EECS 4412 A Data Mining

Project Phase 2

Group Member:

Trevor Ramtahal (216771875)

Ka Hei Chan (216328387)

Yongjie Ba (215122856)

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Professor: Habib-ur Rehman

Appendix A

I **Yongjie Ba** student ID # **215122856** acknowledge that I have contributed at least 30% time and effort to the preparation of this report and work discussed herein.

Student Signature: Y rie bu

I *Trevor Ramtahal* student ID # 216771875 acknowledge that I have contributed at least 30% time and effort to the preparation of this report and work discussed herein.

Student Signature: Twoy Romtohal

I *Ka Hei Chan* student ID # 216328387 acknowledge that I have contributed at least 30% time and effort to the preparation of this report and work discussed herein.

Student_Signature: Ka Hei Oun

Task 1: Describing the Data

For the Family status data set:

| sex | age | address | famsize | Pstatus | Medu | Fedu | Mjob | Fjob | guardian | failures | famsup | famrel | absences | G1 | G2 | G3 | |
|--------|-----|--------------|------------|---------|------|------|-------------------------|-------------------|----------|----------|---------------|--------|----------|----|----|----|---------------|
| F | | 18 U | GT3 | Α | | 4 | 4 at_home | teacher | mother | | 0 no | | 4 | 6 | 5 | 6 | 6 |
| F | | 17 U | GT3 | T | | 1 | 1 at_home | other | father | | 0 yes | | 5 | 4 | 5 | 5 | 6 |
| F | | 15 U | LE3 | T | | 1 | 1 at_home | other | mother | | 3 no | | 4 | 10 | 7 | 8 | 10¦ |
| F | | 15 U | GT3 | T | | 4 | 2 health | services | mother | | 0 yes | | 3 | 2 | 15 | 14 | 15 |
| F | | 16 U | GT3 | T | | 3 | 3 other | other | father | | 0 yes | | 4 | 4 | 6 | 10 | 10Î |
| M | | 16 U | LE3 | T | | 4 | 3 services | other | mother | | 0 yes | | 5 | 10 | 15 | 15 | 15 |
| M | | 16 U | LE3 | T | | 2 | 2 other | other | mother | | 0 no | | 4 | 0 | 12 | 12 | 11 6 19 |
| F | | 17 U | GT3 | Α | | 4 | 4 other | teacher | mother | | 0 yes | | 4 | 6 | 6 | 5 | 6 |
| М | | 15 U | LE3 | Α | | 3 | 2 services | other | mother | | 0 yes | | 4 | 0 | 16 | 18 | 19 |
| М | | 15 U | GT3 | T | | 3 | 4 other | other | mother | | 0 yes | | 5 | 0 | 14 | 15 | 15i |
| F | | 15 U | GT3 | T | | 4 | 4 teacher | health | mother | | 0 yes | | 3 | 0 | 10 | 8 | 9 |
| F | | 15 U | GT3 | T | | 2 | 1 services | other | father | | 0 yes | | 5 | 4 | 10 | 12 | 12 |
| М | | 15 U | LE3 | T | | 4 | 4 health | services | father | | 0 yes | | 4 | 2 | 14 | 14 | 14 |
| M | | 15 U | GT3 | Ť | | 4 | 3 teacher | other | mother | | 0 yes | | 5 | 2 | 10 | 10 | 11 |
| M | | 15 U | GT3 | A | | 2 | 2 other | other | other | | 0 yes | | 4 | 0 | 14 | 16 | 16 |
| F | | 16 U | GT3 | T | | 4 | 4 health | other | mother | | 0 yes | | 4 | 4 | 14 | 14 | 16 14 |
| F | | 16 U | GT3 | Ť | | 4 | 4 services | services | mother | | 0 yes | | 3 | 6 | 13 | 14 | 14 |
| F | | 16 U | GT3 | Ť | | 3 | 3 other | other | mother | | 0 yes | | 5 | 4 | 8 | 10 | 10 |
| M | | 17 U | GT3 | Ť | | 3 | 2 services | services | mother | | 3 yes | | 5 | 16 | 6 | 5 | 5 |
| M | | 16 U | LE3 | Ť | | 4 | 3 health | other | father | | 0 no | | 3 | 4 | 8 | 10 | 10 |
| М | | 15 U | GT3 | Ť | | 4 | 3 teacher | other | mother | | 0 no | | 4 | 0 | 13 | 14 | 15 |
| M | | 15 U | GT3 | Ť | | 4 | 4 health | health | father | | 0 yes | | 5 | 0 | 12 | 15 | 15 |
| M | | 16 U | LE3 | Ť | | 4 | 2 teacher | other | mother | | 0 no | | 4 | 2 | 15 | 15 | 16 |
| M | | 16 U | LE3 | Ť | | 2 | 2 other | other | mother | | 0 yes | | 5 | 0 | 13 | 13 | 12 |
| IC IVI | | 15 R | GT3 | Ť | | 2 | 4 services | health | mother | | 0 yes | | 4 | 2 | 10 | 9 | 12 8 |
| - | | 16 U | GT3 | Ť | | 2 | 2 services | services | mother | | 2 yes | | 1 | 14 | 6 | 9 | - °i |
| M | | 15 U | GT3 | Ť | | 2 | 2 other | other | mother | | | | 4 | 2 | 12 | 12 | 8 11 |
| M | | 15 U | GT3 | Ť | | 4 | 2 health | | mother | | 0 yes 0 no | | 2 | 4 | 15 | 16 | 15 |
| M | | 16 U | LE3 | A | | 3 | 4 services | services other | mother | | | | 5 | 4 | 11 | 11 | 11 |
| M | | 16 U | GT3 | T | | 4 | | | | | 0 yes | | 4 | 16 | 10 | 12 | 11 |
| M | | | | T | | 4 | 4 teacher | teacher | mother | | 0 yes | | | 0 | | 11 | 40 |
| M | | 15 U 15 U | GT3 GT3 | T | | 4 | 4 health | services | mother | | 0 yes | | 4 | 0 | 9 | 16 | 12 17 |
| M | | | | T | | 4 | 4 services 3 teacher | services | mother | | 0 yes | | 4 | 0 | | | 1/j |
| | | 15 R | GT3 | | | | | at_home | mother | | 0 yes | | - | • | 17 | 16 | 16 |
| M | | 15 U | LE3 | T | | 3 | 3 other | other | mother | | 0 no | | 5 | 0 | 8 | 10 | 12 15 6 |
| M | | 16 U | GT3 | T | | 3 | 2 other | other | mother | | 0 yes | | 5 | 0 | 12 | 14 | 15 |
| F | | 15 U | GT3 | T | | 2 | 3 other | other | father | | 0 yes | | 3 | 0 | 8 | 7 | 61 |
| M | | 15 U | LE3 | Ţ | | 4 | 3 teacher | services | mother | | 0 yes | | 5 | 2 | 15 | 16 | 18 |
| M | | 16 R | GT3 | A | | 4 | 4 other | teacher | mother | | 0 yes | | 2 | 7 | 15 | 16 | 15 |
| F | | 15 R | GT3 | T | | 3 | 4 services | health | mother | | 0 yes | | 4 | 2 | 12 | 12 | 11 |
| F | | 15 R | GT3 | T | | 2 | 2 at_home | other | mother | | 0 yes | | 4 | 8 | 14 | 13 | 13 |
| ļF . | | 16 U | LE3 | T | | 2 | 2 other | other | mother | | 1 yes | | 3 | 25 | 7 | 10 | 11¦ |
| M | | 15 U | LE3 | T | | 4 | 4 teacher | other | other | | 0 yes | | 5 | 8 | 12 | 12 | 12 |
| M | | 15 U | GT3 | T | | 4 | 4 services | teacher | father | | 0 yes | | 4 | 2 | 19 | 18 | 18 |
| M | | 15 U | GT3 | T | | 2 | 2 services | services | father | | 0 yes | | 5 | 0 | 8 | 8 | 11 9 6 |
| F | | 16 U | LE3 | T | | 2 | 2 other | at_home | father | | 1 no | | 4 | 14 | 10 | 10 | 9i |
| F | | 15 U | LE3 | Α | | 4 | 3 other | other | mother | | 0 yes | | 5 | 8 | 8 | 8 | 6 |
| F | | 16 U | LE3 | Α | | 3 | 3 other | services | mother | | 0 yes | | 2 | 12 | 11 | 12 | 11 |
| М | | 16 U | GT3 | T | | 4 | 3 health | services | mother | | 0 no | | 4 | 4 | 19 | 19 | 20 |
| М | | 15 U | GT3 | T | | 4 | 2 teacher | other | mother | | 0 yes | | 4 | 2 | 15 | 15 | 14 |
| F | | 15 U | GT3 | T | | 4 | 4 services | teacher | father | | 1 yes | | 4 | 2 | 7 | 7 | 7 |

Adding a new feature: Ave G - the average grade of G1, G2 and G3

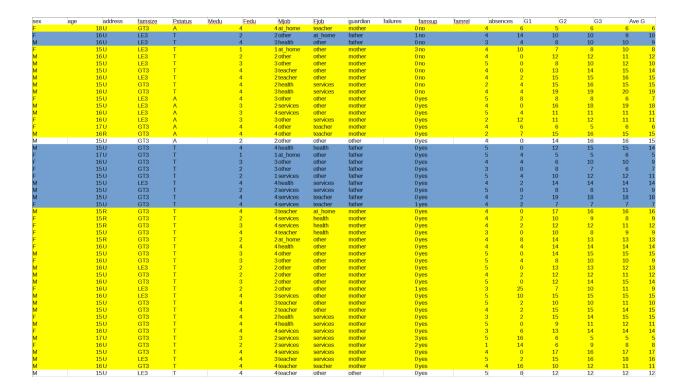
Then classify the data:

Set yellow background colour to the student whose guardian is the mother;

Set blue background colour to the student whose guardian is the father;

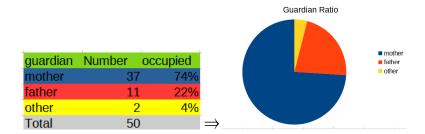
Set white background colour to the student whose guardian is the other;

For detail please see the picture below;



Based on the dataset, we can get:

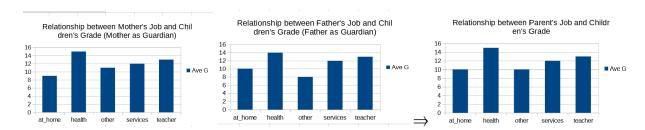
1. The guardian ratio between father, mother and others:



2. The average grade of students whose guardian is a mother or father and then get the average:

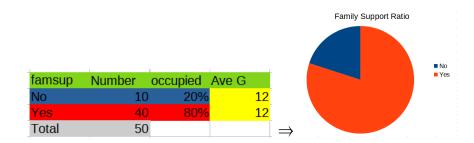
| 1 | guardian | Miob | Ave G | | guardian | Fiob | Ave G | | P_Job | Ave G | |
|-----|----------|----------|-------|----|----------|----------|-------|------------------|----------|-------|----|
| - 1 | mother | at home | | 9 | father | at home | | 10 | at_home | | 10 |
| | | health | | 15 | | health | | 14 | health | | 15 |
| | | other | | 11 | | other | | 8 | other | | 10 |
| | | services | | 12 | | services | | 12 | services | | 12 |
| | | teacher | | 13 | | teacher | | $13 \Rightarrow$ | teacher | | 13 |

Then we can transform the tables into graphs:

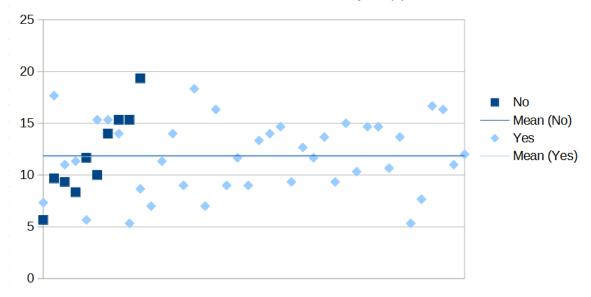


We can find that the guardian did less influence on students' grades.

3. We can also know whether a student has or has not their family support and also support impacts their grades:



Student Grade with Family Support



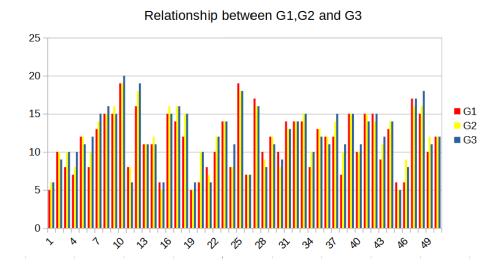
And we can find that the mean of yes and no are the same.

=> Family support did less influence on students' grades.

4. We can get the relationship between students' absences and their grades:

Usually high average grade means fewer absences.

5. The closely Relationship between G1, G2 and G3



The G3 is high related to G1 and G2.

6. The relationship between the relationship of students' parents and students' Grade

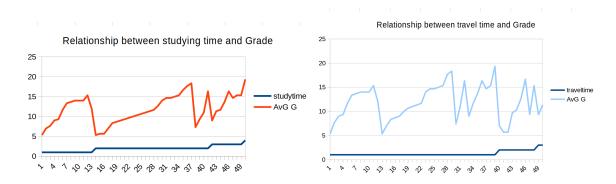
| Pstatus | Ave G |
|---------|-------|
| Α | 11.17 |
| Т | 11.91 |
| | |

For the data set of students' spend time on:

| sex | age | address | traveltime | studytime | failures | romantic | freetime | goout | Dalc | Walc | absences | G1 | G2 | G3 |
|-----|-----|---------|------------|-----------|----------|----------|----------|-------|------|------|----------|----|------|------|
| F | | U | | 2 | 2 | 0 no | | 3 | 4 1 | 1 | 1 6 | | i (| 6 |
| F | 17 | U | | 1 | 2 | 0 no | | 3 | 3 1 | 1 | 1 4 | | 5 | 5 6 |
| F | 15 | U | | 1 | 2 | 3 no | | 3 | 2 2 | 3 | 3 10 | 7 | 7 | 3 10 |
| F | 15 | U | | 1 | 3 | 0 yes | | 2 | 2 1 | 1 | 1 2 | 15 | 1- | 1 15 |
| F | 16 | U | | 1 | 2 | 0 no | | 3 | 2 1 | 2 | 2 4 | | 5 1 | |
| M | 16 | U | | 1 | 2 | 0 no | | 4 | 2 1 | 2 | 2 10 | 15 | j 1 | 5 15 |
| M | 16 | U | | 1 | 2 | 0 no | | 4 | 4 1 | 1 | 1 0 | 12 | 2 1: | 2 11 |
| F | 17 | U | | 2 | 2 | 0 no | | 1 | 4 1 | 1 | 1 6 | (| 6 | 5 6 |
| M | 15 | U | | 1 | 2 | 0 no | | 2 | 2 1 | 1 | 1 0 | 16 | 1 | 3 19 |
| M | 15 | U | | 1 | 2 | 0 no | | 5 | 1 1 | 1 | 1 0 | 14 | 1 | 5 15 |
| F | 15 | U | | 1 | 2 | 0 no | | 3 | 3 1 | 2 | 2 0 | 10 |) | 3 9 |
| F | 15 | U | | 3 | 3 | 0 no | | 2 | 2 1 | 1 | 1 4 | 10 | 1: | 2 12 |
| M | 15 | U | | 1 | 1 | 0 no | | 3 | 3 1 | 3 | 3 2 | 14 | 1 1 | 1 14 |
| M | 15 | U | | 2 | 2 | 0 no | | 4 | 3 1 | 2 | 2 2 | 10 | 1 |) 11 |
| M | 15 | U | | 1 | 3 | 0 yes | | 5 | 2 1 | 1 | 1 0 | 14 | 1 | 5 16 |
| F | 16 | U | | 1 | 1 | 0 no | | 4 | 4 1 | 2 | 2 4 | 14 | 1- | 1 14 |
| F | 16 | U | | 1 | 3 | 0 no | | 2 | 3 1 | 2 | 2 6 | 13 | 3 1- | 1 14 |
| F | 16 | U | | 3 | 2 | 0 no | | 3 | 2 1 | 1 | 1 4 | | 3 1 | 10 |
| M | 17 | U | | 1 | 1 | 3 no | | 5 | 5 2 | . 4 | 1 16 | (| 6 | 5 5 |
| M | 16 | U | | 1 | 1 | 0 no | | 1 | 3 1 | 3 | 3 4 | | 3 1 |) 10 |
| M | 15 | U | | 1 | 2 | 0 no | | 4 | 1 1 | 1 | 1 0 | 13 | 3 1 | 1 15 |
| M | 15 | U | | 1 | 1 | 0 no | | 4 | 2 1 | 1 | 1 0 | 12 | 2 1 | 5 15 |
| M | 16 | U | | 1 | 2 | 0 no | | 5 | 1 1 | 3 | 3 2 | 15 | 1 | 5 16 |
| M | 16 | U | | 2 | 2 | 0 no | | 4 | 4 2 | . 4 | 1 0 | 13 | 1: | 3 12 |
| F | 15 | R | | 1 | 3 | 0 no | | 3 | 2 1 | 1 | 1 2 | 10 |) | 9 8 |
| F | 16 | U | | 1 | 1 | 2 no | | 2 | 2 1 | 3 | 3 14 | | 6 | 9 8 |
| M | 15 | U | | 1 | 1 | 0 no | | 2 | 2 1 | 2 | 2 2 | 12 | 2 1: | |
| M | 15 | U | | 1 | 1 | 0 no | | 2 | 4 2 | . 4 | 4 | 15 | 1 | |
| M | 16 | U | | 1 | 2 | 0 no | | 3 | 3 1 | | | | | |
| M | 16 | U | | 1 | 2 | 0 yes | | 4 | 5 5 | 5 | 16 | 10 | 1: | |
| M | 15 | U | | 1 | 2 | 0 no | | 4 | 2 3 | 4 | 1 0 | 9 | 1 | |
| M | 15 | | | 2 | 2 | 0 no | | 3 | 1 1 | 1 | 1 0 | | | |
| M | 15 | | | 1 | 2 | 0 yes | | 5 | 2 1 | | | | | |
| M | | U | | 1 | 2 | 0 no | | 3 | 2 1 | | | | | |
| M | | U | | 1 | 1 | 0 no | | 4 | 3 1 | | | | | |
| F | 15 | | | 2 | 1 | 0 no | | 5 | 1 1 | | | | | 7 6 |
| M | 15 | | | 1 | 3 | 0 no | | 4 | 3 1 | | | | | |
| M | | R | | 2 | 3 | 0 yes | | 4 | 3 1 | | | | | |
| F | | R | | 1 | 3 | 0 no | | 3 | 2 1 | | | | | |
| F | 15 | | | 1 | 1 | 0 no | | 3 | 1 1 | | | | | |
| F | 16 | | | 2 | 2 | 1 yes | | 3 | 3 1 | | | | | |
| M | | U | | 1 | 1 | 0 yes | | 4 | 3 2 | | | | | |
| M | 15 | | | 1 | 2 | 0 no | | 3 | 3 1 | | | | | |
| M | 15 | | | 1 | 1 | 0 no | | 4 | 1 1 | | | | | 3 11 |
| F | 16 | | | 2 | 2 | 1 no | | 3 | 3 2 | | | | | |
| F | 15 | | | 1 | 2 | 0 yes | | 2 | 2 1 | | | | | 3 6 |
| F | | U | | 1 | 2 | 0 no | | 3 | 5 1 | | | | | |
| M | 16 | | | 1 | 4 | 0 no | | 2 | 2 1 | | | | | |
| M | | U | | 1 | 2 | 0 no | | 3 | 3 2 | | | | | |
| F | 15 | U | | 1 | 2 | 1 no | | 4 | 4 1 | 1 | 1 2 | 1 | 7 | 7 |

Based on the dataset, we can get:

1. Relationship between students who spent more time on studying and their grades, and relationship between students who spend more time on travel and their grades:



As the time spent on studying increases, the student's grade is also increasing, also decreasing as travelling time increases overall.

2. The ratio of students who has a romantic relationship or not:

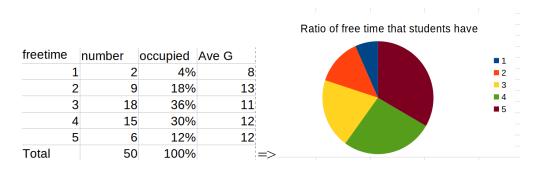
romantic: (nominal, binary) student in a romantic relationship, 'yes' or 'no'

| romantic | number | occupied | Ave G | |
|----------|--------|----------|-------|---|
| yes | 8 | 16% | 1 | 3 |
| no | 42 | 84% | 1 | 2 |
| Total | 50 | | | |

3. The ratio of free time that students have:

freetime: (interval) student's free time after school, ranked from 1 - very low up

to 5 - very high



Based on this table, we can easily get the ratio of free time that students have and also know that less free time does not mean high grade.

4. The relationship between Dalc, Walc and students' Grade:

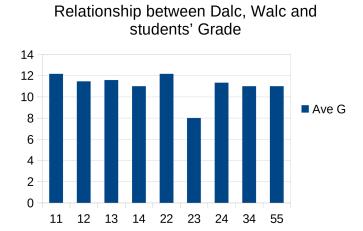
Dalc - (interval) same ranking system as freetime, but for how much alcohol the student consumes on workdays

Walc - (interval) same ranking system as freetime, but for how much alcohol the student consumes on weekends

DW - Dalc and Walc: if both Dalc and Walc is 1, then DW is 11.

| DW | number | Ave G |
|----|--------|-------|
| 11 | 28 | 12 |
| 12 | 8 | 11 |
| 13 | 4 | 12 |
| 14 | 1 | 11 |
| 22 | 2 | 12 |
| 23 | 1 | 8 |
| 24 | | 11 |
| 34 | 1 | 11 |
| 55 | 1 | 11 |

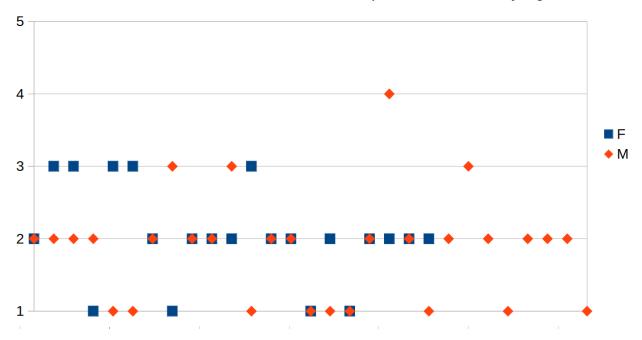
Then we can translate the data into graphics:



Based on the graph, the students who consume alcohol on weekends will get more grades.

5. The distribution of both males and females spent studying

Distribution of Male and Female spent time on studying

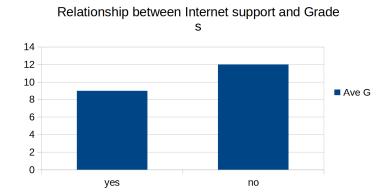


For the data set of support that students received:

| sex | age | address | failures schoolsup | famsup | paid | activities | internet | absences | G1 | G2 | G3 |
|---------|-----|---------|--------------------|--------|------|------------|----------|----------|------|----|---------------|
| F | | 18 U | 0 yes | no | no | no | no | | 6 5 | 6 | 6 |
| F | | 17 U | 0 no | yes | no | no | yes | | 4 5 | 5 | 6 |
| F | | 15 U | 3 yes | no | yes | no | yes | 1 | 0 7 | 8 | 10 |
| F | | 15 U | 0 no | yes | yes | yes | yes | | 2 15 | 14 | 15 |
| F | | 16 U | 0 no | yes | yes | no | no | | 4 6 | 10 | 10 |
| M | | 16 U | 0 no | yes | yes | yes | yes | 1 | 0 15 | 15 | 15 |
| M | | 16 U | 0 no | no | no | no | yes | | 0 12 | 12 | |
| F | | 17 U | 0 yes | yes | no | no | no | | 6 | 5 | 6 |
| М | | 15 U | 0 no | yes | yes | no | yes | | 0 16 | | 19 |
| M | | 15 U | 0 no | yes | yes | yes | yes | | 0 14 | 15 | 15 |
| F | | 15 U | 0 no | yes | yes | no | yes | | 0 10 | | 9 |
| F | | 15 U | 0 no | yes | no | yes | yes | | 4 10 | 12 | |
| М | | 15 U | 0 no | yes | yes | yes | yes | | 2 14 | | 14 |
| M | | 15 U | 0 no | yes | yes | no | yes | | 2 10 | 10 | 11 |
| M | | 15 U | 0 no | yes | no | no | yes | | 0 14 | 16 | 16 |
| ļF | | 16 U | 0 no | yes | no | no | yes | | 4 14 | 14 | 14 |
| ļΕ | | 16 U | 0 no | yes | yes | yes | yes | | 6 13 | | 14 |
| F | | 16 U | 0 yes | yes | no | yes | no | | 4 8 | | 10 |
| M | | 17 U | 3 no | yes | no | yes | yes | 1 | | | 5 |
| М | | 16 U | 0 no | no | yes | yes | yes | | 4 8 | | 10 |
| M | | 15 U | 0 no | no | no | no | yes | | 0 13 | | 15 |
| M | | 15 U | 0 no | yes | yes | no | yes | | 0 12 | | 15 |
| М | | 16 U | 0 no | no | no | yes | yes | | 2 15 | | 16 |
| M | | 16 U | 0 no | yes | no | yes | yes | | 0 13 | | 12 |
| }F | | 15 R | 0 yes | yes | yes | yes | yes | | 2 10 | | 8 |
| ļF | | 16 U | 2 no | yes | yes | no | yes | 1- | | | |
| M | | 15 U | 0 no | yes | yes | no | yes | | 2 12 | | 11 |
| M | | 15 U | 0 no | no | yes | no | yes | | 4 15 | | 15 |
| M | | 16 U | 0 yes | yes | no | yes | yes | | 4 11 | | 11 |
| M | | 16 U | 0 no | yes | yes | yes | yes | 1 | | | 11 |
| М | | 15 U | 0 no | yes | yes | no | yes | | 0 9 | | 12 |
| М | | 15 U | 0 no | yes | no | yes | yes | | 0 17 | | 17 |
| M | | 15 R | 0 no | yes | no | yes | yes | | 0 17 | | 16 |
| M | | 15 U | 0 no | no | no | yes | yes | | 3 0 | | 12 |
| ļΜ | | 16 U | 0 no | yes | yes | no | yes | | 0 12 | | 12 15 6 |
| F | | 15 U | 0 no | yes | no | yes | no | | 3 0 | | 6 |
| M. | | 15 U | 0 no | yes | no | yes | yes | | 2 15 | | 18 |
| M | | 16 R | 0 no | yes | no | yes | yes | | 7 15 | | 15 |
| F | | 15 R | 0 yes | yes | yes | yes | yes | | 2 12 | | 11 |
| ΙF | | 15 R | 0 yes | yes | yes | yes | no | | 8 14 | | 13 |
| F | | 16 U | 1 no | yes | no | yes | yes | 2 | | | 11 |
| М | | 15 U | 0 no | yes | no | no | yes | | 8 12 | | 12 |
| M | | 15 U | 0 no | yes | no | yes | yes | | 2 19 | | 18 |
| M | | 15 U | 0 yes | yes | no | no | yes | | 3 0 | | 11 9 |
| Ľ. | | 16 U | 1 yes | no | no | yes | yes | 1. | | | 9 |
| - - | | 15 U | 0 yes | yes | yes | yes | yes | | 8 8 | | 6 |
| l+ | | 16 U | 0 no | yes | no | no | yes | 1: | | | 11 |
| M | | 16 U | 0 no | no | no | yes | yes | | 4 19 | | 20 |
| M | | 15 U | 0 no | yes | yes | no | no | | 2 15 | | 14 |
| E | | 15 U | 1 yes | yes | no | yes | yes | | 2 7 | 7 | 7 |

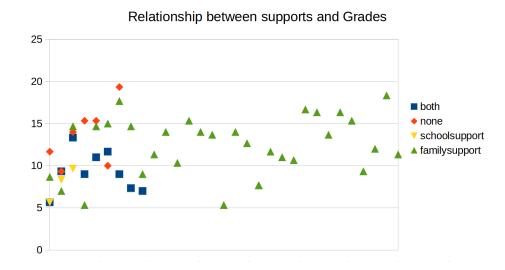
Based on the data set, we can get:

1. The relationship between Internet support and Grades:



By using this graph, we can quickly get how the internet influences students' grades.

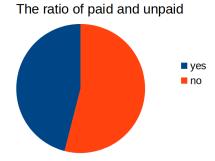
2. The relationship between family support or school support or both or none and students' grades:



Using this technology, we can quickly figure out how different support influences students' grades.

3. The ratio of paid courses and unpaid courses and their influences on students' grades:

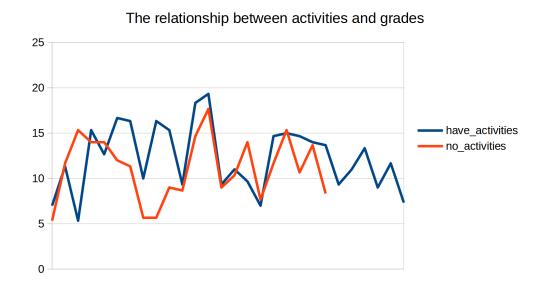
| paid | number | occupied | Ave G |
|------|--------|------------|-------|
| yes | 23 | 46% | 12.71 |
| no | 27 | 54% | 11.01 |
| | | | |
| | | difference | 8.49% |



We can find that the students who paid for extra courses will have 2.83% grades higher than the students who are not paid for extra courses.

Also by using this chart and table, we can quickly know whats the occupied of paid courses on total students courses. And we also know the impact that having paid courses on students' grades.

4. The relationship between having or not having activities and their grades:



By using this graph, we can quickly get what's the difference between having activities and not having activities that impact students' grades.

Task 2: Basic Statistical Analysis

In 216771875-216328387-215122856-T2.csv

The selected dimensions are bold below.

Nominal: sex, address, Pstatus, Mjob, Fjob, guardian, famsup

Ordinal: **Medu**, Fedu, famrel

Interval: famsize

Ratio: age, failures, absences, G1, G2, G3

For preprocessing, by using simple random sampling (get 50 objects from the original dataset) will help us save more time.

| Number | Occupied | |
|------------------|-------------------------------|--|
| 4 | 8% | |
| 8 | 16% | |
| 16 | 32% | |
| 13 | 26% | |
| 9 | 18% | |
| 50 | | |
| | | |
| can not determin | | |
| other | | |
| | 4 8 16 13 9 50 | |

For the mother's job, other have more numbers, so the mode of Mjob is other.

But the mean can not be determined, because the attribute of a mother's job is nominal.

| Medu | | Number |
|-------|---|--------|
| | 0 | 0 |
| | 1 | 2 |
| | 2 | 12 |
| | 3 | 10 |
| | 4 | 26 |
| Total | | 50 |
| | | |
| Mean | | 3.2 |
| Mode | | 4 |

Based on the table we can quickly get the mode is 4, which is post-secondary education

also the mean of the mother's education level is $3.2 (\approx 3)$, which is secondary education.

| famsize | Number | |
|---------|-------------|---------|
| GT3 | 34 | |
| LE3 | 16 | |
| Total | 50 | |
| | | |
| Mean | can not det | termine |
| Mode | GT3 | |

By counting the dataset, we can get there are 34 students whose family members are greater than 3, and 16 students whose family members are less than or equal to 3, so the mode is GT3, but we can not get the mean since this interval can not be calculated in numbers.

| G3 | |
|------------|-------|
| Total Mark | 604 |
| Number | 50 |
| Mean | 12.08 |
| Mode | 11 |

For G3, we can get the mean of 12.08 (\approx 12), and after sorting the G3 of all students, we can quickly get the mode of G3 is 11.

Task 3: Standardization and Normalization

In 216771875-216328387-215122856-T3.csv

The selected dimensions are:

Interval: freetime

Ratio: **G3**

Z-score standardization formula: z-score = x - mean / stdev; (x: the element to be compared;

mean: mathematical mean of the attribute; stdev: standard deviation of attribute)

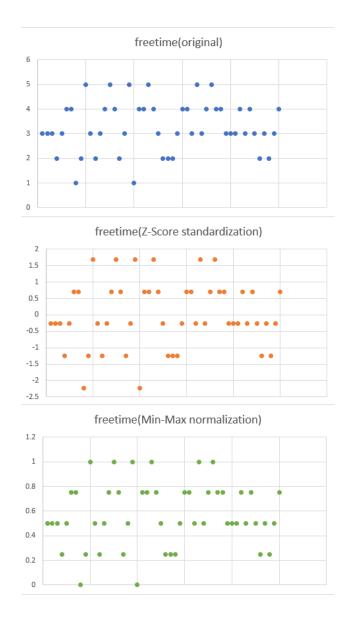
Min-Max normalization formula: xscale = x-xmin/(xmax-xmin); (x: element to be compared;

xmin: minimum value in attribute; xmax: maximum value of an attribute)

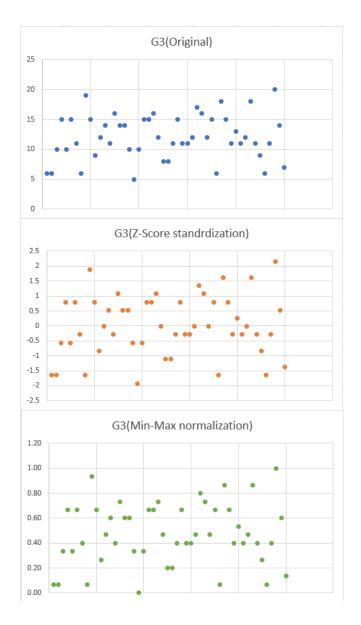
For the attribute freetime, after performing Z-score standardization and Min-Max normalization as seen in the graph, although from the original data we can analyze the attribute based on the

occurrence of the value, after the processes we can analyze how many elements are close/far

from the mathematical mean.



Similar to freetime, the G3 attribute also benefits on the two processes. By standardization and normalization, every datapoint has the same scale thus each feature is equally important.



Despite having a similar scatter pattern, the two processes should result in better scatter (more distinct than the original) if the range of values of the attributes were larger. On another hand, as both attributes do not have outlier due to the small range, performing Min-Max normalization and Z-score standardization results in a clear chart

Task 4: Principal Component Analysis

For task 4, the dataset regarding family-grade relations has the most, i.e. 14 attributes, said dataset is chosen. In order to ease the calculation process, we have dropped several insignificant attributes and resulted in the below spreadsheet, focusing on mother/father's education level, family's relationship, failures, and the three grades.

| Medu | Fedu | failures | famrel | G1 | G2 | G3 |
|------|------|----------|--------|----------|----------|----|
| 4 | 4 | 0 | 4 | 5 | 6 | |
| 1 | 1 | 0 | 5 | 5 | 5 | 6 |
| 1 | 1 | 3 | 4 | 7 | 8 | 10 |
| 4 | 2 | 0 | 3 | 15 | 14 | 15 |
| 3 | 3 | 0 | 4 | 6 | 10 | 10 |
| 4 | 3 | 0 | 5 | 15 | 15 | |
| 2 | 2 | 0 | 4 | 12 | 12 | |
| 4 | 4 | 0 | 4 | 6 | 5 | |
| 3 | 2 | 0 | 4 | 16 | 18 | |
| 3 | 4 | 0 | 5 | 14 | 15 | 15 |
| 4 | 4 | 0 | 3 | 10 | 8 | |
| 2 | 1 | 0 | 5 | 10 | 12 | |
| 4 | 4 | 0 | 4 | 14 | 14 | |
| 4 | 3 | 0 | 5 | 10 | 10 | 11 |
| 2 | 2 | 0 | 4 | 14 | 16 | |
| 4 | 4 | 0 | 4 | 14 | 14 | 14 |
| 4 | 4 | 0 | 3 | 13 | 14 | 14 |
| 3 | 3 | 0 | 5 | 8 | 10 | 10 |
| 3 | 2 | 3 | 5 | 6 | 5 | |
| 4 | 3 | 0 | 3 | 8 | 10 | |
| 4 | 3 | 0 | 4 | 13 | 14 | 15 |
| 4 | 4 | 0 | 5 | 12 | 15 | |
| 4 | 2 | 0 | 4 | 15 | 15 | |
| 2 | 2 | 0 | 5 | 13 | 13 | |
| 2 | 4 | 0 | 4 | 10 | 9 | |
| 2 | 2 | 2 | 1 | 6 | 9 | |
| 2 | 2 | 0 | 4 | 12 | 12 | |
| 4 | 2 | 0 | 2 | 15 | 16 | |
| 3 | 4 | 0 | 5 | 11 | 11 | 11 |
| 4 | 4 | 0 | 4 | 10 | 12 | |
| 4 | 4 | 0 | 5 | 9 | 11 | 12 |
| 4 | 4 | 0 | 4 | 17 | 16 | |
| 4 | 3 | 0 | 4 | 17 | 16 | |
| 3 | 3 | 0 | 5 | 8 | 10 | |
| 3 | 2 | 0 | 5 | 12 | 14 | 15 |
| 2 | 3 | 0 | 3 | 8 | 7 | |
| 4 | 3 | 0 | 5 | 15 | 16 | |
| 4 | 4 | 0 | 2 | 15 | 16 | |
| 3 | 4 | 0 | 4 | 12 | 12 | |
| 2 | 2 | 0 | 3 | 14 7 | 13 10 | |
| | 4 | 1 | 5 | | | |
| 4 | 4 | 0 | | 12 19 | 12 | |
| 2 | 2 | 0 | 5 | 8 | 18 8 | |
| 2 | 2 | 1 | 4 | 10 | 10 | |
| 4 | 3 | 0 | 5 | 8 | 8 | |
| 3 | 3 | 0 | 2 | 11 | 12 | |
| 4 | 3 | 0 | 4 | 19 | 19 | |
| 4 | 2 | 0 | 4 | 15 | 15 | |
| 4 | 4 | 1 | 4 | 7 | 7 | 7 |

To perform PCA, standardization is needed to compare every component in the same scale.

Using the Z-score standardization result in the dataset below:

| After Standa | | £=:1 | £1 | 01 | C2 | C2 |
|--------------|-------|----------|--------|-------|-------|----------|
| Medu | | failures | famrel | G1 | G2 | G3 |
| 0.85 | 1.13 | -0.33 | -0.04 | -1.73 | -1.67 | -1.65 |
| -2.35 | -2.01 | -0.33 | 1.02 | -1.73 | -1.95 | -1.65393 |
| -2.35 | -2.01 | 4.14 | -0.04 | -1.18 | -1.11 | -0.56582 |
| 0.85 | -0.96 | -0.33 | -1.11 | 0.99 | 0.58 | 0.794323 |
| -0.21 | 0.08 | -0.33 | -0.04 | -1.46 | -0.55 | -0.56582 |
| 0.85 | 0.08 | -0.33 | 1.02 | 0.99 | 0.86 | 0.794323 |
| -1.28 | -0.96 | -0.33 | -0.04 | 0.17 | 0.02 | -0.29379 |
| 0.85 | 1.13 | -0.33 | -0.04 | -1.46 | -1.95 | -1.65393 |
| -0.21 | -0.96 | -0.33 | -0.04 | 1.26 | 1.70 | 1.882437 |
| -0.21 | 1.13 | -0.33 | 1.02 | 0.72 | 0.86 | 0.794323 |
| 0.85 | 1.13 | -0.33 | -1.11 | -0.37 | -1.11 | -0.83785 |
| -1.28 | -2.01 | -0.33 | 1.02 | -0.37 | 0.02 | -0.02176 |
| 0.85 | 1.13 | -0.33 | -0.04 | 0.72 | 0.58 | 0.522295 |
| 0.85 | 0.08 | -0.33 | 1.02 | -0.37 | -0.55 | -0.29379 |
| -1.28 | -0.96 | -0.33 | -0.04 | 0.72 | 1.14 | 1.066352 |
| 0.85 | 1.13 | -0.33 | -0.04 | 0.72 | 0.58 | 0.522295 |
| 0.85 | 1.13 | -0.33 | -1.11 | 0.45 | 0.58 | 0.522295 |
| -0.21 | 0.08 | -0.33 | 1.02 | -0.91 | -0.55 | -0.56582 |
| -0.21 | -0.96 | 4.14 | 1.02 | -1.46 | -1.95 | -1.92596 |
| 0.85 | 0.08 | -0.33 | -1.11 | -0.91 | -0.55 | -0.56582 |
| 0.85 | 0.08 | -0.33 | -0.04 | 0.45 | 0.58 | 0.794323 |
| 0.85 | 1.13 | -0.33 | 1.02 | 0.17 | 0.86 | 0.794323 |
| 0.85 | -0.96 | -0.33 | -0.04 | 0.99 | 0.86 | 1.066352 |
| -1.28 | -0.96 | -0.33 | 1.02 | 0.45 | 0.30 | -0.02176 |
| -1.28 | 1.13 | -0.33 | -0.04 | -0.37 | -0.83 | -1.10988 |
| -1.28 | -0.96 | 2.65 | -3.24 | -1.46 | -0.83 | -1.10988 |
| -1.28 | -0.96 | -0.33 | -0.04 | 0.17 | 0.02 | -0.29379 |
| 0.85 | -0.96 | -0.33 | -2.18 | 0.99 | 1.14 | 0.794323 |
| -0.21 | 1.13 | -0.33 | 1.02 | -0.10 | -0.26 | -0.29379 |
| 0.85 | 1.13 | -0.33 | -0.04 | -0.37 | 0.02 | -0.29379 |
| 0.85 | 1.13 | -0.33 | 1.02 | -0.64 | -0.26 | -0.02176 |
| 0.85 | 1.13 | -0.33 | -0.04 | 1.53 | 1.14 | 1.33838 |
| 0.85 | 0.08 | -0.33 | -0.04 | 1.53 | 1.14 | 1.066352 |
| -0.21 | 0.08 | -0.33 | 1.02 | -0.91 | -0.55 | -0.02176 |
| -0.21 | -0.96 | -0.33 | 1.02 | 0.17 | 0.58 | 0.794323 |
| -1.28 | 0.08 | -0.33 | -1.11 | -0.91 | -1.39 | -1.65393 |
| 0.85 | 0.08 | -0.33 | 1.02 | 0.99 | 1.14 | 1.610409 |
| 0.85 | 1.13 | -0.33 | -2.18 | 0.99 | 1.14 | |
| -0.21 | 1.13 | -0.33 | -0.04 | 0.17 | 0.02 | -0.29379 |
| -1.28 | -0.96 | -0.33 | -0.04 | 0.72 | 0.30 | 0.250266 |
| -1.28 | -0.96 | 1.16 | -1.11 | -1.18 | -0.55 | -0.29379 |
| 0.85 | 1.13 | -0.33 | 1.02 | 0.17 | 0.02 | -0.02176 |
| 0.85 | 1.13 | -0.33 | -0.04 | 2.08 | 1.70 | 1.610409 |
| -1.28 | -0.96 | -0.33 | 1.02 | -0.91 | -1.11 | -0.29379 |
| -1.28 | -0.96 | 1.16 | -0.04 | -0.37 | -0.55 | -0.83785 |
| 0.85 | 0.08 | -0.33 | 1.02 | -0.91 | -1.11 | -1.65393 |
| -0.21 | 0.08 | -0.33 | -2.18 | -0.10 | 0.02 | -0.29379 |
| 0.85 | 0.08 | -0.33 | -0.04 | 2.08 | 1.98 | 2.154466 |
| 0.85 | -0.96 | -0.33 | -0.04 | 0.99 | 0.86 | 0.522295 |
| 0.85 | 1.13 | 1.16 | -0.04 | -1.18 | -1.39 | -1.3819 |

After standardization, we can compute the covariance matrix for the data set by using

$$Cov(X,Y) = \frac{\sum_{i=1}^{n} (X_i - \overline{X})(Y_i - \overline{Y})}{n-1}$$

Resulting the covariance matrix below:

| Pearson's r | Medu V | Fedu | failures | famrel | G1 | G3 | G2 |
|-------------|--------|--------|----------|--------|--------|--------|--------|
| Medu | 0.11 | 0.642 | -0.355 | -0.032 | 0.385 | 0.349 | 0.357 |
| Fedu | 0.642 | - | -0.346 | 0.004 | 0.122 | 0.036 | 0.075 |
| failures | -0.355 | -0.346 | - | -0.141 | -0.404 | -0.363 | -0.396 |
| famrel | -0.032 | 0.004 | -0.141 | - | -0.045 | 0.016 | -0.071 |
| G1 | 0.385 | 0.122 | -0.404 | -0.045 | - | 0.896 | 0.936 |
| G3 | 0.349 | 0.036 | -0.363 | 0.016 | 0.896 | - | 0.962 |
| G2 | 0.357 | 0.075 | -0.396 | -0.071 | 0.936 | 0.962 | - |

After computing covariance, we then can calculate the Eigenvalue

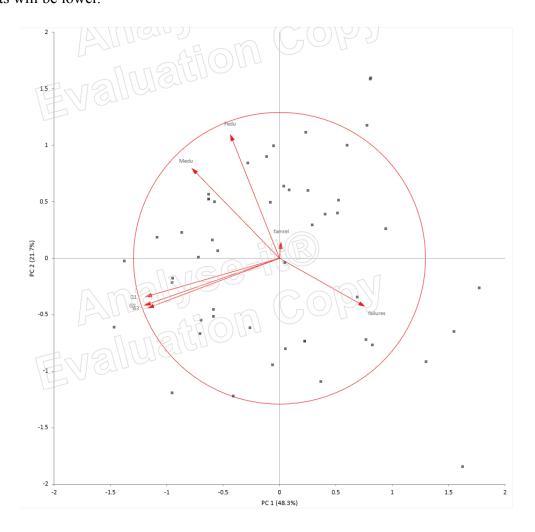
$$\det(\mathbf{A} - \lambda \mathbf{I}) = 0$$

by solving lamda, then choose the largest two lambda as principal components.

Using Microsoft Excel add-ins, analyse-it, the result table shows that the first component has a strong relation toward the three grades and average relation toward Medu, Fedu, and failures. The second component shows strong relation toward Medu and Fedu. Hence these two components are chosen and highlighted.

| | Component | | | | | | |
|----------|-----------|--------|--------|--------|--------|--------|--------|
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 |
| Medu | -0.328 | 0.503 | -0.172 | -0.350 | 0.697 | -0.036 | -0.033 |
| Fedu | -0.185 | 0.690 | -0.121 | -0.125 | -0.672 | 0.084 | 0.024 |
| failures | 0.321 | -0.267 | -0.302 | -0.842 | -0.155 | -0.022 | -0.025 |
| famre | 0.006 | 0.092 | 0.928 | -0.354 | -0.017 | -0.039 | -0.061 |
| G1 | -0.503 | -0.214 | -0.024 | -0.073 | -0.154 | -0.786 | 0.231 |
| G2 | -0.507 | -0.262 | -0.038 | -0.057 | -0.115 | 0.194 | -0.787 |
| G3 | -0.495 | -0.277 | 0.039 | -0.139 | -0.037 | 0.578 | 0.567 |
| | | | | | | | |

The below biplot indicates the relations between different attributes. In terms of the length of the vector, we can discover the following G1, G2, and G3. Medu and Fedu have the second strongest representation, and famrel has the least representation. Apart from that, it also indicates that the G1, G2 and G3 have a positive correlation, indicating that if a student's grade is already high in the first exam, he/she will tend to score high in the following two exams. On the other hand, it shows that Medu has a stronger positive correlation with the grades than Fedu, showing that a mother's educational level tends to contribute more to the student's academic achievement in this dataset. Furthermore, it shows that the failures attribute has a negative correlation with the grades, which is common knowledge, indicating that with more failures the grades of the students will be lower.



Task 5: Similarity Measurement

- The object type to be used for this task will be from timer.csv, or the table from task one which was focused on what students did in their free time. To fit the requirements of the task, the 5 numeric dimensions to be picked are the following
 - Age (ratio), which describes the student's age
 - Studytime (ordinal), which describes numerically the amount of study time a student has
 - Freetime (interval), describes numerically the amount of free time a student has
 - Absences (ratio), describes the amount of time a student has been absent
 - G3 (ratio), the numeric value of the student's final grade
- As one of the 5 questions within task one was how the ratio of a student's study time and free time affect their grades (G3), the focus will be on these three attributes
- The filtered data can be found in 216771875-216328387-215122856-T5Data.csv

The following CSV files for euclidean and cosine distance matrices were formed using the file EECS4412 Task5.ipynb (using Jupyter Notebook, the sklearn imported functions).

- 216771875-216328387-215122856-T5EU.csv
- 216771875-216328387-215122856-T5CO.csv
- 216771875-216328387-215122856-T5MA.csv

Notes:

when checking the jupyter notebook file the
 216771875-216328387-215122856-T5Data.csv file must be in the same folder/directory
 as the notebook file, keep all files from TASK 5 folder within the TASK 5 folder

- Images of the code (if necessary) and the code itself can be found within the TASK_5 folder, all imports are stated at the beginning of the code and consists of only imports available when installing Jupyter Notebook and Anaconda

Task 6: Classification

- The original classifier to be used with the data table from task 5 would have been G3, and through using regression and a decision tree classifier for testing, we would have possibly been able to predict how study time and free time affect a student's grades. However, as we are only working with 50 objects, this becomes infeasible as it would be basically impossible to create an accurate classifier with 20 different outcomes. Instead, as shown in EECS4412 Task6.ipynb we have done the following preprocessing steps:
 - As we want to answer the question of how study time and free time affect one's grades, we will be removing age and absences from the table
 - To give our classifier an easier time, we need to reduce the total number of possible classifications. Based on the value of G3, a new value is added to each object, called rating. The rating could have one of 3 values, 'Failing' if G3 is between 0 and 10, 'Passing' if G3 is between 10 and 15, and 'Strongly_Passing' if G3 is between 15 and 20. With the reduction of classes, with our total of 50 objects, the classifier should have an easier time making classification versus having the original 20 from G3
 - G3 is removed as it is basically redundant
- Thus, our final table is located in 216771875-216328387-215122856-T6Data.csv and contains the attributes studytime, freetime and rating.

- Decision Tree Classification Analysis

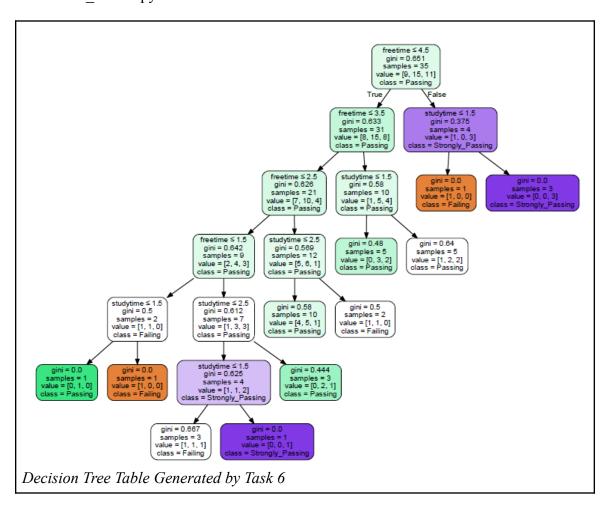
- The split we decided to go with for the data set was a 70 30 split, 70 for training data and 30 for testing, the code for all the decision tree classification can be found within EECS4412 Task6.ipynb
- The Accuracy of this decision tree in our case happens to be 0.666. This indicates to us that there isn't a correlation between how a student spends their time and their grades, thus these attributes are 1, unusable (at least by themselves) to predict a student's grade, and 2, have no to little effect on a student's grades
- The Accuracy is affected by the data sample used for training. Take for instance the snippet of code below from EECS4412 Task6.ipynb

- If we change the random_state to 1, we will receive an accuracy of 0.333, while a random_state of 2 would be 0.533. While increasing the amount of training data may assist in having better accuracy, the best possible classifier developed was is random_state of 3, with an accuracy of 0.666
- On top of this, analysis from task 1 (page 7-8) has indicated that study time and free time have no guaranteed effect on one's grades as well.

 While there is a slight upward trend in terms of study time versus grades, there may not be enough data (mainly in the case of the number of students who spend the most time studying) to accurately train the

classifier to determine one's grades. Free time also appears to have little correlation with grades, and other than students with the least free time, they all had about the same average for grades. With more data on students who spend more time, we may be able to accurately see that upward trend in our analysis, and train our classifier better.

- The Decision Tree generated can be viewed below, or will be within the task 6 folder (decision_tree_graph.pdf), or will be generated once EECS4412 Task6.ipynb is run:



Notes

- Images of the code and the code itself (for jupyter notebook) can be found within the TASK_6 folder all imports are from the jupyter and anaconda installation except for the ones notated with #graphviz
- Much like task 5, keep all files from the TASK_6 folder within the TASK_6 folder for the Jupyter Notebook code to work properly
 - To install graphviz, in the Anaconda Navigator type the following
 - conda install graphviz
 - conda install python-graphviz
 - More Documentation at https://graphviz.readthedocs.io/en/stable/manual.html
 - More Troubleshooting can be found from
 https://stackoverflow.com/questions/33433274/anaconda-graphviz-cant-import-aft
 er-installation
 - ALTERNATIVELY, the code snippet (from Task6_IMG04.png) can be commented out and the two import lines annotated with #graphviz can be removed, code will function but no visual graph will be created

Task 7: Progress on your Objectives

1. From table 1, based on the quality of family relationships and the parents' cohabitation status, we had raised the question of "How does the quality of family relationships affect the students' performance and would students be taken care of by single parents who struggle on their performances?"

From our analysis in task 1 (page 6), we can see that a good relationship between students' parents will lead to higher grades for students.

| | Pstatus | Ave G |
|--------|---------|-------|
| | Α | 11.17 |
| Found: | Т | 11.91 |
| | | |

A: students' parents apart from each other;

T: students' parents living together;

- 2. In table 1, based on the attributes of family information, we have asked the question "Is there a positive correlation between parents' high level of education or whose job is teachers and students' performance?" From our analysis in task 1 (page 4-5), we can see that parents with jobs we typically associate with requiring education such as health and teaching had a higher average grade in comparison to other jobs and parents who stayed at home. While this does answer the question we can possibly use this data for further analysis.
- 3. From table 2, we originally raised the question of "how does the ratio between time spent on studying and time spent on not studying affect the students' performances?" Through the analysis, we have done on task 1 (page 7-8) and using the attributes of study time and free time in the decision tree classifier in task 6, we can come to the conclusion that study time and free time are not major contributing factors to a student's grades with our current data. We can come to the conclusion due to two main reasons, one is the fact that while there is an upward trend between study time and grades, we may not have enough evidence within 50 students to say this is without a doubt accurate and there is little correlation between free time and grade. Secondly, our decision tree classifier had a max accuracy of 66.66%, pointing to the idea that these attributes are not the best to use for predicting grades. From our analysis on page 7, we have also found that lower travelling time resulted in higher

- grades, so we can instead ask "how does the ratio between time spent on studying and time spent travelling affect the students' performances?"
- 4. In table 3, we can find whether extra paid classes have a positive, neutral or negative effect on the students' performances by asking the question "Is there a correlation between extra paid classes and the students' performances?" From our analysis in task 1, we can see there is a positive correlation between having paid for classes and grades, as paying resulted in a higher average grade compared to not doing so. This answers the question we raised and we can use this data further to aid in future attempts at classification.
- 5. In table 3, we have a question: "Whether it is more beneficial for their performance if students accept support not only from school but also from family?" Through analysis done in task one (page 11), we can make several observations
 - a. Having only school support typically resulted in lower grades
 - b. Having no support typically had higher grades, family support and no support are comparable in their averages
 - c. Having support from both school and family typically resulted in lower grades

 This somewhat answers our question with our limited results from 50 students, which we
 can use to assist in a classifier.