

EDA OF STUDENTS PERFORMANCE EXAM DATASET

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

```
[2]: df=pd.read_csv("D:\stud.csv")
df.head()
```

```
[2]:
```

	gender	race_ethnicity	parental_level_of_education	lunch	
0	female	group B	bachelor's degree	standard	
1	female	group C	some college	standard	
2	female	group B	master's degree	standard	
3	male	group A	associate's degree	free/reduced	
4	male	group C	some college	standard	

	test_preparation_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

```
[3]: ## Summary of the dataset
df.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

```

```

[4]: ## checking Descriptive Statistic Summary of dataset
df.describe()

```

```

[4]:      math_score  reading_score  writing_score
count  1000.00000    1000.000000    1000.000000
mean     66.08900      69.169000     68.054000
std      15.16308     14.600192     15.195657
min       0.00000     17.000000     10.000000
25%      57.00000     59.000000     57.750000
50%      66.00000     70.000000     69.000000
75%      77.00000     79.000000     79.000000
max     100.00000    100.000000    100.000000

```

Isights And Observation 1. From the above description of numerical data, all means values are very close to each other between 66 and 69 2. All the standard deviation close to each other between 14 and 15 3. While there is minimum 0 for maths some other having 10 and 17 value

```

[5]: ## List down all the dataset column names
df.columns

```

```

[5]: Index(['gender', 'race_ethnicity', 'parental_level_of_education', 'lunch',
          'test_preparation_course', 'math_score', 'reading_score',
          'writing_score'],
          dtype='object')

```

```

[6]: ## Missing values in the dataset
df.isnull().sum()

```

```

[6]: 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

```

There are no null or missing values

```

[7]: ## Duplicates Record
df.duplicated()

```

```

[7]: 0    False
     1    False

```

```

2      False
3      False
4      False
...
995    False
996    False
997    False
998    False
999    False
Length: 1000, dtype: bool

```

```
[ ]: There are no duplicates values
```

```
[10]: ## For sseeing the duplicated records values
df[df.duplicated()]
```

```
[10]: Empty DataFrame
Columns: [gender, race_ethnicity, parental_level_of_education, lunch,
test_preparation_course, math_score, reading_score, writing_score]
Index: []
```

```
[11]: ## Remove the duplicates
df.drop_duplicates(inplace=True)
```

```
[12]: ## Checking the number of unqiues values of each columns
df.nunique()
```

```
[12]: 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
```

```
[13]: [feature for feature in df.columns if df[feature].dtype=='O']
```

```
[13]: ['gender',
      'race_ethnicity',
      'parental_level_of_education',
      'lunch',
      'test_preparation_course']
```

```
[34]: # Seggregate numerical and categorical values
numerical_features=[feature for feature in df.columns if df[feature].dtype!='O']
```

```
categorical_feature=[feature for feature in df.columns if df[feature].
↳dtype=='O']
```

```
[35]: numerical_features
```

```
[35]: ['math_score', 'reading_score', 'writing_score']
```

```
[36]: categorical_feature
```

```
[36]: ['gender',
      'race_ethnicity',
      'parental_level_of_education',
      'lunch',
      'test_preparation_course']
```

```
[37]: ## Aggregate the total score with mean
df['total_score']=(df['math_score']+df['reading_score']+df['writing_score'])
df['average']=df['total_score']/3
df.head()
```

```
[37]:
```

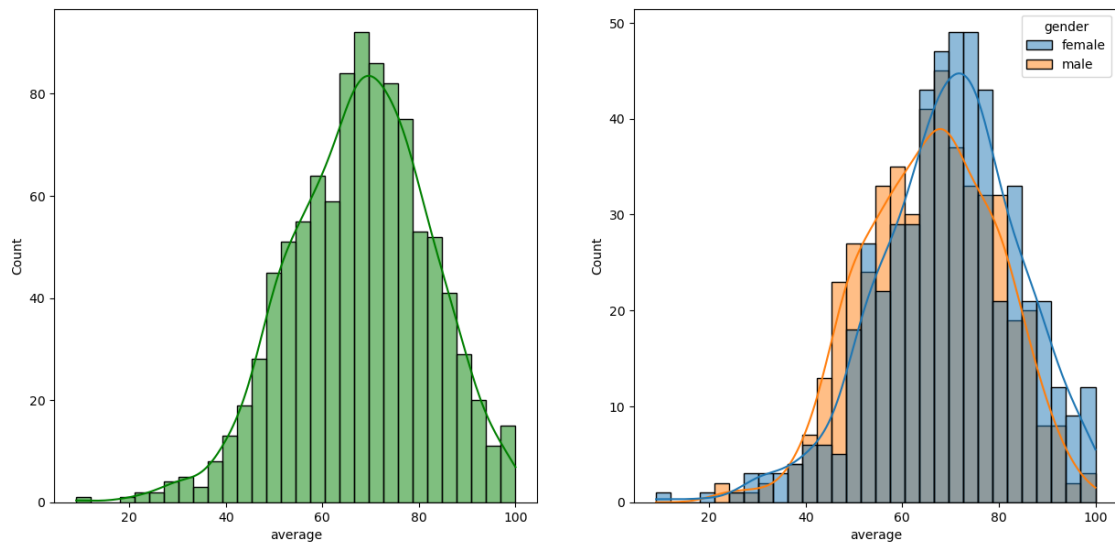
	gender	race_ethnicity	parental_level_of_education	lunch	\
0	female	group B	bachelor's degree	standard	
1	female	group C	some college	standard	
2	female	group B	master's degree	standard	
3	male	group A	associate's degree	free/reduced	
4	male	group C	some college	standard	

	test_preparation_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	

	total_score	average
0	218	72.666667
1	247	82.333333
2	278	92.666667
3	148	49.333333
4	229	76.333333

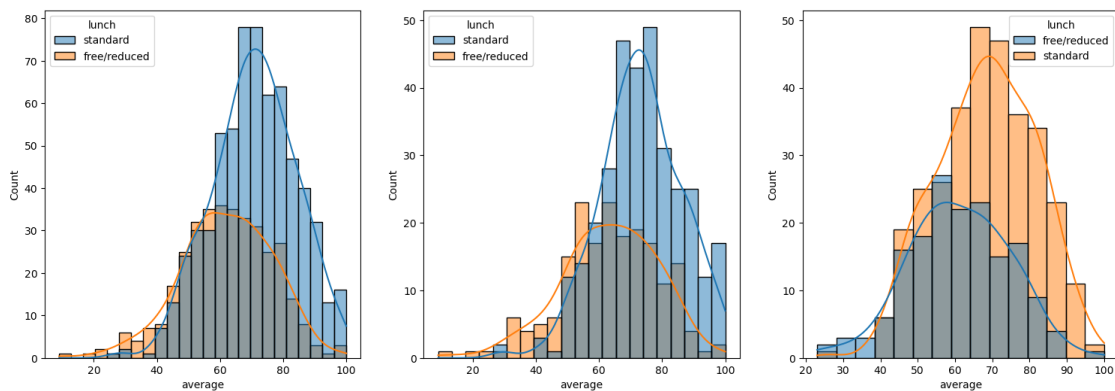
```
[38]: ## Explore More Visualization
fig,axis=plt.subplots(1,2,figsize=(15,7))
plt.subplot(121)
sns.histplot(data=df,x='average',bins=30,kde=True,color='g')
plt.subplot(122)
sns.histplot(data=df,x='average',bins=30,kde=True,hue='gender')
```

```
plt.show()
```



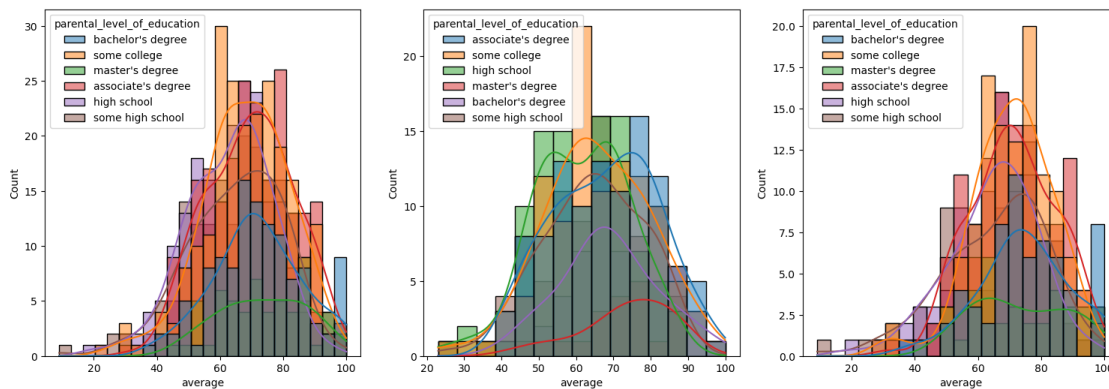
Insights 1. Female students tends to perform well than male students

```
[39]: plt.subplots(1,3,figsize=(25,6))
plt.subplot(141)
sns.histplot(data=df,x='average',kde=True,hue='lunch')
plt.subplot(142)
sns.histplot(data=df[df.gender=='female'],x='average',kde=True,hue='lunch')
plt.subplot(143)
sns.histplot(data=df[df.gender=='male'],x='average',kde=True,hue='lunch')
plt.show()
```



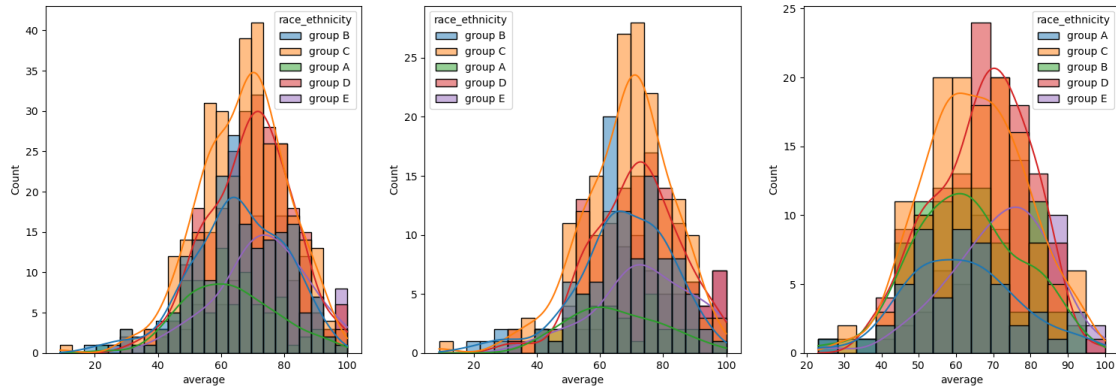
Insights 1. Standard Lunch helps students to perform well in exam 2. Standard Lunch helps perform well in exams be it a male or female

```
[40]: plt.subplots(1,3,figsize=(25,6))
plt.subplot(141)
ax =sns.histplot(data=df,x='average',kde=True,hue='parental_level_of_education')
plt.subplot(142)
ax =sns.histplot(data=df[df.
    ↳gender=='male'],x='average',kde=True,hue='parental_level_of_education')
plt.subplot(143)
ax =sns.histplot(data=df[df.
    ↳gender=='female'],x='average',kde=True,hue='parental_level_of_education')
plt.show()
```



Insights 1. In general parents education don't help student to perform well in exam 2. Second plot shows that the parents whose education is of associate's degree or master's degree their male child perform well in exam 3. Third plot we can see there is no effect of parent's education on female students

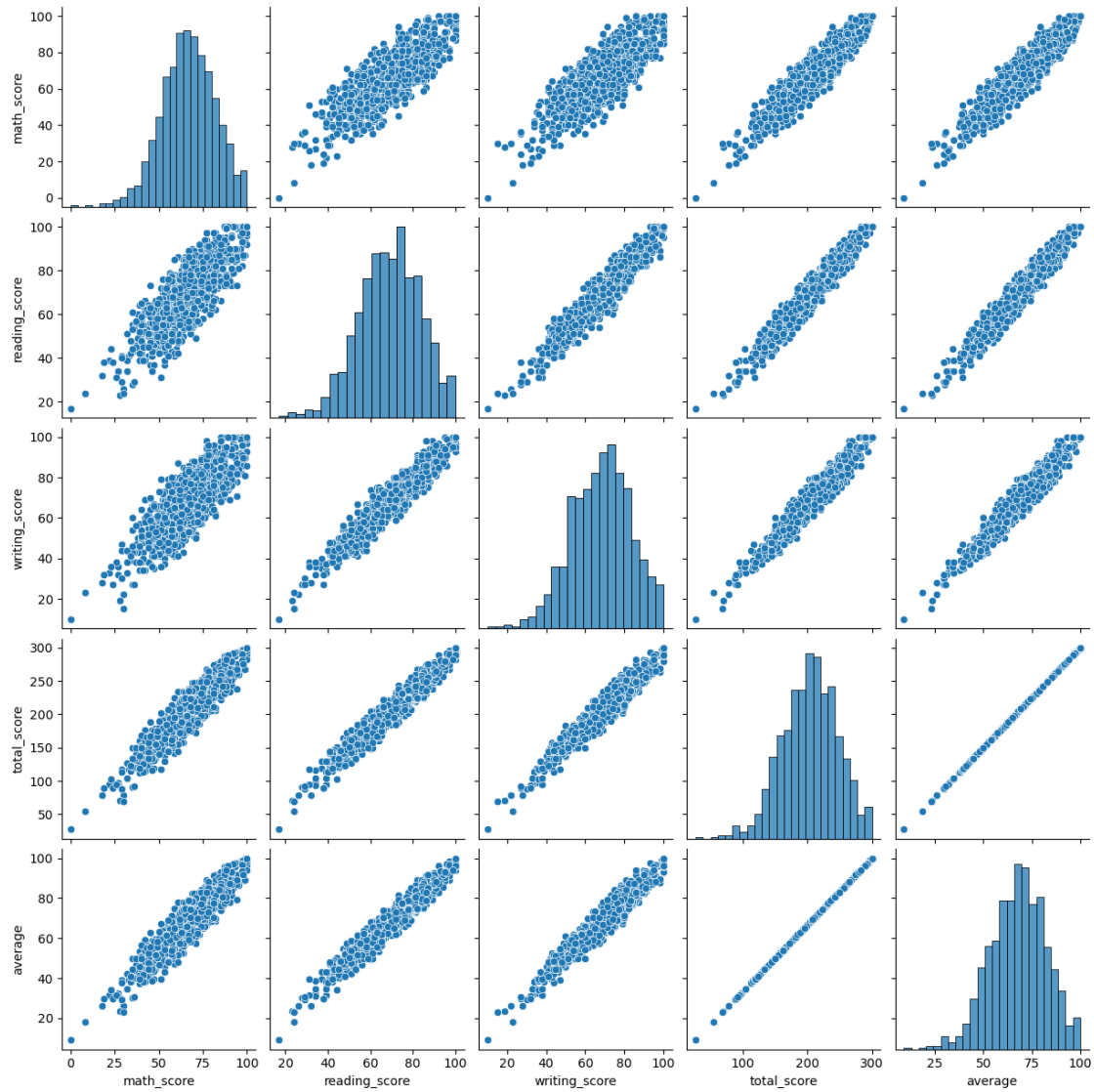
```
[41]: plt.subplots(1,3,figsize=(25,6))
plt.subplot(141)
ax =sns.histplot(data=df,x='average',kde=True,hue='race_ethnicity')
plt.subplot(142)
ax =sns.histplot(data=df[df.
    ↳gender=='female'],x='average',kde=True,hue='race_ethnicity')
plt.subplot(143)
ax =sns.histplot(data=df[df.
    ↳gender=='male'],x='average',kde=True,hue='race_ethnicity')
plt.show()
```



Insights 1. Students of group A and group B tends to perform poorly in exam 2. Students of group A and group B tends to perform poorly in exam irrespective of whether they male or female

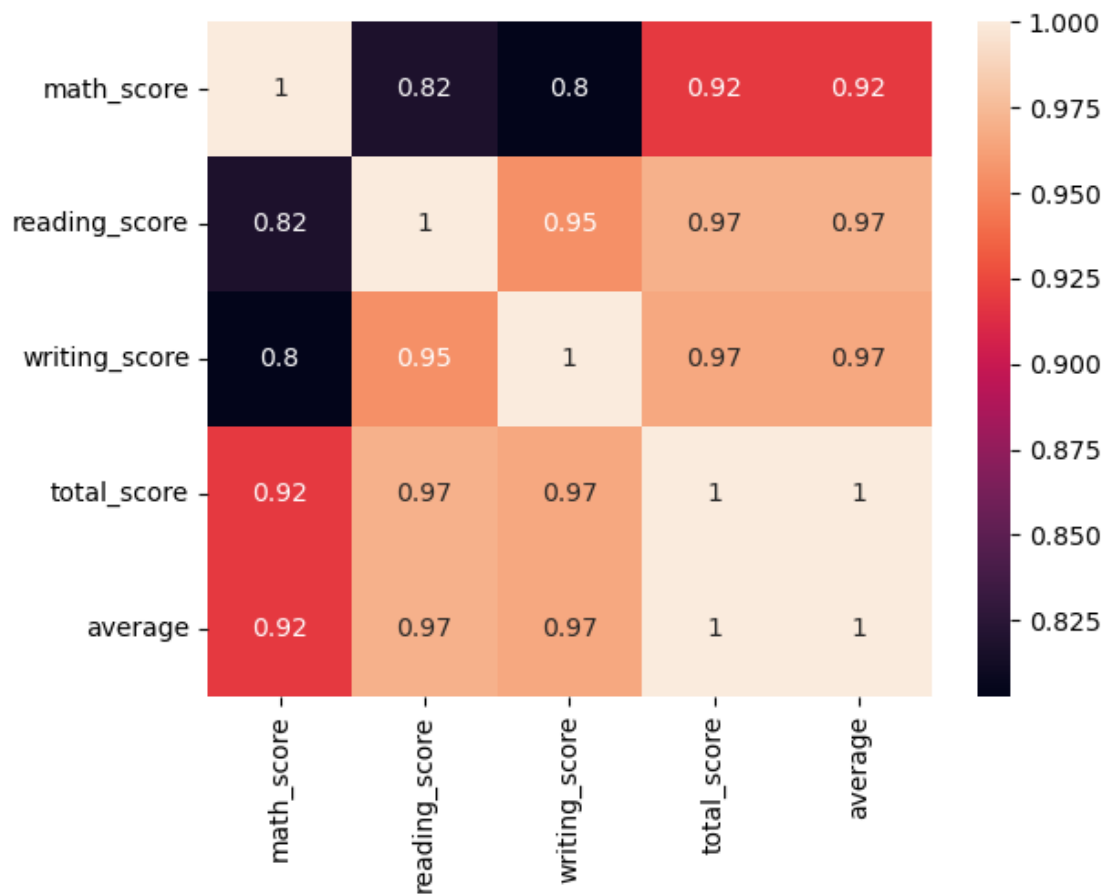
```
[43]: sns.pairplot(df)
```

```
[43]: <seaborn.axisgrid.PairGrid at 0x222601ef5b0>
```



```
[46]: sns.heatmap(df.corr(), annot=True)
```

```
[46]: <AxesSubplot:>
```

[]: