

**HO CHI MINH UNIVERSITY OF TECHNOLOGY**  
**OFFICE FOR INTERNATIONAL STUDY PROGRAM**



**PROBABILITY AND STATISTIC**  
**PROJECT REPORT**

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# **1.PROBLEM**

This data approach student achievement in secondary education of two Portuguese schools. The Data attributes include student grades, demographic, social and school related features) and it was collected by using school reports and questionnaires.

Attribute Information:

- sex - student's sex (binary: 'F' - female or 'M' - male)
- age - student's age (numeric: from 15 to 22)
- studytime - weekly study time (numeric: 1 - <2 hours, 2 - 2 to 5 hours, 3 - 5 to 10 hours, or 4- >10 hours)
- failures - number of past class failures (numeric: n if  $1 \leq n < 3$ , else 4)
- higher - wants to take higher education (binary: yes or no)
- absences - number of school absences (numeric: from 0 to 93)

# these grades are related with the course subject, Math or Portuguese:

- G1 - first period grade (numeric: from 0 to 20)
- G2 - second period grade (numeric: from 0 to 20)
- G3 - final grade (numeric: from 0 to 20, output target)

Steps:

1. Import data: grade.csv
2. Data cleaning: NA (Not available)
3. Data visualization
  - a. Transformation
  - b. Descriptive statistics for each of the variables
  - c. Graphs: hist, boxplot, pairs.
4. Fitting linear regression models: We want to explore what factors may affect the final grade.
5. Predictions

## **2.SOLUTION:**

### **2.1. Import data :**

At first, installing the libraries for commands and functions is needed to solve the problem in a clear way.

```
▶ install.packages("dplyr")
install.packages("GGally")
install.packages("corrplot")
install.packages("ggpubr")
install.packages("broom")

[ ] library(ggplot2)
library(devtools)
library(GGally)
library(corrplot)
library(dplyr)
library(broom)
library(ggpubr)
```

After building a group of libraries, inputting the dataset and organizing the variables or factors from the dataset in columns are the following steps.

```
[ ] #https://drive.google.com/file/d/1xBHBU-hB6K4xQv4UTFEzcvjyQKqWWjpZ/view?usp=sharing

[ ] system("gdown --id 1xBHBU-hB6K4xQv4UTFEzcvjyQKqWWjpZ")

[ ] gradeData <- read.table("grade.csv", header = TRUE, sep = ",")
View(gradeData)
```

### **2.2. Data cleaning: NA**

Locating the null value in any factors and replacing them is the significant stage in data cleaning. In order to complete this step, by using the “summary” command.

## summary(gradeData)

```
sex               age      studytime      failures
Length:395      Min.   :15.0    Min.   :1.000   Min.   :0.0000
Class :character 1st Qu.:16.0    1st Qu.:1.000   1st Qu.:0.0000
Mode  :character Median :17.0    Median :2.000   Median :0.0000
                Mean  :16.7    Mean  :2.035   Mean  :0.3342
                3rd Qu.:18.0    3rd Qu.:2.000   3rd Qu.:0.0000
                Max.   :22.0    Max.   :4.000   Max.   :3.0000

higher           absences           G1           G2
Length:395      Min.   : 0.000   Min.   : 3.00   Min.   : 0.00
Class :character 1st Qu.: 0.000   1st Qu.: 8.00   1st Qu.: 9.00
Mode  :character Median : 4.000   Median :11.00   Median :11.00
                Mean  : 5.709   Mean  :10.91   Mean  :10.72
                3rd Qu.: 8.000   3rd Qu.:13.00   3rd Qu.:13.00
                Max.   :75.000   Max.   :19.00   Max.   :19.00
                NA's      :5

G3
Min.   : 0.00
1st Qu.: 8.00
Median :11.00
Mean   :10.42
3rd Qu.:14.00
Max.   :20.00
```

From the result, there are 5 NA values in G2 column, so the next step is the change in those values into the median calculated by rest values in this column.

```
[ ] gradeData[is.na(gradeData)] = median(gradeData$G2, na.rm = TRUE)
    head(gradeData)
```

## 2.3. Data visualization

### 2.3.1. Transformation

To utilize R program to calculate, all factors or values from the dataset must be transferred to numeric type. Before the transformation process is coded, several implies are established for thorough understanding.

- School: GP = 0
- School: MS = 1
- Sex: Male = 0
- Sex: Female = 1
- Address: U = 0
- Address: R = 1
- Famsize: GT3 = 0
- Famsize: LE3 = 1

- Pstatus: A = 0
- Pstatus: T = 1
- Jobs: at\_home = 0
- Jobs: services = 1
- Jobs: teacher = 2
- Jobs: health = 3
- Jobs: other = 4
- Reason: course = 0
- Reason: home = 1
- Reason: reputation = 2
- Reason: other = 3
- Guardian: father = 0
- Guardian: mother = 1
- Guardian: other = 3
- Everything else: no = 0
- Everything else: yes = 1

And then,

```
[ ] gradeData[gradeData == "GP"] <- 0
    gradeData[gradeData == "MS"] <- 1

    gradeData[gradeData == "M"] <- 0
    gradeData[gradeData == "F"] <- 1

    gradeData[gradeData == "U"] <- 0
    gradeData[gradeData == "R"] <- 1

    gradeData[gradeData == "GT3"] <- 0
    gradeData[gradeData == "LE3"] <- 1

    gradeData[gradeData == "A"] <- 0
    gradeData[gradeData == "T"] <- 1

    gradeData[gradeData == "at_home"] <- 0
    gradeData[gradeData == "services"] <- 1
    gradeData[gradeData == "teacher"] <- 2
    gradeData[gradeData == "health"] <- 3
    gradeData$Mjob[gradeData$Mjob == "other"] <- 4
    gradeData$Fjob[gradeData$Fjob == "other"] <- 4

    gradeData[gradeData == "course"] <- 0
    gradeData[gradeData == "home"] <- 1
    gradeData[gradeData == "reputation"] <- 2
    gradeData$reason[gradeData$reason == "other"] <- 3
```

```

gradeData[gradeData == "father"] <- 0
gradeData[gradeData == "mother"] <- 1
gradeData$guardian[gradeData$guardian == "other"] <- 3

gradeData[gradeData == "yes"] <- 0
gradeData[gradeData == "no"] <- 1

head(gradeData)

```

A data frame: 6 × 34

	X	school	sex	age	address	famsize	Pstatus	Medu	Fedu	Mjob	...	famrel	freetime	goout	Dalc	Walc	health	absences	G1	G2	G3
	<int>	<chr>	<chr>	<int>	<chr>	<chr>	<chr>	<int>	<int>	<chr>	...	<int>	<int>	<int>	<int>	<int>	<int>	<int>	<int>	<dbl>	<int>
1	1	0	1	18	0	0	0	4	4	0	...	4	3	4	1	1	3	6	5	6	6
2	2	0	1	17	0	0	1	1	1	0	...	5	3	3	1	1	3	4	5	11	6
3	3	0	1	15	0	1	1	1	1	0	...	4	3	2	2	3	3	10	7	8	10
4	4	0	1	15	0	0	1	4	2	3	...	3	2	2	1	1	5	2	15	14	15
5	5	0	1	16	0	0	1	3	3	4	...	4	3	2	1	2	5	4	6	10	10
6	6	0	0	16	0	1	1	4	3	1	...	5	4	2	1	2	5	10	15	11	15

### 2.3.2. statistics for each of the variables

After the data cleaning and transformation have been done, *class(gradedata* and *summary* command is used to form all the variables into the separate table containing calculating information such as min, 1st Qu., median, mean, 3rd Qu., and max.

```
[ ] class(gradeData$school) <- "numeric"
class(gradeData$sex) <- "numeric"
class(gradeData$address) <- "numeric"
class(gradeData$famsize) <- "numeric"
class(gradeData$Pstatus) <- "numeric"
class(gradeData$Mjob) <- "numeric"
class(gradeData$Fjob) <- "numeric"
class(gradeData$reason) <- "numeric"
class(gradeData$guardian) <- "numeric"
class(gradeData$schoolsup) <- "numeric"
class(gradeData$famsup) <- "numeric"
class(gradeData$paid) <- "numeric"
class(gradeData$activities) <- "numeric"
class(gradeData$nursery) <- "numeric"
class(gradeData$higher) <- "numeric"
class(gradeData$internet) <- "numeric"
class(gradeData$romantic) <- "numeric"

summary(gradeData)
```

Here is the result:

X	school	sex	age
Min. : 1.0	Min. :0.0000	Min. :0.0000	Min. :15.0
1st Qu.: 99.5	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:16.0
Median :198.0	Median :0.0000	Median :1.0000	Median :17.0
Mean :198.0	Mean :0.1165	Mean :0.5266	Mean :16.7
3rd Qu.:296.5	3rd Qu.:0.0000	3rd Qu.:1.0000	3rd Qu.:18.0
Max. :395.0	Max. :1.0000	Max. :1.0000	Max. :22.0
address	famsize	Pstatus	Medu
Min. :0.0000	Min. :0.0000	Min. :0.0000	Min. :0.000
1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:1.0000	1st Qu.:2.000
Median :0.0000	Median :0.0000	Median :1.0000	Median :3.000
Mean :0.2228	Mean :0.2886	Mean :0.8962	Mean :2.749
3rd Qu.:0.0000	3rd Qu.:1.0000	3rd Qu.:1.0000	3rd Qu.:4.000
Max. :1.0000	Max. :1.0000	Max. :1.0000	Max. :4.000
Fedu	Mjob	Fjob	reason
Min. :0.000	Min. :0.000	Min. :0.000	Min. :0.000
1st Qu.:2.000	1st Qu.:1.000	1st Qu.:1.000	1st Qu.:0.000
Median :2.000	Median :2.000	Median :4.000	Median :1.000
Mean :2.522	Mean :2.241	Mean :2.762	Mean :1.081
3rd Qu.:3.000	3rd Qu.:4.000	3rd Qu.:4.000	3rd Qu.:2.000
Max. :4.000	Max. :4.000	Max. :4.000	Max. :3.000
guardian	traveltime	studytime	failures
Min. :0.0000	Min. :1.000	Min. :1.000	Min. :0.0000
1st Qu.:1.0000	1st Qu.:1.000	1st Qu.:1.000	1st Qu.:0.0000
Median :1.0000	Median :1.000	Median :2.000	Median :0.0000
Mean :0.9342	Mean :1.448	Mean :2.035	Mean :0.3342
3rd Qu.:1.0000	3rd Qu.:2.000	3rd Qu.:2.000	3rd Qu.:0.0000
Max. :3.0000	Max. :4.000	Max. :4.000	Max. :3.0000

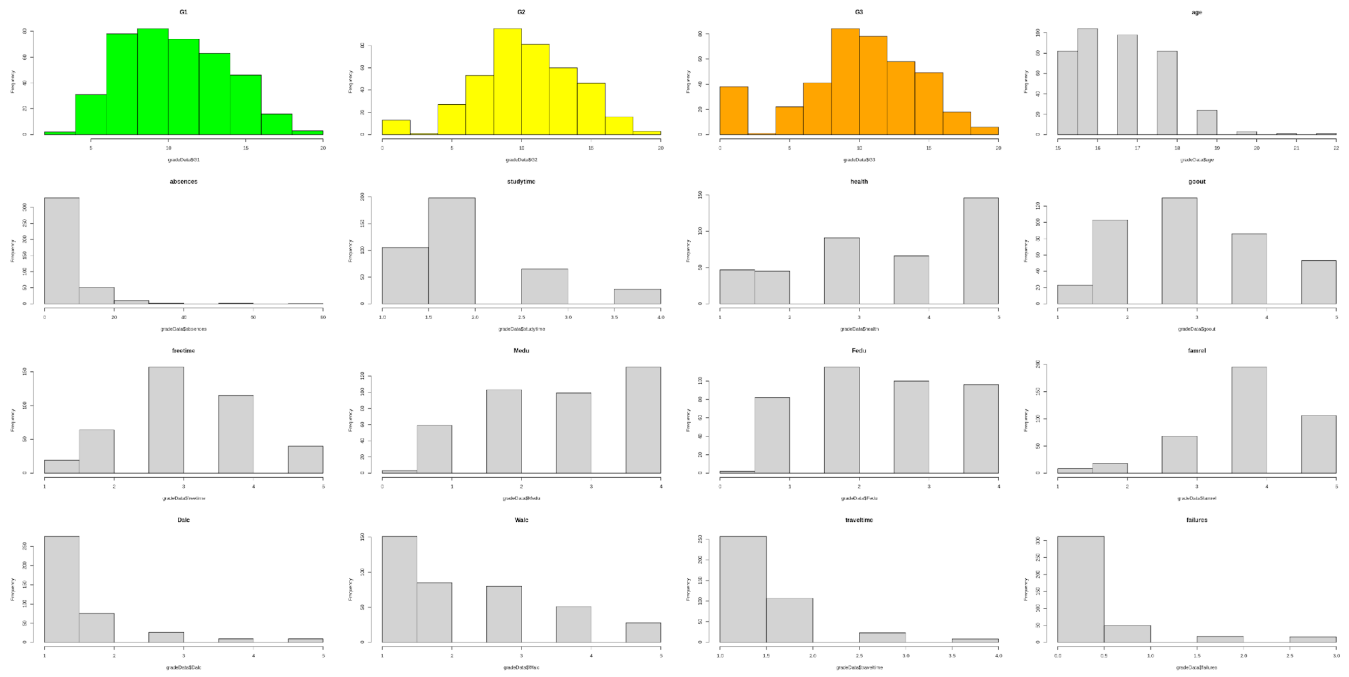


schoolsup	famsup	paid	activities		
Min. :0.0000	Min. :0.0000	Min. :0.0000	Min. :0.0000		
1st Qu.:1.0000	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.:0.0000		
Median :1.0000	Median :0.0000	Median :1.0000	Median :0.0000		
Mean :0.8709	Mean :0.3873	Mean :0.5418	Mean :0.4911		
3rd Qu.:1.0000	3rd Qu.:1.0000	3rd Qu.:1.0000	3rd Qu.:1.0000		
Max. :1.0000	Max. :1.0000	Max. :1.0000	Max. :1.0000		
nursery	higher	internet	romantic	G2	G3
Min. :0.0000	Min. :0.00000	Min. :0.0000	Min. :0.0000	Min. : 0.00	Min. : 0.00
1st Qu.:0.0000	1st Qu.:0.00000	1st Qu.:0.0000	1st Qu.:0.0000	1st Qu.: 9.00	1st Qu.: 8.00
Median :0.0000	Median :0.00000	Median :0.0000	Median :1.0000	Median :11.00	Median :11.00
Mean :0.2051	Mean :0.05063	Mean :0.1671	Mean :0.6658	Mean :10.72	Mean :10.42
3rd Qu.:0.0000	3rd Qu.:0.00000	3rd Qu.:0.0000	3rd Qu.:1.0000	3rd Qu.:13.00	3rd Qu.:14.00
Max. :1.0000	Max. :1.00000	Max. :1.0000	Max. :1.0000	Max. :19.00	Max. :20.00
famrel	freetime	goout	Dalc		
Min. :1.000	Min. :1.000	Min. :1.000	Min. :1.000		
1st Qu.:4.000	1st Qu.:3.000	1st Qu.:2.000	1st Qu.:1.000		
Median :4.000	Median :3.000	Median :3.000	Median :1.000		
Mean :3.944	Mean :3.235	Mean :3.109	Mean :1.481		
3rd Qu.:5.000	3rd Qu.:4.000	3rd Qu.:4.000	3rd Qu.:2.000		
Max. :5.000	Max. :5.000	Max. :5.000	Max. :5.000		
Walc	health	absences	G1		
Min. :1.000	Min. :1.000	Min. : 0.000	Min. : 3.00		
1st Qu.:1.000	1st Qu.:3.000	1st Qu.: 0.000	1st Qu.: 8.00		
Median :2.000	Median :4.000	Median : 4.000	Median :11.00		
Mean :2.291	Mean :3.554	Mean : 5.709	Mean :10.91		
3rd Qu.:3.000	3rd Qu.:5.000	3rd Qu.: 8.000	3rd Qu.:13.00		
Max. :5.000	Max. :5.000	Max. :75.000	Max. :19.00		

### 2.3.3. Graphs: hist, boxplot, pairs

#### *Hist*

```
[ ] options(repr.plot.width=30, repr.plot.height=15)
par(mfrow=c(4,4))
hist(gradeData$G1, main = "G1", col = "green")
hist(gradeData$G2, main = "G2", col = "yellow")
hist(gradeData$G3, main = "G3", col = "orange")
hist(gradeData$age, main = "age")
hist(gradeData$absences, main = "absences")
hist(gradeData$studytime, main = "studytime")
hist(gradeData$health, main = "health")
hist(gradeData$goout, main = "goout")
hist(gradeData$freetime, main = "freetime")
hist(gradeData$Medu, main = "Medu")
hist(gradeData$Fedu, main = "Fedu")
hist(gradeData$famrel, main = "famrel")
hist(gradeData$Dalc, main = "Dalc")
hist(gradeData$Walc, main = "Walc")
hist(gradeData$traveltime, main = "traveltime")
hist(gradeData$failures, main = "failures")
```



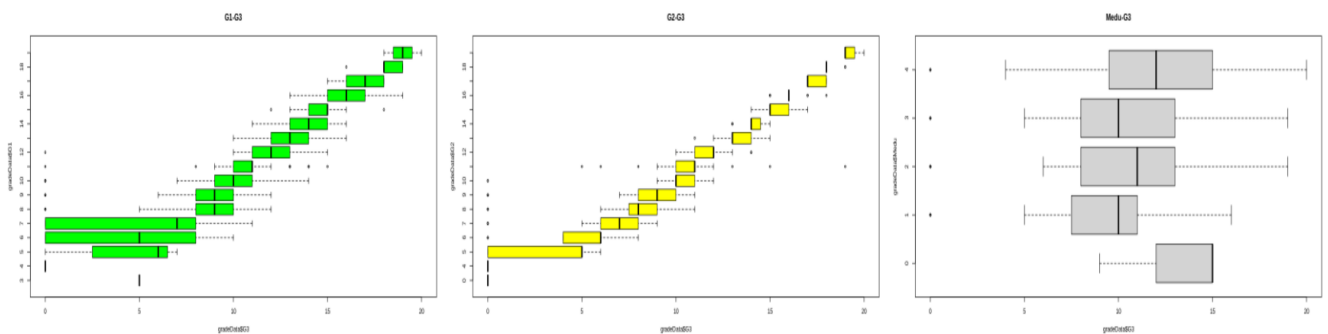
## Boxplot

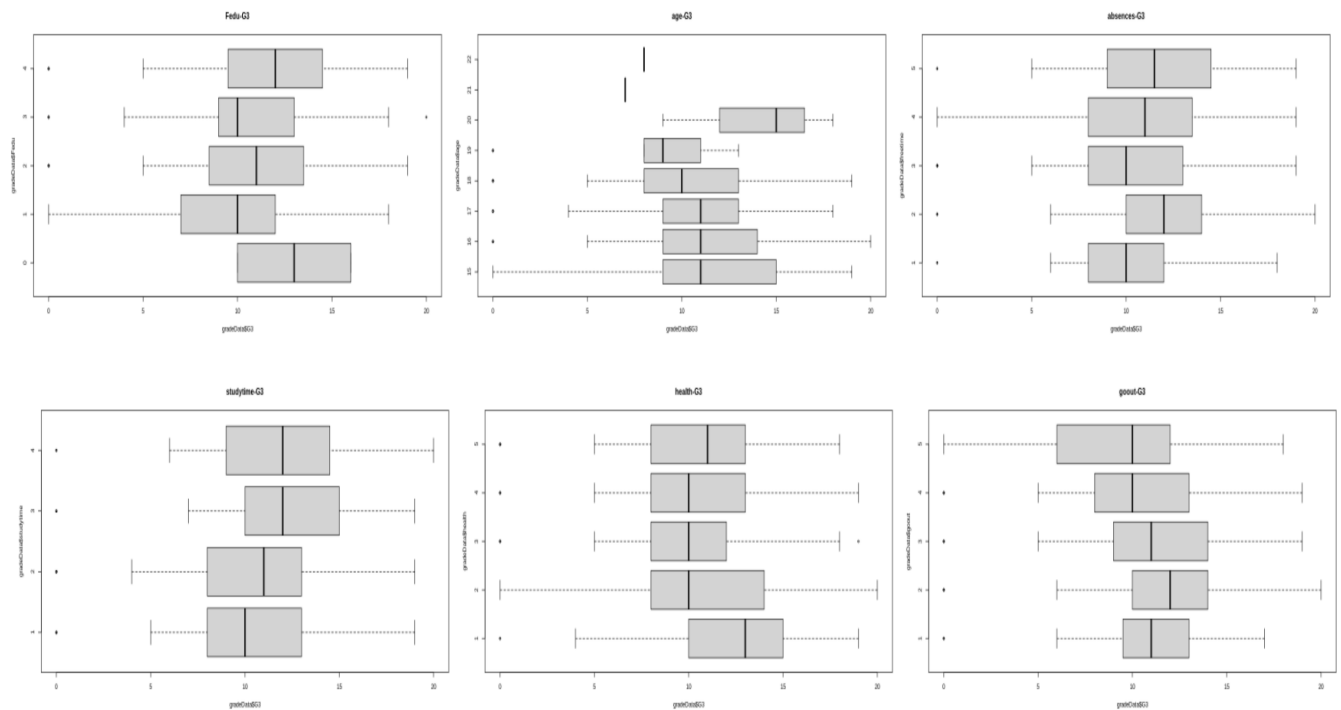
Using *boxplot* command to compare final grade G3 with G1, G2, Medu, Fedu, age, absences, studytime, health and goout.

```

▶ options(repr.plot.width=30, repr.plot.height=15)
par(mfrow=c(3,3))
boxplot(gradeData$G3 ~ gradeData$G1, horizontal = TRUE, main = "G1-G3", col = "green")
boxplot(gradeData$G3 ~ gradeData$G2, horizontal = TRUE, main = "G2-G3", col = "yellow")
boxplot(gradeData$G3 ~ gradeData$Medu, horizontal = TRUE, main = "Medu-G3")
boxplot(gradeData$G3 ~ gradeData$Fedu, horizontal = TRUE, main = "Fedu-G3")
boxplot(gradeData$G3 ~ gradeData$age, horizontal = TRUE, main = "age-G3")
boxplot(gradeData$G3 ~ gradeData$freetime, horizontal = TRUE, main = "absences-G3")
boxplot(gradeData$G3 ~ gradeData$studytime, horizontal = TRUE, main = "studytime-G3")
boxplot(gradeData$G3 ~ gradeData$health, horizontal = TRUE, main = "health-G3")
boxplot(gradeData$G3 ~ gradeData$goout, horizontal = TRUE, main = "goout-G3")

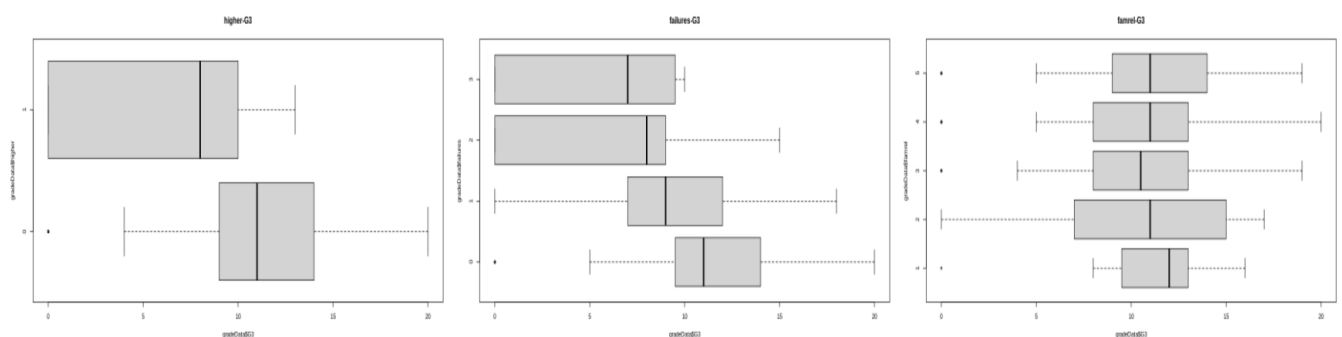
```

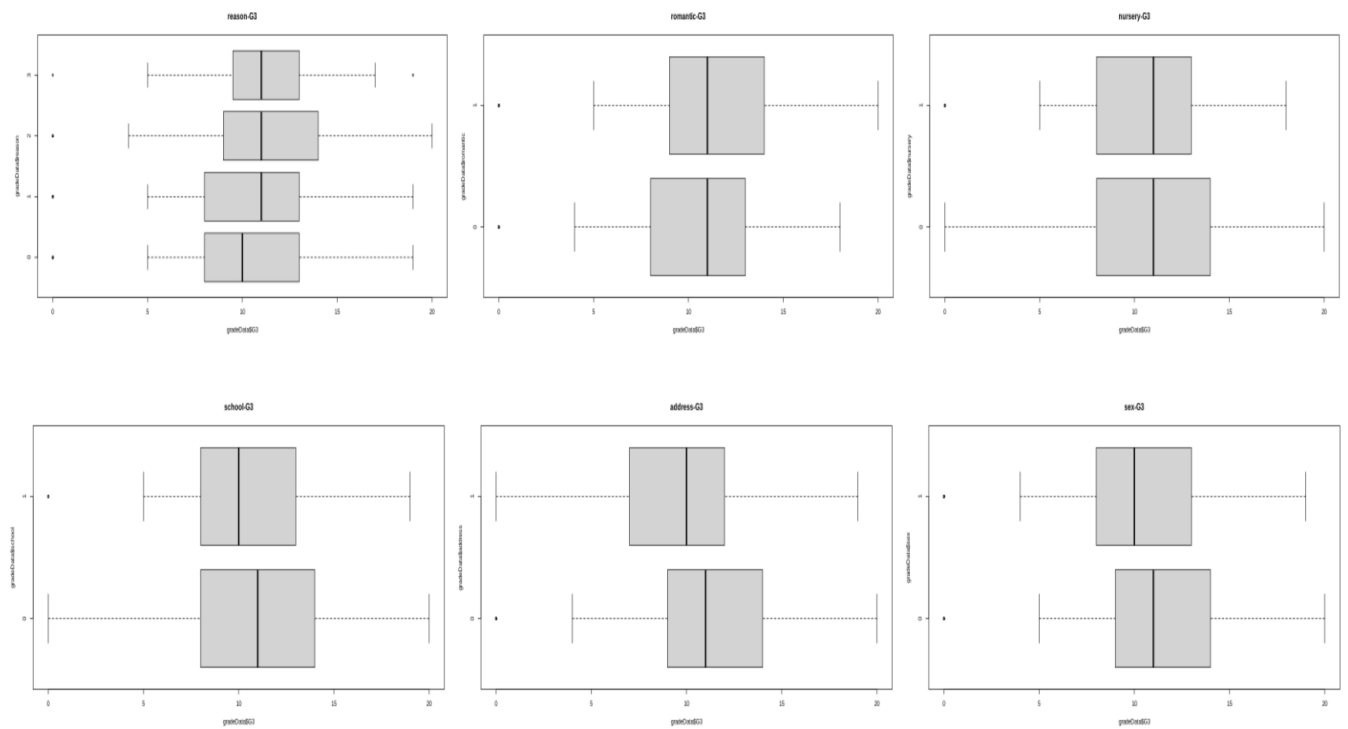




Continuously, using *boxplot* command to compare final grade G3 with school, address, sex, higher, failures, famrel, reason, romantic and nursery.

```
[ ] options(repr.plot.width=30, repr.plot.height=15)
par(mfrow=c(3,3))
boxplot(gradeData$G3 ~ gradeData$school, horizontal = TRUE, main = "school-G3")
boxplot(gradeData$G3 ~ gradeData$address, horizontal = TRUE, main = "address-G3")
boxplot(gradeData$G3 ~ gradeData$sex, horizontal = TRUE, main = "sex-G3")
boxplot(gradeData$G3 ~ gradeData$higher, horizontal = TRUE, main = "higher-G3")
boxplot(gradeData$G3 ~ gradeData$failures, horizontal = TRUE, main = "failures-G3")
boxplot(gradeData$G3 ~ gradeData$famrel, horizontal = TRUE, main = "famrel-G3")
boxplot(gradeData$G3 ~ gradeData$reason, horizontal = TRUE, main = "reason-G3")
boxplot(gradeData$G3 ~ gradeData$romantic, horizontal = TRUE, main = "romantic-G3")
boxplot(gradeData$G3 ~ gradeData$nursery, horizontal = TRUE, main = "nursery-G3")
```

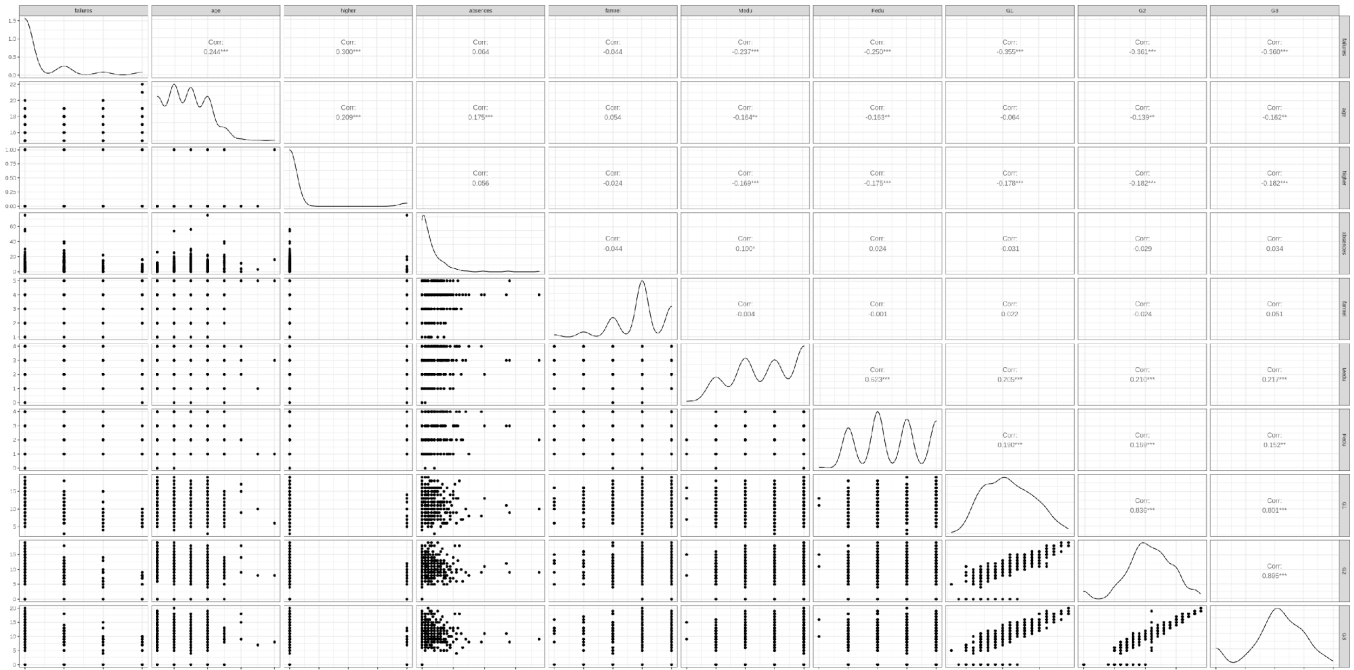




## Pairs

Using *pairs* command to show the statistical relationship between variables (failures, age, higher, absences, famrel, Medu, Fedu, G1, G2 and G3).

```
[ ] options(repr.plot.width=30, repr.plot.height=15)
    ggpairs(subData) + theme_bw()
```



### 2.3.4. Fitting linear regression models

First, using below command to confirm that G3 is a function of the other values and *data = grade* to confirm that R has to compute on dataset called grade.

```
[ ] LinearModel <- lm(G3 ~ ., data=gradeData)
summary(LinearModel)
```

```
[ ] Call:
lm(formula = G3 ~ ., data = gradeData)

Residuals:
    Min       1Q   Median       3Q      Max
-7.5690 -0.6073  0.2500  1.0744  5.7061

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept) -3.827865    2.388814  -1.602  0.10996
X            -0.004157    0.001663  -2.499  0.01291 *
school1      0.939343    0.402766   2.332  0.02025 *
sex1        -0.224026    0.238265  -0.940  0.34774
age          0.043876    0.143198   0.306  0.75948
address1    -0.027359    0.276990  -0.099  0.92138
famsize1     0.092757    0.231310   0.401  0.68866
Pstatus1    -0.298755    0.341233  -0.876  0.38189
Medu         0.108200    0.153445   0.705  0.48119
Fedu        -0.158381    0.131043  -1.209  0.22762
Mjob1        0.316354    0.376445   0.840  0.40127
Mjob2        0.175186    0.491379   0.357  0.72166
Mjob3        0.006884    0.527828   0.013  0.98960
Mjob4        0.266285    0.337450   0.789  0.43058
Fjob1       -0.183476    0.501502  -0.366  0.71469
Fjob2       -0.115774    0.612794  -0.189  0.85026
Fjob3        0.373380    0.678883   0.550  0.58267
Fjob4        0.023485    0.484967   0.048  0.96140
reason1     -0.110051    0.261144  -0.421  0.67371
reason2      0.180781    0.272287   0.664  0.50717
reason3      0.342835    0.387481   0.885  0.37688
guardian1    0.241905    0.257476   0.940  0.34811
```

Based on p-value, constructing 6 models more by eliminating one by one variable from the low p-value to the lowest.

```
[ ] LinearModel_1 <- lm(G3 ~ X +school+ famrel + absences + G1 + G2 , data = gradeData)
LinearModel_2 <- lm(G3 ~ school + famrel + absences + G1 + G2, data= gradeData)
LinearModel_3 <- lm(G3 ~ famrel + absences + G1 + G2, data = gradeData)
LinearModel_4 <- lm(G3 ~ absences + G1 + G2, data = gradeData)
LinearModel_5 <- lm(G3 ~ G1 + G2, data = gradeData)
LinearModel_6 <- lm(G3 ~ G2, data = gradeData)
```

```
[ ] anova(LinearModel_6,LinearModel_5,LinearModel_4,LinearModel_3,LinearModel_2,LinearModel_1,LinearModel)
```

A anova: 7 × 6

	Res.Df	RSS	Df	Sum of Sq	F	Pr(>F)
	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1	393	1642.932	NA	NA	NA	NA
2	392	1565.603	1	77.328443	20.6497771	7.590171e-06
3	391	1534.502	1	31.101716	8.3053980	4.195276e-03
4	390	1495.395	1	39.106886	10.4430976	1.347272e-03
5	389	1494.942	1	0.452328	0.1207896	7.283874e-01
6	388	1425.370	1	69.572059	18.5785134	2.118998e-05
7	352	1318.155	36	107.215052	0.7952970	7.961416e-01

Then, by *anova* command, the comparison between regression models are built.

Observing the Anova data table from the model 1 to 7, the result has illustrated that the model 2 seems to be the finest model to be built a fitting linear regression model compared to other models because of the p-values (the model 2 has smallest value,  $p_2 \sim 0.019$ ).

model 2: `G3 ~ school + famrel + absences + G1 + G2`

Then, having the fitting model below:

```
[ ] guardian3 -0.052696 0.474736 -0.111 0.91168
traveltime 0.101313 0.160785 0.630 0.52903
studytime -0.099203 0.137499 -0.721 0.47109
failures -0.193218 0.167203 -1.156 0.24863
schoolsup1 -0.449382 0.326983 -1.374 0.17021
famsup1 -0.125593 0.230105 -0.546 0.58554
paid1 -0.256815 0.226228 -1.135 0.25706
activities1 0.323779 0.210157 1.541 0.12430
nursery1 0.221102 0.259206 0.853 0.39424
higher1 -0.247778 0.513630 -0.482 0.62982
internet1 0.096594 0.294700 0.328 0.74328
romantic1 0.209608 0.225673 0.929 0.35362
famrel 0.347329 0.116769 2.975 0.00314 **
freetime 0.025411 0.112307 0.226 0.82113
goout -0.015578 0.107024 -0.146 0.88436
Dalc -0.212028 0.156003 -1.359 0.17497
Walc 0.210583 0.117030 1.799 0.07281 .
health 0.044192 0.076199 0.580 0.56232
absences 0.041264 0.013654 3.022 0.00269 **
G1 0.305410 0.060386 5.058 6.85e-07 ***
G2 0.873046 0.052079 16.764 < 2e-16 ***
---
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 1.935 on 352 degrees of freedom
Multiple R-squared: 0.8406, Adjusted R-squared: 0.8216
F-statistic: 44.2 on 42 and 352 DF, p-value: < 2.2e-16
```

$$G3 = -3.77114 + 0.93638 \times G2 + 0.23115 \times G1 + 0.35501 \times \text{famrel} \\ + 0.03726 \times \text{absences} + 0.10628 \times \text{school1}$$

```
[ ] summary(LinearModel_2)
```

Call:

```
lm(formula = G3 ~ school + famrel + absences + G1 + G2, data = gradeData)
```

Residuals:

	Min	1Q	Median	3Q	Max
	-9.3242	-0.4523	0.2072	1.0080	7.3526

Coefficients:

	Estimate	Std. Error	t value	Pr(> t )	
(Intercept)	-3.77114	0.56316	-6.696	7.49e-11	***
school1	0.10628	0.30980	0.343	0.73173	
famrel	0.35501	0.11080	3.204	0.00147	**
absences	0.03726	0.01241	3.002	0.00285	**
G1	0.23115	0.05443	4.247	2.72e-05	***
G2	0.93638	0.04870	19.226	< 2e-16	***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

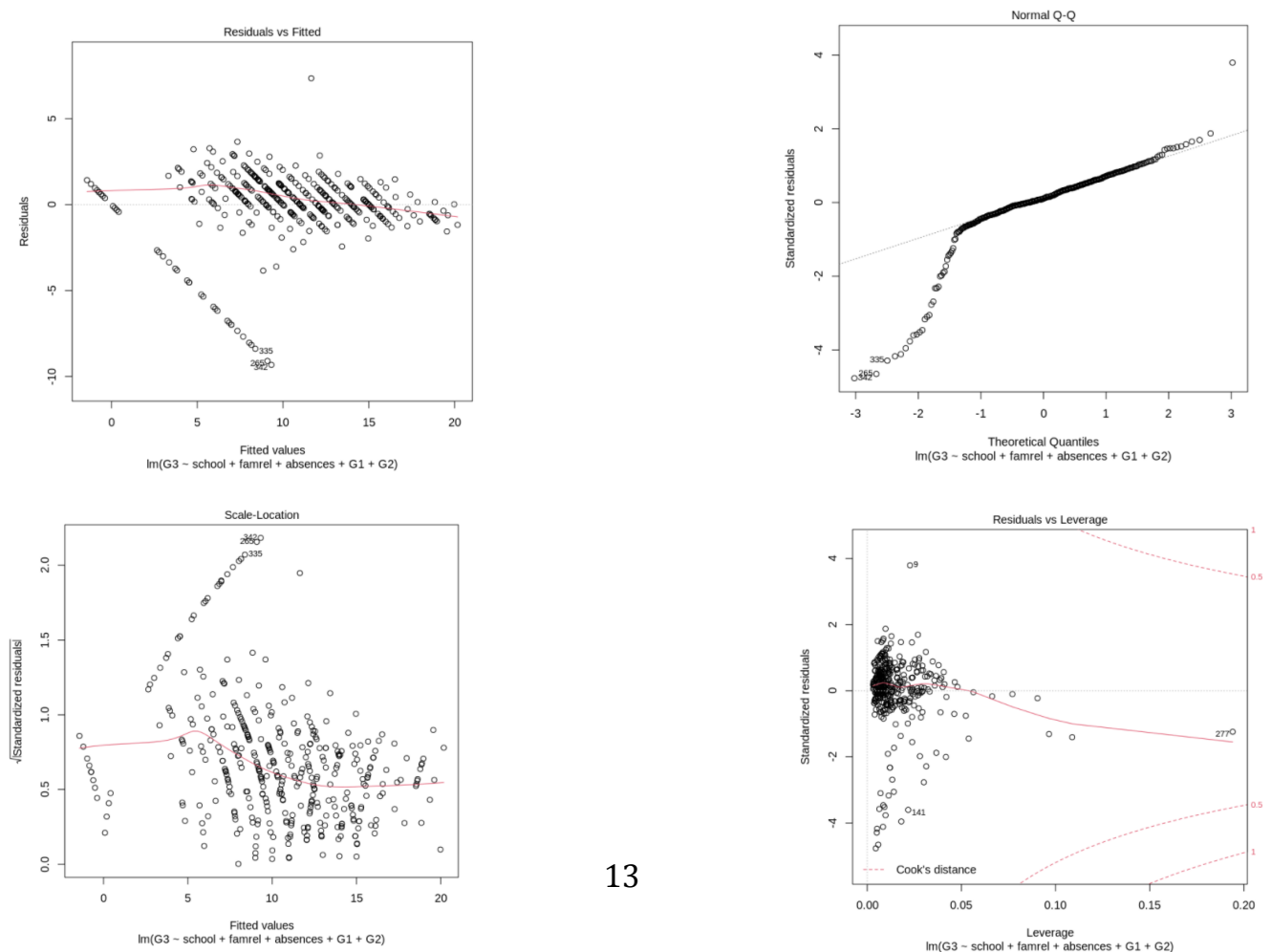
Residual standard error: 1.96 on 389 degrees of freedom

Multiple R-squared: 0.8192, Adjusted R-squared: 0.8169

F-statistic: 352.6 on 5 and 389 DF, p-value: < 2.2e-16

Following that, plotting that model:

```
[ ] plot(LinearModel_2)
```



## 2.4. Predictions

### 2.4.1. Evaluation

First, in order to evaluate whether those students passed or failed based on final grade, the condition order: *if their final grade is not less than 10, they are passed*, is used to *evaluate*. After that step, the prediction data also is built as the same function above but predict\_G3.

```
[ ] evaluate = gradeData$G3
evaluate = ifelse(evaluate >=10, "pass", "fail")
observe = table(evaluate)
View (observe)
```

```
evaluate
fail pass
130 265
```

```
[ ] Predict_G3 = predict(LinearModel_2, gradeData)
Predict_G3 = ifelse(Predict_G3 >=10, "pass", "fail")
observe = table(Predict_G3)
View (observe)
```

```
Predict_G3
fail pass
185 210
```

The percent error for students who failed is  $\frac{185-130}{130} \times 100\% = 42.31\%$

The percent error for students who passed is  $\frac{265-210}{265} \times 100\% = 20.75\%$



### 2.4.2. Prediction a new data

First, creating a data frame to predict the final grade. As below, the new data frame is given as an example

```
newd = data.frame(school = 1, famrel = 5, absences = 20, G1 = 10, G2 = 11)
```

Then, using *predict* command to compute G3 (final grade) from the others factor in the data frame.

```
G3_predict = predict(LinearModel_2, newd)
```

And using *round* command to round the result

```
round(G3_predict, digits = 4)
```

```
1: 11.4671
```

Finally, the final result computed by R is 11.4671.

## **REFERENCES**

Our source code:

<https://colab.research.google.com/drive/1zOCpF4MARuGzPXpdl-peQ7Amtbc-aNv?usp=sharing> (we run directly on the google collab and then converting to the R file)

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