

Meghana_Nadig_Assignment6

Problem 1:

```
# Importing Data
```

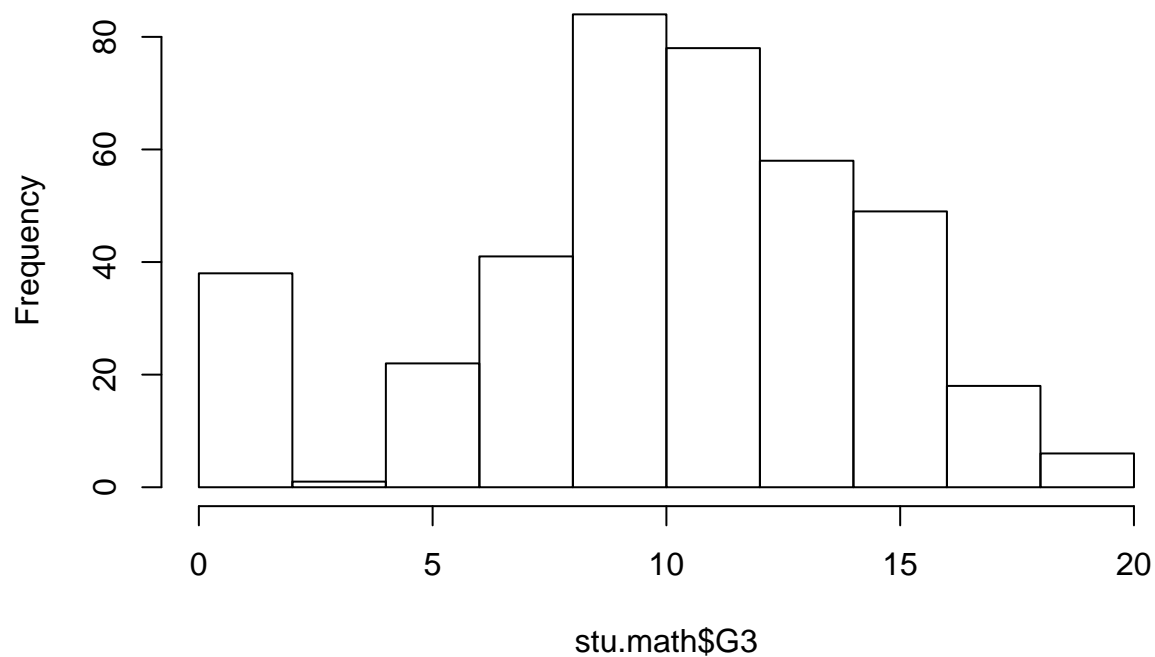
```
stu.math <- read.csv("C:/Users/Meghana Nadig/Downloads/student-mat.csv",sep=";",header=TRUE,stringsAsFactors=TRUE)
str(stu.math)
```

```
## 'data.frame': 395 obs. of 33 variables:
## $ school : Factor w/ 2 levels "GP","MS": 1 1 1 1 1 1 1 1 1 1 ...
## $ sex : Factor w/ 2 levels "F","M": 1 1 1 1 1 2 2 1 2 2 ...
## $ age : int 18 17 15 15 16 16 16 17 15 15 ...
## $ address : Factor w/ 2 levels "R","U": 2 2 2 2 2 2 2 2 2 2 ...
## $ famsize : Factor w/ 2 levels "GT3","LE3": 1 1 2 1 1 2 2 1 2 1 ...
## $ Pstatus : Factor w/ 2 levels "A","T": 1 2 2 2 2 2 2 1 1 2 ...
## $ Medu : int 4 1 1 4 3 4 2 4 3 3 ...
## $ Fedu : int 4 1 1 2 3 3 2 4 2 4 ...
## $ Mjob : Factor w/ 5 levels "at_home","health",...: 1 1 1 2 3 4 3 3 4 3 ...
## $ Fjob : Factor w/ 5 levels "at_home","health",...: 5 3 3 4 3 3 3 5 3 3 ...
## $ reason : Factor w/ 4 levels "course","home",...: 1 1 3 2 2 4 2 2 2 2 ...
## $ guardian : Factor w/ 3 levels "father","mother",...: 2 1 2 2 1 2 2 2 2 2 ...
## $ traveltime: int 2 1 1 1 1 1 1 2 1 1 ...
## $ studytime : int 2 2 2 3 2 2 2 2 2 2 ...
## $ failures : int 0 0 3 0 0 0 0 0 0 0 ...
## $ schoolsup : Factor w/ 2 levels "no","yes": 2 1 2 1 1 1 1 2 1 1 ...
## $ famsup : Factor w/ 2 levels "no","yes": 1 2 1 2 2 2 1 2 2 2 ...
## $ paid : Factor w/ 2 levels "no","yes": 1 1 2 2 2 2 1 1 2 2 ...
## $ activities: Factor w/ 2 levels "no","yes": 1 1 1 2 1 2 1 1 1 2 ...
## $ nursery : Factor w/ 2 levels "no","yes": 2 1 2 2 2 2 2 2 2 2 ...
## $ higher : Factor w/ 2 levels "no","yes": 2 2 2 2 2 2 2 2 2 2 ...
## $ internet : Factor w/ 2 levels "no","yes": 1 2 2 2 1 2 2 1 2 2 ...
## $ romantic : Factor w/ 2 levels "no","yes": 1 1 1 2 1 1 1 1 1 1 ...
## $ famrel : int 4 5 4 3 4 5 4 4 4 5 ...
## $ freetime : int 3 3 3 2 3 4 4 1 2 5 ...
## $ goout : int 4 3 2 2 2 2 4 4 2 1 ...
## $ Dalc : int 1 1 2 1 1 1 1 1 1 1 ...
## $ Walc : int 1 1 3 1 2 2 1 1 1 1 ...
## $ health : int 3 3 3 5 5 5 3 1 1 5 ...
## $ absences : int 6 4 10 2 4 10 0 6 0 0 ...
## $ G1 : int 5 5 7 15 6 15 12 6 16 14 ...
## $ G2 : int 6 5 8 14 10 15 12 5 18 15 ...
## $ G3 : int 6 6 10 15 10 15 11 6 19 15 ...
```

```
# Checking the distribution of G3
```

```
hist(stu.math$G3)
```

Histogram of stu.math\$G3



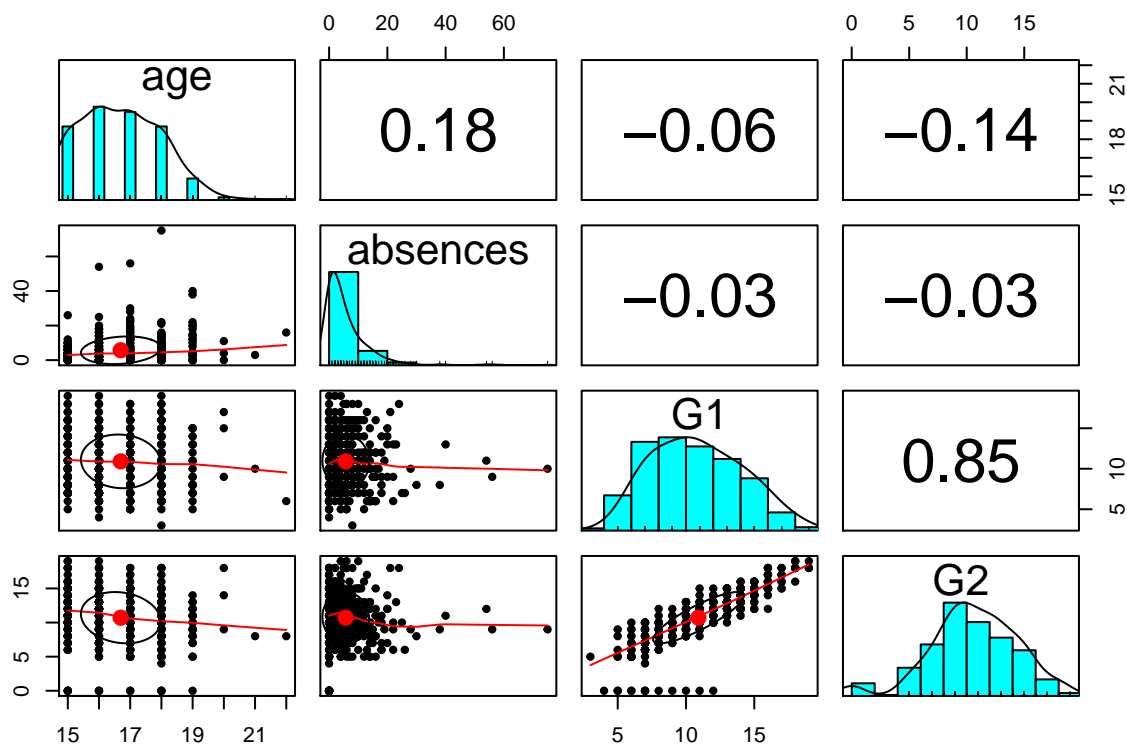
```
#install.packages("psych")
```

```
library(psych)
```

```
# Creating a Scatter Plot
```

```
# Selecting 4 continuous variables: "age", "absences", "G1", "G2"
```

```
pairs.panels(stu.math[c("age", "absences", "G1", "G2")])
```



```
# Building a Regression Model
# Selecting 4 variables: "paid"(extra paid classes within the course subject),"age" (student's age),"G1"
# Converting categorical variable to binary indicator (Dummy code)
stu.math$paid.bin <- ifelse(stu.math$paid == "yes", 1, 0)

stu.math$paid.bin

##      [1] 0 0 1 1 1 1 0 0 1 1 1 0 1 1 0 0 1 0 0 1 0 1 0 0 1 1 1 1 0 1 1 0 0 0 1
##     [36] 0 0 0 1 1 0 0 0 0 0 1 0 0 1 0 1 1 0 1 1 1 1 0 0 0 0 0 0 1 0 0 0 1 1 0
##     [71] 1 0 0 0 1 1 0 1 0 0 1 1 1 1 0 1 0 0 1 0 1 1 0 1 0 1 0 1 0 1 1 0 0 1 1
##    [106] 0 1 0 1 1 0 0 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 0 1 0 0 0 0 0 0 0 0 0 0
##    [141] 0 0 1 0 0 1 0 1 0 0 0 0 1 0 0 1 0 0 1 0 0 0 0 0 0 0 0 0 1 1 1 0 1 1 0 1
##    [176] 1 1 0 0 0 1 1 0 0 1 0 1 1 0 0 0 0 1 0 0 0 0 1 0 1 0 1 1 0 1 1 1 1 1 1
##    [211] 1 1 1 1 1 1 1 1 1 1 1 0 0 0 1 0 0 0 1 1 0 0 1 0 1 1 0 1 0 0 0 0 1 0 0 1
##    [246] 0 0 0 0 1 0 0 0 0 0 0 1 0 0 1 1 1 0 0 1 1 1 1 0 1 1 1 1 1 1 1 0 1 0 1
##    [281] 1 0 1 0 0 1 1 1 1 1 1 1 0 1 1 1 1 1 1 1 0 1 1 1 0 1 0 1 0 0 0 0 0 1 0
##    [316] 0 1 1 1 1 1 0 1 0 0 0 0 0 1 1 0 0 0 0 0 0 1 1 0 1 1 0 0 0 0 0 0 1 1 0
##    [351] 0 1 0 0 1 0 1 0 0 0 0 0 0 1 0 1 1 1 1 1 0 1 0 0 0 0 1 1 1 1 1 0 0 0 1
##    [386] 1 1 0 1 0 1 0 0 0 0

#install.packages("caTools")
library(caTools)

split <- sample.split(stu.math, SplitRatio = 0.8)

training <- subset(stu.math, split == TRUE)
```

```

testing <- subset(stu.math, split == FALSE)

# Training the model

m <- lm(G3 ~ paid.bin + age + G1 + G2, data = training)

summary(m)

##
## Call:
## lm(formula = G3 ~ paid.bin + age + G1 + G2, data = training)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.3026 -0.3516  0.3175  1.1026  3.5827
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.29837    1.60270   0.186  0.85244
## paid.bin      0.05722    0.23528   0.243  0.80801
## age          -0.14078    0.09087  -1.549  0.12235
## G1             0.18344    0.06456   2.841  0.00479 **
## G2             0.97039    0.05671  17.110 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.045 on 310 degrees of freedom
## Multiple R-squared:  0.8097, Adjusted R-squared:  0.8072
## F-statistic: 329.7 on 4 and 310 DF, p-value: < 2.2e-16

Backfitting (p-value)

# Removing paid.bin (highest p-value) from the model

m1 <- lm(G3 ~ age + G1 + G2, data = training)

summary(m1)

##
## Call:
## lm(formula = G3 ~ age + G1 + G2, data = training)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.3263 -0.3460  0.3223  1.1195  3.5571
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  0.33389    1.59362   0.210  0.83418
## age          -0.14160    0.09067  -1.562  0.11940
## G1             0.18194    0.06417   2.835  0.00488 **
## G2             0.97217    0.05615  17.314 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##

```

```
## Residual standard error: 2.042 on 311 degrees of freedom
## Multiple R-squared:  0.8096, Adjusted R-squared:  0.8078
## F-statistic: 440.9 on 3 and 311 DF,  p-value: < 2.2e-16
```

Removing age (second highest p-value) from the model

```
m2 <- lm(G3 ~ G1 + G2, data = training)
```

```
summary(m2)
```

```
##
## Call:
## lm(formula = G3 ~ G1 + G2, data = training)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -9.5074 -0.3159  0.3341  1.0313  3.8096
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -2.07773    0.39414  -5.272 2.53e-07 ***
## G1           0.17440    0.06414   2.719 0.00691 **
## G2           0.98412    0.05576  17.650 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 2.047 on 312 degrees of freedom
## Multiple R-squared:  0.8081, Adjusted R-squared:  0.8069
## F-statistic: 657.1 on 2 and 312 DF,  p-value: < 2.2e-16
```

Regression Equation = $-1.98 + 0.18G1 + 0.96G2$

The p-value backward elimination technique is used.

Predicting G3

```
G3_pred <- predict(m2,testing)
G3_pred
```

```
##      3      9      17      23      24      28      31
## 7.015993 18.426742 13.967082 15.299995 12.982966 16.284111 10.317139
##      37      43      51      57      58      62      65
## 16.284111 18.949938 12.808567 15.125597 15.125597 7.539189 9.507421
##      71      77      85      91      92      96      99
## 14.951198 10.665936 9.333022 6.031877 17.442626 8.984225 13.618284
##      105     111     119     125     126     130     133
## 18.426742 19.759656 6.380674 6.206275 12.982966 18.775540 12.459770
##      139     145     153     159     160     164     167
## 12.173248 -1.205734 9.507421 15.648792 11.475653 9.507421 9.507421
##      173     179     187     193     194     198     201
## 11.014733 7.539189 11.650052 7.015993 8.174508 8.348906 16.458510
##      207     213     221     227     228     232     235
## 6.031877 12.808567 4.873362 15.474394 10.840335 10.665936 6.380674
##      241     247     255     261     262     266     269
## 11.824451 11.824451 11.126856 18.601141 7.190391 17.617025 8.523305
##      275     281     289     295     296     300     303
## 9.507421 7.190391 14.315879 13.157364 12.173248 15.474394 12.347647
```

```
##      309      315      323      329      330      334      337
## 12.347647 13.331763 10.665936 8.523305 14.141480 7.190391 13.157364
##      343      349      357      363      364      368      371
## 15.474394 14.951198 12.808567 10.665936 15.474394 5.047760 6.031877
##      377      383      391
## 14.315879 10.665936 8.348906
```

95% Confidence Interval

```
# Residual Standard error is 2.023
```

```
# Finding 95% CI
```

```
G3_pred[1] - 1.96*2.023
```

```
##      3
```

```
## 3.050913
```

```
G3_pred[1] + 1.96*2.023
```

```
##      3
```

```
## 10.98107
```

Therefore, the 95% CI is between 3.06 and 11

RMSE

```
# Testing Model
```

```
sqerr_test <- (stu.math[33] - G3_pred)^2
```

```
msqerr_test <- mean(sqerr_test)
```

```
rmse_test <- sqrt(msqerr_test)
```

```
rmse_test
```

```
## [1] 5.963049
```

Therefore, the RMSE for testing model is 5.77

```
# Training Model
```

```
m3 <- lm(G3 ~ G1 + G2, data = testing)
```

```
G3_pred_train <- predict(m2,training)
```

```
sqerr_train <- (stu.math[33] - G3_pred_train)^2
```

```
msqerr_train <- mean(sqerr_train)
```

```
rmse_train <- sqrt(msqerr_train)
```

```
rmse_train
```

```
## [1] 5.969406
```

Therefore, the RMSE for training model is 6.22

Problem 2:

```
stu.math1 <- stu.math
```

```
# Adding new column
```

```
stu.math1$paid.bin <- NULL
```

```
stu.math1["PF"] <- NA
```

```
stu.math1$PF <- ifelse(stu.math1$G3 < 10, "F", "P")
```

```
stu.math1
```

```
##      school sex age address famsize Pstatus Medu Fedu      Mjob      Fjob
```

## 1	GP	F	18	U	GT3	A	4	4	at_home	teacher
## 2	GP	F	17	U	GT3	T	1	1	at_home	other
## 3	GP	F	15	U	LE3	T	1	1	at_home	other
## 4	GP	F	15	U	GT3	T	4	2	health	services
## 5	GP	F	16	U	GT3	T	3	3	other	other
## 6	GP	M	16	U	LE3	T	4	3	services	other
## 7	GP	M	16	U	LE3	T	2	2	other	other
## 8	GP	F	17	U	GT3	A	4	4	other	teacher
## 9	GP	M	15	U	LE3	A	3	2	services	other
## 10	GP	M	15	U	GT3	T	3	4	other	other
## 11	GP	F	15	U	GT3	T	4	4	teacher	health
## 12	GP	F	15	U	GT3	T	2	1	services	other
## 13	GP	M	15	U	LE3	T	4	4	health	services
## 14	GP	M	15	U	GT3	T	4	3	teacher	other
## 15	GP	M	15	U	GT3	A	2	2	other	other
## 16	GP	F	16	U	GT3	T	4	4	health	other
## 17	GP	F	16	U	GT3	T	4	4	services	services
## 18	GP	F	16	U	GT3	T	3	3	other	other
## 19	GP	M	17	U	GT3	T	3	2	services	services
## 20	GP	M	16	U	LE3	T	4	3	health	other
## 21	GP	M	15	U	GT3	T	4	3	teacher	other
## 22	GP	M	15	U	GT3	T	4	4	health	health
## 23	GP	M	16	U	LE3	T	4	2	teacher	other
## 24	GP	M	16	U	LE3	T	2	2	other	other
## 25	GP	F	15	R	GT3	T	2	4	services	health
## 26	GP	F	16	U	GT3	T	2	2	services	services
## 27	GP	M	15	U	GT3	T	2	2	other	other
## 28	GP	M	15	U	GT3	T	4	2	health	services
## 29	GP	M	16	U	LE3	A	3	4	services	other
## 30	GP	M	16	U	GT3	T	4	4	teacher	teacher
## 31	GP	M	15	U	GT3	T	4	4	health	services
## 32	GP	M	15	U	GT3	T	4	4	services	services
## 33	GP	M	15	R	GT3	T	4	3	teacher	at_home
## 34	GP	M	15	U	LE3	T	3	3	other	other
## 35	GP	M	16	U	GT3	T	3	2	other	other
## 36	GP	F	15	U	GT3	T	2	3	other	other
## 37	GP	M	15	U	LE3	T	4	3	teacher	services
## 38	GP	M	16	R	GT3	A	4	4	other	teacher
## 39	GP	F	15	R	GT3	T	3	4	services	health
## 40	GP	F	15	R	GT3	T	2	2	at_home	other
## 41	GP	F	16	U	LE3	T	2	2	other	other
## 42	GP	M	15	U	LE3	T	4	4	teacher	other
## 43	GP	M	15	U	GT3	T	4	4	services	teacher
## 44	GP	M	15	U	GT3	T	2	2	services	services
## 45	GP	F	16	U	LE3	T	2	2	other	at_home
## 46	GP	F	15	U	LE3	A	4	3	other	other
## 47	GP	F	16	U	LE3	A	3	3	other	services
## 48	GP	M	16	U	GT3	T	4	3	health	services
## 49	GP	M	15	U	GT3	T	4	2	teacher	other
## 50	GP	F	15	U	GT3	T	4	4	services	teacher
## 51	GP	F	16	U	LE3	T	2	2	services	services
## 52	GP	F	15	U	LE3	T	4	2	health	other
## 53	GP	M	15	U	LE3	A	4	2	health	health
## 54	GP	F	15	U	GT3	T	4	4	services	services

## 55	GP	F	15	U	LE3	A	3	3	other	other
## 56	GP	F	16	U	GT3	A	2	1	other	other
## 57	GP	F	15	U	GT3	A	4	3	services	services
## 58	GP	M	15	U	GT3	T	4	4	teacher	health
## 59	GP	M	15	U	LE3	T	1	2	other	at_home
## 60	GP	F	16	U	GT3	T	4	2	services	other
## 61	GP	F	16	R	GT3	T	4	4	health	teacher
## 62	GP	F	16	U	GT3	T	1	1	services	services
## 63	GP	F	16	U	LE3	T	1	2	other	services
## 64	GP	F	16	U	GT3	T	4	3	teacher	health
## 65	GP	F	15	U	LE3	T	4	3	services	services
## 66	GP	F	16	U	LE3	T	4	3	teacher	services
## 67	GP	M	15	U	GT3	A	4	4	other	services
## 68	GP	F	16	U	GT3	T	3	1	services	other
## 69	GP	F	15	R	LE3	T	2	2	health	services
## 70	GP	F	15	R	LE3	T	3	1	other	other
## 71	GP	M	16	U	GT3	T	3	1	other	other
## 72	GP	M	15	U	GT3	T	4	2	other	other
## 73	GP	F	15	R	GT3	T	1	1	other	other
## 74	GP	M	16	U	GT3	T	3	1	other	other
## 75	GP	F	16	U	GT3	T	3	3	other	services
## 76	GP	M	15	U	GT3	T	4	3	teacher	other
## 77	GP	M	15	U	GT3	T	4	0	teacher	other
## 78	GP	F	16	U	GT3	T	2	2	other	other
## 79	GP	M	17	U	GT3	T	2	1	other	other
## 80	GP	F	16	U	GT3	T	3	4	at_home	other
## 81	GP	M	15	U	GT3	T	2	3	other	services
## 82	GP	M	15	U	GT3	T	2	3	other	other
## 83	GP	F	15	U	LE3	T	3	2	services	other
## 84	GP	M	15	U	LE3	T	2	2	services	services
## 85	GP	F	15	U	GT3	T	1	1	other	other
## 86	GP	F	15	U	GT3	T	4	4	services	services
## 87	GP	F	16	U	LE3	T	2	2	at_home	other
## 88	GP	F	15	U	GT3	T	4	2	other	other
## 89	GP	M	16	U	GT3	T	2	2	services	other
## 90	GP	M	16	U	LE3	A	4	4	teacher	health
## 91	GP	F	16	U	GT3	T	3	3	other	other
## 92	GP	F	15	U	GT3	T	4	3	services	other
## 93	GP	F	16	U	LE3	T	3	1	other	other
## 94	GP	F	16	U	GT3	T	4	2	teacher	services
## 95	GP	M	15	U	LE3	T	2	2	services	health
## 96	GP	F	15	R	GT3	T	1	1	at_home	other
## 97	GP	M	16	R	GT3	T	4	3	services	other
## 98	GP	F	16	U	GT3	T	2	1	other	other
## 99	GP	F	16	U	GT3	T	4	4	other	other
## 100	GP	F	16	U	GT3	T	4	3	other	at_home
## 101	GP	M	16	U	GT3	T	4	4	services	services
## 102	GP	M	16	U	GT3	T	4	4	services	teacher
## 103	GP	M	15	U	GT3	T	4	4	services	other
## 104	GP	F	15	U	GT3	T	3	2	services	other
## 105	GP	M	15	U	GT3	A	3	4	services	other
## 106	GP	F	15	U	GT3	A	3	3	other	health
## 107	GP	F	15	U	GT3	T	2	2	other	other
## 108	GP	M	16	U	GT3	T	3	3	services	other

## 109	GP	M	15	R	GT3	T	4	4	other	other
## 110	GP	F	16	U	LE3	T	4	4	health	health
## 111	GP	M	15	U	LE3	A	4	4	teacher	teacher
## 112	GP	F	16	R	GT3	T	3	3	services	other
## 113	GP	F	16	U	GT3	T	2	2	at_home	other
## 114	GP	M	15	U	LE3	T	4	2	teacher	other
## 115	GP	M	15	R	GT3	T	2	1	health	services
## 116	GP	M	16	U	GT3	T	4	4	teacher	teacher
## 117	GP	M	15	U	GT3	T	4	4	other	teacher
## 118	GP	M	16	U	GT3	T	3	3	other	services
## 119	GP	M	17	R	GT3	T	1	3	other	other
## 120	GP	M	15	U	GT3	T	3	4	other	other
## 121	GP	F	15	U	GT3	T	1	2	at_home	services
## 122	GP	M	15	U	GT3	T	2	2	services	services
## 123	GP	F	16	U	LE3	T	2	4	other	health
## 124	GP	M	16	U	GT3	T	4	4	health	other
## 125	GP	F	16	U	GT3	T	2	2	other	other
## 126	GP	M	15	U	GT3	T	3	4	services	services
## 127	GP	F	15	U	LE3	A	3	4	other	other
## 128	GP	F	19	U	GT3	T	0	1	at_home	other
## 129	GP	M	18	R	GT3	T	2	2	services	other
## 130	GP	M	16	R	GT3	T	4	4	teacher	teacher
## 131	GP	F	15	R	GT3	T	3	4	services	teacher
## 132	GP	F	15	U	GT3	T	1	1	at_home	other
## 133	GP	F	17	U	LE3	T	2	2	other	other
## 134	GP	F	16	U	GT3	A	3	4	services	other
## 135	GP	M	15	R	GT3	T	3	4	at_home	teacher
## 136	GP	F	15	U	GT3	T	4	4	services	at_home
## 137	GP	M	17	R	GT3	T	3	4	at_home	other
## 138	GP	F	16	U	GT3	A	3	3	other	other
## 139	GP	M	16	U	LE3	T	1	1	services	other
## 140	GP	F	15	U	GT3	T	4	4	teacher	teacher
## 141	GP	M	15	U	GT3	T	4	3	teacher	services
## 142	GP	M	16	U	LE3	T	2	2	services	services
## 143	GP	F	15	U	GT3	T	4	4	teacher	services
## 144	GP	F	16	U	LE3	T	1	1	at_home	at_home
## 145	GP	M	17	U	GT3	T	2	1	other	other
## 146	GP	F	15	U	GT3	T	1	1	other	services
## 147	GP	F	15	U	GT3	T	3	2	health	services
## 148	GP	F	15	U	GT3	T	1	2	at_home	other
## 149	GP	M	16	U	GT3	T	4	4	teacher	teacher
## 150	GP	M	15	U	LE3	A	2	1	services	other
## 151	GP	M	18	U	LE3	T	1	1	other	other
## 152	GP	M	16	U	LE3	T	2	1	at_home	other
## 153	GP	F	15	R	GT3	T	3	3	services	services
## 154	GP	M	19	U	GT3	T	3	2	services	at_home
## 155	GP	F	17	U	GT3	T	4	4	other	teacher
## 156	GP	M	15	R	GT3	T	2	3	at_home	services
## 157	GP	M	17	R	LE3	T	1	2	other	other
## 158	GP	F	18	R	GT3	T	1	1	at_home	other
## 159	GP	M	16	R	GT3	T	2	2	at_home	other
## 160	GP	M	16	U	GT3	T	3	3	other	services
## 161	GP	M	17	R	LE3	T	2	1	at_home	other
## 162	GP	M	15	R	GT3	T	3	2	other	other

## 163	GP	M	16	U	LE3	T	1	2	other	other
## 164	GP	M	17	U	GT3	T	1	3	at_home	services
## 165	GP	M	17	R	LE3	T	1	1	other	services
## 166	GP	M	16	U	GT3	T	3	2	services	services
## 167	GP	M	16	U	GT3	T	2	2	other	other
## 168	GP	F	16	U	GT3	T	4	2	health	services
## 169	GP	F	16	U	GT3	T	2	2	other	other
## 170	GP	F	16	U	GT3	T	4	4	health	health
## 171	GP	M	16	U	GT3	T	3	4	other	other
## 172	GP	M	16	U	GT3	T	1	0	other	other
## 173	GP	M	17	U	LE3	T	4	4	teacher	other
## 174	GP	F	16	U	GT3	T	1	3	at_home	services
## 175	GP	F	16	U	LE3	T	3	3	other	other
## 176	GP	M	17	U	LE3	T	4	3	teacher	other
## 177	GP	F	16	U	GT3	T	2	2	services	other
## 178	GP	M	17	U	GT3	T	3	3	other	other
## 179	GP	M	16	R	GT3	T	4	2	teacher	services
## 180	GP	M	17	U	GT3	T	4	3	other	other
## 181	GP	M	16	U	GT3	T	4	3	teacher	other
## 182	GP	M	16	U	GT3	T	3	3	services	other
## 183	GP	F	17	U	GT3	T	2	4	services	services
## 184	GP	F	17	U	LE3	T	3	3	other	other
## 185	GP	F	16	U	GT3	T	3	2	other	other
## 186	GP	M	17	U	GT3	T	3	3	services	services
## 187	GP	M	16	U	GT3	T	1	2	services	services
## 188	GP	M	16	U	LE3	T	2	1	other	other
## 189	GP	F	17	U	GT3	A	3	3	health	other
## 190	GP	M	17	R	GT3	T	1	2	at_home	other
## 191	GP	F	16	U	GT3	T	2	3	services	services
## 192	GP	F	17	U	GT3	T	1	1	at_home	services
## 193	GP	M	17	U	GT3	T	1	2	at_home	services
## 194	GP	M	16	R	GT3	T	3	3	services	services
## 195	GP	M	16	U	GT3	T	2	3	other	other
## 196	GP	F	17	U	LE3	T	2	4	services	services
## 197	GP	M	17	U	GT3	T	4	4	services	teacher
## 198	GP	M	16	R	LE3	T	3	3	teacher	other
## 199	GP	F	17	U	GT3	T	4	4	services	teacher
## 200	GP	F	16	U	LE3	T	4	4	teacher	teacher
## 201	GP	F	16	U	GT3	T	4	3	health	other
## 202	GP	F	16	U	GT3	T	2	3	other	other
## 203	GP	F	17	U	GT3	T	1	1	other	other
## 204	GP	F	17	R	GT3	T	2	2	other	other
## 205	GP	F	16	R	GT3	T	2	2	services	services
## 206	GP	F	17	U	GT3	T	3	4	at_home	services
## 207	GP	F	16	U	GT3	A	3	1	services	other
## 208	GP	F	16	U	GT3	T	4	3	teacher	other
## 209	GP	F	16	U	GT3	T	1	1	at_home	other
## 210	GP	F	17	R	GT3	T	4	3	teacher	other
## 211	GP	F	19	U	GT3	T	3	3	other	other
## 212	GP	M	17	U	LE3	T	4	4	services	other
## 213	GP	F	16	U	GT3	A	2	2	other	other
## 214	GP	M	18	U	GT3	T	2	2	services	other
## 215	GP	F	17	R	LE3	T	4	4	services	other
## 216	GP	F	17	U	LE3	T	3	2	other	other

## 217	GP	F	17	U	GT3	T	4	3	other	other
## 218	GP	M	18	U	LE3	T	3	3	services	health
## 219	GP	F	17	U	GT3	T	2	3	at_home	other
## 220	GP	F	17	U	GT3	T	2	2	at_home	at_home
## 221	GP	F	17	R	GT3	T	2	1	at_home	services
## 222	GP	F	17	U	GT3	T	1	1	at_home	other
## 223	GP	F	16	U	GT3	T	2	3	services	teacher
## 224	GP	M	18	U	GT3	T	2	2	other	other
## 225	GP	F	16	U	GT3	T	4	4	teacher	services
## 226	GP	F	18	R	GT3	T	3	1	other	other
## 227	GP	F	17	U	GT3	T	3	2	other	other
## 228	GP	M	17	U	LE3	T	2	3	services	services
## 229	GP	M	18	U	LE3	T	2	1	at_home	other
## 230	GP	F	17	U	GT3	A	2	1	other	other
## 231	GP	F	17	U	LE3	T	4	3	health	other
## 232	GP	M	17	R	GT3	T	2	2	other	other
## 233	GP	M	17	U	GT3	T	4	4	teacher	teacher
## 234	GP	M	16	U	GT3	T	4	4	health	other
## 235	GP	M	16	U	LE3	T	1	1	other	other
## 236	GP	M	16	U	GT3	T	3	2	at_home	other
## 237	GP	M	17	U	LE3	T	2	2	other	other
## 238	GP	F	16	U	GT3	T	2	1	other	other
## 239	GP	F	17	R	GT3	T	2	1	at_home	services
## 240	GP	M	18	U	GT3	T	2	2	other	services
## 241	GP	M	17	U	LE3	T	4	3	health	other
## 242	GP	M	17	R	LE3	A	4	4	teacher	other
## 243	GP	M	16	U	LE3	T	4	3	teacher	other
## 244	GP	M	16	U	GT3	T	4	4	services	services
## 245	GP	F	18	U	GT3	T	2	1	other	other
## 246	GP	M	16	U	GT3	T	2	1	other	other
## 247	GP	M	17	U	GT3	T	2	3	other	other
## 248	GP	M	22	U	GT3	T	3	1	services	services
## 249	GP	M	18	R	LE3	T	3	3	other	services
## 250	GP	M	16	U	GT3	T	0	2	other	other
## 251	GP	M	18	U	GT3	T	3	2	services	other
## 252	GP	M	16	U	GT3	T	3	3	at_home	other
## 253	GP	M	18	U	GT3	T	2	1	services	services
## 254	GP	M	16	R	GT3	T	2	1	other	other
## 255	GP	M	17	R	GT3	T	2	1	other	other
## 256	GP	M	17	U	LE3	T	1	1	health	other
## 257	GP	F	17	U	LE3	T	4	2	teacher	services
## 258	GP	M	19	U	LE3	A	4	3	services	at_home
## 259	GP	M	18	U	GT3	T	2	1	other	other
## 260	GP	F	17	U	LE3	T	2	2	services	services
## 261	GP	F	18	U	GT3	T	4	3	services	other
## 262	GP	M	18	U	GT3	T	4	3	teacher	other
## 263	GP	M	18	R	GT3	T	3	2	other	other
## 264	GP	F	17	U	GT3	T	3	3	other	other
## 265	GP	F	18	U	GT3	T	2	2	at_home	services
## 266	GP	M	18	R	LE3	A	3	4	other	other
## 267	GP	M	17	U	GT3	T	3	1	services	other
## 268	GP	F	18	R	GT3	T	4	4	teacher	other
## 269	GP	M	18	U	GT3	T	4	2	health	other
## 270	GP	F	18	R	GT3	T	2	1	other	other

## 271	GP	F	19	U	GT3	T	3	3	other	services
## 272	GP	F	18	U	GT3	T	2	3	other	services
## 273	GP	F	18	U	LE3	T	1	1	other	other
## 274	GP	M	17	R	GT3	T	1	2	at_home	at_home
## 275	GP	F	17	U	GT3	T	2	4	at_home	health
## 276	GP	F	17	U	LE3	T	2	2	services	other
## 277	GP	F	18	R	GT3	A	3	2	other	services
## 278	GP	M	18	U	GT3	T	4	4	teacher	services
## 279	GP	F	18	U	GT3	T	4	4	health	health
## 280	GP	M	18	U	LE3	T	4	3	teacher	services
## 281	GP	M	17	U	LE3	A	4	1	services	other
## 282	GP	M	17	U	LE3	A	3	2	teacher	services
## 283	GP	F	18	R	LE3	T	1	1	at_home	other
## 284	GP	F	18	U	GT3	T	1	1	other	other
## 285	GP	F	17	U	GT3	T	2	2	other	other
## 286	GP	M	17	U	GT3	T	1	1	other	other
## 287	GP	F	18	U	GT3	T	2	2	at_home	at_home
## 288	GP	F	17	U	GT3	T	1	1	services	teacher
## 289	GP	M	18	U	GT3	T	2	1	services	services
## 290	GP	M	18	U	LE3	A	4	4	teacher	teacher
## 291	GP	M	18	U	GT3	T	4	2	teacher	other
## 292	GP	F	17	U	GT3	T	4	3	health	services
## 293	GP	F	18	U	LE3	T	2	1	services	at_home
## 294	GP	F	17	R	LE3	T	3	1	services	other
## 295	GP	M	18	R	LE3	T	3	2	services	other
## 296	GP	M	17	U	GT3	T	3	3	health	other
## 297	GP	F	19	U	GT3	T	4	4	health	other
## 298	GP	F	18	U	LE3	T	4	3	other	other
## 299	GP	F	18	U	GT3	T	4	3	other	other
## 300	GP	M	18	U	LE3	T	4	4	teacher	teacher
## 301	GP	F	18	U	LE3	A	4	4	health	other
## 302	GP	M	17	U	LE3	T	4	4	other	teacher
## 303	GP	F	17	U	GT3	T	4	2	other	other
## 304	GP	F	17	U	GT3	T	3	2	health	health
## 305	GP	M	19	U	GT3	T	3	3	other	other
## 306	GP	F	18	U	GT3	T	2	4	services	at_home
## 307	GP	M	20	U	GT3	A	3	2	services	other
## 308	GP	M	19	U	GT3	T	4	4	teacher	services
## 309	GP	M	19	R	GT3	T	3	3	other	services
## 310	GP	F	19	U	LE3	T	1	1	at_home	other
## 311	GP	F	19	U	LE3	T	1	2	services	services
## 312	GP	F	19	U	GT3	T	2	1	at_home	other
## 313	GP	M	19	U	GT3	T	1	2	other	services
## 314	GP	F	19	U	LE3	T	3	2	services	other
## 315	GP	F	19	U	GT3	T	1	1	at_home	health
## 316	GP	F	19	R	GT3	T	2	3	other	other
## 317	GP	F	18	U	GT3	T	2	1	services	other
## 318	GP	F	18	U	GT3	T	4	3	other	other
## 319	GP	F	17	R	GT3	T	3	4	at_home	services
## 320	GP	F	18	U	GT3	T	4	4	teacher	other
## 321	GP	F	17	U	GT3	A	4	3	services	services
## 322	GP	F	17	U	GT3	T	2	2	other	other
## 323	GP	F	17	R	LE3	T	2	2	services	services
## 324	GP	F	17	U	GT3	T	3	1	services	services

## 325	GP	F	17	U	LE3	T	0	2	at_home	at_home
## 326	GP	M	18	U	GT3	T	4	4	other	other
## 327	GP	M	17	U	GT3	T	3	3	other	services
## 328	GP	M	17	R	GT3	T	2	2	services	other
## 329	GP	F	17	U	GT3	T	4	4	teacher	services
## 330	GP	F	17	U	GT3	T	4	4	teacher	teacher
## 331	GP	M	18	U	LE3	T	2	2	other	other
## 332	GP	F	17	R	GT3	T	2	4	at_home	other
## 333	GP	F	18	U	GT3	T	3	3	services	services
## 334	GP	F	18	U	LE3	T	2	2	other	other
## 335	GP	F	18	R	GT3	T	2	2	at_home	other
## 336	GP	F	17	U	GT3	T	3	4	services	other
## 337	GP	F	19	R	GT3	A	3	1	services	at_home
## 338	GP	F	17	U	GT3	T	3	2	other	other
## 339	GP	F	18	U	LE3	T	3	3	services	services
## 340	GP	F	17	R	GT3	A	3	2	other	other
## 341	GP	F	19	U	GT3	T	2	1	services	services
## 342	GP	M	18	U	GT3	T	4	4	teacher	services
## 343	GP	M	18	U	LE3	T	3	4	services	other
## 344	GP	F	17	U	GT3	A	2	2	at_home	at_home
## 345	GP	F	18	U	GT3	T	2	3	at_home	other
## 346	GP	F	18	U	GT3	T	3	2	other	services
## 347	GP	M	18	R	GT3	T	4	3	teacher	services
## 348	GP	M	18	U	GT3	T	4	3	teacher	other
## 349	GP	F	17	U	GT3	T	4	3	health	other
## 350	MS	M	18	R	GT3	T	3	2	other	other
## 351	MS	M	19	R	GT3	T	1	1	other	services
## 352	MS	M	17	U	GT3	T	3	3	health	other
## 353	MS	M	18	U	LE3	T	1	3	at_home	services
## 354	MS	M	19	R	GT3	T	1	1	other	other
## 355	MS	M	17	R	GT3	T	4	3	services	other
## 356	MS	F	18	U	GT3	T	3	3	services	services
## 357	MS	F	17	R	GT3	T	4	4	teacher	services
## 358	MS	F	17	U	LE3	A	3	2	services	other
## 359	MS	M	18	U	LE3	T	1	1	other	services
## 360	MS	F	18	U	LE3	T	1	1	at_home	services
## 361	MS	F	18	R	LE3	A	1	4	at_home	other
## 362	MS	M	18	R	LE3	T	1	1	at_home	other
## 363	MS	F	18	U	GT3	T	3	3	services	services
## 364	MS	F	17	U	LE3	T	4	4	at_home	at_home
## 365	MS	F	17	R	GT3	T	1	2	other	services
## 366	MS	M	18	R	GT3	T	1	3	at_home	other
## 367	MS	M	18	U	LE3	T	4	4	teacher	services
## 368	MS	F	17	R	GT3	T	1	1	other	services
## 369	MS	F	18	U	GT3	T	2	3	at_home	services
## 370	MS	F	18	R	GT3	T	4	4	other	teacher
## 371	MS	F	19	U	LE3	T	3	2	services	services
## 372	MS	M	18	R	LE3	T	1	2	at_home	services
## 373	MS	F	17	U	GT3	T	2	2	other	at_home
## 374	MS	F	17	R	GT3	T	1	2	other	other
## 375	MS	F	18	R	LE3	T	4	4	other	other
## 376	MS	F	18	R	GT3	T	1	1	other	other
## 377	MS	F	20	U	GT3	T	4	2	health	other
## 378	MS	F	18	R	LE3	T	4	4	teacher	services

## 379	MS	F	18	U	GT3	T	3	3	other	other
## 380	MS	F	17	R	GT3	T	3	1	at_home	other
## 381	MS	M	18	U	GT3	T	4	4	teacher	teacher
## 382	MS	M	18	R	GT3	T	2	1	other	other
## 383	MS	M	17	U	GT3	T	2	3	other	services
## 384	MS	M	19	R	GT3	T	1	1	other	services
## 385	MS	M	18	R	GT3	T	4	2	other	other
## 386	MS	F	18	R	GT3	T	2	2	at_home	other
## 387	MS	F	18	R	GT3	T	4	4	teacher	at_home
## 388	MS	F	19	R	GT3	T	2	3	services	other
## 389	MS	F	18	U	LE3	T	3	1	teacher	services
## 390	MS	F	18	U	GT3	T	1	1	other	other
## 391	MS	M	20	U	LE3	A	2	2	services	services
## 392	MS	M	17	U	LE3	T	3	1	services	services
## 393	MS	M	21	R	GT3	T	1	1	other	other
## 394	MS	M	18	R	LE3	T	3	2	services	other
## 395	MS	M	19	U	LE3	T	1	1	other	at_home
##	reason	guardian	traveltime	studytime	failures	schoolsup	famsup			
## 1	course	mother	2	2	0	yes	no			
## 2	course	father	1	2	0	no	yes			
## 3	other	mother	1	2	3	yes	no			
## 4	home	mother	1	3	0	no	yes			
## 5	home	father	1	2	0	no	yes			
## 6	reputation	mother	1	2	0	no	yes			
## 7	home	mother	1	2	0	no	no			
## 8	home	mother	2	2	0	yes	yes			
## 9	home	mother	1	2	0	no	yes			
## 10	home	mother	1	2	0	no	yes			
## 11	reputation	mother	1	2	0	no	yes			
## 12	reputation	father	3	3	0	no	yes			
## 13	course	father	1	1	0	no	yes			
## 14	course	mother	2	2	0	no	yes			
## 15	home	other	1	3	0	no	yes			
## 16	home	mother	1	1	0	no	yes			
## 17	reputation	mother	1	3	0	no	yes			
## 18	reputation	mother	3	2	0	yes	yes			
## 19	course	mother	1	1	3	no	yes			
## 20	home	father	1	1	0	no	no			
## 21	reputation	mother	1	2	0	no	no			
## 22	other	father	1	1	0	no	yes			
## 23	course	mother	1	2	0	no	no			
## 24	reputation	mother	2	2	0	no	yes			
## 25	course	mother	1	3	0	yes	yes			
## 26	home	mother	1	1	2	no	yes			
## 27	home	mother	1	1	0	no	yes			
## 28	other	mother	1	1	0	no	no			
## 29	home	mother	1	2	0	yes	yes			
## 30	home	mother	1	2	0	no	yes			
## 31	home	mother	1	2	0	no	yes			
## 32	reputation	mother	2	2	0	no	yes			
## 33	course	mother	1	2	0	no	yes			
## 34	course	mother	1	2	0	no	no			
## 35	home	mother	1	1	0	no	yes			
## 36	other	father	2	1	0	no	yes			

## 37	home	mother	1	3	0	no	yes
## 38	reputation	mother	2	3	0	no	yes
## 39	course	mother	1	3	0	yes	yes
## 40	reputation	mother	1	1	0	yes	yes
## 41	home	mother	2	2	1	no	yes
## 42	home	other	1	1	0	no	yes
## 43	course	father	1	2	0	no	yes
## 44	course	father	1	1	0	yes	yes
## 45	course	father	2	2	1	yes	no
## 46	course	mother	1	2	0	yes	yes
## 47	home	mother	1	2	0	no	yes
## 48	reputation	mother	1	4	0	no	no
## 49	home	mother	1	2	0	no	yes
## 50	other	father	1	2	1	yes	yes
## 51	course	mother	3	2	0	no	yes
## 52	other	mother	1	2	0	no	yes
## 53	other	father	2	1	1	no	no
## 54	course	mother	1	1	0	yes	yes
## 55	other	mother	1	1	0	no	no
## 56	other	mother	1	2	0	no	no
## 57	reputation	mother	1	2	0	no	yes
## 58	reputation	mother	1	2	0	no	yes
## 59	home	father	1	2	0	yes	yes
## 60	course	mother	1	2	0	no	yes
## 61	other	mother	1	2	0	no	yes
## 62	course	father	4	1	0	yes	yes
## 63	reputation	father	1	2	0	yes	no
## 64	home	mother	1	3	0	yes	yes
## 65	reputation	father	1	2	0	yes	no
## 66	course	mother	3	2	0	no	yes
## 67	reputation	mother	1	4	0	no	yes
## 68	course	mother	1	4	0	yes	yes
## 69	reputation	mother	2	2	0	yes	yes
## 70	reputation	father	2	4	0	no	yes
## 71	reputation	father	2	4	0	no	yes
## 72	course	mother	1	4	0	no	no
## 73	reputation	mother	1	2	2	yes	yes
## 74	reputation	mother	1	1	0	no	no
## 75	home	mother	1	2	0	yes	yes
## 76	home	mother	1	2	0	no	yes
## 77	course	mother	2	4	0	no	no
## 78	reputation	mother	1	4	0	no	no
## 79	home	mother	2	1	3	yes	yes
## 80	course	mother	1	2	0	no	yes
## 81	course	father	1	1	0	yes	yes
## 82	home	mother	1	3	0	yes	no
## 83	reputation	mother	1	2	0	no	yes
## 84	home	mother	2	2	0	no	no
## 85	home	father	1	2	0	no	yes
## 86	reputation	father	2	2	2	no	no
## 87	course	mother	1	2	0	no	yes
## 88	reputation	mother	1	3	0	no	yes
## 89	reputation	father	2	2	1	no	no
## 90	reputation	mother	1	2	0	no	yes

## 91	home	mother	1	3	0	no	yes
## 92	reputation	mother	1	1	0	no	no
## 93	home	father	1	2	0	yes	yes
## 94	home	mother	2	2	0	no	yes
## 95	reputation	mother	1	4	0	no	yes
## 96	home	mother	2	4	1	yes	yes
## 97	reputation	mother	2	1	0	yes	yes
## 98	course	mother	1	2	0	no	yes
## 99	reputation	mother	1	1	0	no	no
## 100	course	mother	1	3	0	yes	yes
## 101	other	mother	1	1	0	yes	yes
## 102	other	father	1	3	0	no	yes
## 103	course	mother	1	1	0	no	yes
## 104	home	mother	2	2	0	yes	yes
## 105	course	mother	1	2	0	no	yes
## 106	reputation	father	1	4	0	yes	no
## 107	course	mother	1	4	0	yes	yes
## 108	home	father	1	3	0	no	yes
## 109	home	father	4	4	0	no	yes
## 110	other	mother	1	3	0	no	yes
## 111	course	mother	1	1	0	no	no
## 112	reputation	father	1	3	1	yes	yes
## 113	home	mother	1	2	1	yes	no
## 114	course	mother	1	1	0	no	no
## 115	reputation	mother	1	2	0	no	no
## 116	course	father	1	2	0	no	yes
## 117	reputation	father	2	2	0	no	yes
## 118	home	father	2	1	0	no	no
## 119	course	father	3	2	1	no	yes
## 120	reputation	father	1	1	0	no	no
## 121	course	mother	1	2	0	no	no
## 122	home	father	1	4	0	no	yes
## 123	course	father	2	2	0	no	yes
## 124	course	mother	1	1	0	no	yes
## 125	home	mother	1	2	0	no	no
## 126	home	father	1	1	0	yes	no
## 127	home	mother	1	2	0	yes	no
## 128	course	other	1	2	3	no	yes
## 129	reputation	mother	1	1	2	no	yes
## 130	course	mother	1	1	0	no	no
## 131	course	father	2	3	2	no	yes
## 132	course	mother	3	1	0	no	yes
## 133	course	father	1	1	0	no	yes
## 134	course	father	1	1	0	no	no
## 135	course	mother	4	2	0	no	yes
## 136	course	mother	1	3	0	no	yes
## 137	course	mother	3	2	0	no	no
## 138	course	other	2	1	2	no	yes
## 139	course	mother	1	2	1	no	no
## 140	course	mother	2	1	0	no	no
## 141	course	father	2	4	0	yes	yes
## 142	reputation	father	2	1	2	no	yes
## 143	course	mother	1	3	0	no	yes
## 144	course	mother	1	1	0	no	no

## 145	home	mother	1	1	3	no	yes
## 146	course	father	1	2	0	no	yes
## 147	home	father	1	2	3	no	yes
## 148	course	mother	1	2	0	no	yes
## 149	course	mother	1	1	0	no	yes
## 150	course	mother	4	1	3	no	no
## 151	course	mother	1	1	3	no	no
## 152	course	mother	1	1	1	no	no
## 153	reputation	other	2	3	2	no	yes
## 154	home	mother	1	1	3	no	yes
## 155	course	mother	1	1	0	yes	yes
## 156	course	mother	1	2	0	yes	no
## 157	reputation	mother	1	1	0	no	no
## 158	course	mother	3	1	3	no	yes
## 159	course	mother	3	1	0	no	no
## 160	course	father	1	2	1	no	yes
## 161	course	mother	2	1	2	no	no
## 162	course	mother	2	2	2	yes	yes
## 163	course	mother	2	1	1	no	no
## 164	course	father	1	1	0	no	no
## 165	course	mother	4	2	3	no	no
## 166	course	mother	2	1	1	no	yes
## 167	course	father	1	2	0	no	no
## 168	home	father	1	2	0	no	no
## 169	home	mother	1	2	0	no	yes
## 170	reputation	mother	1	2	0	no	yes
## 171	course	father	3	1	2	no	yes
## 172	reputation	mother	2	2	0	no	yes
## 173	reputation	mother	1	2	0	no	yes
## 174	home	mother	1	2	3	no	no
## 175	reputation	mother	2	2	0	no	yes
## 176	course	mother	2	2	0	no	no
## 177	reputation	mother	2	2	0	no	no
## 178	reputation	father	1	2	0	no	no
## 179	other	mother	1	1	0	no	yes
## 180	course	mother	1	2	0	no	yes
## 181	home	mother	1	2	0	no	yes
## 182	home	mother	1	2	0	no	no
## 183	reputation	father	1	2	0	no	yes
## 184	reputation	mother	1	2	0	no	yes
## 185	reputation	mother	1	2	0	no	yes
## 186	other	mother	1	2	0	no	yes
## 187	other	mother	1	1	0	no	yes
## 188	course	mother	1	2	0	no	no
## 189	reputation	mother	1	2	0	no	yes
## 190	home	mother	1	2	0	no	no
## 191	course	mother	1	2	0	no	no
## 192	course	mother	1	2	0	no	no
## 193	other	other	2	2	0	no	no
## 194	reputation	mother	1	1	0	no	yes
## 195	home	father	2	1	0	no	no
## 196	course	father	1	2	0	no	no
## 197	home	mother	1	1	0	no	no
## 198	home	father	3	1	0	no	yes

## 199	home	mother	2	1	1	no	yes
## 200	reputation	mother	1	2	0	no	yes
## 201	home	mother	1	2	0	no	yes
## 202	reputation	mother	1	2	0	yes	yes
## 203	course	mother	1	2	0	no	yes
## 204	reputation	mother	1	1	0	no	yes
## 205	reputation	mother	2	4	0	no	yes
## 206	home	mother	1	3	1	no	yes
## 207	course	mother	1	2	3	no	yes
## 208	other	mother	1	2	0	no	no
## 209	home	mother	2	1	0	no	yes
## 210	reputation	mother	2	3	0	no	yes
## 211	reputation	other	1	4	0	no	yes
## 212	home	mother	1	2	0	no	yes
## 213	reputation	mother	1	2	0	yes	yes
## 214	home	mother	1	2	1	no	yes
## 215	other	mother	1	1	0	no	yes
## 216	reputation	mother	2	2	0	no	no
## 217	reputation	mother	1	2	2	no	no
## 218	home	father	1	2	1	no	yes
## 219	home	father	2	1	0	no	yes
## 220	course	mother	1	3	0	no	yes
## 221	reputation	mother	2	2	0	no	yes
## 222	reputation	mother	1	3	1	no	yes
## 223	other	mother	1	2	0	yes	no
## 224	home	mother	2	2	0	no	yes
## 225	home	mother	1	3	0	no	yes
## 226	reputation	mother	1	2	1	no	no
## 227	course	mother	1	2	0	no	no
## 228	reputation	father	1	2	0	no	yes
## 229	course	mother	4	2	0	yes	yes
## 230	course	mother	2	3	0	no	no
## 231	reputation	father	1	2	0	no	no
## 232	course	father	2	2	0	no	yes
## 233	reputation	mother	1	2	0	yes	yes
## 234	reputation	father	1	2	0	no	yes
## 235	home	mother	2	2	0	no	yes
## 236	reputation	mother	2	3	0	no	no
## 237	home	father	1	2	0	no	no
## 238	home	mother	1	1	0	no	no
## 239	course	mother	3	2	0	no	no
## 240	reputation	father	1	2	1	no	no
## 241	course	mother	2	2	0	no	no
## 242	course	mother	2	2	0	no	yes
## 243	course	mother	1	1	0	no	no
## 244	course	mother	1	1	0	no	no
## 245	course	other	2	3	0	no	yes
## 246	course	mother	3	1	0	no	no
## 247	course	father	2	1	0	no	no
## 248	other	mother	1	1	3	no	no
## 249	course	mother	1	2	1	no	yes
## 250	other	mother	1	1	0	no	no
## 251	course	mother	2	1	1	no	no
## 252	reputation	other	3	2	0	yes	yes

## 253	other	mother	1	1	1	no	no
## 254	course	mother	2	1	0	no	no
## 255	course	mother	1	1	0	no	no
## 256	course	mother	2	1	1	no	yes
## 257	reputation	mother	1	4	0	no	yes
## 258	reputation	mother	1	2	0	no	yes
## 259	home	mother	1	2	0	no	no
## 260	course	father	1	4	0	no	no
## 261	home	father	1	2	0	no	yes
## 262	course	mother	1	2	0	no	yes
## 263	course	mother	1	3	0	no	no
## 264	home	mother	1	3	0	no	no
## 265	home	mother	1	3	0	no	yes
## 266	reputation	mother	2	2	0	no	yes
## 267	other	mother	1	2	0	no	no
## 268	reputation	mother	2	2	0	no	no
## 269	reputation	father	1	2	0	no	yes
## 270	reputation	mother	2	2	0	no	yes
## 271	home	other	1	2	2	no	yes
## 272	reputation	father	1	4	0	no	yes
## 273	home	mother	2	2	0	no	yes
## 274	home	mother	1	2	0	no	yes
## 275	reputation	mother	2	2	0	no	yes
## 276	course	mother	2	2	0	yes	yes
## 277	home	mother	2	2	0	no	no
## 278	home	mother	2	1	0	no	no
## 279	reputation	father	1	2	1	yes	yes
## 280	course	mother	2	1	0	no	no
## 281	home	mother	2	1	0	no	no
## 282	home	mother	1	1	1	no	no
## 283	reputation	mother	2	4	0	no	yes
## 284	home	mother	2	2	0	yes	no
## 285	course	mother	1	2	0	no	yes
## 286	reputation	father	1	2	0	no	no
## 287	other	mother	1	3	0	no	yes
## 288	reputation	mother	1	3	0	no	yes
## 289	reputation	mother	1	3	0	no	no
## 290	reputation	mother	1	2	0	no	yes
## 291	home	mother	1	2	0	no	yes
## 292	reputation	mother	1	3	0	no	yes
## 293	reputation	mother	1	2	1	no	no
## 294	reputation	mother	2	4	0	no	yes
## 295	reputation	mother	2	3	0	no	yes
## 296	home	mother	1	1	0	no	yes
## 297	reputation	other	2	2	0	no	yes
## 298	home	other	2	2	0	no	yes
## 299	reputation	father	1	4	0	no	yes
## 300	home	mother	1	1	0	no	yes
## 301	home	mother	1	2	0	no	yes
## 302	home	father	2	1	0	no	no
## 303	reputation	mother	2	3	0	no	yes
## 304	reputation	father	1	4	0	no	yes
## 305	home	other	1	2	1	no	yes
## 306	reputation	other	1	2	1	no	yes

## 307	course	other	1	1	0	no	no
## 308	reputation	other	2	1	1	no	yes
## 309	reputation	father	1	2	1	no	no
## 310	reputation	other	1	2	1	yes	yes
## 311	home	other	1	2	1	no	no
## 312	other	other	3	2	0	no	yes
## 313	course	other	1	2	1	no	no
## 314	reputation	other	2	2	1	no	yes
## 315	home	other	1	3	2	no	no
## 316	reputation	other	1	3	1	no	no
## 317	course	mother	2	2	0	no	yes
## 318	course	mother	1	3	0	no	yes
## 319	course	father	1	3	0	no	yes
## 320	course	mother	1	2	0	no	yes
## 321	course	mother	1	2	0	no	yes
## 322	course	mother	1	2	0	no	yes
## 323	course	mother	1	3	0	no	yes
## 324	course	father	1	3	0	no	yes
## 325	home	father	2	3	0	no	no
## 326	course	mother	1	3	0	no	no
## 327	reputation	mother	1	1	0	no	no
## 328	course	mother	4	1	0	no	yes
## 329	course	mother	1	3	0	no	yes
## 330	course	mother	2	3	0	no	yes
## 331	course	mother	1	4	0	no	yes
## 332	course	father	1	3	0	no	yes
## 333	home	mother	1	2	0	no	no
## 334	home	other	1	2	0	no	no
## 335	course	mother	2	4	0	no	no
## 336	course	mother	1	3	0	no	no
## 337	home	other	1	3	1	no	no
## 338	home	mother	1	2	0	no	yes
## 339	home	mother	1	4	0	no	yes
## 340	home	mother	1	2	0	no	yes
## 341	home	other	1	3	1	no	no
## 342	home	father	1	2	1	no	yes
## 343	home	mother	1	2	0	no	no
## 344	home	father	1	2	1	no	yes
## 345	course	mother	1	3	0	no	yes
## 346	other	mother	1	3	0	no	no
## 347	course	mother	1	3	0	no	no
## 348	course	mother	1	3	0	no	yes
## 349	reputation	mother	1	3	0	no	yes
## 350	course	mother	2	1	1	no	yes
## 351	home	other	3	2	3	no	no
## 352	course	mother	2	2	0	no	yes
## 353	course	mother	1	1	1	no	no
## 354	home	other	3	1	1	no	yes
## 355	home	mother	2	2	0	no	yes
## 356	course	father	1	2	0	no	yes
## 357	other	father	2	2	0	no	yes
## 358	reputation	mother	2	2	0	no	no
## 359	home	father	2	1	0	no	no
## 360	course	father	2	3	0	no	no

## 361	course	mother	3	2	0	no	no		
## 362	other	mother	2	2	1	no	no		
## 363	other	mother	2	2	0	no	yes		
## 364	course	mother	1	2	0	no	yes		
## 365	course	father	2	2	0	no	no		
## 366	course	mother	2	2	0	no	yes		
## 367	other	mother	2	3	0	no	no		
## 368	reputation	mother	3	1	1	no	yes		
## 369	course	father	2	1	0	no	yes		
## 370	other	father	3	2	0	no	yes		
## 371	home	other	2	2	2	no	no		
## 372	other	father	3	1	0	no	yes		
## 373	home	mother	1	3	0	no	no		
## 374	course	mother	1	1	0	no	no		
## 375	reputation	mother	2	3	0	no	no		
## 376	home	mother	4	3	0	no	no		
## 377	course	other	2	3	2	no	yes		
## 378	course	mother	1	2	0	no	no		
## 379	home	mother	1	2	0	no	no		
## 380	reputation	mother	1	2	0	no	yes		
## 381	home	father	1	2	0	no	no		
## 382	other	mother	2	1	0	no	no		
## 383	home	father	2	2	0	no	no		
## 384	other	mother	2	1	1	no	no		
## 385	home	father	2	1	1	no	no		
## 386	other	mother	2	3	0	no	no		
## 387	reputation	mother	3	1	0	no	yes		
## 388	course	mother	1	3	1	no	no		
## 389	course	mother	1	2	0	no	yes		
## 390	course	mother	2	2	1	no	no		
## 391	course	other	1	2	2	no	yes		
## 392	course	mother	2	1	0	no	no		
## 393	course	other	1	1	3	no	no		
## 394	course	mother	3	1	0	no	no		
## 395	course	father	1	1	0	no	no		
##	paid activities	nursery	higher	internet	romantic	famrel	freetime	goout	
## 1	no	no	yes	yes	no	no	4	3	4
## 2	no	no	no	yes	yes	no	5	3	3
## 3	yes	no	yes	yes	yes	no	4	3	2
## 4	yes	yes	yes	yes	yes	yes	3	2	2
## 5	yes	no	yes	yes	no	no	4	3	2
## 6	yes	yes	yes	yes	yes	no	5	4	2
## 7	no	no	yes	yes	yes	no	4	4	4
## 8	no	no	yes	yes	no	no	4	1	4
## 9	yes	no	yes	yes	yes	no	4	2	2
## 10	yes	yes	yes	yes	yes	no	5	5	1
## 11	yes	no	yes	yes	yes	no	3	3	3
## 12	no	yes	yes	yes	yes	no	5	2	2
## 13	yes	yes	yes	yes	yes	no	4	3	3
## 14	yes	no	yes	yes	yes	no	5	4	3
## 15	no	no	yes	yes	yes	yes	4	5	2
## 16	no	no	yes	yes	yes	no	4	4	4
## 17	yes	yes	yes	yes	yes	no	3	2	3
## 18	no	yes	yes	yes	no	no	5	3	2

## 19	no	yes	yes	yes	yes	no	5	5	5
## 20	yes	yes	yes	yes	yes	no	3	1	3
## 21	no	no	yes	yes	yes	no	4	4	1
## 22	yes	no	yes	yes	yes	no	5	4	2
## 23	no	yes	yes	yes	yes	no	4	5	1
## 24	no	yes	yes	yes	yes	no	5	4	4
## 25	yes	yes	yes	yes	yes	no	4	3	2
## 26	yes	no	no	yes	yes	no	1	2	2
## 27	yes	no	yes	yes	yes	no	4	2	2
## 28	yes	no	yes	yes	yes	no	2	2	4
## 29	no	yes	yes	yes	yes	no	5	3	3
## 30	yes	yes	yes	yes	yes	yes	4	4	5
## 31	yes	no	no	yes	yes	no	5	4	2
## 32	no	yes	yes	yes	yes	no	4	3	1
## 33	no	yes	yes	yes	yes	yes	4	5	2
## 34	no	yes	no	yes	yes	no	5	3	2
## 35	yes	no	no	yes	yes	no	5	4	3
## 36	no	yes	yes	yes	no	no	3	5	1
## 37	no	yes	yes	yes	yes	no	5	4	3
## 38	no	yes	yes	yes	yes	yes	2	4	3
## 39	yes	yes	yes	yes	yes	no	4	3	2
## 40	yes	yes	yes	yes	no	no	4	3	1
## 41	no	yes	no	yes	yes	yes	3	3	3
## 42	no	no	no	yes	yes	yes	5	4	3
## 43	no	yes	yes	yes	yes	no	4	3	3
## 44	no	no	yes	yes	yes	no	5	4	1
## 45	no	yes	yes	yes	yes	no	4	3	3
## 46	yes	yes	yes	yes	yes	yes	5	2	2
## 47	no	no	yes	yes	yes	no	2	3	5
## 48	no	yes	yes	yes	yes	no	4	2	2
## 49	yes	no	yes	yes	no	no	4	3	3
## 50	no	yes	no	yes	yes	no	4	4	4
## 51	yes	no	yes	yes	yes	no	4	3	3
## 52	yes	no	yes	yes	yes	no	4	3	3
## 53	no	no	yes	yes	no	no	5	5	5
## 54	yes	no	yes	yes	yes	no	3	3	4
## 55	yes	no	yes	yes	yes	no	5	3	4
## 56	yes	yes	yes	yes	yes	yes	5	3	4
## 57	yes	yes	yes	yes	yes	no	4	3	2
## 58	no	yes	yes	yes	no	no	3	2	2
## 59	no	yes	yes	yes	yes	no	4	3	2
## 60	no	no	yes	yes	yes	no	4	2	3
## 61	no	yes	yes	yes	no	no	2	4	4
## 62	no	yes	no	yes	yes	yes	5	5	5
## 63	no	yes	yes	yes	yes	no	4	4	3
## 64	yes	yes	yes	yes	yes	no	3	4	4
## 65	no	yes	yes	yes	yes	yes	4	4	4
## 66	no	yes	yes	yes	yes	no	5	4	3
## 67	no	yes	no	yes	yes	yes	1	3	3
## 68	yes	no	yes	yes	yes	no	4	3	3
## 69	yes	no	yes	yes	yes	no	4	1	3
## 70	no	no	no	yes	yes	no	4	4	2
## 71	yes	no	yes	yes	yes	no	4	3	2
## 72	no	no	yes	yes	yes	no	3	3	3

## 73	no	no	no	yes	yes	yes	3	3	4
## 74	no	yes	yes	yes	no	no	5	3	2
## 75	yes	yes	yes	yes	yes	no	4	3	3
## 76	yes	yes	yes	yes	yes	no	4	3	3
## 77	no	yes	yes	yes	yes	no	3	4	3
## 78	yes	no	yes	yes	yes	yes	5	2	3
## 79	no	yes	yes	no	yes	no	4	5	1
## 80	no	no	yes	yes	yes	no	2	4	3
## 81	yes	yes	no	yes	yes	yes	3	2	2
## 82	yes	no	no	yes	yes	no	5	3	2
## 83	yes	no	yes	yes	yes	no	4	4	4
## 84	yes	yes	yes	yes	yes	no	5	3	3
## 85	no	yes	no	yes	yes	no	4	3	2
## 86	yes	no	yes	yes	yes	yes	4	4	4
## 87	no	no	yes	yes	no	no	4	3	4
## 88	no	yes	yes	yes	yes	no	5	3	3
## 89	yes	yes	no	yes	yes	no	4	4	2
## 90	no	no	yes	yes	no	no	4	1	3
## 91	yes	no	yes	yes	yes	yes	4	3	3
## 92	yes	yes	yes	yes	yes	no	4	5	5
## 93	no	no	yes	yes	no	no	3	3	3
## 94	yes	yes	yes	yes	yes	no	5	3	3
## 95	no	yes	yes	yes	yes	no	4	3	4
## 96	yes	yes	yes	yes	yes	no	3	1	2
## 97	no	yes	no	yes	yes	no	3	3	3
## 98	yes	no	yes	yes	no	yes	4	3	5
## 99	no	yes	no	yes	yes	no	5	3	4
## 100	yes	no	yes	yes	yes	no	5	3	5
## 101	yes	yes	yes	yes	yes	no	4	5	5
## 102	no	yes	yes	yes	yes	yes	4	4	3
## 103	no	yes	no	yes	yes	no	5	3	3
## 104	yes	no	yes	yes	yes	no	4	3	5
## 105	yes	yes	yes	yes	yes	no	5	4	4
## 106	no	no	yes	yes	no	no	4	3	3
## 107	yes	no	yes	yes	yes	no	5	1	2
## 108	no	yes	yes	yes	yes	no	5	3	3
## 109	yes	yes	yes	yes	yes	yes	1	3	5
## 110	yes	yes	yes	yes	yes	yes	5	4	5
## 111	no	yes	yes	yes	yes	no	5	5	3
## 112	no	yes	yes	yes	yes	no	4	1	2
## 113	no	yes	yes	yes	yes	no	3	1	2
## 114	no	no	yes	yes	yes	no	3	5	2
## 115	no	yes	yes	yes	yes	yes	5	4	2
## 116	no	yes	yes	yes	yes	no	5	4	4
## 117	no	yes	yes	yes	no	no	4	4	3
## 118	no	yes	yes	yes	yes	no	5	4	2
## 119	no	yes	yes	yes	yes	no	5	2	4
## 120	no	no	yes	yes	yes	no	3	4	3
## 121	no	no	no	yes	yes	no	3	2	3
## 122	yes	yes	yes	yes	yes	no	5	5	4
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##	Dalc Walc health absences G1 G2 G3 PF								

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## 3	2	3	3	10	7	8	10	P
## 4	1	1	5	2	15	14	15	P
## 5	1	2	5	4	6	10	10	P
## 6	1	2	5	10	15	15	15	P
## 7	1	1	3	0	12	12	11	P
## 8	1	1	1	6	6	5	6	F
## 9	1	1	1	0	16	18	19	P
## 10	1	1	5	0	14	15	15	P
## 11	1	2	2	0	10	8	9	F
## 12	1	1	4	4	10	12	12	P
## 13	1	3	5	2	14	14	14	P
## 14	1	2	3	2	10	10	11	P
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## 20	1	3	5	4	8	10	10	P
## 21	1	1	1	0	13	14	15	P
## 22	1	1	5	0	12	15	15	P
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## 26	1	3	5	14	6	9	8	F
## 27	1	2	5	2	12	12	11	P
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## 37	1	1	4	2	15	16	18	P
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## 45	2	2	5	14	10	10	9	F
## 46	1	1	5	8	8	8	6	F
## 47	1	4	3	12	11	12	11	P
## 48	1	1	2	4	19	19	20	P
## 49	2	2	5	2	15	15	14	P
## 50	1	1	3	2	7	7	7	F
## 51	2	3	4	2	12	13	13	P
## 52	1	1	5	2	11	13	13	P
## 53	3	4	5	6	11	11	10	P
## 54	2	3	5	0	8	10	11	P

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## 64	2	4	4	2	10	9	9	F
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## 74	2	2	5	2	12	12	14	P
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## 87	1	2	2	4	8	7	6	F
## 88	1	3	1	4	13	14	14	P
## 89	1	1	3	12	11	10	10	P
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## 91	1	3	4	0	7	7	8	F
## 92	1	3	1	4	16	17	18	P
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## 101	5	5	4	14	7	7	5	F
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## 104	1	1	2	26	7	6	6	F
## 105	1	1	1	0	16	18	18	P
## 106	1	1	4	10	10	11	11	P
## 107	1	1	3	8	7	8	8	F
## 108	1	1	5	2	16	18	18	P

## 109	3	5	1	6 10 13 13	P
## 110	1	1	4	4 14 15 16	P
## 111	1	1	4	6 18 19 19	P
## 112	1	1	2	0 7 10 10	P
## 113	1	1	5	6 10 13 13	P
## 114	1	1	3	10 18 19 19	P
## 115	1	1	5	8 9 9 9	F
## 116	1	2	5	2 15 15 16	P
## 117	1	1	2	2 11 13 14	P
## 118	1	1	5	0 13 14 13	P
## 119	1	4	5	20 9 7 8	F
## 120	1	2	4	6 14 13 13	P
## 121	1	2	1	2 16 15 15	P
## 122	1	2	5	6 16 14 15	P
## 123	1	2	5	2 13 13 13	P
## 124	1	4	5	18 14 11 13	P
## 125	1	1	5	0 8 7 8	F
## 126	3	2	5	0 13 13 12	P
## 127	1	1	1	0 7 10 11	P
## 128	1	1	5	2 7 8 9	F
## 129	1	2	4	0 7 4 0	F
## 130	2	5	4	8 18 18 18	P
## 131	2	2	5	0 12 0 0	F
## 132	1	2	4	0 8 0 0	F
## 133	1	3	5	12 10 13 12	P
## 134	1	4	5	16 12 11 11	P
## 135	1	1	5	0 9 0 0	F
## 136	1	1	5	0 11 0 0	F
## 137	2	4	5	0 10 0 0	F
## 138	1	1	5	0 4 0 0	F
## 139	1	3	5	0 14 12 12	P
## 140	1	1	5	0 16 16 15	P
## 141	1	1	3	0 7 9 0	F
## 142	2	2	2	8 9 9 9	F
## 143	1	1	5	2 9 11 11	P
## 144	3	3	1	2 14 14 13	P
## 145	1	2	5	0 5 0 0	F
## 146	1	2	5	0 8 11 11	P
## 147	1	1	3	0 6 7 0	F
## 148	1	1	5	2 10 11 11	P
## 149	2	1	5	0 7 6 0	F
## 150	2	5	5	0 8 9 10	P
## 151	2	5	4	0 6 5 0	F
## 152	3	5	5	6 12 13 14	P
## 153	2	3	3	8 10 10 10	P
## 154	1	1	4	0 5 0 0	F
## 155	1	1	4	0 11 11 12	P
## 156	1	1	1	2 11 8 8	F
## 157	3	3	5	8 16 12 13	P
## 158	1	5	4	6 9 8 10	P
## 159	1	2	3	2 17 15 15	P
## 160	4	4	5	4 10 12 12	P
## 161	2	2	5	0 7 6 0	F
## 162	1	4	3	6 5 9 7	F

## 163	2	4	5	0	7	0	0	F
## 164	1	4	2	2	10	10	10	P
## 165	1	5	5	0	5	8	7	F
## 166	1	1	2	16	12	11	12	P
## 167	2	4	4	4	10	10	10	P
## 168	1	1	3	0	14	15	16	P
## 169	1	1	4	0	6	7	0	F
## 170	1	1	3	0	14	14	14	P
## 171	2	4	2	0	6	5	0	F
## 172	1	1	3	2	13	15	16	P
## 173	1	3	5	0	13	11	10	P
## 174	1	1	3	0	8	7	0	F
## 175	1	1	4	4	10	11	9	F
## 176	4	4	4	4	10	9	9	F
## 177	1	4	5	2	13	13	11	P
## 178	1	4	4	4	6	5	6	F
## 179	3	4	3	10	10	8	9	F
## 180	1	1	2	4	10	10	11	P
## 181	2	3	3	10	9	8	8	F
## 182	1	2	3	2	12	13	12	P
## 183	2	3	5	0	16	17	17	P
## 184	2	3	1	56	9	9	8	F
## 185	1	2	1	14	12	13	12	P
## 186	2	3	4	12	12	12	11	P
## 187	1	2	3	2	11	12	11	P
## 188	1	2	5	0	15	15	15	P
## 189	1	3	3	6	8	7	9	F
## 190	1	5	3	4	8	9	10	P
## 191	1	1	2	10	11	12	13	P
## 192	1	1	3	0	8	8	9	F
## 193	4	5	5	12	7	8	8	F
## 194	3	4	5	8	8	9	10	P
## 195	1	1	3	0	13	14	14	P
## 196	1	1	5	0	14	15	15	P
## 197	1	2	5	4	17	15	16	P
## 198	3	5	3	8	9	9	10	P
## 199	2	3	2	24	18	18	18	P
## 200	1	2	3	0	9	9	10	P
## 201	1	5	2	2	16	16	16	P
## 202	1	3	4	6	8	10	10	P
## 203	1	3	1	4	9	9	10	P
## 204	1	2	3	18	7	6	6	F
## 205	1	1	5	6	10	10	11	P
## 206	3	4	5	28	10	9	9	F
## 207	2	2	4	5	7	7	7	F
## 208	1	1	1	10	11	12	13	P
## 209	1	4	5	6	9	9	10	P
## 210	1	1	4	6	7	7	7	F
## 211	1	2	3	10	8	8	8	F
## 212	4	5	3	13	12	12	13	P
## 213	1	1	4	0	12	13	14	P
## 214	2	4	5	15	6	7	8	F
## 215	1	2	3	12	8	10	10	P
## 216	1	3	1	2	14	15	15	P

## 217	2	4	1	22	6	6	4	F
## 218	2	4	4	13	6	6	8	F
## 219	1	4	3	3	7	7	8	F
## 220	1	1	4	4	9	10	10	P
## 221	1	2	5	2	6	6	6	F
## 222	1	1	5	0	6	5	0	F
## 223	1	1	3	2	16	16	17	P
## 224	5	5	4	0	12	13	13	P
## 225	1	1	5	0	13	13	14	P
## 226	1	1	4	16	9	8	7	F
## 227	1	3	3	10	16	15	15	P
## 228	1	3	3	2	12	11	12	P
## 229	4	5	3	14	10	8	9	F
## 230	1	2	3	10	12	10	12	P
## 231	1	2	3	14	13	13	14	P
## 232	1	1	1	4	11	11	11	P
## 233	1	3	2	14	11	9	9	F
## 234	2	4	1	2	14	13	13	P
## 235	1	1	5	18	9	7	6	F
## 236	1	3	2	10	11	9	10	P
## 237	5	5	4	4	14	13	13	P
## 238	1	1	5	20	13	12	12	P
## 239	1	1	3	2	13	11	11	P
## 240	3	5	2	0	7	7	0	F
## 241	1	4	5	14	12	12	12	P
## 242	2	3	4	2	10	11	12	P
## 243	1	1	3	0	6	0	0	F
## 244	1	2	5	0	13	12	12	P
## 245	1	1	3	0	7	0	0	F
## 246	1	1	4	6	18	18	18	P
## 247	1	1	2	4	12	12	13	P
## 248	5	5	1	16	6	8	8	F
## 249	1	3	5	8	3	5	5	F
## 250	2	4	5	0	13	15	15	P
## 251	2	4	5	0	6	8	8	F
## 252	1	3	2	6	7	10	10	P
## 253	2	5	5	4	6	9	8	F
## 254	1	3	3	0	8	9	8	F
## 255	2	4	5	0	8	12	12	P
## 256	1	2	5	2	7	9	8	F
## 257	1	1	4	6	14	12	13	P
## 258	1	1	1	12	11	11	11	P
## 259	1	2	4	8	15	14	14	P
## 260	1	1	2	0	10	9	0	F
## 261	1	3	2	21	17	18	18	P
## 262	1	1	3	2	8	8	8	F
## 263	1	1	3	1	13	12	12	P
## 264	1	1	4	4	10	9	9	F
## 265	1	1	3	0	9	10	0	F
## 266	3	4	1	13	17	17	17	P
## 267	3	4	5	2	9	9	10	P
## 268	2	2	4	8	12	10	11	P
## 269	1	3	5	10	10	9	10	P
## 270	1	2	3	0	6	0	0	F

## 271	3	3	5	15	9	9	9	F
## 272	1	3	2	4	15	14	14	P
## 273	1	1	3	2	11	11	11	P
## 274	2	2	1	2	15	14	14	P
## 275	1	1	1	2	10	10	10	P
## 276	2	3	5	6	12	12	12	P
## 277	1	1	5	75	10	9	9	F
## 278	1	4	3	22	9	9	9	F
## 279	1	1	4	15	9	8	8	F
## 280	1	2	1	8	10	11	10	P
## 281	2	4	5	30	8	8	8	F
## 282	3	4	3	19	11	9	10	P
## 283	1	1	3	1	12	12	12	P
## 284	1	1	4	4	8	9	10	P
## 285	1	2	5	4	10	9	11	P
## 286	1	2	4	2	12	10	11	P
## 287	1	2	2	5	18	18	19	P
## 288	1	1	3	6	13	12	12	P
## 289	1	3	2	6	15	14	14	P
## 290	1	1	2	9	15	13	15	P
## 291	1	4	5	11	12	11	11	P
## 292	1	2	3	0	15	15	15	P
## 293	1	1	5	12	12	12	13	P
## 294	1	1	3	6	18	18	18	P
## 295	1	1	4	8	14	13	14	P
## 296	1	3	5	4	14	12	11	P
## 297	2	3	2	0	10	9	0	F
## 298	1	2	2	10	10	8	8	F
## 299	1	1	3	0	14	13	14	P
## 300	2	2	1	5	16	15	16	P
## 301	1	1	4	14	12	10	11	P
## 302	2	2	5	0	11	11	10	P
## 303	1	1	3	0	15	12	14	P
## 304	1	2	5	0	17	17	18	P
## 305	1	1	3	20	15	14	13	P
## 306	1	1	3	8	14	12	12	P
## 307	1	1	5	0	17	18	18	P
## 308	1	1	4	38	8	9	8	F
## 309	1	2	5	0	15	12	12	P
## 310	1	3	3	18	12	10	10	P
## 311	2	2	3	0	9	9	0	F
## 312	1	1	2	20	14	12	13	P
## 313	2	2	4	3	13	11	11	P
## 314	1	2	1	22	13	10	11	P
## 315	1	1	3	14	15	13	13	P
## 316	1	1	3	40	13	11	11	P
## 317	1	2	1	0	8	8	0	F
## 318	1	1	5	9	9	10	9	F
## 319	2	5	5	0	11	11	10	P
## 320	3	3	5	2	11	11	11	P
## 321	1	2	5	23	13	13	13	P
## 322	1	1	3	12	11	9	9	F
## 323	2	2	3	3	11	11	11	P
## 324	2	3	5	1	12	14	15	P

## 325	2	3	2	0 16 15 15	P
## 326	2	2	3	3 9 12 11	P
## 327	3	5	5	3 14 15 16	P
## 328	5	5	4	8 11 10 10	P
## 329	1	3	4	7 10 9 9	F
## 330	1	2	4	4 14 14 14	P
## 331	2	4	5	2 9 8 8	F
## 332	1	1	5	7 12 14 14	P
## 333	1	1	4	0 7 0 0	F
## 334	1	1	2	0 8 8 0	F
## 335	1	1	4	0 10 9 0	F
## 336	1	3	5	16 16 15 15	P
## 337	1	2	5	12 14 13 13	P
## 338	2	3	2	0 7 8 0	F
## 339	1	1	1	7 16 15 17	P
## 340	2	3	2	4 9 10 10	P
## 341	1	3	3	4 11 12 11	P
## 342	2	2	2	0 10 10 0	F
## 343	1	3	5	11 16 15 15	P
## 344	1	2	4	0 9 8 0	F
## 345	1	2	3	4 11 10 10	P
## 346	2	3	1	7 13 13 14	P
## 347	1	2	4	9 16 15 16	P
## 348	2	3	5	0 10 10 9	F
## 349	1	3	4	0 13 15 15	P
## 350	5	5	5	10 11 13 13	P
## 351	3	3	2	8 8 7 8	F
## 352	2	3	3	2 13 13 13	P
## 353	2	3	3	7 8 7 8	F
## 354	3	3	5	4 8 8 8	F
## 355	1	3	2	4 13 11 11	P
## 356	1	1	5	0 10 9 9	F
## 357	1	2	5	4 12 13 13	P
## 358	1	2	5	2 12 12 11	P
## 359	1	2	3	4 10 10 10	P
## 360	1	1	4	0 18 16 16	P
## 361	1	4	5	0 13 13 13	P
## 362	2	3	5	2 13 12 12	P
## 363	1	3	3	0 11 11 10	P
## 364	1	1	1	0 16 15 15	P
## 365	1	2	3	0 12 11 12	P
## 366	2	4	3	4 10 10 10	P
## 367	2	2	5	0 13 13 13	P
## 368	1	2	1	0 7 6 0	F
## 369	1	2	4	0 11 10 10	P
## 370	4	2	5	10 14 12 11	P
## 371	1	1	3	4 7 7 9	F
## 372	2	3	3	3 14 12 12	P
## 373	1	1	3	8 13 11 11	P
## 374	1	3	1	14 6 5 5	F
## 375	1	1	1	0 19 18 19	P
## 376	1	2	4	2 8 8 10	P
## 377	1	1	3	4 15 14 15	P
## 378	3	4	2	4 8 9 10	P

```
## 379 1 2 1 0 15 15 15 P
## 380 2 3 1 17 10 10 10 P
## 381 1 4 2 4 15 14 14 P
## 382 1 3 5 5 7 6 7 F
## 383 1 1 3 2 11 11 10 P
## 384 1 3 5 0 6 5 0 F
## 385 4 3 3 14 6 5 5 F
## 386 1 3 4 2 10 9 10 P
## 387 2 2 5 7 6 5 6 F
## 388 1 2 5 0 7 5 0 F
## 389 1 1 1 0 7 9 8 F
## 390 1 1 5 0 6 5 0 F
## 391 4 5 4 11 9 9 9 F
## 392 3 4 2 3 14 16 16 P
## 393 3 3 3 3 10 8 7 F
## 394 3 4 5 0 11 12 10 P
## 395 3 3 5 5 8 9 9 F
```

Dummy code

```
stu.math1$PF.dv <- ifelse(stu.math1$PF == "P", 1, 0)
```

```
stu.math1$PF.dv
```

```
## [1] 0 0 1 1 1 1 1 0 1 1 0 1 1 1 1 1 1 0 1 1 1 1 1 0 0 1 1 1 1 1 1 1 1
## [36] 0 1 1 1 1 1 1 1 1 0 0 1 1 1 0 1 1 1 1 1 1 1 0 1 1 1 0 0 1 1 1 0 0 1
## [71] 1 1 0 1 1 1 1 1 1 0 1 1 0 1 1 0 0 1 1 0 0 1 0 1 1 1 1 1 0 0 1 1 0 1
## [106] 1 0 1 1 1 1 1 1 1 0 1 1 1 0 1 1 1 1 1 0 1 1 0 0 1 0 0 1 1 0 0 0 0 1 1
## [141] 0 0 1 1 0 1 0 1 0 1 0 1 0 1 1 0 1 0 1 1 1 1 0 0 0 1 0 1 1 1 0 1 0 1 1 0 0
## [176] 0 1 0 0 1 0 1 1 0 1 1 1 1 0 1 1 0 0 1 1 1 1 1 1 1 1 1 1 1 0 1 0 0 1 1 0
## [211] 0 1 1 0 1 1 0 0 0 1 0 0 1 1 1 0 1 1 1 0 1 1 1 0 1 0 1 1 1 1 0 1 1 0 1 0
## [246] 1 1 0 0 1 0 1 0 0 1 0 1 1 1 0 1 0 1 0 1 0 0 1 1 1 1 0 0 1 1 1 1 1 0 0 0 1
## [281] 0 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 0 0 1 1 1 1 1 1 1 1 1 1 0 1 1 0 1 1 1 1
## [316] 1 0 0 1 1 1 0 1 1 1 1 1 1 1 0 1 0 1 0 1 0 0 0 1 1 0 1 1 1 0 1 0 1 1 1 0 1 1
## [351] 0 1 0 0 1 0 1 1 1 1 1 1 1 1 1 1 1 0 1 1 0 1 1 0 1 1 1 1 1 1 1 1 0 1 0 0 0
## [386] 1 0 0 0 0 0 1 0 1 0
```

Converting categorical variable to binary indicator (Dummy code)

```
stu.math1$paid.dv <- ifelse(stu.math1$paid == "yes", 1, 0)
```

```
stu.math1$paid.dv
```

```
## [1] 0 0 1 1 1 1 0 0 1 1 1 0 1 1 0 0 1 0 0 1 0 1 0 0 1 1 1 1 0 1 1 0 0 0 1
## [36] 0 0 0 1 1 0 0 0 0 0 1 0 0 1 0 1 1 0 1 1 1 1 0 0 0 0 0 0 1 0 0 0 1 1 0
## [71] 1 0 0 0 1 1 0 1 0 0 1 1 1 1 0 1 0 0 1 0 1 1 0 1 0 1 0 1 0 1 1 0 0 1 1
## [106] 0 1 0 1 1 0 0 0 0 0 0 0 0 0 0 0 1 1 0 1 0 0 0 0 1 0 0 0 0 0 0 0 0 0 0
## [141] 0 0 1 0 0 1 0 1 0 0 0 0 1 0 0 1 0 0 0 1 0 0 0 0 0 0 0 1 1 1 0 1 1 0 1
## [176] 1 1 0 0 0 1 1 0 0 1 0 1 1 0 0 0 0 1 0 0 0 0 1 0 1 0 1 1 0 1 1 1 1 1 1
## [211] 1 1 1 1 1 1 1 1 1 1 1 0 0 0 1 0 0 0 1 1 0 0 1 0 1 1 0 1 0 0 0 0 1 0 0 1
## [246] 0 0 0 0 1 0 0 0 0 0 0 1 0 0 1 1 1 0 0 1 1 1 1 1 0 1 1 1 1 1 1 0 1 0 1
## [281] 1 0 1 0 0 1 1 1 1 1 1 1 0 1 1 1 1 1 1 1 0 1 1 1 0 1 0 1 0 0 0 0 0 1 0
## [316] 0 1 1 1 1 1 0 1 0 0 0 0 0 1 1 0 0 0 0 0 0 1 1 0 1 1 0 0 0 0 0 0 1 1 0
## [351] 0 1 0 0 1 0 1 0 0 0 0 0 0 1 0 1 1 1 1 1 0 1 0 0 0 0 1 1 1 1 1 0 0 0 1
## [386] 1 1 0 1 0 1 0 0 0 0
```

```
# Splitting the data
```

```
split_data <- sample.split(stu.math1, SplitRatio = 0.8)
```

```
train <- subset(stu.math1, split == TRUE)
```

```
test <- subset(stu.math1, split == FALSE)
```

```
# Training the model
```

```
model <- glm(PF.dv ~ paid.dv + age + G1 + G2, data = train, family = binomial)
```

```
summary(model)
```

```
##
```

```
## Call:
```

```
## glm(formula = PF.dv ~ paid.dv + age + G1 + G2, family = binomial,
```

```
## data = train)
```

```
##
```

```
## Deviance Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -2.60033 -0.08259  0.00665  0.13992  2.31790
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error z value Pr(>|z|)
```

```
## (Intercept) -13.10930     3.98027  -3.294 0.000989 ***
```

```
## paid.dv      -0.07451     0.45224  -0.165 0.869137
```

```
## age         -0.26583     0.19080  -1.393 0.163532
```

```
## G1           0.22458     0.15746   1.426 0.153805
```

```
## G2           1.68524     0.29520   5.709 1.14e-08 ***
```

```
## ---
```

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

```
##
```

```
## (Dispersion parameter for binomial family taken to be 1)
```

```
##
```

```
## Null deviance: 412.34 on 314 degrees of freedom
```

```
## Residual deviance: 124.22 on 310 degrees of freedom
```

```
## AIC: 134.22
```

```
##
```

```
## Number of Fisher Scoring iterations: 8
```

```
Backfitting (p-value)
```

```
# Removing paid.dv (highest p-value) from the model
```

```
model2 <- glm(PF.dv ~ age + G1 + G2, data = train, family = binomial)
```

```
summary(model2)
```

```
##
```

```
## Call:
```

```
## glm(formula = PF.dv ~ age + G1 + G2, family = binomial, data = train)
```

```
##
```

```
## Deviance Residuals:
```

```
##      Min       1Q   Median       3Q      Max
```

```
## -2.61203 -0.08120 0.00657 0.13995 2.33155
##
## Coefficients:
##             Estimate Std. Error z value Pr(>|z|)
## (Intercept) -13.1589    3.9673  -3.317 0.00091 ***
## age         -0.2641    0.1904  -1.387 0.16556
## G1           0.2240    0.1573   1.424 0.15447
## G2           1.6838    0.2949   5.709 1.14e-08 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 412.34  on 314  degrees of freedom
## Residual deviance: 124.25  on 311  degrees of freedom
## AIC: 132.25
##
## Number of Fisher Scoring iterations: 8
# Removing age (second highest p-value) from the model

model3 <- glm(PF.dv ~ G1 + G2, data = train, family = binomial)

summary(model3)

##
## Call:
## glm(formula = PF.dv ~ G1 + G2, family = binomial, data = train)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.55447 -0.06986  0.00668  0.11728  2.20993
##
## Coefficients:
##             Estimate Std. Error z value Pr(>|z|)
## (Intercept) -17.6393    2.6210  -6.730 1.70e-11 ***
## G1           0.1587    0.1492   1.063  0.288
## G2           1.7524    0.2939   5.963 2.48e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##    Null deviance: 412.34  on 314  degrees of freedom
## Residual deviance: 126.25  on 312  degrees of freedom
## AIC: 132.25
##
## Number of Fisher Scoring iterations: 8
# Removing G1 from the model

model4 <- glm(PF.dv ~ G2, data = train, family = binomial)

summary(model4)
```

```
##
## Call:
## glm(formula = PF.dv ~ G2, family = binomial, data = train)
##
## Deviance Residuals:
##      Min       1Q   Median       3Q      Max
## -2.57761  -0.06740   0.00644   0.10703   2.20446
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -17.3321     2.5563  -6.780  1.2e-11 ***
## G2           1.8743     0.2737   6.848  7.5e-12 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 412.34  on 314  degrees of freedom
## Residual deviance: 127.40  on 313  degrees of freedom
## AIC: 131.4
##
## Number of Fisher Scoring iterations: 8
```

The p-value backward elimination technique is used.

Predicting pass or fail

```
PF_pred <- predict(model4, test, type = "response")
PF_pred
```

```
##           3           9           17           23           24
## 8.805336e-02 9.999999e-01 9.998647e-01 9.999792e-01 9.991193e-01
##          28          31          37          43          51
## 9.999968e-01 9.639204e-01 9.999968e-01 9.999999e-01 9.991193e-01
##          57          58          62          65          71
## 9.999792e-01 9.999792e-01 8.805336e-02 8.039196e-01 9.999792e-01
##          77          85          91          92          96
## 9.639204e-01 8.039196e-01 1.460116e-02 9.999995e-01 8.039196e-01
##          99         105         111         119         125
## 9.998647e-01 9.999999e-01 1.000000e+00 1.460116e-02 1.460116e-02
##         126         130         133         139         145
## 9.991193e-01 9.999999e-01 9.991193e-01 9.942887e-01 2.970080e-08
##         153         159         160         164         167
## 8.039196e-01 9.999792e-01 9.942887e-01 8.039196e-01 8.039196e-01
##         173         179         187         193         194
## 9.639204e-01 8.805336e-02 9.942887e-01 8.805336e-02 3.861957e-01
##         198         201         207         213         221
## 3.861957e-01 9.999968e-01 1.460116e-02 9.991193e-01 2.268757e-03
##         227         228         232         235         241
## 9.999792e-01 9.639204e-01 9.639204e-01 1.460116e-02 9.942887e-01
##         247         255         261         262         266
## 9.942887e-01 9.942887e-01 9.999999e-01 8.805336e-02 9.999995e-01
##         269         275         281         289         295
## 3.861957e-01 8.039196e-01 8.805336e-02 9.998647e-01 9.991193e-01
##         296         300         303         309         315
## 9.942887e-01 9.999792e-01 9.942887e-01 9.942887e-01 9.991193e-01
```

```
##           323           329           330           334           337
## 9.639204e-01 3.861957e-01 9.998647e-01 8.805336e-02 9.991193e-01
##           343           349           357           363           364
## 9.999792e-01 9.999792e-01 9.991193e-01 9.639204e-01 9.999792e-01
##           368           371           377           383           391
## 2.268757e-03 1.460116e-02 9.998647e-01 9.639204e-01 3.861957e-01
```

Regression Equation

$$P(\text{Pass}) = 1/(1 + e^{-(-17.6 + 1.9 \cdot G2)})$$

```
# For age = 15 and G2 = 10, P(Pass) = 80%
```

```
new <- data.frame(G2=10)
res <- predict(model4,new,type = "response")
res
```

```
##           1
## 0.8039196
```

Accuracy of the model

```
# Creating Confusion Matrix
```

```
(table(ActualValue=test$PF.dv, PredictedValue=PF_pred>0.5))
```

```
##           PredictedValue
## ActualValue FALSE TRUE
##           0      16      0
##           1       5     59
```

Therefore, accuracy of model is $(22+55)/(22+55+5+0) = 93.9\%$

Problem 3:

Collecting and Exploring the data

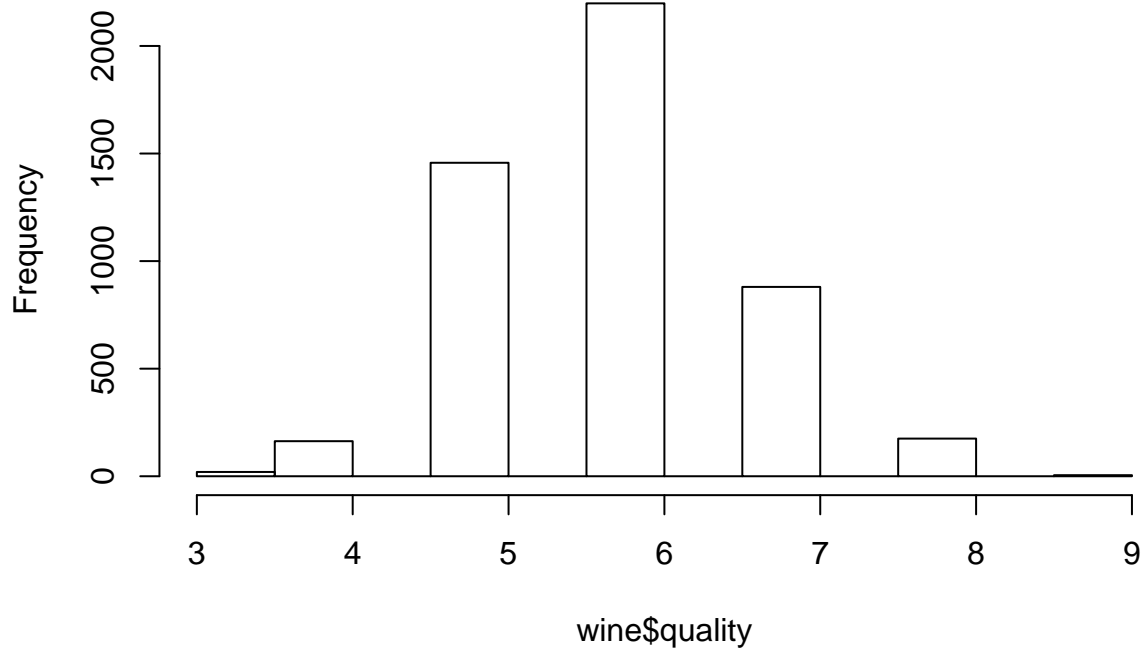
```
wine <- read.csv("C:/Users/Meghana Nadig/Downloads/whitewines.csv")
```

```
str(wine)
```

```
## 'data.frame':   4898 obs. of  12 variables:
## $ fixed.acidity      : num  7 6.3 8.1 7.2 7.2 8.1 6.2 7 6.3 8.1 ...
## $ volatile.acidity   : num  0.27 0.3 0.28 0.23 0.23 0.28 0.32 0.27 0.3 0.22 ...
## $ citric.acid        : num  0.36 0.34 0.4 0.32 0.32 0.4 0.16 0.36 0.34 0.43 ...
## $ residual.sugar     : num  20.7 1.6 6.9 8.5 8.5 6.9 7 20.7 1.6 1.5 ...
## $ chlorides          : num  0.045 0.049 0.05 0.058 0.058 0.05 0.045 0.045 0.049 0.044 ...
## $ free.sulfur.dioxide: num  45 14 30 47 47 30 30 45 14 28 ...
## $ total.sulfur.dioxide: num  170 132 97 186 186 97 136 170 132 129 ...
## $ density            : num  1.001 0.994 0.995 0.996 0.996 ...
## $ pH                 : num  3 3.3 3.26 3.19 3.19 3.26 3.18 3 3.3 3.22 ...
## $ sulphates          : num  0.45 0.49 0.44 0.4 0.4 0.44 0.47 0.45 0.49 0.45 ...
## $ alcohol            : num  8.8 9.5 10.1 9.9 9.9 10.1 9.6 8.8 9.5 11 ...
## $ quality            : int  6 6 6 6 6 6 6 6 6 6 ...
```

```
hist(wine$quality)
```


Histogram of wine\$quality



```
# Splitting data into training and testing dataset
```

```
wine_train <- wine[1:3750, ]
```

```
wine_test <- wine[3751:4898, ]
```

Training model on data

```
#install.packages("rpart")
```

```
library(rpart)
```

```
m.rpart <- rpart(quality ~ ., data = wine_train)
```

```
m.rpart
```

```
## n= 3750
```

```
##
```

```
## node), split, n, deviance, yval
```

```
## * denotes terminal node
```

```
##
```

```
## 1) root 3750 3140.06000 5.886933
```

```
## 2) alcohol< 10.85 2473 1510.66200 5.609381
```

```
## 4) volatile.acidity>=0.2425 1406 740.15080 5.402560
```

```
## 8) volatile.acidity>=0.4225 182 92.99451 4.994505 *
```

```
## 9) volatile.acidity< 0.4225 1224 612.34560 5.463235 *
```

```
## 5) volatile.acidity< 0.2425 1067 631.12090 5.881912 *
```

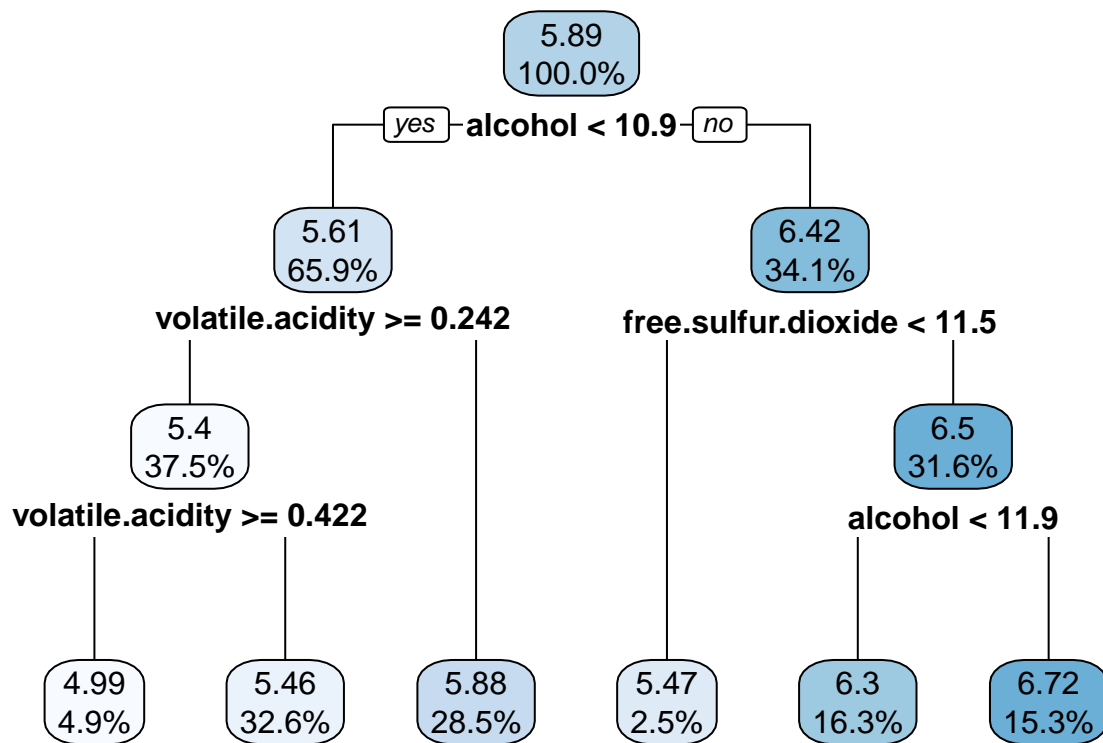
```
## 3) alcohol>=10.85 1277 1069.95800 6.424432
```

```
##      6) free.sulfur.dioxide< 11.5 93    99.18280 5.473118 *
##      7) free.sulfur.dioxide>=11.5 1184  879.99920 6.499155
##      14) alcohol< 11.85 611  447.38130 6.296236 *
##      15) alcohol>=11.85 573  380.63180 6.715532 *
```

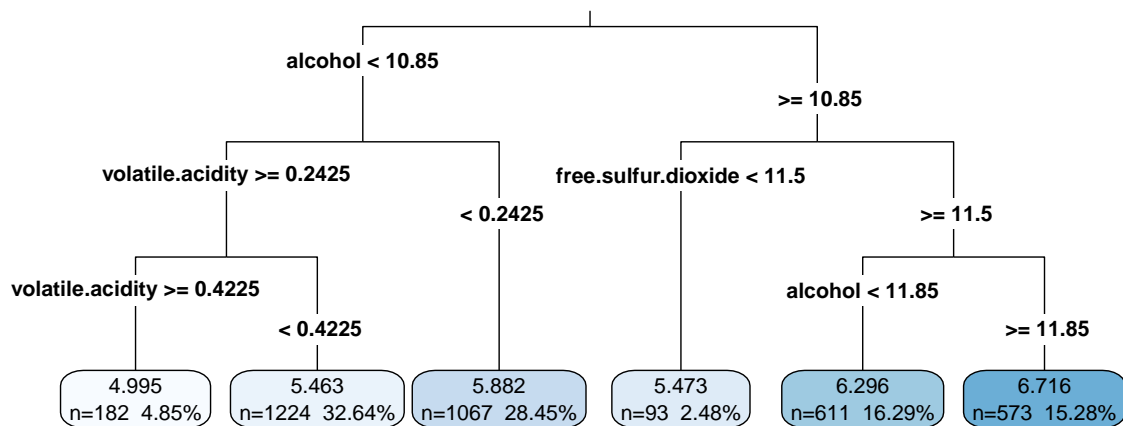
Visualizing decision trees

```
#install.packages("rpart.plot")
library(rpart.plot)

rpart.plot(m.rpart, digits = 3)
```



```
rpart.plot(m.rpart, digits = 4, fallen.leaves = TRUE,
           type = 3, extra = 101)
```



Evaluating model performance

```
p.rpart <- predict(m.rpart, wine_test)
```

```
summary(p.rpart)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    4.995   5.463   5.882   5.999   6.296   6.716
```

```
summary(wine_test$quality)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    3.000   5.000   6.000   5.848   6.000   8.000
```

```
cor(p.rpart, wine_test$quality)
```

```
## [1] 0.4931608
```

Measuring performance with mean absolute error

```
MAE <- function(actual, predicted) {
  mean(abs(actual - predicted))
}
```

```
MAE(p.rpart, wine_test$quality)
```

```
## [1] 0.5732104
```

```
# Mean quality rating in the training data
```

```
mean(wine_train$quality)
```

```
## [1] 5.886933
```

```
# Mean quality rating in the testing data
```

```
MAE(5.87, wine_test$quality)
```

```
## [1] 0.5815679
```

Improving model performance

```
library(RWeka)
```

```
##
```

```
## Attaching package: 'RWeka'
```

```
## The following object is masked from 'package:caTools':
```

```
##
```

```
##      LogitBoost
```

```
m.m5p <- M5P(quality ~ ., data = wine_train)
```

```
m.m5p
```

```
## M5 pruned model tree:
```

```
## (using smoothed linear models)
```

```
##
```

```
## alcohol <= 10.85 : LM1 (2473/77.476%)
```

```
## alcohol > 10.85 :
```

```
## |   free.sulfur.dioxide <= 20.5 :
```

```
## | |   free.sulfur.dioxide <= 10.5 : LM2 (81/104.574%)
```

```
## | |   free.sulfur.dioxide > 10.5 : LM3 (224/87.002%)
```

```
## |   free.sulfur.dioxide > 20.5 : LM4 (972/84.073%)
```

```
##
```

```
## LM num: 1
```

```
## quality =
```

```
## 0.0777 * fixed.acidity
```

```
## - 2.3087 * volatile.acidity
```

```
## + 0.0732 * residual.sugar
```

```
## + 0.0022 * free.sulfur.dioxide
```

```
## - 155.0175 * density
```

```
## + 0.6462 * pH
```

```
## + 0.7923 * sulphates
```

```
## + 0.0758 * alcohol
```

```
## + 156.2102
```

```
##
```

```
## LM num: 2
```

```
## quality =
```

```
## -0.0314 * fixed.acidity
```

```
## - 0.3415 * volatile.acidity
```

```
## + 1.7929 * citric.acid
```

```
## + 0.1316 * residual.sugar
```

```
## - 0.2456 * chlorides
```

```
## + 0.1212 * free.sulfur.dioxide
```

```
## - 178.6281 * density
```

```
## + 0.054 * pH
```

```
## + 0.1392 * sulphates
## + 0.0108 * alcohol
## + 180.6069
##
## LM num: 3
## quality =
## -0.2019 * fixed.acidity
## - 2.3804 * volatile.acidity
## - 1.0851 * citric.acid
## + 0.0905 * residual.sugar
## - 0.2456 * chlorides
## + 0.0041 * free.sulfur.dioxide
## - 177.078 * density
## + 0.054 * pH
## + 0.0868 * sulphates
## + 0.0108 * alcohol
## + 183.5076
##
## LM num: 4
## quality =
## 0.0004 * fixed.acidity
## - 0.0325 * volatile.acidity
## + 0.0957 * residual.sugar
## - 5.9702 * chlorides
## + 0.0002 * free.sulfur.dioxide
## - 172.3931 * density
## + 1.0123 * pH
## + 1.1653 * sulphates
## + 0.1542 * alcohol
## + 171.6842
##
## Number of Rules : 4
```

```
summary(m.m5p)
```

```
##
## === Summary ===
##
## Correlation coefficient          0.5932
## Mean absolute error             0.5804
## Root mean squared error         0.7367
## Relative absolute error         83.3671 %
## Root relative squared error     80.507 %
## Total Number of Instances      3750
```

```
p.m5p <- predict(m.m5p, wine_test)
```

```
summary(p.m5p)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##    4.170   5.646   6.032   6.079   6.501   7.913
```

```
# Finding correlation
```

```
cor(p.m5p, wine_test$quality)
```

```
## [1] 0.531723
```

```

# Mean absolute error
MAE(wine_test$quality, p.m5p)

## [1] 0.5660352

RSME of the model
RSME <- function(actual, predicted) {
  sqrt(mean((actual - predicted)^2))
}

RSME(wine_test$quality,p.m5p)

## [1] 0.7191548

RSME

## function(actual, predicted) {
## sqrt(mean((actual - predicted)^2))
## }

```

Therefore, the RSME of the model is 0.72