

```
In [454]: cat_enc1.head(10)
```

```
Out[454]:
```

	c1
0	40.0
1	10.0
2	6.0
3	5.0
4	27.0
5	7.0
6	4.0
7	7.0
8	2.0
9	1.0

```
In [455]: cat_enc11.head(10)
```

```
Out[455]:
```

	c1
0	2534.0
1	2534.0
2	2534.0
3	2534.0
4	2534.0
5	2534.0
6	3544.0
7	3544.0
8	2534.0
9	2534.0

Масштабирование данных

```
In [456]: from sklearn.preprocessing import MinMaxScaler, StandardScaler, Normalizer
```

MinMax

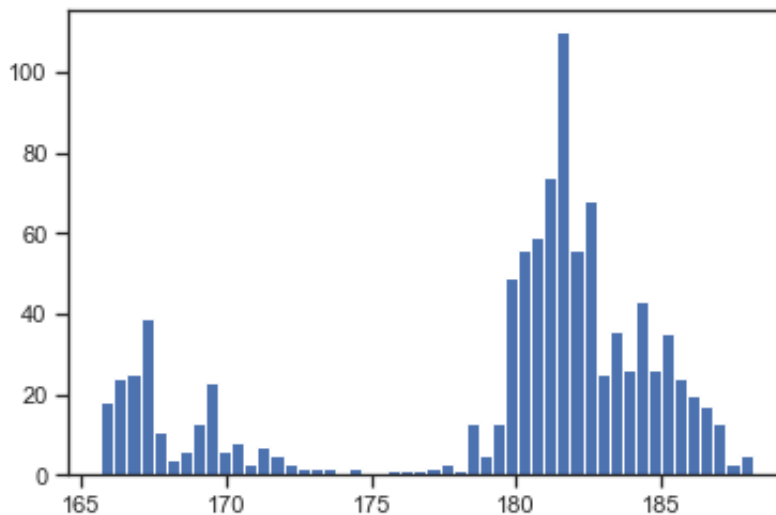
```
In [457]: # data = pd.read_csv('googleplaystore.csv', sep=",")
strategies[0], test_num_impute(strategies[0])
sc1 = MinMaxScaler()
sc1_data = sc1.fit_transform(data[['depth']])
```

```
In [458]: strategies[0], test_num_impute1(strategies[0])
sc11 = MinMaxScaler()
sc1_data1 = sc11.fit_transform(data1[['agegp']])
```

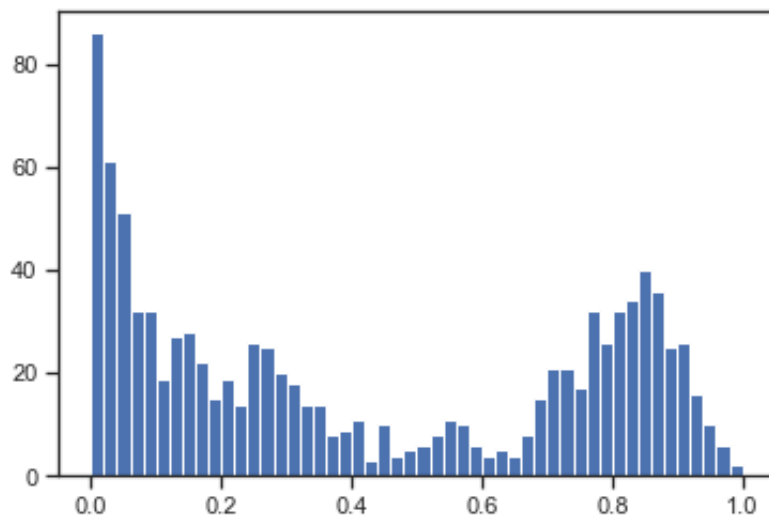
```
In [459]: strategies[0], test_num_impute1(strategies[0])
sc111 = MinMaxScaler()
sc1_data11 = sc111.fit_transform(data1[['ncases']])
```

```
In [460]: plt.hist(data['long'], 50)
plt.show()
```

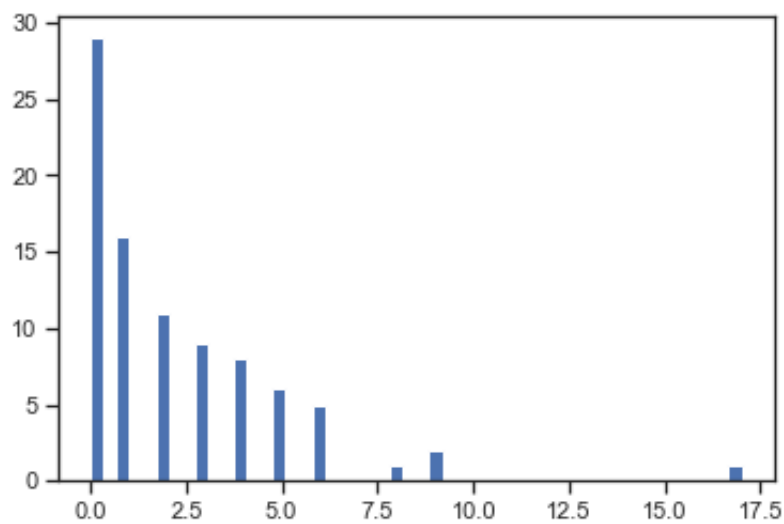
```
/usr/local/lib/python3.7/site-packages/numpy/lib/histograms.py:839
: RuntimeWarning: invalid value encountered in greater_equal
  keep = (tmp_a >= first_edge)
/usr/local/lib/python3.7/site-packages/numpy/lib/histograms.py:840
: RuntimeWarning: invalid value encountered in less_equal
  keep &= (tmp_a <= last_edge)
```



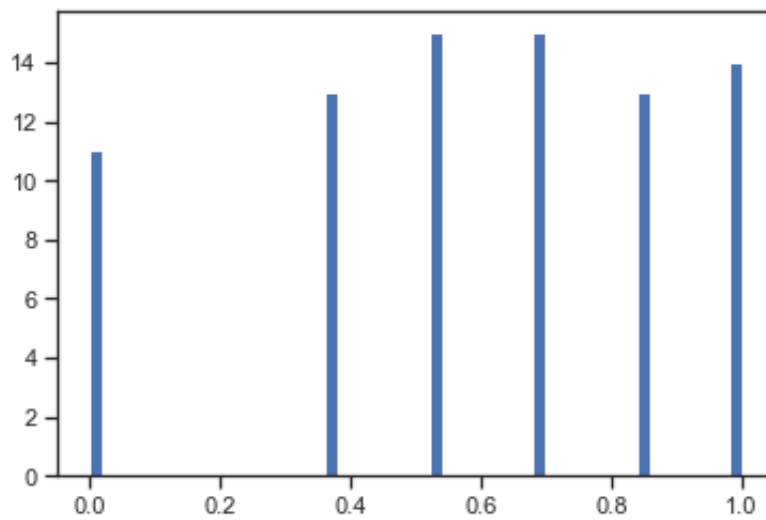
```
In [461]: plt.hist(scl_data, 50)  
plt.show()
```



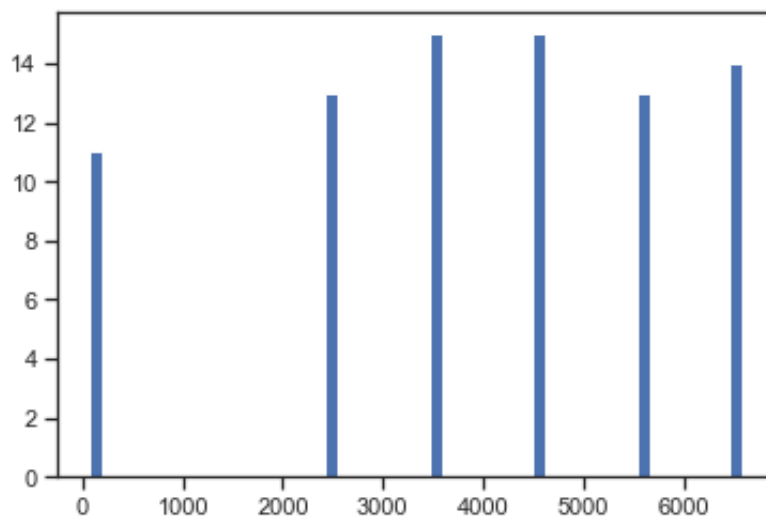
```
In [462]: plt.hist(data1['ncases'], 50)  
plt.show()
```



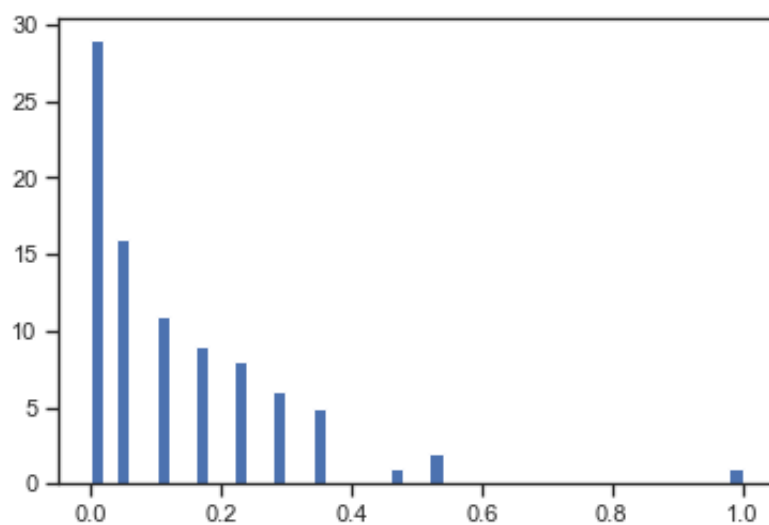
```
In [463]: plt.hist(scl_data1, 50)  
plt.show()
```



```
In [464]: plt.hist(data1['agegp'], 50)  
plt.show()
```



```
In [465]: plt.hist(sc1_data11, 50)
plt.show()
```



Z-оценка

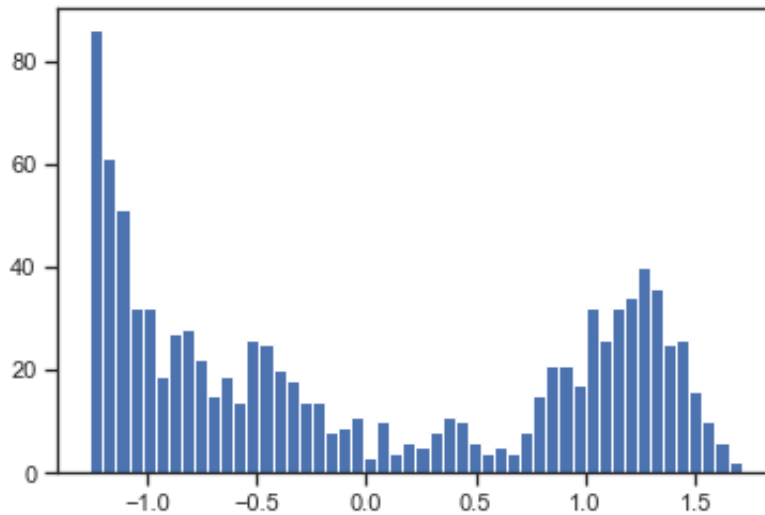
```
In [466]: sc2 = StandardScaler()
sc2_data = sc2.fit_transform(data[['depth']])
```

```
In [467]: sc21 = StandardScaler()
sc2_data1 = sc21.fit_transform(data1[['ncontrols']])
```

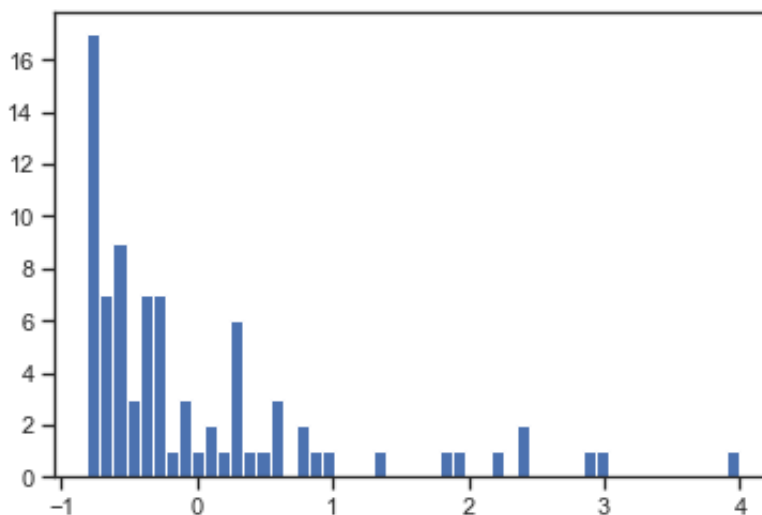


```
In [468]: plt.hist(sc2_data, 50)
plt.show()
```

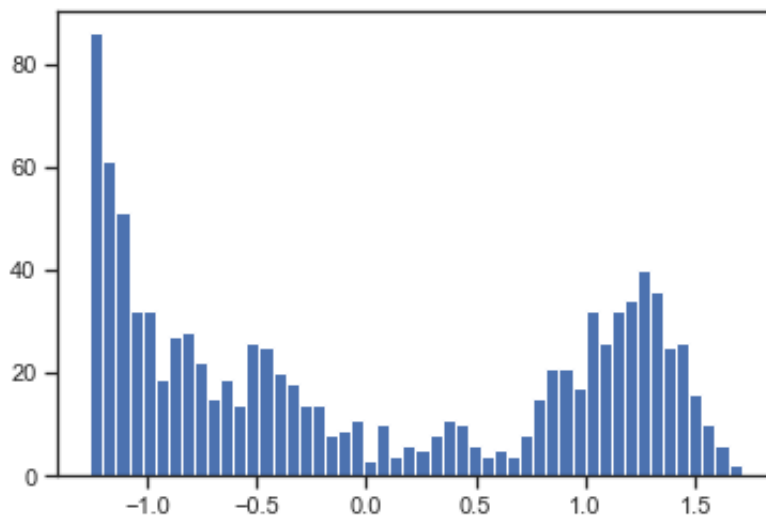
```
/usr/local/lib/python3.7/site-packages/numpy/lib/histograms.py:839
: RuntimeWarning: invalid value encountered in greater_equal
  keep = (tmp_a >= first_edge)
/usr/local/lib/python3.7/site-packages/numpy/lib/histograms.py:840
: RuntimeWarning: invalid value encountered in less_equal
  keep &= (tmp_a <= last_edge)
```



```
In [469]: plt.hist(sc2_data1, 50)
plt.show()
```



```
In [470]: plt.hist(sc2_data, 50)  
plt.show()
```

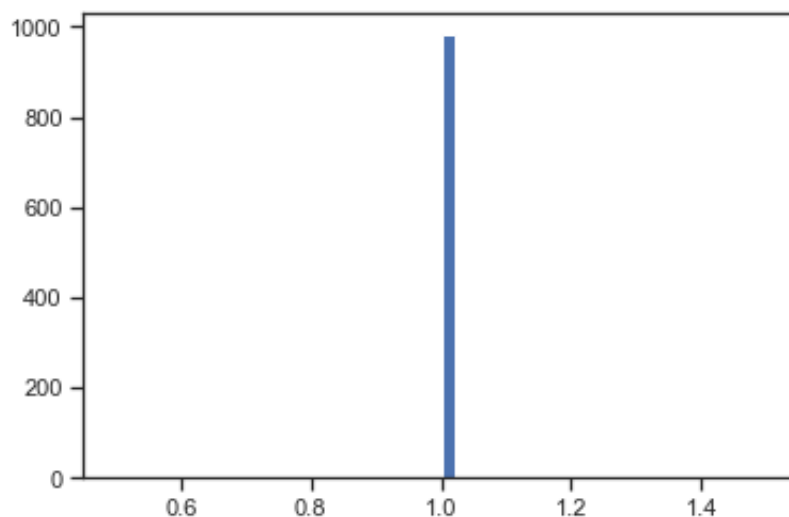


Нормализация

```
In [471]: sc3 = Normalizer()  
sc3_data = sc3.fit_transform(data_new_2[['depth']])
```

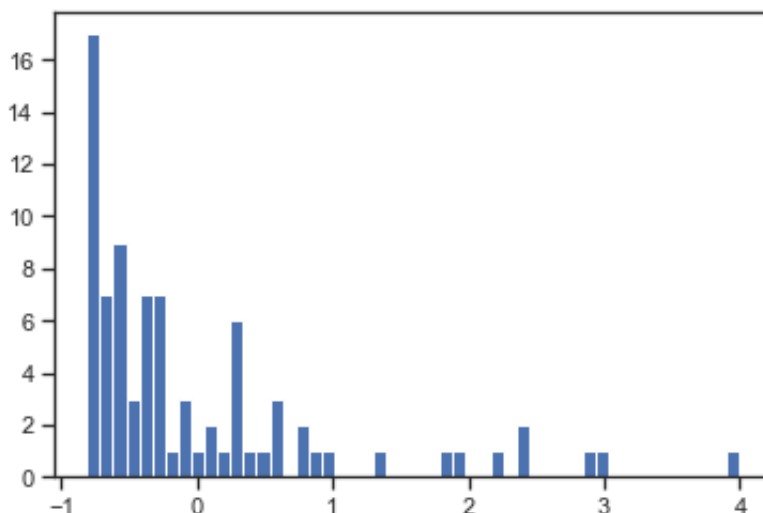
```
In [472]: sc31 = StandardScaler()  
sc3_data1 = sc31.fit_transform(data1[['ncontrols']])
```

```
In [473]: plt.hist(sc3_data, 50)  
plt.show()
```



```
In [474]: plt.hist(sc3_data1, 50)  
plt.show()
```

```
/usr/local/lib/python3.7/site-packages/numpy/lib/histograms.py:839  
: RuntimeWarning: invalid value encountered in greater_equal  
  keep = (tmp_a >= first_edge)  
/usr/local/lib/python3.7/site-packages/numpy/lib/histograms.py:840  
: RuntimeWarning: invalid value encountered in less_equal  
  keep &= (tmp_a <= last_edge)
```



Вывод:

В процессе выполнения данной работы были изучены методы обработки пропусков в данных, кодирования категориальных признаков и масштабирования данных.

```
In [ ]:
```