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Assignment 6
10/31/2019

```
hurricane.mat <- read.csv('C:\\Users\\mgleyzer\\Downloads\\Hurricane cat 4 and 5  
corrected.csv' )
```

```
# delete the NA values in the first two rows
```

```
hurricane.mat1 <- hurricane.mat[-c(1,2),]
```

```
# function takes the hurricane.mat1, fits a linear model,
```

```
And makes a prediction based on given bootstrapped values of the statistic
```

```
lmfunc <- function(vec0,mat=hurricane.mat1){
```

```
  # Coerce hurricane.mat1 into a data frame
```

```
  mat00<-as.data.frame(hurricane.mat1)
```

```
  # Make a data frame to be used for the linear model fitting
```

```
  # The rows of the matrix to be used in the linear fit
```

```
# Determined by the bootstrapped indices
```

```
  mat0<-as.data.frame(hurricane.mat1[vec0,])
```

```
  # Fit a linear model with predictors(AMO , decade)
```

```
  # and response variable as cat45, using the mat0 data
```

```
  lm.str<-lm(cat45~AMO+decade,data=mat0)
```

```
  # Get the 13th row which corresponds to the 1990 decade
```

```
  predict.mat<-mat00[c(13,13) , ]
```

```
  # Predict new response values using fitted linear model and prediction
```

```
  # matrix
```

```
  out<-predict(lm.str,predict.mat)
```

```
# Return the column with the predicted response variable
```

```
  out[1]}
```

```
Jackknife<-function(vec0,statfunc=sd){
```

```
  # Assign the length of vec0 to n1
```

```
  n1<-length(vec0)
```

```
  # Initialize the variable jackvec
```

```
  jackvec<-NULL
```

```
  # Calculate the initial standard deviation of vec0
```

```
  mu0<-statfunc(vec0)
```

```
  # Iterate the length of vec0
```

```
  for(i in 1:n1){
```

```
    # Find the standard deviation of
```

```
    # vec0, when omitting the ith element
```

```
    mua<-statfunc(vec0[-i])
```

```
    # Create vector jackvec composed of jackknifed elements.
```

```
    # Those elements represent the bootstrapped values' individual effects on standard  
deviation
```

```

jackvec<-c(jackvec, n1*(mu0)-(n1-1)*mua)}
# Find the bias of each jackknife estimate compared to the original
# estimate
jackmean<- mean(jackvec)
jackbias<-mean(jackvec)-mu0
# Find the standard deviation of the jackknifed effect estimates
jacksd<-sd(jackvec)
# List out all the results
list(mu0=mu0,jackbias=jackbias,jacksd=jacksd)}

```

```

My.bootstrap.ci<-function(data.str,nboot=10000,alpha=0.05, stat.func = lmfunc){
#Assign length of the data(no. of rows) to n0
n0<-length(data.str[,1])
# Make vec0 the vector holding the indices of the rows
vec0<-c(1:n0)
#Assign to mean0 the value returned by the lmfunc function
mean0<-lmfunc(vec0)
#Find the standard deviation of the data using the jackknife function. Assign it to
variable
#sd0
sd0<-Jackknife(vec0 , statfunc = stat.func) $ jacksd
#Initialize the bootvec object
bootvec<-NULL
#Iterate 10000 times
for( i in 1:nboot){
#Create bootstrap vector of data
#By sampling with replacement from given vector v0
vecb<-sample(vec0,replace=T)
#Apply the linear model function to the bootstrap vector
meanb<-stat.func(vecb)
#Find the standard deviation of the bootstrap vector
sdb<-Jackknife(vecb , statfunc = stat.func) $ jacksd
#Bootstrap vector of test statistics(to be used for confidence intervals)
bootvec<-c(bootvec,(meanb-mean0)/(sdb/sqrt(n0))))}
#Determine lower quantile of bootvec test statistic data
lq<-quantile(bootvec,alpha/2)
#Determine upper quantile of bootvec test statistic data
uq<-quantile(bootvec,1-alpha/2)
#Find lower bound for normal distribution confidence interval
LB<-mean0-(sd0/sqrt(n0))*uq
#Find upper bound for normal distribution confidence interval
UB<-mean0-(sd0/sqrt(n0))*lq

```

```

#Find bootstrapped lower bound for confidence interval
NLB<-(mean0-(sd0/sqrt(n0))*qt(1-alpha/2,n0-1)
#Find bootstrapped upper bound for confidence interval
NUB<-(mean0+(sd0/sqrt(n0))*qt(1-alpha/2,n0-1 )
#Store both confidence intervals(normal theory, bootstrapped) in a list
list(bootstrap.confidence.interval=c(LB,UB),normal.confidence.interval=c(NLB,NUB))}

```

```
My.bootstrap.ci(hurricane.mat1 , nboot= 10000, alpha = 0.05, stat.func = lmfunc)
```

Results:

```
$bootstrap.confidence.interval
```

```

      15      15
14.13991 18.38291

```

```
$normal.confidence.interval
```

```

      15      15
13.75571 18.10972

```