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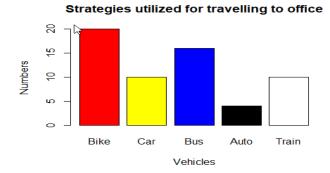
ITA0448

ASSIGNMENT - 5

1. i) Describe how histogram charts are created in R. Create a histogram chart for the below given age attribute.

Age: 5,45,23,30,33,32,34,35,42,41,28,29

ii) Create a 3D Pie Chart for the dataset "political Knowledge" with suitable labels and colour.



col = c("red", "orange", "green"), explode = 0.1)

INPUT

i)age <- c(5, 45, 23, 30, 33, 32, 34, 35, 42, 41, 28, 29)
hist(age, main = "Age Distribution", xlab = "Age", ylab = "Frequency")
ii)library(plotrix)
political_knowledge <- c("Low", "Low", "Medium", "Medium", "High", "High")
pie3D(table(political_knowledge), main = "Political Knowledge",</pre>

2. Write R code for the below output Figure 1 shows Bike is assigned red, car is assigned yellow, bus is assigned blue, auto is assigned black, and train is assigned white. Mention the parameters used in the below barchart.

```
# Load the mtcars dataset
data(mtcars)
# Create a boxplot graph of mpg by cyl
boxplot(mpg ~ cyl, data = mtcars, xlab = "Number of Cylinders", ylab = "Miles per Gallon",
    main = "Relationship between MPG and Cylinders")
# Define a recursive function to find the sum of n natural numbers
sum natural <- function(n) {</pre>
 if (n == 0) { # Base case: sum of 0 numbers is 0
  return(0)
 } else { # Recursive case: sum of n numbers is n + sum of (n-1) numbers
  return(n + sum natural(n-1))
}
}
# Test the function with n = 10
sum_natural(10) # Output: 55
3. Create a Boxplot graph for the relation between "mpg"(miles per galloon) and
"cyl(number of Cylinders) for the dataset "mtcars" available in R Environment.
v)Write R program to find the sum of Natural Numbers using Recursion
INPUT
# Load the mtcars dataset
data(mtcars)
# Create a boxplot graph of mpg by cyl
boxplot(mpg ~ cyl, data = mtcars, xlab = "Number of Cylinders", ylab = "Miles per Gallon",
    main = "Relationship between MPG and Cylinders")
# Define a recursive function to find the sum of n natural numbers
sum_natural <- function(n) {</pre>
```

```
if (n == 0) { # Base case: sum of 0 numbers is 0
  return(0)
 } else { # Recursive case: sum of n numbers is n + sum of (n-1) numbers
  return(n + sum_natural(n-1))
 }
}
# Test the function with n = 10
sum_natural(10)
OUTPUT
55
4. a. Melt 'airquality' data set which inbuild dataset in 'R' and display as a long – format
data?
b. Melt air quality data and specify month and day to be "ID variables"?
c. Cast the molten 'airquality' data set.
d. Use cast function appropriately and compute the average of Ozone, Solar, Wind
and temperature per month?
e. Create a boxplot for ozone reading of 'airquality' dataset. Add title, label and color.
INPUT
a)library(reshape2)
data(mtcars)
airquality_melt <- melt(airquality)
b)airquality_melt <- melt(airquality, id.vars = c("Month", "Day"))
head(airquality_melt)
c)airquality_cast <- dcast(airquality_melt, Month ~ variable)
head(airquality_cast)
```

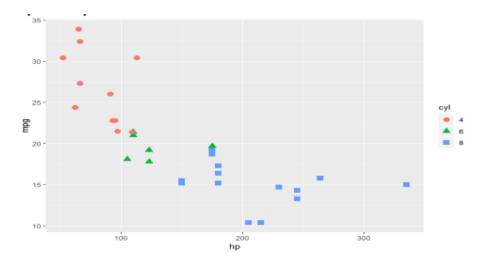
```
d)library(plyr)
airquality_avg <- cast(airquality_melt, Month ~ variable, mean)
head(airquality_avg)
e)library(ggplot2)
ggplot(data = airquality, aes(x = "", y = Ozone)) +
 geom_boxplot(fill = "lightblue", color = "blue") +
 labs(title = "Boxplot of Ozone Readings", y = "Ozone Reading")
head(airquality melt)
5. a. Write a program for creating a pie-chart in R using the input vector (21,62,10,53).
Provide labels for the chart as 'London', 'New York', 'Singapore', 'Mumbai'. Add a
title to the chart as 'city pie-chart' and add a legend at the top right corner of the chart.
b. Using linear regression analysis establish a relationship between height and weight
of a person using the input vector given below.
# Values of height
151, 174, 138, 186, 128, 136, 179, 163, 152, 131
# Values of weight.
63, 81, 56, 91, 47, 57, 76, 72, 62, 48 Predict the weight of a person with height 170.
c. Visualize the regression graphically.
d. Call 'mtcars' which is built in dataset in 'R' and plot distribution of mpg feature.
Make x axis range from 10 to 35 and plot title as "More trends in 70's Vehicles".
e. Find statistical summary of the 'mtcars' dataset.
INPUT
# Input vector
> # Input vector
> input_vector <- c(21, 62, 10, 53)
> # Labels for chart
> labels <- c("London", "New York", "Singapore", "Mumbai")</pre>
> # Create pie chart
```

> pie(input_vector, labels = labels, main = "City Pie Chart")
> # Add legend

- 6. There is a popular built-in data set in R called "mtcars" (Motor Trend Car Road Tests), which is retrieved from the 1974 Motor Trend US Magazine.
- (i) Find the dimension of the data set & amp; Give the statistical summary of the features.
- (ii)Create correlation matrix between mpg vs all other features and print the high 3 correlated Features(both +ve and -ve)
- (iii)Plot the Box plot for "mpg" group by "cyl" feature.
- (iv)Create a scatter plot graph for the relation between "mpg"(miles per gallon) and "hp"(horse power) group by cyl(number ofcylinder)

 Sample Output:
- (v) Generate a multiple regression model to establish the relationship between "mpg" as a response variable with "disp","hp" and "wt" as predictor variables. Plot the

regression line. Find the MSE of the model.



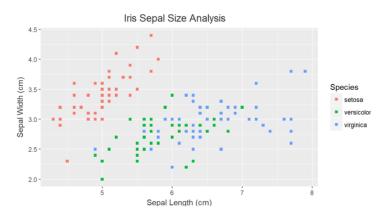
INPUT

```
data(mtcars)
> # Dimensions of the data set
  dim(mtcars)
[1] 32 11
> # Statistical summary of the features
> summary(mtcars)
                                          disp
                                                                              drat
      mpg
                                                             hp
                           :4.000
                                                               52.0
 Min.
         :10.40
                   Min.
                                     Min.
                                             : 71.1
                                                       Min.
                                                                         Min.
.760
                   1st Qu.:4.000
                                     1st Qu.:120.8
 1st Qu.:15.43
                                                       1st Qu.: 96.5
                                                                         1st Qu.:3
.080
                   Median :6.000
                                     Median :196.3
                                                       Median :123.0
 Median :19.20
                                                                         Median:3
.695
 Mean
         :20.09
                   Mean
                           :6.188
                                     Mean
                                             :230.7
                                                       Mean
                                                               :146.7
                                                                         Mean
                                                                                 :3
.597
 3rd Qu.:22.80
                   3rd Qu.:8.000
                                     3rd Qu.:326.0
                                                       3rd Qu.:180.0
                                                                         3rd Qu.:3
.920
         :33.90
                           :8.000
                                             :472.0
                                                               :335.0
 Max.
                   Max.
                                     Max.
                                                       Max.
                                                                         Max.
                                                                                 :4
.930
       wt
                        qsec
                                                                            gear
                                             :0.0000
                                                                :0.0000
                           :14.50
 Min.
         :1.513
                   Min.
                                     Min.
                                                        Min.
                                                                           Min.
:3.000
                   1st Qu.:16.89
                                     1st Qu.:0.0000
                                                        1st Qu.:0.0000
 1st Qu.:2.581
                                                                           1st Qu.
:3.000
                                                        Median :0.0000
                   Median :17.71
                                     Median :0.0000
 Median :3.325
                                                                           Median
:4.000
 Mean
         :3.217
                   Mean
                           :17.85
                                     Mean
                                             :0.4375
                                                        Mean
                                                                :0.4062
                                                                           Mean
:3.688
                   3rd Qu.:18.90
                                     3rd Qu.:1.0000
                                                        3rd Qu.:1.0000
                                                                           3rd Qu.
 3rd Qu.:3.610
:4.000
         :5.424
                           :22.90
                                     Max.
                                             :1.0000
                                                                :1.0000
                                                                           Max.
 Max.
                   Max.
                                                        Max.
:5.000
      carb
         :1.000
 Min.
 1st Qu.:2.000
 Median :2.000
         :2.812
 Mean
 3rd Qu.:4.000
         :8.000
 Max.
> # Correlation matrix between mpg and other features
> correlations <- cor(mtcars)</pre>
> # Sort the correlations by the absolute values
> correlations_sorted <- sort(abs(correlations[,'mpg']), decreasing = TRUE</pre>
> # Print the top three highest correlated features
> correlations_sorted[1:3]
```

mpg wt cyl 1.0000000 0.8676594 0.8521620

- 7. (i) Use melt and cast function to find mean of numeric data in dataset based on Species group.
- (ii) Generate a suitable plot which summaries statistical parameter of Sepal.Width based on Species group
- (iii) Generate scatter plot between Sepal.Length vs Sepal.Width grouped by Specias.

Sample Output:



INPUT
a)# Load the reshape2 package
library(reshape2)

Melt the iris data set
melted_iris <- melt(iris, id.vars = 'Species')</pre>

Cast the molten data set to find the mean of the numeric variables based on the species group mean_iris <- dcast(melted_iris, Species ~ variable, mean)

View the mean_iris data frame

mean_iris

b)# Load the ggplot2 package

library(ggplot2)

Create a box plot of Sepal.Width grouped by Species

ggplot(iris, aes(x = Species, y = Sepal.Width)) +

geom_boxplot() +

ggtitle("Box plot of Sepal.Width by Species")

c)# Create a scatter plot of Sepal.Length vs Sepal.Width grouped by Species

ggplot(iris, aes(x = Sepal.Length, y = Sepal.Width, color = Species)) +

geom_point() +

ggtitle("Scatter plot of Sepal.Length vs Sepal.Width by Species")

8. A) Heights(in cm) of father and son are given as follows

Father(X): 150 152 155 157 160 161 164 165

Son (Y): 154 156 158 159 160 162 161 164

Fit a regression line parameters to predict the height of son given the height of father.

Write R code for same.

B) Fit a regression line parameters distribution with the following data

X	0	1	2	3	4	5
Y	142	156	69	27	5	1

INPUT

Heights of father and son

father_height <- c(150, 152, 155, 157, 160, 161, 164, 165)

son_height <- c(154, 156, 158, 159, 160, 162, 161, 164)

Fit a regression line to predict the height of son given the height of father

regression_model <- Im(son_height ~ father_height)

Print the summary of the regression model summary(regression_model)

B)# Fit a regression line to the data regression_model <- Im(y ~ x, data = data_df)

Print the summary of the regression model summary(regression_model)