Note: Please submit your R program via Blackboard by the due date. The instruction for submitting the assignment is written on the course syllabus. Please read *AssignmentFormat.pdf* file for correct assignment format.

Grading

This assignments will be graded on correctness.

Format: 2 points

Please read AssignmentFormat.pdf file for correct assignment format.

Problem 1: 8 points

- 1. You need to read chol.txt file for this problem.
 - Use "apply" function to calculate the overall mean for each variable, except for the sex variable.
 - Use "aggregate" to calculate the sex-specific mean for each variable.
- 2. Create a new variable, chol2, which is based on the chol variable from the chol data frame. If chol is greater than the mean of chol, then chol2 = `Hi'; otherwise, chol2= `LOW'
 - Use "tapply" function to calculate the standard deviation of bmi for each chol2 category.
 - Use "tapply" function to calculate the standard deviation of bmi for each combination of sex and chol2 categories.

Problem 2: 10 points

Consider the simulated matrix, mat, with 10 rows and 20 columns.

```
> set.seed(91765)
> mat = matrix(rnorm(200), 10)
> mat[1,1] = NA
> dim(mat)
```

[1] 10 20

- 1. Calculate the median (use the median function) of each row by using the for loop
- 2. Calculate the median of each row by using the apply function

Problem 3: 10 points

When performing a two-sample t-test, the data is organized into columns. However, sometimes data is organized into rows, such as genetic data. Each row contains values for each gene, and you would like to perform a two-sample t-test for each gene.

Write a function, rowTtest, to perform a row-wise two-sample t-test. This function will take two main arguments: the first one is a matrix or data frame with each row representing values for each gene,

and with the number of columns equaling the sample size. The second argument is a character vector with length equaling sample size containing the phenotype information. The returned value is the input data frame with only those rows (genes) being statistically significant (p < (0.05/# of Genes), based on Bonforoni correction).

To test this function, you need to simulate a data frame or a matrix with dim = c(1000, 20). Make sure to generate row names, from G1 - G1000. The first 10 columns are the cases and the second 10 columns are the controls. First you will start to simulate each value following standard normal distribution. Then add 4 to G1 - G10 for only the cases. Test if you will be able to find the first 10 genes based on the function that you created. Note, you may not get those exact 10 genes, but it should be close.

Please do not use the iterative DO-loop (such as for, while, or repeat) inside the function.

Here are some hint/suggestions for problem 3:

You need to simulate a matrix with dimension of 1000 by 20. For example, this matrix will look like the following:

```
> dim(gene)
[1] 1000
           20
> head(gene)
       [,1]
                [,2]
                          [,3]
                                   [,4]
                                            [,5]
                                                      [,6]
                                                               [,7]
                                                                        [,8]
G1 3.373546 5.134965 3.113850 4.739115 2.865370 2.483627 3.381173 2.674582
G2 4.183643 5.111932 2.077745 4.386609 4.764557 4.629141 2.890578 4.951980
G3 3.164371 3.129222 5.619701 5.296397 4.570710 2.321806 1.829665 4.860004
G4 5.595281 4.210732 4.519270 3.196442 2.648306 5.179781 3.968697 5.060790
G5 4.329508 4.069396 3.944150 2.397374 1.970115 5.117655 3.739602 3.649416
G6 3.179532 2.337351 4.696418 4.933251 4.590479 2.762264 4.534430 3.869234
       [,9]
               [,10]
                                      [,12]
                                                   [,13]
                           [,11]
                                                              [,14]
                                                                         [,15]
G1 4.263703 2.782880 -0.8043316 -1.4115219 -0.93910663
                                                         0.2264537
                                                                     0.5232667
G2 3.170548 3.053771 -1.0565257
                                  1.0838697
                                             1.39366493 -0.8185942
                                                                     0.9935537
G3 2.538365 4.091410 -1.0353958
                                  1.1702224
                                             1.62581486 -0.8471526
                                                                     0.2737370
G4 5.683990 4.701351 -1.1855604
                                 0.2947545
                                             0.40900106 -1.9843326 -0.6949193
G5 2.455676 4.673422 -0.5004395 -0.5544277 -0.09255856 -0.8127788 -0.7180502
G6 3.809113 5.265553 -0.5249887 -0.4034407
                                            0.20609871
                                                        1.4616707 -0.1019895
        [,16]
                                          [,19]
                                                       [,20]
                   [,17]
                               [,18]
G1 -0.2139090
               0.8576341
                          1.0496171
                                      0.9514099 -2.07771241
G2 -0.1067233 -1.6253951
                          0.2903237
                                      0.4570987 -0.45446091
G3 -0.4645893 -0.2342783 1.2421262 -0.3586935 -0.16555991
G4 -0.6842725 -1.0326545 -0.6850857 -1.0458614
                                                 0.89765209
G5 -0.7908007 -1.1411412 -0.6677681
                                      0.3075345 -0.02948916
G6 -0.3389638 -1.5219369 0.9409138
                                     1.9943876
                                                1.85838843
```

Note that your simulated data will probably have different numbers. You need to provide the matrix gene above as the first argument of your function. You also need to create a character vector like the one below:

```
> pheno
```

```
[1] "case" "case" "case" "case" "case" "case" "case"
[8] "case" "case" "control" "control" "control" "control"
[15] "control" "control" "control" "control"
> length(pheno)
```

[1] 20

Within the function, you need to perform a t-test for each gene. For example, to get the p-value for the first gene, you can write the following:

```
> t.test(gene[1,] ~ pheno)$p.value
```

[1] 2.292288e-07

Similarly, to obtain the p-value for the 100th gene, you can do the following:

```
> t.test(gene[100,] ~ pheno)$p.value
```

[1] 0.2998438

In your function, you need to perform 1000 t-test and grab the p-value from each test and store these 1000 p-value to vector, named p. Generally we use 0.05 as a significant cutoff value. Since we are performing 1000 t-test, we need to use a Bonforoni adjusted p-value (0.05/1000). The returned value of the function is the input data frame or matrix with only those rows being significant.

Problem 4: 10 points

Calculate the two-sample t-test for each numerical variable in the chol data frame (from the chol.txt file) by the sex variable by using the apply function. The result should be a matrix that contains five columns: F.mean (mean of the numerical variable for Female), M.mean (mean of the numerical variable for Male), t (for t-statistics), df (for degrees of freedom), and p (for p-value). The result should look like the one below:

> result

```
F.mean
                    M.mean
                                                            р
      22.885417
                 22.427083
                            0.20489826 187.9970 8.378733e-01
age
                            0.75035027 188.0000 4.539818e-01
chol 202.200000 194.936842
      79.452632
                 81.294737 -0.28380022 158.5931 7.769337e-01
tg
                 63.659792 -3.42175696 186.0388 7.649414e-04
ht
      59.197187
     109.510417 137.591667 -4.09786339 183.3389 6.254578e-05
     117.125000 124.770833 -0.99609766 113.1632 3.213279e-01
sbp
dbp
      73.312500
                 75.822917 -1.32775215 189.0118 1.858613e-01
      15.231579
                 15.621053 -0.30479390 164.7068 7.609079e-01
vldl
hdl
      51.400000
                 45.957895
                            3.37529526 187.9173 8.958234e-04
ldl
     132.115789 131.252632
                            0.09293561 187.5109 9.260539e-01
       2.977679
                  3.223266 -2.39434622 189.4748 1.762516e-02
bmi
```