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EXPERIMENT 06

A manufacturer of pins knows that 2% of his products are defective. If he sells pins in boxes of 20 and find the number of boxes containing

(i) at least 2 defective

(ii) exactly 2 defective

(iii) at most 2 defective

pins in a consignment of 1000 boxes

(iv) plot the distribution

#(v) $E(x)$

#(vi) variance of X ?

number of trials

$m = 20$

m

#probability of success

$ps = 0.02$

poisson parameter

$\lambda = m \cdot ps$

λ

(i) at least 2 defective

$p1 = \sum(dpois(2:m, \lambda))$

$p1$

(i) number of boxes containing at least two defectives

$\text{round}(1000 \cdot p1)$

exactly two defectives

$p2 = dpois(2, \lambda)$

p2

(ii) number of boxes containing exactly 2 defectives

round(1000*p2)

at most 2 defectives

p3 = sum(dpois(0:2, lambda))

p3

(iii) number of boxes containing at most two defectives

round(1000*p3)

(iv) plot the distribution

x1 = 0:m

px1 = dpois(x1, lambda)

px1

plot(x1, px1, type="h", xlab="values of x", ylab = "probability distribution of x", main="Poisson distribution")

#(v) E(x)

Ex1 = weighted.mean(x1, px1)

Ex1

#(vi) variance of x

varx1 = weighted.mean(x1*x1, px1)-(weighted.mean(x1, px1))^2

varx1

#Normal distribution

A company finds that the time taken by one of its engineers to complete or repair job has a normal distribution

with mean 20 minutes and SD 5 minutes. State what proportion of the jobs take

```
# (i) Less than 15 minutes
# (ii) More than 25 minutes
# (iii) Between 15 and 25
minutes # (iv) Plot the
distribution
# (v) Table the distribution
```

```
# Generating the data
```

```
x = seq(0, 40)
```

```
x
```

```
# find the density function of x
```

```
y = dnorm(x, mean=20, sd = 5)
```

```
y
```

```
#plot the normal distribution curve
```

```
plot(x,y,type='l')
```

```
#Proportion of jobs take less than 15 minutes
```

```
p1 = pnorm(15,mean=20, sd = 5)
```

```
p1
```

```
x2 = seq(0,15)
```

```
x2
```

```
y2 = dnorm(x2, mean=20, sd = 5)
```

```
y2
```

```
polygon(c(0, x2, 15), c(0,y2,0), col='yellow')
```

```
#Proportion of jobs that take more than 25 minutes
```

```
p2 = pnorm(40, mean=20, sd=5)- pnorm(25,mean=20,sd=5)
```

```
p2
```

```
x1 = seq(25,40)
```

```
x1
```

```
y1 = dnorm(x1, mean=20, sd=5)
```

```
y1
```

```
polygon(c(25,x1,40), c(0, y1, 0), col='red')
```

```
#Proportion of jobs take between 15 and 25
```

```
p3 = pnorm(25, mean=20, sd=5) - pnorm(15, mean=20, sd=5)
```

```
p3
```

```
x3 = seq(15,25)
```

```
x3
```

```
y3 = dnorm(x3, mean=20, sd=5)
```

```
y3
```

```
polygon(c(15,x3,25), c(0, y3, 0), col='green')
```

```
#Probabilty distribution
```

```
data.frame(p1, p2, p3)
```

OUTPUT

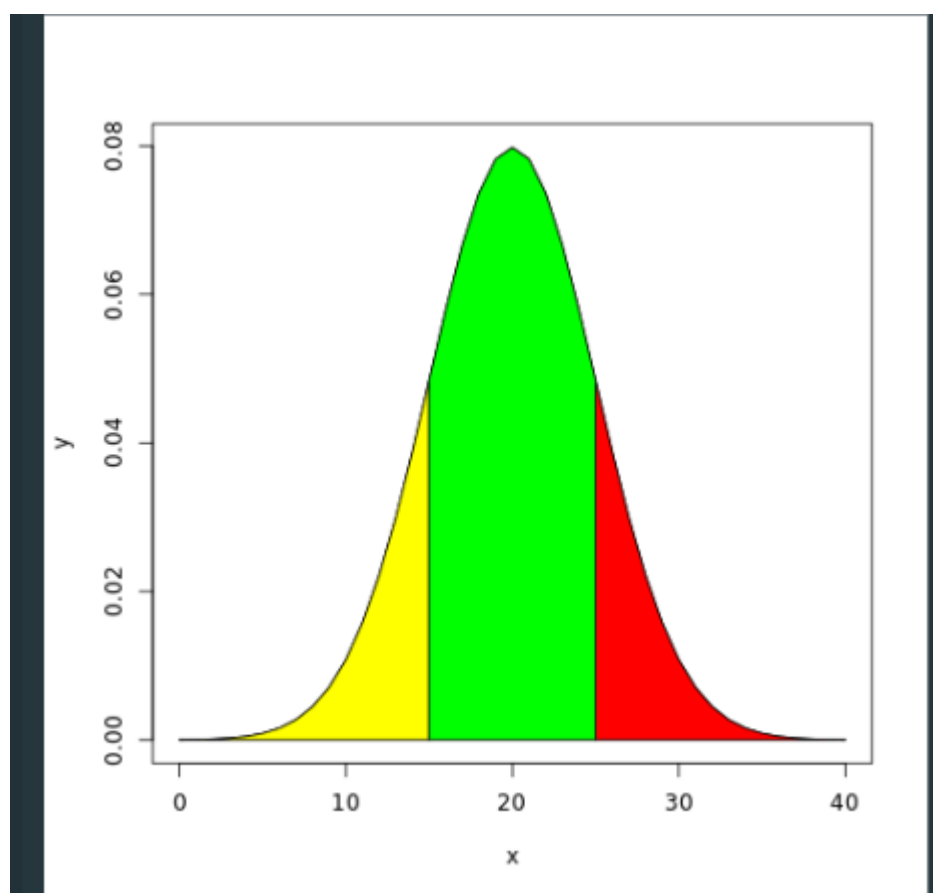
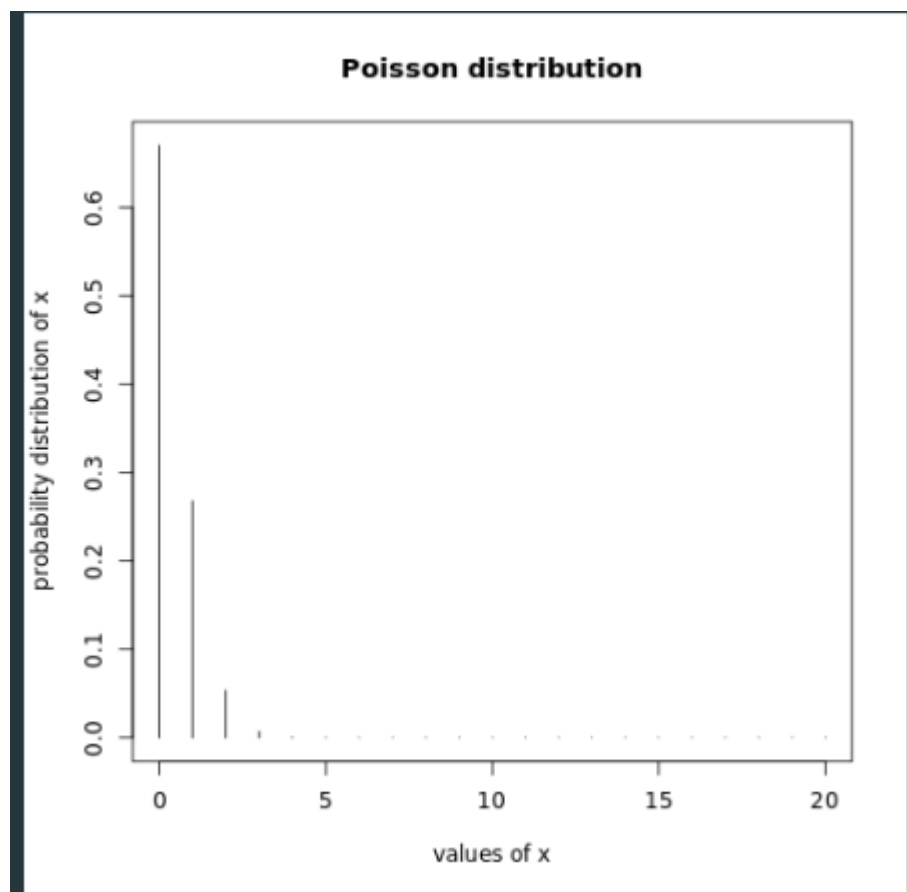
Output

```
[1] 20
[1] 0.4
[1] 0.06155194
[1] 62
[1] 0.0536256
[1] 54
[1] 0.9920737
[1] 992
[1] 6.703200e-01 2.681280e-01 5.362560e-02 7.150080e-03 7.150080e-04
[6] 5.720064e-05 3.813376e-06 2.179072e-07 1.089536e-08 4.842383e-10
[11] 1.936953e-11 7.043466e-13 2.347822e-14 7.224067e-16 2.064019e-17
[16] 5.504051e-19 1.376013e-20 3.237677e-22 7.194838e-24 1.514703e-25
[21] 3.029406e-27
[1] 0.4
[1] 0.4
[1] 0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
[26] 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
[1] 2.676605e-05 5.838939e-05 1.223804e-04 2.464438e-04 4.768176e-04
[6] 8.863697e-04 1.583090e-03 2.716594e-03 4.478906e-03 7.094919e-03
[11] 1.079819e-02 1.579003e-02 2.218417e-02 2.994549e-02 3.883721e-02
[16] 4.839414e-02 5.793831e-02 6.664492e-02 7.365403e-02 7.820854e-02
[21] 7.978846e-02 7.820854e-02 7.365403e-02 6.664492e-02 5.793831e-02
[26] 4.839414e-02 3.883721e-02 2.994549e-02 2.218417e-02 1.579003e-02
[31] 1.079819e-02 7.094919e-03 4.478906e-03 2.716594e-03 1.583090e-03
[36] 8.863697e-04 4.768176e-04 2.464438e-04 1.223804e-04 5.838939e-05
[41] 2.676605e-05
[1] 0.1586553
[1] 0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15
[1] 2.676605e-05 5.838939e-05 1.223804e-04 2.464438e-04 4.768176e-04
[6] 8.863697e-04 1.583090e-03 2.716594e-03 4.478906e-03 7.094919e-03
[11] 1.079819e-02 1.579003e-02 2.218417e-02 2.994549e-02 3.883721e-02
```

```
[6] 8.863697e-04 1.583090e-03 2.716594e-03 4.478906e-03 7.094919e-03
[11] 1.079819e-02 1.579003e-02 2.218417e-02 2.994549e-02 3.883721e-02
[16] 4.839414e-02
[1] 0.1586236
[1] 25 26 27 28 29 30 31 32 33 34 35 36 37 38 39 40
[1] 4.839414e-02 3.883721e-02 2.994549e-02 2.218417e-02 1.579003e-02
[6] 1.079819e-02 7.094919e-03 4.478906e-03 2.716594e-03 1.583090e-03
[11] 8.863697e-04 4.768176e-04 2.464438e-04 1.223804e-04 5.838939e-05
[16] 2.676605e-05
[1] 0.6826895
[1] 15 16 17 18 19 20 21 22 23 24 25
[1] 0.04839414 0.05793831 0.06664492 0.07365403 0.07820854 0.07978846
[7] 0.07820854 0.07365403 0.06664492 0.05793831 0.04839414
      p1      p2      p3
1 0.1586553 0.1586236 0.6826895

[Execution complete with exit code 0]
```

FIGURES



EXPERIMENT 7

#Aim - to understand the testing of hypothesis for large sample tests using R functions

#Suppose the mean weight of King Penguins found in an Antarctic colony last year was 15.4 KG. In a sample of 35 penguins same time this year in the same colony, the mean penguins

#weight is 14.6 kg. Assume the population standard deviation is 2.5 kg. At 0.05 significance level, can we reject the null hypothesis that the mean penguin weight does not differ from last year?

#xbar = 14.6

#mu = 15.4

#alpha = 0.05

#sigma = 2.5

#n = 35

#input the sample mean

xbar=14.6

xbar

#Input the population mean

mu0=15.4

mu0

#Input the standard deviation

sigma = 2.5

sigma


```
#Input the sample size
```

```
n = 35
```

```
#Test statistic
```

```
z = (xbar - mu0)/(sigma/sqrt(n)) z
```

```
#Level of significance
```

```
alpha = 0.05
```

```
alpha
```

```
#Two tailed test critical value
```

```
zhalfalpha = qnorm(1 -
```

```
(alpha/2)) zhalfalpha
```

```
c(-zhalfalpha, zhalfalpha)
```

```
#To find p-value
```

```
pval = 2*pnorm(z)
```

```
pval
```

```
#conclusion
```

```
if (pval > alpha) {
```

```
  print("Accept null
```

```
hypothesis") } else {
```

```
  print("Reject null hypothesis")
```

}

#Problem-2

$$z = (p - P) / (\sqrt{PQ/n})$$

The fatality rate of typhoid patients is believed to be 17.26%. In a certain year 640 patients suffering from typhoid were treated in a metropolitan hospital and only 63 patients died.

Can you consider the efficient? Level of significance is 5%

Input the data

Size of the sample

$$n = 640$$

n

Sample proportion

$$p = 63/n$$

p

Population proportion

$$P = 0.1726$$

P

Probability of failure

$$Q = 1 - P$$

Q

```
#Test statistic
```

```
z = (Sprop-Pprop)/sqrt(Pprop*Q/n)
```

```
z
```

```
#critical value
```

```
E = qnorm(.975)
```

```
E
```

```
#Critical region
```

```
c(-E, E)
```

```
#Confidence interval
```

```
Sprop+c(-E,E)*sqrt(Pprop*(1-Pprop)/  
n)
```

```
#Conclusion
```

```
if (z>-E && z < E) {
```

```
  print("Hospital is not efficient")
```

```
} else {
```

```
  print("Hospital is efficient")
```

```
}
```

OUTPUT

```
[1] 14.6  
[1] 15.4  
[1] 2.5  
[1] -1.893146  
[1] 0.05  
[1] 1.959964  
[1] -1.959964 1.959964  
[1] 0.05833852  
[1] "Accept null hypothesis"
```

```
[1] 640
[1] 0.0984375
[1] 0.1726
[1] 0.8274
[1] -4.964736
[1] 1.959964
[1] -1.959964 1.959964
[1] 0.06915985 0.12771515
[1] "Hospital is efficient"
```

EXPERIMENT8

#Testing of hypothesis

In a random sample of size 500, them mean is found to be 20. In another independent sample

of size 400, the mean is 15. Could the samples have been drawn from the same population with SD 4?

#input the sample mean

xbar = 20

xbar

ybar=15

ybar

input the standard deviation

sigma = 4

input the sample size

n1 = 500

n1

```
n2 = 400
```

```
n2
```

```
# Test statistic
```

```
z = (xbar-ybar)/(sigma*sqrt((1/n1)+(1/n2)))
```

```
z
```

```
#Level of significance
```

```
alpha = 0.05
```

```
alpha
```

```
# Two tailed critical value
```

```
zalpha = qnorm(1 - (alpha/2))
```

```
zalpha
```

```
#conclusion
```

```
if (z <= zalpha)
```

```
{
```

```
  print("Accept Null Hypothesis")
```

```
}else
```

```
{
```

```
  print("Reject Null Hypothesis")
```

```
}
```

```
#-----
```

Testing of hypothesis - Two sample proportion test

$z = (p1 - p2)/\sqrt{PQ(1/n1 + 1/n2)}$

In a large city A, 20% of a random sample of 900 school boys had a slight

physical defect. In another large city B, 18.5% of a random sample of 1600 boys had the same defect.

Is the difference between the proportions significant

Input the sample proportions

p1 = 0.20

p1

p2 = 0.185

p2

Input the sample sizes

n1 = 900

n1

n2 = 1600

n2

#Population proportion

$P = (n1*p1 + n2*p2)/(n1+n2)$

P

$Q = 1 - P$

Q

#Test statistic

```
z = (p1-p2)/sqrt(P*Q*((1/n1)+(1/
n2))) z
```

```
alpha=0.05
```

```
alpha
```

```
#Two-tailed critical value
```

```
zalpha = qnorm(1 - (alpha/2)) zalpha
```

```
# conclusion
```

```
if (z<=zalpha) {print("Accept Null Hypothesis")} else {print("Reject Null
Hypothesis")}
```

OUTPUT

Output

```
[1] 20
[1] 15
[1] 500
[1] 400
[1] 18.6339
[1] 0.05
[1] 1.959964
[1] "Reject Null Hypothesis"
[1] 0.2
[1] 0.185
[1] 900
[1] 1600
[1] 0.1904
[1] 0.8096
[1] 0.9169249
[1] 0.05
[1] 1.959964
[1] "Accept Null Hypothesis"
```

```
[Execution complete with exit code 0]
```

EXPERIMENT 9

#T-Test

$t = (\bar{x} - \mu) / (\sigma / \sqrt{n})$

Two independent samples of sizes 8 and 7 are given contained the following values

Sample 1 - 19 17 15 21 16 18 16 14

Sample 2 - 15 14 15 19 15 18 16 20

Is the difference between the sample means significant

Problem 1

#input the data

sample1=c(19, 17, 15, 21, 16, 18, 16, 14)

sample1

sample2 = c(15, 14, 15, 19, 15, 18, 16, 20)

sample2

output using t-distribution

t = t.test(sample1, sample2)

t

#test-statistic

cv = t\$statistic

cv

#critical value

tv = qt(0.975, 14)

tv

#conclusion

if (cv <= tv) {print("Accept Ho")} else {print("Reject Ho")}

Problem 2

The following data relate to the marks obtained by 10 students in two test, one held at the beginning

of a year and the other at the end of the year after intensive coaching. Do the data indicate that the

students have got benefited by coaching

Test-1 19 17 15 21 16 18 16 14 19 20

Test-2 15 14 15 19 15 18 16 20 22 19

#input the data

test1 = c(19, 17, 15, 21, 16, 18, 16, 14, 19, 20)

test1

test2 = c(15, 14, 15, 19, 15, 18, 16, 20, 22, 19)

test2

output using t-distribution

t2 = t.test(test1, test2)

t2

#test-statistic

cv2 = t2\$statistic

cv2

```
#critical value
```

```
tv2 = qt(0.975, 18)
```

```
tv2
```

```
#conclusion
```

```
if (cv2 <= tv2) {print("Accept Ho")} else {print("Reject Ho")}
```

```
#-----
```

```
#F Test
```

```
# Problem: 3 (F-Test)
```

```
#Two independent samples of sizes 8 and 7 contained the following values:
```

```
# Sample 1: 19 17 15 21 16 18 16 14
```

```
# Sample 2: 15 14 15 19 15 18 16 20
```

```
sample1 = c(19, 17, 15, 21, 16, 18, 16, 14)
```

```
sample1
```

```
sample2 = c(15, 14, 15, 19, 15, 18, 16, 20)
```

```
sample2
```

```
#output using t-distribution
```

```
f = var.test(sample1, sample2)
```

```
f
```

```
#F Test to compare two variances
```

```

# Test statistic
cv = f$statistic
cv

tv = qf(0.95, 7, 7)
tv

#conclusion
if (cv <= tv) {print("Accept Ho")} else {print("Reject Ho")}

```

OUTPUT

Output

```

[1] 19 17 15 21 16 18 16 14
[1] 15 14 15 19 15 18 16 20

Welch Two Sample t-test

data: sample1 and sample2
t = 0.44721, df = 13.989, p-value = 0.6616
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -1.898128  2.898128
sample estimates:
mean of x mean of y
    17.0    16.5

t
0.4472136
[1] 2.144787
[1] "Accept Ho"
[1] 19 17 15 21 16 18 16 14 19 20
[1] 15 14 15 19 15 18 16 20 22 19

Welch Two Sample t-test

data: test1 and test2
t = 0.18042, df = 17.555, p-value = 0.8589
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
 -2.133217  2.533217
sample estimates:

```

```

sample estimates:
mean of x mean of y
    17.5    17.3

      t
0.1804154
[1] 2.100922
[1] "Accept Ho"
[1] 19 17 15 21 16 18 16 14
[1] 15 14 15 19 15 18 16 20

      F test to compare two variances

data: sample1 and sample2
F = 1.0588, num df = 7, denom df = 7, p-value = 0.9418
alternative hypothesis: true ratio of variances is not equal to 1
95 percent confidence interval:
 0.2119805 5.2887274
sample estimates:
ratio of variances
      1.058824

      F
1.058824
[1] 3.787044
[1] "Accept Ho"

```

EXPERIMENT 10

Problem-1

#Five coins are tossed 256 times. The number of heads observed by binomial distribution is given by below.

Examine if the coins are unbiased by employing chi-square goodness of fit.

No of heads 0 1 2 3 4 5

Frequency 5 35 75 84 45 12

number of coins

n = 5

n

level of significance

alpha = 0.05

alpha

N = 256

N

P = 0.5

P

x = c(0:n);x

obf = c(5, 35, 75, 84, 45, 12)

obf

exf = (dbinom(x, n, P)*256)

exf

sum(obf)

sum(exf)

chisq <- sum((obf-exf)^2/exf)

chisq

cv = chisq;cv

#Critical value using chisq-distribution

tv = qchisq(1-alpha, n);tv

```
#Hypothesis conclusion
```

```
if (cv <= tv) {print("Accept H0/Fit is good")} else {print("Reject H0/Fit is not  
good.")}
```

```
# Problem 2
```

```
# From the following information state whether the condition of the child is  
associated with the condition of the house
```

```
# Condition of the child   Condition of the house clean   Condition of the  
house dirty
```

```
#   Clean                      69                      51
```

```
#   Fairly Clean               81                      20
```

```
#   Dirty                      35                      44
```

```
# Input the data
```

```
data <- matrix(c(69, 51, 81, 20, 35, 44), ncol=2, byrow=T)
```

```
data
```

```
l = length(data);l
```

```
#Output by chisq distribution
```

```
cv = chisq.test(data)
```

```
cv
```

```
#p-value
```

```
cv = cv$p.value
```

```
cv
```

alpha = 0.05

#Hypothesis

conclusion

```
if (cv > alpha) {print("Attributes are independent")} else {print("Attributes are not independent")}
```

OUTPUT

Output

```
[1] 5
[1] 0.05
[1] 256
[1] 0.5
[1] 0 1 2 3 4 5
[1] 5 35 75 84 45 12
[1] 8 40 80 80 40 8
[1] 256
[1] 256
[1] 4.8875
[1] 4.8875
[1] 11.0705
[1] "Accept H0/Fit is good"
      [,1] [,2]
[1,]   69   51
[2,]   81   20
[3,]   35   44
[1] 6

      Pearson's Chi-squared test

data:  data
X-squared = 25.629, df = 2, p-value = 2.721e-06

[1] 2.72114e-06
[1] "Attributes are not independent"
```

EXPERIMENT 11

Completely randomised design

Problem: A car rental agency, which uses 5 different brands of tyres in the process of deciding the brand of tyre

to purchase as standard equipment for its fleet, finds that each of 5 tyres of each brand

last the following number of kilometers

A B C D E

36 46 35 45 41

37 39 42 36 39

42 35 37 39 37

38 37 43 35 35

47 43 38 32 38

Test the hypothesis that the five brands have almost the same average life

One way ANOVA

Test of tyres

A = c(36, 37, 42, 38, 47)

B = c(46, 39, 35, 37, 43)

C = c(35, 42, 37, 43, 38)

D = c(45, 36, 39, 35, 32)

E = c(41, 39, 37, 35, 38)

group <- data.frame(cbind(A,B,C,D,E))

group

summary(group)

#stack vector from data frame

stgr <- stack(group);stgr

#Completely randomized design


```
crd <- aov (values~ind, data=stgr)
```

```
#ANOVA table
```

```
summary(crd)
```

```
# There is no difference in the average life of
```

```
tyres # Visualization of data
```

```
boxplot(group, ylab="Average life of tyres in kilometers", main="Brands of Tyres")
```

```
# Two way ANOVA
```

```
# Randomized block design
```

```
# The following table gives monthly sales in thousand rupees of a certain firm in the 3 states by its four salesmen
```

```
# States    Salesmen
```

```
#      I  II  III  IV
```

```
# A      6  5   3   8
```

```
# B      8  9   6   5
```

```
# C     10  7   8   7
```

```
# Setup the analysis of variance table and test whether there is any significant difference
```

```
#(i) between the salesmen
```

```
# (ii) between sales in the states
```

```
#Monthly sales of the states
```

```
StateA = c(6,5,3,8)
```

```
StateA
```

```
StateB = c(8, 9, 6, 5)
```

```
StateB
```

```
StateC = c(10, 7, 8, 7)
```

```
StateC
```

```
#Frame the data set
```

```
Group <- data.frame(cbind(StateA, StateB, StateC))
```

```
Group
```

```
Sales = c(t(as.matrix(Group))); Sales
```

```
f = c("State A", "State B", "State C")
```

```
f
```

```
g = c("Salesman 1", "Salesman 2", "Salesman 3", "Salesman 4")
```

```
g
```

```
# Number of columns
```

```
k = ncol(Group)
```

```
k
```

```
# Number of rows
```

```
n = nrow(Group)
```

```
n
```

```
#Generate factor levels of states
```

```
States = gl(k, 1, n*k, factor(f))
```

States

```
# Generate factor levels of Salesmen
```

```
Salesmen = gl(n, k, n*k, factor(g))
```

```
Salesmen
```

```
#ANOVA Table
```

```
anova = aov(Sales~States +
```

```
Salesmen) summary(anova)
```

```
# Reject the states
```

```
# Reject the salesmen
```

OUTPUT

Output

	A	B	C	D	E
1	36	46	35	45	41
2	37	39	42	36	39
3	42	35	37	39	37
4	38	37	43	35	35
5	47	43	38	32	38

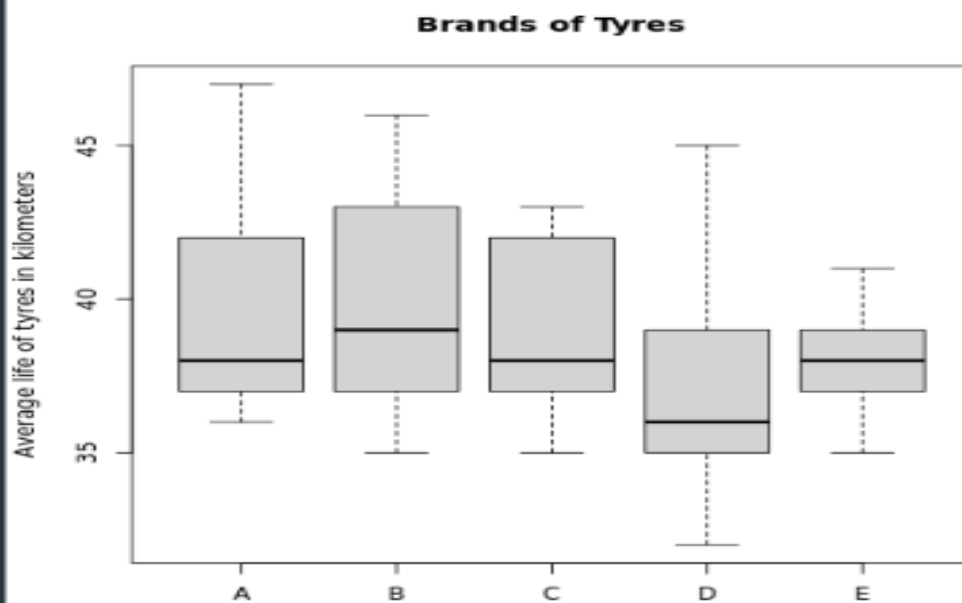
	A	B	C	D	E
Min.	:36	Min. :35	Min. :35	Min. :32.0	Min. :35
1st Qu.:	37	1st Qu.:37	1st Qu.:37	1st Qu.:35.0	1st Qu.:37
Median :	38	Median :39	Median :38	Median :36.0	Median :38
Mean :	40	Mean :40	Mean :39	Mean :37.4	Mean :38
3rd Qu.:	42	3rd Qu.:43	3rd Qu.:42	3rd Qu.:39.0	3rd Qu.:39
Max. :	47	Max. :46	Max. :43	Max. :45.0	Max. :41
values	ind				

1	36	A
2	37	A
3	42	A
4	38	A
5	47	A
6	46	B
7	39	B
8	35	B
9	37	B
10	43	B
11	35	C
12	42	C
13	37	C
14	43	C
15	38	C
16	45	D
17	36	D
18	39	D
19	35	D
20	32	D
21	41	E
22	39	E
23	37	E
24	35	E
25	38	E

```

      Df Sum Sq Mean Sq F value Pr(>F)
ind      4    27.4      6.86   0.422  0.791
Residuals 20   325.2     16.26
[1] 6 5 3 8
[1] 8 9 6 5
[1] 10 7 8 7
      StateA StateB StateC
1         6         8      10
2         5         9       7
3         3         6       8
4         8         5       7
[1] 6 8 10 5 9 7 3 6 8 8 5 7
[1] "State A" "State B" "State C"
[1] "Salesman 1" "Salesman 2" "Salesman 3" "Salesman 4"
[1] 3
[1] 4
[1] State A State B State C State A State B State C State A State B State C
[10] State A State B State C
Levels: State A State B State C
[1] Salesman 1 Salesman 1 Salesman 1 Salesman 2 Salesman 2 Salesman 2
[7] Salesman 3 Salesman 3 Salesman 3 Salesman 4 Salesman 4 Salesman 4
Levels: Salesman 1 Salesman 2 Salesman 3 Salesman 4
      Df Sum Sq Mean Sq F value Pr(>F)
States    2 12.667      6.333   1.839  0.238
Salesmen   3  8.333      2.778   0.806  0.535
Residuals  6 20.667      3.444

```



ALL CODES EXP06 - EXP11

```

#no. of trials
m=20
m

#probability of success
ps=0.02

#poisson parameter
lambda=m*ps
lambda

#atleast 2 defectives
p1=sum(dpois(2:m,lambda))
p1

#(i) no. of boxes containing atleast 2 defectives
p2=dpois(2,lambda)
p2

#(ii) no. of boxes containing exactly 2 defectives
round(1000*p2)

#atmost 2 defectives
p3=sum(dpois(0:2,lambda))
p3

#(iv)plot the distribution
x1=0:m
px1=dpois(x1,lambda)
plot(x1,px1,type="h",xlab="values of x",ylab="probability of x",main="poisson
distribution")

#(v)E(x)
Ex1=weighted.mean(x1,px1)
Ex1

#(vi) variance of x
varx1=weighted.mean(x1*x1,px1)-(weighted.mean(x1,px1))^2
varx1

#SAMPLE PROBLEM:
#A company finds the time taken by one of its engineers to complete or repair
job has a normal distribution with mean 20 minutes and S.D 5 minutes. state
what proportion of job takes:
#(i) less than 15 minutes (ii) more than 25 minutes (iii) between 15 & 25
minutes (iv)plot the distribution (v) table the distribution

```

```

#generating the data x
x= seq(0,40)
x

#find the density function of x
y=dnorm(x,mean=20,sd=5)
y

#plot the normal distribution curve
plot(x,y,type='l')

#preparation of jobs take less than 15 minutes
p1=pnorm(15,mean=20,sd=5)
p1

x2=seq(0,15)
x2

y2=dnorm(x2,mean=20,sd=5)
y2
polygon(c(0,x2,15),c(0,y2,0),col='yellow')

#preparation of jobs take more than 25 mins
p2=pnorm(40,mean=20,sd=5)-pnorm(25,mean=20,sd=5)
p2

x1=seq(25,40)
x1
y1=dnorm(x1,mean=20,sd=5)
y1
polygon(c(25,x1,40),c(0,y1,0),col='red')

#proportion of jobs taken b/w 15 and 25 minutes
p3=pnorm(25,mean=20,sd=5)-pnorm(15,mean=20,sd=5)
p3

x3=seq(15,25)
x3

y3=dnorm(x3,mean=20,sd=5)
y3

polygon(c(15,x3,25),c(0,y3,0),col='green')

#probability distribution
data.frame(p1,p2,p3)

#EXPERIMENT 7

```

```

#Input the sample mean
xbar=14.6
xbar

#input the population mean

mu0=15.4
mu0

#input standard deviation
sigma=2.5
sigma

#input the sample size
n=35
n

#test statistics
z=(xbar-mu0)/(sigma/sqrt(n))
z

#level of significance
alpha=0.05
alpha

#two-tailed critical value
zhalfalpha=qnorm(1-(alpha/2))
zhalfalpha

c(-zhalfalpha,zhalfalpha)

#to find p-value
pval=2*pnorm(z)
pval

#conclusion
if(pval>alpha){print("accept null hypothesis")} else{print("reject null
hypothesis")}

#Testing Hypothesis -Large sample proportion test

#input the data
#size of the sample

n=640
n

```



```

#sample proportion
Sprop=63/n
Sprop

#population proportion
Pprop=0.1726
Pprop

#probability of failure
Q=1-Pprop
Q

#test statistics
z=(Sprop-Pprop)/sqrt(Pprop*Q/n)
z

E=qnorm(.975)

#critical region
c(-E,E)

#confidence interval
Sprop+c(-E,E)*sqrt(Pprop*(1-Pprop)/n)

#conclusion
if(z>-E&& z<E){print("hospital is not efficient")} else{print("hospital is
efficient")}

#EXPERIMENT 8
#QUESTION 1
P1=0.20
P1
p2=0.185
p2
n1=900
n1
n2=1600
n2
P=(n1*p1+n2*p2)/(n1+n2)
P
Q=1-P
Q
z=(p1-p2)/sqrt(P*Q*((1/n1)+(1/n2)))
z
alpha=0.05
alpha
#two tailed test value
zalpha=qnorm(1-(alpha/2))

```

```

zalpha
#conclusion:testing of hypothesis fpr large sample tests using R functions has
been explored and concluded
if(z<=zalpha){print("accept null hypothesis")} else{print("reject null
hypothesis")}

#QUESTION 2
sample1=c(19,17,15,21,16,18,16,14)
sample1
sample2=c(15,14,15,19,15,18,16,20)
sample2

#output using t-distribution
t=t.test(sample1,sample2)
t

#test-statistics
cv=t$statistic
cv

#critical value
tv=qt(0.975,14)
tv

#conclusion
if(cv<=tv){print("accept H0")} else{print("reject H0")}

#QUESTION 3

#variance test or F-test
sample1=c(19,17,15,21,16,18,16,14)
sample1
sample2=c(15,14,15,19,15,18,16,20)
sample2
#output using t- distribution
f=var.test(sample1,sample2)
f
#test statistics
tv=qf(0.95,7,7)
tv
#conclusion
if(cv<=tv){print("accept H0")} else{print("reject H0")}

#QUESTION 4
#Testing of hypothesis
# In a random sample of size 500, them mean is found to be 20. In another
independent sample

```

of size 400, the mean is 15. Could the samples have been drawn from the same population with SD 4?

#input the sample mean

xbar = 20

xbar

ybar=15

ybar

input the standard deviation

sigma = 4

input the sample size

n1 = 500

n1

n2 = 400

n2

Test statistic

z = (xbar-ybar)/(sigma*sqrt((1/n1)+(1/n2)))

z

#Level of significance

alpha = 0.05

alpha

Two tailed critical value

zalpha = qnorm(1 - (alpha/2))

zalpha

#conclusion

if (z <= zalpha)

{

 print("Accept Null Hypothesis")

}else

{

 print("Reject Null Hypothesis")

}

#EXPERIMENT 9

#T-Test

t = (xbar - u)/(sigma/sqrt(n))

Two independent samples of sizes 8 and 7 are given contained the following values

Sample 1 - 19 17 15 21 16 18 16 14

Sample 2 - 15 14 15 19 15 18 16 20

```

# Is the difference between the sample means significant
# Problem 1
#input the data
sample1=c(19, 17, 15, 21, 16, 18, 16, 14)
sample1

sample2 = c(15, 14, 15, 19, 15, 18, 16, 20)
sample2

# output using t-distribution
t = t.test(sample1, sample2)
t

#test-statistic
cv = t$statistic
cv

#critical value
tv = qt(0.975, 14)
tv

#conclusion
if (cv <= tv) {print("Accept Ho")} else {print("Reject Ho")}

# Problem 2
# The following data relate to the marks obtained by 10 students in two test,
one held at the beginning
# of a year and the other at the end of the year after intensive coaching. Do
the data indicate that the
# students have got benefited by coaching
# Test-1 19 17 15 21 16 18 16 14 19 20
# Test-2 15 14 15 19 15 18 16 20 22 19

#input the data
test1 = c(19, 17, 15, 21, 16, 18, 16, 14, 19, 20)
test1

test2 = c(15, 14, 15, 19, 15, 18, 16, 20, 22, 19)
test2

# output using t-distribution
t2 = t.test(test1, test2)
t2

#test-statistic
cv2 = t2$statistic
cv2

```

```

#critical value
tv2 = qt(0.975, 18)
tv2

#conclusion
if (cv2 <= tv2) {print("Accept Ho")} else {print("Reject Ho")}

#-----
#F Test
# Problem: 3 (F-Test)
#Two independent samples of sizes 8 and 7 contained the following values:
# Sample 1: 19 17 15 21 16 18 16 14
# Sample 2: 15 14 15 19 15 18 16 20

sample1 = c(19, 17, 15, 21, 16, 18, 16, 14)
sample1

sample2 = c(15, 14, 15, 19, 15, 18, 16, 20)
sample2

#output using t-distribution
f = var.test(sample1, sample2)
f

#F Test to compare two variances
# Test statistic
cv = f$statistic
cv

tv = qf(0.95, 7, 7)
tv

#conclusion
if (cv <= tv) {print("Accept Ho")} else {print("Reject Ho")}

#EXPERIMENT 10
# Problem-1
#Five coins are tossed 256 times. The number of heads observed by binomial
distribution is given by below.
# Examine if the coins are unbiased by employing chi-square goodness of fit.
# No of heads 0 1 2 3 4 5
# Frequency 5 35 75 84 45 12

# number of coins
n = 5
n

# level of significance

```

```

alpha = 0.05
alpha

N = 256
N

P = 0.5
P

x = c(0:n);x

obf = c(5, 35, 75, 84, 45, 12)
obf

exf = (dbinom(x, n, P)*256)
exf

sum(obf)

sum(exf)

chisq <- sum((obf-exf)^2/exf)
chisq
cv = chisq;cv

#Critical value using chisq-distribution
tv = qchisq(1-alpha, n);tv

#Hypothesis conclusion
if (cv <= tv) {print("Accept H0/Fit is good")} else {print("Reject H0/Fit is
not good.")}

# Problem 2
# From the following information state whether the condition of the child is
associated with the condition of the house
# Condition of the child    Condition of the house clean    Condition of the
house dirty
#      Clean                69                                51
#      Fairly Clean         81                                20
#      Dirty                35                                44

# Input the data
data <- matrix(c(69, 51, 81, 20, 35, 44), ncol=2, byrow=T)
data

l = length(data);l

#Output by chisq distribution

```

```

cv = chisq.test(data)
cv

#p-value
cv = cv$p.value
cv
alpha = 0.05
#Hypothesis conclusion
if (cv > alpha) {print("Attributes are independent")} else {print("Attributes
are not independent")}

#EXPERIMENT 11

# Completely random mixed design

# Problem: A car rental agency, which uses 5 different brands of tyres in the
process of deciding the brand of tyre
# to purchase as standard equipment for its fleet, finds that each of 5 tyres
of each brand
# last the following number of kilometers
# A B C D E
# 36 46 35 45 41
# 37 39 42 36 39
# 42 35 37 39 37
# 38 37 43 35 35
# 47 43 38 32 38
# Test the hypothesis that the five brands have almost the same average life
# One way ANOVA
# Test of tyres
A = c(36, 37, 42, 38, 47)
B = c(46, 39, 35, 37, 43)
C = c(35, 42, 37, 43, 38)
D = c(45, 36, 39, 35, 32)
E = c(41, 39, 37, 35, 38)
group <- data.frame(cbind(A,B,C,D,E))
group

summary(group)
#stack vector from data frame
stgr <- stack(group);stgr

#Completely randomized design
crd <- aov (values~ind, data=stgr)
#ANOVA table
summary(crd)
# There is no difference in the average life of tyres
# Visualization of data

```

```

boxplot(group, ylab="Average life of tyres in kilometers", main="Brands of Tyres")

# Two way ANOVA
# Randomized block design
# The following table gives monthly sales in thousand rupees of a certain firm
in the 3 states by its four salesmen

# States      Salesmen
#           I    II   III   IV
#   A        6    5    3    8
#   B        8    9    6    5
#   C       10    7    8    7

# Setup the analysis of variance table and test whether there is any
significant difference
#(i) between the salesmen
# (ii) between sales in the states

#Monthly sales of the states
StateA = c(6,5,3,8)
StateA

StateB = c(8, 9, 6, 5)
StateB

StateC = c(10, 7, 8, 7)
StateC

#Frame the data set
Group <- data.frame(cbind(StateA, StateB, StateC))
Group

Sales = c(t(as.matrix(Group))); Sales

f = c("State A", "State B", "State C")
f
g = c("Salesman 1", "Salesman 2", "Salesman 3", "Salesman 4")
g
# Number of columns
k = ncol(Group)
k
# Number of rows
n = nrow(Group)
n

#Generate factor levels of states
States = gl(k, 1, n*k, factor(f))

```


States

```
# Generate factor levels of Salesmen  
Salesmen = gl(n, k, n*k, factor(g))  
Salesmen
```

```
#ANOVA Table
```

```
anova = aov(Sales~States + Salesmen)
```

```
summary(anova)
```

```
# Reject the states
```

```
# Reject the salesmen
```
