

Assignment 2 - Data Wrangling II (Data Science)

Atharva Taras (TE A - 73)

Create an “Academic performance” dataset of students and perform the following operations using Python.

Dataset - <https://www.kaggle.com/code/bhartiprasad17/student-academic-performance-analysis/input>
(<https://www.kaggle.com/code/bhartiprasad17/student-academic-performance-analysis/input>)

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

```
In [2]: 1 data = pd.read_csv(r'A:\Python Projects\College Practicals\DS\StudentsPerformance
```

```
In [3]: 1 data.head()
```

```
Out[3]:
```

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score
0	female	group B	bachelor's degree	standard	none	72	72	74
1	female	group C	some college	standard	completed	69	90	88
2	female	group B	master's degree	standard	none	90	95	93
3	male	group A	associate's degree	free/reduced	none	47	57	44
4	male	group C	some college	standard	none	76	78	75

```
In [4]: 1 data.shape
```

```
Out[4]: (1000, 8)
```

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.

```
In [5]: 1 data.isnull().sum()
```

```
Out[5]: gender                0
        race/ethnicity        0
        parental level of education  0
        lunch                  0
        test preparation course  0
        math score             0
        reading score          0
        writing score           0
        dtype: int64
```

```
In [6]: 1 data.info()
```

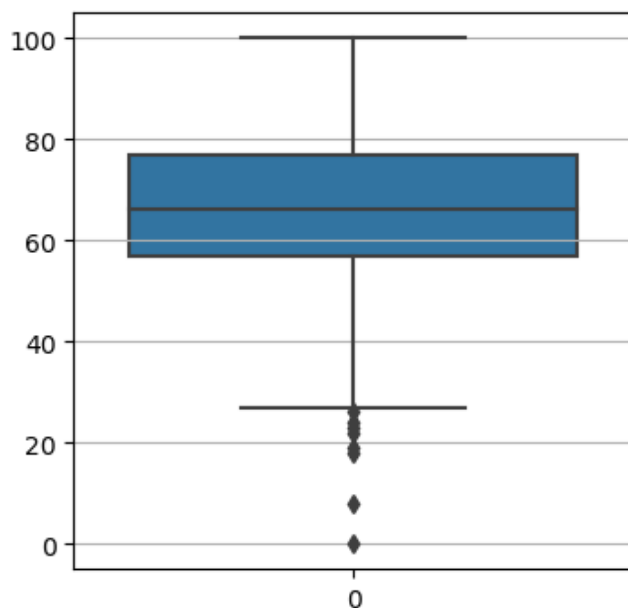
```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 1000 entries, 0 to 999  
Data columns (total 8 columns):  
#   Column                                Non-Null Count  Dtype  
---  ---                                -  
0   gender                                1000 non-null   object  
1   race/ethnicity                        1000 non-null   object  
2   parental level of education          1000 non-null   object  
3   lunch                                1000 non-null   object  
4   test preparation course              1000 non-null   object  
5   math score                           1000 non-null   int64  
6   reading score                        1000 non-null   int64  
7   writing score                         1000 non-null   int64  
dtypes: int64(3), object(5)  
memory usage: 62.6+ KB
```

No missing values/inconsistencies found

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

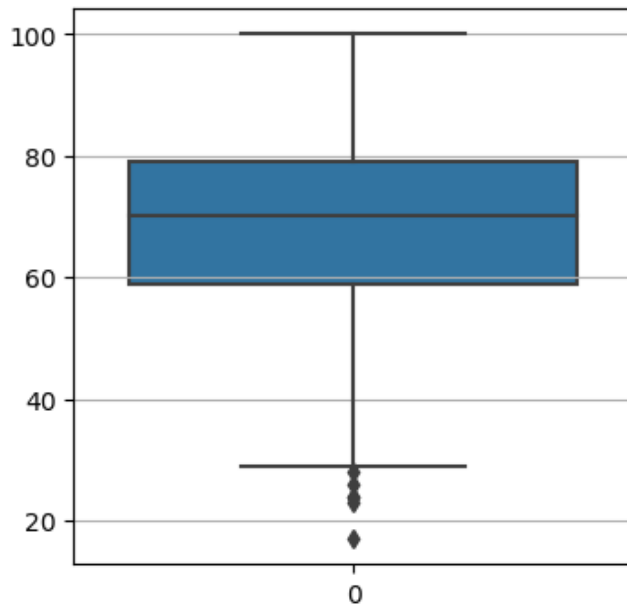
```
In [7]: 1 plt.figure(figsize=(4, 4))  
2 plt.grid()  
3 sns.boxplot(data=data['math score'])
```

Out[7]: <AxesSubplot: >



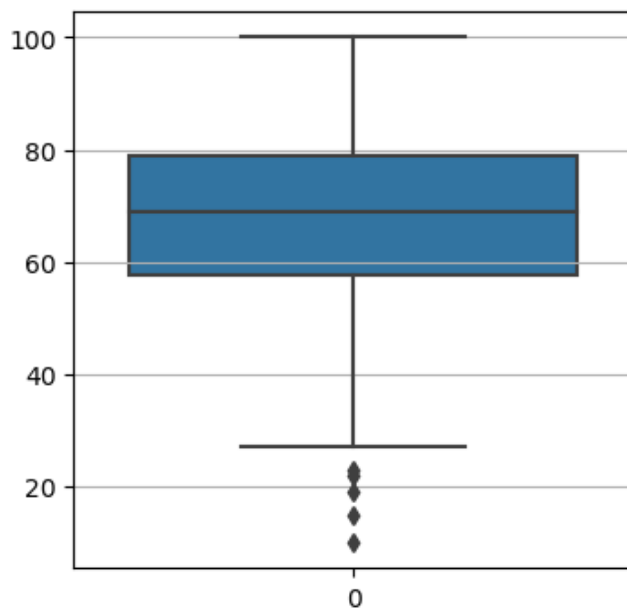
```
In [8]: 1 plt.figure(figsize=(4, 4))
        2 plt.grid()
        3 sns.boxplot(data=data['reading score'])
```

Out[8]: <AxesSubplot: >



```
In [9]: 1 plt.figure(figsize=(4, 4))
        2 plt.grid()
        3 sns.boxplot(data=data['writing score'])
```

Out[9]: <AxesSubplot: >



Since student marks can range from 0-100, outlier values cannot be dropped

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.

Adding a separate column for pass/fail based on aggregate score of three subjects - math, reading and writing

```
In [10]: 1 marks = []
2
3 for i in range(0, data.shape[0]):
4     tmp = []
5
6     tmp.append(data['math score'][i])
7     tmp.append(data['reading score'][i])
8     tmp.append(data['writing score'][i])
9
10    aggregate = (sum(tmp)/len(tmp)).round(2)
11
12    marks.append(aggregate)
```

```
In [11]: 1 data['Aggregate'] = marks
```

```
In [12]: 1 data.head(2)
```

```
Out[12]:
```

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score	Aggregate
0	female	group B	bachelor's degree	standard	none	72	72	74	72.67
1	female	group C	some college	standard	completed	69	90	88	82.33

Setting aggregate pass marks = 35

```
In [13]: 1 tmp1ist = []
2
3 for score in data['Aggregate']:
4
5     if score > 34:
6         tmp1ist.append(True)
7
8     else:
9         tmp1ist.append(False)
10
11 data['Passed'] = tmp1ist
```

```
In [14]: 1 data.head(2)
```

```
Out[14]:
```

	gender	race/ethnicity	parental level of education	lunch	test preparation course	math score	reading score	writing score	Aggregate	Passed
0	female	group B	bachelor's degree	standard	none	72	72	74	72.67	True
1	female	group C	some college	standard	completed	69	90	88	82.33	True

```
In [15]: 1 data.skew(numeric_only=True)
```

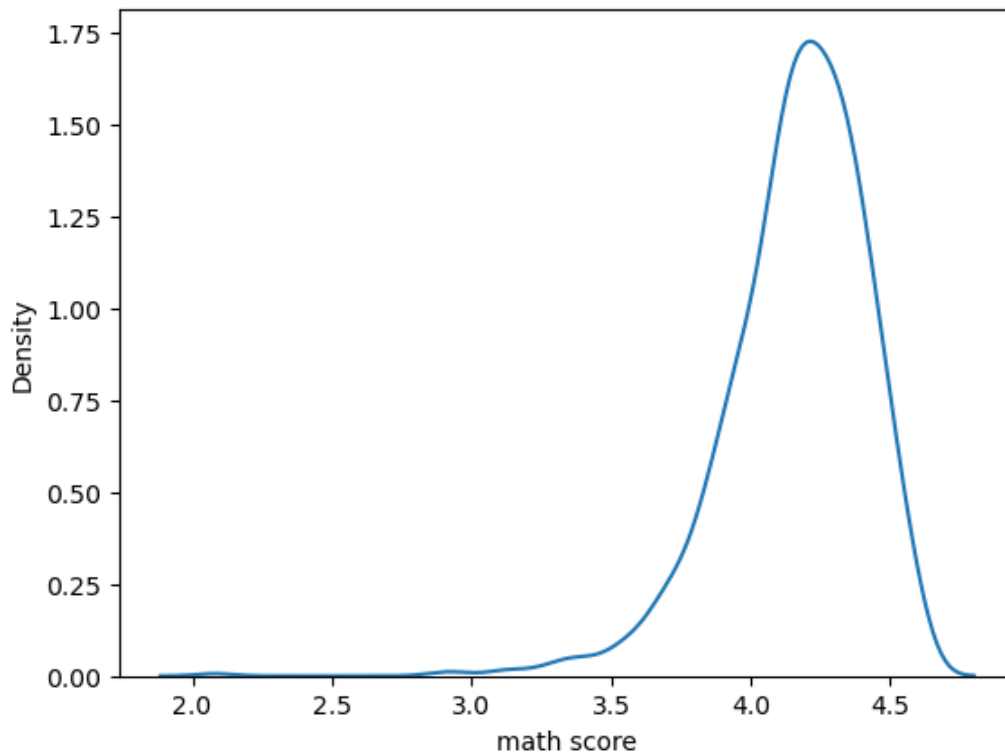
```
Out[15]: math score      -0.278935
reading score -0.259105
writing score -0.289444
Aggregate     -0.299042
Passed        -7.992087
dtype: float64
```

Converting to a normal distribution

```
In [16]: 1 t = np.log(data['math score'])
        2 sns.kdeplot(t)
```

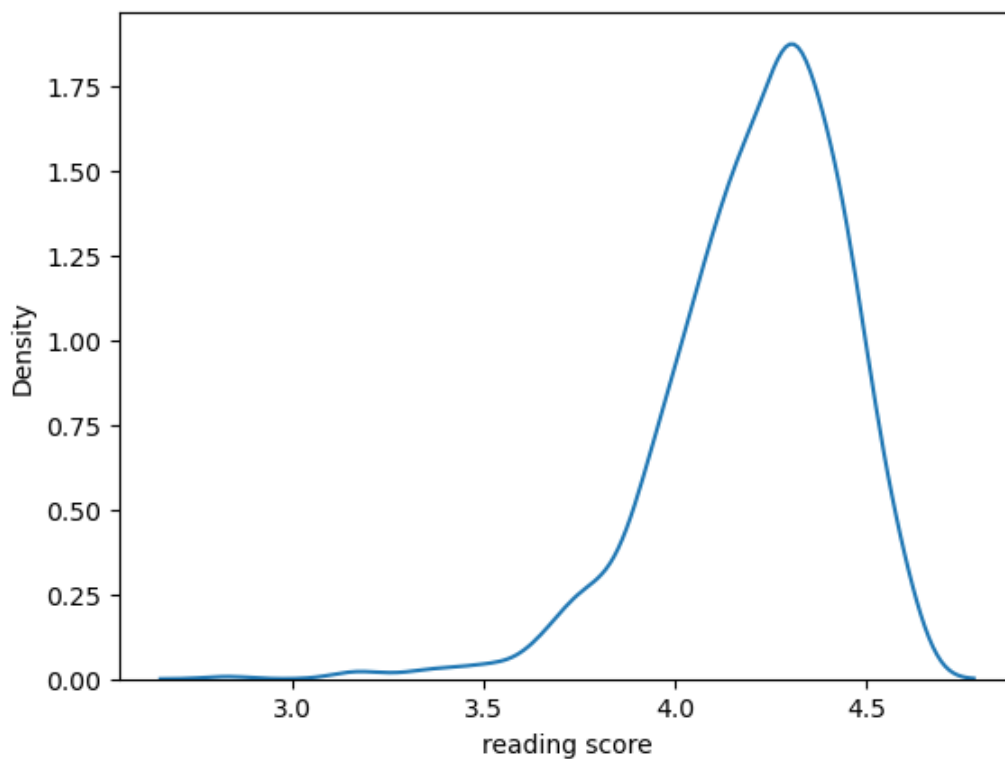
A:\Applications\Anaconda\lib\site-packages\pandas\core\arraylike.py:402: RuntimeWarning: divide by zero encountered in log
result = getattr(ufunc, method)(*inputs, **kwargs)

Out[16]: <AxesSubplot: xlabel='math score', ylabel='Density'>



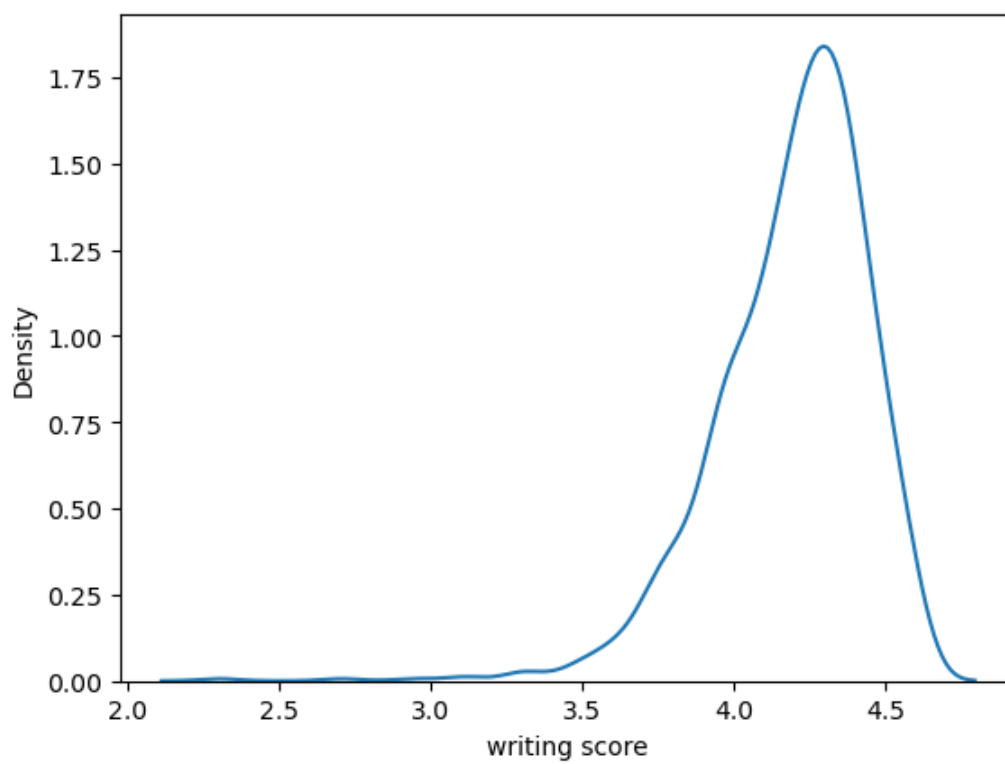
```
In [17]: 1 t = np.log(data['reading score'])
        2 sns.kdeplot(t)
```

Out[17]: <AxesSubplot: xlabel='reading score', ylabel='Density'>



```
In [18]: 1 t = np.log(data['writing score'])  
        2 sns.kdeplot(t)
```

Out[18]: <AxesSubplot: xlabel='writing score', ylabel='Density'>



```
In [ ]: 1
```