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Analysis of Environmental Data

sd(pen_boot\$t)
[1] 1.0047

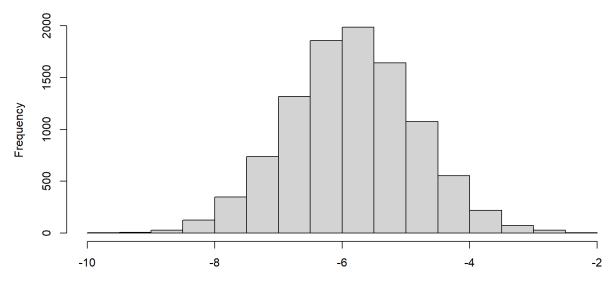
Lab 8

require(palmerpenguins)
 penguin_dat= droplevels(subset(penguins, species != 'Gentoo'))
 install.packages("simpleboot")
 require(simpleboot)

 adelie= subset(penguin_dat, species == 'Adelie')
 chinstrap= subset(penguin_dat, species== 'Chinstrap')

 pen_boot= two.boot(adelie\$flipper_length_mm, chinstrap\$flipper_length_mm, mean, R= 10000, na.rm= TRUE)

Histogram of 10000 bootstrap differences in mean penguin flipper length



Difference in mean flipper length of Adelie and Chinstrap penguins

2.

quantile(pen_boot\$t, c(0.025, 0.975))2.5% 97.5%-7.819310 -3.874961

- 4. The bootstrap resampled differences in means are pretty normally distributed as seen in the histogram, where the highest frequencies are right around -6 declining fairly evenly on both sides of the histogram. Both the mean and median are almost identical (both right around -5.85) which further supports the claim that these data are normally distributed.
- 5. pen ecdf= ecdf(pen boot\$t)
- 6. 1- pen_ecdf(-4.5) [1] 0.0872
- 7. pen_ecdf(-8) [1] 0.0161
- 8. The null hypothesis is that there is no difference in mean flipper length between the two species of penguins.

The alternative hypothesis is that there is a significant difference in the means of the two species of penguins.

```
9. veg<- read.csv("~/environmental_data/data/vegdata.csv")
dat_tree = droplevels(subset(veg, treatment %in% c("control", "clipped")))
wilcox.test(pine ~ treatment, data = dat_tree, alternative = "less")
```

Wilcoxon rank sum test with continuity correction

```
data: pine by treatment
W = 48, p-value = 0.9598
alternative hypothesis: true location shift is less than 0
```

10. veg<- read.csv("~/environmental_data/data/vegdata.csv")

```
dat_tree = droplevels(subset(veg, treatment %in% c("control", "clipped")))
wilcox.test(pine ~ treatment, data = dat_tree, alternative = "less")
control= droplevels(subset(dat_tree, treatment == 'control'))
clipped= droplevels(subset(dat_tree, treatment == 'clipped'))
tree_boot = two.boot(clipped$pine, control$pine, mean, R= 10000, na.rm= TRUE)
boot.ci(tree_boot)
```

BOOTSTRAP CONFIDENCE INTERVAL CALCULATIONS
Based on 10000 bootstrap replicates

```
CALL:
boot.ci(boot.out = tree_boot)
```

```
Intervals:
Level Normal Basic
95% (3.25, 28.81) (2.38, 27.62)

Level Percentile BCa
95% (4.38, 29.62) (5.00, 30.62)

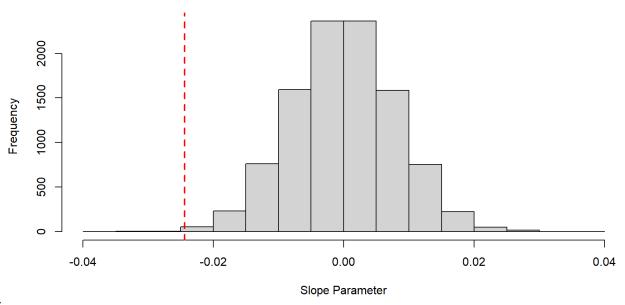
11. quantile(tree_boot$t, c(0.025, 0.975))
2.5% 97.5%
4.375 29.625
```

12. Simpsons diversity is measure of species richness and evenness which considers the overall number of species and the relative abundance of each species.

```
13. # Calculate the sample mean and sd:
    s_sidi_mean = mean(dat_all$s.sidi, na.rm = TRUE)
    s sidi sd = sd(dat all$s.sidi, na.rm = TRUE)
    # Use the subset-by-name symbol ($) to create a
    # new column of z-standardized values.
    dat_all$s.sidi.standardized = (dat_all$s.sidi - s_sidi_mean)/s_sidi_sd
14. dat 1 =
     subset(
      dat all,
      select = c(b.sidi, s.sidi))
    m = 10000
    result mc = numeric(m)
    for(i in 1:m)
     index_1 = sample(nrow(dat_1), replace = TRUE)
     index_2 = sample(nrow(dat_1), replace = TRUE)
     dat_resampled_i =
      data.frame(
       b.sidi = dat 1$b.sidi[index 1],
       s.sidi = dat_1$s.sidi[index_2]
      )
     fit_resampled_i = lm(b.sidi ~ s.sidi, data = dat_resampled_i)
```

```
slope_resampled_i = coef(fit_resampled_i)[2]
print(slope_resampled_i)
# ... your loop code ...
result_mc[i] = coef(fit_resampled_i)[2]
}
```

Julian's Null Distribution of Regression Slope



15.

16. quantile(result_mc, c(.05))
5%
-0.01329749

The observed slope is less than the critical value

17. The results of the simulation suggest weak evidence for a negative relationship between vegetation diversity and bird diversity. The majority of simulations produced slope parameters more positive than the observed slope.

```
18. m = 10000
  result_boot = numeric(m)
  for(i in 1:m)
  {
    index_1 = sample(nrow(dat_1), replace = TRUE)

    dat_boot = dat_1[index_1, ]

  fit_bs1 = lm(b.sidi ~ s.sidi, data = dat_boot)
```

```
# ... your loop code ...

result_boot[i] = coef(fit_bs1)
}
```