HW6

Helped by: 106034054 106070020

Q1

(a) Visualize Verizon's response times for ILEC vs. CLEC customers

```
data <- read.csv("verizon.csv", header = TRUE)

ILEC_time <- subset(data, Group == 'ILEC')

CLEC_time <- subset(data, Group == 'CLEC')

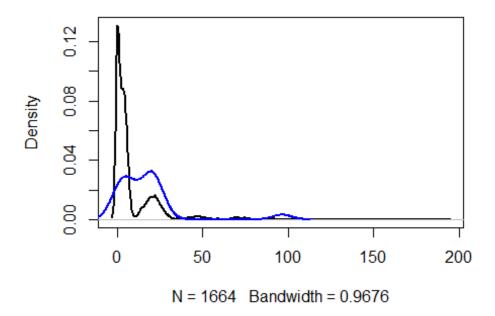
ILEC <- ILEC_time$Time

CLEC <- CLEC_time$Time

plot(density(ILEC), lwd = '2')

lines(density(CLEC), lwd = '2', col = 'blue')</pre>
```

density.default(x = ILEC)



- (b) test the difference between the mean of ILEC sample response times versus the mean of CLEC sample response times.
 - (i) What are the appropriate null and alternative hypotheses in this case?

 H_0 :ILEC sample mean >= CLEC sample mean H_1 : ILEC sample mean < CLEC sample mean

(ii) Based on output, would you reject the null hypothesis or not?

```
t.test(ILEC, CLEC, alternative = 'less',
      var.equal = FALSE,conf.level = 0.99)
##
##
   Welch Two Sample t-test
##
## data: ILEC and CLEC
## t = -1.9834, df = 22.346, p-value = 0.02987
## alternative hypothesis: true difference in means is less th
an 0
## 99 percent confidence interval:
##
       -Inf 2.130858
## sample estimates:
## mean of x mean of y
## 8.411611 16.509130
```

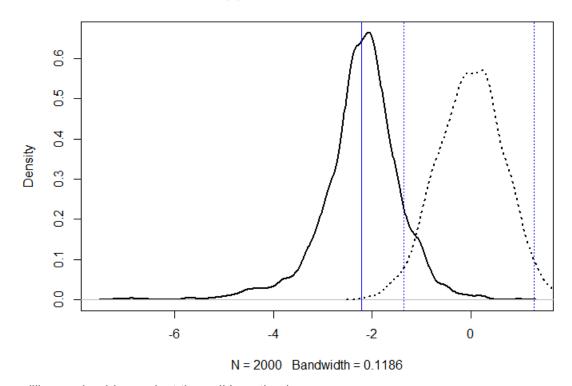
As the p-value is bigger than 1%, we should not reject H_0

- (c) Bootstrapped samples of ILEC against samples of CLEC & samples of ILEC against the original ILEC sample
 - (i) Plot a distribution of the bootstrapped null t-values and alternative t-values, adding vertical lines to show the 5% rejection zone of the null distribution.

```
boot_result_1 <-replicate(2000,boot_t_test(ILEC, CLEC)$statistic)
boot_result_2 <- replicate(2000,boot_sampleO_only(ILEC, ILEC)$sta
tistic)

plot(density(boot_result_1),lwd = 2, main = 'bootstrapped null an
d alternative t-values')
abline(v = mean(boot_result_1),col = 'blue')
lines(density(boot_result_2),lwd = 2,lty = 'dotted')
abline(v = quantile(boot_result_2,probs = 0.025),col = 'blue',lty
= 'dotted')
abline(v = quantile(boot_result_2,probs = 0.975),col = 'blue',lty
= 'dotted')</pre>
```

bootstrapped null and alternative t-values



(ii) should we reject the null hypothesis

Yes, we should reject the null hypothesis because median of alternative is out of the 95% null lines.

Q2 Variance

(a) null and alternative hypotheses in this case?

 H_0 :ILEC variance = CLEC variance

 H_1 : ILEC variance < CLEC variance

(b) traditional statistical methods

(i) F-statistic of the ratio of variances

```
data <- read.csv("verizon.csv", header = TRUE)</pre>
ILEC time <- subset(data, Group == 'ILEC')</pre>
CLEC_time <- subset(data, Group == 'CLEC')</pre>
ILEC <- ILEC time$Time</pre>
CLEC <- CLEC time$Time
var.test(ILEC, CLEC, alternative = 'less')
##
## F test to compare two variances
##
## data: CLEC and ILEC
## F = 1.7627, num df = 22, denom df = 1663, p-value = 0.01582
## alternative hypothesis: true ratio of variances is greater th
an 1
## 95 percent confidence interval:
## 1.138356
                  Inf
## sample estimates:
## ratio of variances
##
            1,762717
```

(ii) cut-off value of F

```
qf(0.95, df1 = length(CLEC)-1, df2 = length(ILEC)-1)
## [1] 1.548476
```

- (iii) Can we reject the null hypothesis

 Reject the null hypothesis. As the F-value in (ii) is bigger than the cut-off value which means in the cut-off zone.
- (c) bootstrapping
 - (i) bootstrapped values of the F-statistic, <u>for both null and alternative</u> <u>hypotheses</u>.

```
f_test_boot <- function(larger_sd_sample, smaller_sd_sample){
    resample_sd1 <- sample(larger_sd_sample, length(larger_sd_sample),
    replace = TRUE)
    resample_sd2 <-sample(smaller_sd_sample,length(smaller_sd_sample),
    replace = TRUE)
    f_alt <- var(resample_sd1)/var(resample_sd2)</pre>
```

```
f_null <- var(resample_sd1)/var(larger_sd_sample)
  c(f_alt, f_null)
}
f_boot <- replicate(2000, f_test_boot(CLEC, ILEC))
f_alts <- f_boot[1,]
f_nulls <- f_boot[2,]</pre>
```

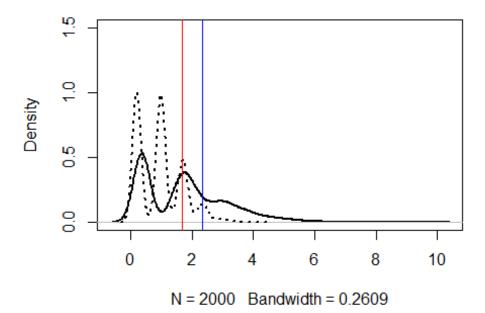
(ii) 95% cutoff value according to the bootstrapped null values of F?

```
boot_var_95 <-quantile(f_nulls, probs = 0.95)
boot_var_95
## 95%
## 2.360329</pre>
```

(iii) a visualization of the null and alternative distributions of the bootstrapped Fstatistic, with vertical lines at the cutoff value of F nulls.

```
plot(density(f_alts),lwd = 2, main = 'bootstrapped null and
  alternative F test', ylim = c(0,1.5))
lines(density(f_nulls), lwd = 2,lty = 'dotted')
abline(v = boot_var_95,col = 'blue')
abline(v = median(f_alts), col = 'red')
```

bootstrapped null and alternative F test



(iv) bootstrap results suggest about the null hypothesis?

We should not reject the null hypothesis because the median of the alternative is inside the 95% zone.

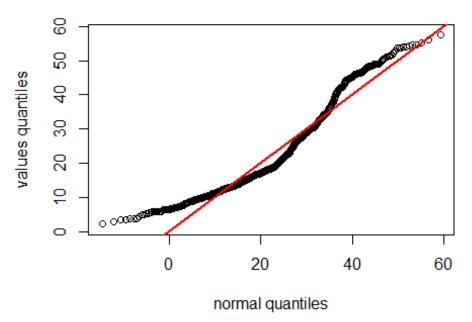
Q3

(a) Let's create a function to see if key statistics/assumptions of *normality* are met in our distributions.

```
norm_qq_plot<- function(values){
  probs1000 <- seq(0, 1, 0.001)
  q_vals <- quantile(values, probs = probs1000)
  q_norm <- qnorm(probs1000, mean = mean(values), sd =
  sd(values))
  plot(q_norm, q_vals, xlab = 'normal quantiles', ylab =
  'values quantiles')
  abline(a=0, b= 1, col = 'red', lwd = 2)
}</pre>
```

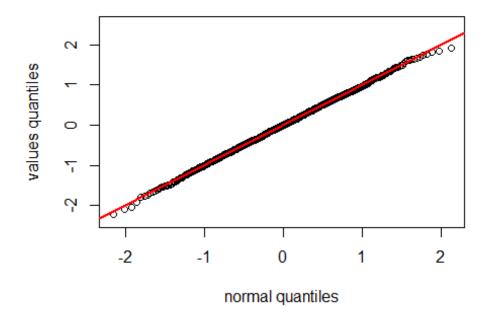
(b) running it against the values of our d123 distribution





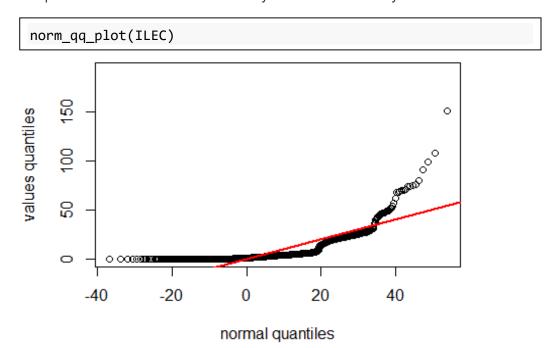
The dots doesn't lie on the red line so it is not a normal distribution.

(c) Use your normal Q-Q plot function to check if the bootstrapped distribution of null t-values in question 1c was normally distributed. What's your conclusion?

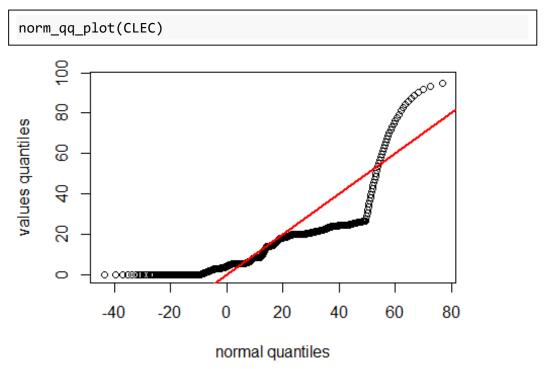


The dots lie on the red line so it is not a normal distribution.

(d) Use your normal Q-Q plot function to check if the two samples we compared in question 2 could have been normally distributed. What's your conclusion?



The dots doesn't lie on the red line so it is not a normal distribution.



The dots doesn't lie on the red line so it is not a normal distribution.