# LAB ASSIGNMENT 1

Submitted by: Kamya Mehra 102213026

3CO35

# QUESTION 1, 2, 3:-

```
> #Question 1
> A = 23.4
> B = 45
> C = 678
>
> #Question 2
> A
[1] 23.4
> B
[1] 45
> C
[1] 678
>
> #Question 3
> list1= list(A=23.4, B=45, C=678)
```

```
> list1[- 3]
$A
[1] 23.4

$B
[1] 45
> |
```

# QUESTION 4:-

```
Console Terminal × Background Jobs ×

R 84.4.1 · ~/ ↔

> #Question 4

> # "I am learning R"

> |
```

# QUESTION 5:-

```
Console Terminal × Background Jobs ×

R R 4.4.1 · ~/  
> #Question 5
> firstname = "MyName"
> lastname = "MySurname"
> firstname
[1] "MyName"
> lastname
[1] "MySurname"
> |
```

# QUESTION 6:-

# QUESTION 7:-

```
Console Terminal × Background Jobs ×

R R 4.4.1 · ~/ ~

> A = 23.4

> B = 45

> C = 678

> A+B+C

[1] 746.4

> A-B-C

[1] -699.6

> A*B*C

[1] 713934

> A/B/C

[1] 0.0007669617

> |
```

# QUESTION 8:-

```
Console Terminal ×
                  Background Jobs ×
                                                -6
R 4.4.1 · ~/ ≈
> #Question 8
> \exp(49)
[1] 1.907347e+21
> log(49)
[1] 3.89182
> log10(49)
[1] 1.690196
> log2(49)
[1] 5.61471
> pi
[1] 3.141593
> sqrt(49)
[1] 7
```

# QUESTION 9:-

```
Console Terminal × Background Jobs ×

R 4.4.1 · ~/ ↔

> #Question 9

> 23+(4.5*2.3)/10

[1] 24.035

> 456/12-log(90)

[1] 33.50019

> exp(5)+12/(5^6)

[1] 148.4139

> sqrt(45)*12/3

[1] 26.83282

> |
```

# LAB ASSIGNMENT 2

Submitted By: Kamya Mehra 102213026 3CO35

Here are some snippets of the output produced while working in the swirl library: -

```
Console Terminal × Background Jobs ×
R 4.4.1 · ~/ ≈
> main()
| Returning to the main menu...
| Would you like to continue with one of these lessons?
1: R Programming Vectors
2: No. Let me start something new.
Selection: 2
| Please choose a course, or type 0 to exit swirl.
1: R Programming
2: Take me to the swirl course repository!
Selection: 1
| Please choose a lesson, or type 0 to return to course menu.
1: Basic Building Blocks
                             2: Workspace and Files
                                                          3: Sequences of Numbers
                             5: Missing Values
                                                          6: Subsetting Vectors
4: Vectors
7: Matrices and Data Frames 8: Logic
                                                          9: Functions
10: lapply and sapply
                            11: vapply and tapply
                                                         12: Looking at Data
                            14: Dates and Times
13: Simulation
                                                         15: Base Graphics
Selection: 2
                                                                                         0%
| In this lesson, you'll learn how to examine your local workspace in R and begin to explore
the relationship between your workspace and the file system of your machine.
                                                                                          3%
| Because different operating systems have different conventions with regards to things like
 file paths, the outputs of these commands may vary across machines.
```

```
| Please choose a lesson, or type 0 to return to course menu.
1: Basic Building Blocks
                              2: Workspace and Files
                                                           3: Sequences of Numbers
                              5: Missing Values
                                                          6: Subsetting Vectors
4: Vectors
7: Matrices and Data Frames 8: Logic
                                                           9: Functions
10: lapply and sapply
                            11: vapply and tapply
                                                          12: Looking at Data
13: Simulation
                            14: Dates and Times
                                                         15: Base Graphics
Selection: 7
                                                                                           0%
                                                                                       | In this lesson, we'll cover matrices and data frames. Both represent 'rectangular' data
| types, meaning that they are used to store tabular data, with rows and columns.
| The main difference, as you'll see, is that matrices can only contain a single class of data,
| while data frames can consist of many different classes of data.
| Let's create a vector containing the numbers 1 through 20 using the `:` operator. Store the
 result in a variable called my_vector.
```

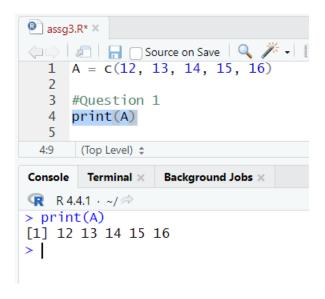
# LAB ASSIGNMENT 3

Submitted By: Kamya Mehra 102213026 3CO35

```
A = c(12,13,14,15,16)
```

This line needs to be executed with each code snippet as "A" is being used

# Question 1



```
assg3.R* ×
1 A = c(12, 13, 14, 15, 16)
  3 #Question 1
  4 # print(A)
  6 #Question 2
  7
    sum = sum(A)
  8 print (sum)
  9
 10
     #Ouaction 2
 8:12
     (Top Level) $
Console Terminal ×
               Background Jobs ×
R 4.4.1 · ~/ ≈
> sum = sum(A)
> print (sum)
[1] 70
```

```
10
      #Question 3
      product = prod(A)
 11
 12
      print (product)
 13
 11:1
      (Top Level) $
Console Terminal ×
                   Background Jobs ×
R 4.4.1 · ~/ ≈
> product = prod(A)
> print (product)
[1] 524160
> |
```

# Question 4

```
TD
  14
      #Question 4
 15
      maximum = max(A)
 16
      print(maximum)
 17
      minimum = min(A)
 18
      print(minimum)
 19
 20
 18:15
      (Top Level) $
Console Terminal ×
                  Background Jobs ×
R 4.4.1 · ~/ ≈
> maximum = max(A)
> print(maximum)
[1] 16
> minimum = min(A)
> print(minimum)
[1] 12
> |
```

```
19
  20
      #Question 5
      range\_array = range(A)
  22
      print (range_array)
  23
  24
      #Question 6
  25
     # mean_array = mean(A)
 22:20
      (Top Level) $
Console Terminal ×
                  Background Jobs ×
R 4.4.1 · ~/ ≈
> range_array = range(A)
> print (range_array)
[1] 12 16
> |
```

```
24 #Ouestion 6
      mean\_array = mean(A)
  25
  26 variance_array = var(A)
  27
      stddev\_array = sd(A)
  28
      print (mean_array)
      print (variance_array)
  29
      print (stddev_array)
  30
  31
 30:21
      (Top Level) $
       Terminal ×
Console
                  Background Jobs ×
R 4.4.1 · ~/ ≈
> mean_array = mean(A)
> variance_array = var(A)
> stddev_array = sd(A)
> print (mean_array)
[1] 14
> print (variance_array)
[1] 2.5
> print (stddev_array)
[1] 1.581139
>
```

```
assg3.R* ×
30 # print (stddev_array)
 31
 32 #Question 7
 B = sort(A)
 34 print (B)
    C = sort(A, decreasing=TRUE)
 35
 36
     print(C)
 37
 38
    #Question 8
 36:9
     (Top Level) $
Console Terminal ×
                Background Jobs ×
R 4.4.1 · ~/ ≈
> B = sort(A)
> print (B)
[1] 12 13 14 15 16
> C = sort(A, decreasing=TRUE)
> print(C)
[1] 16 15 14 13 12
```

A warning message is displayed since the 1:20 range has 20 elements, but if 3 rows and 4 columns are specified, only 12 elements can be accommodated

```
38 #Question 8
     mat = matrix(1:20, nrow=3, ncol=4)
 40 print(mat)
 41
 42 #Question 9
 43 \# CW = cbind(A,B,C)
 40:11 (Top Level) $
Console Terminal × Background Jobs ×
R 4.4.1 · ~/ ≈
> mat = matrix(1:20, nrow=3, ncol=4)
Warning message:
In matrix(1:20, nrow = 3, ncol = 4):
 data length [20] is not a sub-multiple or multiple of the number of rows [3]
> print(mat)
     [,1] [,2] [,3] [,4]
       1 4 7
2 5 8
[1,]
                  8
                      11
[2,]
[3,]
       3 6 9
                      12
> |
```

```
42 #Question 9
      CW = cbind(A,B,C)
  44 print(CW)
  45
      RW = rbind(A,B,C)
 46 print (RW)
 47
 48 #Question 10
 49 # arr_{rw} = RW[2:3,]
 43:1 (Top Level) $
Console Terminal ×
                  Background Jobs ×
R 4.4.1 · ~/ ≈
> CW = cbind(A,B,C)
> print(CW)
     A B C
[1,] 12 12 16
[2,] 13 13 15
[3,] 14 14 14
[4,] 15 15 13
[5,] 16 16 12
> RW = rbind(A,B,C)
> print (RW)
  [,1] [,2] [,3] [,4] [,5]
        13
             14
                   15
                         16
   12
В
    12
         13
              14
                   15
                         16
C
    16
         15
              14
                   13
                         12
> |
```

```
48
      #Question 10
  49
      arr_rw = RW[2:3,]
  50
      print(arr_rw)
  51
  52
      #Question 11
                 CWF 17 #Connact b
 50:14
      (Top Level) $
Console Terminal × Background Jobs ×
R 4.4.1 · ~/ ≈
> arr_rw = RW[2:3,]
> print(arr_rw)
  [,1] [,2] [,3] [,4] [,5]
    12
         13 14
                   15
                         16
C
    16
         15
               14
                    13
                          12
> |
```

# Question 11

```
#Question 11

arr_cw = CW[,1] #Correct but 1:4 gives error since 4Th column isn't present in CW matrix

print(arr_cw)

64

Framinal × Background Jobs ×

R 4.4.1 ~/

> arr_cw = CW[,1] #Correct but 1:4 gives error since 4Th column isn't present in CW matrix

print(arr_cw)

1 12 13 14 15 16

> |
```

# LAB ASSIGNMENT 4.1

# Submitted By: Kamya Mehra 102213026 3CO35

```
#Question 1
  20
       DF = data.frame(
          PatientID = c(1, 2, 3, 4), AdmDate = c("10/15/2009", "11/01/2009", "10/21/2009", "10/28/2009"),
  21
  22
          Age = c(25, 34, 28, 52),
Diabetes = c("Type 1", "Type 2", "Type 1", "Type 1"),
Status = c("Poor", "Improved", "Excellent", "Poor")
  23
  24
  25
  26
  27
       DF
  28
 27:3
       (Top Level) $
Console Terminal ×
                       Background Jobs ×
> #Question 1
> DF = data.frame(
    PatientID = c(1, 2, 3, 4),

AdmDate = c("10/15/2009", "11/01/2009", "10/21/2009", "10/28/2009"),
     Age = c(25, 34, 28, 52),
Diabetes = c("Type 1", "Type 2", "Type 1", "Type 1"),
Status = c("Poor", "Improved", "Excellent", "Poor")
+
+ )
> DF
  PatientID
                    AdmDate Age Diabetes
                                                    Status
         1 10/15/2009 25 Type 1
                                                      Poor
2
            2 11/01/2009 34
                                     Type 2 Improved
3
            3 10/21/2009 28 Type 1 Excellent
4
            4 10/28/2009 52 Type 1
                                                     Poor
>
```

```
#Question 2(A)
      Subset1 = DF[c("PatientID","Age")]
      Subset1
  31
  32
  33
     # Question 2(B)
  34
      type1_data = subset(DF, Diabetes=="Type 1")
  35
      type1_data
  36
  37
      #Question 2(C)
      poor_data = subset(DF, Status=="Poor")
  38
  39
      poor_data
  40
      nrow(poor_data)
 /11
     (Top Level) $
 40:16
Console Terminal ×
                 Background Jobs ×
R 4.4.1 · ~/ ≈
> Subset1
  PatientID Age
1
          1 25
2
          2
             34
3
          3
             28
4
          4 52
> # Question 2(B)
> type1_data = subset(DF, Diabetes=="Type 1")
> type1_data
               AdmDate Age Diabetes
  PatientID
                                        Status
1
          1 10/15/2009 25
                             Type 1
                                          Poor
3
          3 10/21/2009 28
                              Type 1 Excellent
4
          4 10/28/2009 52
                             Type 1
                                          Poor
> #Question 2(C)
> poor_data = subset(DF, Status=="Poor")
> poor_data
               AdmDate Age Diabetes Status
  PatientID
1
          1 10/15/2009 25
                             Type 1
                                       Poor
          4 10/28/2009 52
                             Type 1
                                       Poor
> nrow(poor_data)
[1] 2
> |
```

```
42 #Question 2(D)
 43
     summary(DF)
 44
 45 #Question 2(E)
     mean(DF$Age)
 46
 47
46:13
     (Top Level) $
Console Terminal × Background Jobs ×
R 4.4.1 · ~/ ≈
> #Question 2(D)
> summary(DF)
  PatientID
                AdmDate
                                    Age
                                                Diabetes
                                                                   Status
Min. :1.00
             Length:4
                                Min. :25.00 Length:4
                                                               Length:4
             Class :character
1st Qu.:1.75
                                1st Qu.:27.25
                                              Mode :character
Median :2.50
                                Median :31.00
                                Mean :34.75
Mean :2.50
3rd Qu.:3.25
                                3rd Qu.:38.50
Max. :4.00
                                Max. :52.00
> #Question 2(E)
> mean(DF$Age)
[1] 34.75
```

```
#Question 3
57
    a = c(12, 14, 16, 20)
    matrix_2d = matrix(1:10, nrow = 5, ncol = 2)
    s = c('First', 'Second', 'Third')
60
    MyList = list(
61
     Title = "My First List",
62
      Criteria = list(
63
64
        Age_Vector = a,
65
        Matrix_2D = matrix_2d,
66
        Score_Vector = s
67
68
69
    print(MyList)
70
    print(MyList$Criteria)
71
    print(MyList$Criteria$Age_Vector)
```

```
$Title
[1] "My First List"
$Criteria
$Criteria$Age_Vector
[1] 12 14 16 20
$Criteria$Matrix_2D
   [,1] [,2]
[1,]
     1
         6
[2,]
      2
            7
[3,]
      3 8
     4 9
[4,]
     5 10
[5,]
$Criteria$Score_Vector
[1] "First" "Second" "Third"
> print(MyList$Criteria)
$Age_Vector
[1] 12 14 16 20
$Matrix_2D
  [,1] [,2]
[1,]
      1 6
[2,]
      2
            7
[3,]
      3 8
[4,]
      4
          9
     5 10
[5,]
$Score_Vector
[1] "First" "Second" "Third"
> print(MyList$Criteria$Age_Vector)
[1] 12 14 16 20
> |
```

# LAB ASSIGNMENT 4.2

# Submitted By: Kamya Mehra 102213026 3CO35

# Question 1

```
21
  22
       #Question 1
  23
      vec1 \leftarrow seq(1.3, 4.9, by = 0.3)
  24
      print(vec1)
  25
      vec2 \leftarrow rep(1:4, times = 5)
      print(vec2)
  26
  27
      vec3 < - seq(14, 0, by = -2)
  28
      print(vec3)
  29
      vec4 \leftarrow rep(c(5, 12, 13, 20), each = 2)
  30
      print(vec4)
  31
 30:12
       (Top Level) $
       Terminal ×
                    Background Jobs X
Console
R 4.4.1 · ~/ ≈
> #Question 1
> \text{vec1} < - \text{seq}(1.3, 4.9, \text{by} = 0.3)
> print(vec1)
 [1] 1.3 1.6 1.9 2.2 2.5 2.8 3.1 3.4 3.7 4.0 4.3 4.6 4.9
> vec2 <- rep(1:4, times = 5)
> print(vec2)
 [1] 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4 1 2 3 4
> \text{vec3} < - \text{seq}(14, 0, \text{by} = -2)
> print(vec3)
[1] 14 12 10 8 6 4
> \text{vec4} < -\text{rep}(c(5, 12, 13, 20), \text{ each} = 2)
> print(vec4)
[1] 5 5 12 12 13 13 20 20
```

```
## Summary_by_species - aggregate(cbind(Sepal.Width, Sepal.Length) - Species, data = iris, FUN = function(x) c(mean = mean(x), sd = sd(x)))
### print(summary_by_species)
### Power of the state of the
```

```
#Question 4
  49
      data(mtcars)
  50
  51
      str(mtcars)
 52
      names (mtcars)
 53
 50:1
      (Top Level) $
                 Background Jobs ×
Console Terminal ×
R 4.4.1 · ~/ ≈
> data(mtcars)
> str(mtcars)
'data.frame':
                32 obs. of 11 variables:
$ mpg : num 21 21 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 ...
 $ cyl : num  6  6  4  6  8  6  8  4  4  6 ...
              160 160 108 258 360 ...
 $ disp: num
              110 110 93 110 175 105 245 62 95 123 ...
 $ hp : num
 $ drat: num 3.9 3.9 3.85 3.08 3.15 2.76 3.21 3.69 3.92 3.92 ...
 $ wt : num 2.62 2.88 2.32 3.21 3.44 ...
 $ qsec: num 16.5 17 18.6 19.4 17 ...
 $ vs : num 0 0 1 1 0 1 0 1 1 1 ...
 $ am : num 1 1 1 0 0 0 0 0 0 0 ...
 $ gear: num 4 4 4 3 3 3 3 4 4 4 ...
$ carb: num 4 4 1 1 2 1 4 2 2 4 ...
> names(mtcars)
 [1] "mpg" "cyl" "disp" "hp" "drat" "wt"
                                                "gsec" "vs"
                                                               "am"
                                                                      "gear" "carb"
```

```
54
      subset_cyl_geq_5 = mtcars[mtcars$cyl >= 5, ]
  55
      print(subset_cyl_geq_5)
  56
 54:1
      (Top Level) $
Console Terminal ×
                  Background Jobs ×
R 4.4.1 · ~/ ≈
> subset_cyl_geq_5 = mtcars[mtcars$cyl >= 5, ]
> print(subset_cyl_geq_5)
                      mpg cyl
                               disp hp drat
                                                  wt qsec vs am gear carb
                     21.0
                            6 160.0 110 3.90 2.620 16.46
Mazda RX4
                                                            0
                                                               1
                                                                     4
Mazda RX4 Wag
                            6 160.0 110 3.90 2.875 17.02
                                                                          4
                     21.0
                                                            0
                                                                1
                                                                     4
Hornet 4 Drive
                     21.4
                            6 258.0 110 3.08 3.215 19.44
                                                            1
                                                                0
                                                                     3
                                                                          1
Hornet Sportabout
                     18.7
                            8 360.0 175 3.15 3.440 17.02
                                                            0
                                                                0
                                                                     3
                                                                          2
Valiant
                     18.1
                            6 225.0 105 2.76 3.460 20.22
                                                            1
                                                                0
                                                                     3
                                                                          1
Duster 360
                     14.3
                            8 360.0 245 3.21 3.570 15.84
                                                            0
                                                                0
                                                                     3
                                                                          4
Merc 280
                     19.2
                            6 167.6 123 3.92 3.440 18.30
                                                                0
                                                                     4
                                                            1
                                                                          4
                     17.8
                            6 167.6 123 3.92 3.440 18.90 1
                                                                0
                                                                     4
Merc 280C
                                                                          4
                     16.4
                            8 275.8 180 3.07 4.070 17.40
                                                                0
                                                                     3
                                                                          3
Merc 450SE
                                                            0
Merc 450SL
                     17.3
                            8 275.8 180 3.07 3.730 17.60
                                                            0
                                                                0
                                                                     3
                                                                          3
                     15.2
                            8 275.8 180 3.07 3.780 18.00
                                                                0
                                                                     3
                                                                          3
Merc 450SLC
                                                            0
Cadillac Fleetwood 10.4
                            8 472.0 205 2.93 5.250 17.98
                                                            0
                                                                0
                                                                     3
                                                                          4
Lincoln Continental 10.4
                            8 460.0 215 3.00 5.424 17.82
                                                            0
                                                                0
                                                                     3
                                                                          4
Chrysler Imperial
                     14.7
                            8 440.0 230 3.23 5.345 17.42
                                                            0
                                                                0
                                                                     3
                                                                          4
                            8 318.0 150 2.76 3.520 16.87
Dodge Challenger
                     15.5
                                                            0
                                                                0
                                                                     3
                                                                          2
AMC Javelin
                     15.2
                            8 304.0 150 3.15 3.435 17.30
                                                            0
                                                                0
                                                                     3
                                                                          2
Camaro Z28
                     13.3
                            8 350.0 245 3.73 3.840 15.41
                                                            0
                                                                0
                                                                     3
                                                                          4
Pontiac Firebird
                     19.2
                            8 400.0 175 3.08 3.845 17.05
                                                            0
                                                                0
                                                                     3
                                                                          2
                                                                     5
Ford Pantera L
                     15.8
                            8 351.0 264 4.22 3.170 14.50
                                                            0
                                                                1
                                                                          4
                                                                     5
Ferrari Dino
                     19.7
                            6 145.0 175 3.62 2.770 15.50
                                                            0
                                                               1
                                                                          6
Maserati Bora
                     15.0
                            8 301.0 335 3.54 3.570 14.60
                                                            0
                                                               1
                                                                     5
                                                                          8
> 50
  57
      first_10_{cars} = mtcars[1:10, ]
  58
      print(first_10_cars)
  59
  60
      honda cars = mtcars[grep("Honda", rownames(mtcars)), ]
 57:1
       (Top Level) $
Console Terminal ×
                  Background Jobs ×
R 4.4.1 · ~/ €
> first_10_cars = mtcars[1:10, ]
> print(first_10_cars)
                    mpg cyl disp hp drat
                                               wt qsec vs am gear carb
Mazda RX4
                          6 160.0 110 3.90 2.620 16.46
                                                                       4
                   21.0
                                                         0
                                                            1
                                                                  4
Mazda RX4 Wag
                          6 160.0 110 3.90 2.875 17.02
                   21.0
                                                         0
                                                            1
                          4 108.0 93 3.85 2.320 18.61
Datsun 710
                   22.8
                                                         1
                                                                       1
                                                             1
                   21.4
                          6 258.0 110 3.08 3.215 19.44
                                                             0
                                                                  3
Hornet 4 Drive
                                                         1
                                                                       1
                          8 360.0 175 3.15 3.440 17.02
                                                                       2
Hornet Sportabout 18.7
                                                         0
                                                             0
                                                                  3
Valiant
                   18.1
                          6 225.0 105 2.76 3.460 20.22
                                                         1
                                                             0
                                                                  3
                                                                       1
Duster 360
                   14.3
                          8 360.0 245 3.21 3.570 15.84
                                                         0
                                                             0
                                                                  3
                                                                       4
                                   62 3.69 3.190 20.00
                                                                       2
Merc 240D
                   24.4
                          4 146.7
                                                         1
                                                             0
                                                                  4
Merc 230
                   22.8
                          4 140.8 95 3.92 3.150 22.90
                                                         1
                                                             0
                                                                  4
                                                                       2
                          6 167.6 123 3.92 3.440 18.30 1
                                                                  4
                                                                       4
Merc 280
                   19.2
> |
```

```
honda_cars = mtcars[grep("Honda", rownames(mtcars)), ]
print(honda_cars)

Console Terminal × Background Jobs ×

R R 4.4.1 · ~/ ~

> honda_cars = mtcars[grep("Honda", rownames(mtcars)), ]
print(honda_cars)
    mpg cyl disp hp drat wt qsec vs am gear carb
Honda Civic 30.4 4 75.7 52 4.93 1.615 18.52 1 1 4 2

>
```

# LAB ASSIGNMENT 5

# Submitted By: Kamya Mehra 102213026 3CO35

#### Question 1

```
#Question 1
 data = read.csv("daily_show_quests.csv")
 head(data, n=10)
    <db1> <chr>
                                                 <chr> <chr> <chr>
                                                 1/11/99 Acting Michael J. Fox
 1 1999 actor
                                              1/12/99 Comedy Sandra Bernhard
1/13/99 Acting Tracey Ullman
1/14/99 Acting Gillian Anderson
1/18/99 Acting David Alan Grier
1/19/99 Acting William Baldwin
 2 <u>1</u>999 Comedian
3 <u>1</u>999 television actress
 4 1999 film actress
 5 1999 actor
 6 <u>1</u>999 actor
                                               1/20/99 Musician Michael Stipe
 7 <u>1</u>999 Singer-lyricist
                                           1/21/99 Media Carmen Electra
1/25/99 Acting Matthew Lillard
1/26/99 Comedy David Cross
8 1999 model
9 1999 actor
10 <u>1</u>999 stand-up comedian
```

#### Question 2

```
# Q2: Display the column names and rename them
colnames(daily_show_guests) # Display current column names

# Rename the columns
data <- daily_show_guests %>%
   rename(
   YEAR = year,
   GoogleKnowlege_Occupation = job,
   Show = date,
   Group = category,
   Raw_Guest_List = guest_name
)
```

```
# Q3: Create a report having YEAR, Show (date), and Raw_Guest_List (guest_name)
report <- daily_show_guests %>%
  select(YEAR, Show, Raw_Guest_List)
 # Print the first 10 records of the report
head(report, 10)
 head(report, 10)
 A tibble: 10 \times 3
                 Raw_Guest_List
   YEAR Show
  <db1> <chr>
                 <chr>
   1999 1/11/99 Michael J. Fox
   1999 1/12/99 Sandra Bernhard
   1999 1/13/99 Tracey Ullman
   1999 1/14/99 Gillian Anderson
   1999 1/18/99 David Alan Grier
   1999 1/19/99 William Baldwin
   1999 1/20/99 Michael Stipe
   1999 1/21/99 Carmen Electra
   1999 1/25/99 Matthew Lillard
   1999 1/26/99 David Cross
Question 4
# Q4: Use "select" to print all records except YEAR
filtered_data <- daily_show_guests %>%
  select(-YEAR)
# Print the first 10 records of the filtered data
nead(filtered_data, 10)
head(filtered_data, 10)
<sup>4</sup> A tibble: 10 × 4
   GoogleKnowlege_Occupation Show
                                               Raw Guest List
                                      Group
                              <chr>
                                      <chr>
                                                <chr>
                                               Michael J. Fox
 1 actor
                              1/11/99 Acting
 2 Comedian
                                               Sandra Bernhard
                              1/12/99 Comedy
                              1/13/99 Acting
 3 television actress
                                               Tracey Ullman
4 film actress
                                               Gillian Anderson
                              1/14/99 Acting
5 actor
                              1/18/99 Acting
                                               David Alan Grier
                                               William Baldwin
                              1/19/99 Acting
6 actor
                              1/20/99 Musician Michael Stipe
7 Singer-lyricist
8 model
                              1/21/99 Media
                                               Carmen Electra
                              1/25/99 Acting
                                               Matthew Lillard
9 actor
                              1/26/99 Comedy
                                               David Cross
O stand-up comedian
E
Question 5
```

```
# Q5: Extract the list of people who are "actor" and whose name contains "ABC"
extracted_data <- daily_show_guests %>%
  filter(GoogleKnowlege_Occupation == "actor" & grepl("ABC", Raw_Guest_List, ignore.case = TRUE)) %>%
  select(Raw_Guest_List)
# Print the extracted data
extracted_data
Question 6
# Q6: Arrange the records in the order of the Show (date)
sorted_dates <- daily_show_guests %>%
  arrange (Show)
# Print the first 10 sorted records
nead(sorted_dates, 10)
> # Print the first 10 sorted records
> head(sorted_dates, 10)
# A tibble: 10 \times 5
    YEAR GoogleKnowlege_Occupation Show
                                        Group
                                                       Raw_Guest_List
   <db1> <chr>
                                  <chr>
                                          <chr>
                                                         <chr>
                                  1/1/07 Acting
 1 2007 actress
                                                        Meryl Streep
 2 2007 author
                                  1/1/07 Media
                                                        Sam Sheridan
 3 2008 Author
                                  1/1/08 Media
                                                        Peggy Noonan
4 2008 Consultant
                                  1/1/08 Political Aide Tim Gunn
5 2008 television host
                                  1/1/08 Media
                                                        Conan O'Brien
 6 2000 football player
                                  1/10/00 Athletics
                                                         Joe Montana
   <u>2</u>001 singer
                                  1/10/01 Musician
                                                       Vitamin C
8 <u>2</u>002 actor
                                  1/10/02 Acting
                                                        Jack Black
9 <u>2</u>005 lawyer
                                  1/10/05 Misc
                                                        John Grisham
```

1/10/06 Acting

Albert Brooks

10 <u>2</u>006 actor

>

# #question /| daily\_show\_guests <- daily\_show\_guests %>% mutate(Experience = "Work Experience")

# Print the first 10 records to verify the new column head(daily\_show\_guests, 10)

# # Print the first 10 records to verify the new column head(daily\_show\_guests, 10) A tibble: 10 $\times$ 6

	YEAR	GoogleKnowlege_Occupation	Show	Group	Raw_Guest_List	Experience
	<db7></db7>	<chr></chr>	<chr></chr>	<chr></chr>	<chr></chr>	<chr></chr>
1	<u>1</u> 999	actor	1/11/99	Acting	Michael J. Fox	Work Experience
2	1999	Comedian	1/12/99	Comedy	Sandra Bernhard	Work Experience
3	<u>1</u> 999	television actress	1/13/99	Acting	Tracey Ullman	Work Experience
4	1999	film actress	1/14/99	Acting	Gillian Anderson	Work Experience
5	1999	actor	1/18/99	Acting	David Alan Grier	Work Experience
6	1999	actor	1/19/99	Acting	William Baldwin	Work Experience
7	1999	Singer-lyricist	1/20/99	Musician	Michael Stipe	Work Experience
8	1999	model	1/21/99	Media	Carmen Electra	Work Experience
9	1999	actor	1/25/99	Acting	Matthew Lillard	Work Experience
0	1999	stand-up comedian	1/26/99	Comedy	David Cross	Work Experience

# LAB ASSIGNMENT 6

# Submitted By: Kamya Mehra 102213026 3CO35

```
    Untitled1* 
    ■ Untitled4* 
    ■ Untitled2* 
    ■ Untitled3* 
    ■ Unt
  Run
         1 library(dplyr)
           3 set.seed(123) # For reproducibility
           4 data <- data.frame(</pre>
                          Country = rep(paste0("Country", 1:20), each = 1),
Continent = rep(c("Asia", "Europe", "Africa", "Americas", "Oceania"), each = 4),
          5
          6
                          Year = rep(2001:2020, times = 1),
                         LifeExp = runif(20, 50, 80),
Pop = sample(1e6:1e7, 20, replace = TRUE),
gdpPerc = runif(20, 1000, 50000)
          8
          9
       10
      11 )
       12
      13
     //Question 1.1
unique_countries <- data %>%
                  group_by(Continent) %>%
     16
                          summarise(UniqueCountries = n_distinct(Country))
      17
       18 unique_countries
     19
    77:37 (Top Level) $
 Console Terminal × Background Jobs ×
 R 4.4.1 · ~/ ≈
> unique_countries <- data %>%
 + group_by(Continent) %>%
          summarise(UniqueCountries = n_distinct(Country))
> unique_countries
# A tibble: 5 \times 2
      Continent UniqueCountries
1 Africa
                                                                                                4
2 Americas
3 Asia
                                                                                                4
4 Europe
5 Oceania
```

```
20
      //Question 1.2
      lowest_gdp_europe <- data %>%
  21
        filter(Continent == "Europe") %>%
  22
  23
        filter(gdpPerc == min(gdpPerc))
  24
      lowest_gdp_europe
  25
 77:37
      (Top Level) $
Console
       Terminal ×
                 Background Jobs X
> lowest_gdp_europe <- data %>%
    filter(Continent == "Europe") %>%
    filter(gdpPerc == min(gdpPerc))
> lowest_gdp_europe
   Country Continent Year LifeExp
                                       Pop gdpPerc
1 Country8
              Europe 2008 76.77257 2498010 10196.86
```

```
Run
 26 //Question 1.3
 27 avg_life_exp <- data %>%
     group_by(Continent, Year) %>%
 28
 29
       summarise(AverageLifeExp = mean(LifeExp))
 30 avg_life_exp
77:37 (Top Level) $
Console Terminal × Background Jobs ×
> avg_life_exp <- data %>%
+ group_by(Continent, Year) %>%
+ summarise(AverageLifeExp = mean(LifeExp))
summarise() has grouped output by 'Continent'. You can override using the `.groups` argument.
> avg_life_exp
# A tibble: 20 \times 3
# Groups:
           Continent [5]
  Continent Year AverageLifeExp
   <chr>
             <int>
                             \langle dh 1 \rangle
1 Africa
              2009
                              66.5
2 Africa
              2010
                              63.7
3 Africa
              <u>2</u>011
                              78.7
4 Africa
              <u>2</u>012
                              63.6
5 Americas
              2013
                              70.3
6 Americas
              <u>2</u>014
                              67.2
7 Americas
              2015
                              53.1
8 Americas
                              77.0
              2016
9 Asia
              2001
                              58.6
10 Asia
              <u>2</u>002
                              73.6
              2003
11 Asia
                              62.3
12 Asia
              <u>2</u>004
                              76.5
              2005
13 Europe
                              78.2
14 Europe
              2006
                              51.4
15 Europe
              2007
                              65.8
              2008
16 Europe
                              76.8
17 Oceania
              2017
                              57.4
18 Oceania
              <u>2</u>018
                              51.3
19 Oceania
              2019
                              59.8
20 Oceania
                              78.6
              2020
```

```
//Question 1.4
     data <- data %>%
  34
        mutate(TotalGDP = Pop * gdpPerc)
  35
  36
  37
     top5_gdp_countries <- data %>%
  38
        group_by(Country) %>%
  39
        summarise(TotalGDP = sum(TotalGDP)) %>%
  40
        top_n(5, TotalGDP)
 41
     top5_gdp_countries
 77:37
       (Top Level) $
       Terminal ×
                  Background Jobs X
R 4.4.1 · ~/ ≈
> data <- data %>%
    mutate(TotalGDP = Pop * gdpPerc)
+
     Country Continent Year LifeExp
                                           Pop
                                                 gdpPerc
                                                             TotalGDP
1
                  Asia 2001 58.62733 5691993 44436.984 252935001747
    Country1
2
    Country2
                  Asia 2002 73.64915 1402857
                                               9577.580
                                                         13435974954
                  Asia 2003 62.26931 9972029 7404.089
3
                                                          73833789097
    Country3
                  Asia 2004 76.49052 6077584 33001.994 200572392690
4
    Country4
5
    Country5
                Europe 2005 78.21402 3766694 17832.307
                                                          67168844313
                Europe 2006 51.36669 8617355 33181.148 285933733953
6
    Country6
7
                Europe 2007 65.84316 1402312 16698.289 23416210879
    Country7
8
    Country8
                Europe 2008 76.77257 2498010 10196.865
                                                         25471870349
                Africa 2009 66.54305 9789812 39332.421 385057004789
9
    Country9
10 Country10
                Africa 2010 63.69844 3187228 5586.154
                                                          17804347553
                Africa 2011 78.70500 5462663 23872.173 130405636378
11 Country11
                Africa 2012 63.60002 5933182 26063.768 154641076392
12 Country12
13 Country13 Americas 2013 70.32712 7344696 30399.459 223274784948
14 Country14 Americas 2014 67.17900 6927064 17308.353 119896072260
15 Country15
              Americas 2015 53.08774 3466068 24942.039
16 Country16 Americas 2016 76.99475 4906782 47769.218 234393136816
17 Country17
               Oceania 2017 57.38263 1990323 24662.217
                                                         49085778644
               Oceania 2018 51.26179 7138629 44627.161 318576744874
18 Country18
19 Country19
               Oceania 2019 59.83762 6599061 45807.471 302286296432
20 Country20
               Oceania 2020 78.63511 3548894 30828.014 109405354391
 > top5_gdp_countries <- data %>%
     group_by(Country) %>%
     summarise(TotalGDP = sum(TotalGDP)) %>%
 +
     top_n(5, TotalGDP)
 > top5_gdp_countries
 # A tibble: 5 \times 2
   Country
                   TotalGDP
    <chr>
 1 Country1 <u>252</u>935<u>001</u>747.
 2 Country18 <u>318</u>576<u>744</u>874.
 3 Country19 302286296432.
 4 Country6 <u>285</u>933<u>733</u>953.
5 Country9 385057004789.
```

```
//Question 1.5
     high_life_exp <- data %>%
 44
 45
      filter(LifeExp >= 80)
 46
     high_life_exp
 47
77:37
      (Top Level) $
Console Terminal × Background Jobs ×
R 4.4.1 · ~/ ≈
> high_life_exp <- data %>%
   filter(LifeExp >= 80)
> high_life_exp
[1] Country Continent Year
                                   LifeExp
                                                        gdpPerc
                                                                  TotalGDP
                                              Pop
<0 rows> (or 0-length row.names)
```

```
48 //Question 1.6
 49 correlation <- data %>%
  50
        group_by(Country) %>%
        summarise(Correlation = cor(LifeExp, gdpPerc)) %>%
  51
  52
        arrange(desc(abs(Correlation))) %>%
  53
        slice(1:10)
 54 correlation
 77:37
     (Top Level) $
Console Terminal × Background Jobs ×
R 4.4.1 · ~/ A
> correlation <- data %>%
    group_by(Country) %>%
    summarise(Correlation = cor(LifeExp, gdpPerc)) %>%
    arrange(desc(abs(Correlation))) %>%
    slice(1:10)
> correlation
# A tibble: 10 \times 2
   Country Correlation
   <chr>>
                   \langle db 1 \rangle
1 Country1
                      NA
 2 Country10
                      NA
 3 Country11
                      NA
4 Country12
                      NA
 5 Country13
                      NA
6 Country14
                      NA
 7 Country15
                      NA
8 Country16
                      NA
9 Country17
                      NA
10 Country18
                      NA
```

```
//Question 1.7
     highest_avg_pop <- data %>%
        filter(Continent != "Asia") %>%
 58
 59
        group_by(Continent, Year) %>%
 60
       summarise(AvgPop = mean(Pop)) %>%
 61
       arrange(desc(AvgPop)) %>%
 62
       slice(1)
 63 highest_avg_pop
 77:37 (Top Level) $
Console Terminal × Background Jobs ×
R 4.4.1 · ~/ ≈
> highest_avg_pop <- data %>%
    filter(Continent != "Asia") %>%
    group_by(Continent, Year) %>%
   summarise(AvgPop = mean(Pop)) %>%
    arrange(desc(AvgPop)) %>%
   slice(1)
summarise()` has grouped output by 'Continent'. You can override using the `.groups` argument.
> highest_avg_pop
# A tibble: 4 \times 3
# Groups: Continent [4]
 Continent Year AvgPop
  <chr>
            <int>
1 Africa
             2009 9789812
             2013 7<u>344</u>696
2 Americas
3 Europe
             2006 8617355
             2018 7138629
4 Oceania
```

```
65
      //Question 1.8
      consistent_pop <- data %>%
  67
        group_by(Country) %>%
  68
        summarise(PopSD = sd(Pop)) %>%
  69
        arrange(PopSD) %>%
  70
        slice(1:3)
  71
     consistent_pop
  72
 77:37
       (Top Level) $
Console
       Terminal ×
                   Background Jobs X
> consistent_pop <- data %>%
    group_by(Country) %>%
    summarise(PopSD = sd(Pop)) %>%
    arrange(PopSD) %>%
    slice(1:3)
> consistent_pop
# A tibble: 3 \times 2
  Country
            PopSD
             \langle db 1 \rangle
  <chr>
1 Country1
                NA
2 Country10
                NA
3 Countrv11
```

```
//Question 1.9
         data <- data %>%
           arrange(Country, Year) %>%
    75
    76
           group_by(Country) %>%
    77
           mutate(PopChange = Pop - lag(Pop),
    78
                     LifeExpChange = LifeExp - lag(LifeExp))
         data
    80
    81
        decrease_pop_increase_lifeexp <- data %>%
    82
           filter(!is.na(PopChange) & !is.na(LifeExpChange) & PopChange < 0 & LifeExpChange > 0)
    83
         decrease_pop_increase_lifeexp
   87:1
         (Top Level) ±
  Console Terminal × Background Jobs ×
  R 4.4.1 · ~/ €
 > data <- data %>%
      arrange(Country, Year) %>%
       group_by(Country) %>%
      mutate(PopChange = Pop - lag(Pop),
               LifeExpChange = LifeExp - lag(LifeExp))
 # A tibble: 20 \times 9
                 Country [20]
 # Groups:
                  Continent Year LifeExp
                                                       Pop gdpPerc
                                                                             TotalGDP PopChange LifeExpChange
      <chr>
                   <chr>
                                <int>
                                          \langle db 1 \rangle
                                                     <int>
                                                               \langle db 1 \rangle
                                                                                 <db7>
                                                                                              <int>
   1 Country1 Asia
                                 2001
                                           58.6 5<u>691</u>993
                                                              <u>44</u>437. <u>252</u>935<u>001</u>747.
                                                                                                  NA
                                                                                                                    NA
                                           63.7 3<u>187</u>228
78.7 5<u>462</u>663
   2 Country10 Africa
                                 2010
                                                              <u>5</u>586.
                                                                        <u>17</u>804<u>347</u>553.
                                                                                                  NA
                                                                                                                     NA
     Country11 Africa
                                 2011
                                                              <u>23</u>872. <u>130</u>405<u>636</u>378.
                                                                                                  NA
                                                                                                                    NA
   4 Country12 Africa
                                 <u>2</u>012
                                           63.6 5<u>933</u>182
                                                              <u>26</u>064. <u>154</u>641<u>076</u>392.
   5 Country13 Americas
                                 <u>2</u>013
                                           70.3 7344696
                                                              <u>30</u>399. <u>223</u>274<u>784</u>948.
                                                                                                  NA
                                                                                                                     NA
                                 <u>2</u>014
                                           67.2 6927064
  6 Country14 Americas
                                                              <u>17</u>308. <u>119</u>896<u>072</u>260.
                                                                                                  NA
                                                                                                                     NA
                                 <u>2</u>015
  7 Country15 Americas
                                           53.1 3<u>466</u>068
                                                              <u>24</u>942.
                                                                        <u>86</u>450<u>802</u>018.
                                                                                                  NΑ
                                                                                                                     NA
  8 Country16 Americas
                                 <u>2</u>016
                                           77.0 4906782
                                                              <u>47</u>769. <u>234</u>393<u>136</u>816.
                                                                                                  NA
                                                                                                                     NA
  9 Country17 Oceania
                                           57.4 1990323
                                                              <u>24</u>662. <u>49</u>085<u>778</u>644.
                                 <u>2</u>017
                                                             <u>44</u>627. <u>318</u>576<u>744</u>874. 
<u>45</u>807. <u>302</u>286<u>296</u>432.
 10 Country18 Oceania
                                 <u>2</u>018
                                           51.3 7138629
                                                                                                  NA
                                                                                                                     NA
                                           59.8 6<u>599</u>061
 11 Country19 Oceania
                                 2019
                                                                                                  NΑ
                                                                                                                     NΑ
                                                               <u>9</u>578.
 12 Country2 Asia
                                 <u>2</u>002
                                           73.6 1402857
                                                                       <u>13</u>435<u>974</u>954.
                                                                                                  NA
                                                                                                                     NA
 13 Country20 Oceania
                                 <u>2</u>020
                                           78.6 3<u>548</u>894
                                                              <u>30</u>828. <u>109</u>405<u>354</u>391.
                                                                                                  NA
                                                                                                                     NA
 14 Country3 Asia
                                 <u>2</u>003
                                           62.3 9972029
                                                              <u>7</u>404. <u>73</u>833<u>789</u>097.
                                                                                                                     NA
 15 Country4
                                 2004
                                           76.5 6<u>077</u>584
                                                              <u>33</u>002. <u>200</u>572<u>392</u>690.
                                                                                                  NA
                                                                                                                     NA
                  Asia
                                           78.2 3<u>766</u>694
 16 Country5
                  Europe
                                 2005
                                                              17832.
                                                                        <u>67</u>168<u>844</u>313.
                                                                                                  NΔ
                                                                                                                    NA
17 Country6 Europe
                                 <u>2</u>006
                                           51.4 8<u>617</u>355
                                                             33181. 285933<u>733</u>953.
                                                                                                  NA
                                                                                                                     NΑ
                                         65.8 1402312 16698. 23416210879.
76.8 2498010 10197. 25471870349.
                               <u>2</u>007
18 Country7 Europe
                                                                                               NA
                                                                                                                NA
19 Country8 Europe
                               2008
                                                                                               NA
                                                                                                                NA
20 Country9 Africa
                               <u>2</u>009
                                         66.5 9789812 39332. 385057004789.
> decrease_pop_increase_lifeexp <- data %>%
    filter(!is.na(PopChange) & !is.na(LifeExpChange) & PopChange < 0 & LifeExpChange > 0)
> decrease_pop_increase_lifeexp
# A tibble: 0 \times 9
# Groups: Country [0]
# i 9 variables: Country <chr>, Continent <chr>, Year <int>, LifeExp <dbl>, Pop <int>, gdpPerc <dbl>,
     TotalGDP <dbl>, PopChange <int>, LifeExpChange <dbl>
```

```
//Question 2.1
                           library(dplyr)
                        library(ggplot2)
            88
            89
            90
            91 DataSet <- data.frame(
                                MedID = 1:10,
            92
                                Med_Name = paste("Medicine", 1:10),
Company = sample(c("CompanyA", "CompanyB", "CompanyC"), 10, replace = TRUE),
Manf_year = sample(2015:2024, 10, replace = TRUE),
Exp_date = as.Date(sample(seq(as.Date('2025/01/01'), as.Date('2030/01/01'), by="day"), 10)),
            93
            94
            95
            96
            97
                                Quantity_in_stock = sample(50:200, 10, replace = TRUE),
                                Sales = sample(1000:5000, 10, replace = TRUE)
            98
        100
        101 write.csv(DataSet, "DataSet.csv", row.names = FALSE)
      121:1
                      (Top Level) $
    Console Terminal × Background Jobs ×
    R 4.4.1 · ~/ ≈
  > library(dplyr)
  > library(ggplot2)
  > DataSet <- data.frame(
                MedID = 1:10,
             Med_Name = paste("Medicine", 1:10),

Company = sample(c("CompanyA", "CompanyB", "CompanyC"), 10, replace = TRUE),

Manf_year = sample(2015:2024, 10, replace = TRUE),

Exp_date = as.Date(sample(seq(as.Date('2025/01/01'), as.Date('2030/01/01'), by="day"), 10)),

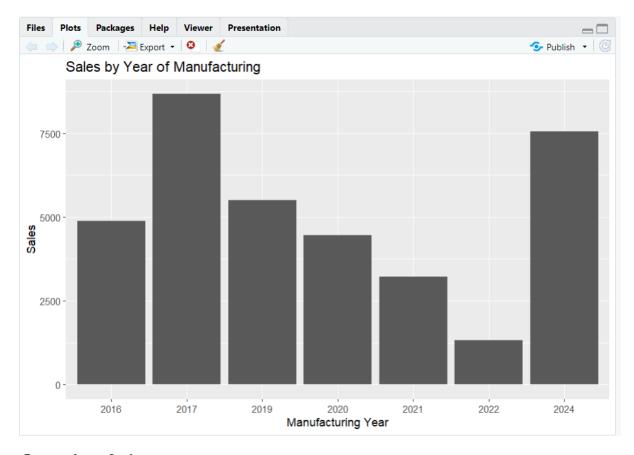
Output it is a stack as manufactor of the complace of the
               Quantity_in_stock = sample(50:200, 10, replace = TRUE),
                Sales = sample(1000:5000, 10, replace = TRUE)
 > write.csv(DataSet, "DataSet.csv", row.names = FALSE)
Question 2.2
           104
                              //Question 2.2
                                 DataSet <- read.csv("DataSet.csv")</pre>
           105
           106
                                head(DataSet, 4)
```

```
107
 108
 121:1
      (Top Level) $
Console Terminal × Background Jobs ×
R 4.4.1 · ~/ ≈
                                         Exp_date Quantity_in_stock Sales
 MedID
          Med_Name Company Manf_year
      1 Medicine 1 CompanyC
                              2017 2028-10-17
                                                                 103
      2 Medicine 2 CompanyB
                                 2021 2025-08-24
                                                                 159 3207
3
      3 Medicine 3 CompanyC
                                 2020 2026-09-02
                                                                 143 4462
4
      4 Medicine 4 CompanyA
                                 2024 2025-11-26
                                                                 128 3202
```

```
T08
 109
      //Question 2.3
 110 tail(DataSet, 4)
 121:1
     (Top Level) $
Console Terminal ×
                 Background Jobs ×
R 4.4.1 · ~/ ♠
> tail(DataSet, 4)
   MedID
            Med_Name Company Manf_year
                                           Exp_date Quantity_in_stock Sales
                                   2022 2025-07-31
          Medicine 7 CompanyC
                                                                   156
                                                                       1325
                                    2017 2026-11-17
8
       8 Medicine 8 CompanyB
                                                                   184
                                                                       4855
                                    2024 2029-12-14
                                                                   200 4351
9
      9 Medicine 9 CompanyA
                                    2016 2027-03-25
                                                                   151 4871
10
      10 Medicine 10 CompanyC
```

### Question 2.4

```
//Question 2.5
ggplot(DataSet, aes(x = as.factor(Manf_year), y = Sales)) +
  geom_bar(stat="identity") +
  labs(title="Sales by Year of Manufacturing", x="Manufacturing Year", y="Sales")
```

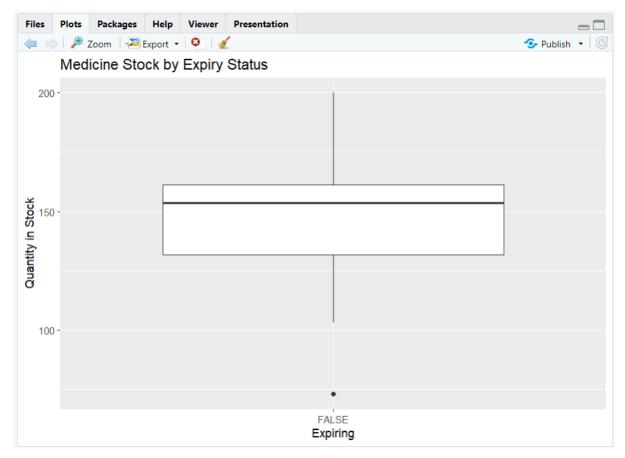


```
123 //Question 2.6
 124 DataSet %>%
 125
        group_by(Company) %>%
 126
        summarise(Count = n()) %>%
       filter(Count > 1)
 127
> DataSet %>%
 group_by(Company) %>%
   summarise(Count = n()) %>%
+ filter(Count > 1)
# A tibble: 3 \times 2
 Company Count
  <chr>
         <int>
1 CompanyA
              2
2 CompanyB
3 CompanyC
```

```
129 //Question 2.7
130 unique(DataSet$Med_Name)
131

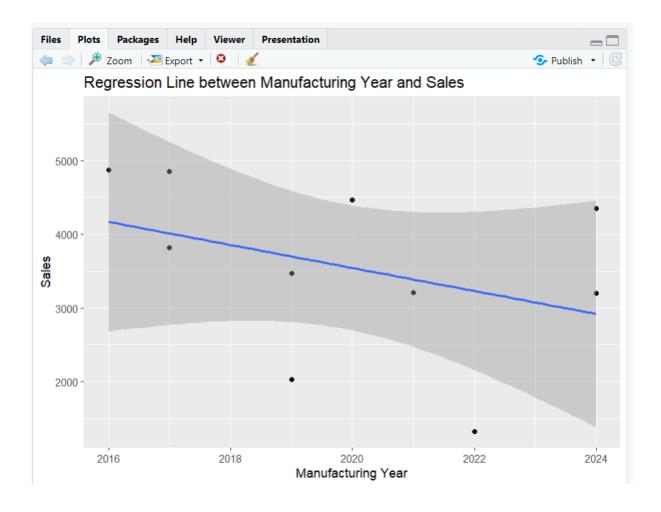
> unique(DataSet$Med_Name)
[1] "Medicine 1" "Medicine 2" "Medicine 3" "Medicine 4" "Medicine 5" "Medicine 6" "Medicine 7"
[8] "Medicine 8" "Medicine 9" "Medicine 10"
```

```
//Question 2.8
DataSet %>%
134 mutate(Expiring = Exp_date < Sys.Date()) %>%
135 ggplot(aes(x = as.factor(Expiring), y = Quantity_in_stock)) +
136 geom_boxplot() +
137 labs(title="Medicine Stock by Expiry Status", x="Expiring", y="Quantity in Stock")
```



# **Question 2.9**

```
//Question 2.9
140 mean(DataSet$Quantity_in_stock)
> mean(DataSet$Quantity_in_stock)
[1] 145.9
```



# LAB ASSIGNMENT 7

## Submitted By: Kamya Mehra 102213026 3CO35

```
//Question 1
library(dplyr)
set.seed(123)
SUB1 <- sample(50:100, 20, replace=TRUE)
SUB2 <- sample(50:100, 20, replace=TRUE)
SUB3 <- sample(50:100, 20, replace=TRUE)
MARKS <- data.frame(SUB1, SUB2, SUB3)
# a, b
MARKS$Total <- apply(MARKS, 1, sum)
# C
st.err <- function(x) {
  return(sd(x) / sqrt(length(x)))
standard_errors <- apply(MARKS[,1:3], 2, st.err)</pre>
# d
MARKS[,1:3] \leftarrow MARKS[,1:3] + 0.25
standard_errors
      SUB1 SUB2 SUB3 Total
    80.25 57.25 72.25
                           209
    64.25 75.25 76.25
                           215
   100.25 56.25 56.25
                          212
    63.25 91.25 76.25
                          230
5
    52.25 58.25 81.25
                          191
    91.25 68.25 87.25
6
                           246
    99.25 85.25 74.25
                           258
    92.25 63.25 83.25
                           238
9
    86.25 66.25 78.25
                           230
10
    63.25 92.25 54.25
                           209
11
    74.25 88.25 57.25
                           219
    75.25 61.25 61.25
12
                          197
13 76.25 64.25 62.25
                           202
   54.25 81.25 67.25
14
                           202
15 100.25 91.25 82.25
                           273
16 76.25 94.25 76.25
                           246
    77.25 56.25 74.25
17
                          207
    58.25 58.25 87.25
                          203
    78.25 90.25 70.25
19
                          238
20 84.25 59.25 64.25
                          207
> standard_errors
     SUB1
              SUB2
                        SUB3
3.322095 3.265671 2.272577
```

```
25 //Question 2
 26 V1 <- MARKS$SUB1
 27 V2 <- MARKS$SUB2
 28 V3 <- MARKS$SUB3
 29
 30
 31 vectors <- list(V1, V2, V3)</pre>
 32 sums <- lapply(vectors, sum)</pre>
 33 sums
> V1 <- MARKS$SUB1
> V2 <- MARKS$SUB2
> V3 <- MARKS$SUB3
> vectors <- list(V1, V2, V3)</pre>
> sums <- lapply(vectors, sum)</pre>
> sums
[[1]]
[1] 1547
[[2]]
[1] 1458
[[3]]
[1] 1442
```

# **Question 3**

```
35
36 //Question 3|
37 TOTAL_SUM <- sapply(vectors, sum)
38 TOTAL_SUM

> TOTAL_SUM <- sapply(vectors, sum)
> TOTAL_SUM

[1] 1547 1458 1442
```

```
40 //Question 4
41 squared_values <- sapply(vectors, function(x) x^2)
42 squared_values
43
```

```
> squared_values <- sapply(vectors, function(x) x/
> squared_values
           [,1]
                    [,2]
                              [,3]
 [1,]
       6440.062 3277.562 5220.062
       4128.062 5662.562 5814.062
 [2,]
 [3,] 10050.062 3164.062 3164.062
 [4,]
       4000.562 8326.562 5814.062
       2730.062 3393.062 6601.562
 [5,]
 [6,]
       8326.562 4658.062 7612.562
       9850.562 7267.562 5513.062
 [7,]
       8510.062 4000.562 6930.562
 [8,]
       7439.062 4389.062 6123.062
 [9,]
[10,]
       4000.562 8510.062 2943.062
[11,]
       5513.062 7788.062 3277.562
[12,]
       5662.562 3751.562 3751.562
       5814.062 4128.062 3875.062
[13,]
      2943.062 6601.562 4522.562
[14,]
[15,] 10050.062 8326.562 6765.062
      5814.062 8883.062 5814.062
[16,]
       5967.562 3164.062 5513.062
[17,]
[18,]
       3393.062 3393.062 7612.562
[19,]
       6123.062 8145.062 4935.062
[20,]
      7098.062 3510.562 4128.062
```

```
//Question 5
 44
 45
     I <- rep(1:4, each=5)
 46
    MARKS$I <- I
 47
     Ι
 48
 49
 50
     mean_by_index <- tapply(MARKS$SUB1, MARKS$I, mean)</pre>
 51
     sd_by_index <- tapply(MARKS$SUB1, MARKS$I, sd)</pre>
 52
     mean_by_index
 53 sd_by_index
54
> I <- rep(1:4, each=5)
> MARKS$I <- I
> I
 [1] 1 1 1 1 1 2 2 2 2 2 3 3 3 3 3 4 4 4 4 4
> mean_by_index <- tapply(MARKS$SUB1, MARKS$I, mean)</pre>
> sd_by_index <- tapply(MARKS$SUB1, MARKS$I, sd)</pre>
> mean_by_index
    1
          2
72.05 86.45 76.05 74.85
> sd_by_index
        1
                   2
18.660118 13.773162 16.315637 9.787747
```

```
57 //Question 6
  58 * f \leftarrow function(x, y) {
  return(x / y)
  60 - }
  61 result <- mapply(f, V1, V2)
  62 result
 62
10.000110 13.// JUL 10.JLJUJ/ 3./0// 4/
> f <- function(x, y) {</pre>
   return(x / y)
+ }
> result <- mapply(f, V1, V2)</pre>
> result
  [1] \ \ 1.4017467 \ \ 0.8538206 \ \ 1.7822222 \ \ 0.6931507 \ \ 0.8969957 \ \ 1.3369963 \ \ 1.1642229 \ \ 1.4584980 \ \ 1.3018868 
 [10] \ \ 0.6856369 \ \ 0.8413598 \ \ 1.2285714 \ \ 1.1867704 \ \ 0.6676923 \ \ 1.0986301 \ \ 0.8090186 \ \ 1.3733333 \ \ 1.00000000 
[19] 0.8670360 1.4219409
```

```
65 //Question 7
    66
         data("Seatbelts")
   67
   68
   69
        seatbelts_sum <- apply(Seatbelts, 2, sum)
         seatbelts_mean <- apply(Seatbelts, 2, mean)</pre>
         seatbelts_sd <- apply(Seatbelts, 2, sd)</pre>
   71
   72
   73
        seatbelts_sum
   74
        seatbelts_mean
   75
        seatbelts_sd
   76
> data("Seatbelts")
> seatbelts_sum <- apply(Seatbelts, 2, sum)
> seatbelts_mean <- apply(Seatbelts, 2, mean)
> seatbelts_sd <- apply(Seatbelts, 2, sd)</pre>
> seatbelts_sum
DriversKilled
                    drivers
                                   front
                                                 rear
                                                                kms
                                                                      PetrolPrice
                                                                                      VanKilled
 2.357800e+04 3.206990e+05 1.607460e+05 7.703200e+04 2.878772e+06 1.989581e+01 1.739000e+03
         law
 2.300000e+01
> seatbelts_mean
DriversKilled
                    drivers
                                   front
                                                  rear
                                                                kms
                                                                      PetrolPrice
                                                                                      VanKilled
 1.228021e+02 1.670307e+03 8.372188e+02 4.012083e+02 1.499360e+04 1.036240e-01 9.057292e+00
         law
 1.197917e-01
> seatbelts sd
                                                                      PetrolPrice
DriversKilled
                    drivers
                                   front
                                                                                      VanKilled
                                                 rear
                                                                kms
 2.537989e+01 2.896110e+02 1.750990e+02 8.310221e+01 2.938049e+03 1.217583e-02 3.636903e+00
          law
 3.255667e-01
> |
```