# RK1 Андрианов А.А. ИУ5-23М

### Вариант Задачи №1 - 1

Для набора данных проведите кодирование одного (произвольного) категориального признака с использованием метода "count (frequency) encoding".

#### Вариант Задачи №2 - 21

Для набора данных проведите масштабирование данных для одного (произвольного) числового признака с использованием масштабирования по медиане.

#### Дополнительные требования по группе:

Для студентов групп ИУ5-23M, ИУ5И-23M - для произвольной колонки данных построить график "Ящик с усами (boxplot)".

## ▼ Импорт необходимых библиотек

```
1 import numpy as np
 2 import matplotlib.pyplot as plt
 3 import seaborn as sns
 4 import pandas as pd
5 import plotly.express as px
 6 import seaborn as sns
7 from matplotlib.pyplot import figure
8 from IPython.display import Image
9 !pip install category_encoders
10 from category_encoders.count import CountEncoder as ce_CountEncoder
     Looking in indexes: <a href="https://pypi.org/simple">https://us-python.pkg.dev/colab-wheels/public/simple/</a>
     Collecting category_encoders
       Downloading category_encoders-2.5.0-py2.py3-none-any.whl (69 kB)
                                           69 kB 3.7 MB/s
     Requirement already satisfied: patsy>=0.5.1 in /usr/local/lib/python3.7/dist-packages (from category_encoders) (0.5.2)
     Requirement already satisfied: scipy>=1.0.0 in /usr/local/lib/python3.7/dist-packages (from category_encoders) (1.4.1)
     Requirement already satisfied: statsmodels>=0.9.0 in /usr/local/lib/python3.7/dist-packages (from category_encoders) (0.10.2)
     Requirement already satisfied: numpy>=1.14.0 in /usr/local/lib/python3.7/dist-packages (from category_encoders) (1.21.6)
     Requirement already satisfied: pandas>=1.0.5 in /usr/local/lib/python3.7/dist-packages (from category_encoders) (1.3.5)
     Requirement already satisfied: scikit-learn>=0.20.0 in /usr/local/lib/python3.7/dist-packages (from category_encoders) (1.0.2)
     Requirement already satisfied: pytz>=2017.3 in /usr/local/lib/python3.7/dist-packages (from pandas>=1.0.5->category_encoders) (
     Requirement already satisfied: python-dateutil>=2.7.3 in /usr/local/lib/python3.7/dist-packages (from pandas>=1.0.5->category_e
     Requirement already satisfied: six in /usr/local/lib/python3.7/dist-packages (from patsy>=0.5.1->category_encoders) (1.15.0)
     Requirement already satisfied: joblib>=0.11 in /usr/local/lib/python3.7/dist-packages (from scikit-learn>=0.20.0->category_enco
     Requirement already satisfied: threadpoolctl>=2.0.0 in /usr/local/lib/python3.7/dist-packages (from scikit-learn>=0.20.0->categ
     Installing collected packages: category-encoders
     Successfully installed category-encoders-2.5.0
     /usr/local/lib/python3.7/dist-packages/statsmodels/tools/_testing.py:19: FutureWarning: pandas.util.testing is deprecated. Use
       import pandas.util.testing as tm
    4
```

### Задача 1 (№1)

Для набора данных проведите кодирование одного (произвольного) категориального признака с использованием метода "count (frequency) encoding".

```
1 # Подключение к gogle диску
2 data = pd.read_csv('data.csv', sep=",")
3 data.head()
```

```
type_school school_accreditation gender interest residence parent_age
                                                                 Less
      0
             Academic
                                                                            Urban
                                                                                              56
                                                     Male
                                                            Interested
                                                                 Less
      1
             Academic
                                                     Male
                                                                            Urban
                                                                                              57
                                                            Interested
                                                           Very
      2
             Academic
                                               B Female
                                                                            Urban
                                                                                              50
 1 data_features = list(zip(
 2 # признаки
 3 [i for i in data.columns],
 4 zip(
       # типы колонок
 6
       [str(i) for i in data.dtypes],
 7
       # проверим есть ли пропущенные значения
 8
       [i for i in data.isnull().sum()]
 9)))
10 # Признаки с типом данных и количеством пропусков
11 data_features
     [('type_school', ('object', 0)),
  ('school_accreditation', ('object', 0)),
      ('gender', ('object', 0)),
('interest', ('object', 0)),
('residence', ('object', 0)),
('parent_age', ('int64', 0)),
      ('parent_salary', ('int64', 0)), ('house_area', ('float64', 0)),
       ('average_grades', ('float64', 0)),
       ('parent_was_in_college', ('bool', 0)),
      ('in_college', ('bool', 0))]
 1 data.isnull().sum()
     type_school
     school\_accreditation
     gender
                                  0
     interest
     residence
                                  0
     parent_age
                                  0
     parent_salary
     house_area
                                  0
     average_grades
                                  0
     parent_was_in_college
                                  0
     {\tt in\_college}
                                  0
     dtype: int64
 1 ce_CountEncoder1 = ce_CountEncoder()
 2 data_COUNT_ENC = ce_CountEncoder1.fit_transform(data[data.columns.difference(['in_college'])])
```

#### 1 data\_COUNT\_ENC

	average_grades	gender	house_area	interest	parent_age	parent_salary	pare
0	84.09	515	83.0	229	56	6950000	
1	86.91	515	76.8	229	57	4410000	
2	87.43	485	80.6	324	50	6500000	
3	82.12	515	78.2	324	49	6600000	
4	86.79	485	75.1	324	57	5250000	
995	85.99	485	63.6	324	49	7420000	
996	89.72	485	84.3	229	51	7480000	
997	79.56	515	75.2	229	49	5550000	
998	87.18	515	105.8	261	53	5840000	
999	86.13	515	69.1	100	50	2940000	
1000 rows × 10 columns							
4							<b>&gt;</b>

```
1 data['gender'].unique()
   array(['Male', 'Female'], dtype=object)
```

```
1 data_COUNT_ENC['gender'].unique()
    array([515, 485])

1 ce_CountEncoder2 = ce_CountEncoder(normalize=True)
2 data_FREQ_ENC = ce_CountEncoder2.fit_transform(data[data.columns.difference(['in_college'])])
```

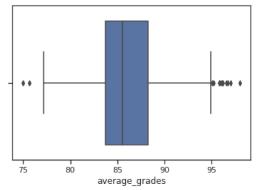
1 data\_FREQ\_ENC

	average_grades	gender	house_area	interest	parent_age	parent_salary	pare
0	84.09	0.515	83.0	0.229	56	6950000	
1	86.91	0.515	76.8	0.229	57	4410000	
2	87.43	0.485	80.6	0.324	50	6500000	
3	82.12	0.515	78.2	0.324	49	6600000	
4	86.79	0.485	75.1	0.324	57	5250000	
995	85.99	0.485	63.6	0.324	49	7420000	
996	89.72	0.485	84.3	0.229	51	7480000	
997	79.56	0.515	75.2	0.229	49	5550000	
998	87.18	0.515	105.8	0.261	53	5840000	
999	86.13	0.515	69.1	0.100	50	2940000	
1000 rows × 10 columns							
4							•

## ▼ Диаграмма "Ящик с усами"

```
1 #строим·график·"ящик·с·усами"
2 sns.boxplot(x=data["average_grades"])
```

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f33c440fa10>



## ▼ Задача 2 (№21)

Для набора данных проведите масштабирование данных для одного (произвольного) числового признака с использованием масштабирования по медиане.

```
1 import pandas as pd
2 import matplotlib.pyplot as plt
3 import numpy as np
4 import seaborn as sns
5 import scipy.stats as stats
6 import datetime
7 from sklearn.preprocessing import RobustScaler
8 from sklearn.preprocessing import MaxAbsScaler
9 from sklearn.preprocessing import MinMaxScaler
10 from sklearn.neighbors import KNeighborsClassifier
```

- 11 from sklearn.feature\_selection import SelectFromModel
- 12 from sklearn.linear\_model import LogisticRegression
- 13 from sklearn.feature\_selection import VarianceThreshold
- 14 import joblib
- 15 # from mlxtend.feature\_selection import SequentialFeatureSelector as SFS
- 16 from category\_encoders.one\_hot import OneHotEncoder as ce\_OneHotEncoder

#### 1 data.head()

	type_school	$school\_accreditation$	gender	interest	residence	parent_age	pai
0	Academic	А	Male	Less Interested	Urban	56	
1	Academic	А	Male	Less Interested	Urban	57	
2	Academic	В	Female	Very Interested	Urban	50	
3	Vocational	В	Male	Very Interested	Rural	49	
4	Academic	А	Female	Very Interested	Urban	57	
1							
4							-

#### 1 data.describe()

	parent_age	parent_salary	house_area	average_grades
cou	nt 1000.000000	1.000000e+03	1000.000000	1000.000000
mea	an 52.208000	5.381570e+06	74.515300	86.097200
sto	3.500427	1.397546e+06	15.293346	3.378738
mi	n 40.000000	1.000000e+06	20.000000	75.000000
25%	% 50.000000	4.360000e+06	64.600000	83.737500
50%	% 52.000000	5.440000e+06	75.500000	85.575000
75%	% 54.000000	6.382500e+06	84.825000	88.262500
ma	x 65.000000	1.000000e+07	120.000000	98.000000

```
1 data_new2 = data[["parent_age", "parent_salary", "house_area", "average_grades"]]
```

```
1 def arr_to_df(arr_scaled):
2     res = pd.DataFrame(arr_scaled, columns=data_new2.columns)
3     return res

1 def draw_graph(col_list, data1, data2, label1, label2):
2     fig, (ax1, ax2) = plt.subplots(ncols = 2, figsize=(20,6))
3     ax1.set_title(label1)
4     sns.kdeplot(data=data1[col_list], ax=ax1)
5     ax2.set_title(label2)
6     sns.kdeplot(data=data2[col_list], ax=ax2)
7     plt.show()

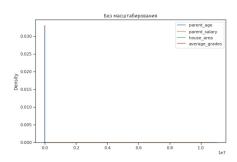
1 rs = RobustScaler()
2 data_median_scale_arr = rs.fit_transform(data_new2)
```

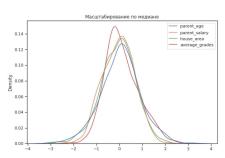
3 data\_median\_scale = arr\_to\_df(data\_median\_scale\_arr)

4 data\_median\_scale.describe()

	parent_age	parent_salary	house_area	average_grades	7
count	1000.000000	1000.000000	1000.000000	1.000000e+03	
mean	0.052000	-0.028890	-0.048687	1.154033e-01	

1 draw\_graph(["parent\_age", "parent\_salary", "house\_area", "average\_grades"], data, data\_median\_scale,'Без масштабирования', 'Масшта





# ▼ Диаграмма "Ящик с усами"

- 1 #строим график "ящик с усами"
- 2 sns.boxplot(x=data\_median\_scale["parent\_salary"])

<matplotlib.axes.\_subplots.AxesSubplot at 0x7f33c44c5750>

