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ФАКУЛЬТЕТ Информатика и системы управления

КАФЕДРА Системы обработки информации и управления

## Отчёт по лабораторной работе №2

По дисциплине:  
«Технологии машинного обучения»

Выполнил:

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## Задание

1. 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.)
2. Для выбранного датасета (датасетов) на основе материалов [лекции](#) решить следующие задачи:
  - обработку пропусков в данных;
  - кодирование категориальных признаков;
  - масштабирование данных.

## ЛР №2

### Импорт библиотек

```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
from pandas.plotting import scatter_matrix
import warnings
warnings.filterwarnings('ignore')
sns.set(style="ticks")
%matplotlib inline
```

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

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

```
Out[3]:
```

	country	iso_code	date	total_vaccinations	people_vaccinated	people_fully_vaccinated
0	Albania	ALB	2021-01-10	0.0	0.0	NaN
1	Albania	ALB	2021-01-11	NaN	NaN	NaN
2	Albania	ALB	2021-01-12	128.0	128.0	NaN
3	Albania	ALB	2021-01-13	188.0	188.0	NaN
4	Albania	ALB	2021-01-14	266.0	266.0	NaN

```
In [4]: data.dtypes
```

```
Out[4]: country                object
iso_code                      object
date                          object
total_vaccinations            float64
people_vaccinated              float64
people_fully_vaccinated        float64
daily_vaccinations_raw         float64
daily_vaccinations             float64
total_vaccinations_per_hundred float64
people_vaccinated_per_hundred  float64
people_fully_vaccinated_per_hundred float64
daily_vaccinations_per_million float64
vaccines                       object
source_name                    object
source_website                  object
dtype: object
```

```
In [5]: data.isnull().sum()
# проверим есть ли пропущенные значения
```

```
Out[5]: country          0
iso_code          272
date              0
total_vaccinations 1214
people_vaccinated 1615
people_fully_vaccinated 2277
daily_vaccinations_raw 1583
daily_vaccinations 135
total_vaccinations_per_hundred 1214
people_vaccinated_per_hundred 1615
people_fully_vaccinated_per_hundred 2277
daily_vaccinations_per_million 135
vaccines          0
source_name       0
source_website    0
dtype: int64
```

```
In [6]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3555 entries, 0 to 3554
Data columns (total 15 columns):
#   Column                                  Non-Null Count  Dtype
---  ---
0   country                                3555 non-null   object
1   iso_code                              3283 non-null   object
2   date                                  3555 non-null   object
3   total_vaccinations                    2341 non-null   float64
4   people_vaccinated                     1940 non-null   float64
5   people_fully_vaccinated                1278 non-null   float64
6   daily_vaccinations_raw                 1972 non-null   float64
7   daily_vaccinations                    3420 non-null   float64
8   total_vaccinations_per_hundred         2341 non-null   float64
9   people_vaccinated_per_hundred          1940 non-null   float64
10  people_fully_vaccinated_per_hundred     1278 non-null   float64
11  daily_vaccinations_per_million          3420 non-null   float64
12  vaccines                               3555 non-null   object
13  source_name                            3555 non-null   object
14  source_website                         3555 non-null   object
dtypes: float64(9), object(6)
memory usage: 416.7+ KB
```

## Обработка пропусков

```
In [7]: # Удаляем столбцы, которые не несут значимой информации
data.drop(['source_name', 'source_website'], axis = 1, inplace = True)
```

```
In [8]: data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3555 entries, 0 to 3554
Data columns (total 13 columns):
#   Column                                     Non-Null Count  Dtype
---  -
0   country                                   3555 non-null   object
1   iso_code                                 3283 non-null   object
2   date                                     3555 non-null   object
3   total_vaccinations                       2341 non-null   float64
4   people_vaccinated                       1940 non-null   float64
5   people_fully_vaccinated                 1278 non-null   float64
6   daily_vaccinations_raw                 1972 non-null   float64
7   daily_vaccinations                     3420 non-null   float64
8   total_vaccinations_per_hundred          2341 non-null   float64
9   people_vaccinated_per_hundred           1940 non-null   float64
10  people_fully_vaccinated_per_hundred     1278 non-null   float64
11  daily_vaccinations_per_million          3420 non-null   float64
12  vaccines                                3555 non-null   object
dtypes: float64(9), object(4)
memory usage: 361.2+ KB

```

```

In [9]: # Заполняем отсутствующие значения
data['total_vaccinations'] = data['total_vaccinations'].replace(0,np.nan)
data['total_vaccinations'] = data['total_vaccinations'].fillna(data['total_

```

```

In [10]: data.head()

```

```

Out[10]:
   country iso_code date total_vaccinations people_vaccinated people_fully_vaccinated
0  Albania   ALB  2021-01-10      1.508878e+06              0.0                NaN
1  Albania   ALB  2021-01-11      1.508878e+06              NaN                NaN
2  Albania   ALB  2021-01-12      1.280000e+02             128.0                NaN
3  Albania   ALB  2021-01-13      1.880000e+02             188.0                NaN
4  Albania   ALB  2021-01-14      2.660000e+02             266.0                NaN

```

```

In [11]: data.isnull().sum()
# проверим есть ли пропущенные значения

```

```

Out[11]:
country              0
iso_code            272
date                0
total_vaccinations   0
people_vaccinated    1615
people_fully_vaccinated 2277
daily_vaccinations_raw 1583
daily_vaccinations   135
total_vaccinations_per_hundred 1214
people_vaccinated_per_hundred 1615
people_fully_vaccinated_per_hundred 2277
daily_vaccinations_per_million 135
vaccines              0
dtype: int64

```

## ЛР №2

### Импорт библиотек

```
In [1]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
from sklearn.impute import SimpleImputer
from sklearn.model_selection import train_test_split
```

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

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

```
Out[3]:
```

	PassengerId	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	A/5 21171	7.2500
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	female	38.0	1	0	PC 17599	71.2833
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	STON/O2. 3101282	7.9250
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	113803	53.1000
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	373450	8.0500

```
In [4]: data['Embarked'].value_counts()
```

```
Out[4]: S    644
C    168
Q     77
Name: Embarked, dtype: int64
```

```
In [5]: # Кодировать признаки Pclass и Embarked в отдельные столбцы
data = pd.get_dummies(data, columns=['Pclass', 'Embarked'])
```

```
In [6]: # Пол кодируем в 1/0
data['IsMale'] = data.Sex.replace({'female':0, 'male':1})
data.drop('Sex', axis = 1, inplace = True)
```

```
In [7]: data.head()
```

```
Out[7]:
```

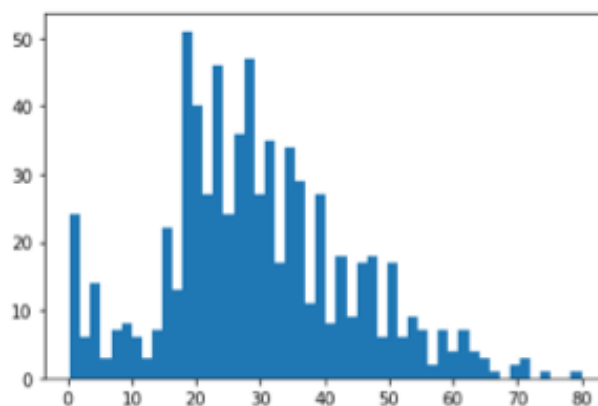
	PassengerId	Survived	Name	Age	SibSp	Parch	Ticket	Fare	Cabin	Pclass
0	1	0	Braund, Mr. Owen Harris	22.0	1	0	A/5 21171	7.2500	NaN	
1	2	1	Cumings, Mrs. John Bradley (Florence Briggs Th...	38.0	1	0	PC 17599	71.2833	C85	
2	3	1	Heikkinen, Miss. Laina	26.0	0	0	STON/O2. 3101282	7.9250	NaN	
3	4	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	35.0	1	0	113803	53.1000	C123	
4	5	0	Allen, Mr. William Henry	35.0	0	0	373450	8.0500	NaN	

## Масштабирование значений

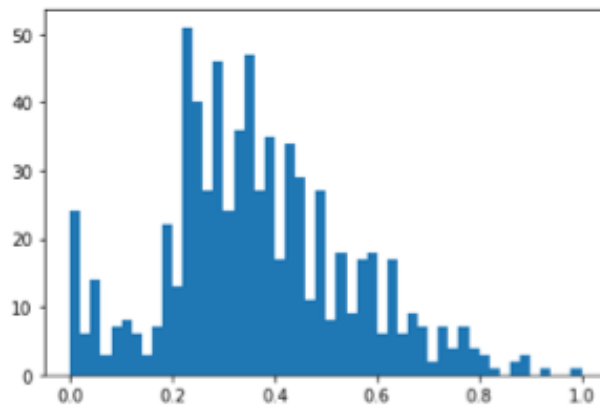
```
In [8]: from sklearn.preprocessing import StandardScaler, MinMaxScaler, StandardSc
```

```
In [9]: scl = MinMaxScaler()  
scl_data = scl.fit_transform(data[['Age']])
```

```
In [10]: plt.hist(data['Age'], 50)  
plt.show()
```



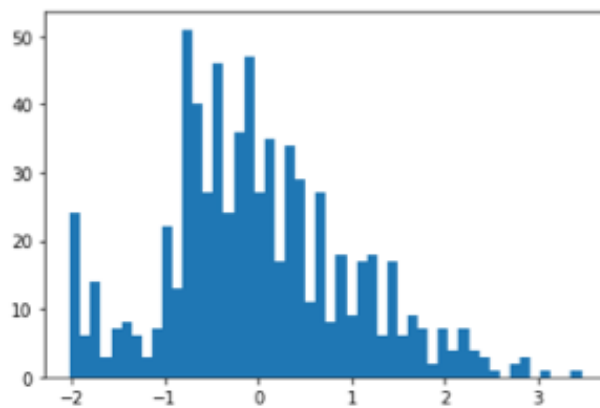
```
In [11]: plt.hist(scl_data, 50)  
plt.show()
```



```
In [12]: # Удаляем столбцы, которые не несут значимой информации
data.drop(['Cabin', 'Name', 'Ticket'], axis = 1, inplace = True)
```

```
In [13]: sc2 = StandardScaler()
sc2_data = sc2.fit_transform(data[['Age']])
```

```
In [14]: plt.hist(sc2_data, 50)
plt.show()
```



```
In [15]: data.head()
```

```
Out[15]:
```

	PassengerId	Survived	Age	SibSp	Parch	Fare	Pclass_1	Pclass_2	Pclass_3	Embarked
0	1	0	22.0	1	0	7.2500	0	0	1	
1	2	1	38.0	1	0	71.2833	1	0	0	
2	3	1	26.0	0	0	7.9250	0	0	1	
3	4	1	35.0	1	0	53.1000	1	0	0	
4	5	0	35.0	0	0	8.0500	0	0	1	