

# Practical 2

## Data condensation using RStudio

Summarize the Data:-

```
summary(data)
```

Variable	Min	1st Qu.	Median	Mean	3rd Qu.	Max
gender	0.000	0.000	0.000	0.482	1.000	1.000
race_ethnicity	-	-	-	-	-	-
parental_level_of_education	-	-	-	-	-	-
lunch	0.000	0.000	1.000	0.645	1.000	1.000
test_preparation_course	0.000	0.000	0.000	0.358	1.000	1.000
math_score	0.00	57.00	66.00	66.09	77.00	100.00
reading_score	17.00	59.00	70.00	69.17	79.00	100.00
writing_score	10.00	57.75	69.00	68.05	79.00	100.00
total_score	27.0	175.0	205.0	203.3	233.0	300.0
average_score	9.00	58.33	68.33	67.77	77.67	100.00

Explore Relationships Between Variables

1 Distribution of Scores

Overall statistics

```
> overall_stats <- data.frame( Mean = colMeans(studentdata[, c("math_score", "reading_score",  
"writing_score")]),Variance = apply(studentdata[, c("math_score", "reading_score", "writing_score")],  
2, var) )
```

```
> print("Overall Statistics:")
```

```
> print(overall_stats)
```

Variable	Mean	Variance
math_score	66.089	229.9190
reading_score	69.169	213.1656
writing_score	68.054	230.9080

# Step 1: Print Gender Distribution

```
cat("Gender Distribution:\n")
```

```
gender_distribution <- table(studentdata$gender)
```

```
print(gender_distribution)
```

# Step 2: Print Test Preparation Course Distribution

```
cat("\nTest Preparation Course Distribution:\n")
```

```
test_preparation_course_distribution <- table(studentdata$test_preparation_course)
```

```
print(test_preparation_course_distribution)
```

# Step 3: Print Parental Education Level Distribution

```
cat("\nParental Education Level Distribution:\n")
```

```
parental_education_distribution <- table(studentdata$parental_level_of_education)
```

```
print(parental_education_distribution)
```

# Step 4: Print Lunch Distribution

```
cat("\nLunch Distribution:\n")
```

```
lunch_distribution <- table(studentdata$lunch)
```

```
print(lunch_distribution)
```

ouput:-

### **Gender Distribution**

#### **Gender Count**

Male    518

Female 482

---

### **Test Preparation Course Distribution**

#### **Course Completion Status Count**

Not Completed            642

Completed                358

---

### Parental Education Level Distribution

Education Level	Count
-----------------	-------

Associate's Degree	222
--------------------	-----

Bachelor's Degree	118
-------------------	-----

High School	196
-------------	-----

Master's Degree	59
-----------------	----

Some College	226
--------------	-----

Some High School	179
------------------	-----

---

### Lunch Distribution

Lunch Type	Count
------------	-------

Standard Lunch	355
----------------	-----

Free/Reduced Lunch	645
--------------------	-----

Condense the data:-

```
unique_gender <- length(unique(studentdata$gender))
> unique_test_preparation <- length(unique(studentdata$test_preparation_course))
> unique_parental_education <- length(unique(studentdata$parental_level_of_education))
> unique_lunch <- length(unique(studentdata$lunch))
>
> # Quantitative variables: Calculate summary statistics
> math_summary <- summary(studentdata$math_score)
> reading_summary <- summary(studentdata$reading_score)
> writing_summary <- summary(studentdata$writing_score)
cat("Qualitative Variables:\n")
cat("Unique Genders:", unique_gender, "\n")
cat("Unique Test Preparation Courses:", unique_test_preparation, "\n")
cat("Unique Parental Education Levels:", unique_parental_education, "\n")
```

```
cat("Unique Lunch Types:", unique_lunch, "\n\n")
cat("Quantitative Variables:\n")
cat("Math Score Summary:\n", math_summary, "\n")
cat("Reading Score Summary:\n", reading_summary, "\n")
cat("Writing Score Summary:\n", writing_summary, "\n")
output:-
```

Here is the **output in table format** as requested:

**Qualitative Variables Table**

Variable	Unique Values
Gender	2
Test Preparation Courses	2
Parental Education Levels	6
Lunch Types	2

---

**Quantitative Variables Table**

Statistic	Math Score	Reading Score	Writing Score
Min	0	17	10
1st Quartile	57	59	57.75
Median	66	70	69
Mean	66.089	69.169	68.054
3rd Quartile	77	79	79
Max	100	100	100

This provides a clear and structured view of your summarized data.

## Use of Data Condensation

1. **Simplifies Data:** Condensing data reduces complexity by summarizing large datasets into key metrics, making it easier to interpret.
2. **Highlights Trends:** It helps identify patterns, such as averages, ranges, or distributions, at a glance.
3. **Improves Decision-Making:** Summarized data enables quicker and more informed decisions, especially in areas like performance analysis or resource allocation.
4. **Efficient Comparisons:** Condensation facilitates comparisons across variables or groups without overwhelming details.
5. **Data Visualization:** Summarized information is ideal for creating charts or tables to present insights effectively.

In essence, data condensation transforms raw data into actionable insights, aiding both analysis and communication.