# Bootstrapping in R – A Tutorial

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# Bootstrapping

- Resampling technique with replacement
  - "The population is to the sample as the sample is to the bootstrap samples"
- Allows estimation of the sampling distribution of a statistic
  - Confidence intervals, bias, variance, etc.





## Procedure

- Resample a dataset a given number of times
- Calculate a statistic from each sample
- Accumulate the results and calculate sample distribution of the statistic





## Objective

- Calculate a series of linear regressions to determine which variable or combination of variables best explains the volume of black cherry trees
  - Comparisons made using coefficient of determination (R-squared)
- Bootstrap the linear regressions (for each bootstrap sample) to determine 95% confidence intervals of their respective Rsquared values





## Data

- "trees" dataset (included in R)
- Volume (cubic feet), girth (diameter in inches, measured at breast height), and height (feet) measurements of 31 felled black cherry trees

help(trees)





# **Code Walkthrough**





- Load the boot library
  - Contains functions to conduct bootstrapping

```
library(boot)
```

Investigate the "trees" dataset

```
head(trees)
```

```
Girth Height Volume
  8.3
          70
               10.3
  8.6
          65
               10.3
8.8
               10.2
          63
10.5
          72
               16.4
10.7
               18.8
          81
10.8
          83
               19.7
```



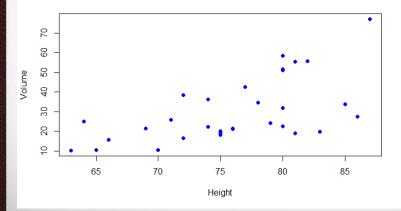


## Explore relationships between volume, girth, and height

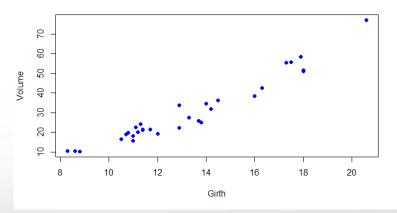
```
plot(trees$Volume~trees$Height, main = 'Black Cherry Tree Volume
Relationship', xlab = 'Height', ylab = 'Volume', pch = 16, col =
'blue')
```

```
plot(trees$Volume~trees$Girth, main = 'Black Cherry Tree Volume
Relationship', xlab = 'Girth', ylab = 'Volume', pch = 16, col =
'blue')
```

### **Black Cherry Tree Volume Relationship**



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- Create a function that will calculate a statistic (or multiple statistics) on each bootstrap sample
- Function syntax in R

```
foo = function(parameter, parameter, marameter,) {
  bar = *do something to data passed as parameters*
  return(bar)
}
```





 Statistic-calculation function for the boot package takes two specific parameters (simple example) and will be applied to each bootstrap sample

```
sample_mean = function(data, indices){
  sample = data[indices, ]
  bar = mean(sample)
  return(bar)
}
```

Calculate the mean of the bootstrap sample

Creates the bootstrap sample (i.e., subset the provided data by the "indices" parameter). "indices" is automatically provided by the "boot" function; this is the sampling with replacement portion of bootstrapping

### Or, more concisely:

```
sample_mean = function(data, indices) {
  return(mean(data[indices]))
}
```



- Create a function to calculate linear regressions of several variable combinations and return their respective R-squared values
  - Height only,
  - Girth only
  - Girth / height ratio
  - Girth and height
  - Girth, height, and girth / height ratio
- Note that we are calculating (and returning) multiple statistics simultaneously

These statistics will be calculated for each bootstrap sample

```
volume_estimate = function(data, indices) {
    d = data[indices, ]
    H_relationship = lm(d$Volume~d$Height, data = d)
    H_r_sq = summary(H_relationship)$r.square
    G_relationship = lm(d$Volume~d$Girth, data = d)
    G_r_sq = summary(G_relationship)$r.square
    G_H_ratio = d$Girth / d$Height
    G_H_relationship = lm(d$Volume~G_H_ratio, data = d)
    G_H_r_sq = summary(G_H_relationship)$r.square
    combined_relationship = lm(d$Volume~d$Height + d$Girth, data = d)
    combined_r_sq = summary(combined_relationship)$r.square
    combined_2_relationship = lm(d$Volume~d$Height + d$Girth + G_H_ratio, data = d)
    combined_2_r_sq = summary(combined_2_relationship)$r.square
    relationships = c(H_r_sq, G_r_sq, G_H_r_sq, combined_r_sq, combined_2_r_sq)
    return(relationships)
}
```





## Conduct the bootstrapping

- Use "boot" function

results = boot(data = trees, statistic = volume\_estimate, R = 5000)

Dataset from which statistics will be calculated

Function we created to calculate statistics on each bootstrap sample

Number of bootstrap samples (i.e., iterations)





## View some calculated statistics of boot object

print(results)

#### ORDINARY NONPARAMETRIC BOOTSTRAP

```
call:
boot(data = trees, statistic = volume_estimate, R = 5000)
```

Bootstrap Statistics :

original bias std. error t1\* 0.3579026 0.0024051943 0.12025420 t2\* 0.9353199 0.0005495767 0.01751679 t3\* 0.7309204 0.0025156062 0.08064029 t4\* 0.9479500 0.0032851681 0.01210484 t5\* 0.9732894 0.0005447157 0.01042662

t\* corresponds to index of "relationships" vector (e.g., t1\* refers to height only R-squared value





## Plot the boot objects

## - Provides histogram and Q-Q plot

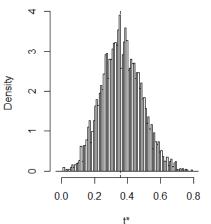
$$plot(results, index = 1)$$

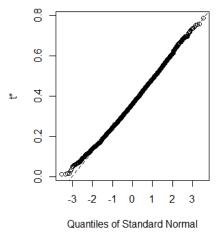
The index parameter corresponds to the indices of the vector ("relationships") returned by the "volume\_estimation" function (e.g., index 1 is the first item in the vector, which is the height only R-squared value)

Height only R-squared distribution:











- Calculate 95% confidence intervals for each of the bootstrapped R-squared values
  - Using "Bias Corrected and Accelerated" (BCa) method

Specify index corresponding to position in vector for each statistic

```
confidence_interval_H = boot.ci(results, index = 1, conf = 0.95, type = 'bca')
print(confidence_interval_H)
ci_H = confidence_interval_H$bca[ , c(4, 5)]
print(ci_H)
```

Store confidence intervals in a variable in order to plot later

```
BOOTSTRAP CONFIDENCE INTERVAL CALCULATIONS
Based on 5000 bootstrap replicates

CALL:
boot.ci(boot.out = results, conf = 0.95, type = "bca", index = 1)

Intervals:
Level BCa
95% ( 0.1415,  0.6123 )
Calculations and Intervals on Original Scale
> ci_H = confidence_interval_H$bca[ , c(4, 5)]
> print(ci_H)

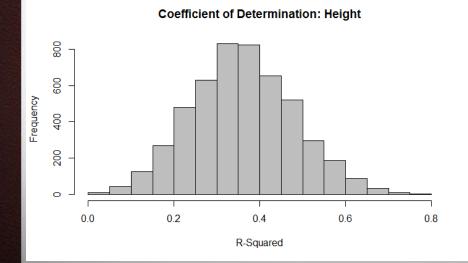
0.1414861 0.6122950
```

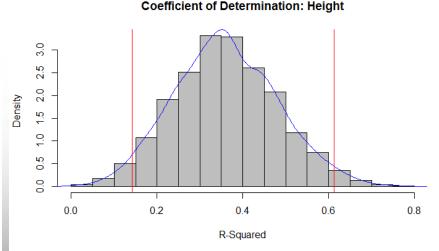


- View histograms (frequency and density)
- Add kernel density line (blue)
- Add 95% confidence intervals (red)

```
hist(results$t[,1], main = 'Coefficient of Determination: Height', xlab = 'R-
    Squared', col = 'grey')
hist(results$t[,1], main = 'Coefficient of Determination: Height', xlab = 'R-
    Squared', col = 'grey', prob = T)
lines(density(results$t[,1]), col = 'blue')
abline(v = ci_H, col = 'red')
```

Note syntax to call desired sample distribution





## Can also call the entire sample distribution to further manipulate, save, etc.

results\$t[, 1]————

Access the sample statistics of each bootstrap sample

R-squared values of height only linear regression:

Subset to particular statistic; first column of the boot object "t" corresponds to the first item in the vector returned by the "volume\_esitmate" function

```
> results$t[,1]
```

```
[1] 0.207990443 0.363816239 0.579971818 0.423443272 0.336572704 0.417656521 0.251820295 0.343777274 [9] 0.270477273 0.480302587 0.564330760 0.474092665 0.174531538 0.300817972 0.502245182 0.359519760 [17] 0.367795668 0.435299147 0.243218209 0.180413913 0.428146329 0.568726861 0.399806911 0.195195281 [25] 0.255877036 0.416366115 0.315921685 0.541198595 0.272757355 0.628962441 0.350397269 0.192770891 [33] 0.266364939 0.310743438 0.613576574 0.696147632 0.488130237 0.388040468 0.344063541 0.399933017 [41] 0.255363943 0.395594597 0.318028661 0.391665068 0.356077907 0.188440159 0.421280357 0.072206043 [49] 0.449664202 0.462657862 0.413759773 0.446951604 0.369800075 0.468153637 0.182068140 0.375718017 [57] 0.151727603 0.237096695 0.293074927 0.476329686 0.308111480 0.218648993 0.265019573 0.204667380 [65] 0.651896672 0.639127085 0.478180644 0.315661237 0.630257581 0.426617868 0.352848563 0.333865284
```





## Results

- Linear regression with explanatory variables of girth, height, and girth / height ratio provided best coefficients of determination to model the volume of black cherry trees
- 5,000 sample bootstrap allowed estimation of R-squared sampling distribution
  - Could have also bootstrapped values of coefficients, additional models, etc.

Estimating Black Cherry Tree Volume - Linear Regression Coefficients of Determination				
	Original Value	Bias	Std. Error	95% Confidence Interval
Height Only	0.3579026	0.002405194	0.1202542	0.1414861 - 0.6122950
Girth Only	0.9353199	0.000549577	0.01751679	0.8770796 - 0.9582597
Girth / Height	0.7309204	0.002515606	0.08064029	0.4782823 - 0.8421099
Girth and Height	0.94795	0.003285168	0.01210484	0.9052392 - 0.9647783
Girth, Height, and Girth / Height	0.9732894	0.000544716	0.01042662	0.9418756 - 0.9868528





### References

http://www.statmethods.net/advstats/bootstrapping.html
http://www.mayin.org/ajayshah/KB/R/documents/boot.html
http://www.r-bloggers.com/bootstrap-example/
http://cran.r-project.org/web/packages/boot/boot.pdf

