

Мартынова П.В ИУ5-61Б Вариант №13

```
import numpy as np
import pandas as pd
from typing import Dict, Tuple
from scipy import stats
from IPython.display import Image
from sklearn.datasets import load_iris, load_boston
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, balanced_accuracy_score
from sklearn.metrics import precision_score, recall_score, f1_score,
classification_report
from sklearn.metrics import confusion_matrix
from sklearn.metrics import mean_absolute_error, mean_squared_error,
mean_squared_log_error, median_absolute_error, r2_score
from sklearn.metrics import roc_curve, roc_auc_score
from sklearn.preprocessing import MinMaxScaler
from sklearn.datasets import make_blobs, make_circles
from sklearn.model_selection import cross_val_score, cross_validate
from sklearn.linear_model import LogisticRegression
from sklearn.ensemble import RandomForestClassifier
from sklearn.svm import SVC, NuSVC, LinearSVC, OneClassSVM, SVR,
NuSVR, LinearSVR
from sklearn.pipeline import make_pipeline
from sklearn.model_selection import RandomizedSearchCV
from sklearn.ensemble import AdaBoostClassifier
from sklearn import svm
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set(style="ticks")

data = pd.read_csv('marvel-wikia-data.csv', sep=",")

data.head()
```

	page_id	name \
0	1678	Spider-Man (Peter Parker)
1	7139	Captain America (Steven Rogers)
2	64786	Wolverine (James \"Logan\" Howlett)
3	1868	Iron Man (Anthony \"Tony\" Stark)
4	2460	Thor (Thor Odinson)

	urlslug	ID \
0	\\/Spider-Man_(Peter_Parker)	Secret Identity
1	\\/Captain_America_(Steven_Rogers)	Public Identity
2	\\/Wolverine_(James_%22Logan%22_Howlett)	Public Identity
3	\\/Iron_Man_(Anthony_%22Tony%22_Stark)	Public Identity
4	\\/Thor_(Thor_Odinson)	No Dual Identity

	ALIGN	EYE	HAIR	SEX	GSM	\
0	Good Characters	Hazel Eyes	Brown Hair	Male Characters	NaN	
1	Good Characters	Blue Eyes	White Hair	Male Characters	NaN	
2	Neutral Characters	Blue Eyes	Black Hair	Male Characters	NaN	
3	Good Characters	Blue Eyes	Black Hair	Male Characters	NaN	
4	Good Characters	Blue Eyes	Blond Hair	Male Characters	NaN	

	ALIVE	APPEARANCES	FIRST APPEARANCE	Year
0	Living Characters	4043.0	Aug-62	1962.0
1	Living Characters	3360.0	Mar-41	1941.0
2	Living Characters	3061.0	Oct-74	1974.0
3	Living Characters	2961.0	Mar-63	1963.0
4	Living Characters	2258.0	Nov-50	1950.0

```
data.isnull().sum()
```

```

page_id      0
name         0
urlslug      0
ID           3770
ALIGN        2812
EYE          9767
HAIR         4264
SEX          854
GSM          16286
ALIVE         3
APPEARANCES  1096
FIRST APPEARANCE 815
Year         815
dtype: int64

```

```
data.shape
```

```
(16376, 13)
```

```
data.pop('GSM')
```

```

0      NaN
1      NaN
2      NaN
3      NaN
4      NaN

```

```

...
16371   NaN
16372   NaN
16373   NaN
16374   NaN
16375   NaN

```

```
Name: GSM, Length: 16376, dtype: object
```

```
data.shape
```

```
(16376, 12)
```

```
data = data.dropna(axis=0, how='any')
data.shape
```

```
(4402, 12)
```

```
data.head()
```

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4	\/Thor_(Thor_Odinson)	No Dual Identity

	ALIGN	EYE	HAIR	SEX \
0	Good Characters	Hazel Eyes	Brown Hair	Male Characters
1	Good Characters	Blue Eyes	White Hair	Male Characters
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4	Good Characters	Blue Eyes	Blond Hair	Male Characters

	ALIVE	APPEARANCES	FIRST APPEARANCE	Year
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3	Living Characters	2961.0	Mar-63	1963.0
4	Living Characters	2258.0	Nov-50	1950.0

Кодируем категориальные признаки

```
data.dtypes
```

page_id	int64
name	object
urlslug	object
ID	object
ALIGN	object
EYE	object
HAIR	object
SEX	object
ALIVE	object
APPEARANCES	float64

```

FIRST APPEARANCE    object
Year                float64
dtype: object

```

```

from sklearn.preprocessing import LabelEncoder, OneHotEncoder
le = LabelEncoder()
df_int = le.fit_transform(data['name'])
data['name'] = df_int
df_int = le.fit_transform(data['urlslug'])
data['urlslug'] = df_int
df_int = le.fit_transform(data['ID'])
data['ID'] = df_int
df_int = le.fit_transform(data['ALIGN'])
data['ALIGN'] = df_int
df_int = le.fit_transform(data['EYE'])
data['EYE'] = df_int
df_int = le.fit_transform(data['HAIR'])
data['HAIR'] = df_int
df_int = le.fit_transform(data['SEX'])
data['SEX'] = df_int
df_int = le.fit_transform(data['ALIVE'])
data['ALIVE'] = df_int
df_int = le.fit_transform(data['FIRST APPEARANCE'])
data['FIRST APPEARANCE'] = df_int
data.head()

```

	page_id	name	urlslug	ID	ALIGN	EYE	HAIR	SEX	ALIVE
0	1678	3738	3738	3	1	8	5	3	1
1	7139	624	624	2	1	3	20	3	1
2	64786	4302	4302	2	2	3	2	3	1
3	1868	1785	1785	2	1	3	2	3	1
4	2460	3942	3942	1	1	3	3	3	1

	FIRST APPEARANCE	Year
0	80	1962.0
1	425	1941.0
2	622	1974.0
3	434	1963.0
4	548	1950.0

Масштабируем числовые данные

```

sc1 = MinMaxScaler()
data['page_id'] = sc1.fit_transform(data[['page_id']])
data['APPEARANCES'] = sc1.fit_transform(data[['APPEARANCES']])

```

```
data['Year'] = scl.fit_transform(data[['Year']])
data.head()
```

	page_id	name	urlslug	ID	ALIGN	EYE	HAIR	SEX	ALIVE
APPEARANCES	\								
0	0.000866	3738	3738	3	1	8	5	3	1
1	0.008108	624	624	2	1	3	20	3	1
2	0.084554	4302	4302	2	2	3	2	3	1
3	0.001118	1785	1785	2	1	3	2	3	1
4	0.001903	3942	3942	1	1	3	3	3	1

	FIRST APPEARANCE	Year
0	80	0.310811
1	425	0.027027
2	622	0.472973
3	434	0.324324
4	548	0.148649

```
data['ALIVE'].unique()
```

```
array([1, 0])
```

```
target = data['ALIVE']
```

```
data_X_train, data_X_test, data_y_train, data_y_test =
train_test_split(
    data, target, test_size=0.2, random_state=1)
```

```
data_X_train.shape, data_y_train.shape
```

```
((3521, 12), (3521,))
```

```
data_X_test.shape, data_y_test.shape
```

```
((881, 12), (881,))
```

```
np.unique(target)
```

```
array([0, 1])
```

Логистическая регрессия

```
model = LogisticRegression()
```

```
model.fit(data_X_train, data_y_train)
```

```
C:\Users\Asus\anaconda3\lib\site-packages\sklearn\linear_model\
_logistic.py:763: ConvergenceWarning: lbfgs failed to converge
(status=1):
```

STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

```
n_iter_i = _check_optimize_result(
    LogisticRegression()

data_y_pred = model.predict(data_X_test)
accuracy_score(data_y_test, data_y_pred)

1.0

f1_score(data_y_test, data_y_pred, average='micro')

1.0
```

Случайный лес

```
model_2 = RandomForestClassifier()
model_2.fit(data_X_train, data_y_train)

RandomForestClassifier()

data_y_pred = model_2.predict(data_X_test)
accuracy_score(data_y_test, data_y_pred)

1.0

f1_score(data_y_test, data_y_pred, average='micro')

1.0
```