# practical2

### April 18, 2024

1 Create an "Academic performance" dataset of students and perform the following operations using Python.

```
[1]: import pandas as pd
     import numpy as np
[2]: df = pd.read_csv('lego_sets.csv')
[3]:
     df.head(10)
[3]:
       set_id
                                                       theme
                                                                           subtheme
                                      name
                                            year
           1-8
     0
                          Small house set
                                            1970
                                                   Minitalia
                                                                                NaN
     1
          2-8
                         Medium house set
                                            1970
                                                   Minitalia
                                                                                NaN
     2
          3-6
                                            1970
                                                   Minitalia
                                                                                NaN
                        Medium house set
     3
          4-4
                          Large house set
                                            1970
                                                   Minitalia
                                                                                NaN
                                                                       Model Maker
     4
          4-6
                 Mini House and Vehicles
                                            1970
                                                   Samsonite
     5
        078-1
                Roadway Base Plate 50X50
                                            1970
                                                   Samsonite
                                                                      Supplemental
        104-1
     6
                  4.5V Replacement Motor
                                            1970
                                                               Supplemental / 4.5V
                                                      Trains
     7
        126-1
                 Steam Locomotive (Push)
                                            1970
                                                      Trains
                                                                               4.5v
        157-3
                    4 Car Auto Transport
                                            1970
                                                   Samsonite
                                                                        Model Maker
        242 - 4
                           Big Model Book
                                            1970
                                                       Books
                                                                               LEGO
                                  pieces
            themeGroup category
                                           minifigs
                                                      agerange_min
                                                                     US_retailPrice
     0
               Vintage
                          Normal
                                     67.0
                                                NaN
                                                                NaN
                                                                                 NaN
                          Normal
                                    109.0
     1
               Vintage
                                                NaN
                                                                NaN
                                                                                 NaN
     2
                                                NaN
               Vintage
                          Normal
                                    158.0
                                                                NaN
                                                                                 NaN
     3
               Vintage
                          Normal
                                    233.0
                                                NaN
                                                                NaN
                                                                                 NaN
     4
               Vintage
                          Normal
                                      NaN
                                                NaN
                                                                NaN
                                                                                 NaN
     5
               Vintage
                          Normal
                                      1.0
                                                 NaN
                                                                NaN
                                                                                 NaN
     6
           Modern day
                          Normal
                                      1.0
                                                 NaN
                                                                NaN
                                                                                 NaN
     7
           Modern day
                          Normal
                                     60.0
                                                 NaN
                                                                NaN
                                                                                 NaN
     8
               Vintage
                          Normal
                                     65.0
                                                 NaN
                                                                NaN
                                                                                 NaN
        Miscellaneous
                            Book
                                      NaN
                                                 NaN
                                                                                 NaN
                                                                NaN
                              bricksetURL
     0
          https://brickset.com/sets/1-8
     1
          https://brickset.com/sets/2-8
```

```
2
    https://brickset.com/sets/3-6
3
    https://brickset.com/sets/4-4
4
    https://brickset.com/sets/4-6
 https://brickset.com/sets/078-1
 https://brickset.com/sets/104-1
7
 https://brickset.com/sets/126-1
8 https://brickset.com/sets/157-3
  https://brickset.com/sets/242-4
                                       thumbnailURL \
0
    https://images.brickset.com/sets/small/1-8.jpg
1
    https://images.brickset.com/sets/small/2-8.jpg
2
    https://images.brickset.com/sets/small/3-6.jpg
3
    https://images.brickset.com/sets/small/4-4.jpg
4
5
  https://images.brickset.com/sets/small/078-1.jpg
  https://images.brickset.com/sets/small/104-1.jpg
7
  https://images.brickset.com/sets/small/126-1.jpg
 https://images.brickset.com/sets/small/157-3.jpg
  https://images.brickset.com/sets/small/242-4.jpg
                                            imageURL
0
    https://images.brickset.com/sets/images/1-8.jpg
1
    https://images.brickset.com/sets/images/2-8.jpg
2
    https://images.brickset.com/sets/images/3-6.jpg
3
     https://images.brickset.com/sets/images/4-4.jpg
4
  https://images.brickset.com/sets/images/078-1.jpg
5
  https://images.brickset.com/sets/images/104-1.jpg
7
  https://images.brickset.com/sets/images/126-1.jpg
  https://images.brickset.com/sets/images/157-3.jpg
  https://images.brickset.com/sets/images/242-4.jpg
```

2 1. Scan all variables for missing values and inconsistencies. If there are missing values and/or inconsistencies, use any of the suitable techniques to deal with them.

18457 non-null

name

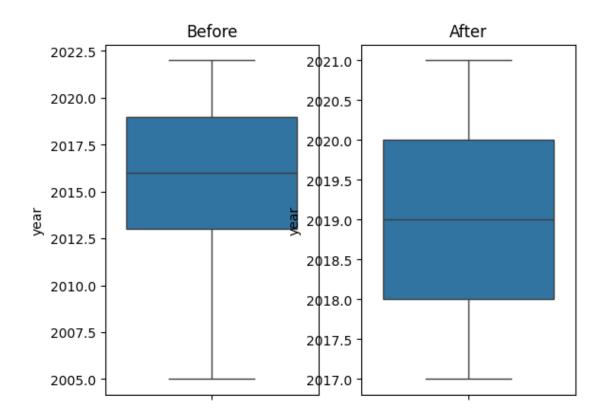
object

```
2
                          18457 non-null
                                          int64
         year
     3
         theme
                         18457 non-null object
     4
         subtheme
                         14901 non-null
                                          object
     5
         themeGroup
                          18455 non-null
                                          object
     6
         category
                          18457 non-null
                                          object
     7
         pieces
                          14533 non-null float64
         minifigs
                         8399 non-null
                                          float64
         agerange_min
                         6787 non-null
                                          float64
     10 US_retailPrice 6982 non-null
                                          float64
     11 bricksetURL
                          18457 non-null object
     12 thumbnailURL
                         17451 non-null
                                          object
         imageURL
                         17451 non-null
     13
                                          object
    dtypes: float64(4), int64(1), object(9)
    memory usage: 2.0+ MB
[5]: df.isnull().sum()
[5]: set_id
                           0
    name
                           0
     year
                           0
     theme
                           0
     subtheme
                        3556
     themeGroup
                           2
                           0
     category
    pieces
                        3924
    minifigs
                       10058
     agerange_min
                       11670
    US_retailPrice
                       11475
     bricksetURL
                           0
     thumbnailURL
                        1006
     imageURL
                        1006
     dtype: int64
[6]: new_df = df['subtheme'].replace(np.nan,0)
                                                   #fill null avalue with O
[7]: new_df.isnull().sum()
[7]: 0
[8]: df.dropna(axis=0,inplace=True)
[9]: df.info()
    <class 'pandas.core.frame.DataFrame'>
    Index: 2669 entries, 6107 to 18029
    Data columns (total 14 columns):
         Column
                         Non-Null Count Dtype
```

```
0
     \mathtt{set\_id}
                     2669 non-null
                                      object
                      2669 non-null
                                      object
 1
     name
 2
                     2669 non-null
                                      int64
     year
 3
     theme
                     2669 non-null
                                      object
 4
                     2669 non-null
                                      object
     subtheme
 5
     themeGroup
                     2669 non-null
                                      object
 6
     category
                     2669 non-null
                                      object
 7
     pieces
                     2669 non-null
                                      float64
 8
    minifigs
                     2669 non-null
                                      float64
     agerange_min
 9
                     2669 non-null
                                      float64
 10 US_retailPrice 2669 non-null
                                      float64
    bricksetURL
                     2669 non-null
                                      object
 11
 12 thumbnailURL
                      2669 non-null
                                      object
 13 imageURL
                     2669 non-null
                                      object
dtypes: float64(4), int64(1), object(9)
memory usage: 312.8+ KB
```

3 2. Scan all numeric variables for outliers. If there are outliers, use any of the suitable techniques to deal with them.

```
[10]: import seaborn as sb
      import matplotlib.pyplot as plt
[22]: ###warnings.filterwarnings("ignore")
      fig,axis = plt.subplots(1,2)
      max_val = df.year.quantile(0.95)
      min_val = df.year.quantile(0.5)
      print("Before Shape", df.shape)
      df2 = df[(df['year']>min_val) & (df['year']<max_val)]</pre>
      print("After Shape", df2.shape)
      sb.boxplot(df['year'], orient = 'v', ax=axis[0])
      axis[0].title.set_text("Before")
      sb.boxplot(df2['year'], orient = 'v', ax=axis[1])
      axis[1].title.set_text("After")
      plt.show
     Before Shape (2669, 14)
     After Shape (1003, 14)
[22]: <function matplotlib.pyplot.show(close=None, block=None)>
```



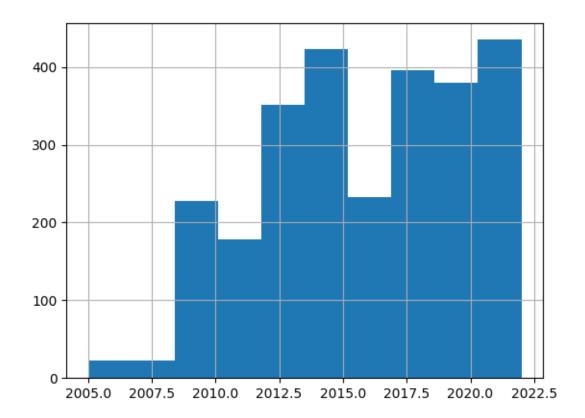
3. Apply data transformations on at least one of the variables. The purpose of this transformation should be one of the following reasons: to change the scale for better understanding of the variable, to convert a non-linear relation into a linear one, or to decrease the skewness and convert the distribution into a normal distribution.

2]:		set_id			name	year	theme	subtheme \	\
	6107	7236-1		Police	Car	2005	City	Police	
	6254	10144-1		Sandcra	wler	2005	Star Wars	Episode IV	
	6407	4962-1		Baby	Zoo	2006	Duplo	LEGO Ville	
	6445	6209-1		Sla	ve I	2006	Star Wars	Episode V	
	6447	6211-1 I	mperial St	ar Destr	oyer	2006	Star Wars	Episode IV	
		themeGroup	category	pieces	mini	figs	agerange_min	n US_retailPr	rice
	6107	Modern day	Normal	59.0		1.0	5.0	) 5	5.99
	6254	Licensed	Normal	1669.0		11.0	12.0	139	9.99
	6407	Pre-school	Normal	18.0		1.0	2.0	) 9	9.99
	6445	Licensed	Normal	537.0		5.0	8.0	) 49	9.99
	6447	Licensed	Normal	1367.0		9.0	9.0	99	9.99

https://brickset.com/sets/7236-1

6107

```
6254 https://brickset.com/sets/10144-1
     6407
            https://brickset.com/sets/4962-1
     6445
            https://brickset.com/sets/6209-1
     6447
            https://brickset.com/sets/6211-1
                                                thumbnailURL \
     6107 https://images.brickset.com/sets/small/7236-1.jpg
     6254 https://images.brickset.com/sets/small/10144-1...
     6407 https://images.brickset.com/sets/small/4962-1.jpg
     6445 https://images.brickset.com/sets/small/6209-1.jpg
     6447 https://images.brickset.com/sets/small/6211-1.jpg
                                                    imageURL
     6107 https://images.brickset.com/sets/images/7236-1...
     6254 https://images.brickset.com/sets/images/10144-...
     6407 https://images.brickset.com/sets/images/4962-1...
     6445 https://images.brickset.com/sets/images/6209-1...
     6447 https://images.brickset.com/sets/images/6211-1...
[13]:
     import sklearn
[14]: from sklearn.preprocessing import StandardScaler
[15]: scaler = StandardScaler()
     x = df[['agerange_min', 'year', 'US_retailPrice', 'minifigs']]
     scaledf = scaler.fit_transform(x)
     print(scaledf)
     [[-0.62750576 -2.73682016 -0.67971696 -0.86203015]
      [ 2.00112431 -2.73682016 1.57016224 2.95855848]
      [-1.7540615 -2.48543244 -0.61255639 -0.86203015]
      2.40966941 1.81238189]
      [ 0.49904998  1.5367711
                                0.56275364
                                           3.34061735]
      [ 0.49904998  1.5367711
                                1.23435938 3.72267621]]
[16]: df.year.hist()
[16]: <Axes: >
```



# [17]: import scipy.stats as stats

[18]: sb.distplot(df['year'], bins=40)

C:\Users\kumar\AppData\Local\Temp\ipykernel\_9060\3404656447.py:1: UserWarning:

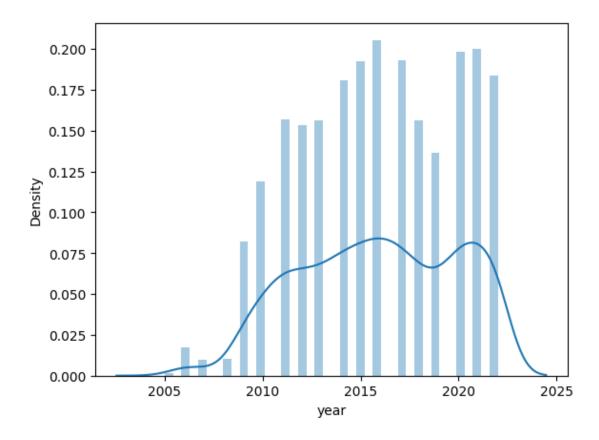
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sb.distplot(df['year'], bins=40)

[18]: <Axes: xlabel='year', ylabel='Density'>



[19]: df['year'].skew() #check the skewness

#### [19]: -0.17892940223703785

to reduce skewness we have 4 methods. 1. log

[20]: log = np.log(df['year'])
print(log.skew())

#### -0.18203775462443883

## [21]: sb.distplot(log, bins=40)

C:\Users\kumar\AppData\Local\Temp\ipykernel\_9060\3503255974.py:1: UserWarning:

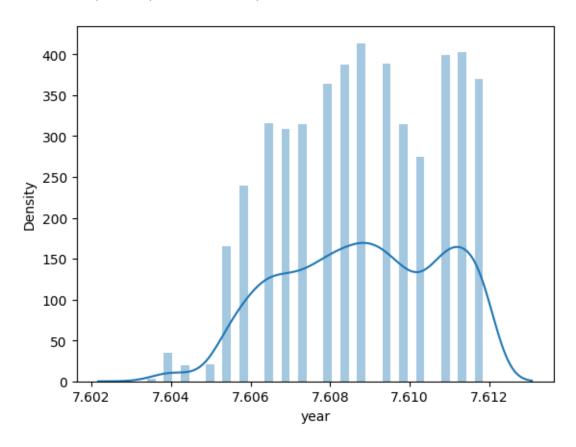
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sb.distplot(log, bins=40)

[21]: <Axes: xlabel='year', ylabel='Density'>



[]: