TMO-RK1

April 18, 2021

```
0.0.1
                     №1
                     , 5-62,
    0.0.2
                                  19.
                                          3.
        №3.
                      (label encoding, one hot encoding)
                                5-62
[1]: import numpy as np
     import pandas as pd
     import seaborn as sns
     import matplotlib.pyplot as plt
     from sklearn.impute import SimpleImputer
     from sklearn.preprocessing import *
    0.1
[2]: data = pd.read_csv('marvel-wikia-data.csv', sep=",")
[3]: #
     data.head()
[3]:
        page_id
                                                  name
     0
           1678
                            Spider-Man (Peter Parker)
                     Captain America (Steven Rogers)
     1
           7139
     2
                 Wolverine (James \"Logan\" Howlett)
          64786
     3
           1868
                   Iron Man (Anthony \"Tony\" Stark)
           2460
                                  Thor (Thor Odinson)
     4
                                                                  ID
                                         urlslug
                                                                    \
     0
                     \/Spider-Man_(Peter_Parker)
                                                    Secret Identity
              \/Captain_America_(Steven_Rogers)
                                                    Public Identity
     1
     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
                                    FYF.
                                               HAIR
                                                                   SEX GSM \
```

```
Male Characters
     0
           Good Characters
                             Hazel Eyes
                                          Brown Hair
                                                                        NaN
     1
                                                      Male Characters
           Good Characters
                              Blue Eyes
                                          White Hair
                                                                        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
                            APPEARANCES FIRST APPEARANCE
                                                              Year
                     ALIVE
       Living Characters
                                  4043.0
                                                   Aug-62
                                                            1962.0
     1 Living Characters
                                                   Mar-41
                                                            1941.0
                                  3360.0
     2 Living Characters
                                                   Oct-74
                                                            1974.0
                                  3061.0
     3 Living Characters
                                                   Mar-63
                                  2961.0
                                                            1963.0
     4 Living Characters
                                 2258.0
                                                   Nov-50
                                                            1950.0
[4]: #
     data.dtypes
[4]: page_id
                            int64
    name
                           object
                           object
     urlslug
     ID
                           object
     ALIGN
                           object
     EYE
                           object
     HAIR
                           object
     SEX
                           object
     GSM
                           object
     ALIVE
                           object
     APPEARANCES
                          float64
     FIRST APPEARANCE
                           object
     Year
                          float64
     dtype: object
    0.2
[5]: #
     data.describe()
[5]:
                             APPEARANCES
                                                   Year
                  page_id
             16376.000000
                            15280.000000
                                           15561.000000
     count
            300232.082377
                               17.033377
                                            1984.951803
     mean
     std
            253460.403399
                               96.372959
                                              19.663571
```

1939.000000

1974.000000

1990.000000

2000.000000

2013.000000

1.000000

1.000000

3.000000

8.000000

4043.000000

min

25%

50%

75%

max

1025.000000

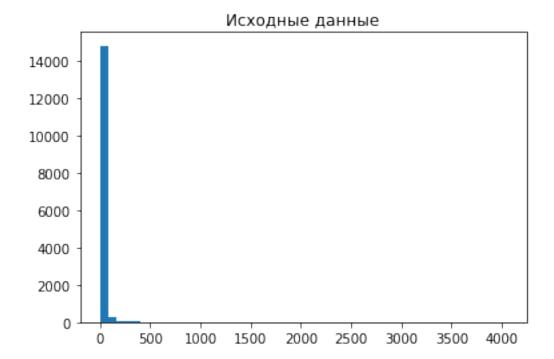
28309.500000

282578.000000

509077.000000

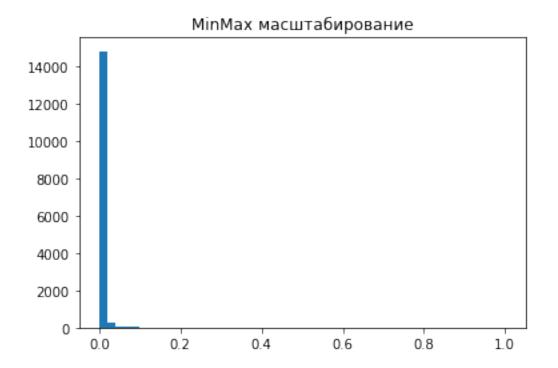
755278.000000

0.2.1 - 1 4043

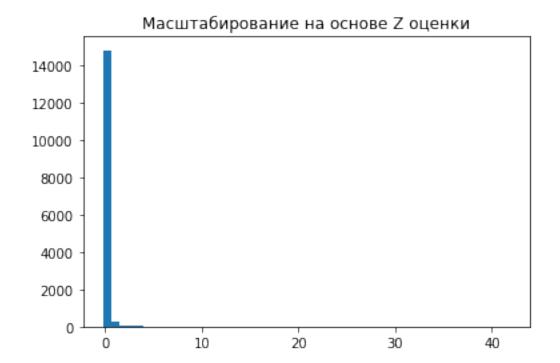


0.2.2 MinMax - 0 1

```
[7]: sc1 = MinMaxScaler()
    sc1_data = sc1.fit_transform(data[['APPEARANCES']])
    plt.hist(sc1_data, 50)
    plt.title("MinMax ")
    plt.show()
```



```
0.2.3 Z - -3 3
```



0.3

0.3.1 Label encoding

```
[9]: #
                           "Unknown"
     imp2 = SimpleImputer(missing_values=np.nan, strategy='constant',__
      data['EYE'] = imp2.fit_transform(data[['EYE']])
     types = data['EYE']
     types.unique()
 [9]: array(['Hazel Eyes', 'Blue Eyes', 'Brown Eyes', 'Green Eyes', 'Grey Eyes',
            'Yellow Eyes', 'Gold Eyes', 'Red Eyes', 'Black Eyeballs',
            'Amber Eyes', 'Variable Eyes', 'Unknown', 'Black Eyes',
            'White Eyes', 'Orange Eyes', 'Silver Eyes', 'Purple Eyes',
            'Pink Eyes', 'One Eye', 'Violet Eyes', 'Multiple Eyes',
            'Magenta Eyes', 'Yellow Eyeballs', 'No Eyes', 'Compound Eyes'],
           dtype=object)
[10]: #label encoding
     le = LabelEncoder()
```

```
[11]: np.unique(data_le)
[11]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13, 14, 15, 16,
             17, 18, 19, 20, 21, 22, 23, 24])
[12]: le.inverse_transform(data_le)
[12]: array(['Hazel Eyes', 'Blue Eyes', 'Blue Eyes', ..., 'Black Eyes',
             'Unknown', 'Unknown'], dtype=object)
     0.3.2 One hot encoding
[13]: pd.get_dummies(data['EYE']).head()
[13]:
         Amber Eyes
                     Black Eyeballs
                                      Black Eyes Blue Eyes
                                                              Brown Eyes
      1
                                   0
                                               0
                                                           1
                                                                        0
      2
                  0
                                   0
                                               0
                                                           1
                                                                        0
      3
                  0
                                   0
                                               0
                                                           1
                                                                        0
      4
                  0
                                   0
                                               0
                                                           1
                                                                        0
         Compound Eyes Gold Eyes Green Eyes Grey Eyes Hazel Eyes
      0
                     0
      1
                                 0
                                             0
                                                         0
      2
                     0
                                 0
                                             0
                                                         0
                                                                      0
      3
                     0
                                 0
                                             0
                                                         0
                                                                      0
                     0
                                 0
                                             0
                                                         0
                    Purple Eyes Red Eyes
                                           Silver Eyes Unknown
         Pink Eyes
                                                                  Variable Eyes \
      0
                 0
                               0
                                         0
                                                       0
                                                                0
                                                                                0
      1
      2
                 0
                               0
                                         0
                                                       0
                                                                0
                                                                                0
      3
                 0
                               0
                                         0
                                                       0
                                                                0
                                                                                0
                 0
                               0
                                         0
                                                       0
                                                                0
                                                                                0
         Violet Eyes
                     White Eyes Yellow Eyeballs Yellow Eyes
      0
                   0
                   0
                                0
                                                  0
                                                               0
      1
      2
                   0
                                0
                                                  0
                                                               0
      3
                   0
                                0
                                                  0
                                                               0
                   0
                                0
                                                  0
                                                               0
```

data_le = le.fit_transform(types)

```
[5 rows x 25 columns]
```

0.4

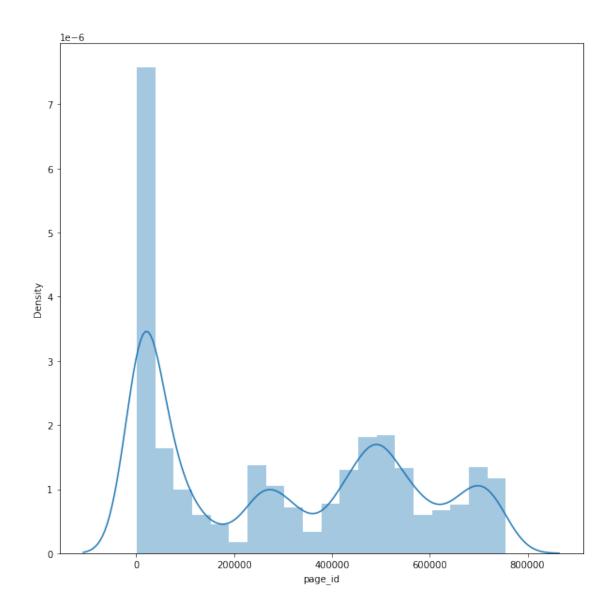
,

```
[14]: fig, ax = plt.subplots(figsize=(10,10))
sns.distplot(data['page_id'])
```

/usr/local/lib/python3.9/site-packages/seaborn/distributions.py:2557: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

[14]: <AxesSubplot:xlabel='page_id', ylabel='Density'>



[]: