Lab0

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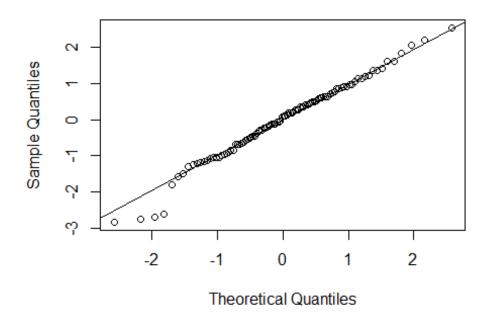
Lab 0 Problems

Question 1.a

Generate a random sample of 100 from t-distribtion, degree of freedom 10.

```
Check qqnorm(); qqline(), Shapiro test. Remarks
sample.t <- rt(100, df=10)
qqnorm(sample.t)
qqline(sample.t)</pre>
```

Normal Q-Q Plot



```
shapiro.test(sample.t)
##
## Shapiro-Wilk normality test
##
```

```
## data: sample.t
## W = 0.98385, p-value = 0.2618
```

- 1. Null hypothesis (H0 sample data is normal)
- 2. Alternative hypothesis (H1 sample data is not normal)

Since the pvalue > 0.05, which means that the null hypothesis that values are normal is accepted and the alternative hypothesis is rejected. ***

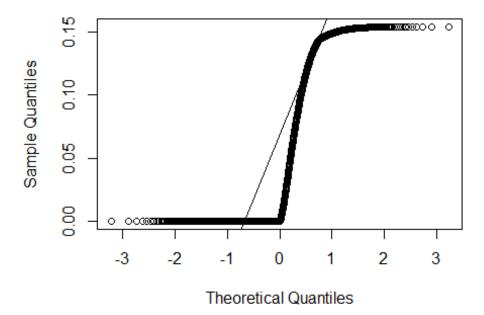
Question 1.b

Generate a random sample of 100 from a Chi-square distribution, df =5.

Perform same procedures as in (a). Remarks.

```
sample.1b <-seq(-4, 4, 0.01)
sample.chi = dchisq(sample.1b, df=5)
qqnorm(sample.chi)
qqline(sample.chi)</pre>
```

Normal Q-Q Plot



```
shapiro.test(sample.chi)
##
## Shapiro-Wilk normality test
##
```

```
## data: sample.chi
## W = 0.728, p-value < 2.2e-16
```

- 1. Null hypothesis (H0 sample data is normal)
- 2. Alternative hypothesis (H1 sample data is not normal)

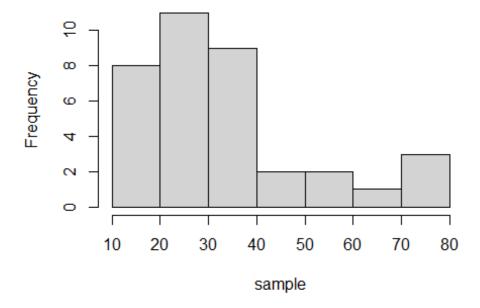
Since the pvalue < 0.05, which means that the null hypothesis that values are normal is rejected and the alternative hypothesis is accepted.

Question 2.a

Write a 95%-CI for the population mean.

Confidence interval for population mean can be determined using the following

Histogram of sample



```
n<-length(sample)
xbar<-mean(sample)</pre>
```

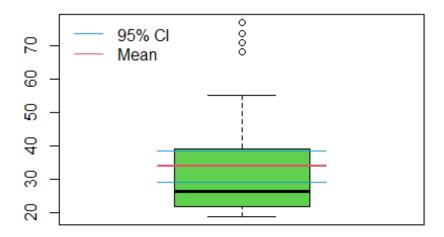
```
s<-sd(sample)
sprintf("mean %f, std dev %f, sample size %d", xbar, s, n)
## [1] "mean 33.800000, std dev 16.897489, sample size 36"

margin <- qt(0.95,df=n-1)*s/sqrt(n)
low_ci <- xbar - margin

high_ci <- xbar + margin
sprintf("confidence interval for the population mean is (%f, %f)", low_ci, high_ci)

## [1] "confidence interval for the population mean is (29.041745, 38.558255)"

boxplot(sample, col = 3)
lines(c(0.75,1.25),c(low_ci,low_ci),col=4)
lines(c(0.75,1.25),c(xbar,xbar),col=2,lwd=2)
lines(c(0.75,1.25),c(high_ci,high_ci),col=4)
legend("topleft", c("95% CI", "Mean"), lty=1,col = c(4, 2),bty ="n")</pre>
```



What assumption about population for the work, suppose the sample is random.

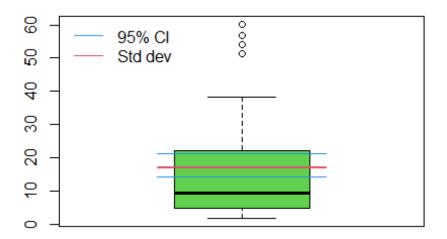
The assumptions would be that the data is normal and the errors are constant.

Question 2.b

Write a 95%- CI for population standard deviation.

Confidence interval for population standard deviation can be determined using the following formula confidence interval: $CI = [\sqrt{(n-1)s^2/X^2(\alpha/2)}, \sqrt{(n-1)s^2/X^2(1-\alpha/2)}]$

```
sample <-c(26.4,23.5,25.4,22.9,25.2,39.2,25.5,31.9,26.0,44.6,35.5,38.6,
           30.1,31.0,30.8,32.8,47.7,39.1,55.3,50.7,73.8,71.1,68.4,77.1,
           19.4,19.3,18.7,19.0,23.2,21.3,23.2,19.9,18.9,19.8,19.6,21.9)
n <- length(sample)</pre>
xbar<-mean(sample)</pre>
s<-sd(sample)</pre>
sprintf("population sd %f, size %d, mean %f", s, n, xbar)
## [1] "population sd 16.897489, size 36, mean 33.800000"
left <- qchisq(p=.05, df=n-1, lower.tail=FALSE)</pre>
right <- qchisq(p=.95, df=n-1, lower.tail=FALSE)
low_ci \leftarrow sqrt((n-1)*s^2/left)
high_ci <- sqrt((n-1)*s^2/right)
sprintf("Confidence interval for population std dev is %f, %f", low_ci,
high_ci)
## [1] "Confidence interval for population std dev is 14.165551, 21.091275"
boxplot(sample-s, col = 3)
lines(c(0.75,1.25),c(low_ci,low_ci),col=4)
lines(c(0.75,1.25),c(s,s),col=2,lwd=2)
lines(c(0.75,1.25),c(high_ci,high_ci),col=4)
legend("topleft", c("95% CI", "Std dev"), lty=1,col = c(4, 2),bty ="n")
```

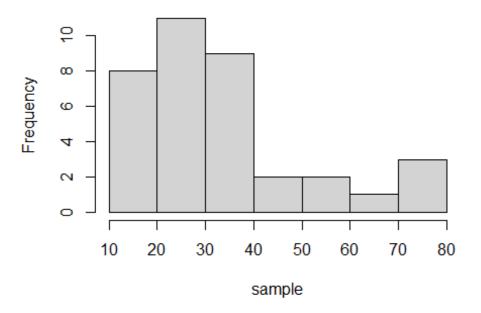


A confidence interval for a standard deviation is a range of values that is likely to contain true population standard deviation with a certain level of confidence.

Question 3

Run the following code

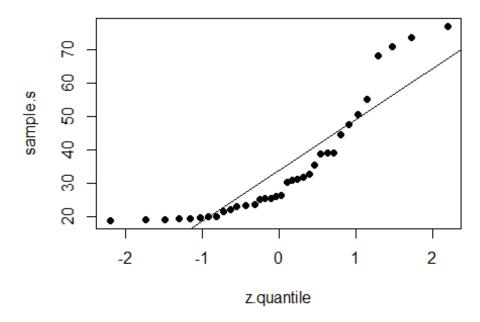
Histogram of sample



```
sample.s <-sort(sample) #sort data increasing</pre>
rank <- rank(sample.s) #rank data from 1 to 36</pre>
size <- length(sample.s) # size of data</pre>
p <- (rank-.5)/size #cummulative prob of data
z.quantile <- qnorm(p) # Standard Normal quantiles with such probability
data <- cbind(rank, sample.s, p, z.quantile)</pre>
data
##
         rank sample.s
                                  p z.quantile
##
    [1,]
          1.0
                   18.7 0.01388889 -2.20041058
##
    [2,]
          2.0
                   18.9 0.04166667 -1.73166440
##
    [3,]
          3.0
                   19.0 0.06944444 -1.47994139
##
          4.0
    [4,]
                   19.3 0.09722222 -1.29754293
##
          5.0
                   19.4 0.12500000 -1.15034938
    [5,]
    [6,]
##
          6.0
                   19.6 0.15277778 -1.02459239
##
                   19.8 0.18055556 -0.91324993
    [7,]
          7.0
##
    [8,]
          8.0
                   19.9 0.20833333 -0.81221780
##
    [9,]
          9.0
                   21.3 0.23611111 -0.71886805
                   21.9 0.26388889 -0.63140189
## [10,] 10.0
                   22.9 0.29166667 -0.54852228
## [11,] 11.0
## [12,] 12.5
                   23.2 0.33333333 -0.43072730
## [13,] 12.5
                   23.2 0.33333333 -0.43072730
## [14,] 14.0
                   23.5 0.37500000 -0.31863936
## [15,] 15.0
                   25.2 0.40277778 -0.24616365
## [16,] 16.0
                   25.4 0.43055556 -0.17495994
## [17,] 17.0
                   25.5 0.45833333 -0.10463346
```

```
## [18,] 18.0
                  26.0 0.48611111 -0.03482132
## [19,] 19.0
                  26.4 0.51388889
                                    0.03482132
## [20,] 20.0
                  30.1 0.54166667
                                    0.10463346
## [21,] 21.0
                  30.8 0.56944444
                                    0.17495994
## [22,] 22.0
                  31.0 0.59722222
                                    0.24616365
## [23,] 23.0
                  31.9 0.62500000
                                    0.31863936
## [24,] 24.0
                  32.8 0.65277778
                                    0.39283081
## [25,] 25.0
                  35.5 0.68055556
                                    0.46925288
## [26,] 26.0
                  38.6 0.70833333
                                    0.54852228
## [27,] 27.0
                  39.1 0.73611111
                                    0.63140189
## [28,] 28.0
                  39.2 0.76388889
                                    0.71886805
## [29,] 29.0
                  44.6 0.79166667
                                    0.81221780
## [30,] 30.0
                  47.7 0.81944444
                                    0.91324993
## [31,] 31.0
                  50.7 0.84722222
                                    1.02459239
## [32,] 32.0
                  55.3 0.87500000
                                    1.15034938
## [33,] 33.0
                  68.4 0.90277778
                                    1.29754293
## [34,] 34.0
                  71.1 0.93055556
                                    1.47994139
## [35,] 35.0
                  73.8 0.95833333
                                    1.73166440
## [36,] 36.0
                  77.1 0.98611111
                                    2.20041058
#scatterplot of x=Z quantiles, y= data sorted
plot(x=z.quantile, y=sample.s, pch=16, main="QQ Plot")
abline(lm(sample.s ~ z.quantile))
```





The QQ plot suggests that the data is not normal