

## Assignment 9 (S-520)

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1.

- (a) The experimental unit is a person. i.e. aerobic students.
- (b) The experimental units belong to one population, i.e., aerobic students. (1- sample location)
- (c) Two measurements were taken on each experimental unit:
  - i. Number of watts expended during protocol S (30-minute ride on the first week)
  - ii. Number of watts expended during protocol D (30-minute ride on the second week)

(d) Let  $S_i$  be the score on protocol S for student  $i$ , and let  $D_i$  denote score on protocol D for student  $i$ .  $\mu$

Then,  $X_i = D_i - S_i$  is the random variable of interest. We are interested on drawing inferences about  $\mu$ .

(e)  $\mu > 0$  iff  $D_i > S_i$ . Thus, to test the theory in favor of dynamic stretches we might want to test  $H_0 : \mu \leq 0$

vs.  $H_1 : \mu > 0$ .

2.C-1: 2-sample location problem

- (a) The experimental unit is a middle-aged man.
- (b) The experimental units belong to one of two populations:
  - i. Type A heavy men.
  - ii. Type B heavy men.
- (c) One measurement (cholesterol level) were taken on each experimental unit.

(d) Let  $X_i$  denote the cholesterol level for man  $i$  (Type A).

Let  $Y_j$  denote the cholesterol level for man  $j$  (Type B).

Then,  $X_1; X_2; \dots; X_{n1} \sim P_1$ ;  $Y_1; Y_2; \dots; Y_{n2} \sim P_2$ .

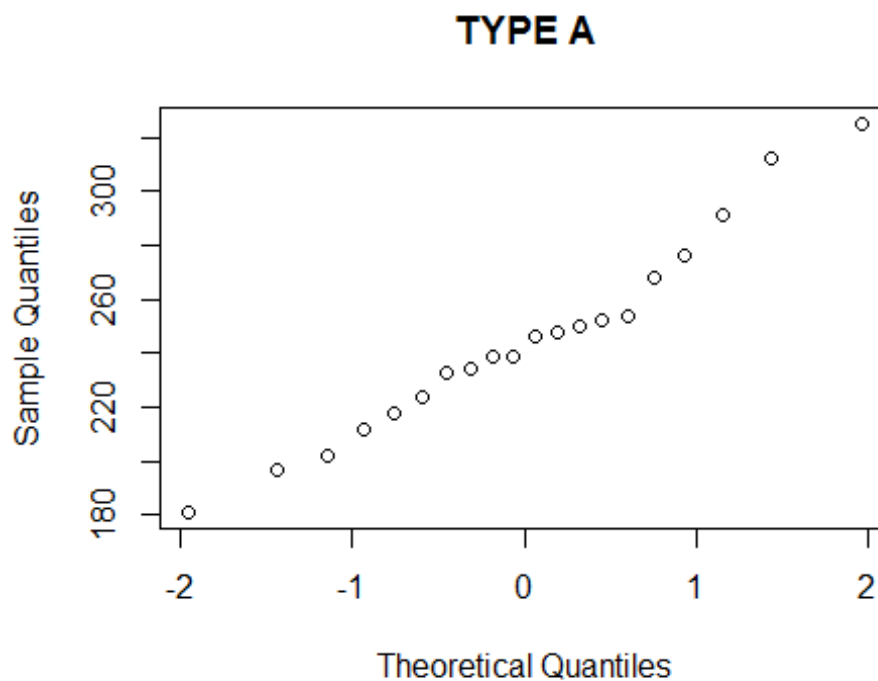
We are interested on drawing inferences about  $\Delta = \mu_1 - \mu_2$

(e)  $\Delta > 0$  iff  $\mu_1 > \mu_2$ . Thus, to document that Type A have higher cholesterol than Type B, we might want

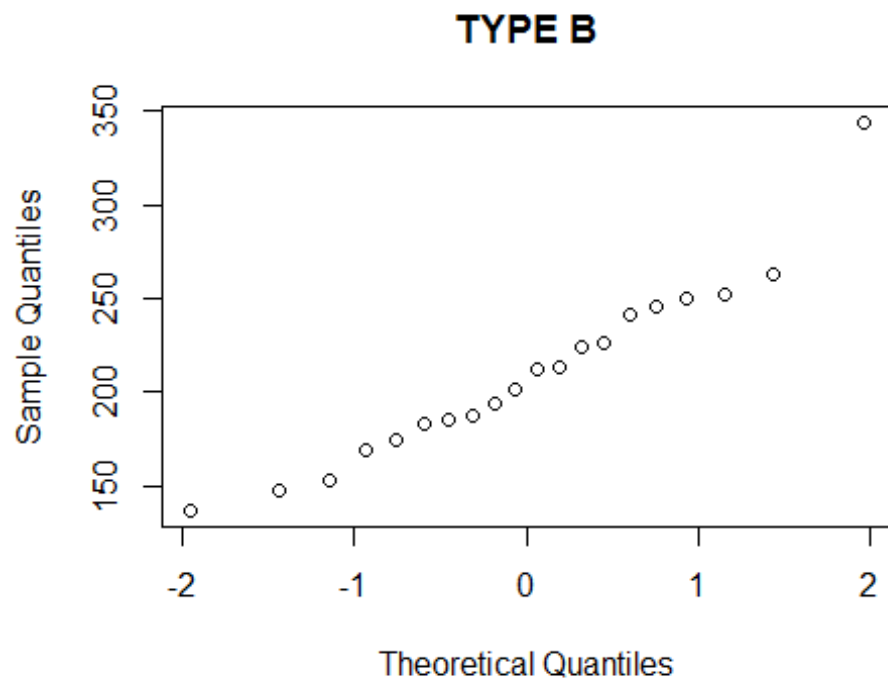
to test  $H_0 : \Delta \leq 0$  vs.  $H_a : \Delta > 0$ .

C2

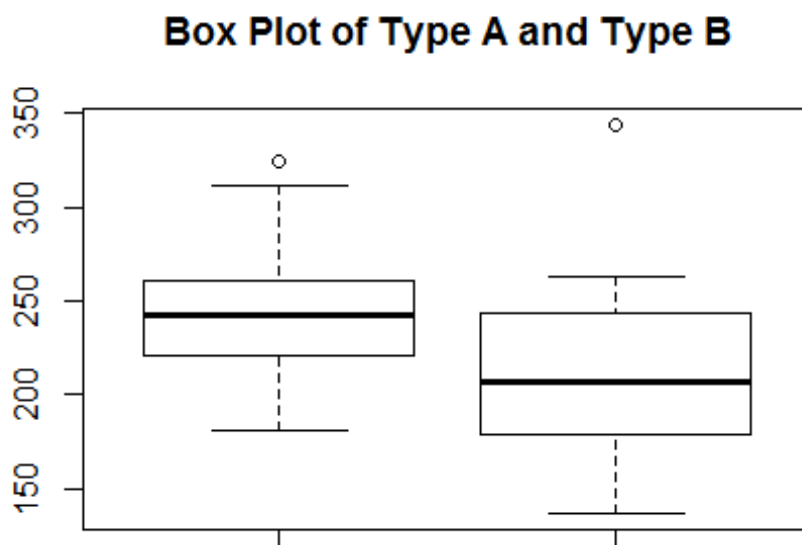
```
typea<- c(233,291,312,250,246,197,268,224,239,239,254,276,234,181,248,252,
          202,218,212,325)
typeb<- c(344,185,263,246,224,212,188,250,148,169,226,175,242,252,153,183,
          137,202,194,213)
qqnorm(typea,main = "TYPE A")
```



```
qqnorm(typeb,main = "TYPE B")
```



```
boxplot(typea,typeb,main="Box Plot of Type A and Type B")
```



*# QQplot for both Type A and Type B suggests some values may be inconsistent  
# with normal distribution specially largest in each set as seen in boxplot.*

```

a=IQR(typea)/sqrt(var(typea))
a
## [1] 0.9552842

b=IQR(typeb)/sqrt(var(typeb))
b
## [1] 1.282584

# Ratio for Type B suggest sample more close to normal distribution but
# also has large outlier hence I would not assume data was drawn from
# normal distribution although there is slight chance of being picked up
# from normal distribution.
delta<- mean(typea)- mean(typeb)
n1=length(typea)
n2=length(typeb)
va=var(typea)/n1
vb=var(typeb)/n2
se=sqrt(va+vb)
nu<- (va+vb)^2/(va^2/(n1-1)+vb^2/(n2-1))
# If we let alpha= 0.05 then
1- pt(2.5621,nu)
## [1] 0.00740548

# 0.007405 < 0.05 = alpha -> reject H0
# b) We want 90% confidence interval for delta,
qt=qt(0.95,nu)
lower=delta-qt*se
upper=delta+qt*se
lower
## [1] 11.84155
upper
## [1] 57.65845

```

3.

Let  $\mu_d$  be the mean urinary  $\beta$ -thromboglobulin excretion in diabetic patients.

Let  $\mu_n$  be the mean urinary  $\beta$ -thromboglobulin excretion in normal patients.

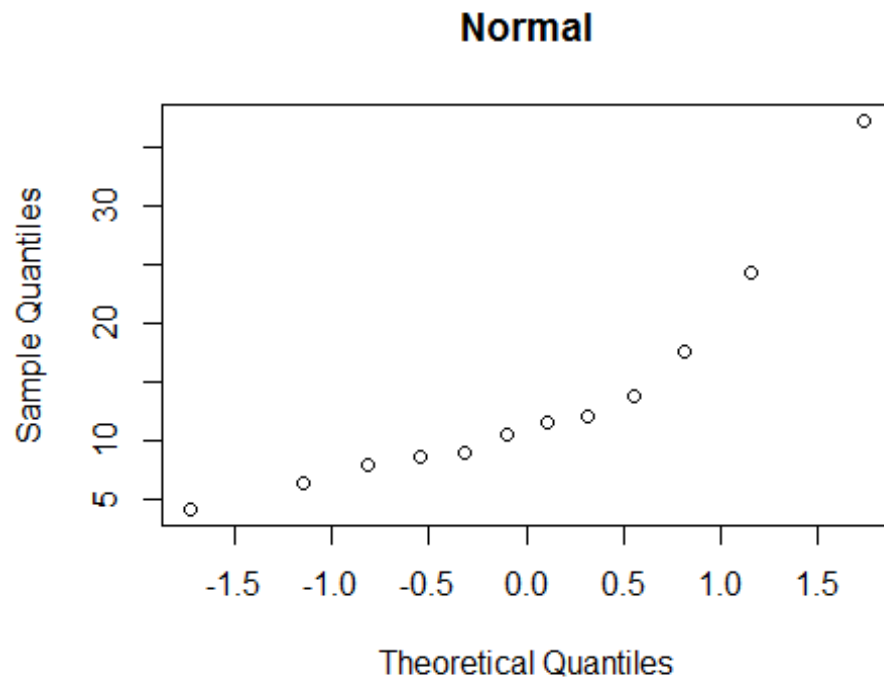
$\delta = \mu_d - \mu_n$  be the difference in mean urinary  $\beta$ -thromboglobulin excretion of diabetic and normal patients respectively.

### Hypothesis Test

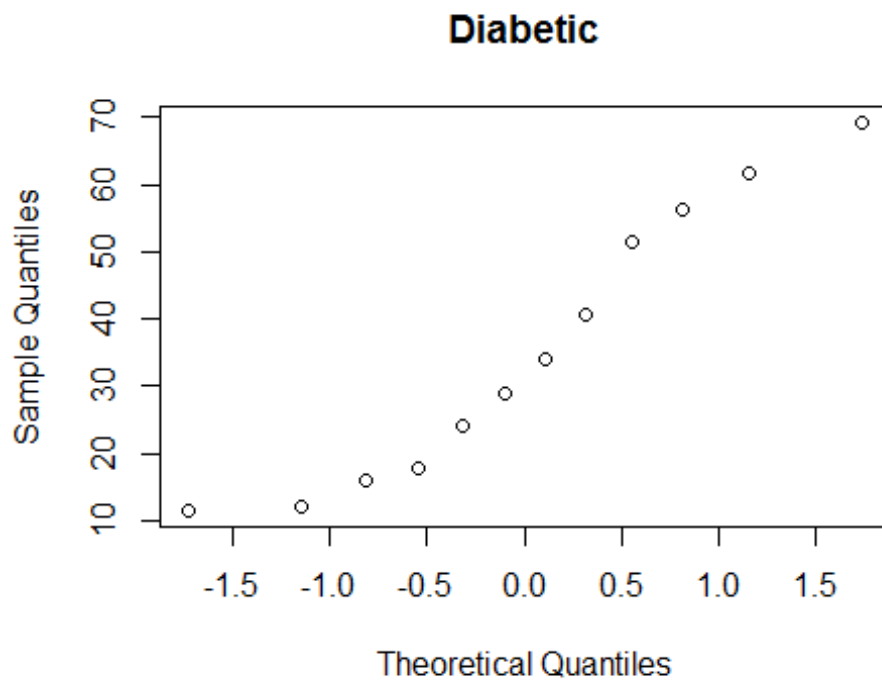
$H_0 : \delta \leq 0$

$H_a : \delta > 0$

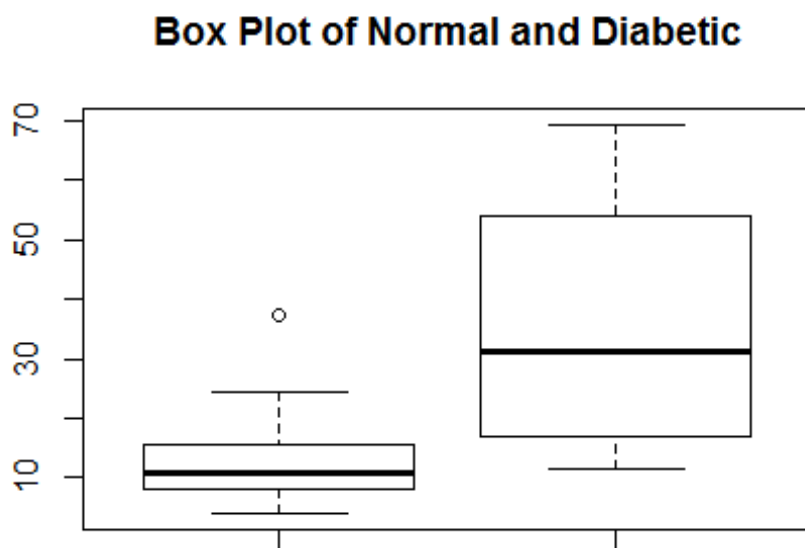
```
normal<- c(4.1,6.3,7.8,8.5,8.9,10.4,11.5,12.0,13.8,17.6,24.3,37.2)
diabetic<- c(11.5,12.1,16.1,17.8,24.0,28.8,33.9,40.7,51.3,56.2,61.7,69.2)
qqnorm(normal,main="Normal")
```



```
qqnorm(diabetic,main = "Diabetic")
```

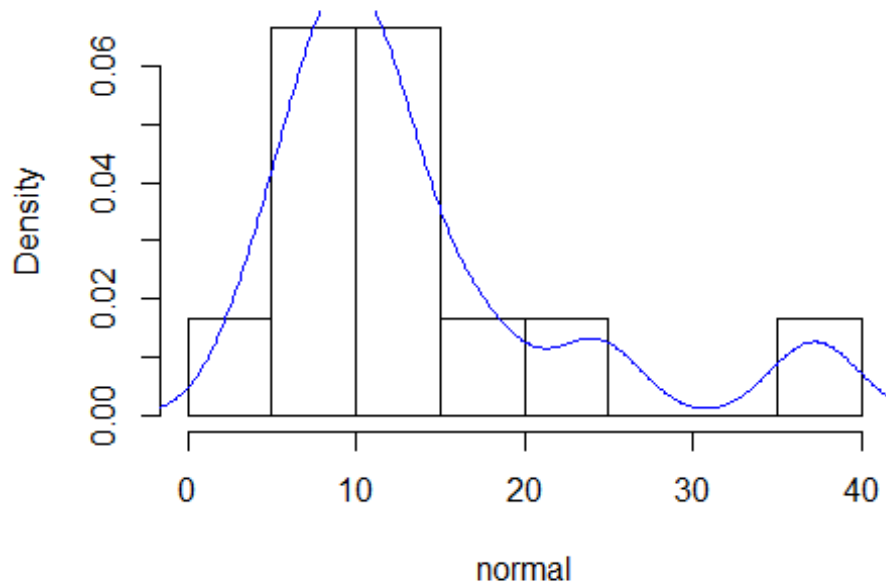


```
boxplot(normal,diabetic,main="Box Plot of Normal and Diabetic")
```



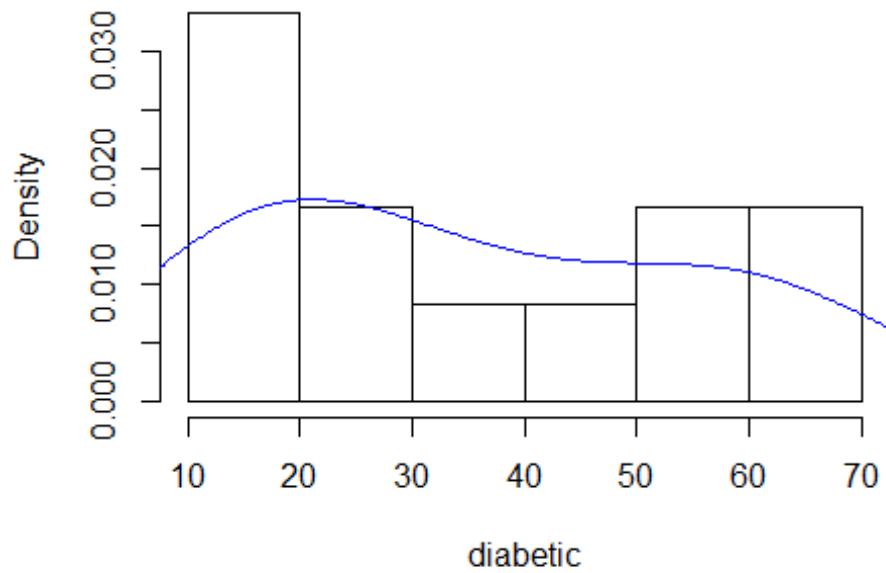
```
hist(normal,prob=TRUE)  
lines(density(normal),col="blue")
```

**Histogram of normal**

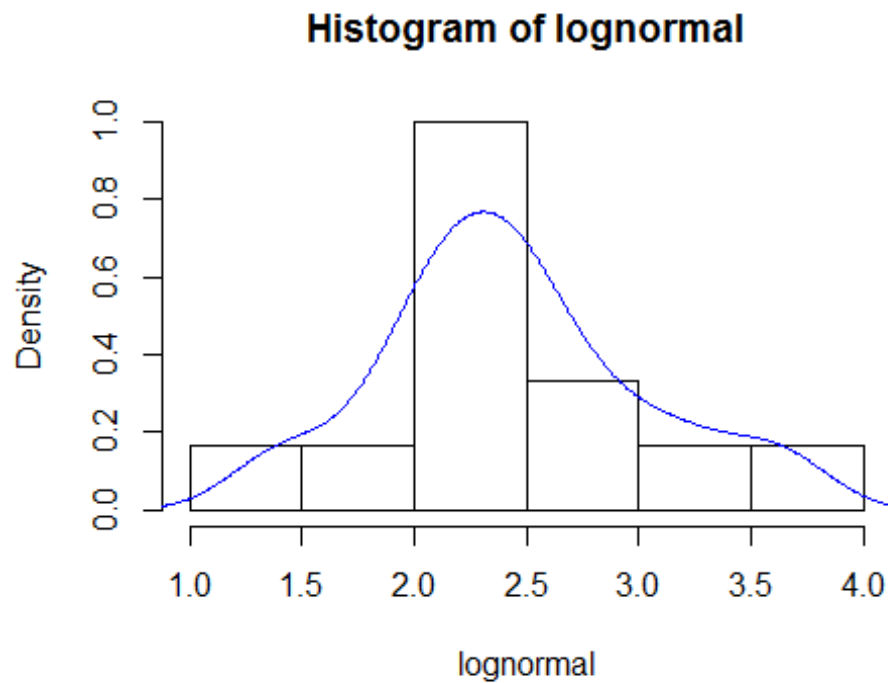


```
hist(diabetic,prob=TRUE)  
lines(density(diabetic),col="blue")
```

**Histogram of diabetic**

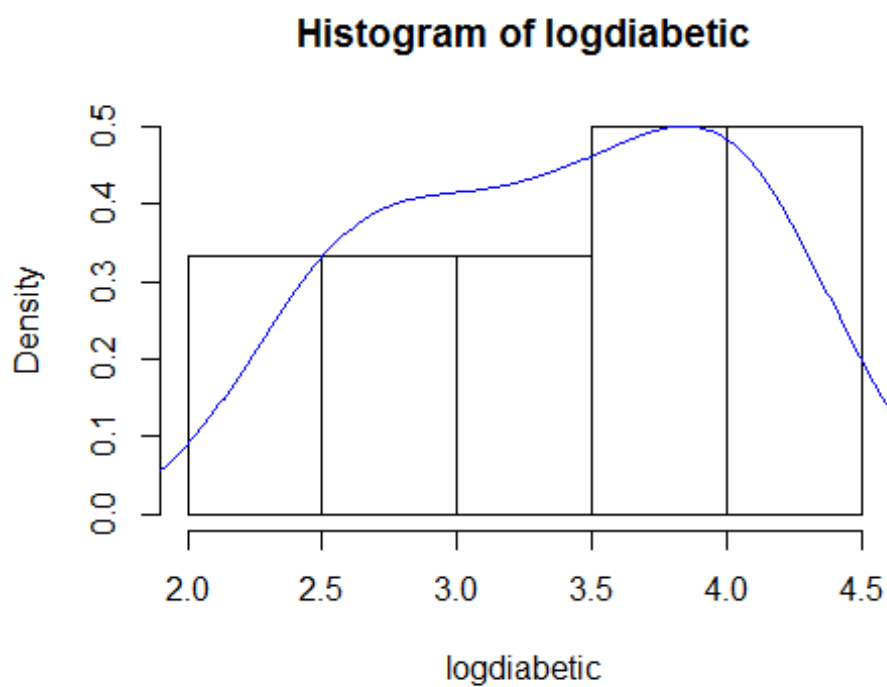


```
# 1 )
#After seeing qqplot,boxplot and histogram we can say that samples are
# not drawn from normal distribution.
# 2)
# (a) Natural Logarithm
lognormal<- log(normal)
logdiabetic<- log(diabetic)
hist(lognormal,prob=TRUE)
lines(density(lognormal),col="blue")
```

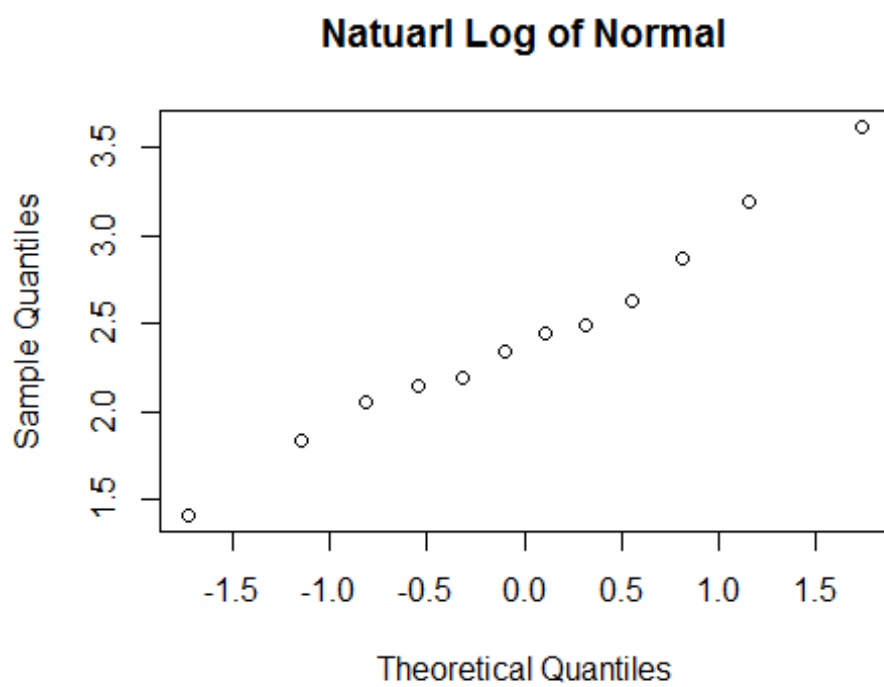


```
hist(logdiabetic,prob=TRUE)
lines(density(logdiabetic),col="blue")
```

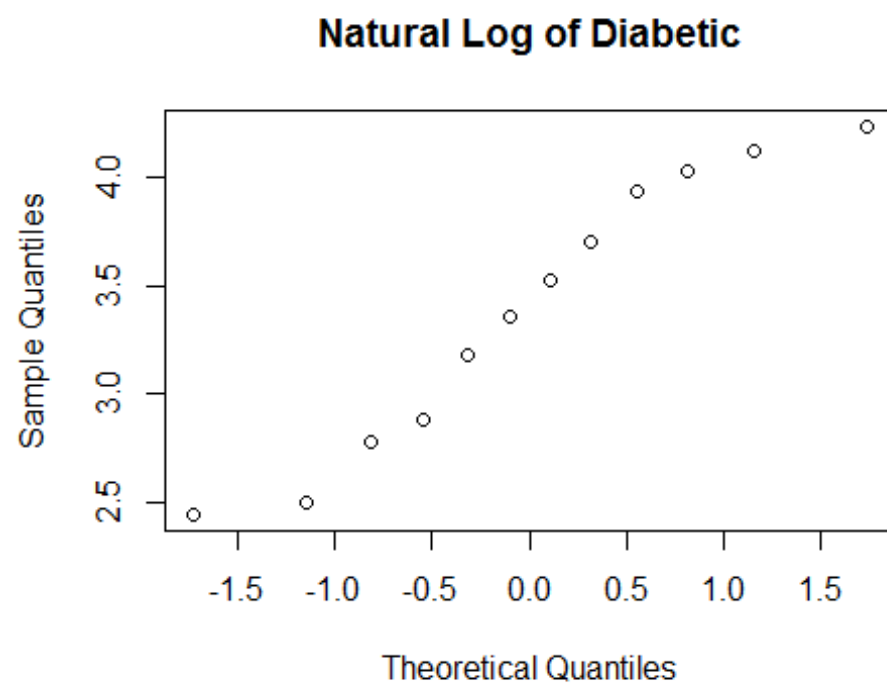




```
qqnorm(lognormal,main="Natural Log of Normal")
```

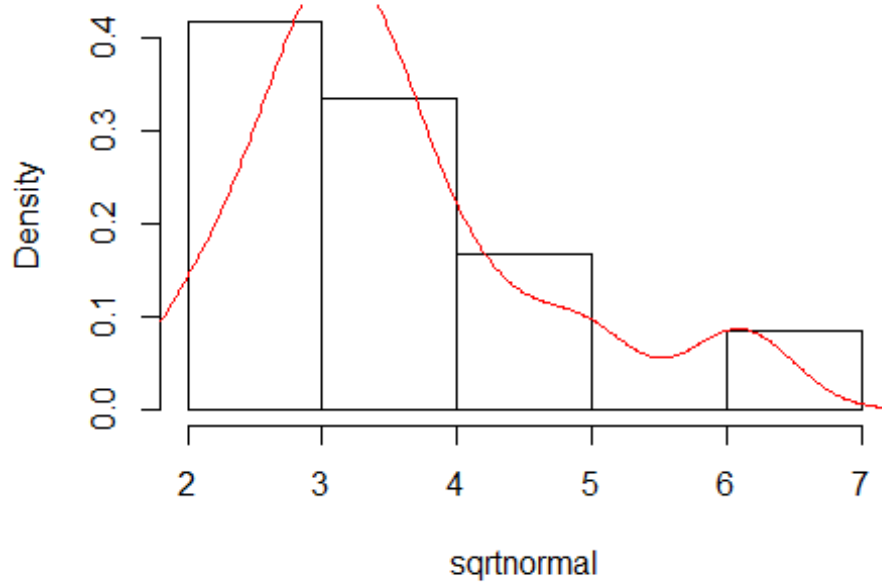


```
qqnorm(logdiabetic,main = "Natural Log of Diabetic")
```



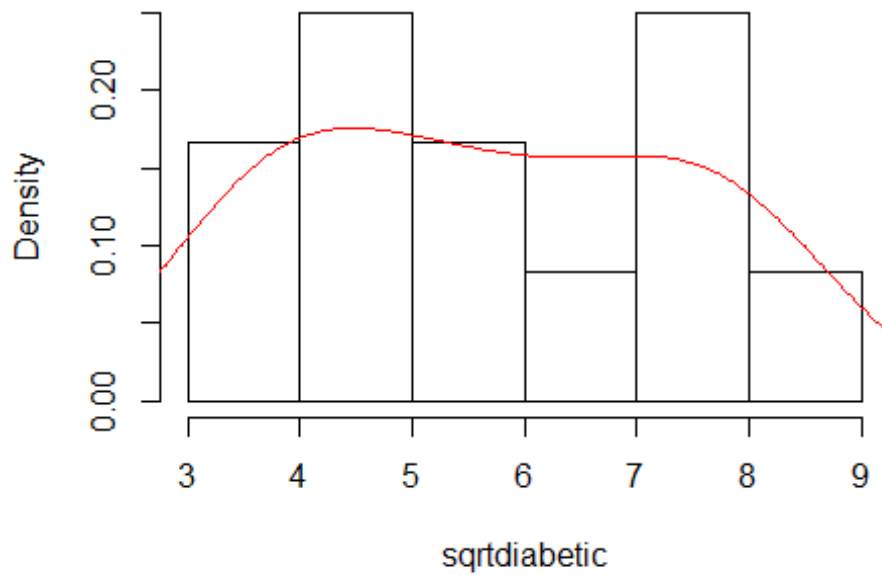
```
# (b) Square Root  
sq rtnormal<-sqrt(normal)  
sq rtdiabetic<-sqrt(diabetic)  
hist(sq rtnormal,prob=TRUE)  
lines(density(sq rtnormal),col="red")
```

**Histogram of sqrtnormal**



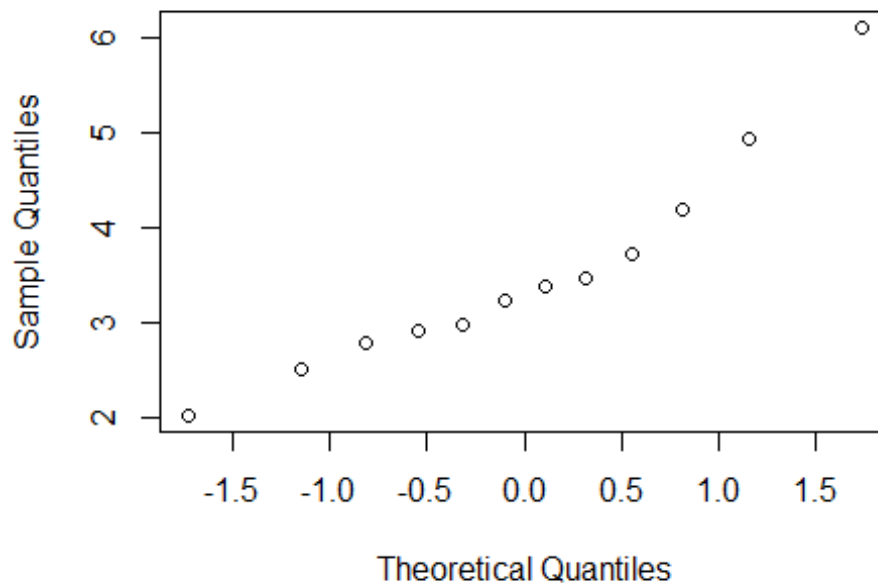
```
hist(sqrtdiabetic,prob=TRUE)  
lines(density(sqrtdiabetic),col="red")
```

**Histogram of sqrtdiabetic**



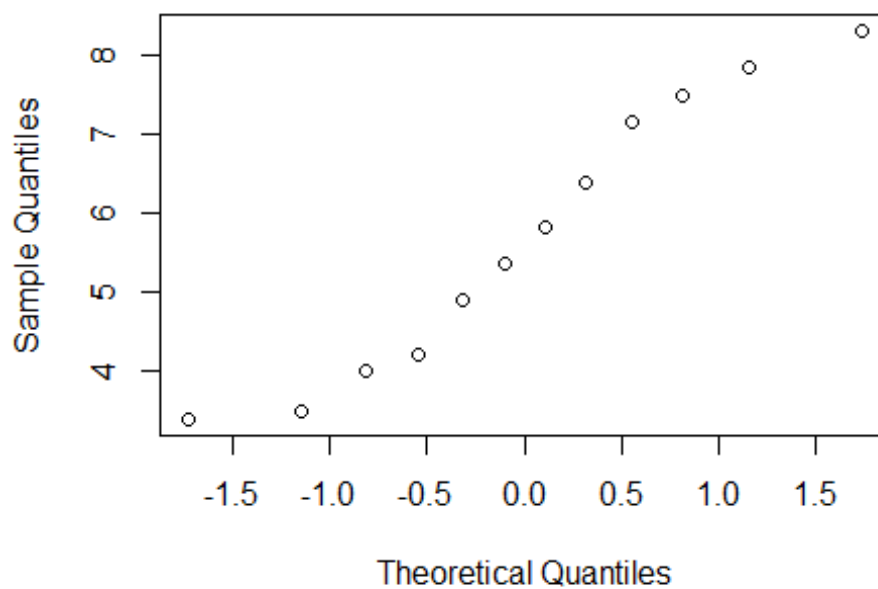
```
qqnorm(sqrtnormal,main="Square root of Normal")
```

### Square root of Normal



```
qqnorm(sqrtdiabetic,main ="Square root of Diabetic")
```

### Square root of Diabetic



```
# I would prefer log transformation over square root transformation since  
# log transformation is more symmetric to normal distribution.
```

```

# 3)
# As seen from histograms, density plots and qqplots, Log transformed
# measurements appear closer to normal distribution.
# 4)
# Welch's t-test
Delta = mean(logdiabetic) - mean(lognormal)
se = sqrt(var(logdiabetic)/12 + var(lognormal)/12)
Tw = Delta/se
nu = (var(logdiabetic)/12+var(lognormal)/12)^2/((var(logdiabetic)/12)^2/11+(v
ar(lognormal)/12)^2/11)
Pvalue = 2*(1-pt(abs(Tw),df=nu))
Pvalue

## [1] 0.0009776127

# Welch 95% confidence interval
q = qt(0.975, df=nu)
lower = Delta - q*se
upper = Delta + q*se
CI<-c(lower,upper)
CI

## [1] 0.4352589 1.4792986

# Since P-value is quite low we can reject H0 in favor of Ha.

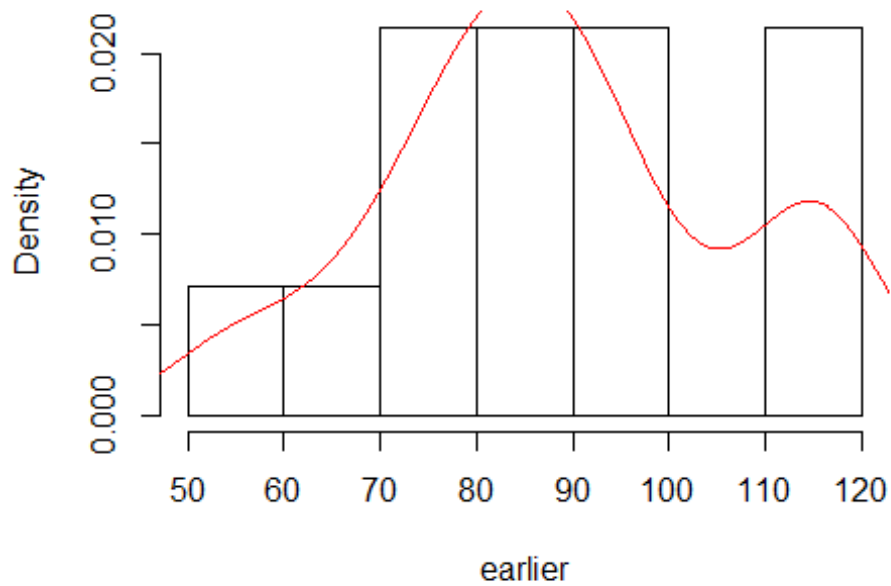
4.

Let  $\mu_e$  be the mean movie length in ealier years (1956)
Let  $\mu_t$  be the mean movie length in Todays years (1996)
 $\delta = \mu_t - \mu_e$  be the difference in mean movie length time respectively. Hypothesis Test
 $H_0$  be the hypothesis that  $\delta \leq 0$ 
 $H_a$  be the hypothesis that  $\delta > 0$ 

earlier<-c(74,114,114,87,92,55,67,118,79,83,79,92,99,87)
now<-c(70,98,90,95,88,108,110,96,91,88,120,96,90,90)
hist(earlier,prob=TRUE)
lines(density(earlier),col="red")

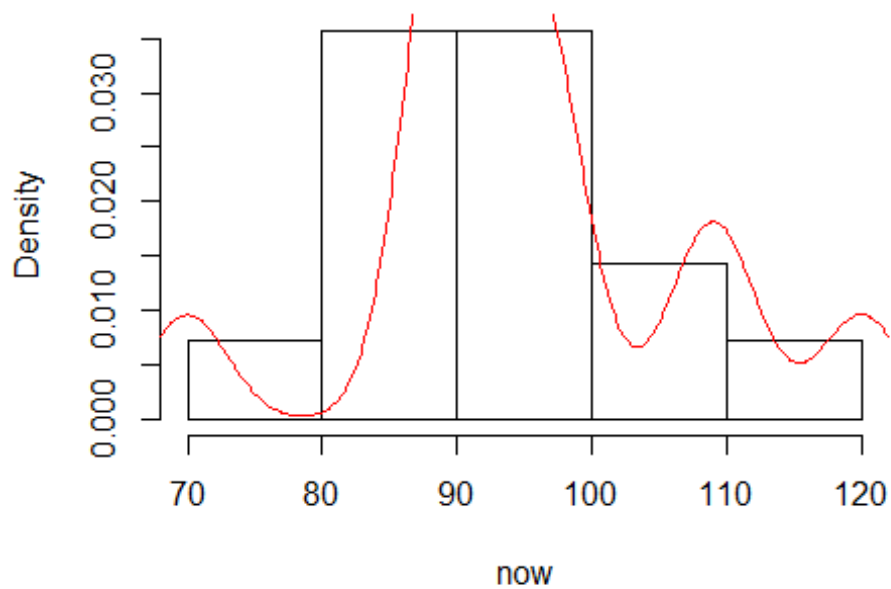
```

**Histogram of earlier**



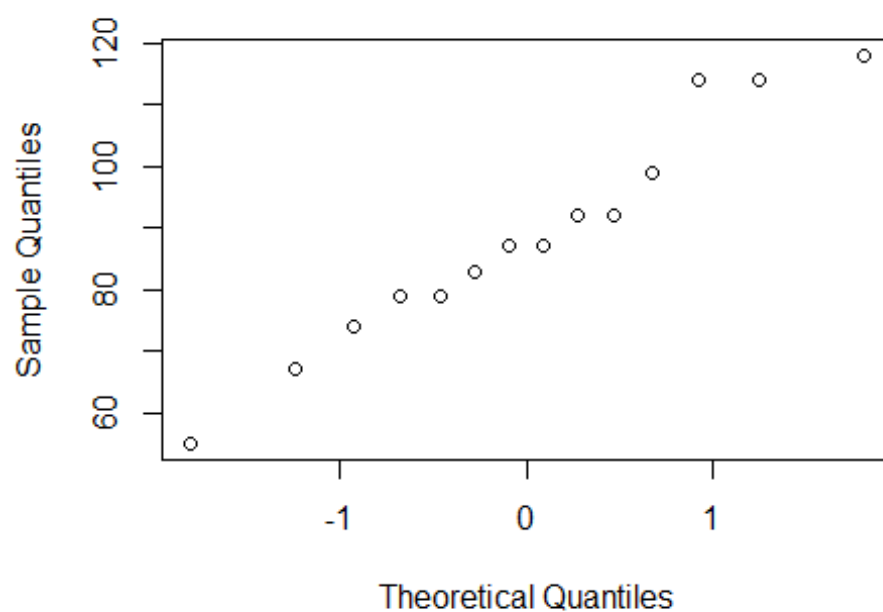
```
hist(now,prob=TRUE)  
lines(density(now),col="red")
```

**Histogram of now**



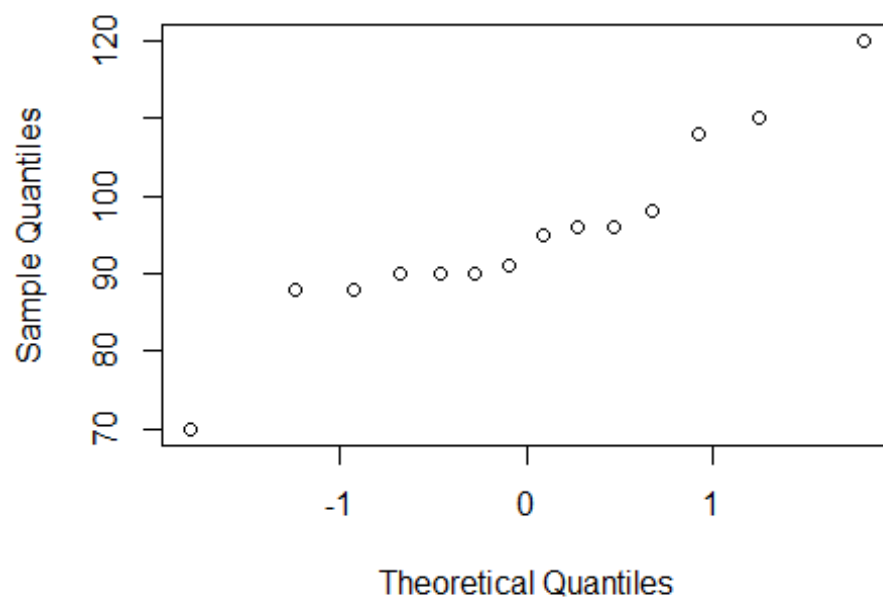
```
qqnorm(earlier,main="Feature Films made in 1956")
```

### Feature Films made in 1956



```
qqnorm(now,main="Feature Films made in 1996")
```

### Feature Films made in 1996



```
t.test(now,earlier)
```

```
##
## Welch Two Sample t-test
##
## data: now and earlier
## t = 1.105, df = 22.395, p-value = 0.2809
## alternative hypothesis: true difference in means is not equal to 0
## 95 percent confidence interval:
## -5.623821 18.480963
## sample estimates:
## mean of x mean of y
## 95.00000 88.57143

# P-value =0.2809 which shows that mean length of movies in 1956 is more
# than mean length of movies in 1996 with 28% probability assuming null
# hypothesis is true.
# 95% Confidence Interval is (-5.623,18.480) shows that difference in movie
# is not necessarily above 0.
# 0.01 and 0.05 are ideal significance values in hypothesis testing which
# is read as probability of null hypothesis being true.
# After conducting welch's t-test which is based on normality of samples
# datasets are normally distributed as seen from plots to conduct experiment.
```