EXPLORATORY DATA ANALYSIS

#importing various libraries

import pandas as pd
import numpy as np
import seaborn as sns
import statistics as sc

#importing the csv file

data = pd.read_csv("students.csv")

(A) Understanding the Data

data.head()

0 1 2 3 4	gender female female female male	race/ethnici group group group group group) B) C) B) A	bach ma	l of education nelor's degree some college aster's degree ciate's degree some college	lunch \ standard standard standard free/reduced standard	
0 1 2 3 4	test pre	comple r r	ione	n score 72 69 90 47 76	reading score 72 90 95 57 78	writing score 74 88 93 44 75	
da	data.tail()						
<pre>gender race/ethnicity parental level of education lunch \</pre>					on		
99	•	e gro	oup E		master's degre	ee standard	
99	6 mal	e gro	oup C		high schoo	ol free/reduced	
99	7 femal	e gro	oup C		high schoo	ol free/reduced	
99	8 femal	e gro	oup D		some colle	ge standard	
99	9 femal	e gro	oup D		some colle	ge free/reduced	
	test p	reparation o	course ma	ath score	e reading sco	re writing score	
99	5	comp	leted	88	3 !	99 95	

996	none	62	55	55			
997	completed	59	71	65			
998	completed	68	78	77			
999	none	77	86	86			
data.shape	data.shape						
(1000, 8)							
data.nunique	e()						
gender 2 race/ethnicity 5 parental level of education 6 lunch 2 test preparation course 2 math score 81 reading score 72 writing score 77 dtype: int64							
<pre>couple_columns= data[['gender', 'math score', 'reading score', 'writing score']] couple_columns</pre>							
gender female female female male male male male male male male	math score reading 72 69 90 47 76 88 62 59 68 77	ng score writ: 72 90 95 57 78 99 55 71 78 86	ing score 74 88 93 44 75 95 55 65 77 86				

[1000 rows x 4 columns]

(B) Types of EDA with Examples

1. Univariate Non- Graphical #MEAN, STANDARD DEVIATION, QUARTILES

data.describe()

```
reading score
       math score
                                  writing score
count 1000.00000
                     1000.000000
                                    1000.000000
mean
         66.08900
                       69.169000
                                      68.054000
         15.16308
                       14.600192
                                      15.195657
std
min
         0.00000
                       17.000000
                                      10.000000
25%
         57.00000
                       59,000000
                                      57.750000
         66.00000
50%
                       70.000000
                                      69.000000
75%
         77.00000
                       79.000000
                                      79.000000
        100.00000
                      100.000000
                                     100.000000
max
#MEDIAN
math = data['math score'].median()
reading = data['reading score'].median()
writing = data['writing score'].median()
print('math score median = ', math)
print('reading score median = ', reading)
print('writing score median = ', writing )
math score median = 66.0
reading score median = 70.0
writing score median = 69.0
#MODE
math1 = data['math score'].mode()
reading1 = data['reading score'].mode()
writing1 = data['writing score'].mode()
print('math score mode = ', math1)
print('reading score mode = ', reading1)
print('writing score mode = ', writing1)
math score mode = 0
Name: math score, dtype: int64
reading score mode = 0
Name: reading score, dtvpe: int64
writing score mode = 0
Name: writing score, dtype: int64
#MEAN
math2 = data['math score'].mean()
reading2 = data['reading score'].mean()
writing2 = data['writing score'].mean()
print('math score mode = ', math2)
print('reading score mode = ', reading2)
print('writing score mode = ', writing2)
```

```
math score mode = 66.089
reading score mode = 69.169
writing score mode = 68.054
```

2. Multivariate Non - Graphical

M = pd.read_csv("students.csv")
M

lung		race/ethnicity p	parental leve	l of education	
lunc 0	female	group B	bac	helor's degree	standard
1	female	group C		some college	standard
2	female	group B	m	aster's degree	standard
3	male	group A	asso	ciate's degree	free/reduced
4	male	group C		some college	standard
995	female	group E	m	aster's degree	standard
996	male	group C		high school	free/reduced
997	female	group C		high school	free/reduced
998	female	group D		some college	standard
999	female	group D	D some college		free/reduced
	test pre	eparation course	math score	reading score	writing score
0		none	72	72	74
1		completed	69	90	88
2		none	90	95	93
3		none	47	57	44
4		none	76	78	75

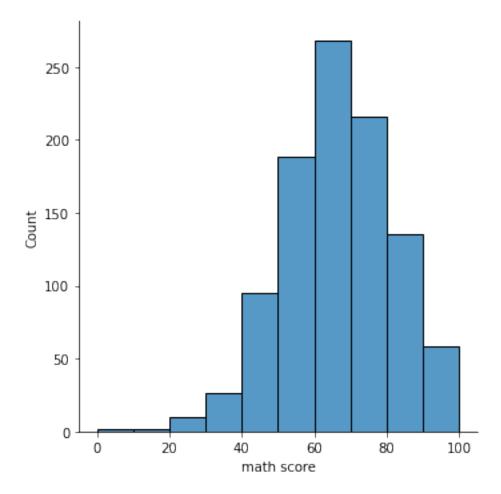
995	completed	88	99	95
996	none	62	55	55
997	completed	59	71	65
998	completed	68	78	77
999	none	77	86	86

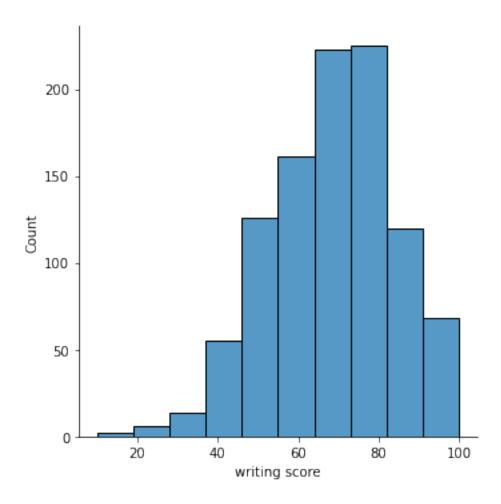
```
[1000 rows x 8 columns]
cross_tab = pd.crosstab(M.gender,M.lunch, normalize = 'index')
cross_tab
lunch free/reduced standard
gender
female    0.364865    0.635135
male    0.344398    0.655602
```

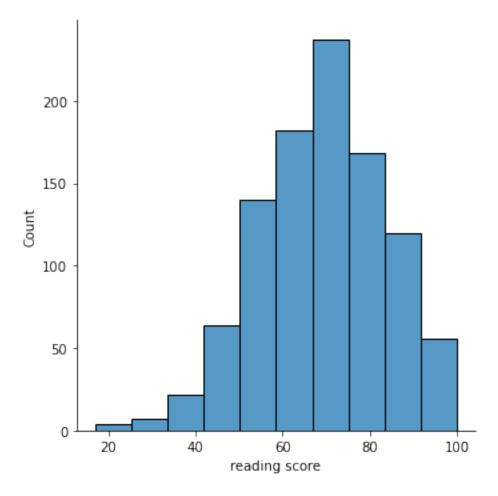
3. Univariate Graphical

#HISTOGRAM

```
sns.displot(data['math score'], bins = 10)
sns.displot(data['writing score'], bins = 10)
sns.displot(data['reading score'], bins = 10)
<seaborn.axisgrid.FacetGrid at 0x7fa4f0c8a970>
```



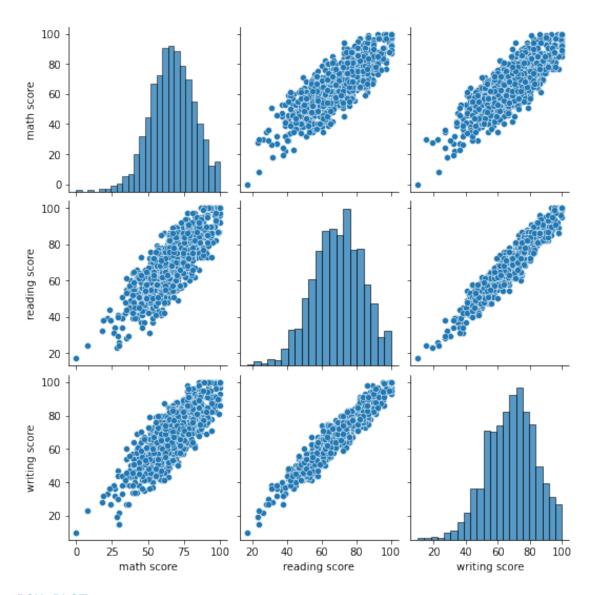




#PAIR PLOT

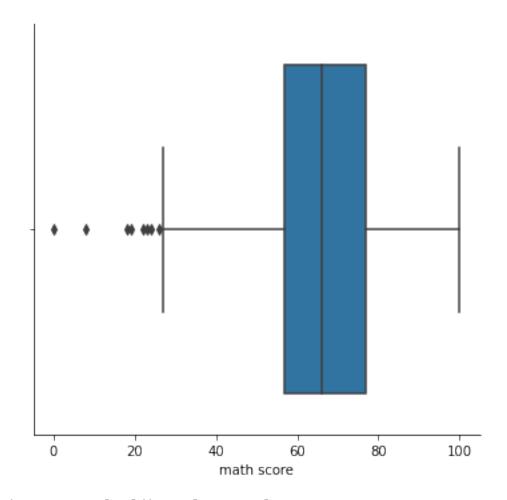
sns.pairplot(data)

<seaborn.axisgrid.PairGrid at 0x7fa5231a24c0>



#BOX PLOT

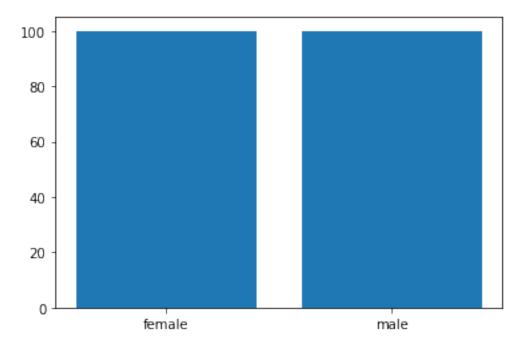
sns.catplot(x = 'math score', kind = 'box', data= data)
<seaborn.axisgrid.FacetGrid at 0x7fa512941790>



import matplotlib.pyplot as plt

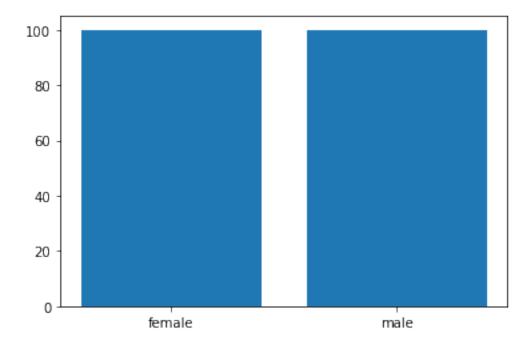
```
#BAR GRAPHS
x = data['gender']
y = data['math score']
plt.bar(x,y)
```

<BarContainer object of 1000 artists>



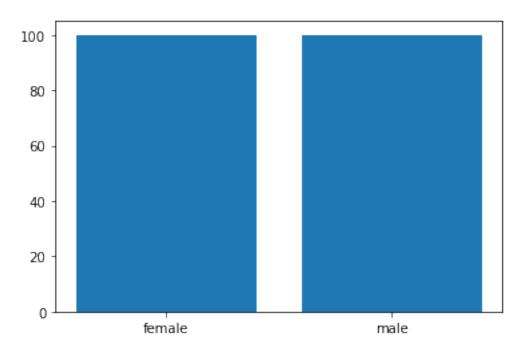
```
x = data['gender']
y = data['reading score']
plt.bar(x,y)
```

<BarContainer object of 1000 artists>



```
x = data['gender']
y = data['writing score']
plt.bar(x,y)
```

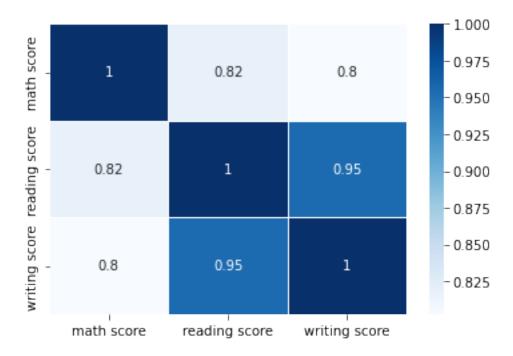
<BarContainer object of 1000 artists>



4. Multivariate Graphical #HEAT MAP

sns.heatmap(data.corr(), annot = True, linewidth = 0.5, cmap =
'Blues')

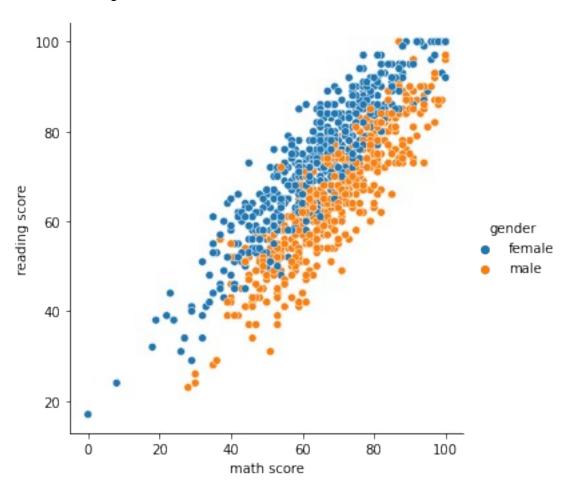
<AxesSubplot:>



#SCATTER PLOT

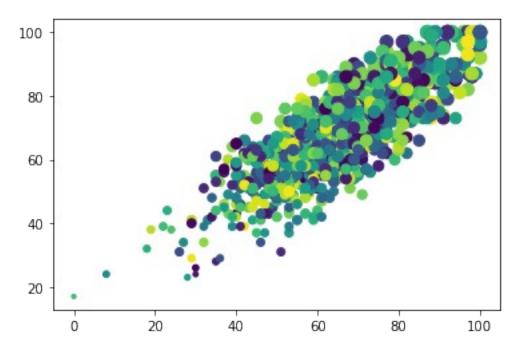
```
sns.relplot(x= 'math score', y = 'reading score', hue = 'gender',
data= data)
```

<seaborn.axisgrid.FacetGrid at 0x7fa5015ce070>



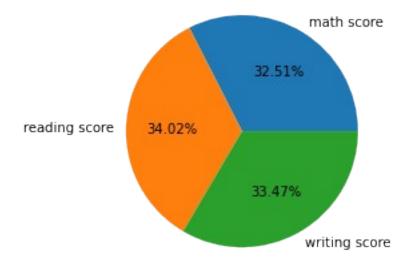
#BUBBLE CHART

```
m = data['math score']
r = data['reading score']
w = data['writing score']
colours = np.random.rand(1000)
plt.scatter(m,r,w, c = colours)
plt.show()
```



#Pie Chart
#(for the average score in all subjects)

```
val_ue = (math2, reading2, writing2)
lab_el = ('math score', 'reading score', 'writing score')
plt.pie(val_ue, labels = lab_el, autopct = '%2.2f%%')
plt.show()
```



(C) CONCLUSION

- 1. About the Data: A sample data set 'students.csv' was downloaded from Kaggle. The data set gives basic information related to 999 students along with marks in three different domains. A total of 8 columns make up the data set.
- 2. Exploratory data analysis was performed on the data and the following conclusions were drawn:

####2.1) The entire class of students performed the best in Reading followed by Writing and then Math. The lowest mark in math is 0, for the other two subjects, the lowest mark > 0. Hence, the mean for maths is the least. However, in each subject, the maximum marks obtained is 100 [Mean and Pie Chart]

####2.2)Most people have scored between 60 to 70 marks in Maths, 60-80
in case of Writing and 70 - 80 in case of Reading [Mode and Histogram]
####2.3)No gender disparity was seen in terms of scoring highest marks
in any subject [Bar Graph]

####2.4)However, female collectivey have performed better in case of Reading as compared to Maths. For male, the relationship is vice-versa [Scatter Plot]

####2.5)Majority of the class population has scored above 40 in all subjects. Implying that, if we assume 40 to be the cut off marks for qualifying the exams, the majority of class has passed the exams [Bubble Chart]