

```

library(ggplot2)
#
# Session times for websites A and B
times1 <- read.csv('web_page_data.csv')
times1[,2] <- times1[,2] * 100
str(times1)

## 'data.frame':    36 obs. of  2 variables:
## $ Page: Factor w/ 2 levels "Page A","Page B": 1 2 1 2 1 2 1 2 1 2 ...
## $ Time: num  21 253 35 71 67 85 211 246 132 149 ...

head(times1)

##      Page Time
## 1 Page A   21
## 2 Page B 253
## 3 Page A   35
## 4 Page B   71
## 5 Page A   67
## 6 Page B   85

# number of session times for each website
table(times1$Page)

##
## Page A Page B
##    21    15

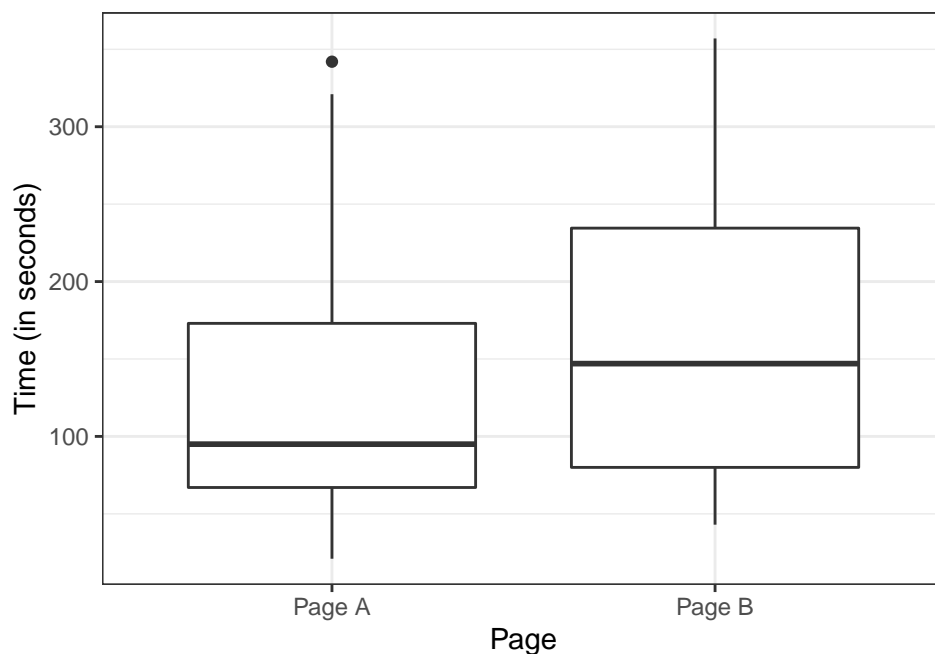
# average session time for each website
aux = tapply(times1$Time,times1$Page,mean)
aux

##    Page A    Page B
## 126.3333 162.0000

obs_diff = aux[2]-aux[1]

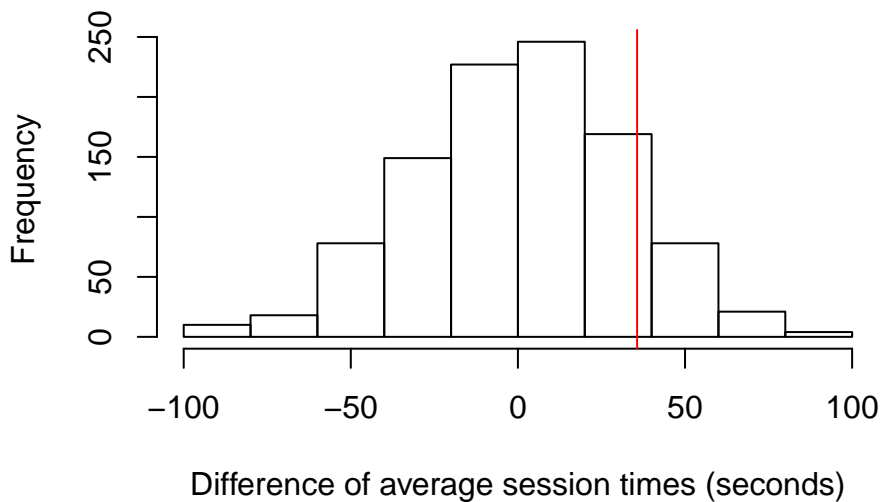
# boxplot comparing the distribution of session times
ggplot(times1, aes(x=Page, y=Time)) +
  geom_boxplot() +
  labs(y='Time (in seconds)') +
  theme_bw()

```



```
#
# how likely is this difference to be observed if average session times are equal?
#
# PERMUTATION TEST
#
# randomly group 21 and 15 times, then find difference of their average times
# repeat many times finding the distribution of differences of average times
# if obs_diff is out of the range of the distribution
# then conclude that the true difference of average session times is not zero
# and therefore one webpage results in longer session times, on average.
#
# note: setdiff(x,y) collects those elements in x but not in y
#
# x: vector of numeric values
# n1: group A
# n2: group B
#
# find difference (mean of group B values - mean of group A values)
#
function1 <- function(x, n1, n2)
{
  n <- n1 + n2
  idx_b <- sample(1:n, n1)
  idx_a <- setdiff(1:n, idx_b)
  mean_diff <- mean(x[idx_b]) - mean(x[idx_a])
  return(mean_diff)
}

# dist of 1000 differences between session times of two groups randomly chosen
set.seed(1)
differences <- rep(0, 1000)
times = times1$Time
for(i in 1:1000) differences[i] = function1(times, 21, 15)
hist(differences, xlab='Difference of average session times (seconds)', main='')
abline(v = obs_diff, col='red')
```



```
# proportion of TRUE in a sequence of logical obs
aux = c(TRUE,TRUE,TRUE,FALSE,TRUE,TRUE)
aux

## [1] TRUE TRUE TRUE FALSE TRUE TRUE
sum(aux)

## [1] 5
mean(aux)

## [1] 0.8333333
# proportion of session times beyond redline (p-value approximation)
mean(differences > obs_diff)

## [1] 0.138
#
# t-test (p-value)
t.test(Time ~ Page, data=times1, alternative='less' )

##
## Welch Two Sample t-test
##
## data: Time by Page
## t = -1.0983, df = 27.693, p-value = 0.1408
## alternative hypothesis: true difference in means is less than 0
## 95 percent confidence interval:
##      -Inf 19.59674
## sample estimates:
## mean in group Page A mean in group Page B
##      126.3333      162.0000
#
# p-value agrees with that of permutation test
#
test1 = t.test(Time ~ Page, data=times1, alternative='less' )
str(test1)

## List of 10
## $ statistic : Named num -1.1
## ..- attr(*, "names")= chr "t"
## $ parameter : Named num 27.7
```

```
## ..- attr(*, "names")= chr "df"
## $ p.value : num 0.141
## $ conf.int : num [1:2] -Inf 19.6
## ..- attr(*, "conf.level")= num 0.95
## $ estimate : Named num [1:2] 126 162
## ..- attr(*, "names")= chr [1:2] "mean in group Page A" "mean in group Page B"
## $ null.value : Named num 0
## ..- attr(*, "names")= chr "difference in means"
## $ stderr : num 32.5
## $ alternative: chr "less"
## $ method : chr "Welch Two Sample t-test"
## $ data.name : chr "Time by Page"
## - attr(*, "class")= chr "htest"
```

```
test1$p.value
```

```
## [1] 0.1407622
```

```
#
#
# PROPORTIONS
#
# p1, p2 population proportion of bath soap sales with designs 1 and 2
# First design is better if p1 > p2
```

```
d0 = read.csv("Xm13-09.csv",header=T)
head(d0)
```

```
## Supermarket1 Supermarket2
## 1 4255 4255
## 2 4255 9077
## 3 4255 8855
## 4 8855 9077
## 5 7118 8855
## 6 9077 4255
```

```
#
# table shows scanner codes for different brands
# focus on 9077
tail(d0)
```

```
## Supermarket1 Supermarket2
## 1033 NA 3745
## 1034 NA 8855
## 1035 NA 3745
## 1036 NA 8855
## 1037 NA 9077
## 1038 NA 8855
```

```
# number of sales in the Supermarkets is different
```

```
# rename cols
names(d0) = c("s1","s2")
dim(d0)
```

```
## [1] 1038 2
```

```
# soap brands purchased at Supermarket 1
table(d0$s1)
```

```
##
```

```
## 3745 4255 7118 8855 9077
## 99 218 163 244 180
# 180 bath soaps sold in Supermarket 1
#
# soap brands purchased at Supermarket 2
table(d0$s2)

##
## 3745 4255 7118 8855 9077
## 134 228 218 303 155
# 155 bath soaps sold in Supermarket 2
#
x = c(180,155)

# number of purchases at each Supermarket
sum(table(d0$s1))

## [1] 904
sum(table(d0$s2))

## [1] 1038
# number of soaps from all brands sold in 1-week
n1 = sum(table(d0$s1))
n2 = sum(table(d0$s2))
n = c(n1,n2)

# number of bath soaps sold from other brands
others = n-x

# table of counts
d2 = rbind(x,others,n)
rownames(d2) = c("brand","others","total")
colnames(d2) = c("Super1","Super2")
d2

##          Super1 Super2
## brand        180    155
## others       724    883
## total        904   1038

# sample proportions
p1 = 180/904
p1

## [1] 0.199115
p2 = 155/1038
p2

## [1] 0.1493256
obs_diff = p1 - p2
obs_diff

## [1] 0.04978942
#
# Is this difference due to different designs or to random chance?
```

```
# ignore total sales row
```

```
d1 = d2[1:2,]
```

```
d1
```

```
##           Super1 Super2
```

```
## brand       180    155
```

```
## others      724    883
```

```
# sum by row
```

```
apply(d1,1,sum)
```

```
## brand others
```

```
##    335    1607
```

```
# total soaps sold
```

```
sum(apply(d1,1,sum))
```

```
## [1] 1942
```

```
#
```

```
# create a vector of 335 ones, 1607 zeros
```

```
vector1 <- c(rep(0, 1607), rep(1, 335))
```

```
#
```

```
# one to represent a transaction from our brand
```

```
# zero for a transaction from other brand
```

```
#
```

```
# total number of sales by SuperMarket
```

```
d2[3,]
```

```
## Super1 Super2
```

```
##    904    1038
```

```
#
```

```
# PERMUTATION TEST
```

```
#
```

```
# For Supermarket 1, select sample of 904 sales,
```

```
# record number of ones in this sample
```

```
# find proportion of ones in this sample
```

```
# For Supermarket 2,
```

```
# count number of ones in the remaining 1038 rows
```

```
# find proportion of ones in these remaining rows
```

```
#
```

```
# find difference between the 2 proportions
```

```
# repeat 1000 times
```

```
#
```

```
differences <- rep(0, 1000)
```

```
#
```

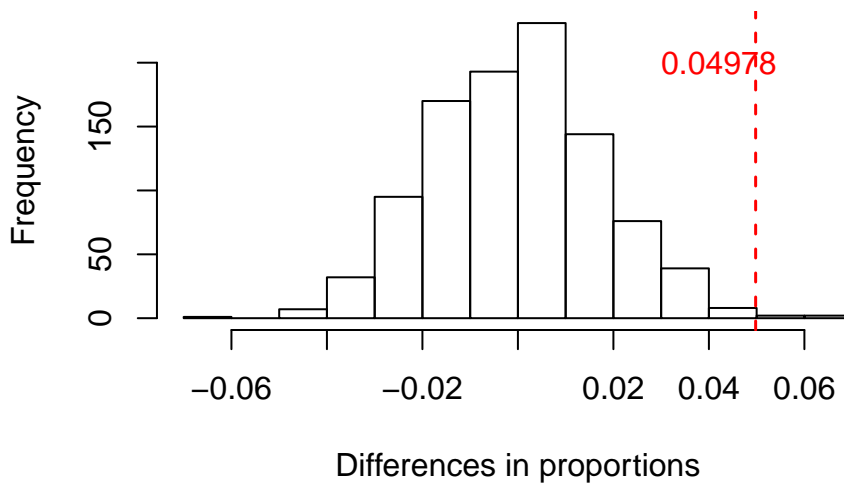
```
set.seed(1)
```

```
for(i in 1:1000) differences[i] = function1(vector1,904,1038)
```

```
hist(differences, xlab='Differences in proportions', main='')
```

```
abline(v = obs_diff, lty=2, lwd=1.5,col='red')
```

```
text("0.04978", x=0.03,y=200,adj=0,col='red')
```



```
# fraction beyond redline (approx p-value)
mean(differences > obs_diff)
```

```
## [1] 0.004
```

```
#
# test on two (or more) proportions
#
x
```

```
## [1] 180 155
```

```
n
```

```
## [1] 904 1038
```

```
prop.test(x,n, alternative="greater")
```

```
##
## 2-sample test for equality of proportions with continuity correction
##
## data: x out of n
## X-squared = 8.0461, df = 1, p-value = 0.00228
## alternative hypothesis: greater
## 95 percent confidence interval:
## 0.02032291 1.00000000
## sample estimates:
## prop 1 prop 2
## 0.1991150 0.1493256
```

```
#
# p-values from random sampling and test on proportions agree
#
# p-value is smaller than alpha
# reject Ho: p1 = p2
# conclude p1 > p2
# Company should use design 1
#
# get p-value alone
#
prop.test(x,n, alternative="greater")$p.value
```

```
## [1] 0.002280073
```