

# Statistical Programming

## Question 3

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Word count: 1623

### 1. Overview

The Student Performance dataset is downloaded from Kaggle Student Performance in Exams (<https://www.kaggle.com/spscientist/students-performance-in-exams>). It consists of 1000 observations on 8 separate variables. 5 columns are categorical, and 3 columns are numeric variables. The variables include:

- Gender Ethnicity
- Parental level of Education
- Lunch
- Test preparation course
- Math score
- Reading score
- Writing score

The inspiration is to understand the influence of the student's background and determine the features which play an essential role in affecting academic performance. The background has to do with their ethnicity, education level of their parents and the type of lunch they have. Other innate characteristics deemed relevant for the students' performance is the gender, whether they had

completed a preparatory course before taking the tests. The students' performance is gauged on their scores obtained in the reading, writing and math tests.

## 2. The Dataset

```
# Read the dataset
```

```
data <- read.csv(file.choose(), header = T)
```

### Library

```
library(dplyr)
```

```
library(ggplot2)
```

```
library(gridExtra)
```

Combining the three scores into the average score as a measure of student performance. Further, we display the overview of data using the first six observations, the new avg.score included. The data contains no missing observations. From the numerical summaries below the reading.scores, writing scores and avg.scores shows left skewness as their means were lesser than the medians while the math.score is slightly right-skewed.

```
# Average score in 3 subjects
```

```
avg.score <- rowSums(data[, 6:8])/3
```

```
data <- cbind(data, avg.score)
```

```
# Preview of the dataset
```

```
head(data)
```

```
## gender race.ethnicity parental.level.of.education lunch
```

```
## 1 female group B bachelor's degree standard
```

```
## 2 female      group C      some college  standard
## 3 female      group B      master's degree  standard
## 4  male       group A      associate's degree free/reduced
## 5  male       group C      some college  standard
## 6 female      group B      associate's degree  standard
## test.preparation.course math.score reading.score writing.score avg.score
## 1             none       72       72       74 72.66667
## 2             completed   69       90       88 82.33333
## 3             none       90       95       93 92.66667
## 4             none       47       57       44 49.33333
## 5             none       76       78       75 76.33333
## 6             none       71       83       78 77.33333
```

**summary**(data)

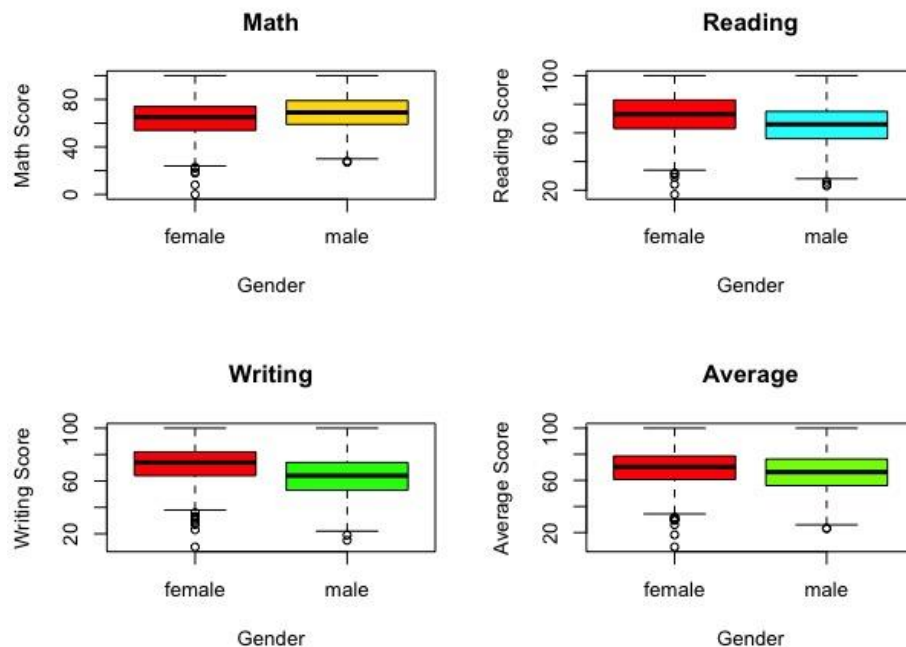
```
## gender      race.ethnicity  parental.level.of.education
## Length:1000    Length:1000    Length:1000
## Class :character Class :character Class :character
## Mode :character Mode :character Mode :character
##
## lunch      test.preparation.course  math.score  reading.score
## Length:1000    Length:1000          Min.   : 0.00  Min.   :17.00
## Class :character Class :character    1st Qu.: 57.00  1st Qu.: 59.00
## Mode :character Mode :character      Median : 66.00  Median : 70.00
##
##                               Mean   : 66.09  Mean   : 69.17
##                               3rd Qu.: 77.00  3rd Qu.: 79.00
##                               Max.    :100.00  Max.    :100.00
## writing.score  avg.score
## Min.   :10.00  Min.   : 9.00
## 1st Qu.: 57.75  1st Qu.: 58.33
## Median : 69.00  Median : 68.33
## Mean   : 68.05  Mean   : 67.77
```

```
## 3rd Qu.: 79.00 3rd Qu.: 77.67
## Max. :100.00 Max. :100.00
```

### 3. Visualisation

The graphical summaries of the data give us the first glimpse of how the variables relate to each other before we proceed to the numerical summaries. The plots give a prior knowledge of the useful elements in our data.

Comparison of Gender attributes to the Marks.



From the plot, it is evident that females tend to have a higher reading and writing score in comparison to males. In contrast, the male has a higher math score than female.

However, we need to use Hypothesis tests are designed to detect whether an “effect” is systematic or is the result of random variation.

*# Comparing the means of Math by Gender - the unpaired samples t-test*

```
var.test(data$math.score ~ data$gender)
```

```
## F test to compare two variances
##
## data: data$math.score by data$gender
## F = 1.1644, num df = 517, denom df = 481, p-value = 0.09016
## alternative hypothesis: true ratio of variances is not equal to 1
## 95 percent confidence interval:
## 0.9764071 1.3877941
## sample estimates:
## ratio of variances
## 1.164396
```

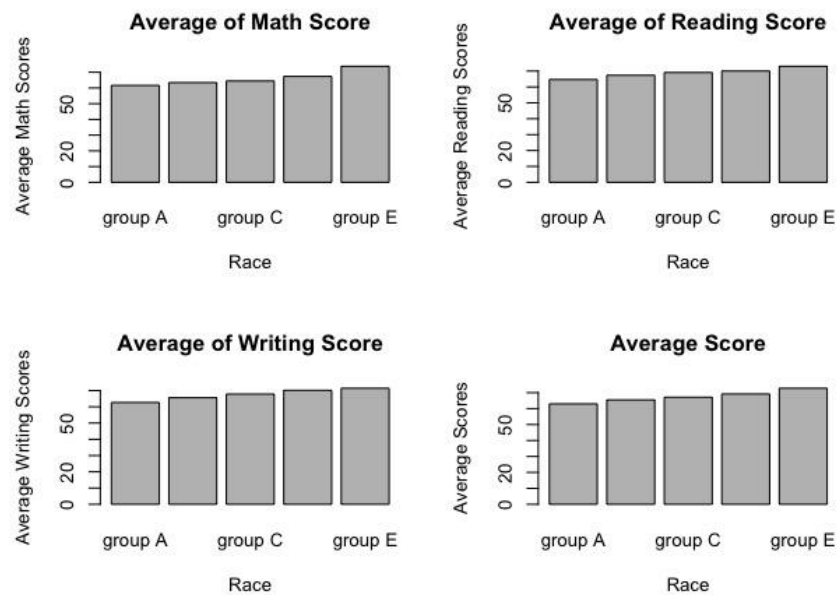
The p-value is greater than 0.05, so do not reject the null hypothesis that the variances are equal. We can thus proceed to perform t-test, assuming the equality of variance.

```
# Perform an unpaired samples t-test using
t.test(data$math.score ~ data$gender, var.equal=TRUE)

## Two Sample t-test
## data: data$math.score by data$gender
## t = -5.3832, df = 998, p-value = 9.12e-08
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -6.952285 -3.237737
## sample estimates:
## mean in group female mean in group male
## 63.63320 68.72822
```

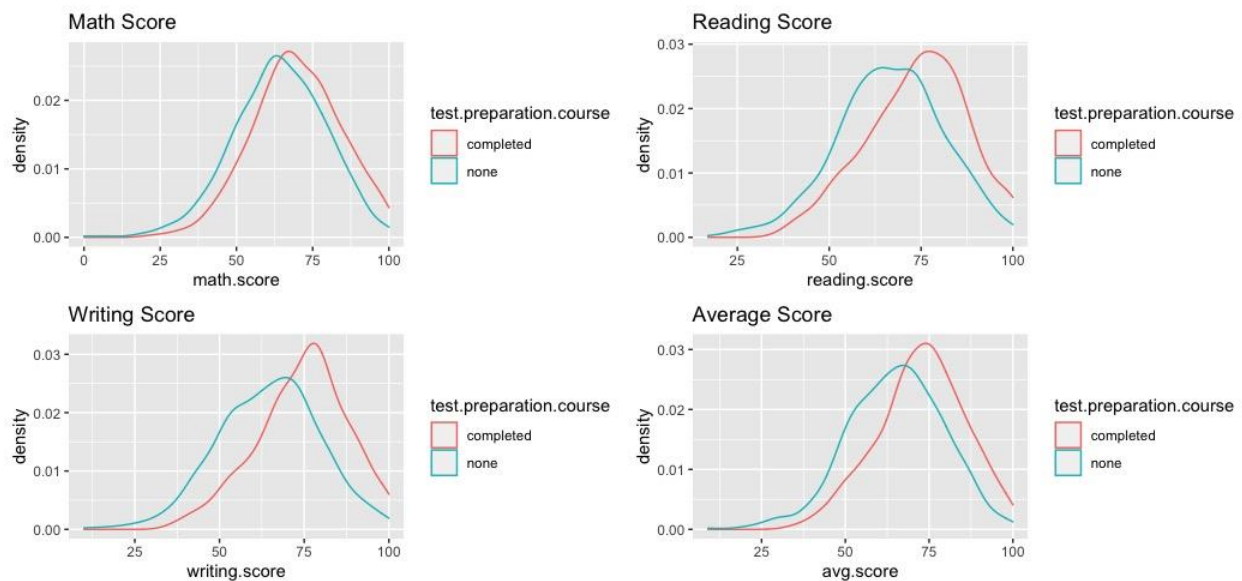
From the t-test, the p-value is (9.12e-08) found to be p-value < 0.001. The p-value is lesser than the significance level(0.05), and the 95% CI of the mean difference does not capture  $H_0 = 0$ . This indicates that mean male score percent is higher mean female score percent, and we reject the null hypothesis ( $H_0$ ) and test is statistically significant.

## Comparison of Race Ethnicity attributes to the Marks.



A boxplot visualisation shows that Group E students scored well compared to all other races,  $E > D > C > B > A$  in perspective.

## Comparison of Test Preparation attributes to the Marks.



We infer from the above plots that the students who completed the test preparation course had higher scores in all the 3 subjects and the average score compared to the students who had not taken any test preparation course.

### Hypothesis Testing

```
# Comparing the means of Reading by Pre-Test Course
var.test(data$reading.score ~ data$test.preparation.course)

##
## F test to compare two variances
##
## data: data$reading.score by data$test.preparation.course
## F = 0.88911, num df = 357, denom df = 641, p-value = 0.2144
## alternative hypothesis: true ratio of variances is not equal to 1
## 95 percent confidence interval:
##  0.7420038 1.0705553
## sample estimates:
## ratio of variances
##      0.8891108
```

The p-value is greater than 0.05, so do not reject the null hypothesis that the variances are equal. Hence we can proceed to t-test:

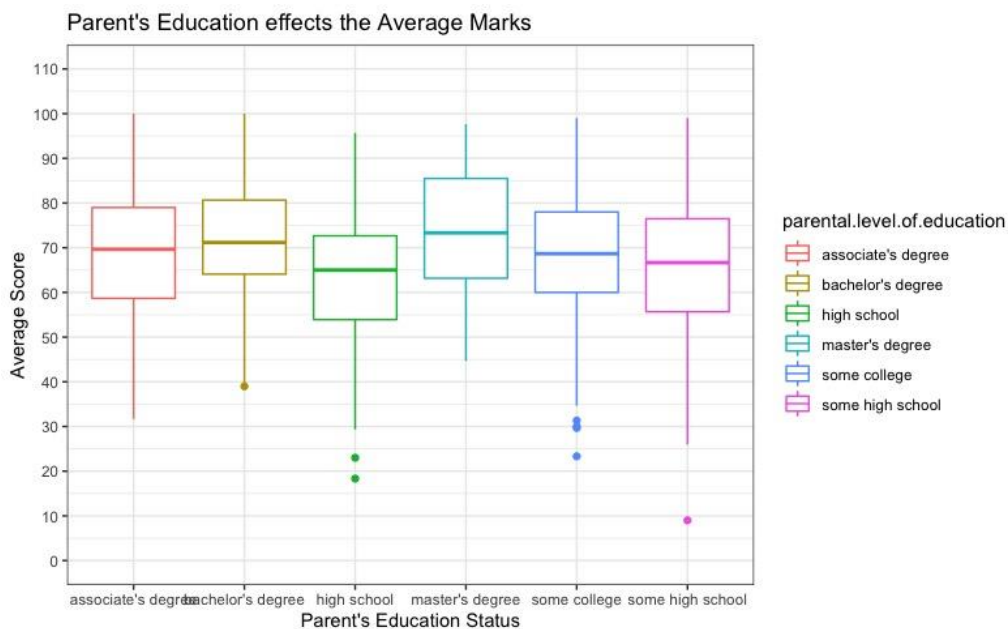
```
##
## Two Sample t-test
##
## data: data$reading.score by data$test.preparation.course
## t = 7.8717, df = 998, p-value = 9.082e-15
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
##  5.524900 9.194274
```

```
## sample estimates:
## mean in group completed    mean in group none
##          73.89385          66.53427
```

From the t-test, the p-value is found to be 9.082e-15. The p-value is lesser than the significance level (0.05), and the 95% CI of the mean difference does not capture  $H_0 = 0$ .

Thus, there is statistical evidence that the mean percentage of students who have taken the pre-test course is higher than the mean percentage of students who have not taken the pre-test course. The test is statistically significant and thus taking a pre-test necessary for improving reading scores.

### Comparison of Parent's Education attributes to the Average Marks.



It is easy to see that students who have highly educated parents (masters degree), also have a higher average score.

The F-test below confirms that the ( $F=10.75$  p-value =  $4.38e-10$ ) above results. The p-value is less than the level of significance 0.05 hence reject the null hypothesis; hence there is a difference in the means of avg.score of students with parents of various learning levels.

```
# ANOVA test to compare parents education and average marks
ANOVA<- aov(data$avg.score ~ data$parental.level.of.education)
summary(ANOVA)
```



```
##               Df Sum Sq Mean Sq F value    Pr(>F)
## data$parental.level.of.education  5 10420 2084.1 10.75 4.38e-10 ***
## Residuals                994 192648  193.8
## --
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

#### 4. Descriptive Statistics

##### Summary on Pre-Test Course determining the average score

```
## # A tibble: 2 x 10
##   test.preparation.cou... Min   Q1  Median   Q3  Max  Mean  SD    n Missing
##   <chr>      <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <int> <int>
## 1 completed      34.3  65   73.5   82.2 100   72.7  13.0  358     0
## 2 none           9   55.4  65.3   75   100   65.0  14.2  642     0
```

The students who took the pre-test had higher average score (72.7) compared to those who took no pre-test course (65.0) as inferred from the boxplots in the above section.

##### Summary on Gender determining the average score

```
## # A tibble: 2 x 10
##   gender Min   Q1  Median   Q3  Max  Mean  SD    n Missing
##   <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <int> <int>
## 1 female  9   60.7  70.3   78.7 100   69.6  14.5  518     0
## 2 male   23   56   66.3   76.2 100   65.8  13.7  482     0
```

Female performed well on the average score (69.6) compared to males (65.8).

##### Summary on Race Ethnicity determining the average score

```
## # A tibble: 2 x 10
##   race.ethnicity Min   Q1  Median   Q3  Max  Mean  SD    n Missing
##   <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <int> <int>
```

## 1 group A	23.3	52	61.3	73	96.3	63.0	14.4	89	0
## 2 group B	18.3	56.7	65	76.8	96.7	65.5	14.7	190	0
## 3 group C	9	57.7	68.3	77	98.7	67.1	13.9	319	0
## 4 group D	31	60.3	70	78.6	99	69.2	13.3	262	0
## 5 group E	26	64.7	73.5	82.4	100	72.8	14.6	140	0

The group E does the best in the average score (72.8) while the group A performs the poorest compared to all the other groups (63).

## 5. Conclusion

We can thus conclude from the data visualisation, t-tests and ANOVA test that Gender, Race/Ethnicity, Test Preparing and Parental level of Education have significant roles to Student Performance. The females perform averagely higher than males in writing, reading and average performance while men outshine them in math. Group E of the ethnicity is ranked top in performance in all the test and average scores. The analysis also shows that the pre-test course is advantageous for improving student performance. Lastly, the students' performance improves with depending on the level parents learning, the students whose parents hold masters degrees had the highest scores.