

Оглавление

- 1. Формулировка задачи
- 2. Основные характеристики датасета
- 3. Обработка пропусков в данных

1) Формулировка задачи:

Для заданного набора данных проведите обработку пропусков в данных для одного категориального и одного количественного признака. Какие способы обработки пропусков в данных для категориальных и количественных признаков Вы использовали? Какие признаки Вы будете использовать для дальнейшего построения моделей машинного обучения и почему?

Подключение библиотек

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

plt.style.use('fivethirtyeight')
%matplotlib inline
```

Загрузка данных в Python

```
In [2]: data = pd.read_csv('rk №1/dc-wikia-data.csv', low_memory=False)
```

2) Основные характеристики датасета

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

Out[3]:

	page_id	name	urlslug	ID	ALIGN	EYE	HAIR	SEX	GSM	ALIVE	APPEARANCES	APPE
0	1422	Batman (Bruce Wayne)	VwikiVBatman_(Bruce_Wayne)	Secret Identity	Good Characters	Blue Eyes	Black Hair	Male Characters	NaN	Living Characters	3093.0	1
1	23387	Superman (Clark Kent)	VwikiVSuperman_(Clark_Kent)	Secret Identity	Good Characters	Blue Eyes	Black Hair	Male Characters	NaN	Living Characters	2496.0	1986
2	1458	Green Lantern (Hal Jordan)	VwikiVGreen_Lantern_(Hal_Jordan)	Secret Identity	Good Characters	Brown Eyes	Brown Hair	Male Characters	NaN	Living Characters	1565.0	1959
3	1659	James Gordon (New Earth)	VwikiVJames_Gordon_(New_Earth)	Public Identity	Good Characters	Brown Eyes	White Hair	Male Characters	NaN	Living Characters	1316.0	1987,
4	1576	Richard Grayson (New Earth)	VwikiVRichard_Grayson_(New_Earth)	Secret Identity	Good Characters	Blue Eyes	Black Hair	Male Characters	NaN	Living Characters	1237.0	1

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

```
Out[4]: (6896, 13)
```

```
In [5]: data.dtypes
```

```
Out[5]: page_id          int64
name              object
urlslug          object
ID               object
ALIGN            object
EYE              object
HAIR             object
SEX              object
GSM              object
ALIVE            object
APPEARANCES      float64
FIRST APPEARANCE object
YEAR             float64
dtype: object
```

```
In [6]: for col in data.columns:
# Количество пустых значений - все значения заполнены
temp_null_count = data[data[col].isnull()].shape[0]
print('{} - {}'.format(col, temp_null_count))
```

```
page_id - 0
name - 0
urlslug - 0
ID - 2013
ALIGN - 601
EYE - 3628
HAIR - 2274
SEX - 125
GSM - 6832
ALIVE - 3
APPEARANCES - 355
FIRST APPEARANCE - 69
YEAR - 69
```

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

```
Out[7]:
```

	page_id	APPEARANCES	YEAR
count	6896.000000	6541.000000	6827.000000
mean	147441.209252	23.625134	1989.766662
std	108388.631149	87.378509	16.824194
min	1380.000000	1.000000	1935.000000
25%	44105.500000	2.000000	1983.000000
50%	141267.000000	6.000000	1992.000000
75%	213203.000000	15.000000	2003.000000
max	404010.000000	3093.000000	2013.000000

3) Обработка пропусков в данных

```
In [8]: data.drop(data[data['APPEARANCES'].isnull()].index, inplace=True)
```

```
In [9]: for col in data.columns:
# Количество пустых значений - все значения заполнены
temp_null_count = data[data[col].isnull()].shape[0]
print('{} - {}'.format(col, temp_null_count))
```

```
page_id - 0
name - 0
```

urlslug - 0
ID - 1883
ALIGN - 566
EYE - 3426
HAIR - 2093
SEX - 114
GSM - 6477
ALIVE - 2
APPEARANCES - 0
FIRST APPEARANCE - 60
YEAR - 60

```
In [10]: from sklearn.impute import SimpleImputer
# Импутация константой
imp = SimpleImputer(missing_values=np.nan, strategy='constant', fill_value='NA')
data_imp = imp.fit_transform(data)
data_imp
```

```
Out[10]: array([[1422, 'Batman (Bruce Wayne)', '\\\\wiki\\\\Batman_(Bruce_Wayne)',
..., 3093.0, '1939, May', 1939.0],
[23387, 'Superman (Clark Kent)',
'\\\\wiki\\\\Superman_(Clark_Kent)', ..., 2496.0, '1986, October',
1986.0],
[1458, 'Green Lantern (Hal Jordan)',
'\\\\wiki\\\\Green_Lantern_(Hal_Jordan)', ..., 1565.0,
'1959, October', 1959.0],
...,
[345590, 'Apollo (Roman God) (New Earth)',
'\\\\wiki\\\\Apollo_(Roman_God)_(New_Earth)', ..., 1.0, 'NA', 'NA'],
[15050, 'Ben Lo (New Earth)', '\\\\wiki\\\\Ben_Lo_(New_Earth)', ...,
1.0, 'NA', 'NA'],
[205584, 'Auctioneer II (New Earth)',
'\\\\wiki\\\\Auctioneer_II_(New_Earth)', ..., 1.0, 'NA', 'NA']],
dtype=object)
```

```
In [11]: data_imp.shape
```

Out[11]: (6541, 13)

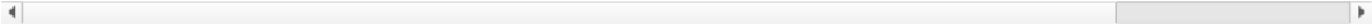
```
In [12]: result = pd.DataFrame(data_imp, columns = ['page_id', 'name', 'urlslug', 'ID', 'ALIGN', 'EYE', 'HAIR', 'SEX', 'GSM', 'ALIV
result
```

Out[12]:

	page_id	name	urlslug	ID	ALIGN	EYE	HAIR	SEX	GSM	ALIVE	APPEARANCES
0	1422	Batman (Bruce Wayne)	VwikiVBatman_(Bruce_Wayne)	Secret Identity	Good Characters	Blue Eyes	Black Hair	Male Characters	NA	Living Characters	3093.0
1	23387	Superman (Clark Kent)	VwikiVSuperman_(Clark_Kent)	Secret Identity	Good Characters	Blue Eyes	Black Hair	Male Characters	NA	Living Characters	2496.0
2	1458	Green Lantern (Hal Jordan)	VwikiVGreen_Lantern_(Hal_Jordan)	Secret Identity	Good Characters	Brown Eyes	Brown Hair	Male Characters	NA	Living Characters	1565.0
3	1659	James Gordon (New Earth)	VwikiVJames_Gordon_(New_Earth)	Public Identity	Good Characters	Brown Eyes	White Hair	Male Characters	NA	Living Characters	1316.0
4	1576	Richard Grayson (New Earth)	VwikiVRichard_Grayson_(New_Earth)	Secret Identity	Good Characters	Blue Eyes	Black Hair	Male Characters	NA	Living Characters	1237.0
...
6536	16094	Mark Antaeus (New Earth)	VwikiVMark_Antaeus_(New_Earth)	Public Identity	Good Characters	Blue Eyes	Black Hair	Male Characters	NA	Deceased Characters	1.0
6537	128000	Jerome Cox (New Earth)	VwikiVJerome_Cox_(New_Earth)	Public Identity	Bad Characters	NA	NA	Male Characters	NA	Living Characters	1.0
6538	345590	Apollo (Roman God) (New Earth)	VwikiVApollo_(Roman_God)_(New_Earth)	NA	Good Characters	NA	NA	Male Characters	NA	Living Characters	1.0

6539	15050	Ben Lo (New Earth)	VwikiVBen_Lo_(New_Earth)	Public Identity	Good Characters	Brown Eyes	Black Hair	Male Characters	NA	Living Characters	1.0
6540	205584	Auctioneer II (New Earth)	VwikiVAuctioneer_II_(New_Earth)	Secret Identity	Bad Characters	NA	White Hair	Male Characters	NA	Living Characters	1.0

6541 rows x 13 columns



Для дальнейшего построения моделей машинного обучения следует использовать количественный признак **"APPEARANCES"** вместе с категориальными признаками, у которых несколько уникальных значений ('ID','ALIGN','EYE','HAIR','SEX','ALIVE')