

student-result-analysis-project

October 29, 2024

```
[360]: #Python Project On Student Result Analysis
import seaborn as sns
import pandas as pd
from matplotlib import pyplot as plt
```

```
[362]: stu=pd.read_csv("Student_Scores.csv")
stu.head(10)
```

```
[362]:
```

	Unnamed: 0	Gender	EthnicGroup	ParentEduc	LunchType	\
0	0	female	NaN	bachelor's degree	standard	
1	1	female	group C	some college	standard	
2	2	female	group B	master's degree	standard	
3	3	male	group A	associate's degree	free/reduced	
4	4	male	group C	some college	standard	
5	5	female	group B	associate's degree	standard	
6	6	female	group B	some college	standard	
7	7	male	group B	some college	free/reduced	
8	8	male	group D	high school	free/reduced	
9	9	female	group B	high school	free/reduced	

	TestPrep	ParentMaritalStatus	PracticeSport	IsFirstChild	NrSiblings	\
0	none	married	regularly	yes	3.0	
1	NaN	married	sometimes	yes	0.0	
2	none	single	sometimes	yes	4.0	
3	none	married	never	no	1.0	
4	none	married	sometimes	yes	0.0	
5	none	married	regularly	yes	1.0	
6	completed	widowed	never	no	1.0	
7	none	married	sometimes	yes	1.0	
8	completed	single	sometimes	no	3.0	
9	none	married	regularly	yes	NaN	

	TransportMeans	WklyStudyHours	MathScore	ReadingScore	WritingScore
0	school_bus	< 5	71	71	74
1	NaN	5 - 10	69	90	88
2	school_bus	< 5	87	93	91
3	NaN	5 - 10	45	56	42

4	school_bus	5 - 10	76	78	75
5	school_bus	5 - 10	73	84	79
6	private	5 - 10	85	93	89
7	private	> 10	41	43	39
8	private	> 10	65	64	68
9	private	< 5	37	59	50

```
[363]: stu=pd.read_csv("Student_Scores.csv")
print(stu.head())
```

	Unnamed: 0	Gender	EthnicGroup	ParentEduc	LunchType	TestPrep	\
0	0	female	NaN	bachelor's degree	standard	none	
1	1	female	group C	some college	standard	NaN	
2	2	female	group B	master's degree	standard	none	
3	3	male	group A	associate's degree	free/reduced	none	
4	4	male	group C	some college	standard	none	

	ParentMaritalStatus	PracticeSport	IsFirstChild	NrSiblings	TransportMeans	\
0	married	regularly	yes	3.0	school_bus	
1	married	sometimes	yes	0.0	NaN	
2	single	sometimes	yes	4.0	school_bus	
3	married	never	no	1.0	NaN	
4	married	sometimes	yes	0.0	school_bus	

	WklyStudyHours	MathScore	ReadingScore	WritingScore
0	< 5	71	71	74
1	5 - 10	69	90	88
2	< 5	87	93	91
3	5 - 10	45	56	42
4	5 - 10	76	78	75

```
[366]: #To Check The Null Values
stu.isnull().sum()
```

```
[366]: Unnamed: 0      0
Gender      0
EthnicGroup 1840
ParentEduc  1845
LunchType   0
TestPrep    1830
ParentMaritalStatus 1190
PracticeSport 631
IsFirstChild 904
NrSiblings  1572
TransportMeans 3134
WklyStudyHours 955
MathScore    0
```

```
ReadingScore      0
WritingScore      0
dtype: int64
```

```
[368]: #So Their Is Null Values Present That We Have To Remove
```

```
[370]: stu=stu.drop("Unnamed: 0",axis=1)
print(stu.head())
```

	Gender	EthnicGroup	ParentEduc	LunchType	TestPrep	\
0	female	NaN	bachelor's degree	standard	none	
1	female	group C	some college	standard	NaN	
2	female	group B	master's degree	standard	none	
3	male	group A	associate's degree	free/reduced	none	
4	male	group C	some college	standard	none	

	ParentMaritalStatus	PracticeSport	IsFirstChild	NrSiblings	TransportMeans	\
0	married	regularly	yes	3.0	school_bus	
1	married	sometimes	yes	0.0	NaN	
2	single	sometimes	yes	4.0	school_bus	
3	married	never	no	1.0	NaN	
4	married	sometimes	yes	0.0	school_bus	

	WklyStudyHours	MathScore	ReadingScore	WritingScore
0	< 5	71	71	74
1	5 - 10	69	90	88
2	< 5	87	93	91
3	5 - 10	45	56	42
4	5 - 10	76	78	75

```
[372]: stu.head(10)
```

```
[372]:
```

	Gender	EthnicGroup	ParentEduc	LunchType	TestPrep	\
0	female	NaN	bachelor's degree	standard	none	
1	female	group C	some college	standard	NaN	
2	female	group B	master's degree	standard	none	
3	male	group A	associate's degree	free/reduced	none	
4	male	group C	some college	standard	none	
5	female	group B	associate's degree	standard	none	
6	female	group B	some college	standard	completed	
7	male	group B	some college	free/reduced	none	
8	male	group D	high school	free/reduced	completed	
9	female	group B	high school	free/reduced	none	

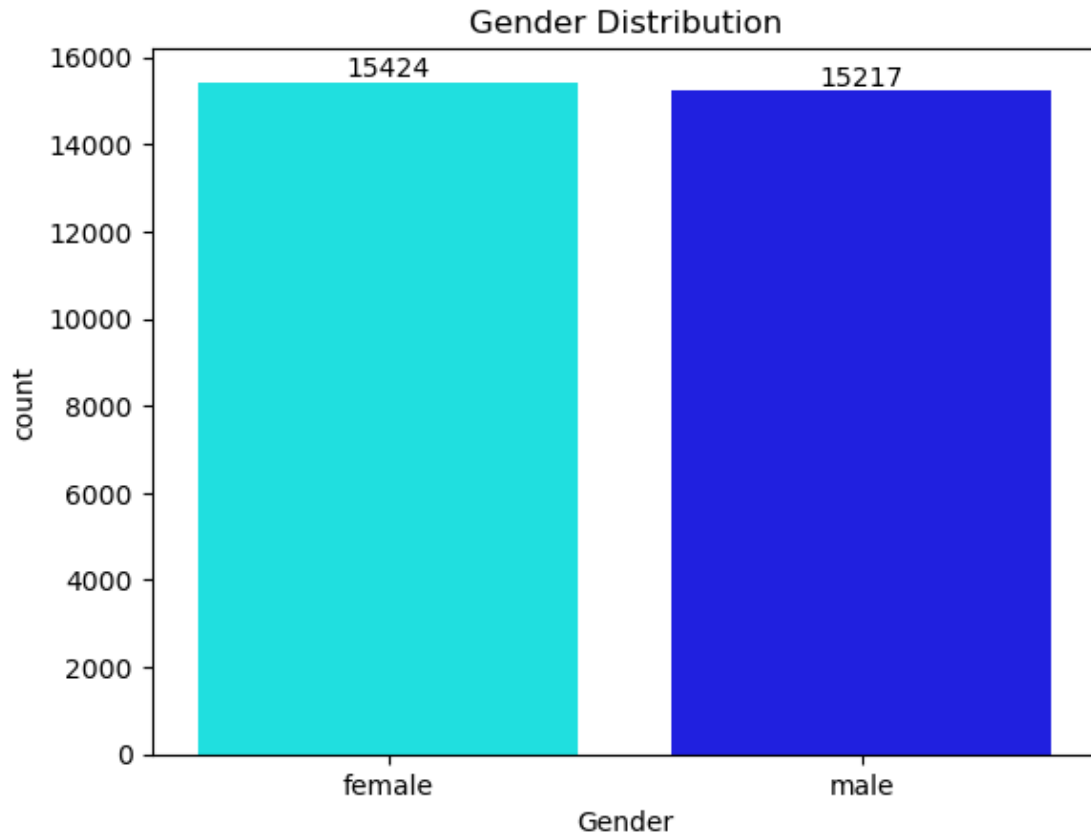
	ParentMaritalStatus	PracticeSport	IsFirstChild	NrSiblings	TransportMeans	\
0	married	regularly	yes	3.0	school_bus	
1	married	sometimes	yes	0.0	NaN	

2	single	sometimes	yes	4.0	school_bus
3	married	never	no	1.0	NaN
4	married	sometimes	yes	0.0	school_bus
5	married	regularly	yes	1.0	school_bus
6	widowed	never	no	1.0	private
7	married	sometimes	yes	1.0	private
8	single	sometimes	no	3.0	private
9	married	regularly	yes	NaN	private

	WklyStudyHours	MathScore	ReadingScore	WritingScore
0	< 5	71	71	74
1	5 - 10	69	90	88
2	< 5	87	93	91
3	5 - 10	45	56	42
4	5 - 10	76	78	75
5	5 - 10	73	84	79
6	5 - 10	85	93	89
7	> 10	41	43	39
8	> 10	65	64	68
9	< 5	37	59	50

[374]: *#Here You Can See That Unnamed Column Is Removed*

```
[376]: plt.title("Gender Distribution")
cp=sns.countplot(data=stu ,x="Gender",hue="Gender",palette=["cyan","blue"])
cp.bar_label(cp.containers[0])
cp.bar_label(cp.containers[1])
plt.show()
```



[378]: *#Student Gender Bar Graph*

[380]: *#Making The Group Of Mean of MathScore,ReadingScore,WritingScore In ParentEduc*
↪Column

[382]: `stu1=stu.groupby("ParentEduc").agg({"MathScore":"mean","ReadingScore":
 ↪"mean","WritingScore":"mean","WritingScore":"mean"})`
`stu1`

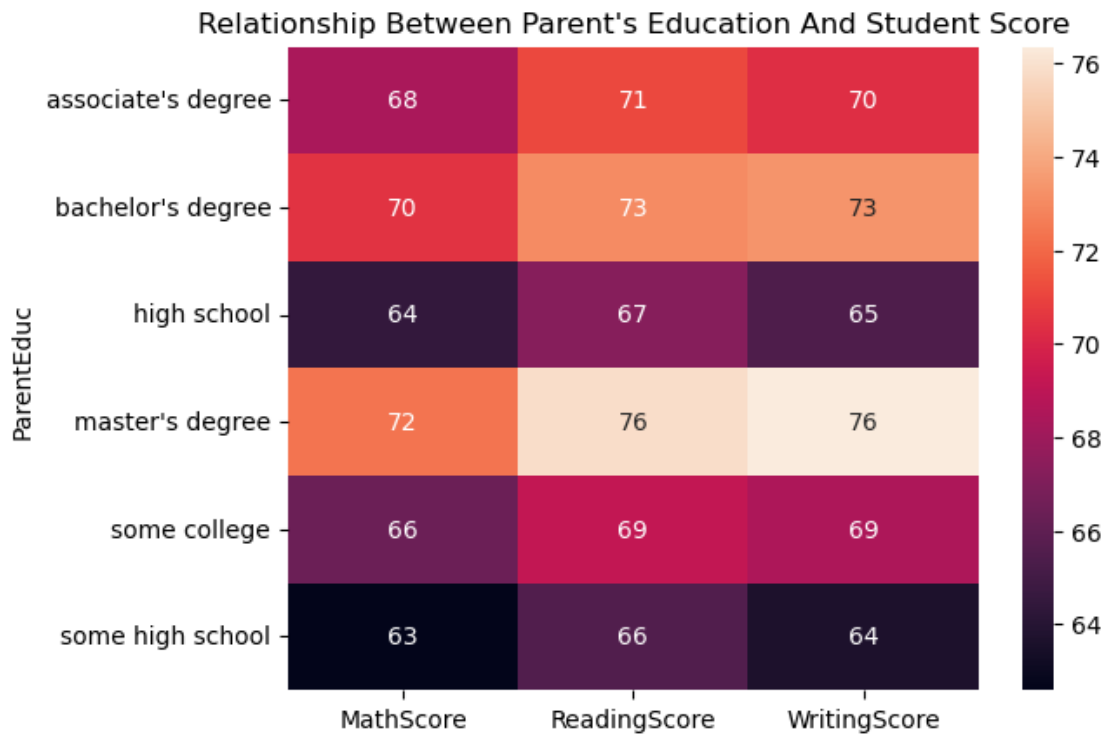
[382]:

	MathScore	ReadingScore	WritingScore
ParentEduc			
associate's degree	68.365586	71.124324	70.299099
bachelor's degree	70.466627	73.062020	73.331069
high school	64.435731	67.213997	65.421136
master's degree	72.336134	75.832921	76.356896
some college	66.390472	69.179708	68.501432
some high school	62.584013	65.510785	63.632409

[384]: *#Here Is The Aggregation*

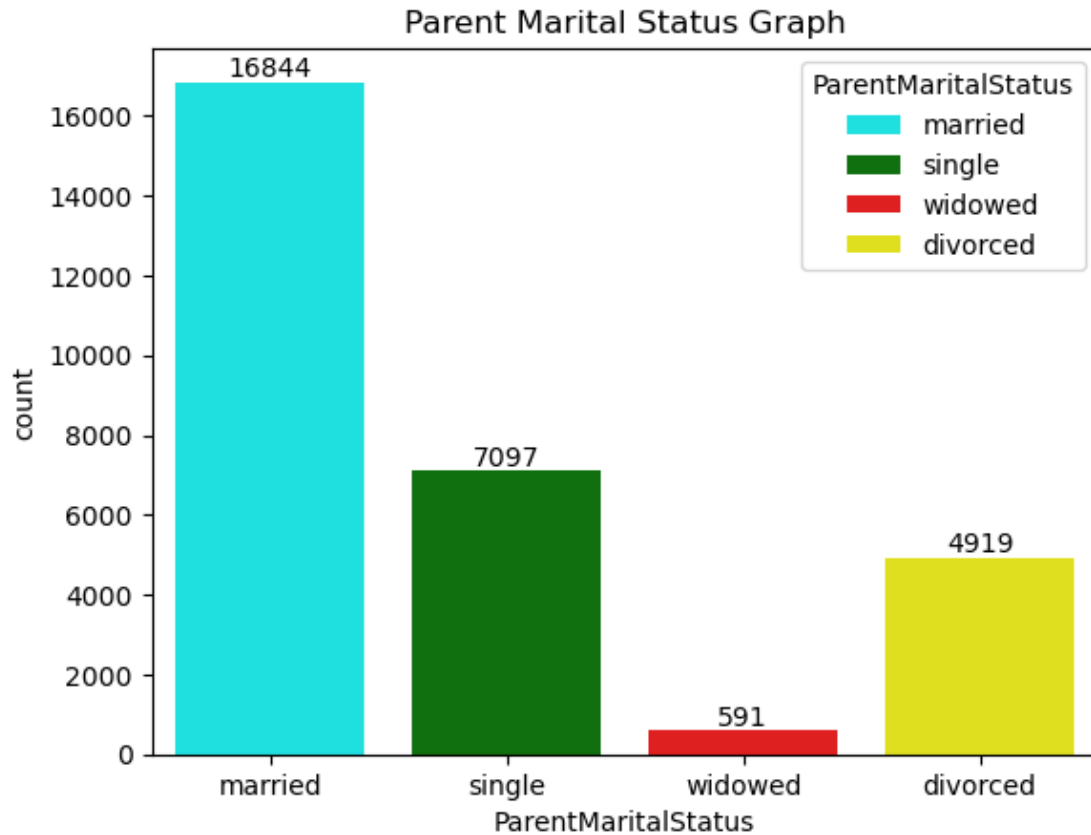
```
[386]: #Lets Make Heatmap
```

```
[388]: plt.title("Relationship Between Parent's Education And Student Score")
sns.heatmap(stu1,annot=True)
plt.show()
```



```
[390]: #By Watching heatmap We Can Say That Education Of Parents Impacting On The
↳Child Education
```

```
[392]: #Making The Group Of Mean of MathScore,ReadingScore,WritingScore In
↳ParentMaritalStatus Column
plt.title("Parent Marital Status Graph")
ms=sns.
↳countplot(data=stu,x="ParentMaritalStatus",hue="ParentMaritalStatus",palette=["cyan",
↳"green", "red", "yellow"])
ms.bar_label(ms.containers[0])
ms.bar_label(ms.containers[1])
ms.bar_label(ms.containers[2])
ms.bar_label(ms.containers[3])
plt.show()
```

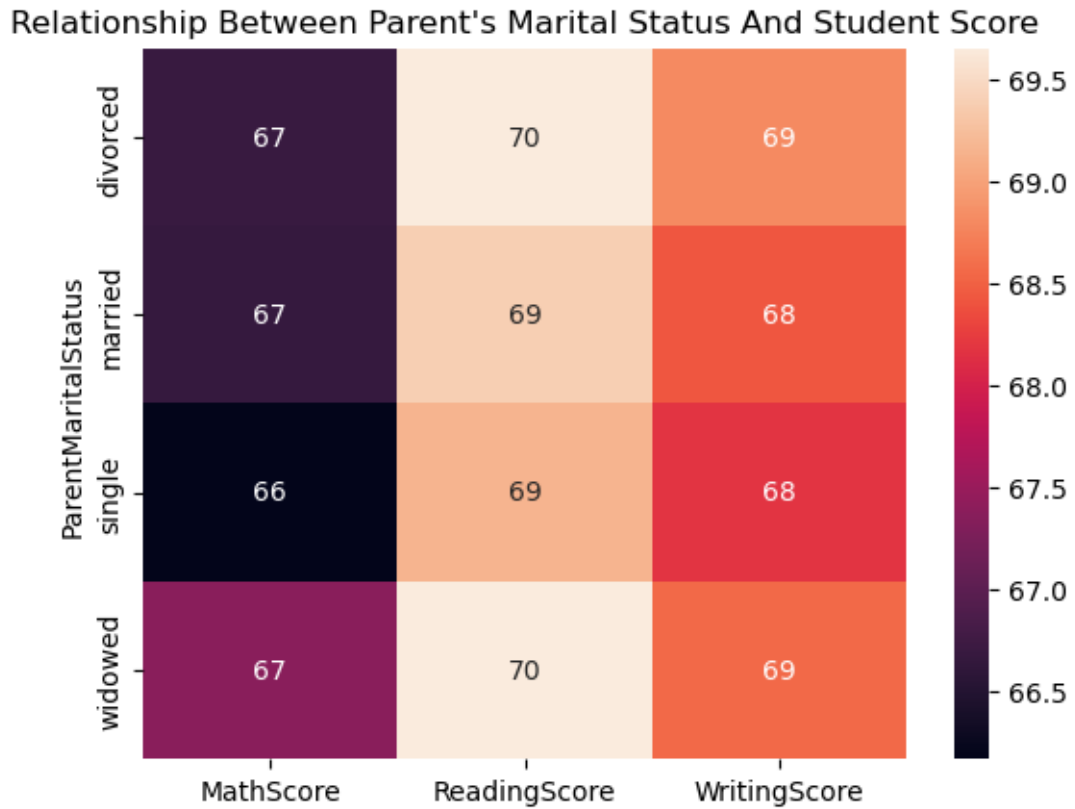


```
[394]: stu2=stu.groupby("ParentMaritalStatus").agg({"MathScore":"mean","ReadingScore":
↪"mean","WritingScore":"mean","WritingScore":"mean"})
stu2
```

```
[394]:
```

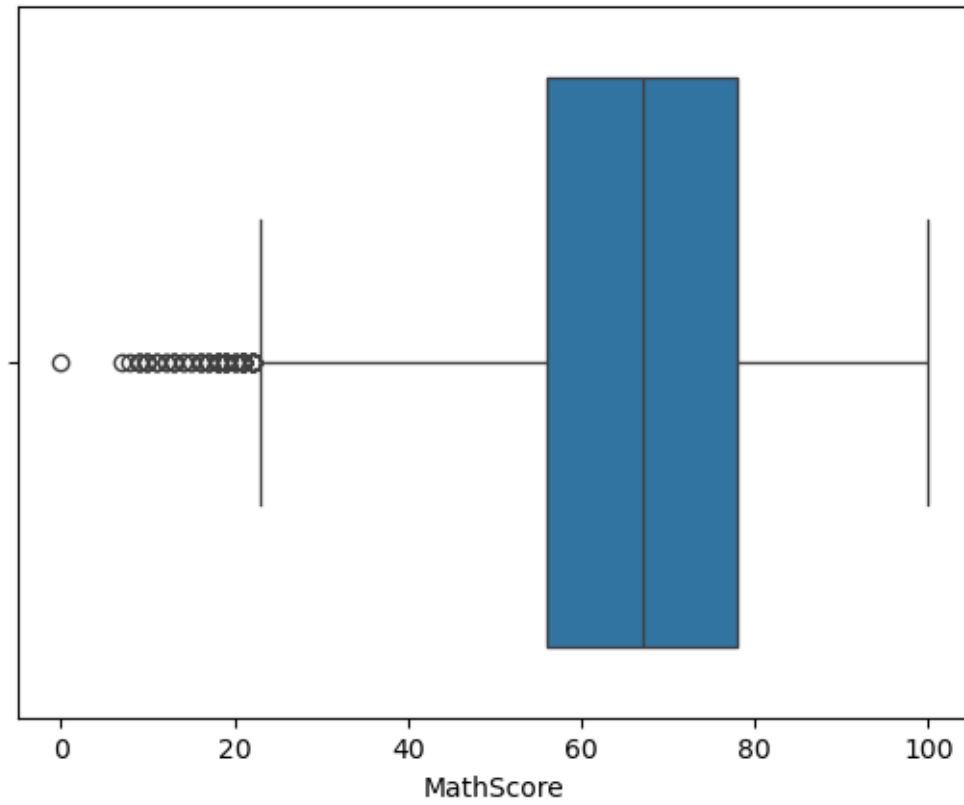
ParentMaritalStatus	MathScore	ReadingScore	WritingScore
divorced	66.691197	69.655011	68.799146
married	66.657326	69.389575	68.420981
single	66.165704	69.157250	68.174440
widowed	67.368866	69.651438	68.563452

```
[396]: plt.title("Relationship Between Parent's Marital Status And Student Score")
sns.heatmap(stu2,annot=True)
plt.show()
```

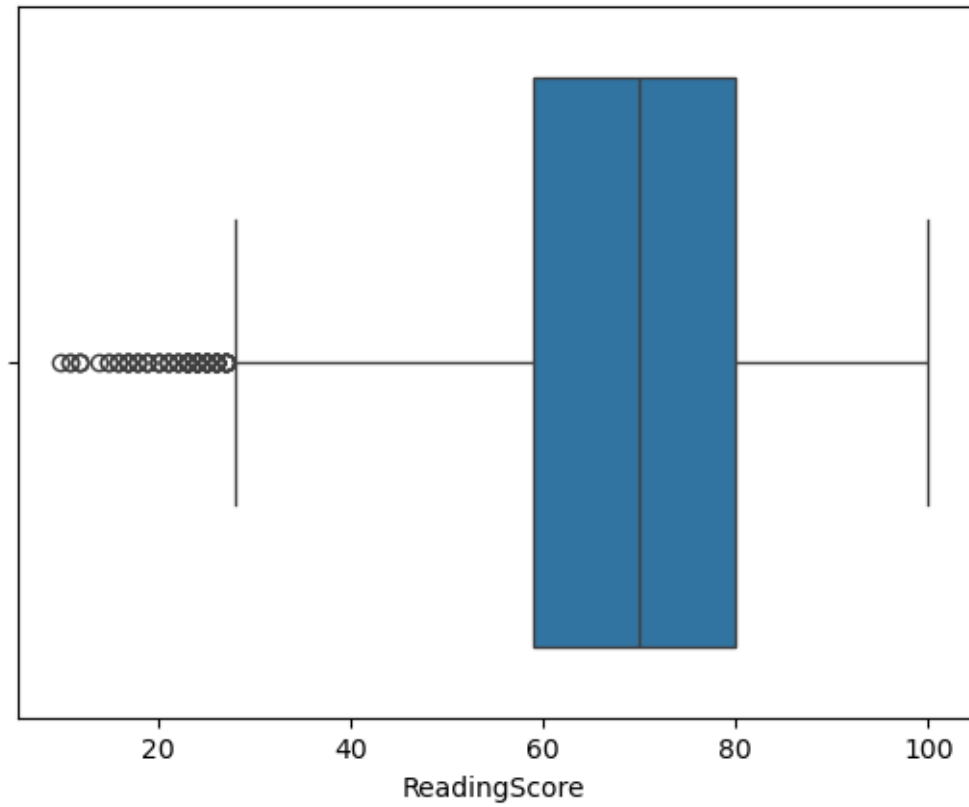


```
[398]: #By Watching heatmap We Can Say That Education Of MaritalStatusParents  
       ↪ Impacting On The Child Eduction
```

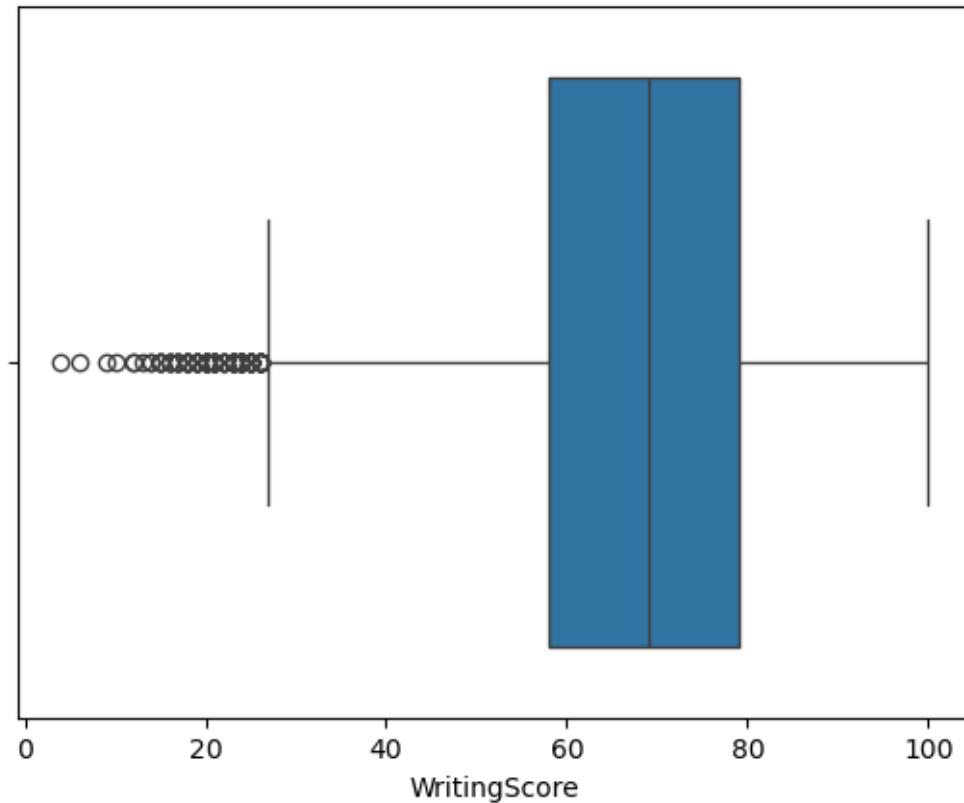
```
[400]: sns.boxplot(data=stu,x="MathScore")  
       plt.show()
```

```
[402]: sns.boxplot(data=stu,x="ReadingScore")  
plt.show()
```



```
[404]: sns.boxplot(data=stu,x="WritingScore")  
plt.show()
```



```
[406]: #Boxplots Of MathScore,ReadingScore And WritingScore
```

```
[408]: #Now We Have To Check Unique Values Of Ethnic Groups For Distributions
```

```
[301]: print(stu["EthnicGroup"].unique())
```

```
[nan 'group C' 'group B' 'group A' 'group D' 'group E']
```

```
[326]: #now to check the students list present only in group A By Iloc Function
grpA=stu.loc[(stu["EthnicGroup"]=="group A")].count()
grpA
```

```
[326]: Gender          2219
EthnicGroup          2219
ParentEduc           2078
LunchType            2219
TestPrep             2081
ParentMaritalStatus  2121
PracticeSport        2167
IsFirstChild         2168
NrSiblings           2096
```

```

TransportMeans      1999
WklyStudyHours      2146
MathScore           2219
ReadingScore        2219
WritingScore        2219
dtype: int64

```

```
[321]: #Here Is The Student Info Present In Group A
```

```

[ ]: #lets Make loc Of All Remaining Groups we only want the "ethnicGroup" in pie_
    ↪ chart so for that we
    #merge all the groups in 1 list that makes easy to make a pie chart

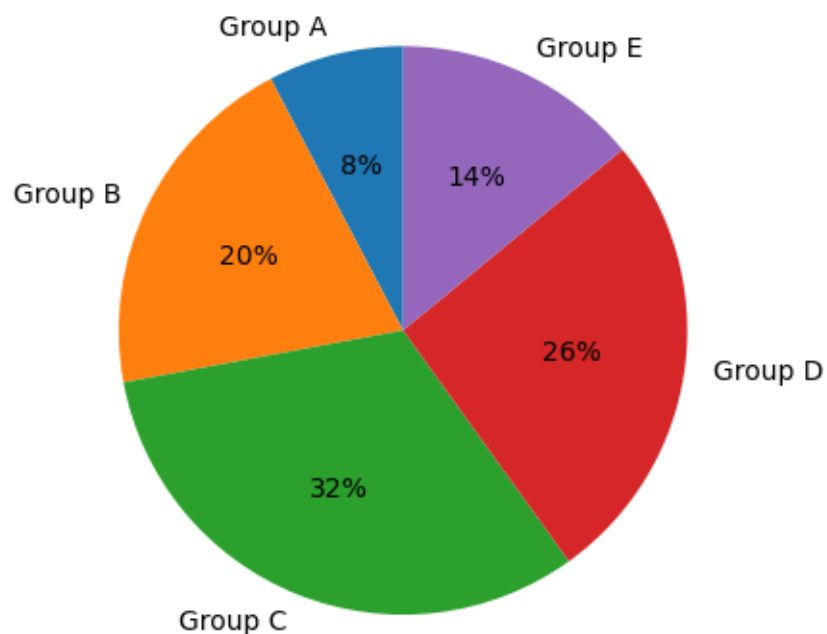
```

```

[358]: grpA=stu.loc[(stu["EthnicGroup"]=="group A")].count()
      grpB=stu.loc[(stu["EthnicGroup"]=="group B")].count()
      grpC=stu.loc[(stu["EthnicGroup"]=="group C")].count()
      grpD=stu.loc[(stu["EthnicGroup"]=="group D")].count()
      grpE=stu.loc[(stu["EthnicGroup"]=="group E")].count()

      lb=["Group A","Group B","Group C","Group D","Group E"]
      mergelist=[grpA["EthnicGroup"],grpB["EthnicGroup"],grpC["EthnicGroup"],grpD["EthnicGroup"],grpE["EthnicGroup"]]
      plt.pie(mergelist,labels=lb,autopct="%0.1f%%",startangle=90)
      plt.show()

```



[]: *#Here Is The Final Pie Chart Of Ethnic Groups Ob The Basis Of 5 Groups!!!*