

Solution Paper 3

2022-04-22

##Q1

#A

mean = 4

variance = 8

##(a)

lambda = mean / variance

alpha = mean*lambda

set.seed(12)

sim.y = rgamma(100,alpha,lambda)

mean(sim.y)

[1] 4.033375

median(sim.y)

[1] 3.485327

f = density(sim.y)

mode.y = f\$x[which.max(f\$y)]

mode.y

[1] 2.252472

##(b)

pop.mean = alpha/lambda

pop.median = qgamma(0.5,alpha,lambda)

pop.mode = (alpha - 1)/lambda

pop.mean;mean(sim.y)

[1] 4

[1] 4.033375

pop.median;median(sim.y)

[1] 3.356694

[1] 3.485327

pop.mode;mode.y

[1] 2

[1] 2.252472

##All the population moments are quite near to the sample / empirical moments.

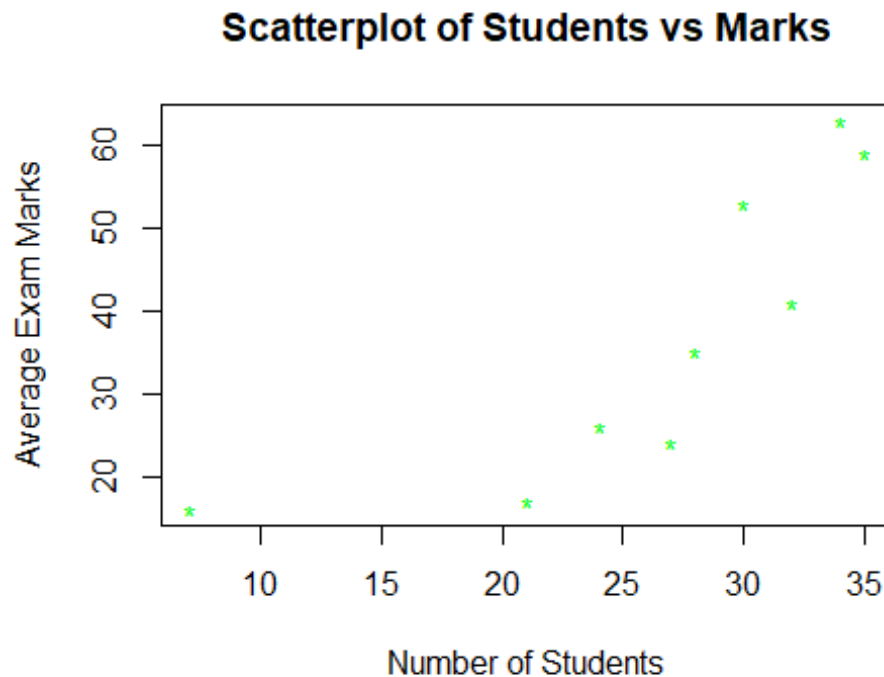
##Q1B

```
n = c(35,32,27,21,34,30,28,24,7)
```

```
m = c(59,41,24,17,63,53,35,26,16)
```

##(a)

```
plot(n,m,  
     col = "green",  
     pch = "*",  
     main = "Scatterplot of Students vs Marks",  
     xlab = "Number of Students",  
     ylab = "Average Exam Marks")
```



#The two variables seem to have a positive correlation.

#The relationship seems to be linear in nature expect when number of students is very low.

##(b)

```
cor.test(n,m, alternative = "greater")
```

```
##
```

```
## Pearson's product-moment correlation
```

```
##
```

```
## data: n and m
```

```
## t = 3.6603, df = 7, p-value = 0.004033
```

```
## alternative hypothesis: true correlation is greater than 0
```

```
## 95 percent confidence interval:
```

```
## 0.4274953 1.0000000
```

```
## sample estimates:
```

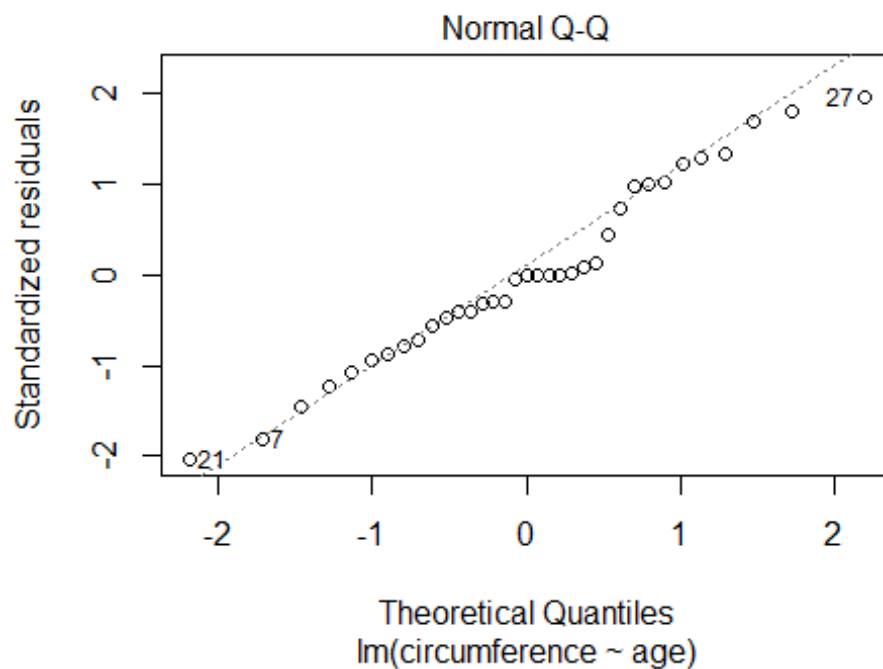
```
##      cor
## 0.8104492
```

#Since the above 95% Confidence Interval contains 0.8 , we can say we have sufficient evidence to accept the H_0 , i.e. $\rho = 0.8$

```
##Q1C
##(i)
lm1 = lm(circumference~age, data = Orange)
##(ii)
summary(lm1)$sigma

## [1] 23.73767

##(iii)
plot(lm1,2)
```



*#Overall, the points seems to lie on the Normal Line
#At Around the median, there is a difference between the theoretical and sample quantiles.
#Overall, we would suggest that errors follow a Normal Distribution by looking at this graph*

```
##Q2A
set.seed(1729)
n = 100
wait.times = -log(runif(n))/0.1
```

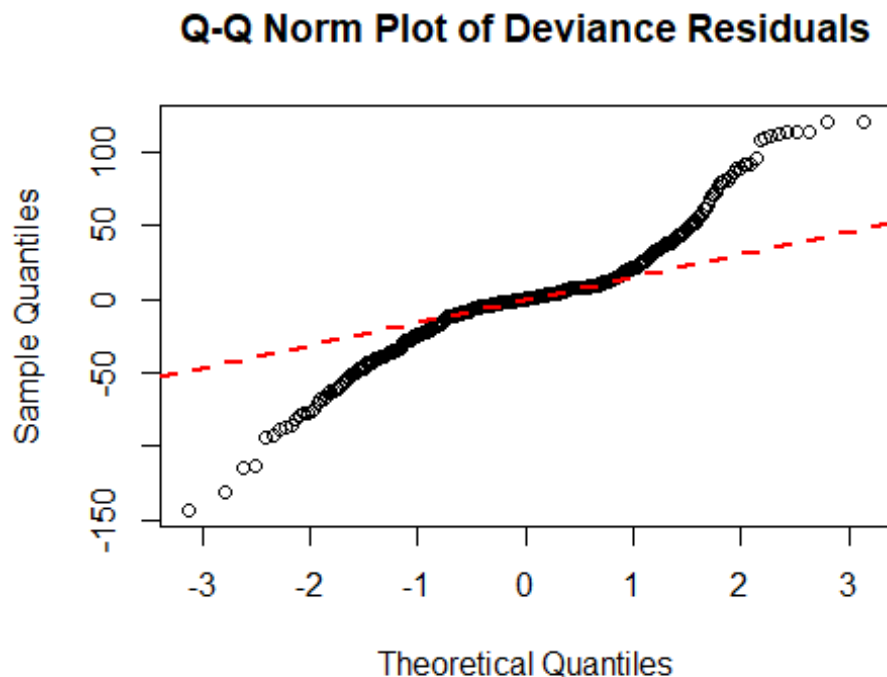
```
##(i)
set.seed(2022)
boot.median = replicate(1000,
                        median(rexp(n,1/mean(wait.times))))

##(ii)
alpha = 0.03
quantile(boot.median,c(alpha/2,1-alpha/2))

##      1.5%      98.5%
## 5.258064 9.675926

##Q2B
glm = glm(weight~Time + I(Time^2) + Diet,
          data = ChickWeight,
          family = gaussian(link = "log"))

##(a)
##If the student does plot(glm,2) , please award 0.5 marks in this subpart
deviance.resid = resid(glm,type = "deviance")
qqnorm(deviance.resid,
       main = "Q-Q Norm Plot of Deviance Residuals")
qqline(deviance.resid,col = "red", lty = 2, lwd = 2)
```



#The Points in the tails (both ends) are quite far off the red line. We can say while deviance residuals are symmetric , they are not normal.

```
##(b)
new.data = data.frame(Time = 12, Diet = "3")
predict(glm,new.data, type = "response")
```

```
##          1
## 150.0973

##Q2C
Males <- c(21,22,28,27,20,23,26,32,25,21,30)
Females <- c(19,18,38,33,24,39,22,29,28,26,30)
##(a)
test = var.test(Males,Females)
test$alternative

## [1] "two.sided"

test$p.value

## [1] 0.0898616

##(b)
#Since the p - value of the test (0.089) which is greater than 5% , we can conclude that the variances are equal.
#Since the two sample t - test requires that the population variances are equal , which is true here.

##Q3A
sachin = c(12,35,65,125,118,48,28,48)
lara = c(75,08,28,107,137,45,36,60)
##(a)
t.test(sachin,lara,
       paired = TRUE)

##
## Paired t-test
##
## data:  sachin and lara
## t = -0.19098, df = 7, p-value = 0.854
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -28.43524  24.18524
## sample estimates:
## mean of the differences
## -2.125

#Since p - values are greater than 5% , we can conclude that the means are equal.
##(b)
var.test(sachin,lara,conf.level = 0.99)$conf.int

## [1] 0.103748 8.190916
## attr(,"conf.level")
## [1] 0.99

##(c)
D = sachin - lara
```

```

ObsT = median(D)
##(d)
sign = c(-1,1)
##(e)
library("gtools")
p = permutations(2,8,c(-1,1),repeats.allowed = T)
p

```

```

##      [,1] [,2] [,3] [,4] [,5] [,6] [,7] [,8]
## [1,] -1 -1 -1 -1 -1 -1 -1 -1
## [2,] -1 -1 -1 -1 -1 -1 -1 1
## [3,] -1 -1 -1 -1 -1 -1 1 -1
## [4,] -1 -1 -1 -1 -1 -1 1 1
## [5,] -1 -1 -1 -1 -1 1 -1 -1
## [6,] -1 -1 -1 -1 -1 1 -1 1
## [7,] -1 -1 -1 -1 -1 1 1 -1
## [8,] -1 -1 -1 -1 -1 1 1 1
## [9,] -1 -1 -1 -1 1 -1 -1 -1
## [10,] -1 -1 -1 -1 1 -1 -1 1
## [11,] -1 -1 -1 -1 1 -1 1 -1
## [12,] -1 -1 -1 -1 1 -1 1 1
## [13,] -1 -1 -1 -1 1 1 -1 -1
## [14,] -1 -1 -1 -1 1 1 -1 1
## [15,] -1 -1 -1 -1 1 1 1 -1
## [16,] -1 -1 -1 -1 1 1 1 1
## [17,] -1 -1 -1 1 -1 -1 -1 -1
## [18,] -1 -1 -1 1 -1 -1 -1 1
## [19,] -1 -1 -1 1 -1 -1 1 -1
## [20,] -1 -1 -1 1 -1 -1 1 1
## [21,] -1 -1 -1 1 -1 1 -1 -1
## [22,] -1 -1 -1 1 -1 1 -1 1
## [23,] -1 -1 -1 1 -1 1 1 -1
## [24,] -1 -1 -1 1 -1 1 1 1
## [25,] -1 -1 -1 1 1 -1 -1 -1
## [26,] -1 -1 -1 1 1 -1 -1 1
## [27,] -1 -1 -1 1 1 -1 1 -1
## [28,] -1 -1 -1 1 1 -1 1 1
## [29,] -1 -1 -1 1 1 1 -1 -1
## [30,] -1 -1 -1 1 1 1 -1 1
## [31,] -1 -1 -1 1 1 1 1 -1
## [32,] -1 -1 -1 1 1 1 1 1
## [33,] -1 -1 1 -1 -1 -1 -1 -1
## [34,] -1 -1 1 -1 -1 -1 -1 1
## [35,] -1 -1 1 -1 -1 -1 1 -1
## [36,] -1 -1 1 -1 -1 -1 1 1
## [37,] -1 -1 1 -1 -1 1 -1 -1
## [38,] -1 -1 1 -1 -1 1 -1 1
## [39,] -1 -1 1 -1 -1 1 1 -1
## [40,] -1 -1 1 -1 -1 1 1 1
## [41,] -1 -1 1 -1 1 -1 -1 -1

```

##	[42,]	-1	-1	1	-1	1	-1	-1	1
##	[43,]	-1	-1	1	-1	1	-1	1	-1
##	[44,]	-1	-1	1	-1	1	-1	1	1
##	[45,]	-1	-1	1	-1	1	1	-1	-1
##	[46,]	-1	-1	1	-1	1	1	-1	1
##	[47,]	-1	-1	1	-1	1	1	1	-1
##	[48,]	-1	-1	1	-1	1	1	1	1
##	[49,]	-1	-1	1	1	-1	-1	-1	-1
##	[50,]	-1	-1	1	1	-1	-1	-1	1
##	[51,]	-1	-1	1	1	-1	-1	1	-1
##	[52,]	-1	-1	1	1	-1	-1	1	1
##	[53,]	-1	-1	1	1	-1	1	-1	-1
##	[54,]	-1	-1	1	1	-1	1	-1	1
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##	[56,]	-1	-1	1	1	-1	1	1	1
##	[57,]	-1	-1	1	1	1	-1	-1	-1
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##	[70,]	-1	1	-1	-1	-1	1	-1	1
##	[71,]	-1	1	-1	-1	-1	1	1	-1
##	[72,]	-1	1	-1	-1	-1	1	1	1
##	[73,]	-1	1	-1	-1	1	-1	-1	-1
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##	[79,]	-1	1	-1	-1	1	1	1	-1
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##	[81,]	-1	1	-1	1	-1	-1	-1	-1
##	[82,]	-1	1	-1	1	-1	-1	-1	1
##	[83,]	-1	1	-1	1	-1	-1	1	-1
##	[84,]	-1	1	-1	1	-1	-1	1	1
##	[85,]	-1	1	-1	1	-1	1	-1	-1
##	[86,]	-1	1	-1	1	-1	1	-1	1
##	[87,]	-1	1	-1	1	-1	1	1	-1
##	[88,]	-1	1	-1	1	-1	1	1	1
##	[89,]	-1	1	-1	1	1	-1	-1	-1
##	[90,]	-1	1	-1	1	1	-1	-1	1
##	[91,]	-1	1	-1	1	1	-1	1	-1

##	[92,]	-1	1	-1	1	1	-1	1	1
##	[93,]	-1	1	-1	1	1	1	-1	-1
##	[94,]	-1	1	-1	1	1	1	-1	1
##	[95,]	-1	1	-1	1	1	1	1	-1
##	[96,]	-1	1	-1	1	1	1	1	1
##	[97,]	-1	1	1	-1	-1	-1	-1	-1
##	[98,]	-1	1	1	-1	-1	-1	-1	1
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##	[102,]	-1	1	1	-1	-1	1	-1	1
##	[103,]	-1	1	1	-1	-1	1	1	-1
##	[104,]	-1	1	1	-1	-1	1	1	1
##	[105,]	-1	1	1	-1	1	-1	-1	-1
##	[106,]	-1	1	1	-1	1	-1	-1	1
##	[107,]	-1	1	1	-1	1	-1	1	-1
##	[108,]	-1	1	1	-1	1	-1	1	1
##	[109,]	-1	1	1	-1	1	1	-1	-1
##	[110,]	-1	1	1	-1	1	1	-1	1
##	[111,]	-1	1	1	-1	1	1	1	-1
##	[112,]	-1	1	1	-1	1	1	1	1
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##	[114,]	-1	1	1	1	-1	-1	-1	1
##	[115,]	-1	1	1	1	-1	-1	1	-1
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##	[122,]	-1	1	1	1	1	-1	-1	1
##	[123,]	-1	1	1	1	1	-1	1	-1
##	[124,]	-1	1	1	1	1	-1	1	1
##	[125,]	-1	1	1	1	1	1	-1	-1
##	[126,]	-1	1	1	1	1	1	-1	1
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## [142,]	1	-1	-1	-1	1	1	-1	1
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## [144,]	1	-1	-1	-1	1	1	1	1
## [145,]	1	-1	-1	1	-1	-1	-1	-1
## [146,]	1	-1	-1	1	-1	-1	-1	1
## [147,]	1	-1	-1	1	-1	-1	1	-1
## [148,]	1	-1	-1	1	-1	-1	1	1
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## [151,]	1	-1	-1	1	-1	1	1	-1
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## [158,]	1	-1	-1	1	1	1	-1	1
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## [162,]	1	-1	1	-1	-1	-1	-1	1
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## [164,]	1	-1	1	-1	-1	-1	1	1
## [165,]	1	-1	1	-1	-1	1	-1	-1
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## [167,]	1	-1	1	-1	-1	1	1	-1
## [168,]	1	-1	1	-1	-1	1	1	1
## [169,]	1	-1	1	-1	1	-1	-1	-1
## [170,]	1	-1	1	-1	1	-1	-1	1
## [171,]	1	-1	1	-1	1	-1	1	-1
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## [173,]	1	-1	1	-1	1	1	-1	-1
## [174,]	1	-1	1	-1	1	1	-1	1
## [175,]	1	-1	1	-1	1	1	1	-1
## [176,]	1	-1	1	-1	1	1	1	1
## [177,]	1	-1	1	1	-1	-1	-1	-1
## [178,]	1	-1	1	1	-1	-1	-1	1
## [179,]	1	-1	1	1	-1	-1	1	-1
## [180,]	1	-1	1	1	-1	-1	1	1
## [181,]	1	-1	1	1	-1	1	-1	-1
## [182,]	1	-1	1	1	-1	1	-1	1
## [183,]	1	-1	1	1	-1	1	1	-1
## [184,]	1	-1	1	1	-1	1	1	1
## [185,]	1	-1	1	1	1	-1	-1	-1
## [186,]	1	-1	1	1	1	-1	-1	1
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## [189,]	1	-1	1	1	1	1	-1	-1
## [190,]	1	-1	1	1	1	1	-1	1
## [191,]	1	-1	1	1	1	1	1	-1

## [192,]	1	-1	1	1	1	1	1	1
## [193,]	1	1	-1	-1	-1	-1	-1	-1
## [194,]	1	1	-1	-1	-1	-1	-1	1
## [195,]	1	1	-1	-1	-1	-1	1	-1
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## [197,]	1	1	-1	-1	-1	1	-1	-1
## [198,]	1	1	-1	-1	-1	1	-1	1
## [199,]	1	1	-1	-1	-1	1	1	-1
## [200,]	1	1	-1	-1	-1	1	1	1
## [201,]	1	1	-1	-1	1	-1	-1	-1
## [202,]	1	1	-1	-1	1	-1	-1	1
## [203,]	1	1	-1	-1	1	-1	1	-1
## [204,]	1	1	-1	-1	1	-1	1	1
## [205,]	1	1	-1	-1	1	1	-1	-1
## [206,]	1	1	-1	-1	1	1	-1	1
## [207,]	1	1	-1	-1	1	1	1	-1
## [208,]	1	1	-1	-1	1	1	1	1
## [209,]	1	1	-1	1	-1	-1	-1	-1
## [210,]	1	1	-1	1	-1	-1	-1	1
## [211,]	1	1	-1	1	-1	-1	1	-1
## [212,]	1	1	-1	1	-1	-1	1	1
## [213,]	1	1	-1	1	-1	1	-1	-1
## [214,]	1	1	-1	1	-1	1	-1	1
## [215,]	1	1	-1	1	-1	1	1	-1
## [216,]	1	1	-1	1	-1	1	1	1
## [217,]	1	1	-1	1	1	-1	-1	-1
## [218,]	1	1	-1	1	1	-1	-1	1
## [219,]	1	1	-1	1	1	-1	1	-1
## [220,]	1	1	-1	1	1	-1	1	1
## [221,]	1	1	-1	1	1	1	-1	-1
## [222,]	1	1	-1	1	1	1	-1	1
## [223,]	1	1	-1	1	1	1	1	-1
## [224,]	1	1	-1	1	1	1	1	1
## [225,]	1	1	1	-1	-1	-1	-1	-1
## [226,]	1	1	1	-1	-1	-1	-1	1
## [227,]	1	1	1	-1	-1	-1	1	-1
## [228,]	1	1	1	-1	-1	-1	1	1
## [229,]	1	1	1	-1	-1	1	-1	-1
## [230,]	1	1	1	-1	-1	1	-1	1
## [231,]	1	1	1	-1	-1	1	1	-1
## [232,]	1	1	1	-1	-1	1	1	1
## [233,]	1	1	1	-1	1	-1	-1	-1
## [234,]	1	1	1	-1	1	-1	-1	1
## [235,]	1	1	1	-1	1	-1	1	-1
## [236,]	1	1	1	-1	1	-1	1	1
## [237,]	1	1	1	-1	1	1	-1	-1
## [238,]	1	1	1	-1	1	1	-1	1
## [239,]	1	1	1	-1	1	1	1	-1
## [240,]	1	1	1	-1	1	1	1	1
## [241,]	1	1	1	1	-1	-1	-1	-1

```
## [242,] 1 1 1 1 -1 -1 -1 1
## [243,] 1 1 1 1 -1 -1 1 -1
## [244,] 1 1 1 1 -1 -1 1 1
## [245,] 1 1 1 1 -1 1 -1 -1
## [246,] 1 1 1 1 -1 1 -1 1
## [247,] 1 1 1 1 -1 1 1 -1
## [248,] 1 1 1 1 -1 1 1 1
## [249,] 1 1 1 1 1 -1 -1 -1
## [250,] 1 1 1 1 1 -1 -1 1
## [251,] 1 1 1 1 1 -1 1 -1
## [252,] 1 1 1 1 1 -1 1 1
## [253,] 1 1 1 1 1 1 -1 -1
## [254,] 1 1 1 1 1 1 -1 1
## [255,] 1 1 1 1 1 1 1 -1
## [256,] 1 1 1 1 1 1 1 1
```

```
##(f)
```

```
n = ncol(p)
```

```
dif = rep(0,n)
```

```
for(i in 1:n){
```

```
  dif[i] = median(D*p[i,])
```

```
}
```

```
##(g)
```

```
hist(dif,
```

```
  main = "Histogram of Medians",
```

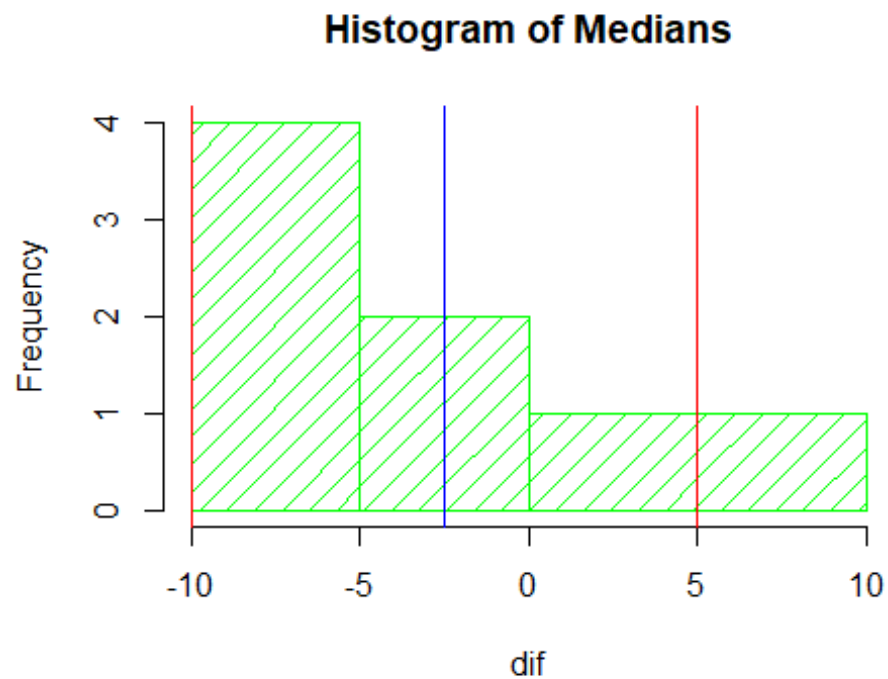
```
  col = "green",
```

```
  density = 10)
```

```
abline(v = quantile(dif,0.025), col = "red")
```

```
abline(v = quantile(dif,0.975), col = "red")
```

```
abline(v = ObsT,col = "blue")
```



#Since the observed value is in between the Critical values, we can conclude that the medians are equal.

##Q3B

#(a)

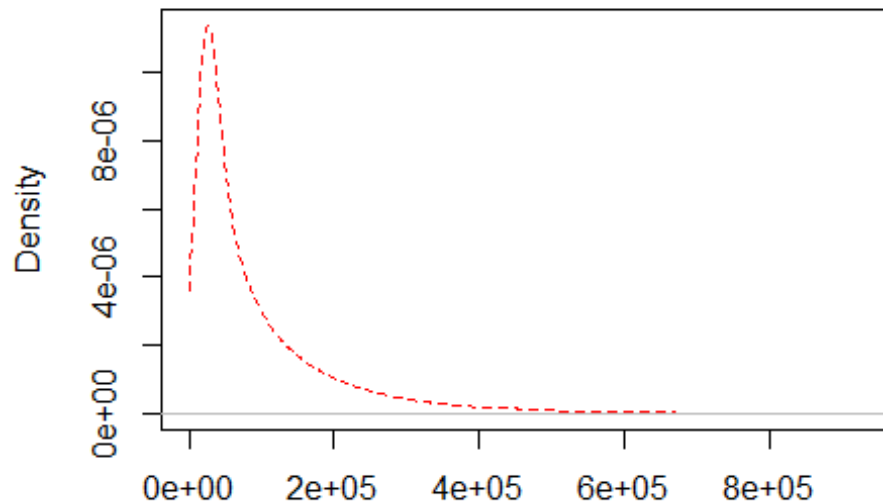
```
claims_exp = read.csv("Claims_Experience.csv")  
nrow(claims_exp)
```

```
## [1] 1436
```

#(b)

```
plot(density(claims_exp$CLAIM, from = 0), col = "red", lty = 2)
```

```
density.default(x = claims_exp$CLAIM, from = 0)
```



N = 1436 Bandwidth = 1.462e+04

#It is clear that CLAIM is +vely skewed. We can use the Log Normal Distribution or Gamma Distribution.

#(c)

```
model1 = glm(CLAIM~LOCATION*JOB*SEX*AGE,  
             data = claims_exp,  
             family = Gamma(link = "inverse"))  
model1$coefficients
```

```
##              (Intercept)  
##              6.648743e-05  
## LOCATIONREGION 02  
##              6.082008e-06  
## LOCATIONREGION 03  
##              -7.057874e-06  
## LOCATIONREGION 04  
##              2.053610e-06  
## LOCATIONREGION 05  
##              -1.279087e-06  
## LOCATIONREGION 06  
##              -1.079271e-05  
## LOCATIONREGION 07  
##              -6.219876e-06  
## LOCATIONREGION 08  
##              3.112934e-05  
## LOCATIONREGION 09  
##              -4.394827e-06
```

```

##          LOCATIONREGION 10
##          2.248405e-05
##          LOCATIONREGION 11
##          5.214718e-06
##          JOBBusiness
##          -8.473181e-05
##          JOBSelfemployed
##          -1.506208e-05
##          JOBService
##          -5.487268e-05
##          SEXM
##          7.394354e-06
##          AGE
##          -9.615268e-07
##          LOCATIONREGION 02:JOBBusiness
##          2.190482e-05
##          LOCATIONREGION 03:JOBBusiness
##          1.487980e-04
##          LOCATIONREGION 04:JOBBusiness
##          -1.784083e-05
##          LOCATIONREGION 05:JOBBusiness
##          3.934269e-05
##          LOCATIONREGION 06:JOBBusiness
##          4.447039e-06
##          LOCATIONREGION 07:JOBBusiness
##          NA
##          LOCATIONREGION 08:JOBBusiness
##          1.253765e-05
##          LOCATIONREGION 09:JOBBusiness
##          -3.945155e-06
##          LOCATIONREGION 10:JOBBusiness
##          -1.803709e-06
##          LOCATIONREGION 11:JOBBusiness
##          NA
##          LOCATIONREGION 02:JOBSelfemployed
##          -9.160112e-06
##          LOCATIONREGION 03:JOBSelfemployed
##          -8.981197e-06
##          LOCATIONREGION 04:JOBSelfemployed
##          -2.233671e-05
##          LOCATIONREGION 05:JOBSelfemployed
##          -1.348598e-05
##          LOCATIONREGION 06:JOBSelfemployed
##          -8.590919e-06
##          LOCATIONREGION 07:JOBSelfemployed
##          -1.591509e-05
##          LOCATIONREGION 08:JOBSelfemployed
##          -2.379111e-05
##          LOCATIONREGION 09:JOBSelfemployed
##          6.703445e-07

```

```
## LOCATIONREGION 10:JOBSelfemployed
## -4.611627e-05
## LOCATIONREGION 11:JOBSelfemployed
## -2.430463e-05
## LOCATIONREGION 02:JOBService
## -2.107660e-05
## LOCATIONREGION 03:JOBService
## -1.997418e-06
## LOCATIONREGION 04:JOBService
## -2.824428e-05
## LOCATIONREGION 05:JOBService
## -4.380992e-05
## LOCATIONREGION 06:JOBService
## -1.495554e-05
## LOCATIONREGION 07:JOBService
## NA
## LOCATIONREGION 08:JOBService
## -1.839124e-05
## LOCATIONREGION 09:JOBService
## 5.542681e-06
## LOCATIONREGION 10:JOBService
## -6.870395e-06
## LOCATIONREGION 11:JOBService
## -2.540112e-06
## LOCATIONREGION 02:SEXM
## 8.127953e-06
## LOCATIONREGION 03:SEXM
## 8.904409e-07
## LOCATIONREGION 04:SEXM
## 1.005682e-05
## LOCATIONREGION 05:SEXM
## 8.190134e-06
## LOCATIONREGION 06:SEXM
## 9.653825e-06
## LOCATIONREGION 07:SEXM
## -1.411381e-05
## LOCATIONREGION 08:SEXM
## 8.774610e-06
## LOCATIONREGION 09:SEXM
## -1.552181e-06
## LOCATIONREGION 10:SEXM
## -2.426235e-06
## LOCATIONREGION 11:SEXM
## -2.214530e-05
## JOBBusiness:SEXM
## 8.130285e-06
## JOBSelfemployed:SEXM
## -6.226583e-06
## JOBService:SEXM
## -1.131110e-05
```

```
##          LOCATIONREGION 02:AGE
##          -1.433937e-07
##          LOCATIONREGION 03:AGE
##          1.529197e-07
##          LOCATIONREGION 04:AGE
##          -7.211924e-08
##          LOCATIONREGION 05:AGE
##          2.842271e-08
##          LOCATIONREGION 06:AGE
##          1.934835e-07
##          LOCATIONREGION 07:AGE
##          1.355919e-07
##          LOCATIONREGION 08:AGE
##          -5.660481e-07
##          LOCATIONREGION 09:AGE
##          1.242373e-07
##          LOCATIONREGION 10:AGE
##          -5.238387e-07
##          LOCATIONREGION 11:AGE
##          -3.955090e-08
##          JOBBusiness:AGE
##          1.296333e-06
##          JOBSelfemployed:AGE
##          2.072594e-07
##          JOBService:AGE
##          8.116343e-07
##          SEXM:AGE
##          -1.212850e-07
##          LOCATIONREGION 02:JOBBusiness:SEXM
##          -2.772917e-05
##          LOCATIONREGION 03:JOBBusiness:SEXM
##          -1.377355e-04
##          LOCATIONREGION 04:JOBBusiness:SEXM
##          1.170428e-06
##          LOCATIONREGION 05:JOBBusiness:SEXM
##          -3.935562e-05
##          LOCATIONREGION 06:JOBBusiness:SEXM
##          NA
##          LOCATIONREGION 07:JOBBusiness:SEXM
##          NA
##          LOCATIONREGION 08:JOBBusiness:SEXM
##          -8.176112e-06
##          LOCATIONREGION 09:JOBBusiness:SEXM
##          4.081533e-06
##          LOCATIONREGION 10:JOBBusiness:SEXM
##          -4.146076e-06
##          LOCATIONREGION 11:JOBBusiness:SEXM
##          NA
##          LOCATIONREGION 02:JOBSelfemployed:SEXM
##          -6.313137e-06
```

```
## LOCATIONREGION 03:JOBSelfemployed:SEXM
## 7.129838e-07
## LOCATIONREGION 04:JOBSelfemployed:SEXM
## 3.451671e-06
## LOCATIONREGION 05:JOBSelfemployed:SEXM
## -1.771808e-06
## LOCATIONREGION 06:JOBSelfemployed:SEXM
## 5.364470e-06
## LOCATIONREGION 07:JOBSelfemployed:SEXM
## 1.815363e-05
## LOCATIONREGION 08:JOBSelfemployed:SEXM
## 2.801365e-06
## LOCATIONREGION 09:JOBSelfemployed:SEXM
## -5.021325e-06
## LOCATIONREGION 10:JOBSelfemployed:SEXM
## -9.044808e-06
## LOCATIONREGION 11:JOBSelfemployed:SEXM
## 7.793583e-06
## LOCATIONREGION 02:JOBService:SEXM
## 1.330813e-05
## LOCATIONREGION 03:JOBService:SEXM
## -1.110880e-05
## LOCATIONREGION 04:JOBService:SEXM
## 1.885062e-08
## LOCATIONREGION 05:JOBService:SEXM
## 1.291790e-06
## LOCATIONREGION 06:JOBService:SEXM
## 2.782997e-05
## LOCATIONREGION 07:JOBService:SEXM
## NA
## LOCATIONREGION 08:JOBService:SEXM
## 3.783515e-06
## LOCATIONREGION 09:JOBService:SEXM
## -1.219085e-06
## LOCATIONREGION 10:JOBService:SEXM
## -6.835778e-06
## LOCATIONREGION 11:JOBService:SEXM
## NA
## LOCATIONREGION 02:JOBBusiness:AGE
## -2.951069e-07
## LOCATIONREGION 03:JOBBusiness:AGE
## -2.479130e-06
## LOCATIONREGION 04:JOBBusiness:AGE
## 3.407004e-07
## LOCATIONREGION 05:JOBBusiness:AGE
## -6.295374e-07
## LOCATIONREGION 06:JOBBusiness:AGE
## NA
## LOCATIONREGION 07:JOBBusiness:AGE
## NA
```

```
##      LOCATIONREGION 08:JOBBusiness:AGE
##      -8.000520e-08
##      LOCATIONREGION 09:JOBBusiness:AGE
##      -1.620948e-08
##      LOCATIONREGION 10:JOBBusiness:AGE
##      1.733688e-07
##      LOCATIONREGION 11:JOBBusiness:AGE
##      NA
##      LOCATIONREGION 02:JOBSelfemployed:AGE
##      2.371257e-07
##      LOCATIONREGION 03:JOBSelfemployed:AGE
##      1.442140e-07
##      LOCATIONREGION 04:JOBSelfemployed:AGE
##      4.340938e-07
##      LOCATIONREGION 05:JOBSelfemployed:AGE
##      2.449317e-07
##      LOCATIONREGION 06:JOBSelfemployed:AGE
##      1.380875e-07
##      LOCATIONREGION 07:JOBSelfemployed:AGE
##      2.510680e-07
##      LOCATIONREGION 08:JOBSelfemployed:AGE
##      5.139940e-07
##      LOCATIONREGION 09:JOBSelfemployed:AGE
##      -8.582010e-09
##      LOCATIONREGION 10:JOBSelfemployed:AGE
##      9.456776e-07
##      LOCATIONREGION 11:JOBSelfemployed:AGE
##      3.689135e-07
##      LOCATIONREGION 02:JOBService:AGE
##      4.125263e-07
##      LOCATIONREGION 03:JOBService:AGE
##      1.496175e-08
##      LOCATIONREGION 04:JOBService:AGE
##      4.757753e-07
##      LOCATIONREGION 05:JOBService:AGE
##      7.007401e-07
##      LOCATIONREGION 06:JOBService:AGE
##      2.244318e-07
##      LOCATIONREGION 07:JOBService:AGE
##      NA
##      LOCATIONREGION 08:JOBService:AGE
##      4.105430e-07
##      LOCATIONREGION 09:JOBService:AGE
##      -1.217710e-07
##      LOCATIONREGION 10:JOBService:AGE
##      2.453950e-07
##      LOCATIONREGION 11:JOBService:AGE
##      NA
##      LOCATIONREGION 02:SEX:AGE
##      -1.576597e-07
```

```
##          LOCATIONREGION 03:SEX:AGE
##          -3.321293e-08
##          LOCATIONREGION 04:SEX:AGE
##          -1.898819e-07
##          LOCATIONREGION 05:SEX:AGE
##          -1.677682e-07
##          LOCATIONREGION 06:SEX:AGE
##          -2.140289e-07
##          LOCATIONREGION 07:SEX:AGE
##          2.233571e-07
##          LOCATIONREGION 08:SEX:AGE
##          -2.325176e-07
##          LOCATIONREGION 09:SEX:AGE
##          -7.995125e-09
##          LOCATIONREGION 10:SEX:AGE
##          1.242166e-07
##          LOCATIONREGION 11:SEX:AGE
##          3.275751e-07
##          JOBBusiness:SEX:AGE
##          -1.188027e-07
##          JOBSelfemployed:SEX:AGE
##          1.675576e-07
##          JOBService:SEX:AGE
##          2.316297e-07
##          LOCATIONREGION 02:JOBBusiness:SEX:AGE
##          4.722078e-07
##          LOCATIONREGION 03:JOBBusiness:SEX:AGE
##          2.287015e-06
##          LOCATIONREGION 04:JOBBusiness:SEX:AGE
##          NA
##          LOCATIONREGION 05:JOBBusiness:SEX:AGE
##          6.530614e-07
##          LOCATIONREGION 06:JOBBusiness:SEX:AGE
##          NA
##          LOCATIONREGION 07:JOBBusiness:SEX:AGE
##          NA
##          LOCATIONREGION 08:JOBBusiness:SEX:AGE
##          2.344789e-07
##          LOCATIONREGION 09:JOBBusiness:SEX:AGE
##          NA
##          LOCATIONREGION 10:JOBBusiness:SEX:AGE
##          NA
##          LOCATIONREGION 11:JOBBusiness:SEX:AGE
##          NA
##          LOCATIONREGION 02:JOBSelfemployed:SEX:AGE
##          6.635080e-08
##          LOCATIONREGION 03:JOBSelfemployed:SEX:AGE
##          -4.193875e-08
##          LOCATIONREGION 04:JOBSelfemployed:SEX:AGE
##          -8.180159e-08
```

```

## LOCATIONREGION 05:JOBSelfemployed:SEX:AGE
## 1.604930e-08
## LOCATIONREGION 06:JOBSelfemployed:SEX:AGE
## -5.609303e-08
## LOCATIONREGION 07:JOBSelfemployed:SEX:AGE
## -3.316507e-07
## LOCATIONREGION 08:JOBSelfemployed:SEX:AGE
## -4.694789e-08
## LOCATIONREGION 09:JOBSelfemployed:SEX:AGE
## 4.151637e-08
## LOCATIONREGION 10:JOBSelfemployed:SEX:AGE
## NA
## LOCATIONREGION 11:JOBSelfemployed:SEX:AGE
## NA
## LOCATIONREGION 02:JOBService:SEX:AGE
## -2.328322e-07
## LOCATIONREGION 03:JOBService:SEX:AGE
## 1.465216e-07
## LOCATIONREGION 04:JOBService:SEX:AGE
## NA
## LOCATIONREGION 05:JOBService:SEX:AGE
## NA
## LOCATIONREGION 06:JOBService:SEX:AGE
## -4.369955e-07
## LOCATIONREGION 07:JOBService:SEX:AGE
## NA
## LOCATIONREGION 08:JOBService:SEX:AGE
## NA
## LOCATIONREGION 09:JOBService:SEX:AGE
## NA
## LOCATIONREGION 10:JOBService:SEX:AGE
## NA
## LOCATIONREGION 11:JOBService:SEX:AGE
## NA

```

#Certain co - effieicents are 0 as there are no claims for this combination and as a result of that , 0 would be predicted and so the co - efficient would be 1/0. Which means the GLM couldnt converge to a single number.

#(d)

```
confint(model1,parm = "SEX")
```

```
## Waiting for profiling to be done...
```

```
##      2.5 %      97.5 %
## -1.799749e-05  3.190735e-05
```

#Since the 95% confidence interval contains the number 0, we can conclude that SEX is not a statistically significant variable.

#(e)

```
intercept.ci = confint(model1,parm = "(Intercept)")
```

```

## Waiting for profiling to be done...

intercept.ci[1] < 0.001 & intercept.ci[2] > 0.001

## 2.5 %
## FALSE

#Since the value 0.001 is not in between the CI , we can conclude that the
Intercept coefficient is not 0.001
#(f)
model0 = glm(CLAIM~1,
             data = claims_exp,
             family = Gamma("inverse"))
#One - Variable Models
model1a = update(model0, .~.+LOCATION)
model1b = update(model0, .~.+JOB)
model1c = update(model0, .~.+SEX)
model1d = update(model0, .~.+AGE)
AIC(model1a)

## [1] 35345.32

AIC(model1b)

## [1] 34079.99

AIC(model1c)

## [1] 35695.06

AIC(model1d)

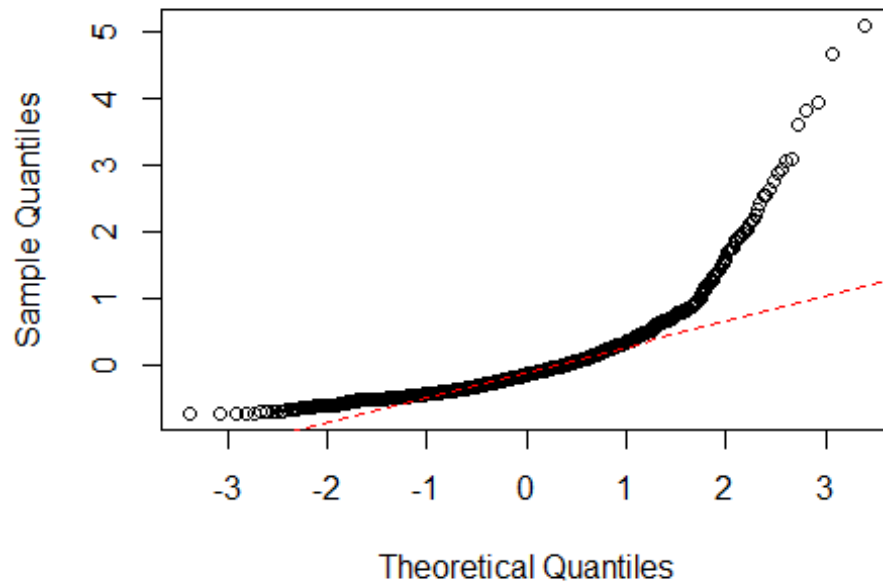
## [1] 33410.99

#Model1D is selected
#Similarly , the candidate will go forward with this method.

#(g)
qqnorm(resid(model1d,"pearson"))
qqline(resid(model1d,"pearson"),col = "red", lty = 2)

```

Normal Q-Q Plot



```
##(h)
sqrt(mean((claims_exp$CLAIM - fitted(model1d))^2))

## [1] 83460.68

##(i)
new_data1 = data.frame(AGE = 34,
                        SEX = "M",
                        LOCATION = "REGION 01",
                        JOB = "Business")
new_Data2 = data.frame(JOB = "Selfemployed",
                        AGE = 26,
                        SEX = "F",
                        LOCATION = "REGION 08")
predict(model1d,new_data1,type = "response")/predict(model1d,new_Data2,type =
"response")

##      1
## 1.235745
```