In [3]:	<pre>data['floor'] = data['floor'].replace('-', np.nan) data = data.astype({'floor': float}) data = data.drop('Unnamed: 0', axis = 1)</pre>
In [4]:	data.shape (6080, 13)
In [5]:	data.dtypes
	city int64 area int64 rooms int64 bathroom int64 bathroom int64 parking spaces int64 floor float64 animal object furniture object rent amount object rent amount object rent amount object trie insurance object total object
In [6]:	<pre>data.isnull().sum()</pre>
Out[6]:	city 0 area 0 rooms 0 bathroom 0 parking spaces 7 floor 1555 animal 0 furniture 0 hoa 0 rent amount 0 property tax firsurance 0 total
In [7]:	<pre>dtype: int64 data.head()</pre>
	city area rooms bathroom parking spaces floor animal furnished property tax fire insurance total 0 1 240 3 3 4 NaN acept furnished R\$0 R\$1,000 R\$1,200 R\$1,21 R\$9,121 1 0 64 2 1 1 0 acept not furnished R\$4,172 R\$5,000 R\$1,21 R\$1,493 2 1 443 5 5 4 3 acept not furnished R\$4,172 R\$5,000 R\$1,216 R\$5,000 R\$1,260 R\$5,160 R\$5,160 R\$1,260 R\$1,260 <t< th=""></t<>
In [8]:	<pre>data_floor = data[['floor']] data_floor.head()</pre>
Out[8]:	floor 0 NaN 1 10.0 2 3.0 3 12.0 4 NaN
In [9]:	<pre>plt.hist(data_floor, 50) plt.xlabel('floor') plt.show()</pre>
	800 - 400 -
In [10]:	<pre>from sklearn.impute import SimpleImputer from sklearn.impute import MissingIndicator</pre>
In [11]:	<pre>indicator = MissingIndicator() mask_missing_values_only = indicator.fit_transform(data_floor) mask_missing_values_only</pre>
out[II].	array([[True],
In [12]:	<pre>def test_num_impute(strategy_param): imp_num = SimpleImputer(strategy=strategy_param) data_num_imp = imp_num.fit_transform(data_floor) return data_num_imp[mask_missing_values_only]</pre>
In [17]:	<pre>data_num_imp = test_num_impute('median') data_num_imp</pre>
046[27].	array([6., 6., 6., 6., 6., 6., 6.]) Преобразование категориальных признаков в числовые
In [27]:	<pre>data_temp = [] data_temp.append(data.columns[6]) data_temp.append(data.columns[7]) data_enc = data[data_temp] data_enc</pre>
Out[27]:	 animal furniture acept furnished acept not furnished
	 acept furnished acept not furnished not acept not furnished
	6075 acept not furnished6076 not acept furnished
	 6077 acept not furnished 6078 not acept not furnished 6079 acept furnished
In [31]:	6080 rows × 2 columns for col in data_enc.columns:
	<pre>print(data_enc[col].unique()) ['acept' 'not acept'] ['furnished' 'not furnished']</pre>
In [33]: In [37]:	from sklearn.preprocessing import LabelEncoder, MinMaxScaler
	<pre>le = LabelEncoder() data_enc_animal = le.fit_transform(data_enc['animal']) data_enc_furniture = le.fit_transform(data_enc['furniture'])</pre>
In [39]: Out[39]:	data_enc_animal, data_enc_furniture (array([0, 0, 0,, 0, 1, 0]), array([0, 1, 0,, 1, 1, 0]))
In [46]:	Macштабирование данных sc1 = MinMaxScaler()
Out[46]:	<pre>sc1_data = sc1.fit_transform(data[['rooms']]) sc1_data array([[0.22222222],</pre>
	[0.44444444],, [0.], [0.22222222], [0.1111111]])
In [47]:	<pre>plt.hist(data['rooms'], 50) plt.show()</pre>
	2000 1750 1500 1000 750 200 250 200 200 200 200 200 200 200 2
In [48]:	<pre>plt.hist(sc1_data, 50) plt.show()</pre>
	2000 - 1750 - 1500 - 1250 - 1000 - 750 - 500 - 250 -
	0 0.0 0.2 0.4 0.6 0.8 1.0

Лабораторная работа №2. Сысоев Александр РТ5-61Б

обработку пропусков в данных;

масштабирование данных.

import numpy as np
import pandas as pd

import seaborn as sns

import matplotlib.pyplot as plt
%matplotlib inline
sns.set(style="ticks")

кодирование категориальных признаков;

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

Выбрать набор данных (датасет), содержащий категориальные признаки и пропуски в данных.

Для выбранного датасета (датасетов) на основе материалов лекции решить следующие задачи:

Импорт библиотек, чтение датасета и вывод основных характеристик

Задание: