Demonstrate the steps for installation of R and R Studio.

Perform the following:

e<-2^3 e2<-2\*\*3

- a) Assign different type of values to variables and display the type of variable. Assign different types such as Double, Integer, Logical, Complex and Character and understand the difference between each data type.
- b) Demonstrate Arithmetic and Logical Operations with simple examples.
- c) Demonstrate generation of sequences and creation of vectors.
- d) Demonstrate Creation of Matrices
- e) Demonstrate the Creation of Matrices from Vectors using Binding Function.
- f) Demonstrate element extraction from vectors, matrices and arrays

rm(list=ls())#Clears the Environment #a) variables and display the type of variable var1<-123 typeof(var1) var2<-123L typeof(var2) var3<-"apple" typeof(var3) var3<-TRUE var4<-FALSE var5<-T var6<-F typeof(var3) typeof(var4) typeof(var5) typeof(var6) var7<- 2+3i typeof(var7) #b)Arithmetic Operations - Inspect the Variables in the workspace a<-1 b<-3 c<-1 sum<-a+b difference<-b-a difference1<-a-b

```
x <- 10/3 #This will store 3.33 in x - FLoating Point Division
10%/%3#Integer DIvision
10% % 3#Remainder after division - Modulus
x1 < -2L
x2 < -2.5
x <- x1+x2 #x will have 4.5 as result
#b) Logical Operations
status1<-a>=b
status2<-a<=b
status3<-!FALSE
status4<-!0
a=6
b=7
a==b
#exponentiaTION
exp(2)
#Arithmetic Operation on Vectors
vec<-seq( from=1, to=9, by=2)
vec1<-1:5
vec2<-6:10
sum_vec<-vec1+vec2
dif_vec<-vec2-vec1
prod_vec<-vec1*vec2#element by element multiplication
#Algebraic operations on vectors act element-wise
x < -c(1,2,3)
y < -c(4,5,6)
x*y # [1] 4 10 18
x+y # [1] 5 7 9
y^x # [1] 4 25 216
c(1,2,3,4) + c(1,2) # [1] 2 4 4 6 #c stands for combine
(1:10)^c(1,2) # [1] 1 4 3 16 5 36 7 64 9 100
2+c(1,2,3) # [1] 3 4 5
2*c(1,2,3) # [1] 2 4 6
(1:10)^2 # [1] 1 4 9 16 25 36 49 64 81 100
v <- c(3,1,TRUE,2+3i)
t < -c(4,1,FALSE,2+3i)
print(v&t) #[1] TRUE TRUE FALSE TRUE
#Create a matrix by binding vectors as rows
a < -c(1,2,3,4,5)
```

```
b < -c(6,7,8,9,10)
rb<-rbind(a,b)
rb
#Create a matrix by binding vectors as columns
cb<-cbind(a,b)
#Accessing Vcetor elements
a[1]
a[1:3]
#CReating Matrices & accessing Values
A <- matrix(1:6, nrow=2, ncol=3,byrow=TRUE)
B <- matrix(2:7, nrow=2, ncol=3,byrow=TRUE)
A
A[1,3]<-0#CHanges the 1st row 3rd column value to 0
A[, 2:3] #References all the rows and columns from 2 to 3
#CReating an Array out of a vector or matrix
arr_a=as.array(a)
arr_a
d<-c(1,2,"Apple",5,6)
list_d=as.list(d)
list d
arr_a[1]
arr_a[1:3]
list_d[1]
list_d[3]
```

Assess the Financial Statement of an Organization being supplied with 2 vectors of data: Monthly Revenue and Monthly Expenses for the Financial Year. You can create your own sample data vector for this experiment) Calculate the following financial metrics:

- a. Profit for each month.
- b. Profit after tax for each month (Tax Rate is 30%).
- c. Profit margin for each month equals to profit after tax divided by revenue.
- d. Good Months where the profit after tax was greater than the mean for the year.
- e. Bad Months where the profit after tax was less than the mean for the year.
- f. The best month where the profit after tax was max for the year.
- g. The worst month where the profit after tax was min for the year.

Note:

a. All Results need to be presented as vectors

- b. Results for Dollar values need to be calculated with \$0.01 precision, but need to be presented in Units of \$1000 (i.e 1k) with no decimal points
- c. Results for the profit margin ratio need to be presented in units of % with no decimal point.
- d. It is okay for tax to be negative for any given month (deferred tax asset)
- e. Generate CSV file for the data.

```
rm(list=ls())
revenue <- c(14574.49, 7606.46, 8611.41, 9175.41, 8058.65, 8105.44, 11496.28, 9766.09, 10305.32, 14379.96,
10713.97, 15433.50)
expenses <- c(12051.82, 5695.07, 12319.20, 12089.72, 8658.57, 840.20, 3285.73, 5821.12, 6976.93, 16618.61,
10054.37, 3803.96)
months<-
c("January", "February", "March", "April", "May", "June", "July", "August", "September", "October", "November", "Decem
ber")
profit <- revenue - expenses</pre>
profit
#Profit after tax for each month
tax<-round(profit*0.3)
tax
profit.after.tax<-(profit-tax)</pre>
profit.after.tax
#Profit margin for each month
profit.margin<-round(profit.after.tax/revenue,2)*100
profit.margin
#Good Months
mean_pat<- mean(profit.after.tax)</pre>
mean_pat
good.months<-profit.after.tax>mean_pat
good.months
#Bad Months
bad.months <-! good.months
bad.months
#The best month
best.month<-profit.after.tax == max(profit.after.tax)
best.month
#The worst month
worst.month<-profit.after.tax == min(profit.after.tax)
worst.month
#Result Presentation according to requirement
revenue.1000<-round(revenue/1000,0)
```

```
revenue.1000
expenses.1000<-round(expenses/1000,0)
expenses.1000
profit.1000<-round(profit/1000,0)
profit.1000
profit.after.tax.1000<-round(profit.after.tax/1000,0)
profit.after.tax.1000
```

Develop a program to create two 3 X 3 matrices A and B and perform the following operations a) Transpose of the matrix b) addition c) subtraction d) multiplication

```
A <- matrix(1:9, nrow=3, ncol=3,byrow=TRUE)#Create MAtrix A
B <- matrix(2:10, nrow=3, ncol=3,byrow=TRUE)#Create MAtrix B
A
В
AT<-t(A)#Transpose of Matrix using t()
AT
mat_sum<-A+B#Add Matrix
mat_sum
mat_dif<-A-B #get difference of Matrix A and B
mat dif
mat_dif1<-B-A
mat_dif1
mat_prod<-A*B#Vector Multiplication
mat_prod
mat_prod1<-A%*%B#Matrix Multiplication
mat_prod1
```

# **Program-4**

Develop a program to find the factorial of given number using recursive function calls.

```
#Method 1:
nfact2<-function(n){
  if(n==1){
    cat("Called nfact2(1)\n")
    return(1)
  }else{
    cat("called nfact2(",n,")\n",sep="")
    return(n*nfact2(n-1))
  }
}
nfact2(6)</pre>
```

Develop an R Program using functions to find all the prime numbers up to a specified number by the method of Sieve of Eratosthenes.

```
primesieve<- function(sieved,unsieved){
  p<-unsieved[1]
  n<-unsieved[length(unsieved)]
  if(p^2 >n){
    return(c(sieved, unsieved))
  }else{
    unsieved<-unsieved[unsieved% % p!=0]
    sieved<-c(sieved,p)
    return(primesieve(sieved,unsieved))
  }
}
primesieve(c(),2:200)</pre>
```

## **Program-6**

The built-in data set mammals contain data on body weight versus brain weight. Develop R commands to:

- a) Find the Pearson and Spearman correlation coefficients. Are they similar?
- b) Plot the data using the plot command.
- c) Plot the logarithm (log) of each variable and see if that makes a difference.

```
rm(list=ls())
install.packages("MASS")
library(MASS)
# Part a: Find the Pearson and Spearman correlation coefficients. Are they similar?
data(mammals)
head(mammals) #it is optional. To view the contents of the "mammals" dataset
pearson_corr <- cor(mammals$brain, mammals$body, method = "pearson")</pre>
spearman corr <- cor(mammals$brain, mammals$body, method = "spearman")
print(paste("Pearson correlation coefficient:", pearson_corr))
print(paste("Spearman correlation coefficient:", spearman_corr))
# Part b: Plot the data using the plot command
plot(mammals$body, mammals$brain, xlab = "Body Weight", ylab = "Brain Weight", main = "Body Weight vs. Brain
Weight")
# Part c: Plot the logarithm (log) of each variable and see if that makes a difference
plot(log(mammals$body), log(mammals$brain), xlab = "log(Body Weight)", ylab = "log(Brain Weight)", main =
"log(Body Weight) vs. log(Brain Weight)")
```

Write an R program to create a Data Frame with following details and do the following operations.

itemCode	itemCategory	itemPrice
1001	Electronics	700
1002	<b>Desktop Supplies</b>	300
1003	Office Supplies	350
1004	USB	400
1005	CD Drive	800

- a) Subset the Data frame and display the details of only those items whose price is greater than or equal to 350.
- b) Subset the Data frame and display only the items where the category is either "Office Supplies" or "Desktop Supplies"
- c) Create another Data Frame called "item-details" with three different fields itemCode, ItemQtyonHand and ItemReorderLvl and merge the two frames

```
rm(list=ls())
item1<- c('1001','Electronics','700')
item2<- c('1002','Desktop Supplies','300')
item3<- c('1003','Office Supplies','350')
item4<- c('1004','USB','400')
item5<- c('1005','CD Drive','800')
itemList = list(item1,item2,item3,item4,item5)
columns<-c('itemCode','itemCategory','itemPrice')
itemData = data.frame(matrix(nrow = 0, ncol = length(columns)))
colnames(itemData) = columns
View(itemData)
for(item in itemList){
itemData[nrow(itemData)+1,]<-item
}
View(itemData)
#Create Required subsets
sub_itemData_gt350<-subset(itemData, itemData$itemPrice >= 350)
View(sub itemData gt350)
sub_itemData_supplies<-subset(itemData, (itemData$itemCategory=='Office Supplies')
|(itemData$itemCategory=='Desktop Supplies'))
View(sub itemData supplies)
#CReate Seond Dataframe
itemDetailsColumn<-c('itemCode','ItemQtyonHand','ItemReorderLvl')
itemDetails = data.frame(matrix(nrow = 0, ncol = length(itemDetailsColumn)))
colnames(itemDetails) = itemDetailsColumn
```

```
View(itemDetails)

itemDetails1<- c('1001','15','2')

itemDetails2<- c('1002','10','2')

itemDetails3<- c('1003','20','4')

itemDetails4<- c('1004','12','5')

itemDetails5<- c('1005','5','1')

itemDetailsList = list(itemDetails1,itemDetails2,itemDetails3,itemDetails4,itemDetails5)

for(itemDetail in itemDetailsList){

itemDetails[nrow(itemDetails)+1,]<-itemDetail

}

View(itemDetails)

#Performing an Inner Join merge

merged <- merge(itemData, itemDetails, by = "itemCode")

merged
```

Let us use the built-in dataset air quality which has Daily air quality measurements in New York, May to September 1973. Create a histogram by using appropriate arguments for the following statements.

- a) Assigning names, using the air quality data set.
- b) Change colors of the Histogram
- c) Remove Axis and Add labels to Histogram
- d) Change Axis limits of a Histogram
- e) Create a Histogram with density and Add Density curve to the histogram

```
View(airquality)
air<-airquality
library(ggplot2)
library("gridExtra")
air$Month <- factor(air$Month,labels=c("May","June","July","August","September"))
gg1 <- ggplot(air,aes(x=1:nrow(air),y=Temp)) +
geom_line(aes(col=Month)) +
geom_point(aes(col=Month,size=Wind)) +
geom_smooth(method="loess",col="black") +
labs(x="Time (days)",y="Temperature (F)")

s<-ggplot(data=air,aes(x=Ozone, y=..density..))
s+geom_histogram(binwidth=10,aes(fill=Month),colour="Black")+
geom_density(lwd = 1, colour = 4,
fill = 4, alpha = 0.25)
```

Design a data frame in R for storing about 20 employee details. Create a CSV file named "input.csv" that defines all the required information about the employee such as id, name, salary, start\_date, dept. Import into R and do the following analysis.

- a) Find the total number rows & columns
- b) Find the maximum salary
- c) Retrieve the details of the employee with maximum salary
- d) Retrieve all the employees working in the IT Department.
- e) Retrieve the employees in the IT Department whose salary is greater than 20000 and write these details into another file "output.csv"

Sl.No	id	name	salary	start_date	department
1	2005001	Clark Kent	15000	01-01-2005	Media
		Bruce			
2	2005002	Wayne	80000	01-01-2005	Executive
3	2005003	Hal Jordan	10000	01-01-2005	Defence
4	2005004	Barry Allen	15000	01-04-2005	Forensics
5	2005005	Wally West	15000	01-04-2005	Media
6	2006001	Arthur Curry	30000	03-05-2006	Catering
7	2007001	Oliver Queen	80000	01-10-2007	Executive
		Richard			
8	2007002	Grayson	5000	01-10-2007	IT
9	2007003	Tim Drake	15000	01-10-2007	IT
10	2007004	Jason Todd	20000	01-10-2007	IT
11	2007005	Tony Stark	90000	15-11-2007	Executive
12	2007006	Steve Rogers	25000	15-11-2007	Sports
13	2007007	Peter Parker	20000	15-11-2007	Media
14	2008001	Clint Barton	40000	02-02-2008	IT
		Bruce			
15	2008002	Banner	30000	02-02-2008	R&D
		Reed			
16	2008003	Richards	35000	02-02-2008	R&D
17	2008004	Ellie Satler	40000	15-02-2008	R&D
		John			
18	2008005	Hammond	90000	01-05-2008	Executive
19	2008006	Ian Malcolm	60000	01-05-2008	Accounts
20	2008007	Alan Grant	60000	01-05-2008	R&D

setwd(' D:/NBA 2013/2.course file/27. R Programming/VTU-RProgramming-master/VTU-RProgramming-master/dataset ')

# Importing the dataset

dataset = read.csv('input.csv')

View (dataset)

typeof(dataset)

dataset

nrow(dataset)

ncol(dataset)

max\_salary = max(dataset\$salary)

max\_salary

```
max_salaries <- subset(dataset, dataset$salary == max(dataset$salary))

max_salaries

IT_Dept<-subset(dataset, dataset$department == 'IT')

IT_Dept

salary_gt_value<-subset(dataset, dataset$salary > 20000)

salary_gt_value

write.csv(salary_gt_value,"output.csv", row.names = FALSE)

#Append new line to dataframe

dataset[nrow(dataset)+1,]<-list(nrow(dataset)+1,'2023001','XYZ','5000','01-01-2023','IT')

View(dataset)
```

Using the built in dataset mtcars which is a popular dataset consisting of the design and fuel consumption patterns of 32 different automobiles. The data was extracted from the 1974 Motor Trend US magazine, and comprises fuel consumption and 10 aspects of automobile design and performance for 32 automobiles (1973-74 models). Format A data frame with 32 observations on 11 variables: [1] mpg Miles/(US) gallon, [2] cyl Number of cylinders [3] disp Displacement (cu.in.), [4] hp Gross horsepower [5] drat Rear axle ratio,[6] wt Weight (lb/1000) [7] qsec 1/4 mile time, [8] vs V/S, [9] am Transmission (0 = automatic, 1 = manual), [10] gear Number of forward gears, [11] carb Number of carburetors

**Answer the following:** 

- a) What is the total number of observations and variables in the dataset?
- b) Find the car with the largest hp and the least hp using suitable functions
- c) Plot histogram / density for each variable and determine whether continuous variables are normally distributed or not. If not, what is their skewness?
- d) What is the average difference of gross horse power(hp) between automobiles with 4 and 8 (to be corrected instead of 3 and 4) number of cylinders(cyl)? Also determine the difference in their standard deviations.
- e) Which pair of variables has the highest Pearson correlation?

```
rm(list=ls())
library(e1071)
library(ggplot2)
mtcars
carData <- mtcars
number.of.observations<-dim(carData)[1]
number.of.variables<-dim(carData)[2]
car.with.largest.hp<-row.names(subset(carData, carData$hp==max(carData$hp)))
car.with.least.hp</ri>
car.with.least.hp<-row.names(subset(carData, carData$hp==min(carData$hp)))
car.with.least.hp
hist(carData$mpg,col="green", border="black",prob = TRUE)
lines(density(carData$mpg),col = "chocolate3")
skewness(carData$mpg,type=2)
kurtosis(carData$mpg,type=2)
```

```
carData.4cyl.hp<-subset(carData,carData$cyl==4)["hp"]
carData.8cyl.hp<-subset(carData,carData$cyl==8)["hp"]
diff.gross.hp<-round(mean(carData.8cyl.hp[,1])-mean(carData.4cyl.hp[,1]))
print(diff.gross.hp)
diff.sd<-round(sd(carData.8cyl.hp[,1])-sd(carData.4cyl.hp[,1]))
print(diff.sd)
print(cor(carData$mpg, carData$cyl, method = 'pearson'))
print(cor(carData$mpg, carData$disp, method = 'pearson'))
print(cor(carData$mpg, carData$drat, method = 'pearson'))
print(cor(carData$mpg, carData$drat, method = 'pearson'))
print(cor(carData$mpg, carData$wt, method = 'pearson'))
print(cor(carData$mpg, carData$wt, method = 'pearson'))
print(cor(carData$mpg, carData$qsec, method = 'pearson'))
```

Demonstrate the progression of salary with years of experience using a suitable data set (You can create your own dataset). Plot the graph visualizing the best fit line on the plot of the given data points. Plot a curve of Actual Values vs. Predicted values to show their correlation and performance of the model. Interpret the meaning of the slope and y-intercept of the line with respect to the given data. Implement using lm function. Save the graphs and coefficients in files. Attach the predicted values of salaries as a new column to the original data set and save the data as a new CSV file.

Sample CSV:

YearsExperience	Salary
1.1	39343
1.3	46205
1.5	37731
2	43525
2.2	39891
2.9	56642
3	60150
3.2	54445
3.2	64445
3.7	57189
3.9	63218
4	55794
4	56957
4.1	57081
4.5	61111
4.9	67938
5.1	66029
5.3	83088
5.9	81363
6	93940
6.8	91738
7.1	98273
7.9	101302
8.2	113812
8.7	109431
9	105582

9.5	116969
9.6	112635
10.3	122391
10.5	121872

```
setwd('PATHNAME')
dataset = read.csv('Salary_Data.csv')
View (dataset)
library(caTools)
set.seed(123)
split = sample.split(dataset$Salary, SplitRatio = 2/3)
training_set = subset(dataset, split == TRUE)
test_set = subset(dataset, split == FALSE)
regressor = lm(formula = Salary ~ YearsExperience, data = training_set)
y_pred = predict(regressor, newdata = test_set)
library(ggplot2)
ggplot() + geom_point(aes(x = training_set$YearsExperience, y = training_set$Salary), colour = 'red') +
 geom_line(aes(x = training_set)YearsExperience, y = predict(regressor, newdata = training_set)),
       colour = 'blue') + ggtitle('Salary vs Experience (Training set)') + xlab('Years of experience') + ylab('Salary')
library(ggplot2)
ggplot() + geom_point(aes(x = test_set$YearsExperience, y = test_set$Salary), colour = 'red') +
 geom_line(aes(x = training_set$YearsExperience, y = predict(regressor, newdata = training_set)),colour = 'blue') +
 ggtitle('Salary vs Experience (Validation set)') + xlab('Years of experience') + ylab('Salary')
```