1. Find the distance between objects 1 and 3 by using the formula provided on the slides. Notice that we have mixed type of attributes.

| Object Identifier | test-1(nominal) | test-2 (ordinal) | test-3 (numeric) | |
|-------------------|-----------------|------------------|------------------|--|
| 1 | A excelle | | t 45 | |
| 2 | В | fair | 22 64 | |
| 3 | С | good | | |
| 4 A | | excellent | 28 | |

1 (A, excellent, 45)

3 (C, good, 64)

Nominal:

m: # of matches, p: total # of variables

m=0,p=1

$$d_{13} = \frac{p - m}{p} = \frac{1 - 0}{1} = 1$$

$$\delta_{13}^{(f)} = 1$$

Ordinal:

fair = 1, good = 2, excellent = 3

fair = 1-1/3-1 = 0, good = 2-1/3-1 = 0.5, excellent = 3-1/3-1 = 1

$$d_{13} = |1 - 0.5| = 0.5$$

$$\delta_{13}^{(f)} = 1$$

Numeric:

$$d_{13} = \frac{|45 - 64|}{64 - 22} = \frac{19}{42}$$

$$\delta_{13}^{(f)} = 1$$

Distance between 1 and 3:

$$d(1,3) = \frac{1+0.5 + \frac{19}{42}}{1+1+1} = \frac{82}{42} = 0.65$$

2. Write a program in any language which can compute Manhattan and Euclidean distances between any two given vectors with any length. You can pass the length to your function, but please don't limit the dimension to 2. You can test your function on vectors you fill in your code without asking user input.

```
#Create function for Manhattan and Euclidean distances.
Manhattan Distance <- function(vector1, vector2) {
  sum(abs(vector1 - vector2))
}
Euclidean Distance <- function(vector1, vector2) {</pre>
  sqrt(sum((vector1 - vector2)^2))
}
#Create 2 vectors.
vector_s1 <- c(9, 7, 3, 10, 25)
vector s2 <- c(4, 10, 30, 22, 100)
#Get the result use the distance function with the vectors
M_Result <- Manhattan_Distance(vector_s1, vector_s2)</pre>
E_Result <- Euclidean_Distance(vector_s1, vector_s2)</pre>
#Print the result
cat("Manhattan Distances: ", M_Result, "\n")
cat("Euclidean Distances: ", E Result, "\n")
Manhattan Distance = |9-4|+|7-10|+|3-30|+|10-22|+|2-100|=122
Euclidean Distance = \sqrt{(9-4)^2 + (7-10)^2 + (3-30)^2 + (10-22)^2 + (2-100)^2}
                   =80.82079
```

```
> #Create function for Manhattan and Euclidean distances
> Manhattan_Distance <- function(vector1, vector2) {</pre>
    sum(abs(vector1 - vector2))
+ }
> Euclidean_Distance <- function(vector1, vector2) {</pre>
    sqrt(sum((vector1 - vector2)^2))
+ }
> #Create 2 vectors
> vector_s1 <- c(9, 7, 3, 10, 25)
> vector_s2 <- c(4, 10, 30, 22, 100)
> #Get the result use the distance function with the vectors
> M_Result <- Manhattan_Distance(vector_s1, vector_s2)</pre>
> E_Result <- Euclidean_Distance(vector_s1, vector_s2)</pre>
> #Print the result
> cat("Manhattan Distances: ", M_Result, "\n")
Manhattan Distances: 122
> cat("Euclidean Distances: ", E_Result, "\n")
Euclidean Distances: 80.82079
```

3. In the table below, determine whether passing a class has a dependency on attendance by using Chi-square test. Please refer to the formula in the slides.

| | Passed | | Failed | | Tatal |
|----------|--------|----------|--------|----------|-------|
| | Real | Expected | Real | Expected | Total |
| Attended | 25.00 | 18.94 | 6.00 | 12.06 | 31.00 |
| Skipped | 8.00 | 14.06 | 15.00 | 8.94 | 23.00 |
| Total | 33.00 | | 21.00 | | 54.00 |

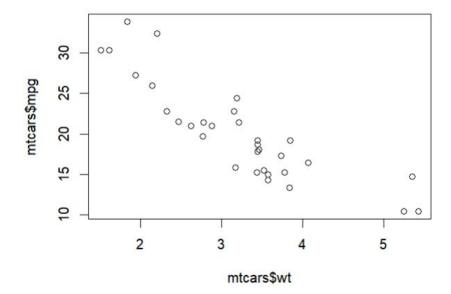
$$\chi^2 = \frac{(25 - 18.94)^2}{18.94} + \frac{(8 - 14.06)^2}{14.06} + \frac{(6 - 12.06)^2}{12.06} + \frac{(15 - 8.94)^2}{8.94} = 11.70$$

11.7 < 25 So passing a class hasn't a dependency on attendance.

4. In R, there is a built-in data frame called mtcars. Please calculate the correlation between mpg and wt attributes of mtcars by using cor() function. Then generate scatter plot based on these two attributes. Your scatter plot should be like the one below. You don't need to submit the image, but R script should be submitted

cor(mtcars\$mpg,mtcars\$wt)

scatter.smooth(x=mtcars\$wt,y=mtcars\$mpg,evaluation = 0)



5. Grad Students Only Write an R or Python script which removes or drops the columns which have more than 75% missing values. Then it should replace the missing values in the remaining columns with the median value of the existing values of that particular column. Download metabolite.csv from Google Drive and use this data set to test your code. Please check the end of this document for some useful R examples and hints. (10 points)

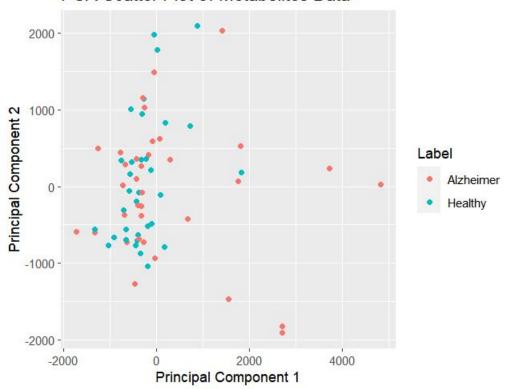
Total 192 column.

There are 4 columns which have more than 75% missing values.

```
Drop column Names:
[1] "Nitro.Tyr"
[1] "PEA"
[1] "Spermine"
[1] "PC.ae.C38.1"
#Start get library and read the files
library(tidyverse)
setwd("C:\\Users\\DELL\\Desktop\\CSC587\\datamining-main\\Rscripts")
data.metabolite <- file.path('data', 'metabolite.csv')
metabolite = read.delim(data.metabolite, sep=',', header = T)
#Get the list of the columns which will not be removed.
columns remain <- NULL
for (i in 1:ncol(metabolite)) {
  missing_proportion <- sum(is.na(metabolite[,i])) / length(metabolite[,i])
  if(missing_proportion>0.75){
    drop column name <- colnames(metabolite)[i]
    print(drop_column_name)
    remain_column_name <- colnames(metabolite)[i]
    columns_remain <- c(columns_remain_remain_column_name)</pre>
  }
}
metabolite <- subset(metabolite, select=columns remain)
#Fill the missing values with the median value in their own columns
for (i in 1:ncol(metabolite)) {
  median_value <- median(metabolite[,i],na.rm = TRUE)</pre>
  for (j in 1:nrow(metabolite)) {
    if(is.na(metabolite[j,i])){
       metabolite[j,i] <- median_value
    }
  }
}
print(metabolite)
```

6. Grad Students Only Please apply PCA on the processed metabolites data and create a scatter plot by using first two principal components in which points are colored based on the Label column. Please submit your code along with your figure in the same file.

PCA Scatter Plot of Metabolites Data



```
PC<sub>1</sub>
                         PC<sub>2</sub>
                                  Label
                  416.88290 Alzheimer
    -179.20084
1
2
   -1323.26135
                 -604.19473 Alzheimer
3
                 -730.06345 Alzheimer
    -647.24841
4
   -1719.11694
                 -586.89472 Alzheimer
5
    -461.84103
                -1273.68357 Alzheimer
6
                   14.82042 Alzheimer
    -728.83748
7
    -104.46543
                 -487.11446
                                Healthy
8
    -664.43109
                 -696.05942
                               Healthy
9
    -401.39663
                 -630.04775
                               Healthy
10
     188.99259
                  831.00122
                               Healthy
11
    -309.52815
                  944.75819
                               Healthy
12
                                Healthy
     -55.18537
                 1981.99127
13
    -231.73627
                   366.51405
                                Healthy
14
    -548.78165
                 1009.50825
                                Healthy
15
    -714.31520
                                Healthy
                 -310.06349
16
      13.64383
                 1786.04919
                                Healthy
17
                               Healthy
      79.76400
                 -113.67639
18
                                Healthy
    -922.76155
                 -666.28719
19
    -667.23962
                  288.07833 Alzheimer
20
   -1036.38724
                 -771.23628
                                Healthy
21
    1828.04151
                  183.84455
                                Healthy
22
    -331.45522
                  355.26604
                                Healthy
      75.35838
23
                  623.97927 Alzheimer
24
    -584.71658
                  -60.24195
                               Healthy.
25
    -349.92694
                               Healthy
                 -876.69197
    -188.15030
26
                 -522.70932
                                Healthy
27
    -450.57621
                 -770.53553
                                Healthy
                                Healthy
28
    -198.28413
                -1044.91877
29
     727.71263
                  791.79114
                               Healthy
   -1325.71445
30
                 -559.53500
                                Healthy
                 1150.90480
                               Healthy
31
    -282.85694
32
    -437.05400
                               Healthy
                 -190.15593
33
     169.46843
                 -793.34038
                               Healthy
34
    -416.02597
                 -707.41408
                                Healthy.
35
    -407.03089
                 -706.82359
                               Healthy
36
    -648.42173
                               Healthy
                 -564.55869
37
    -539.57701
                  315.32588
                                Healthy
38
    -400.24569
                 -249.65816 Alzheimer
20
      OF 01/06
                  501 07670 Alahaiman
```