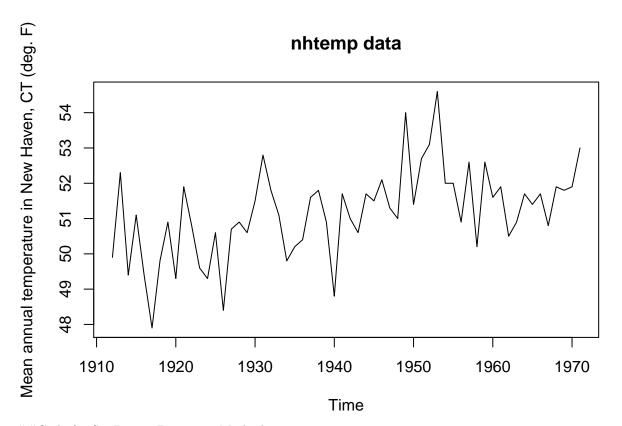
## Bootstrap & Jackknife

Sahil Rangwala 12/13/2019

## R Markdown

```
##Code for Bootstrap Method
#Data is the mean annual temperature in degrees Fahrenheit in New Haven, Connecticut, from 1912 to 1971
#The data is a test data that is offered within Rstudio, hard coded the data due to technical difficult
#Calculating the probability of if the chosen mean annual temperature is less than the median of the da
#Should be around 50% probability, chose the median due to the fact that we know what the output should
#This would help hone in on making sure the code is correct for both the SimDesign and non SimDesign me
     \mathtt{data} = \mathtt{c}(49.9, 52.3, 49.4, 51.1, 49.4, 47.9, 49.8, 50.9, 49.3, 51.9, 50.8, 49.6, 49.3, 50.6, 48.4, 50.7, 50.9, 50.6, 51.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 50.2, 5
    print(median(data))
## [1] 51.2
     thetahat=length(data[data < 51.2]) / length(data) # calculate the theta hat
     bootRet = numeric(B) # vector that houses the bootstrap results
     #Resampling portion
     for(b in 1:B) {
         sampleData = sample(data,length(data),replace=TRUE)
          #Seeing what samples are less than the median
         bootRet[b] = length(sampleData[sampleData < 51.2]) / length(sampleData)</pre>
     }
     #results of the bootstrap method
     mean_bootstrap=mean(bootRet)
     se_bootstrap=sd(bootRet)
     bias_bootstrap=mean(bootRet - thetahat)
print(mean_bootstrap)
## [1] 0.5007833
print(se_bootstrap)
## [1] 0.06670584
print(bias_bootstrap)
## [1] 0.0007833333
#The data chosen
plot(nhtemp, main = "nhtemp data",
```

ylab = "Mean annual temperature in New Haven, CT (deg. F)")



 $\#\#\mathrm{Code}$  for SimDesign Bootstrap Method

```
library(SimDesign)
\mathtt{data} = \mathtt{c}(49.9, 52.3, 49.4, 51.1, 49.4, 47.9, 49.8, 50.9, 49.3, 51.9, 50.8, 49.6, 49.3, 50.6, 48.4, 50.7, 50.9, 50.6, 51.5, 50.9, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 50.8, 5
thetaHat=length(data[data < 51.2]) / length(data) #Same method to calculate thetahat
Design <- expand.grid(condition = c(1)) #Sample chosen is already intialized
Generate <- function(condition, fixed_objects = NULL) {</pre>
             dat <- data
             return (dat)
}
   Analyse <- function(condition, dat, fixed_objects = NULL) {</pre>
             sampleChosen <- sample(dat,length(dat),replace=TRUE)</pre>
            ret <- c(xmean = length(sampleChosen[sampleChosen < 51.2]) / length(sampleChosen)) # calculating th
            return (ret)
}
Summarise <- function(condition, results, fixed_objects = NULL) {</pre>
            ret <- c(mean_Boot = mean(results[,1]), standardErrBoot = sd(results[,1]), biasBoot = mean(results[</pre>
             return (ret)
}
simBootRes <- runSimulation(design=Design, replications=1000, generate=Generate,
                                                                               analyse=Analyse, summarise=Summarise) # replicating the simulation 1000
```

```
##
##
Design row: 1/1;
                  Started: Fri Dec 13 17:08:35 2019;
                                                         Total elapsed time: 0.00s
## Simulation complete. Total execution time: 0.34s
print(simBootRes) # the results of the SimDesign implementation
     condition mean_Boot standardErrBoot biasBoot REPLICATIONS SIM_TIME
## 1
                 0.50075
                               0.06322659 0.00075
                                                                      0.34s
                                                             1000
                     COMPLETED
##
## 1 Fri Dec 13 17:08:35 2019 1452142966
data = c(49.9,52.3,49.4,51.1,49.4,47.9,49.8,50.9,49.3,51.9,50.8,49.6,49.3,50.6,48.4,50.7
Used http://philchalmers.github.io/SimDesign/html/11-Parametric-bootstrap.html to understand general
structure
##Code for Jackknife Method
\mathtt{data} = \mathtt{c}(49.9, 52.3, 49.4, 51.1, 49.4, 47.9, 49.8, 50.9, 49.3, 51.9, 50.8, 49.6, 49.3, 50.6, 48.4, 50.7, 50.9, 50.6, 51.5
 thetaHat = length(data[data<51.2])/length(data) # calculation for theta hat
 n = length(data)
  jackVal = numeric(n) # vector to save the results
  for (i in 1:n) {
    leftOut = data[-i] # leaving out the one specific data sample
    #Calculate the jackknife estimate for the specific sample
    jackVal[i] = length(leftOut [leftOut < 51.2]) / length(leftOut)</pre>
  #results of the Jackknife method
  meanJack = mean(jackVal)
  sumsq = sum((jackVal - mean(jackVal))^2)
  standardErrorJack=sqrt((n - 1) / n) * sqrt(sumsq)
  biasJack=(n - 1) * (mean(jackVal) - thetaHat)
  #return in a form of vector
print(meanJack)
## [1] 0.5
print(standardErrorJack)
## [1] 0.06509446
print(biasJack)
## [1] 0
##Code for SimDesign Jackknife Method
library(SimDesign)
\mathtt{data} = \mathtt{c}(49.9, 52.3, 49.4, 51.1, 49.4, 47.9, 49.8, 50.9, 49.3, 51.9, 50.8, 49.6, 49.3, 50.6, 48.4, 50.7, 50.9, 50.6, 51.5
```

```
Design <- expand.grid(condition = c(1)) # don't need design because the samples are initialized
Generate <- function(condition, fixed_objects = NULL) {</pre>
    dat <- data
    dat
}
Analyse <- function(condition, dat, fixed objects = NULL) {
   thetahat = length(dat[dat < 51.2]) / length(dat) # calculate the theta hat
   n = length(data)
   jackVal = numeric(n) # vector to save the results
   for (i in 1:n) {
      leftOut = data[-i] # leaving out the one specific data sample
      #Calculate the jackknife estimate for the specific sample
      jackVal[i] = length(leftOut [leftOut < 51.2]) / length(leftOut)</pre>
    #results of the Jackknife method
   meanJack = mean(jackVal)
   sumsq = sum((jackVal - mean(jackVal))^2)
    standardErrorJack=sqrt((n - 1) / n) * sqrt(sumsq)
   biasJack=(n - 1) * (mean(jackVal) - thetaHat)
   ret <- c(meanJackknife = meanJack, standardErrorJackknife = standardErrorJack, BiasJackknife = bias
   return(ret)
}
Summarise <- function(condition, results, fixed_objects = NULL) {</pre>
   ret <- c(meanJackknife = mean(results[,1]),</pre>
             standardErrorJackknife = mean(results[,2]),
             BiasJackknife = mean(results[,3]))
   ret
}
SimdDesignResults <- runSimulation(design=Design, replications=1000, generate=Generate,
                         analyse=Analyse, summarise=Summarise)
##
##
Design row: 1/1;
                   Started: Fri Dec 13 17:08:35 2019;
                                                         Total elapsed time: 0.00s
##
## Simulation complete. Total execution time: 0.92s
print(SimdDesignResults)
     condition meanJackknife standardErrorJackknife BiasJackknife
## 1
                         0.5
                                          0.06509446
                                                                 0
             1
##
    REPLICATIONS SIM TIME
                                           COMPLETED
## 1
             1000
                     0.92s Fri Dec 13 17:08:36 2019 2048925918
```

```
##Importance Sampling
#initialize variables
m <- 10000
theta.hat <- se <- numeric(5)
#create function to represent equation
g <- function(x) {</pre>
exp(-x - log(1+x^2)) * (x > 0) * (x < 1)
}
#generate random uniform variables
u <- runif(m)
\#Find\ importance\ function\ and\ calculate\ x
x \leftarrow -\log(1 - u * (1 - \exp(-1)))
#Find Ratio of functions
fg \leftarrow g(x) / (exp(-x) / (1 - exp(-1)))
#take mean of result
theta.hat <- mean(fg)
#calculate standard error
se \leftarrow sd(fg)
theta.hat
## [1] 0.5246061
se
## [1] 0.09718451
##SimDesign Importance Sampling
#Set Sample Size parameter for simulation design
Design <- data.frame(sample_size = c(100, 1000, 10000))</pre>
#Function used to generate data from uniform distribution
Generate <- function(condition, fixed_objects = NULL) {</pre>
    N <- condition$sample_size</pre>
    dat <- runif(N)</pre>
    dat
}
#Calculates Estimates
Analyse <- function(condition, dat, fixed_objects = NULL) {</pre>
    u <- dat
    \#Find\ importance\ function\ and\ calculate\ x
    x \leftarrow -\log(1 - u * (1 - \exp(-1)))
    #Find Ratio of functions
    fg \leftarrow g(x) / (exp(-x) / (1 - exp(-1)))
    #take mean of result
    theta.hat <- mean(fg)
    se \leftarrow sd(fg)
    ret <- c(theta.hat = theta.hat,standardError=se)</pre>
}
#Caculates mean of estimates and standard error
Summarise <- function(condition, results, fixed_objects = NULL) {</pre>
    se=sd(results)
    ret <- c(theta.hat=mean((results[,1])), standardError = mean(results[,2]))</pre>
```

```
ret
}
#Simulation is run 1000 times
results_sd <- runSimulation(design=Design, replications=1000, generate=Generate,
                         analyse=Analyse, summarise=Summarise)
##
##
Design row: 1/3; Started: Fri Dec 13 17:08:36 2019;
                                                         Total elapsed time: 0.00s
##
Design row: 2/3;
                   Started: Fri Dec 13 17:08:36 2019;
                                                         Total elapsed time: 0.31s
##
##
                   Started: Fri Dec 13 17:08:37 2019;
                                                         Total elapsed time: 0.81s
Design row: 3/3;
##
## Simulation complete. Total execution time: 2.85s
results_sd
##
     sample_size theta.hat standardError REPLICATIONS SIM_TIME
## 1
            100 0.5250158
                              0.09660182
                                                  1000
                                                          0.31s
## 2
            1000 0.5247865
                              0.09689087
                                                  1000
                                                          0.51s
## 3
           10000 0.5248178
                              0.09685263
                                                  1000
                                                          2.04s
                    COMPLETED
                                    SEED
## 1 Fri Dec 13 17:08:36 2019 113459029
## 2 Fri Dec 13 17:08:37 2019 1587826100
## 3 Fri Dec 13 17:08:39 2019 346825591
##Monte Carlo Integration
#function to generate monte carlo estimates
pimc=function(a,b){
 m <- 10000
  #generate uniform random variables
  u <- runif(m,a,b)
  #function from integral to be estimated
  g=4*sqrt(1-u^2)
  #find estimate using mean function and taking into account integral limits
  theta.hat=mean(g)*(b-a)
  se=sd(g)
  return(c(theta.hat=theta.hat,sd=se))
 }
a=c(0,0,0)
b=c(1,0.5,.25)
theta.hat=numeric(3)
se=numeric(3)
for(i in 1:length(a)){
  print(pimc(a[i],b[i]))
}
## theta.hat
## 3.1451549 0.8964739
## theta.hat
## 1.9137549 0.1579013
```

```
## theta.hat
## 0.98956111 0.03764206
##SimDesign Monte Carlo Integration
#Set Parameters for simulation design
Design <- data.frame(sample_size = c(10000, 10000, 10000),
                      a = c(0, 0, 0),
                      b = c(1, 0.5, 0.25))
#Function used to generate data from uniform distribution with given parameters
Generate <- function(condition, fixed_objects = NULL) {</pre>
    N <- condition$sample_size</pre>
    11 <- condition$a</pre>
    ul <- condition$b
    dat <- runif(N, 11, u1)</pre>
    dat
}
#Calculates Estimates
Analyse <- function(condition, dat, fixed_objects = NULL) {</pre>
    u <- dat
    #function from integral to be estimated
    g=4*sqrt(1-u^2)
    #find estimate using mean function and taking into account integral limits
    theta.hat <- mean(g) * (condition$b-condition$a)</pre>
    ret <- c(theta_hat = theta.hat,se=sd(g))</pre>
    ret
}
#Caculates mean of estimates and standard error
Summarise <- function(condition, results, fixed_objects = NULL) {</pre>
    se = mean(results[,2])
    ret <- c(theta.hat = mean(results[,1]),standardError= se)</pre>
    ret
}
#Simulation is run 1000 times
results_sd <- runSimulation(design=Design, replications=1000, generate=Generate,
                          analyse=Analyse, summarise=Summarise)
##
##
Design row: 1/3;
                   Started: Fri Dec 13 17:08:39 2019;
                                                          Total elapsed time: 0.00s
##
##
Design row: 2/3;
                   Started: Fri Dec 13 17:08:40 2019;
                                                          Total elapsed time: 1.20s
##
##
Design row: 3/3;
                   Started: Fri Dec 13 17:08:41 2019;
                                                          Total elapsed time: 2.27s
## Simulation complete. Total execution time: 3.19s
results_sd
```

b theta.hat standardError REPLICATIONS SIM\_TIME

##

sample\_size a

##	1	10000 0	1 3.1416441	0.89273105	1000	1.20s
##	2	10000 0 0.	5 1.9132146	0.15803139	1000	1.08s
##	3	10000 0 0.2	5 0.9894844	0.03777168	1000	0.91s
##		CO	MPLETED	SEED		
##	1 Fri	Dec 13 17:08:	40 2019 5236	686787		
##	2 Fri	Dec 13 17:08:	41 2019 394	522631		
##	3 Fri	Dec 13 17:08:	42 2019 1666	278706		