

## lab7.R

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*#Calculate a parametric 95% CI for mean bill length (in mm) for the Gentoo penguins*

*#Q1*

```
require(palmerpenguins)
```

```
## Loading required package: palmerpenguins
```

```
alpha = 0.05
```

```
dat_gentoo = subset(penguins, species == "Gentoo")
```

```
n = length(na.omit(dat_gentoo$bill_length_mm))
```

```
#n = 123
```

*#Q2*

```
ssd = sd(dat_gentoo$bill_length_mm, na.rm = TRUE)
```

```
#ssd = 3.081857
```

*#Q3*

```
alpha = 0.05
```

```
t_crit = abs(qt(alpha / 2, df = n - 1))
```

```
#t_crit = 1.9796
```

*#Q4*

```
sse = ssd / sqrt(n)
```

```
#sse = 0.08847361
```

```
#sse = 0.2778817
```

```
#I had the correct code but copied the wrong number.
```

*#Q5*

```
ci_radius = sse * t_crit
```

```
ci = c(
```

```
  lower = mean(dat_gentoo$bill_length_mm, na.rm = TRUE) - ci_radius,
```

```
  upper = mean(dat_gentoo$bill_length_mm, na.rm = TRUE) + ci_radius)
```

```
print(round(ci, 4))
```

```
##      lower      upper
```

```
## 46.9548 48.0550
```

*#Bootstrap (Q6-9)*

```
#install.packages("boot")
```

```
require(boot)
```

```

## Loading required package: boot

boot_mean = function(x, i)
{
  return(mean(x[i], na.rm = TRUE))
}

myboot =
  boot(
    data = dat_gentoo$bill_length_mm,
    statistic = boot_mean,
    R = 10000)
print(myboot)

##
## ORDINARY NONPARAMETRIC BOOTSTRAP
##
##
## Call:
## boot(data = dat_gentoo$bill_length_mm, statistic = boot_mean,
##       R = 10000)
##
##
## Bootstrap Statistics :
##      original      bias    std. error
## t1* 47.50488 -0.002236324   0.2775417

quantile(
  myboot$t,
  c(0.025, 0.975))

##      2.5%      97.5%
## 46.96748 48.05611

#####

rm(list = ls())

moths = read.csv("/Users/stonehuang/Documents/environmental_data/data/moths.csv")
#Q9
rarefaction_sampler = function(input_dat, n_iterations)
{
  n_input_rows = nrow(input_dat)

  results_out = matrix(
    nrow = n_iterations,
    ncol = n_input_rows)

  # The outer loop: runs once for each bootstrap iteration.  index variable i

```

```

s i
for(i in 1:n_iterations)
{
  # The inner loop: simulates increasing sampling intensity
  # Sampling intensity ranges from 1 site to the complete count of
  # sites in the input data (n)
  for(j in 1:n_input_rows)
  {
    # sample the input data row indices, with replacement
    rows_j = sample(n_input_rows, size = j, replace=TRUE)

    # Creates a new data matrix
    t1 = input_dat[rows_j, ]

    # Calculates the column sums
    t2 = apply(t1, 2, sum)

    # Counts the number of columns in which any moths were observed
    results_out[i, j] = sum(t2 > 0)
  }
}
return(results_out)
}

#Q10
#Using the double loop while keeping track of what row and column should contain
#is the most difficult part about building the function.

#Q11
# Re-read my data:
moths = read.csv("/Users/stonehuang/Documents/environmental_data/data/moths.csv")
rarefact = rarefaction_sampler(moths[, -1], 10000)

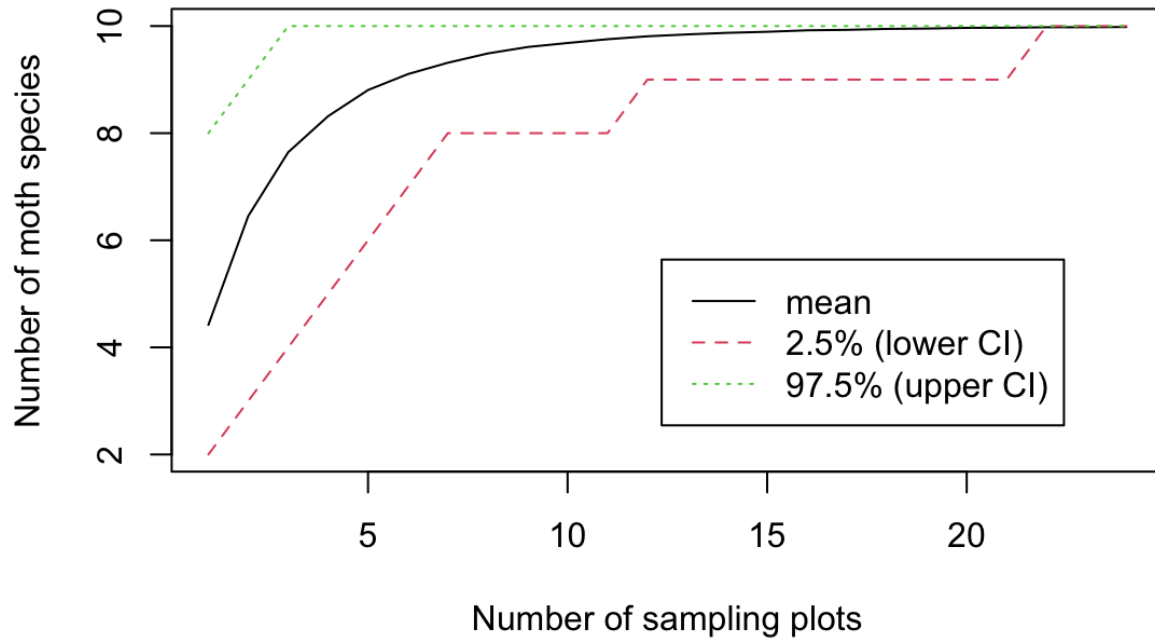
rare_mean = apply(rarefact, 2, mean)
rare_quant = apply(rarefact, 2, quantile, probs=c(0.025, 0.975))
rare = t(rbind(rare_mean, rare_quant))

#Q12
matplot(
  rare,
  type='l',
  xlab='Number of sampling plots',
  ylab='Number of moth species',
  main="Bootstrap rarefaction curve of 10 rare MA moth species")

legend(
  'bottomright',
  legend=c('mean', '2.5% (lower CI)', '97.5% (upper CI)'),
  lty=c(1,2,3), col=c(1,2,3), inset=c(.1,.1))

```

### Bootstrap rarefaction curve of 10 rare MA moth species



#Q13

#I would visit 22 sites if I want to see all of the moth species because both curves of both upper and lower confidence intervals reach 10 at 22 plots.