MATHEMATICAL STATISTICS LAB FILE

MSMA 109

2022-2023



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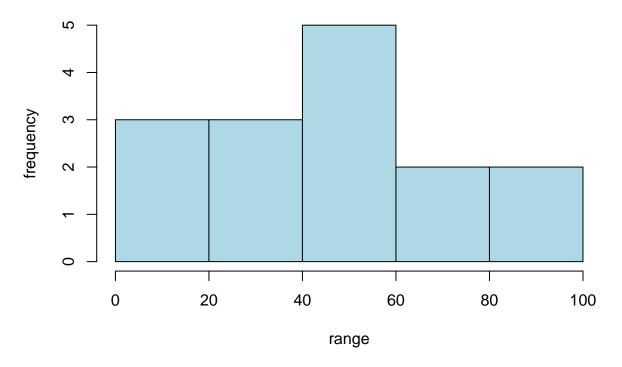
INDEX

S. No.	TITLE	DATE	REMARKS
1	Descriptive Statistics	23/08/2022	
2	Plotting Ogive curve	30/08/2022	
3	Bayes Theorem	06/09/2022	
4	Joint, Marginal, Conditional Probability and CDF	06/09/2022	
5	Binomial Distribution	13/09/2022	
6	Poisson Distribution	20/09/2022	
7	Normal Distribution	11/10/2022	
8	Correlation Analysis	18/10/2022	
9	Regression Analysis	18/10/2022	
10	Hypothesis Testing: t-test	01/11/2022	

Practical-1-Descriptive-Statistics.R

```
#NAME: RITIKA GUPTA
#ROLL NO.:
#COURSE: MSC MATHEMATICS
#Question 1. Create a set of first 5 prime numbers
P=c(2,3,5,7,11)
## [1] 2 3 5 7 11
#Question 2. create a set of 10 random numbers from 50 to 100
R=sample(50:100,10)
## [1] 58 68 60 55 56 86 78 69 77 64
#Question 3. calculate mean, median, mode of data
V=sample(1:100,15)
V
## [1] 33 90 45 59 39 18 51 56 73 92 71 41 32 3 17
mean(V)
## [1] 48
V1=sort(V) #sorting data in ascending order
## [1] 3 17 18 32 33 39 41 45 51 56 59 71 73 90 92
median(V1)
## [1] 45
hist(V1,col = "lightblue",xlab = "range",ylab = "frequency",
main = "Dataset: 15 numbers from 1 to 100")
```

Dataset: 15 numbers from 1 to 100



```
#Question 4. Statistical inferences of data in excel
library(readxl)
data= read_excel("marks_data.xlsx")
View(data)
#mean, median of data in excel
mean1=mean(data$'Marks 1')
median1=median(sort(data$'Marks 1'))
mean2=mean(data$'Marks 2')
median2=median(sort(data$'Marks 2'))
print(paste("Mean of first row = ",mean1))
## [1] "Mean of first row = 181.645161290323"
print(paste("Median of first row = ",median1))
## [1] "Median of first row = 125"
print(paste("Mean of second row = ",mean2))
## [1] "Mean of second row = 78.8064516129032"
print(paste("Median of second row = ",median2))
## [1] "Median of second row = 77"
#standard deviation and variation
sd1=sd(data$'Marks 1')
```

```
sd2=sd(data$'Marks 2')
var1=var(data$'Marks 1')
var2=var(data$'Marks 2')
print(paste("Standard deviation of 1st row = ",sd1))
## [1] "Standard deviation of 1st row = 159.93155794216"
print(paste("Variance of 1st row = ",var1))
## [1] "Variance of 1st row = 25578.1032258065"
print(paste("Standard deviation of 2nd row = ",sd2))
## [1] "Standard deviation of 2nd row = 13.2600134611262"
print(paste("Variance of 2nd row = ",var2))
## [1] "Variance of 2nd row = 175.827956989247"
#kurtosis and skewness
library(moments)
k1=kurtosis(data$'Marks 1')
k2=kurtosis(data$'Marks 2')
s1=skewness(data$'Marks 1')
s2=skewness(data$'Marks 2')
print(paste("Kurtosis of 1st row = ",k1))
## [1] "Kurtosis of 1st row = 2.77608997027731"
print(paste("Skewness of 1st row = ",s1))
## [1] "Skewness of 1st row = 0.958734494896294"
print(paste("Kurtosis of 2nd row = ",k2))
## [1] "Kurtosis of 2nd row = 1.84218052212265"
print(paste("Skewness of 2nd row = ",s2))
## [1] "Skewness of 2nd row = 0.274812460487586"
#sample() takes a sample of the specified size from the elements of x using either
#with or without replacement.
#Argument replace=TRUE when size is greater than length of dataset.
#c() command is a function which combines its arguments to form a vector.
#sort() arranges data given in the argument in an ascending order
#mean() function calculates the arithmetic mean of the data given in its argument.
#median() function calculates the sample median of the data given in its argument.
#hist() function computes a histogram of the given data values.
#xlab, ylab arguments produce axes labels, 'main' argument produces title, and
#col argument is used to give colour to fill the bars of the histogram.
```

```
#read_excel() is used to import excel data in R progam
#sd() function calculates standard deviation of data in its argument
#var() function calculates variance of data in its argument
#kurtosis() function calculates kurtosis of data in its argument
#skewness() function calculates skewness of data in its argument
#*********************************
S=sample(1:10)
## [1] 8 10 5 1 9 3 2 4 6 7
x=1:100
x1=sample(x,10)
x2=sample(50:100,10)
   [1] 65 54 13 5 77 61 22 46 52 67
x2
## [1] 64 68 52 93 71 65 90 99 88 73
length(x1)
## [1] 10
length(x2)
## [1] 10
y=c(2,4,6,8,10)
У
## [1] 2 4 6 8 10
y1=c("H","T")
sample(1:2,10,replace = TRUE)
## [1] 2 1 2 2 2 2 1 1 1 1
#*****************************
```

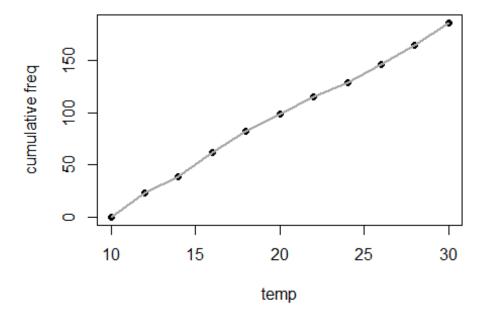
Practical 2: Ogive Curve

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23-08-2022

```
#sample dataset
y=sample(10:30,200,replace = T)
#class interval
\#seq(a,b,c) generates a sequence from a to b by common difference c.
a=seq(10,30,2)
\#cut(x,y) divides the range of x into intervals corresponding to y.
y1=cut(y,a)
#frequency table
f=table(y1)
f
## y1
## (10,12] (12,14] (14,16] (16,18] (18,20] (20,22] (22,24] (24,26] (26,28]
(28,30]
##
        23
                16
                         23
                                 20
                                         17
                                                  16
                                                          14
                                                                  17
                                                                           19
21
#cumulative frequency
cf=cumsum(f)
cf
## (10,12] (12,14] (14,16] (16,18] (18,20] (20,22] (22,24] (24,26] (26,28]
(28,30]
##
        23
                39
                         62
                                 82
                                         99
                                                 115
                                                         129
                                                                 146
                                                                          165
186
s=c(0,cf)
s
##
           (10,12] (12,14] (14,16] (16,18] (18,20] (20,22] (22,24] (24,26]
(26,28]
                23
                         39
                                 62
                                         82
                                                  99
                                                                 129
                                                                          146
##
         0
                                                         115
165
## (28,30]
##
       186
```



Practical 3: Bayes Theorem

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30-08-2022

```
#Q1
#There are two identical urns containing 4 white, 3 red balls and, 3 white, 7
red balls respectively.
#An un is choosen at random and a ball is drawn from it. Find the probability
that the ball is white.
#If the ball drawn is white, what is the probability that it is drawn from
the first urn?
library(LaplacesDemon)
## Warning: package 'LaplacesDemon' was built under R version 4.2.2
PrA=c(1/2,1/2)
PrBA=c(4/7,3/10)
R1=BayesTheorem(PrA, PrBA)
prob_white=(PrA[1]*PrBA[1])+(PrA[2]*PrBA[2])
prob first_white=R1[1]
prob_second_white=R1[2]
print(paste("P(ball is white) = ",prob_white))
## [1] "P(ball is white) = 0.435714285714286"
print(paste("P(I urn | white ball) = ",prob_first_white))
## [1] "P(I urn | white ball) = 0.655737704918033"
print(paste("P(II urn | white ball) = ",prob second white))
## [1] "P(II urn | white ball) = 0.344262295081967"
#What's the probability of going to hell conditional on consorting given that
a person consort.
#6 people consorted out of 9 who went to hell. 5 people consorted out of 7
who went to heaven.
#75% of people go to hell. 25% of people go to heaven.
#A is the event when person goes to heaven #B is the event when person goes
to hell
```

```
PrX=c(3/4,1/4)
PrYX=c(2/3,5/7)
R2=BayesTheorem(PrX,PrYX)
prob hell consort=R2[1]
prob_heaven_consort=R2[2]
print(paste("P(hell | person consorts) = ",prob_hell_consort))
## [1] "P(hell | person consorts) = 0.736842105263158"
print(paste("P(heaven | person consorts) = ",prob_heaven_consort))
## [1] "P(heaven | person consorts) = 0.263157894736842"
#Q3
#Suppose doctors are asked to report the no. of cases of small pox & chicken
pox and
#the symptoms are observed using the result of the survey.
#We find the probability that a patient has spots given that they have
#small pox or chicken pox is 20% and 80% respectively.
#Public health statistic inform us that the small pox in general population
is 0.001 and chicken pox is 0.01.
#What's the probability of the patient having the small pox given that they
have a spot on their face.
#What's the probability of the patient having chicken pox & a spot on their
face.
PrS=c(0.001,0.01)
PrTS=c(0.2,0.8)
R3=BayesTheorem(PrS,PrTS)
Prob small spot=R3[1]
Prob chicken spot=R3[2]
print(paste("P(small pox | spots) = ",Prob_small_spot))
## [1] "P(small pox | spots) = 0.024390243902439"
print(paste("P(chicken pox | spots) = ",Prob_chicken_spot))
## [1] "P(chicken pox | spots) = 0.975609756097561"
```

Practical 4:

Joint, Marginal, Conditional Probabilities and Cumulative Distribution Function

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```
#Joint, Marginal, Conditional Probability and CDF
#Create a joint probability table and use it to compute marginal and
#conditional probabilities
\#P(X=0,Y=1) = 1/3
\#P(X=1,Y=1) = 1/3
\#P(X=-1,Y=1) = 1/3
#we have two random variables X and Y with respective sample spaces
\#X=\{-1,0,1\} and Y=\{-1,0,1\}
#JOINT PROBABILITY
#The joint probability table for these random variables is given by,
P=matrix(c(0,0,0,0,0,1/3,1/3,0,1/3),ncol = 3)
rownames(P)=c("X=-1","X=0","X=1")
colnames(P)=c("Y=-1", "Y=0", "Y=1")
Ρ
       Y=-1
                  Y=0
                             Y=1
## X=-1 0 0.0000000 0.3333333
## X=0 0 0.0000000 0.0000000
## X=1 0 0.3333333 0.3333333
#To display P(X=-1,Y=1), we will display element stored in the
#1st row and 3rd column of p, i.e.,
print(paste("P(-1,1)= ",P[1,3]))
#MARGINAL PROBABILITY
```

```
#Marginal probability of X at x is given by
\#P(X=x) = P(X=x,Y=-1) + P(X=x,Y=0) + P(X=x,Y=1)
Px=apply(P,1,sum)
Px
##
        X=-1
                   X=0
                             X=1
## 0.3333333 0.0000000 0.6666667
#Marginal probability of Y at y is given by
\#P(Y=y) = P(X=-1,Y=y) + P(X=0,Y=y) + P(X=1,Y=y)
Py=apply(P,2,sum)
Ру
##
        Y=-1
                   Y=0
                             Y=1
## 0.0000000 0.3333333 0.6666667
#CONDITIONAL PROBABILITY
#Compute the conditional probability P(X=-1|Y=1) and display it.
P_xy=P[1,3]/Py[3]
print(paste("P(X=-1|Y=1)=",P_xy))
## [1] "P(X=-1|Y=1)= 0.5"
#CUMMULATIVE DISTRIBUTION FUNCTION
#Find the CDF of X
CDF_x=cumsum(Px)
CDF x
        X=-1
##
                   X=0
                             X=1
## 0.3333333 0.3333333 1.0000000
#Find the CDF of Y
CDF_y=cumsum(Py)
CDF_y
##
        Y=-1
                   Y=0
                             Y=1
## 0.0000000 0.333333 1.0000000
```

Practical 5: Binomial Distribution

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```
#Binomial distribution
#Q1. The probability that a person can achieve a target is 3/4. The count
#of tries is 5, what is the probability that he will attain the target at
#least thrice?
#X is the discrete random Variable denoting successful tries
#X can have values: 0,1,2,3,4,5
#To find P(X>=3),
x=3:5
px=sum(dbinom(x,5,3/4))
print(paste("P(X>=3) = ",px))
## [1] "P(X>=3) = 0.896484375"
#Q2. Find the probability of getting at least 5 heads on tossing an
#unbiased coin 6 times.
#Y is a discrete random variable denoting number of successes, i.e. heads
#Y can have values: 0,1,2,3,4,5,6
#To find P(Y>=5),
y=5:6
py=sum(dbinom(y,6,1/2))
print(paste("P(Y>=5) = ",py))
## [1] "P(Y>=5) = 0.109375"
```

Practical 6: Poisson Distribution

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```
#Poisson Distribution
#01. A car hire firm has two cars which it hires out day by day. The number
#of demands of a car on each day is distributed as a poission distribution
#with average number of cars demanded per day as 1.5. Calculate the
proportion of
#days on which neither car is used and the proportion of days on which some
#demand is refused.
\#Let X be a discrete random variable denoting number of cars demanded on a
day
#Possible values of X: 0,1,2,...
#Given, average no. of cars demanded per day, u = 1.5
#To find P(X=0) and P(X>2)
u = 1.5
x=0:2
p1=dpois(x,u)[1]
print(paste("P(neither car is used) = P(X=0) = ",p1))
## [1] "P(neither car is used) = P(X=0) = 0.22313016014843"
p2=1-dpois(x,u)[1]-dpois(x,u)[2]-dpois(x,u)[3]
print(paste("P(some demand is refused) = P(X>2) = ",p2))
## [1] "P(some demand is refused) = P(X>2) = 0.191153169461942"
#Q2. If the probability of a bad reaction from medicine is 0.002, determine
#the chance that out of 1000 persons more than 3 will suffer a bad reaction
#from medicine.
#Let X be a random variable denoting number of people suffering a bad
reaction
#X can have values: 0,1,2,3,...
#To find P(X>3),
n=1000
p=0.002
```

```
U=n*p
X=0:3
prob=1-(dpois(X,U)[1]+dpois(X,U)[2]+dpois(X,U)[3]+dpois(X,U)[4])
print(paste("P(X>3) = ",prob))
## [1] "P(X>3) = 0.142876539501453"
```

Practical 7: Normal Distribution

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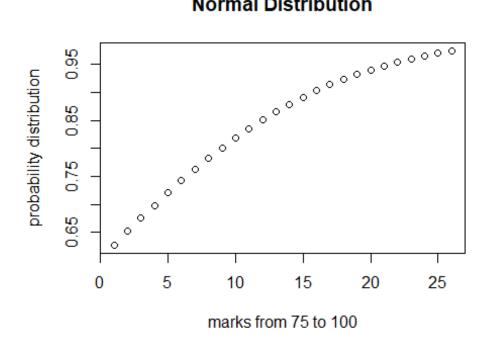
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11-10-2022

```
#Normal Distribution
#01. Suppose x is normally distributed with mean 17.96 and variance 375.67.
#X represents the widget weight. What is the probability that a randomly
#chosen widget weight weighs less than 19?
#To find F(19) = P(X <= 19),
m=17.96
v = 375.67
sd=sqrt(v)
pnorm(19, m, sd)
## [1] 0.521396
#Q2. Assume that the test scores of the exam fit the normal distribution.
#Furthuremore, the mean test score is 70 and the standard deviation is 15.4.
#(a) What is the percentage of students scoring 84 marks or less than in
exam?
#(b) What is the percentage of students scoring 80 marks or more than in
exam?
#(c) What is the percentage of students scoring more than 75 and
#less than equal to 100 in exam?
M=70
SD=15.4
\#(a) To find 100*P(X<=84)
pnorm(84,M,SD)*100
## [1] 81.83489
\#(b) To find 100*P(X>=80)
(1-pnorm(80,M,SD,lower.tail = FALSE))*100
## [1] 74.19441
\#(c) To find 100*P(75<X<=100)
sum(dnorm(75:100,M,SD,log = FALSE))*100
```

```
## [1] 36.12412
F=pnorm(75:100,M,SD)
plot(F,xlab = "marks from 75 to 100",
     ylab = "probability distribution",
     main = "Normal Distribution")
```

Normal Distribution



```
#FUNCTIONS USED
#dnorm() gives the density function
#pnorm() gives the distribution function
```

Practical 8: Correlation Analysis

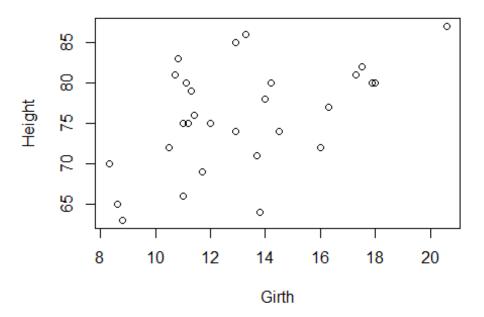
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```
#Correlation Analysis
attach(trees) #calling the pre-existing dataset trees
summary(trees)
##
       Girth
                      Height
                                   Volume
## Min. : 8.30 Min. :63
                               Min.
                                     :10.20
## 1st Qu.:11.05
                   1st Qu.:72 1st Qu.:19.40
## Median :12.90
                   Median: 76 Median: 24.20
## Mean :13.25 Mean :76 Mean :30.17
## 3rd Qu.:15.25
                   3rd Qu.:80 3rd Qu.:37.30
                         :87 Max.
## Max.
         :20.60 Max.
                                      :77.00
#correlation coefficient of girth and height
cor(Girth, Height)
## [1] 0.5192801
#correlation coefficient of volume and height
cor(Volume, Height)
## [1] 0.5982497
#correlation coefficient of girth and volume
cor(Girth, Volume)
## [1] 0.9671194
#scatter plot of height vs girth
plot(Girth, Height, main="scatter plot")
```

scatter plot



Practical 9: Regression Analysis

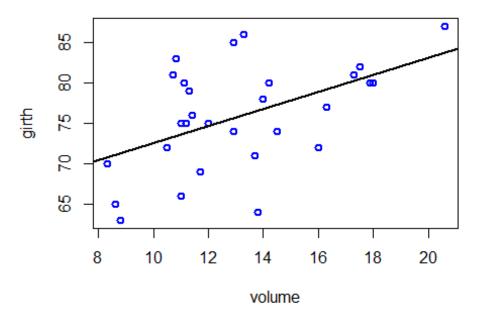
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18-10-2022

```
# Regression Analysis
attach(trees)
summary(trees)
##
       Girth
                       Height
                                    Volume
## Min.
          : 8.30
                   Min.
                          :63
                                Min.
                                       :10.20
## 1st Qu.:11.05
                   1st Qu.:72
                                1st Qu.:19.40
## Median :12.90
                   Median :76
                                Median :24.20
## Mean
         :13.25
                   Mean
                          :76
                                Mean :30.17
## 3rd Qu.:15.25
                   3rd Qu.:80
                                3rd Qu.:37.30
## Max.
         :20.60
                   Max.
                                Max. :77.00
                          :87
model=lm(Girth ~ Volume)
model
##
## Call:
## lm(formula = Girth ~ Volume)
## Coefficients:
## (Intercept)
                    Volume
##
       7.6779
                    0.1846
coefficients(model)
## (Intercept)
                   Volume
##
    7.6778570
                0.1846321
plot(Girth, Height, main = "Regression model Girth vs Volume", col='blue',
     lwd=2,xlab="volume",ylab="girth")
#regression line between girth and height
abline(lm(Height~Girth), lwd=2)
```

Regression model Girth vs Volume



Practical 10: Hypothesis Testing

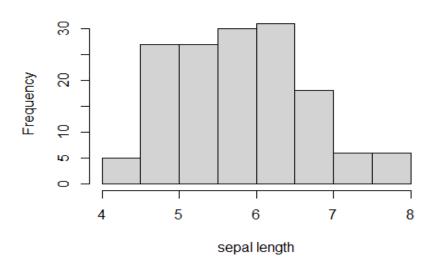
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01-11-2022

```
#Q1. To test mean of data
#sample t-test and confidence intervals are parametric method appropriate
#for examining a single numeric variable
attach(iris) #attach the iris dataset
summary(iris)
##
    Sepal.Length
                   Sepal.Width
                                    Petal.Length
                                                    Petal.Width
## Min.
          :4.300
                          :2.000
                                          :1.000
                                                         :0.100
                   Min.
                                   Min.
                                                  Min.
   1st Qu.:5.100
                   1st Qu.:2.800
                                   1st Qu.:1.600
                                                   1st Qu.:0.300
##
   Median :5.800
                   Median :3.000
                                   Median :4.350
                                                  Median :1.300
## Mean
          :5.843
                   Mean
                          :3.057
                                   Mean
                                         :3.758
                                                  Mean :1.199
   3rd Qu.:6.400
                   3rd Qu.:3.300
                                   3rd Qu.:5.100
                                                   3rd Qu.:1.800
## Max.
          :7.900
                   Max. :4.400
                                   Max. :6.900
                                                  Max.
                                                         :2.500
##
         Species
## setosa
             :50
## versicolor:50
  virginica :50
#(i) Sepal Length of flower is chosen here for mean to be tested
sl=Sepal.Length
#draw the histogram to predict the mean sepal length
hist(sl,xlab="sepal length",main = "Histogram of sepal length")
```

Histogram of sepal length

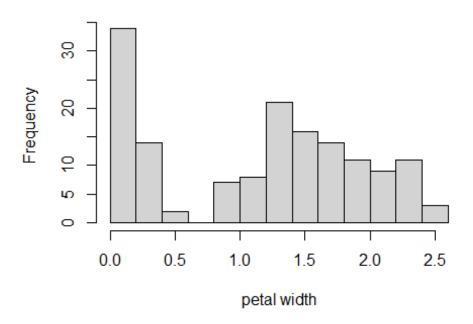


```
# we will test the null hypothesis that mean is 6 ,i.e.,
# Ho:mu=6
# one-sided confidence interval for mu
t.test(sl, mu=6, alt="greater", conf=0.95)
##
## One Sample t-test
##
## data: sl
## t = -2.3172, df = 149, p-value = 0.9891
## alternative hypothesis: true mean is greater than 6
## 95 percent confidence interval:
## 5.731427
                  Inf
## sample estimates:
## mean of x
## 5.843333
#A smaller p-value means that there is stronger evidence in favor of
#the alternative hypothesis.
#There is 95% chance that the mean will lie between 5.731427 to infinity
#p=0.9891 > 0.05, i.e., null hyp is not rejected.
#Also, we can see here that exact value of mean is 5.84333 which is less than 6.
#Thus greater p-value signifies lower evidence to support alternate hypothesis.
#testing for alternate hypothesis being that mean is less than 6
t.test(sl, mu=6, alt="less", conf=0.95)
##
## One Sample t-test
##
## data: sl
## t = -2.3172, df = 149, p-value = 0.01093
## alternative hypothesis: true mean is less than 6
## 95 percent confidence interval:
       -Inf 5.95524
##
## sample estimates:
## mean of x
## 5.843333
\#p=0.01093 < 0.05
#There is 95% chance that the mean will lie between -infinity to 5.95524
#i.e., null hypothesis is rejected.
#Here, we know that exact value of mean is less than 6 which supports
#alternate hypothesis
#two-tailed is by default
#confidence interval can be changed
#store it in an object
TEST_1=t.test(sl, mu=6,alt="greater", conf=0.95)
TEST 2=t.test(sl, mu=6,alt="less", conf=0.95)
```

```
#(ii) Petal Width of flower is chosen here for mean to be tested
pw=Petal.Width

#draw the histogram to predict the mean petal width
hist(pw,xlab="petal width",main = "Histogram of petal width")
```

Histogram of petal width



```
# we will test the null hypothesis that mean is 1 ,i.e.,
# Ho:mu=1
# one-sided confidence interval for mu
t.test(pw, mu=1, alt="less", conf=0.90)
##
##
   One Sample t-test
##
## data: pw
## t = 3.2028, df = 149, p-value = 0.9992
## alternative hypothesis: true mean is less than 1
## 90 percent confidence interval:
        -Inf 1.279448
##
## sample estimates:
## mean of x
## 1.199333
#Here, p=0.9992 > 0.05
#There is 90% chance that the mean will lie between -infinity to 1.279448
#Thus, null hypothesis is rejected.
#testing for alternate hypothesis being that mean is greater than 1
t.test(pw, mu=1, alt="greater", conf=0.90)
##
## One Sample t-test
```

```
##
## data: pw
## t = 3.2028, df = 149, p-value = 0.0008321
## alternative hypothesis: true mean is greater than 1
## 90 percent confidence interval:
## 1.119219
                   Inf
## sample estimates:
## mean of x
## 1.199333
#p=0.00083211 < 0.05, i.e., null hypothesis is not rejected.
#There is 95% chance that the mean will lie between 1.119219 and infinity
#Also, we can see here that exact value of mean is 1.199 > 1
#Thus lower p-value provides stronger evidence in favor of
#the alternative hypothesis.
#Q2. To examine the difference in means of two populations
#we will explore the relationship between mean lung capacity of smokers and
#non-smokers
data1=read.table(file.choose(), header=T) #reads the text file in system
attach(data1)
summary(data1)
##
       LungCap
                                           Height
                                                            Smoke
                           Age
   Min.
           : 0.507
                      Min. : 3.00
                                       Min.
                                             :45.30
                                                        Length:725
##
   1st Qu.: 6.150
                      1st Qu.: 9.00
                                       1st Qu.:59.90
                                                        Class :character
## Median : 8.000
                      Median :13.00
                                       Median :65.40
                                                        Mode :character
   Mean : 7.863
                      Mean :12.33
                                       Mean :64.84
   3rd Qu.: 9.800
                      3rd Qu.:15.00
                                       3rd Qu.:70.30
## Max.
           :14.675
                             :19.00
                                       Max.
                                              :81.80
                      Max.
##
       Gender
                         Caesarean
## Length:725
                        Length:725
                                                LungCap =
                                                             Height =
                                                                    Smoke
                                                                                  Caesarean
                                                        Age
                                                                           Gender
## Class :character
                        Class :character
                                              1
                                                    6.475
                                                                 62.1 no
                                                                           male
## Mode :character
                        Mode :character
                                              2
                                                   10.125
                                                           18
                                                                 74.7 yes
                                                                           female
                                                                                  no
                                              3
                                                    9.550
                                                           16
                                                                           female
                                                                 69.7 no
                                                                                  yes
                                              4
                                                   11.125
                                                           14
                                                                 71.0 no
                                                                           male
                                                                                  no
                                              5
                                                    4.800
                                                                 56.9 no
                                                                           male
                                              6
                                                    6.225
                                                           11
                                                                 58.7 no
                                                                           female
                                                                                  no
                                              7
                                                    4.950
                                                           8
                                                                 63.3 no
                                                                           male
                                                                                  yes
```

#Take two variables for the test, 1st is the LungCap which is numeric variable, #2nd is Smoke which is categorical variable, "yes" means the individual is a #smoker and "no" means he/she is non-smoker

8

9

7.325

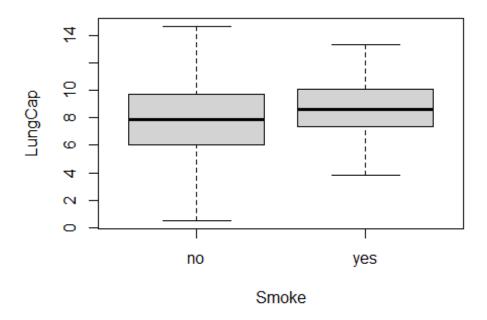
11

70.4 no

70.5 no

male

#examine the plot before conducting the test
boxplot(LungCap ~ Smoke,data1)



```
#test the hypothesis that mean lung capacity of smokers = non smokers
#Ho= mean Lungcap of smokers = mean Lungcap of non smokers
#assume non equal variances
t.test(LungCap~Smoke,data1, mu=0, alt="two.sided", conf=0.95,
       var.equal=FALSE,paired=FALSE)
##
##
   Welch Two Sample t-test
##
## data: LungCap by Smoke
## t = -3.6498, df = 117.72, p-value = 0.0003927
## alternative hypothesis: true difference in means between group no and group y
es is not equal to 0
## 95 percent confidence interval:
   -1.3501778 -0.4003548
## sample estimates:
   mean in group no mean in group yes
##
            7.770188
                              8.645455
#Since p=0.0003927 < 0.05, null hypothesis is rejected,
#i.e., average lung capacity of smokers and non-smokers is not equal
#check variance of two groups
var(data1$LungCap[data1$Smoke=="yes"])
## [1] 3.545292
var(data1$LungCap[data1$Smoke=="no"])
## [1] 7.431694
```