Bootstrap with R

```
> # boot1.R
               Working on the bootstrap
                R --vanilla < boot1.R > boot1.out &
> # Run with
> # grades.dat has 4 columns: ID, Verbal SAT, Math SAT and 1st year GPA
>
> marks <- read.table("grades.dat")</pre>
> n <- length(marks$verbal)</pre>
> n
[1] 200
> marks[1:10,]
   verbal math gpa
      623 509 2.6
1
      454 471 2.3
2
      643 700 2.4
3
4
      585 719 3.0
5
      719 710 3.1
6
      693 643 2.9
7
      571 665 3.1
      646 719 3.3
8
           693 2.3
9
      613
           701 3.3
10
      655
> obscorr <- cor(marks)</pre>
> obscorr
          verbal
                       math
                                  gpa
verbal 1.0000000 0.2746341 0.3224477
math
       0.2746341 1.0000000 0.1942431
       0.3224477 0.1942431 1.0000000
gpa
> # Question: Is the correlation between Verbal SAT and GPA the same as
> # the correlation between math SAT and GPA?
> # What is the sampling distribution of the difference between correlation
> # coefficients?
> #
> obsdiff <- obscorr[3,1]-obscorr[3,2] # Verbal minus math</pre>
> obsdiff
[1] 0.1282046
> # The strategy will be to obtain a 95% bootstrap confidence interval for
> # the difference between verbal correlation and math correlation. This
> # confidence interval will be approximately centered around the observed
> # difference obsdiff = .128. If the confidence interval does not include
> # zero, we will conclude that the observed difference is significantly
> # different from zero.
>
> BOOT <- 1000 ; bsdiff <- NULL ; set.seed(9999)</pre>
> # Accumulate bootstrap values in bsdiff
> # For clarity, do operations in several separate steps inside the loop
> for(i in 1:BOOT)
+
      ł
+
      bootmarks <- marks[sample(1:n,replace=TRUE),] # sample rows with</pre>
+
                                                      # replacement
      bcorr <- cor(bootmarks) # Correlation matrix of bootstrap sample</pre>
+
+
      bdiffer <- bcorr[3,1]-bcorr[3,2] # Differencce between correlation</pre>
+
                                         # coefficients
+
      bsdiff <- c(bsdiff,bdiffer) # Add bdiffer to the end of bsdiff
      } # Next bootstrap sample
> bsdiff <- sort(bsdiff)</pre>
> # Now get lower and upper limits of 95% CI
```

```
> low <- bsdiff[.025*BOOT] ; up <- bsdiff[.975*BOOT + 1]</pre>
> low ; up
[1] -0.03643594
[1] 0.3032818
> (low+up)/2
[1] 0.1334230
> obsdiff
[1] 0.1282046
> write(bsdiff,"bsdiff.dat") # Maybe for later analysis
# boot1.R
             Working on the bootstrap
              R --vanilla < boot1.R > boot1.out &
# Run with
# grades.dat has 4 columns: ID, Verbal SAT, Math SAT and 1st year GPA
marks <- read.table("grades.dat")</pre>
n <- length(marks$verbal)</pre>
n
marks[1:10,]
obscorr <- cor(marks)</pre>
obscorr
# Question: Is the correlation between Verbal SAT and GPA the same as
# the correlation between math SAT and GPA?
# What is the sampling distribution of the difference between correlation
# coefficients?
#
obsdiff <- obscorr[3,1]-obscorr[3,2] # Verbal minus math</pre>
obsdiff
# The strategy will be to obtain a 95% bootstrap confidence interval for
# the difference between verbal correlation and math correlation. This
# confidence interval will be approximately centered around the observed
# difference obsdiff = .128. If the confidence interval does not include
# zero, we will conclude that the observed difference is significantly
# different from zero.
BOOT <- 100 ; bsdiff <- NULL ; set.seed(9999)
# Accumulate bootstrap values in bsdiff
# For clarity, do operations in several separate steps inside the loop
for(i in 1:BOOT)
    {
    bootmarks <- marks[sample(1:n,replace=TRUE),] # sample rows with</pre>
                                                    # replacement
    bcorr <- cor(bootmarks) # Correlation matrix of bootstrap sample</pre>
    bdiffer <- bcorr[3,1]-bcorr[3,2] # Differencce between correlation</pre>
                                       # coefficients
    bsdiff <- c(bsdiff,bdiffer) # Add bdiffer to the end of bsdiff
    } # Next bootstrap sample
bsdiff <- sort(bsdiff)</pre>
# Now get lower and upper limits of 95% CI
low <- bsdiff[.025*BOOT] ; up <- bsdiff[.975*BOOT + 1]</pre>
low ; up
(low+up)/2
obsdiff
write(bsdiff, "bsdiff.dat") # Maybe for later analysis
pdf("bsdiff.pdf") # Send graphics output to pdf file
hist(bsdiff)
```