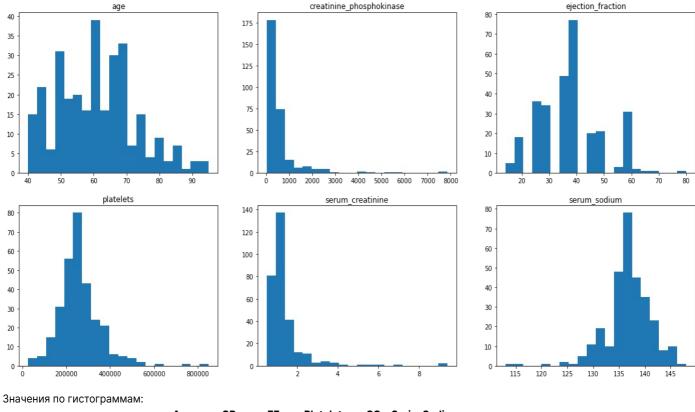
Лабораторная №1

Загрузка данных

```
import pandas as pd
import numpy as np
df = pd.read_csv('heart_failure_clinical_records_dataset.csv')
df = df.drop(columns = ['anaemia', 'diabetes', 'high_blood_pressure', 'sex', 'smoking', 'time', 'DEATH_EVENT'])
print(df) # Вывод датафрейма с данными для лаб. работы. Должно быть 299 наблюдений и 6 признаков
      age
          creatinine_phosphokinase ejection_fraction platelets
0
     75.0
                                582
                                                        265000.00
                                                    38 263358.03
1
    55.0
                               7861
    65.0
                                                    20 162000.00
2
                                146
3
    50.0
                                                    20 210000.00
                                111
    65.0
                                160
                                                    20 327000.00
294 62.0
                                61
                                                    38 155000.00
                                                    38 270000.00
295 55.0
                               1820
296 45.0
                               2060
                                                    60 742000.00
297
    45.0
                               2413
                                                    38 140000.00
298 50.0
                                196
                                                    45 395000.00
     serum_creatinine serum_sodium
0
                  1.9
1
                  1.1
                                136
2
                  1.3
                                129
3
                  1.9
                                137
4
                  2.7
                                116
294
                                143
                 1.1
295
                  1.2
                                139
296
                  0.8
                                138
297
                  1.4
                                140
                                136
298
                  1.6
[299 rows x 6 columns]
import matplotlib.pyplot as plt
plt.rcParams["figure.figsize"] = (20, 10)
n_bins = 20
fig, axs = plt.subplots(2,3)
axs[0, 0].hist(df['age'].values, bins = n_bins)
axs[0, 0].set_title('age')
axs[0, 1].hist(df['creatinine_phosphokinase'].values, bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')
axs[0, 2].hist(df['ejection_fraction'].values, bins = n_bins)
axs[0, 2].set_title('ejection_fraction')
axs[1, 0].hist(df['platelets'].values, bins = n_bins)
axs[1, 0].set_title('platelets')
axs[1, 1].hist(df['serum_creatinine'].values, bins = n_bins)
axs[1, 1].set_title('serum_creatinine')
axs[1, 2].hist(df['serum_sodium'].values, bins = n_bins)
axs[1, 2].set_title('serum_sodium')
plt.show()
```



Age CP EF **Platelets** SC Serim Sodium [40, 96] [0, 8000] [0, 80] [0, 800000] [0, 9] [0, 150] Приблизительный Диапазон 136 Значение, близкое к среднему 60 200 38 220000 1

data = df.to_numpy(dtype='float')

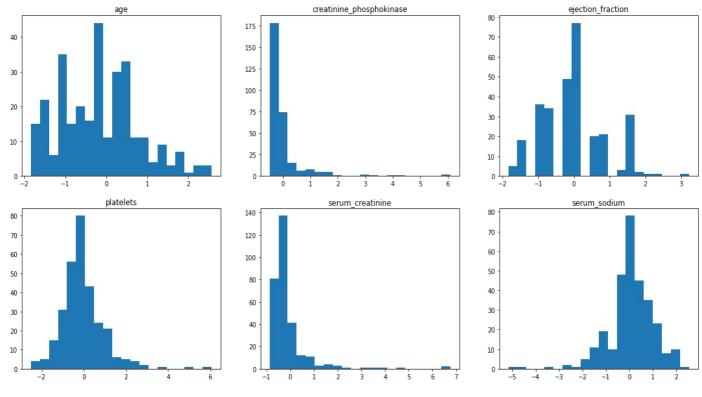
Стандартизация данных

```
from sklearn import preprocessing
scaler = preprocessing.StandardScaler().fit(data[:150,:])
data_scaled = scaler.transform(data)
fig, axs = plt.subplots(2,3)
axs[0, 0].hist(data_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')
axs[0, 1].hist(data_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')
axs[0, 2].hist(data_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')
axs[1, 0].hist(data_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')
```

```
axs[1, 2].hist(data_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')
plt.show()
```

axs[1, 1].hist(data_scaled[:,4], bins = n_bins)

axs[1, 1].set_title('serum_creatinine')



```
fig, axs = plt.subplots(2,3)
```

plt.show()

```
axs[0, 0].hist(data_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')

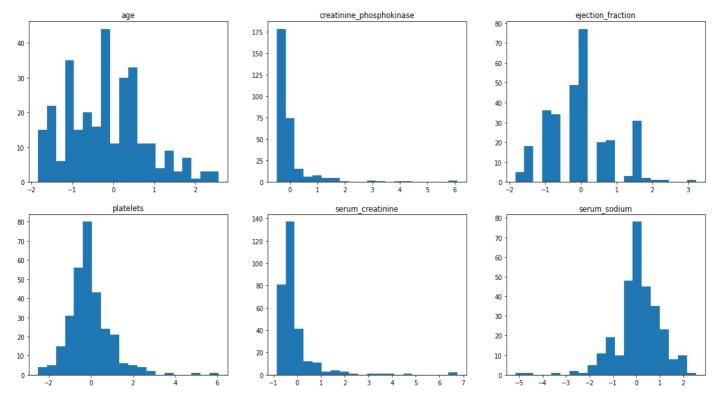
axs[0, 1].hist(data_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')

axs[0, 2].hist(data_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')

axs[1, 0].hist(data_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')

axs[1, 1].hist(data_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')

axs[1, 2].hist(data_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')
```



Что изменилось и почему?

Изменились мат ожидания на 0, а также дисперсии стали сравнимы. Это произошло потому что у кажжого значения набора отняли мат ожидание а и разделили его на СКО. Нормализация произведена с целью, чтобы статистически сравнить величины разных размерностей и порядков.

Расчет параметров:

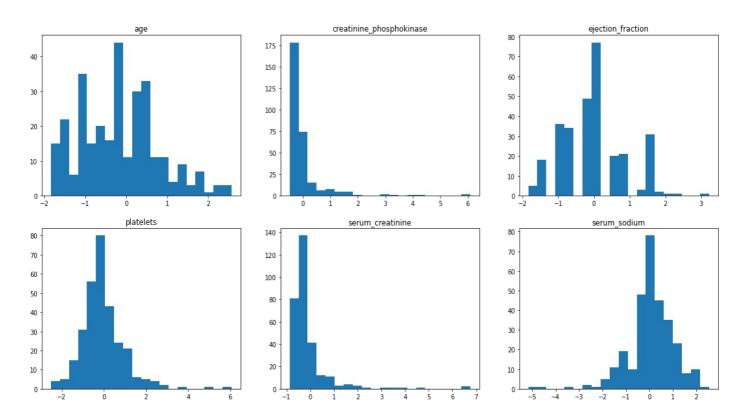
```
mean_age = data[:, 0].mean()
std_age = data[:, 0].std()
mean_cp = data[:, 1].mean()
std_cp = data[:, 1].std()
mean_ef = data[:, 2].mean()
std_ef = data[:, 2].std()
mean_plat = data[:, 3].mean()
std_plat = data[:, 3].std()
mean_sc = data[:, 4].mean()
std_sc = data[:, 4].std()
mean_ss = data[:, 5].mean()
std_ss = data[:, 5].std()
print(f'Age mean (no scaling): {mean age}')
print(f'Age std (no scaling): {std_age}')
print(f'Cp mean (no scaling): {mean_cp}')
print(f'Cp std (no scaling): {std_cp}')
print(f'Ef mean (no scaling): {mean_ef}')
print(f'Ef std (no scaling): {std ef}')
print(f'Plat mean (no scaling): {mean_plat}')
print(f'Plat std (no scaling): {std_plat}')
print(f'Sc mean (no scaling): {mean_sc}')
print(f'Sc std (no scaling): {std_sc}')
print(f'Ss mean (no scaling): {mean_ss}')
print(f'Ss std (no scaling): {std_ss}')
mean_age = data_scaled[:, 0].mean()
std_age = data_scaled[:, 0].std()
mean_cp = data_scaled[:, 1].mean()
std_cp = data_scaled[:, 1].std()
mean_ef = data_scaled[:, 2].mean()
std_ef = data_scaled[:, 2].std()
mean_plat = data_scaled[:, 3].mean()
std_plat = data_scaled[:, 3].std()
mean_sc = data_scaled[:, 4].mean()
std_sc = data_scaled[:, 4].std()
```

```
mean_ss = data_scaled[:, 5].mean()
std_ss = data_scaled[:, 5].std()
print(f'Age mean: {mean age}')
print(f'Age std: {std_age}')
print(f'Cp mean: {mean_cp}')
print(f'Cp std: {std_cp}')
print(f'Ef mean: {mean_ef}')
print(f'Ef std: {std_ef}')
print(f'Plat mean: {mean_plat}')
print(f'Plat std: {std_plat}')
print(f'Sc mean: {mean_sc}')
print(f'Sc std: {std sc}')
print(f'Ss mean: {mean ss}')
print(f'Ss std: {std_ss}')
Age mean (no scaling): 60.83389297658862
Age std (no scaling): 11.874901429842655
Cp mean (no scaling): 581.8394648829432
Cp std (no scaling): 968.6639668032415
Ef mean (no scaling): 38.08361204013378
Ef std (no scaling): 11.815033462318585
Plat mean (no scaling): 263358.02926421404
Plat std (no scaling): 97640.54765451424
Sc mean (no scaling): 1.3938795986622072
Sc std (no scaling): 1.0327786652795918
Ss mean (no scaling): 136.62541806020067
Ss std (no scaling): 4.405092379513557
Age mean: -0.16970362369106984
Age std: 0.9538237876978354
Cp mean: -0.021276750290383013
Cp std: 0.8141790488228113
Ef mean: 0.01050249484809085
Ef std: 0.9061082161919123
Plat mean: -0.035228788194085287
Plat std: 1.0150611342848024
Sc mean: -0.10864080163893569
Sc std: 0.8854288727548568
Ss mean: 0.03790759894920013
Ss std: 0.9703735961735016
import math
mean_age = scaler.mean_[0]
std_age = math.sqrt(scaler.var_[0])
mean_cp = scaler.mean_[1]
std_cp = math.sqrt(scaler.var_[1])
mean_ef = scaler.mean_[2]
std_ef = math.sqrt(scaler.var_[2])
mean_plat = scaler.mean_[3]
std_plat = math.sqrt(scaler.var_[3])
mean_sc = scaler.mean_[4]
std_sc = math.sqrt(scaler.var_[4])
mean_ss = scaler.mean_[5]
std_ss = math.sqrt(scaler.var_[5])
print(f'Age mean: {mean_age}')
print(f'Age std: {std_age}')
print(f'Cp mean: {mean_cp}')
print(f'Cp std: {std_cp}')
print(f'Ef mean: {mean_ef}')
print(f'Ef std: {std_ef}')
print(f'Plat mean: {mean_plat}')
print(f'Plat std: {std_plat}')
print(f'Sc mean: {mean_sc}')
print(f'Sc std: {std_sc}')
print(f'Ss mean: {mean_ss}')
print(f'Ss std: {std_ss}')
```

```
Age mean: 62.9466666666665
Age std: 12.449785361826748
Cp mean: 607.15333333333333
Cp std: 1189.7431752926996
Ef mean: 37.9466666666665
Ef std: 13.039318318923817
Plat mean: 266746.74946666666
Plat std: 96191.79018543586
Sc mean: 1.52060000000000002
Sc std: 1.1664162950393542
Ss mean: 136.45333333333335
Ss std: 4.539583926112858
Формула нормализации: $ z = {{x - u} \over {\sigma}} $$
где x - случайна величина, u - среднее, $\sigma$ - CKO
data_scaled = preprocessing.StandardScaler().fit(data[:,:])
data_scaled = scaler.transform(data)
fig, axs = plt.subplots(2,3)
axs[0, 0].hist(data_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')
axs[0, 1].hist(data_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')
axs[0, 2].hist(data_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')
axs[1, 0].hist(data_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')
axs[1, 1].hist(data_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')
axs[1, 2].hist(data_scaled[:,5], bins = n_bins)
```



axs[1, 2].set_title('serum_sodium')

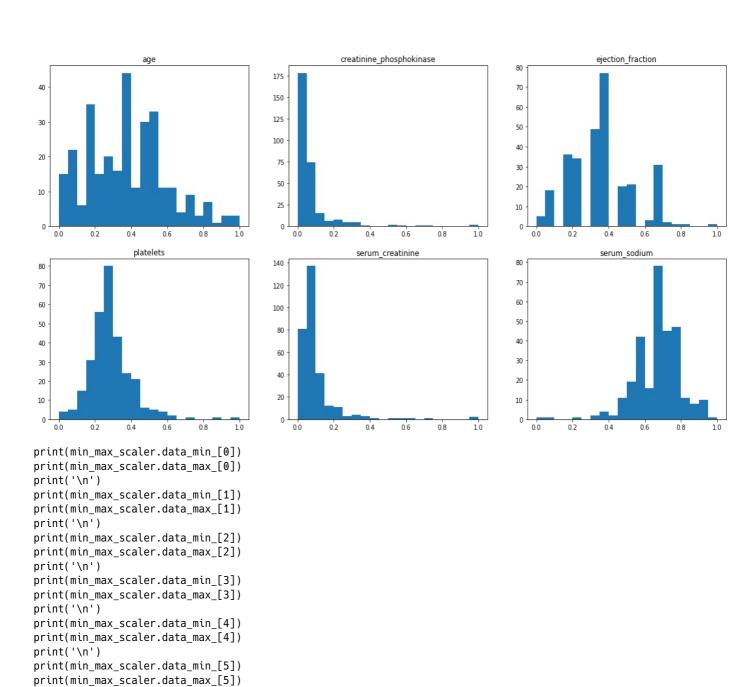


```
import math
```

```
mean age = scaler.mean [0]
std_age = math.sqrt(scaler.var_[0])
mean_cp = scaler.mean_[1]
std_cp = math.sqrt(scaler.var_[1])
mean_ef = scaler.mean_[2]
std_ef = math.sqrt(scaler.var_[2])
mean plat = scaler.mean [3]
std_plat = math.sqrt(scaler.var_[3])
mean_sc = scaler.mean_[4]
std_sc = math.sqrt(scaler.var_[4])
mean ss = scaler.mean [5]
std_ss = math.sqrt(scaler.var_[5])
print(f'Age mean: {mean_age}')
print(f'Age std: {std_age}')
print(f'Cp mean: {mean_cp}')
print(f'Cp std: {std_cp}')
print(f'Ef mean: {mean_ef}')
print(f'Ef std: {std_ef}')
print(f'Plat mean: {mean_plat}')
print(f'Plat std: {std_plat}')
print(f'Sc mean: {mean_sc}')
print(f'Sc std: {std_sc}')
print(f'Ss mean: {mean_ss}')
print(f'Ss std: {std_ss}')
Age mean: 62.9466666666665
Age std: 12.449785361826748
Cp mean: 607.15333333333333
Cp std: 1189.7431752926996
Ef mean: 37.9466666666665
Ef std: 13.039318318923817
Plat mean: 266746.74946666666
Plat std: 96191.79018543586
Sc mean: 1.52060000000000002
Sc std: 1.1664162950393542
Ss mean: 136.45333333333335
Ss std: 4.539583926112858
```

Приведение к диапазону

```
MinMaxScaler
min_max_scaler = preprocessing.MinMaxScaler().fit(data)
data_min_max_scaled = min_max_scaler.transform(data)
fig, axs = plt.subplots(2,3)
axs[0, 0].hist(data_min_max_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')
axs[0, 1].hist(data_min_max_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')
axs[0, 2].hist(data_min_max_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')
axs[1, 0].hist(data_min_max_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')
axs[1, 1].hist(data_min_max_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')
axs[1, 2].hist(data_min_max_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')
plt.show()
```



40.0 95.0

23.0 7861.0

14.0 80.0

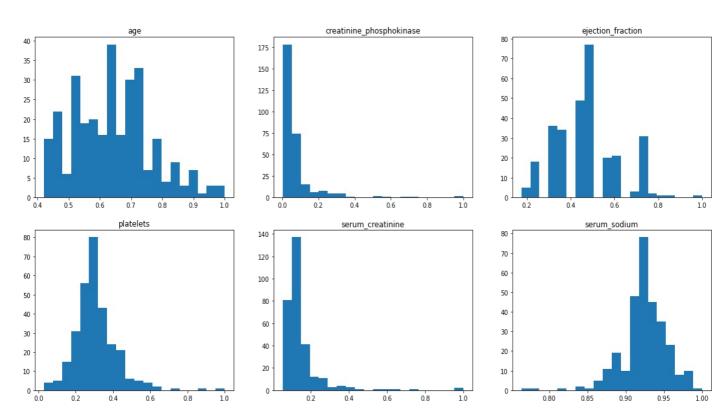
25100.0 850000.0

0.5 9.4

113.0 148.0

MaxAbsScaler

```
max_abs_scaler = preprocessing.MaxAbsScaler().fit(data)
data_max_abs_scaled = max_abs_scaler.transform(data)
fig, axs = plt.subplots(2,3)
axs[0, 0].hist(data_max_abs_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')
axs[0, 1].hist(data_max_abs_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')
axs[0, 2].hist(data_max_abs_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')
axs[1, 0].hist(data_max_abs_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')
axs[1, 1].hist(data_max_abs_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')
axs[1, 2].hist(data_max_abs_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')
plt.show()
```



MaxAbsScaler делит данные на максимальное обсолютное значение, при это разброс не меняется.

RobustScaler

```
robust_scaler = preprocessing.RobustScaler().fit(data)
data_robust_scaled = robust_scaler.transform(data)

fig, axs = plt.subplots(2,3)

axs[0, 0].hist(data_robust_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')

axs[0, 1].hist(data_robust_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')
```

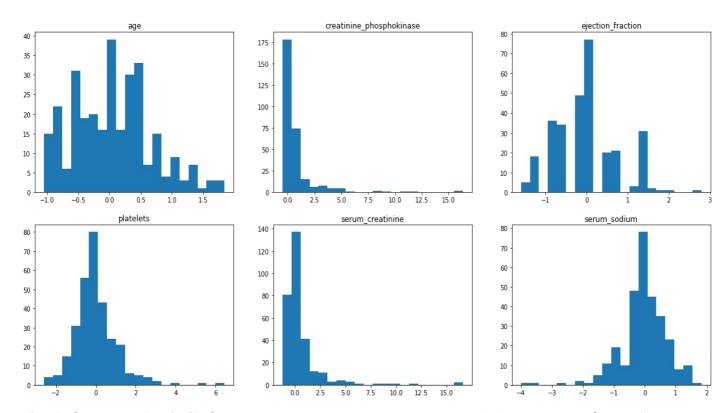
```
axs[0, 2].hist(data_robust_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')

axs[1, 0].hist(data_robust_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')

axs[1, 1].hist(data_robust_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')

axs[1, 2].hist(data_robust_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')

plt.show()
```



RobustScaler похож на StandardScaler, но для стандартизации использует не среднее и СКО, а медиану и IQR (Intequartile range, расстояние между первым и третьи квартилем, т.е. "ширину места, вероятность в которое попасть 50%"). Этот метод прведения более устойчив к выбросам, чем StandardScaler, поэтому так и называется.

Приведение к промежутку [-5, 10]

```
min_max_scaler = preprocessing.MinMaxScaler((-5, 10)).fit(data)
data_min_max_scaled = min_max_scaler.transform(data)
fig, axs = plt.subplots(2,3)
```

```
axs[0, 0].hist(data_min_max_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')

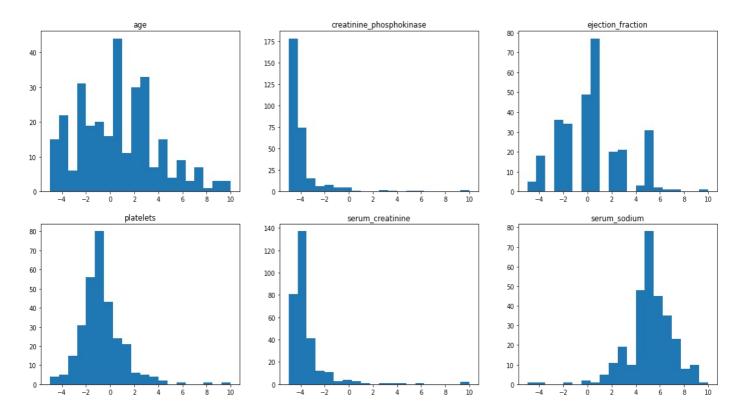
axs[0, 1].hist(data_min_max_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')

axs[0, 2].hist(data_min_max_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')

axs[1, 0].hist(data_min_max_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')

axs[1, 1].hist(data_min_max_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')

axs[1, 2].hist(data_min_max_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')
```



Нелинейные преобразования

```
quantile_transformer = preprocessing.QuantileTransformer(n_quantiles = 100,random_state=0).fit(data)
data_quantile_scaled = quantile_transformer.transform(data)
```

```
fig, axs = plt.subplots(2,3)

axs[0, 0].hist(data_quantile_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')

axs[0, 1].hist(data_quantile_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')

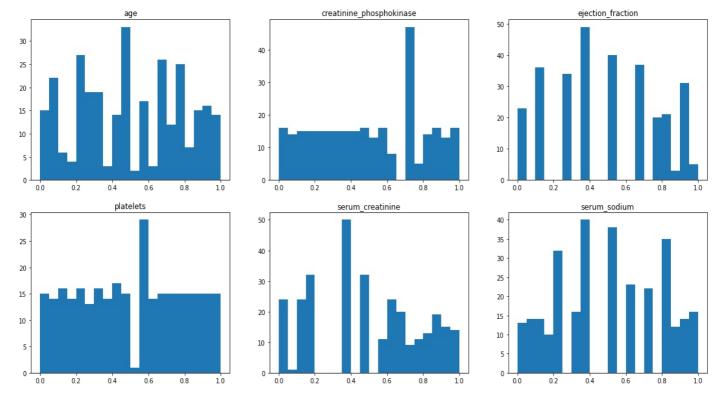
axs[0, 2].hist(data_quantile_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')

axs[1, 0].hist(data_quantile_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')

axs[1, 1].hist(data_quantile_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')

axs[1, 2].hist(data_quantile_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')

plt.show()
```



quantile_transformer = preprocessing.QuantileTransformer(n_quantiles = 10,random_state=0).fit(data)
data_quantile_scaled = quantile_transformer.transform(data)

```
fig, axs = plt.subplots(2,3)

axs[0, 0].hist(data_quantile_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')

axs[0, 1].hist(data_quantile_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')

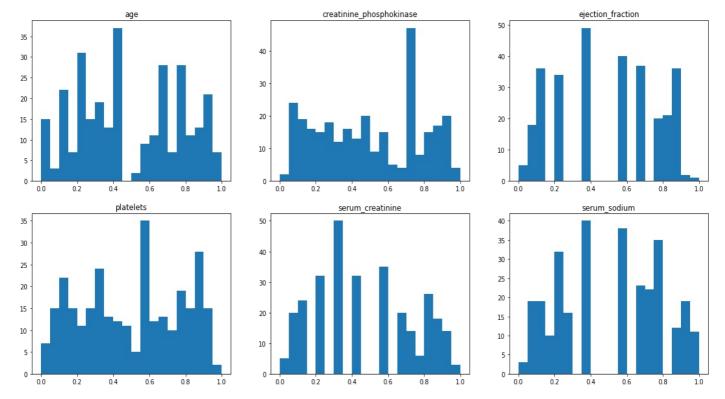
axs[0, 2].hist(data_quantile_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')

axs[1, 0].hist(data_quantile_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')

axs[1, 1].hist(data_quantile_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')

axs[1, 2].hist(data_quantile_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')

plt.show()
```



n_quantiles влияет на то, насколько хорошо приведется исходные данные к равномерному распредлению (в данным случае). Это значение используется для задания размера шага при дискретизации оценочной CDF для исходных данных (CDF используется для приведения к равномерному распределению в QuantileTransformer).

Приведение к номарльному распределению

quantile_transformer = preprocessing.QuantileTransformer(n_quantiles = 100,random_state=0, output_distribution='normal')
data_quantile_scaled = quantile_transformer.transform(data)

```
fig, axs = plt.subplots(2,3)

axs[0, 0].hist(data_quantile_scaled[:,0], bins = n_bins)
axs[0, 0].set_title('age')

axs[0, 1].hist(data_quantile_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')

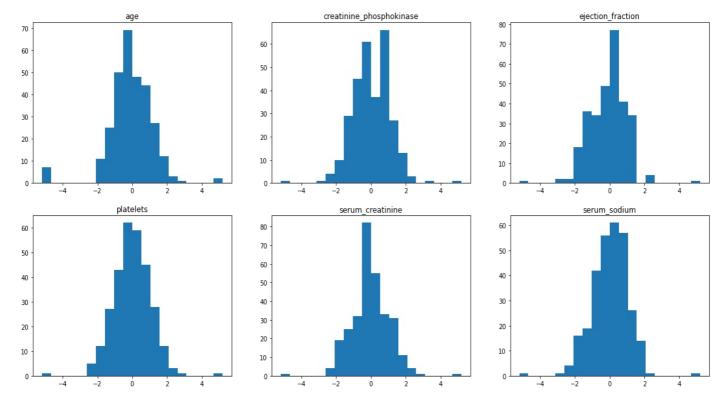
axs[0, 2].hist(data_quantile_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')

axs[1, 0].hist(data_quantile_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')

axs[1, 1].hist(data_quantile_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')

axs[1, 2].hist(data_quantile_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')

plt.show()
```



PowerTransformer

```
data_power_scaled = power_transformer.transform(data)

fig, axs = plt.subplots(2,3)

axs[0, 0].hist(data_power_scaled[:,0], bins = n_bins)c
axs[0, 0].set_title('age')

axs[0, 1].hist(data_power_scaled[:,1], bins = n_bins)
axs[0, 1].set_title('creatinine_phosphokinase')

axs[0, 2].hist(data_power_scaled[:,2], bins = n_bins)
axs[0, 2].set_title('ejection_fraction')

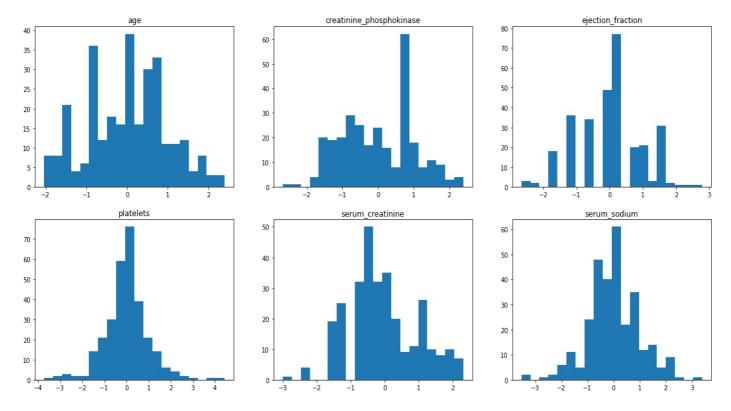
axs[1, 0].hist(data_power_scaled[:,3], bins = n_bins)
axs[1, 0].set_title('platelets')

axs[1, 1].hist(data_power_scaled[:,4], bins = n_bins)
axs[1, 1].set_title('serum_creatinine')

axs[1, 2].hist(data_power_scaled[:,5], bins = n_bins)
axs[1, 2].set_title('serum_sodium')

plt.show()
```

power_transformer = preprocessing.PowerTransformer().fit(data)



Дискретизация признаков

```
biner = preprocessing.KBinsDiscretizer(n_bins=[3, 4, 3, 10, 2, 4], encode='ordinal')
data_bined = biner.fit_transform(data)
```

```
fig, axs = plt.subplots(2,3)

axs[0, 0].hist(data_bined[:,0])
axs[0, 0].set_title('age')

axs[0, 1].hist(data_bined[:,1])
axs[0, 1].set_title('creatinine_phosphokinase')

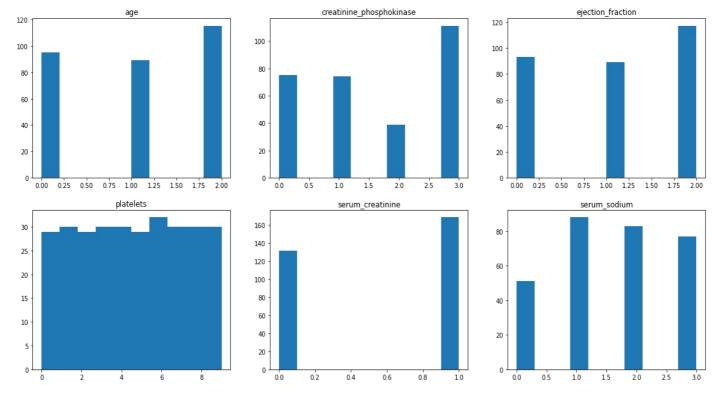
axs[0, 2].hist(data_bined[:,2])
axs[0, 2].set_title('ejection_fraction')

axs[1, 0].hist(data_bined[:,3])
axs[1, 0].set_title('platelets')

axs[1, 1].hist(data_bined[:,4])
axs[1, 1].set_title('serum_creatinine')

axs[1, 2].hist(data_bined[:,5])
axs[1, 2].set_title('serum_sodium')

plt.show()
```



Количество столбиков = количество бинов (промежутков), на которые мы разбили данные.

print(biner.bin_edges_)