### MATHS LAB EXPERIMENT 1 TO 5

NAME: ADARSH GAURAV REG NO: 21BCE1628

#### LAB-1

# Introduction: Understanding Data types; importing/exporting data

#### **CODE**

Χ

```
#Generate data
1:10
#Assign variable name to the value
X = 10
# Combine numeric values into vector
c(1,2,5)
#Arithmetic operations on vectors
a = c(1, 3, 5, 7)
b = c(2, 4, 6, 8)
#Addition
a+b
#Subtraction
a-b
#Constant multiplication
5*a
#Product
a*b
#Division
a/b
#String values
X = as.character(5.2)
```

#### OUTPUT

```
[1] 1 2 3 4 5 6 7 8 9 10
 [1] 1 2 5
[1] 3 7 11 15
[1] -1 -1 -1 -1
  [1] 5 15 25 35
  [1] 2 12 30 56
  [1] 0.5000000 0.7500000 0.8333333 0.8750000
  [1] "5.2"
CODE
#Generate data
1:10
#Assign variable name to the value
X <- 5
# Combine numeric values into vector
c(12,43,14,54,100)
#Arithmetic operations on vectors
a = c(12, 43, 11, 14, 9, 10)
b = c(59, 110, 2, 19, 6, 39)
#Addition
a+b
#Subtraction
a-b
#Constant multiplication
10*a
#Product
a*b
#Division
a/b
```

```
#String values
X = as.character(12.9)
X
#Concatenation of strings
paste("Hello", "World")
```

#### OUTPUT

```
[1] 1 2 3 4 5 6 7 8 9 10
[1] 12 43 14 54 100
function (...) .Primitive("c")
[1] 71 153 13 33 15 49
[1] -47 -67 9 -5 3 -29
[1] 120 430 110 140 90 100
[1] 708 4730 22 266 54 390
[1] 0.2033898 0.3909091 5.50000000 0.7368421 1.5000000 0.2564103
[1] "12.9"
[1] "Hello World"
```

# Computing Summary Statistics /plotting and visualizing data using Tabulation and Graphical Representations

```
CODE
empid
age = c(30, 37, 45, 32, 50, 60, 35, 32, 34, 43, 32, 30, 43, 50, 60)
gender = c(0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 0)
gender
Х
status = c(1, 1, 2, 2, 1, 1, 1, 2, 2, 1, 2, 1, 2, 1, 2)
status
empinfo = data.frame(empid, age, gender, status)
empinfo
empinfo$gender = factor(empinfo$gender, labels=c("male", "female"))
empinfo
empinfo$status = factor(empinfo$status, labels=c("staff", "faculty"))
empinfo$status
empinfo
male = subset(empinfo, empinfo$gender=="male")
female = subset(empinfo, empinfo$gender=="female")
female
staff = subset(empinfo, empinfo$status=="staff")
staff
faculty = subset(empinfo, empinfo$status=="faculty")
faculty
faculty and female = subset(female, female$status=="faculty")
faculty and female
summary(empinfo)
summary(male)
summary(faculty and female)
summary(age)
table1 = table(empinfo$gender)
table1
table2 = table(empinfo$status)
table2
table3 = table(empinfo$gender, empinfo$status)
table3
```

plot(empinfo\$age, type="l", main="Age of employees", xlab="empid", ylab="age in years", col="blue")

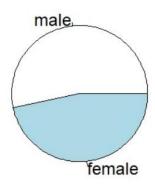
pie(table1)

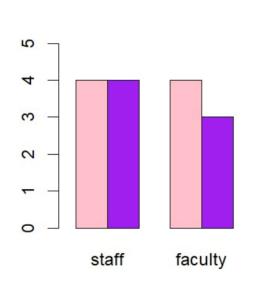
pie(table2)

pie(table3)

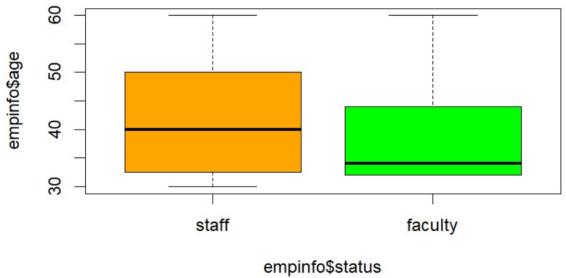
barplot(table3, beside=T, xlim=c(1, 15), ylim=c(0, 5), col=c("pink", "purple")) legend("topright", legend=rownames(table3), fill=c('pink', 'purple'), bty="n") boxplot(empinfo\$age~empinfo\$status, col=c('orange', 'green'))

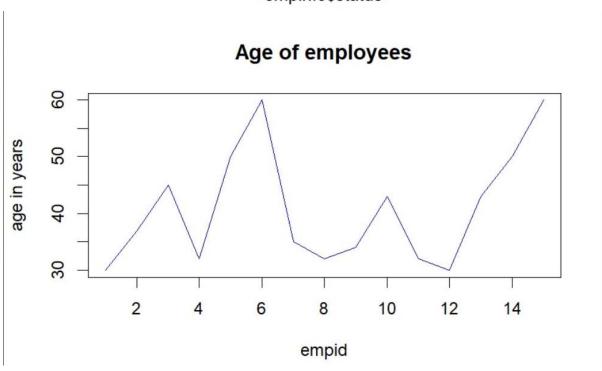
#### OUTPUT









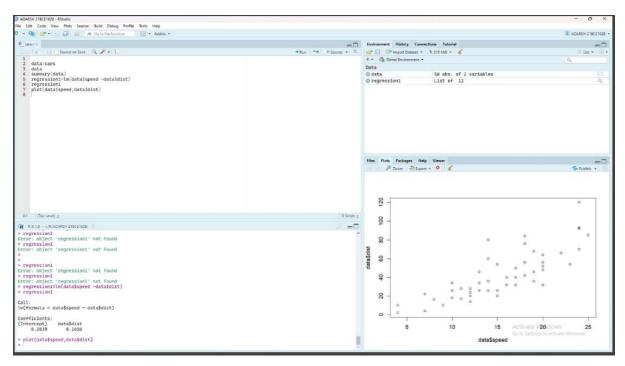


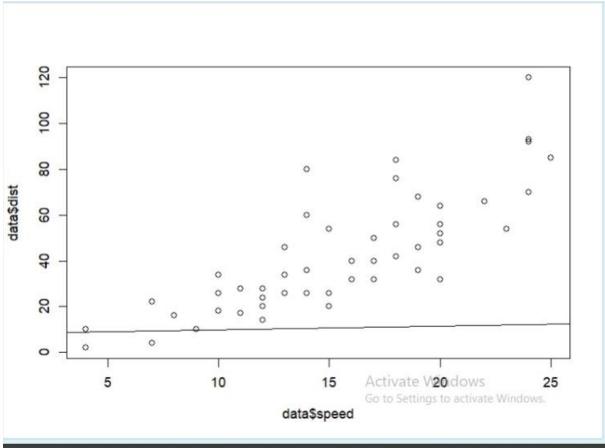
NAME: ADARSH GAURAV REG NO: 21BCE1628

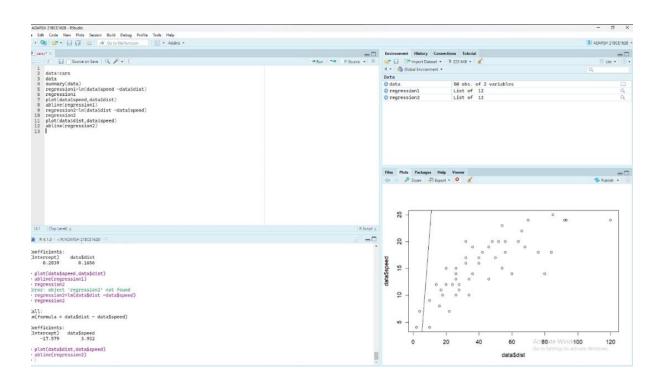
#### LAB-3

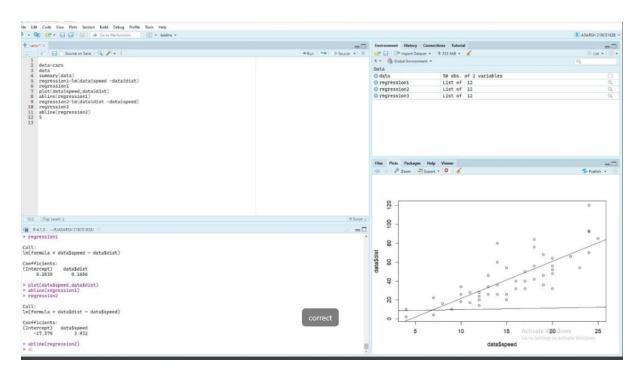
# Applying correlation and simple linear regression model to real data set; computing and interpreting the coefficient of determination

```
data=cars
data
summary(data)
regression1=Im(data$speed ~data$dist)
regression1
plot(data$speed,data$dist)
abline(regression1)
regression2=Im(data$dist ~data$speed)
regression2
plot(data$dist,data$speed)
abline(regression2)
```









weight = c(15,26,27,25,25.5,27,32,18,22,20,26,24)

weight

bmi =c(13.35,16.12,16.74,16,13.59,15.73,15.65,13.85,16.07,12.8,13.65,14.42)

bmi

cor(weight,bmi)

Regression1 = Im(bmi~weight)

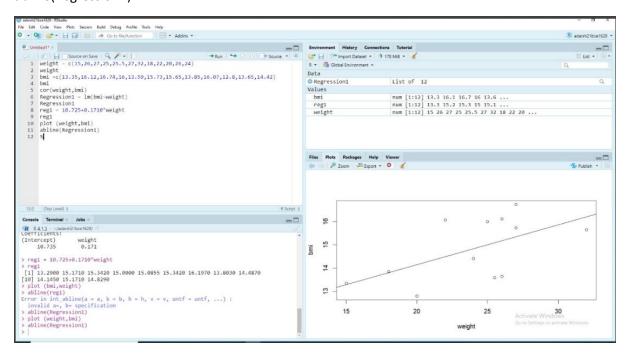
Regression1

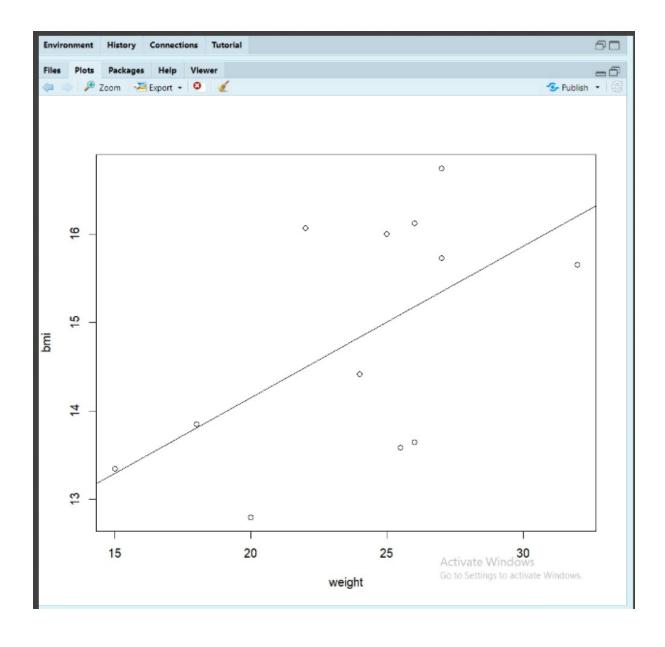
reg1 = 10.725+0.1710\*weight

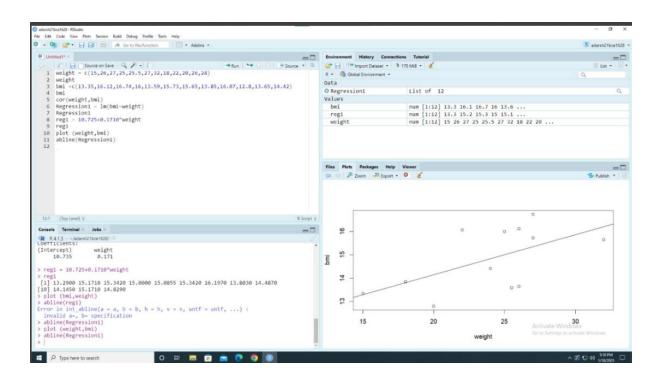
reg1

plot (weight,bmi)

abline(Regression1)







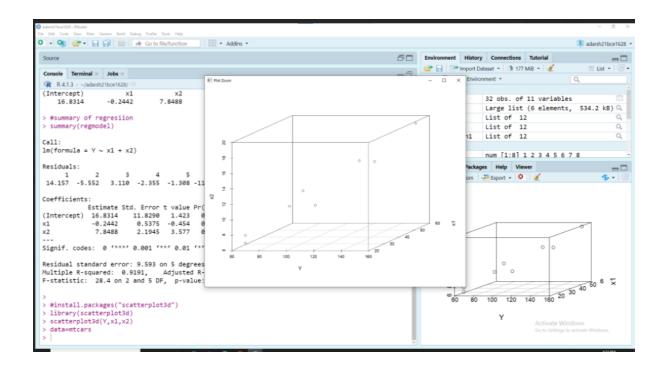
#### LAB 4

# Applying multiple linear regression model to real dataset; computing and interpreting the multiple coefficients of determination

### **CODE**

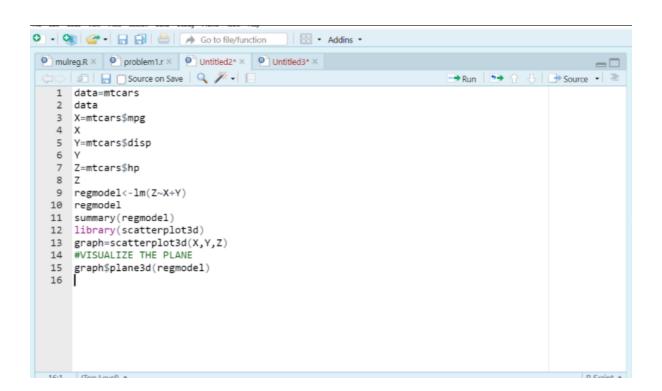
```
Y=c(110,80,70,120,150,90,70,120)
Y
x1=c(30,40,20,50,60,40,20,60)
x1
x2=c(11,10,7,15,19,12,8,14)
x2
# linera regression model of y on x1 and x2
regmodel=lm(Y~x1+x2)
regmodel
#summary of regresiion
summary(regmodel)
#install.packages("scatterplot3d")
library(scatterplot3d)
scatterplot3d(Y,x1,x2)
```

```
> Y=c(110,80,70,120,150,90,70,120)
[1] 110 80 70 120 150 90 70 120
> x1=c(30,40,20,50,60,40,20,60)
[1] 30 40 20 50 60 40 20 60
> x2=c(11,10,7,15,19,12,8,14)
> x2
[1] 11 10 7 15 19 12 8 14
> regmodel=lm(Y1-X1+X2)
Error in stats::model.frame(formula = Y1 - x1 + x2, drop.unused.levels = TRUE) :
 object 'Y1' not found
> regmodel
Call:
lm(formula = y \sim x1 + x2)
Coefficients:
(Intercept)
                  x1
                                x2
                          7.8488
  16.8314
              -0.2442
> regmodel=lm(Y-x1+x2)
Error in formula.default(object, env = baseenv()) : invalid formula
> regmodel
Call:
lm(formula = y \sim x1 + x2)
Coefficients:
```



#### CODE

```
data=mtcars
data
X=mtcars$mpg
X
Y=mtcars$disp
Y
Z=mtcars$hp
Z
regmodel<-Im(Z~X+Y)
regmodel
summary(regmodel)
library(scatterplot3d)
graph=scatterplot3d(X,Y,Z)
#VISUALIZE THE PLANE
graph$plane3d(regmodel)
```



```
mpg cyl alsp np arat
                                                            Wt gsec vs am gear carp
                         21.0 6 160.0 110 3.90 2.620 16.46 0 1 4
21.0 6 160.0 110 3.90 2.875 17.02 0 1 4
Mazda RX4
Mazda RX4 Wag
                                                                                              4
                         22.8 4 108.0 93 3.85 2.320 18.61 1 1
Datsun 710
                                                                                              1
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
                         18.1 6 225.0 105 2.76 3.460 20.22 1 0 3
                        14.3 8 360.0 245 3.21 3.570 15.84 0 0 3
Duster 360
                                                                                              4
Merc 240D
                        2
Merc 230
                                                                                              2
                         19.2 6 167.6 123 3.92 3.440 18.30
17.8 6 167.6 123 3.92 3.440 18.90
Merc 280
                                                                            1 0
                                                                                      4
                                                                                              4
Merc 280C
                                                                            1 0
                                                                                      4
                                                                                              4
                         16.4 8 275.8 180 3.07 4.070 17.40 0 0
                                                                                     3
Merc 450SE
                                                                                              3
                    17.3 8 275.8 180 3.07 3.730 17.60 0 0 3
15.2 8 275.8 180 3.07 3.780 18.00 0 0 3
Merc 450SL
                                                                                              3
Merc 450SLC
                                                                                              3
Cadillac Fleetwood 10.4 8 472.0 205 2.93 5.250 17.98 0 0 3
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
Fiat 128 32.4 4 78.7 66 4.08 2.200 19.47 1 1 4
Honda Civic 30.4 4 75.7 52 4.93 1.615 18.52 1 1 4
Toyota Corolla 33.9 4 71.1 65 4.22 1.835 19.90 1 1 4
Toyota Corona 21.5 4 120.1 97 3.70 2.465 20.01 1 0 3
                                                                                              4
                                                                                              1
                                                                                              2
                                                                                              1
Dodge Challenger 15.5 8 318.0 150 2.76 3.520 16.87 0 0
                                                                                     3
                                                                                              2
AMC Javelin 15.2 8 304.0 150 3.15 3.435 17.30 0 0 3 Camaro Z28 13.3 8 350.0 245 3.73 3.840 15.41 0 0 3
                                                                                              2
Fiat X1-9 27.3 4 79.0 66 4.08 1.935 18.90 1 1 4

Porsche 914-2 26.0 4 120.3 91 4.43 2.140 16.70 0 1 5

Lotus Europa 30.4 4 95.1 113 3.77 1.513 16.90 1 1 5

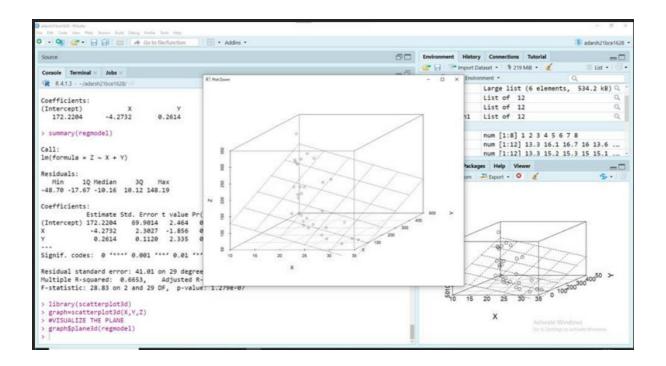
Ford Pantera L 15.8 8 351.0 264 4.22 3.170 14.50 0 1 5

Ferrari Dino 19.7 6 145.0 175 3.62 2.770 15.50 0 1 5

Maserati Rora 15 0 8 301 0 335 3 54 3 570 14 60 0
 Pontiac Firebird 19.2 8 400.0 175 3.08 3.845 17.05 0 0 3
                                                                                              1
                                                                                              2
                                                                                              2
                                                                                              4
                                                                                              6
Maserati Rora
                                                                                              2
```

```
30.4 4 95.1 113 3.77 1.513 16.90 1 1
Lotus Europa
                  15.8 8 351.0 264 4.22 3.170 14.50 0 1 5
Ford Pantera L
                                                                   4
                  19.7 6 145.0 175 3.62 2.770 15.50 0 1 5 15.0 8 301.0 335 3.54 3.570 14.60 0 1 5
Ferrari Dino
                                                                   6
Maserati Bora
                                                                   8
                  21.4 4 121.0 109 4.11 2.780 18.60 1 1
Volvo 142E
> X=mtcars$mpg
[1] 21.0 21.0 22.8 21.4 18.7 18.1 14.3 24.4 22.8 19.2 17.8 16.4 17.3 15.2 10.4 10.4 14.7 32.4
[19] 30.4 33.9 21.5 15.5 15.2 13.3 19.2 27.3 26.0 30.4 15.8 19.7 15.0 21.4
> Y=mtcars$disp
[1] 160.0 160.0 108.0 258.0 360.0 225.0 360.0 146.7 140.8 167.6 167.6 275.8 275.8 275.8 472.0
[16] 460.0 440.0 78.7 75.7 71.1 120.1 318.0 304.0 350.0 400.0 79.0 120.3 95.1 351.0 145.0
[31] 301.0 121.0
> Z=mtcars$hp
 [1] 110 110 93 110 175 105 245 62 95 123 123 180 180 180 205 215 230 66 52 65 97 150
[23] 150 245 175 66 91 113 264 175 335 109
> regmodel<-lm(Z~X+Y)</pre>
> regmodel
lm(formula = Z \sim X + Y)
Coefficients:
(Intercept)
                    X
                           0.2614
               -4.2732
  172.2204
> summary(regmodel)
Call:
```

```
Console Terminal × Jobs ×
R 4.1.3 · ~/adarsh21bce1628/
Coefficients:
(Intercept)
                              0.2614
                -4.2732
  172.2204
> summary(regmodel)
Call:
lm(formula = Z \sim X + Y)
Residuals:
  Min 1Q Median 3Q
-48.70 -17.67 -10.16 10.12 148.19
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
(Intercept) 172.2204 69.9014 2.464 0.0199 *
X -4.2732 2.3027 -1.856 0.0737 .
Y 0.2614 0.1120 2.335 0.0267 *
Signif. codes: 0 "***, 0.001 "**, 0.01 "*, 0.05 "., 0.1 ", 1
Residual standard error: 41.01 on 29 degrees of freedom
Multiple R-squared: 0.6653, Adjusted R-squared: 0.6423
F-statistic: 28.83 on 2 and 29 DF, p-value: 1.279e-07
> library(scatterplot3d)
> graph=scatterplot3d(X,Y,Z)
> #VISUALIZE THE PLANE
> graph$plane3d(regmodel)
```



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#### LAB - 5

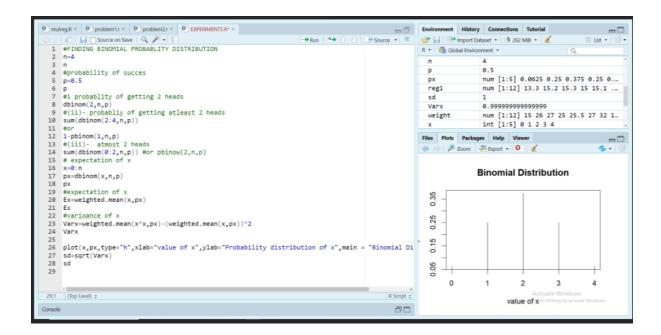
## Fitting the probability distributions: Binomial distribution

#### **CODE**

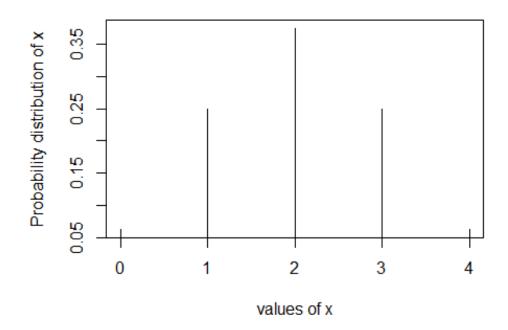
```
#FINDING BINOMIAL PROBABLITY DISTRIBUTION
n=4
#probability of succes
p=0.5
#i probablity of getting 2 heads
dbinom(2,n,p)
#(ii)- probabliy of getting atleast 2 heads
sum(dbinom(2:4,n,p))
#or
1-pbinom(1,n,p)
#(iii)- atmost 2 heads
sum(dbinom(0:2,n,p)) #or pbinow(2,n,p)
# expectation of x
x=0:n
px=dbinom(x,n,p)
px
#expectation of x
Ex=weighted.mean(x,px)
Ex
#varioance of x
```

 $Varx=weighted.mean(x*x,px)-(weighted.mean(x,px))^2$ Varx

plot(x,px,type="h",xlab="value of x",ylab="Probability distribution of x",main = "Binomial Distribution")



#### **Binomial distribution**



## **ALL EXPERIMENT R CODE**

#Experiment-1: Understanding data types, importing and exporting data
#Generate data
#Class Code
#Generate data
1:10
#Assign variable name to the value
X = 10
# Combine numeric values into vector
c(1,2,5)
#Arithmetic operations on vectors
a = c(1, 3, 5, 7)
b = c(2, 4, 6, 8)
#Addition
a+b
#Subtraction
a-b
#Constant multiplication
5*a
#Product
a*b

#Subtraction

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gender = c(0,1,0,1,1,1,0,0,1,0,0,1,1,0,0)

status = c(1,1,2,2,1,1,1,2,2,1,2,1,2,1,2)

gender

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```
status
```

```
#Create dataframe
empinfo = data.frame(empid, age, gender, status)
empinfo
# labeling character to numeric
empinfo$gender = factor(empinfo$gender, labels = c("male", "female"))
empinfo$gender
empinfo$status = factor(empinfo$status, labels = c("staff", "faculty"))
empinfo$status
empinfo
# extract male data
male = subset(empinfo,empinfo$gender=="male")
male
# extract female data
female = subset(empinfo, empinfo$gender == "female")
female
# summary statistics for empinfo data
summary(empinfo)
summary(male)
summary(female)
summary(age)
#creating table (one-way)
table1 = table(empinfo$gender)
table1
table2 = table(empinfo$status)
```

```
table2
```

```
#creating table (two-way)
table3 = table(empinfo$gender, empinfo$status)
table3
#Graphical representation
plot(empinfo$age, type="I", main="Age of employees", xlab="empid", ylab="Age in years", col =
"blue") pie(table1) barplot(table3, beside=T, xlim=c(1,15), ylim=c(0,5), col=c("blue", "red"))
legend("topright", legend=rownames(table3), fill=c('blue', 'red'), bty="n")
boxplot(empinfo$age~empinfo$status, col=c('red', 'blue'))
pie(table1)
barplot(table3, beside=T, xlim=c(1,15), ylim=c(0,5), col=c("blue", "red")) legend("topright",
legend=rownames(table3), fill=c('blue', 'red'), bty="n")
boxplot(empinfo$age~empinfo$status, col=c('red', 'blue'))
#Practice question
#Put data into dataframe
name = c('p1', 'p2', 'p3', 'p4', 'p5', 'p6', 'p7', 'p8', 'p9', 'p10')
age = c(20, 20, 21, 20, 21, 22, 21, 21, 20, 22)
math = c(100, 90, 95, 98, 92, 99, 95, 100, 93, 90)
phy = c(90, 100, 91, 92, 98, 92, 98, 99, 100, 93)
chem = c(93,89,99,92,94,95,99,85,)
# Create dataframe for student marks
studMarks = data.frame(name, age, math, phy, chem)
studMarks
summary(studMarks)
# Plot math marks of students
```

plot(studMarks\$math, type="I", main="Marks in math", xlab="Student", ylab="Marks", col = "blue")

```
#creating table (one-way)
tableAge = table(studMarks$age)
tableAge
#Pie chart for age
pie(tableAge)
tablePhyMarks = table(studMarks$phy)
table Phy Marks
#Bar plot for physics marks
barplot(tablePhyMarks, beside=T, xlim=c(0,10), ylim=c(0,5), col=c("blue", "red"))
#Box plot for math marks and age
boxplot(studMarks$math~studMarks$age, col=c('red', 'blue'))
#Experiment-3: Apply correlation and simple linear regression model to real data
#Class question
#Problem-1
#Import the inbuilt dataset cars
data=cars
data
#Summary of data set
summary(data)
#Variance of speed
v1=var(data$speed)
٧1
```

```
#variance of "dist"
v2=var(data$dist)
v2
#covariance between speed and dist
covariance=cov(data$speed,data$dist)
covariance
#Standard deviation
stdDevSpeed=sd(data$speed)
stdDevSpeed
stdDevDist = sd(data$dist)
stdDevDist
#correlation coefficient
corr = covariance/(stdDevSpeed*stdDevDist)
corr
#or
corr2 = cor(data$speed, data$dist)
corr2
cor.test(data$speed, data$dist)
cor.test(data$speed, data$dist, method="pearson")
cor.test(data$speed, data$dist, method="spearman")
regression1 = Im(data$speed~data$dist)
regression1
```

```
#Plot regression2
plot(data$speed, data$dist)
abline(regression1)
regression2 = Im(data$dist~data$speed)
regression2
#Plot regression1
plot(data$dist, data$speed)
abline(regression2)
#Problem-2
weight = c(15, 26, 27, 25, 25.5, 27, 32, 18, 22, 20, 26, 24)
weight
bmi = c(13.35, 16.12, 16.74, 16, 13.59, 15.73, 15.65, 13.85, 16.07, 12.8, 13.65, 14.42);
bmi
#Apply correlation
cor(weight, bmi)
#Find regression
Regression1 = Im(bmi~weight)
Regression1
reg1 = Im(bmi~weight)
reg1 = 10.735+0.1710*weight
reg1
#Plot regression
plot(weight, bmi)
```

```
abline(Regression1)
```

```
#Practice problem
```

#Vector for age of husbands

ageHusbands = c(23,27,28,29,30,31,33,35,36,39)

#Vector for age of wives

ageWives = c(18,22,23,24,25,26,28,29,30,32)

#Create dataframe

data = data.frame(ageHusbands, ageWives)

#Summary of data set

summary(data)

#Variance of husbands' age

v1=var(data\$ageHusbands)

٧1

#variance of wives' age

v2=var(data\$ageWives)

v2

#covariance between husbands' age and wives' age

covariance=cov(data\$ageHusbands,data\$ageWives)

covariance

**#Standard deviation** 

stdDevHusbAge=sd(data\$ageHusbands)

std Dev Husb Age

stdDevWivesAge = sd(data\$ageWives)

```
stdDevWivesAge
#correlation coefficient
corr = covariance/(stdDevHusbAge*stdDevWivesAge)
corr
#or
corr2 = cor(data$ageHusbands, data$ageWives)
corr2 = cor(data)
cor.test(data$ageHusbands, data$ageWives)
cor.test(data$ageHusbands, data$ageWives, method="pearson")
cor.test(data$ageHusbands, data$ageWives, method="spearman")
regression1 = Im(data$ageHusbands~data$ageWives)
regression1
#Plot regression1
plot(data$ageHusbands, data$ageWives)
abline(regression1)
regression2 = Im(data$ageWives~data$ageHusbands)
regression2
#Plot regression2
plot(data$ageWives, data$ageHusbands)
abline(regression2)
```

#\_\_\_\_\_

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```
#Experiment-4: Apply multiple linear regression model to real data set
#Class question
#Problem-1
# Input the variables
Y = c(110, 80, 70, 120, 150, 90, 70, 120)
Υ
X1 = c(30, 40, 20, 50, 60, 40, 20, 60)
X1
X2 = c(11, 10, 7, 15, 19, 12, 8, 14)
X2
# Linear regression model of Y on X1 and X2\
RegModel = Im(Y^X1+X2) RegModel
# Summary of the data
summary(RegModel)
# install.packages("scatterplot3d")
library(scatterplot3d)
# plot the data set
scatterplot3d(Y,X1,X2)
#Problem-2
data = mtcars
data
X = mtcars$mpg
Χ
Y = mtcars$disp
Υ
Z = mtcars$hp
```

Ζ

```
RegModel <-Im(Z^*X+Y)
RegModel
summary(RegModel)
graph = scatterplot3d(X,Y,Z)
# Visualize the plane
graph$plane3d(RegModel)
#Practice question
Y = c(240,236,290,274,301,316,300,296,267,276,288,261)
X1 = c(25,31,45,60,65,72,80,84,75,60,50,38)
X2 = c(24,21,24,25,25,26,25,25,24,25,25,25)
# Linear regression model of Y on X1 and X2
RegModel = Im(Y^X1+X2)
RegModel
# Summary of the data
summary(RegModel)
library(scatterplot3d)
# plot the data set scatterplot3d(Y,X1,X2)
graph$plane3d(RegModel)
#-----
#Experiment-5: Binomial Distribution
#Class Question
#Four coins are tossed simultaneously. What is the probability of getting
#(i) 2 heads
#(ii) atleast 2 heads
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#(iii) atmost 2 heads

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#(iv) Expectation of x
#(v) Variance of x
#(vi) Visualize the probability distribution
# number of trials
n = 4
n
# probability of success
p = 0.5
р
# (i) probabilty of gettting exactly two heads
dbinom(2,n,p)
# (ii) probabilty of getting atleast two heads
sum(dbinom(2:4, n, p))
# or
1 - pbinom(1, n, p)
# (iii) probability of getting atmost two heads
sum(dbinom(0:2,n,p))
# (iv) Expectation of x
x = 0:n
px = dbinom(x, n, p)
рх
Ex = weighted.mean(x, px)
Ex
```

# (v) variance of x

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varx = weighted.mean(x*x, px) - (weighted.mean(x, px))^2
varx
std <- sqrt(varx)</pre>
std
# (vi) visualize probability distribution
plot(x, px, type = "h", xlab = "values of x", ylab = "Probability distribution of x", main =
    "Binomial distribution")
#Practice question
#The incidence of an occupational disease in an industry is such that the workers have a 20% chance
#of suffering from it. What is the probability that out of 6 workers chosen at random 4 or more will
suffer
#disease?
# number of trials
n = 6
n
# probability of success
p = 0.2
р
sum(dbinom(4:6, n, p))
x = 0:n
px = dbinom(x, n, p)
рх
# visualize probability distribution
plot(x, px, type = "h", xlab = "values of x", ylab = "Probability distribution of x", main =
    "Binomial distribution")
```

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