

# Student Exam Scores Data Insights with PySpark

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## 1 Project Description

This dataset includes scores from three test scores of students at a (fictional) public school and a variety of personal and socio-economic factors that may have interaction effects upon them.

This dataset is fictional and should be used for educational purposes only.

There are a few attributes in the datasets:

1. **Gender:** Gender of the student (male/female).
2. **EthnicGroup:** Ethnic group of the student (group A to E).
3. **ParentEduc:** Parent(s) education background (from some high school to master's degree).
4. **LunchType:** School lunch type (standard or free/reduced).
5. **TestPrep:** Test preparation course followed (completed or none).
6. **ParentMaritalStatus:** Parent(s) marital status (married/single/widowed/divorced).
7. **PracticeSport:** How often the student practices sport (never/sometimes/regularly)).
8. **IsFirstChild:** If the child is the first child in the family or not (yes/no).
9. **NrSiblings:** Number of siblings the student has (0 to 7).
10. **TransportMeans:** Means of transport to school (school bus/private).

11. **WklyStudyHours:** Weekly self-study hours (less than 5 hrs; between 5 and 10 hrs; more than 10 hrs).

12. **MathScore:** Math test score (0-100).

13. **ReadingScore:** Reading test score (0-100).

14. **WritingScore:** Writing test score (0-100).

Here, all those attributes that I am using to find out data offer insights into the academic performance of the students and the different elements that may affect it.

I am using PySpark SQL query technologies to find out each student's performance in different sectors like TestPrep and Ethnic Group.

## 1.1 Implementation Steps

**1.Data Gathering and Preparation :** We gathered our information from kaggle.com,focused on the student scores.We extracted these data from kaggle and saved the two csv files. We cleaned the data and changed the contents column names to lowercase

**2.Data Analysis :**We used pandas , We examined basic information Mathscore and readingscore and writingscore compared with other elements called ethnical group,test prep...etcc. and we used matplotlib for plotting the relationship between the elements

**3.Comparative Analysis:** we have compared the average of each score and compare who is the highest and analysis it.

## 2 Results Summary

There are results of the goals:

**Goal 1: Creating the dataframe**

The screenshot shows a Jupyter Notebook environment with a file explorer on the left and a code editor on the right. The code editor contains the following Python code:

```
Row(amt, b45, c="strings", d=date(2000, 3, 1), e=datetime(2000, 1, 3, 12, 0))
df
[2]: DataFrame[a: bigint, b: double, c: string, d: date, e: timestamp]
[3]: # Creating a dataframe called Student1.
Student1 = spark.read.format('csv').option('header', 'true').load('Original_data_with_more_rows.csv')
Student1.createTempView('Original')
Student1.show()
```

The output of the code shows a preview of the DataFrame with 20 rows. The columns are: \_id, Gender, EthnicGroup, ParentEduc, LunchType, TestPrep, MathScore, ReadingScore, WritingScore. The data is as follows:

_id	Gender	EthnicGroup	ParentEduc	LunchType	TestPrep	MathScore	ReadingScore	WritingScore
0	female	group B	bachelor's degree	standard	none	721	721	741
1	female	group C	some college	standard	completed	699	900	889
2	female	group B	master's degree	standard	none	900	950	930
3	male	group B	associate's degree	free/reduced	none	477	577	444
4	male	group C	some college	standard	none	760	780	750
5	female	group B	associate's degree	standard	none	711	831	780
6	female	group B	some college	standard	completed	680	950	920
7	male	group B	some college/free/reduced	none	none	400	430	390
8	male	group B	high school/free/reduced	completed	none	640	640	670
9	female	group B	high school/free/reduced	none	none	380	600	500
10	male	group B	associate's degree	standard	none	580	540	520
11	male	group B	associate's degree	standard	none	480	520	430
12	female	group B	high school	standard	none	650	810	730
13	male	group A	some college	standard	completed	780	720	700
14	female	group A	master's degree	standard	none	500	530	580
15	female	group C	some high school	standard	none	690	750	780
16	male	group C	high school	standard	none	880	890	860
17	female	group B	some high school/free/reduced	none	none	180	320	280
18	male	group C	master's degree/free/reduced	completed	none	460	420	460
19	female	group B	associate's degree/free/reduced	none	none	540	580	610

**Goal 2:** Determine the count of students based on their ethnic group and whether they underwent test preparation.

```

+ (25)
# Creating an other dataframe called Students
Students = spark.read.format('csv').option('header', 'true').load('Expanded_data_with_more_features.csv')
Students.createOrReplaceTempView('Expanded')
Students.show()

+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| EthnicGroup | ParentEduc | LunchType | TestPrep | ParentMaritalStatus | PracticeSport | IsFirstChild | HasSiblings | TransportMeans | StudyHours | MathScore | ReadingScore | WritingScore |
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| 0 | female | NULL | bachelor's degree | standard | none | married | regularly | yes | 3 | school_bus | < 5 | 7 |
| 1 | female | group C | some college | standard | NULL | married | sometimes | yes | 8 | NULL | 18-May | 6 |
| 2 | female | group B | master's degree | standard | none | single | sometimes | yes | 4 | school_bus | < 5 | 8 |
| 3 | male | group A | associate's degree | free/reduced | none | married | never | no | 1 | NULL | 18-May | 4 |
| 4 | male | group C | some college | standard | none | married | sometimes | yes | 8 | school_bus | 18-May | 7 |
| 5 | female | group B | associate's degree | standard | none | married | regularly | yes | 1 | school_bus | 18-May | 7 |
| 6 | female | group B | some college | standard/completed | widowed | never | no | 1 | private | 18-May | 8 |
| 7 | male | group B | some college | free/reduced | none | married | sometimes | yes | 1 | private | > 18 | 4 |
| 8 | male | group D | high school | free/reduced/completed | single | sometimes | no | 3 | private | > 18 | 6 |
| 9 | female | group B | high school | free/reduced | none | married | regularly | yes | NULL | private | < 5 | 3 |
| 10 | male | group C | associate's degree | standard | none | NULL | sometimes | yes | 1 | private | 18-May | 5 |
| 11 | male | group B | associate's degree | standard | none | divorced | sometimes | yes | 1 | school_bus | 18-May | 4 |
| 12 | female | group B | high school | standard | none | married | regularly | no | 1 | private | 18-May | 6 |

```

**Goal 3:** Assess the overall academic performance of students by examining average scores in Math, Reading, and Writing.

```

+ (45)
Goal 3: Assess the overall academic performance of students by examining average scores in Math, Reading, and Writing.

# Construct the query for average scores
average_scores_query = """
SELECT
    AVG(MathScore) AS AvgMathScore,
    AVG(ReadingScore) AS AvgReadingScore,
    AVG(WritingScore) AS AvgWritingScore
FROM Original
"""

# Execute the query
average_scores_result = spark.sql(average_scores_query)

# Show the result
average_scores_result.show()

+-----+-----+-----+
| AvgMathScore | AvgReadingScore | AvgWritingScore |
+-----+-----+-----+
| 66.74935543879116 | 69.6249796024934 | 68.46832675173786 |
+-----+-----+-----+

```

**Goal 4:** Determine the count of students based on their ethnic group and whether they underwent test preparation.

```

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Construct the query for counting students based on ethnic group and test preparation
count_by_ethnicity_query = """
SELECT
    EthnicGroup,
    TestPrep,
    COUNT(*) AS StudentCount
FROM Original
GROUP BY EthnicGroup, TestPrep
"""

Execute the query
count_by_ethnicity_result = spark.sql(count_by_ethnicity_query)

Show the result
count_by_ethnicity_result.show()

[EthnicGroup, TestPrep, StudentCount]
| group C | completed | 3396 |
| group E | none | 2895 |
| group B | none | 3992 |
| group E | completed | 1479 |
| group A | none | 1574 |
| group D | none | 5273 |
| group B | completed | 2280 |
| group D | completed | 2713 |
| group A | completed | 785 |
| group C | none | 6420 |
  
```

**Goal 5 :** Observing how different ethnic groups perform different academic subjects.

```

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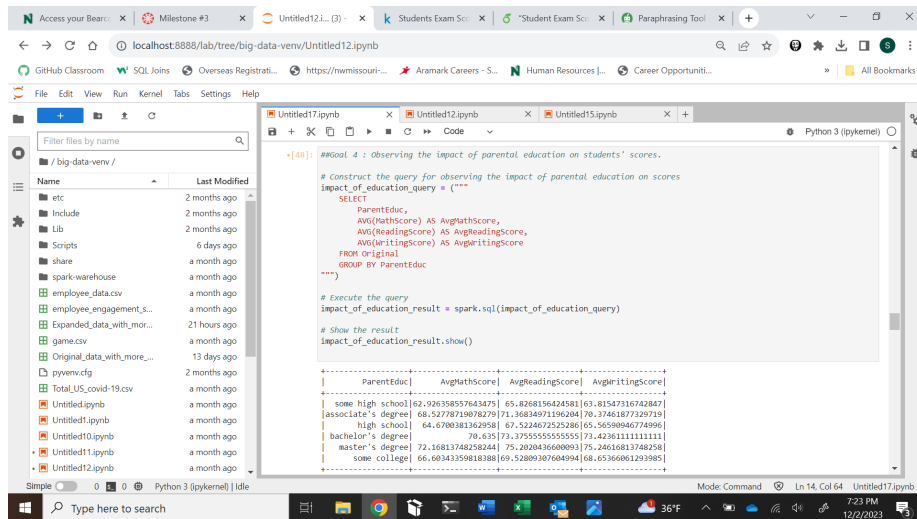
Construct the query for observing academic performance by ethnic group
performance_by_ethnicity_query = """
SELECT
    EthnicGroup,
    AVG(MathScore) AS AvgMathScore,
    AVG(ReadingScore) AS AvgReadingScore,
    AVG(WritingScore) AS AvgWritingScore
FROM Original
GROUP BY EthnicGroup
"""

Execute the query
performance_by_ethnicity_result = spark.sql(performance_by_ethnicity_query)

Show the result
performance_by_ethnicity_result.show()

[EthnicGroup, AvgMathScore, AvgReadingScore, AvgWritingScore]
| group B | 68.2080103592 | 67.586293299742 | 65.9828811369591 |
| group C | 64.88070497147514 | 68.656773411328 | 67.82801548492258 |
| group D | 67.80791384923616 | 70.57337841222139 | 70.83821687953919 |
| group A | 61.1184601873803 | 66.9559135226791 | 65.2246714706228 |
| group E | 75.48064365671642 | 74.49067164179104 | 72.72527985074628 |
  
```

## Goal 6 :Observing the impact of parental education on students' scores.



The screenshot shows a Jupyter Notebook interface with a file explorer on the left and a code editor on the right. The code editor contains a SQL query to observe the impact of parental education on students' scores. The query is as follows:

```
[48]: #Goal 4 : Observing the impact of parental education on students' scores.

# Construct the query for observing the impact of parental education on scores
impact_of_education_query = """
SELECT
    ParentEduc,
    AVG(MathScore) AS AvgMathScore,
    AVG(ReadingScore) AS AvgReadingScore,
    AVG(WritingScore) AS AvgWritingScore
FROM Original
GROUP BY ParentEduc
"""

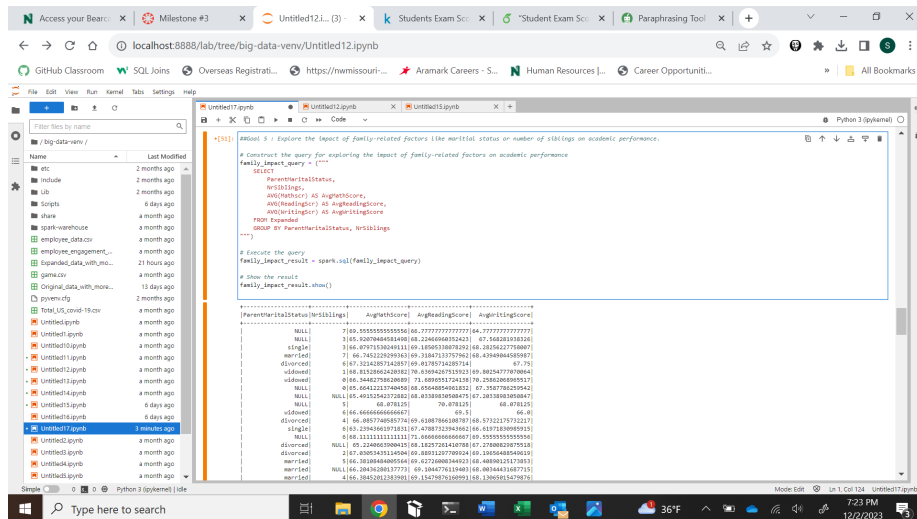
# Execute the query
impact_of_education_result = spark.sql(impact_of_education_query)

# Show the result
impact_of_education_result.show()
```

The output of the query is displayed below the code:

ParentEduc	AvgMathScore	AvgReadingScore	AvgWritingScore
some high school	62.92635857643475	65.8268156424581	63.81547316742847
associate's degree	68.52778719678279	71.36844971196204	70.37461877329719
high school	64.6780381362958	67.5224672525286	65.565084677496
bachelor's degree	70.635	73.37555555555555	73.42361111111111
master's degree	72.16811748258244	75.2028436408091	75.24610813748258
some college	66.6834359818318	69.2389387608494	68.6536686129385

## Goal 7 :Explore the impact of family-related factors like marital status or number of siblings on academic performance.



The screenshot shows a Jupyter Notebook interface with a file explorer on the left and a code editor on the right. The code editor contains a SQL query to explore the impact of family-related factors on academic performance. The query is as follows:

```
[51]: #Goal 5 : Explore the impact of family-related factors like marital status or number of siblings on academic performance.

# Construct the query for exploring the impact of family-related factors on academic performance
family_impact_query = """
SELECT
    ParentMaritalStatus,
    HSiblings,
    AVG(MathScore) AS AvgMathScore,
    AVG(ReadingScore) AS AvgReadingScore,
    AVG(WritingScore) AS AvgWritingScore
FROM Expanded
GROUP BY ParentMaritalStatus, HSiblings
"""

# Execute the query
family_impact_result = spark.sql(family_impact_query)

# Show the result
family_impact_result.show()
```

The output of the query is displayed below the code:

ParentMaritalStatus	HSiblings	AvgMathScore	AvgReadingScore	AvgWritingScore
single	7	68.55555555555556	68.77777777777777	64.77777777777777
single	1	68.55555555555556	68.77777777777777	64.77777777777777
single	0	68.55555555555556	68.77777777777777	64.77777777777777
single	2	68.55555555555556	68.77777777777777	64.77777777777777
single	3	68.55555555555556	68.77777777777777	64.77777777777777
single	4	68.55555555555556	68.77777777777777	64.77777777777777
single	5	68.55555555555556	68.77777777777777	64.77777777777777
single	6	68.55555555555556	68.77777777777777	64.77777777777777
single	8	68.55555555555556	68.77777777777777	64.77777777777777
single	9	68.55555555555556	68.77777777777777	64.77777777777777
single	10	68.55555555555556	68.77777777777777	64.77777777777777
single	11	68.55555555555556	68.77777777777777	64.77777777777777
single	12	68.55555555555556	68.77777777777777	64.77777777777777
single	13	68.55555555555556	68.77777777777777	64.77777777777777
single	14	68.55555555555556	68.77777777777777	64.77777777777777
single	15	68.55555555555556	68.77777777777777	64.77777777777777
single	16	68.55555555555556	68.77777777777777	64.77777777777777
single	17	68.55555555555556	68.77777777777777	64.77777777777777
single	18	68.55555555555556	68.77777777777777	64.77777777777777
single	19	68.55555555555556	68.77777777777777	64.77777777777777
single	20	68.55555555555556	68.77777777777777	64.77777777777777
single	21	68.55555555555556	68.77777777777777	64.77777777777777
single	22	68.55555555555556	68.77777777777777	64.77777777777777
single	23	68.55555555555556	68.77777777777777	64.77777777777777
single	24	68.55555555555556	68.77777777777777	64.77777777777777
single	25	68.55555555555556	68.77777777777777	64.77777777777777
single	26	68.55555555555556	68.77777777777777	64.77777777777777
single	27	68.55555555555556	68.77777777777777	64.77777777777777
single	28	68.55555555555556	68.77777777777777	64.77777777777777
single	29	68.55555555555556	68.77777777777777	64.77777777777777
single	30	68.55555555555556	68.77777777777777	64.77777777777777
single	31	68.55555555555556	68.77777777777777	64.77777777777777
single	32	68.55555555555556	68.77777777777777	64.77777777777777
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single	34	68.55555555555556	68.77777777777777	64.77777777777777
single	35	68.55555555555556	68.77777777777777	64.77777777777777
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single	39	68.55555555555556	68.77777777777777	64.77777777777777
single	40	68.55555555555556	68.77777777777777	64.77777777777777
single	41	68.55555555555556	68.77777777777777	64.77777777777777
single	42	68.55555555555556	68.77777777777777	64.77777777777777
single	43	68.55555555555556	68.77777777777777	64.77777777777777
single	44	68.55555555555556	68.77777777777777	64.77777777777777
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single	51	68.55555555555556	68.77777777777777	64.77777777777777
single	52	68.55555555555556	68.77777777777777	64.77777777777777
single	53	68.55555555555556	68.77777777777777	64.77777777777777
single	54	68.55555555555556	68.77777777777777	64.77777777777777
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single	66	68.55555555555556	68.77777777777777	64.77777777777777
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single	69	68.55555555555556	68.77777777777777	64.77777777777777
single	70	68.55555555555556	68.77777777777777	64.77777777777777
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single	89	68.55555555555556	68.77777777777777	64.77777777777777
single	90	68.55555555555556	68.77777777777777	64.77777777777777
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single	92	68.55555555555556	68.77777777777777	64.77777777777777
single	93	68.55555555555556	68.77777777777777	64.77777777777777
single	94	68.55555555555556	68.77777777777777	64.77777777777777
single	95	68.55555555555556	68.77777777777777	64.77777777777777
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single	98	68.55555555555556	68.77777777777777	64.77777777777777
single	99	68.55555555555556	68.77777777777777	64.77777777777777
single	100	68.55555555555556	68.77777777777777	64.77777777777777
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single	102	68.55555555555556	68.77777777777777	64.77777777777777
single	103	68.55555555555556	68.77777777777777	64.77777777777777
single	104	68.55555555555556	68.77777777777777	64.77777777777777
single	105	68.55555555555556	68.77777777777777	64.77777777777777
single	106	68.55555555555556	68.77777777777777	64.77777777777777
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single	115	68.55555555555556	68.77777777777777	64.77777777777777
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single	120	68.55555555555556	68.77777777777777	64.77777777777777
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single	124	68.55555555555556	68.77777777777777	64.77777777777777
single	125	68.55555555555556	68.77777777777777	64.77777777777777
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single	127	68.55555555555556	68.77777777777777	64.77777777777777
single	128	68.55555555555556	68.77777777777777	64.77777777777777
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single	130	68.55555555555556	68.77777777777777	64.77777777777777
single	131	68.55555555555556	68.77777777777777	64.77777777777777
single	132	68.55555555555556	68.77777777777777	64.77777777777777
single	133	68.55555555555556	68.77777777777777	64.77777777777777
single	134	68.		

**Goal 8:** Observing how the mode of transportation to school relates to weekly study hours drawing a graphs using matplotlib.

The screenshot shows a Jupyter Notebook with the following code:

```

# Notebook 6 : Observing how the mode of transportation to school relates to weekly study hours drawing a graphs using matplotlib.

# Spark SQL query to analyze the impact of family-related factors on academic performance
family_impact_query = """
SELECT
    ParentMaritalStatus,
    COUNT(ReadingScore) AS COUNTMathScore,
    COUNT(ReadingScore) AS COUNTReadingScore,
    COUNT(WritingScore) AS COUNTWritingScore,
    FROM Expanded
    WHERE ParentMaritalStatus IS NOT NULL
    GROUP BY ParentMaritalStatus
"""

# Execute the query
family_impact_result = spark.sql(family_impact_query)

# Convert the result to a Pandas DataFrame for Matplotlib
family_impact_df = family_impact_result.toPandas()

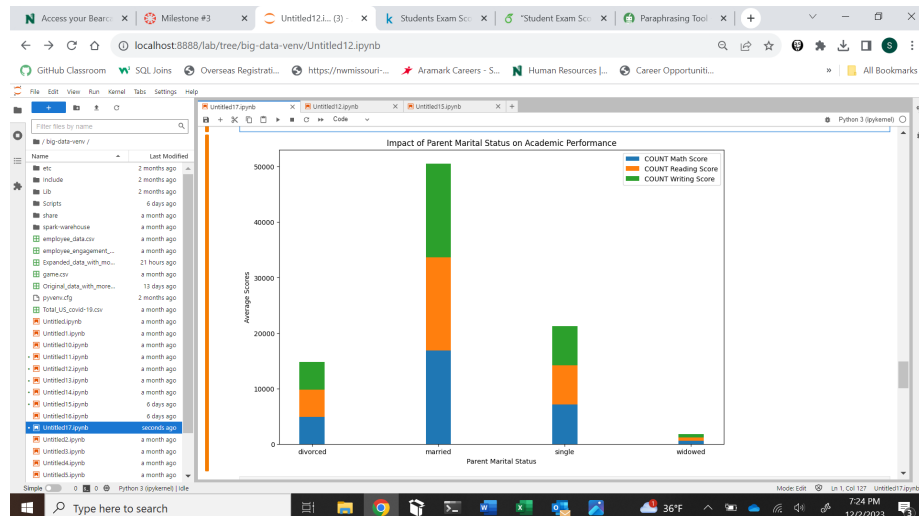
# Plotting the bar chart using Matplotlib
import matplotlib.pyplot as plt

plt.figure(figsize=(12, 8))
bar_width = 0.2
index = range(len(family_impact_df))

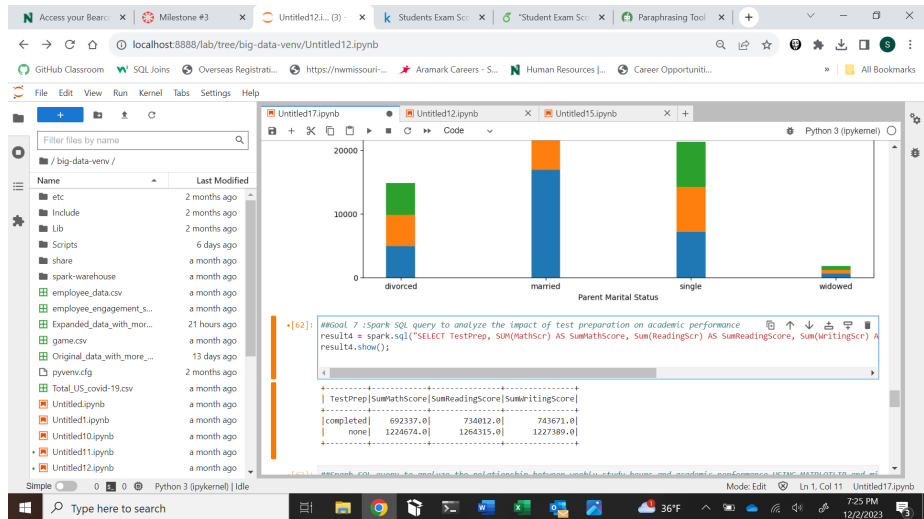
plt.bar(index, family_impact_df['COUNTMathScore'], width=bar_width, label='COUNT Math Score')
plt.bar(index, family_impact_df['COUNTReadingScore'], width=bar_width, label='COUNT Reading Score', bottom=family_impact_df['COUNTMathScore'])
plt.bar(index, family_impact_df['COUNTWritingScore'], width=bar_width, label='COUNT Writing Score', bottom=family_impact_df['COUNTMathScore'])

plt.xlabel('Parent Marital Status')
plt.ylabel('Average Score')
plt.title('Impact of Parent Marital Status on Academic Performance')
plt.xticks(index, family_impact_df['ParentMaritalStatus'])
plt.legend()
plt.show()

```



**Goal 9 :**Determine if students who completed test preparation performed better.





**Goal 10 :**Observing the relationship between weekly study hours and academic performance and also drawing a graphs between weekly study hours and academic performance using matplotlib.

The screenshot shows a Jupyter Notebook with a file explorer on the left and a code editor on the right. The code in the editor is as follows:

```

#Goal 8 : Spark SQL query to analyze the relationship between weekly study hours and academic performance USING MATPLOTLIB and window score of three subjects.
result = spark.sql("SELECT WeeklyStudyHours, MIN(MathScore) AS MINMathScore, MIN(ReadingScore) AS MINReadingScore, MIN(WritingScore) AS MINWritingScore FROM Expanded M
result.show(100)

import matplotlib.pyplot as plt

# Sample data (replace this with your actual data)
study_hours = [1, 2, 3, 4, 5, 6, 7, 8, 9, 10]
avg_math_scores = [78, 75, 80, 85, 90, 92, 88, 86, 78, 75]
avg_reading_scores = [72, 70, 75, 80, 85, 88, 86, 84, 76, 70]
avg_writing_scores = [68, 72, 76, 80, 85, 88, 82, 78, 74, 70]

# Create a scatter plot
plt.figure(figsize=(10, 6))
plt.scatter(study_hours, avg_math_scores, label='MIN Math Score', color='red')
plt.scatter(study_hours, avg_reading_scores, label='MIN Reading Score', color='blue')
plt.scatter(study_hours, avg_writing_scores, label='MIN Writing Score', color='green')

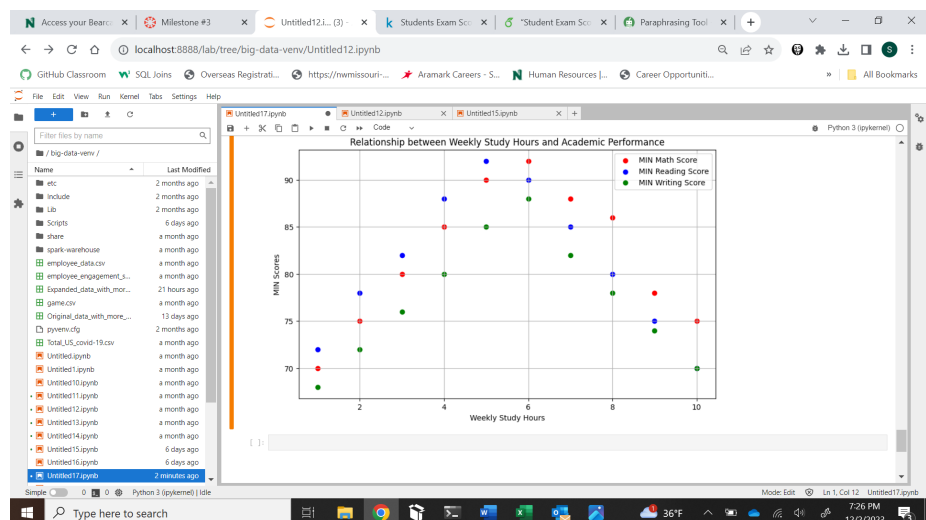
# Customize the plot
plt.title('Relationship between Weekly Study Hours and Academic Performance')
plt.xlabel('Weekly Study Hours')
plt.ylabel('MIN Scores')
plt.legend()
plt.grid(True)

# Show the plot
plt.show()

```

Below the code, the output of the plot is shown as a table:

Weekly Study Hours	MIN Math Score	MIN Reading Score	MIN Writing Score
1-10	78	72	68
< 5	100	100	100
> 10	0	100	10



## 5'VS:

The concept of 5Vs refers to the characteristics of big data that organizations need to consider for effective management and analysis. Here's an explanation of how the 5Vs apply to the dataset you provided:

**1. Volume Analysis :** The size of the dataset is referred to as the volume of data. In the instance, the number of rows and columns in the dataset—which contains data about several students and their characteristics—determines the volume.

**2. Velocity:** The rate at which data is created, analyzed, and updated is referred to as velocity. If the data in our dataset is updated frequently or if new student data is added on a regular basis, this might be relevant.

**3. Variety:** Variety is the range of sources and types of data. The dataset reflects a variety of information types and contains attributes like test scores, gender, ethnicity, and socioeconomic considerations.

**4. Veracity:** The quality and reliability of the data are related to veracity. Ensuring the precision and dependability of data in your dataset, managing absent values, and resolving any differences represent essential elements.

**5. Value:** Value represents the usefulness and meaningfulness of the data. From the dataset, the value lies in extracting insights about student performance, understanding factors influencing academic outcomes, and making informed educational decisions.

## 3 Conclusion

We can conclude for this project that we can calculate the average scores in Math, Reading, and Writing to get an overall understanding of the students' performance where reading score average is highest while compare to other mathscore and writingscore and also Investigate whether the level of parental education has an impact on student scores and in various elements.

## 4 Citations

Provide all the necessary citations for the sources you utilized to finish this project successfully. Save all your work to your GitHub repo and provide the URL.

**Dataset:** Datasets

**GitHub Repo:** <https://github.com/saipuneet/StudentExamScore.git>