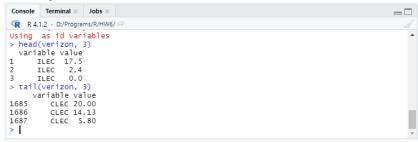
Due: 2022/03/27

Question 1. (a) I pick melt function from reshape library. It's not hard to call. The usage is refer to [1].

(b) To reshape the data with melt function, consider

- (c) Use the command in console:
- > head(verizon, 3)
- > tail(verizon, 3)

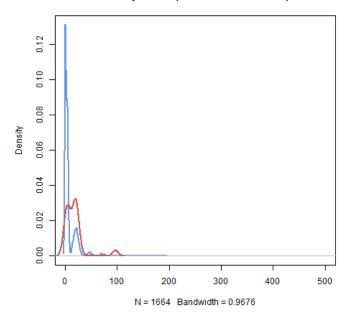
I prefer to show the result by a screenshot.



(d) Simply reuse the codes from the slides.

```
# Question 1 (d)
png(filename = "1d.png")
plot(density(hosts$ILEC$value), col="cornflowerblue", lwd=2, xlim=c(0, 500))
lines(density(hosts$CLEC$value), col="coral3", lwd=2)
legend(300, 0.5, lty=1, c("ILEC", "CLEC"), col=c("coral3", "cornflowerblue"))
dev.off()
```

density.default(x = hosts\$ILEC\$value)



Question 2. (a) H_0 : the mean of response times for CLEC customers is equal to that of ILEC customers; H_1 : The mean of response times for CLEC customers is greater than for ILEC customers.

(b) The code of t-test:

```
# Question 2 (b)

t.test(hosts$CLEC$value, hosts$ILEC$value, alt="greater",

var.equal=FALSE, conf.level=0.99) #(ii)

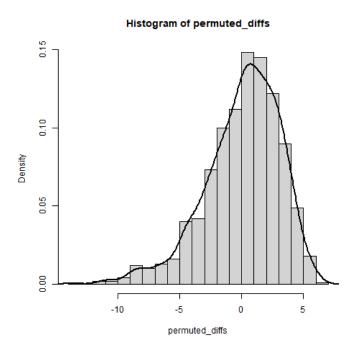
t.test(hosts$CLEC$value, hosts$ILEC$value, alt="greater",

var.equal=TRUE, conf.level=0.99) #(i)
```

- (i) Assuming variances of the two populations are equal, t = 2.6125, p-value = 0.004534. 99% CI is (0.8801387, ∞). Since t is in this interval, we cannot reject H_0 .
- (ii) Assuming variances of the two populations are not equal, t = 1.9834, df = 22.346, p-value = 0.02987. 99% CI is $(-2.130858, \infty)$ Since t is in this interval, we cannot reject H_0 .
- (c) I use the following code to run the permutation test:

(i) The distribution of permuted differences can be drawn by

```
png(filename = "2c.png")
hist(permuted_diffs, breaks = "fd", probability = TRUE)
lines(density(permuted_diffs), lwd=2)
abline(v=observed_diff, lty="dashed")
dev.off()
```



The observed difference is roughly 8.09752, and it is larger than the boundary of this graph. Hence I think it's not suitable to plot on the diagram.

(ii) These *p*-values is computed by:

```
p_1tailed <- sum(permuted_diffs > observed_diff) / nperms
p_2tailed <- sum(abs(permuted_diffs) > observed_diff) / nperms
```

It gives one-tailed p-value = 0, two-tailed p-value = 0.02

(iii) Since the one-tailed p-value = 0, the permuted difference is same as observed difference in all 1000 permutation. Hence one can claim that we cannot reject H_0 directly.

Question 3. (a) I prefer the vectorized form.

```
# Question 3 (a)
gt_eq <- function(a, b)
{
   ifelse(a > b, 1, 0) + ifelse(a == b, 0.5, 0)
}
W <- sum(outer(hosts$CLEC$value, hosts$ILEC$value, FUN = gt_eq))</pre>
```

This gives W = 26820.

(b) The one-tailed *p*-value:

```
# Question 3 (b)

n1 <- length(hosts$CLEC$value) # 23

n2 <- length(hosts$ILEC$value) # 1664

wilcox_p_1tail <- 1 - pwilcox(W, n1, n2)
```

The one-tailed p-value is 0.0003688341.

(c) By calling the built-in wilcox.test() function, say

```
# Question 3 (c)
wilcox.test(hosts$CLEC$value,
hosts$ILEC$value,
alternative = "greater",
conf.level=0.01)
```

The result shows W = 26820 and the one-tailed p-value is 0.0004565. W is same with the result of hand-made approach in (a).

(d) Since both the p-values obtained in (c) and (d) is smaller than 0.01, we cannot reject the null hypothesis.

Question 4. (a) The complete implementation of the required function is:

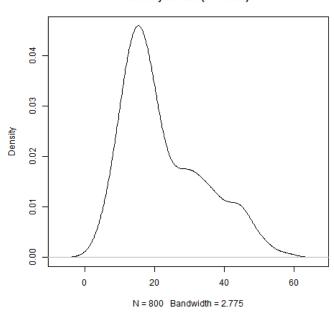
```
# Question 4 (a)
norm_qq_plot <- function(values)
{
    probs1000 <- seq(0, 1, 0.001)
    q_vals <- quantile(values, probs=probs1000)
    q_norm <- qnorm(probs1000, mean=mean(values), sd=sd(values))
    plot(q_norm, q_vals, xlab="normal quantiles", ylab="values quantiles")
    abline(a=0, b=1, col="red", lwd=2)
}</pre>
```

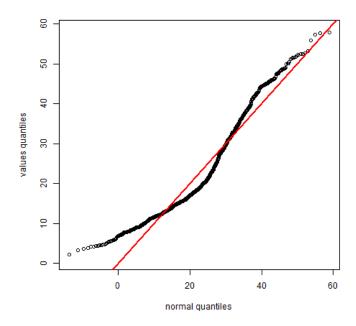
(b) The plots show that my implementation is correct.

```
# Question 4 (b)
set.seed(48763)
d1 <- rnorm(n=500, mean=15, sd=5)
d2 <- rnorm(n=200, mean=30, sd=5)
d3 <- rnorm(n=100, mean=45, sd=5)
d123 <- c(d1, d2, d3)

png(filename = "4b-1.png")
plot(density(d123))
dev.off()
png(filename = "4b-2.png")
norm_qq_plot(d123)
dev.off()</pre>
```

density.default(x = d123)

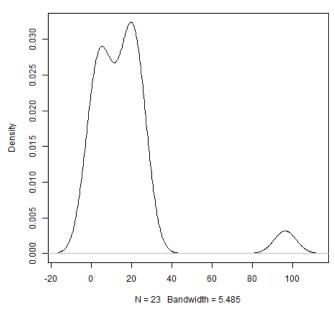


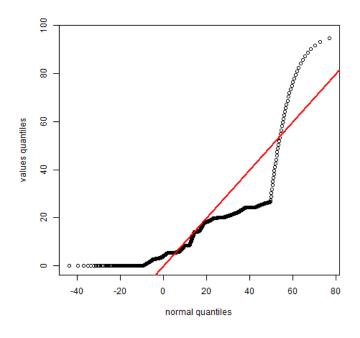


- [2] suggests that d123 is not normally distributed. It looks more likely a "bimodal" distribution.
- (c) Let's plot the density and Q-Q plot first.

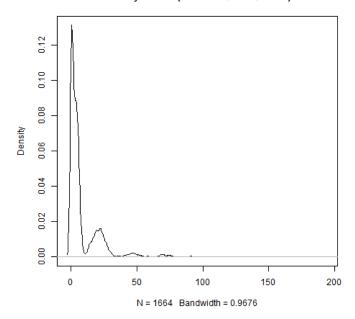
```
# Question 4 (c)
  png(filename = "4c-CLEC-density.png")
  plot(density(hosts$CLEC$value))
   dev.off()
  png(filename = "4c-CLEC-qq.png")
  norm_qq_plot(hosts$CLEC$value)
   dev.off()
  png(filename = "4c-ILEC-density.png")
10
   plot(density(hosts$ILEC$value))
11
   dev.off()
13
   png(filename = "4c-ILEC-qq.png")
14
   norm_qq_plot(hosts$ILEC$value)
15
   dev.off()
```

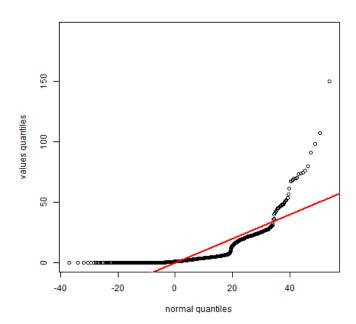
density.default(x = hosts\$CLEC\$value)





density.default(x = hosts\$ILEC\$value)





I really don't think any of CLEC or ILEC follows a normal distribution by any means, neither by the density plot nor the Q-Q plot. Perhapes ILEC follows a bimodal distribution. I guess CLEC may be approximate by a Gaussian-mixture model.

References

- [1] reshape2 (version 1.4.4), RDocumentation, https://www.rdocumentation.org/packages/reshape2/versions/1.4.4/topics/melt
- $[2]\ A\ Q-Q\ Plot\ Dissection\ Kit,\ https://seankross.com/2016/02/29/A-Q-Q-Plot-Dissection-Kit.html$