Bootstrapping ziCPUE data

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Initial setup

The following packages are required:

```
library('dplyr')
library('readxl')
library('ggplot2')
library('broom')
library('RColorBrewer')

Read, clean, and add necessary columns to the table in brett_example.xlsx.

CPUEdf <- read_excel("brett_example.xlsx",1) %>%
    rename(Area = `Effort m^2`, Count = SumOfSPEC) %>%
    filter(Year %in% 1993:2013, !is.na(Year), Month == 10, !is.na(Month)) %>%
    select(Year, Month, Station, Count, Area) %>%
    mutate(CPUE = (Count/Area)*100) %>%
    arrange(Year)
```

Bootstrap Resampling

For seamless bootstrapping with dplyr, the package broom provides the function boostrap. However, this function samples from all rows regardless of grouping. I've hacked the version of the function found here to sample within a given grouping as such: If, for example, you group by a variable called type and in the original sample, types 1, 2, and 3 have sample sizes of 10, 20, and 30, respectively. This function would randomly sample exactly 10 items from the original dataset (with replacement) that are of type 1, 20 of type 2, and 30 of type 3.

```
stratBootstrap <- function(df, m, group) {

# Inner function to sample row numbers from an input vector of sample sizes

# for each group

stratSample <- function(group_n){
   ends <- cumsum(group_n)
   starts <- c(1, ends[-length(ends)] - 1)
   bootRows <- unlist(lapply(seq(length(ends)), function(i){
      sample(seq(starts[i], ends[i]), replace = TRUE)}))
   return(bootRows)
}

df <- df %>% arrange_(group)

n <- nrow(df)</pre>
```

```
group_n <- (df %>%
    group_by_(group) %>%
    summarize(n = n()))[['n']]
  attr(df, "indices") <- replicate(m, stratSample(group_n) - 1,</pre>
                                    simplify = FALSE)
  attr(df, "drop") <- TRUE</pre>
  attr(df, "group_sizes") <- rep(n, m)</pre>
  attr(df, "biggest_group_size") <- n</pre>
  attr(df, "labels") <- data.frame(replicate = 1:m)</pre>
  attr(df, "vars") <- list(quote(replicate))</pre>
  class(df) <- c("grouped_df", "tbl_df", "tbl", "data.frame")</pre>
  return(df)
}
Now for the actual bootstrapping. We'll do 10,000 simulations and calculate the 95% CI via the "percentile" method.
set.seed(9721)
CPUEboot <- CPUEdf %>%
  stratBootstrap(1e4, group = 'Year') %>%
  do(summarize(group_by(., Year), CPUE = mean(CPUE))) %>%
  # To strip attributes:
  as.data.frame %>% as.tbl
# Summary table
CPUEbootSumm <- CPUEboot %>%
  group_by(Year) %>%
  summarize(low = quantile(CPUE, probs = 0.025),
            mid = quantile(CPUE, probs = 0.5),
            high = quantile(CPUE, probs = 0.975))
CPUEbootSumm
## Source: local data frame [19 x 4]
##
##
       Year
                    low
                                 mid
                                            high
##
      (dbl)
                   (dbl)
                               (db1)
                                            (dbl)
## 1
       1993 4.83832211 9.58010489 16.83885783
## 2
      1994 6.99210643 15.83355396 26.44557596
## 3
       1995 8.39890051 20.89424272 35.98938845
      1996 1.31134895 24.82473582 61.64871335
## 4
## 5
      1997 3.57124248 6.70801146 11.27221312
      1999 1.65468194 4.36716710 7.85672566
       2000 0.20093735 0.47319168 0.81233606
## 7
## 8
       2001 0.41001564 0.80018388 1.28325638
## 9
       2002 0.02824304 0.09308116 0.18038384
## 10 2003 0.00000000 0.02769080 0.06139001
## 11 2004 0.33760247 0.95487542 1.76591515
## 12 2005 12.99019720 26.96280502 47.85944003
## 13 2006 0.35316891 0.52012957 0.76636280
## 14 2007 2.74382176 5.31447841 8.78115557
## 15 2008 5.14752570 8.38709651 12.52172003
```

```
## 16 2009 6.21371226 10.32210104 15.49383447
## 17 2010 0.29392622 0.75528146 1.40376504
## 18 2012 0.00000000 0.00000000 0.00000000
## 19 2013 0.00000000 0.02905816 0.06399798
```

Plotting

These are just some aesthetic pieces that I often employ. They're included here in case others are interested.

```
# Minimal ggplot2 theme
plotTheme <- function(base_size = 10, base_family = 'Helvetica') {</pre>
  theme_minimal(base_size = base_size, base_family = base_family) %+replace%
      strip.text = element text(face = 'bold'),
      panel.grid.minor = element_blank(),
      panel.grid.major.x = element_blank(),
      panel.grid.major.y = element_line(color = 'gray50', size = 0.125, linetype = 3),
      panel.border = element rect(fill = NA, color = "gray50"),
      axis.ticks = element_line(color = "gray50"),
      axis.ticks.length = unit(2, 'points'),
      legend.position = 'none'
    )
}
# Color palette for unique values in an input vector
myPalette <- function(inputVector){</pre>
  outPalette <- inputVector %>%
    unique %>%
    length %>%
    colorRampPalette(brewer.pal(8, 'Dark2'))(.)
  return(outPalette)
}
```

Base plot

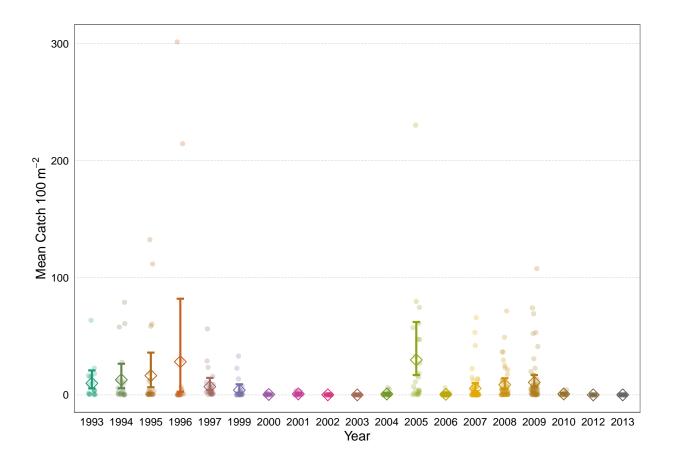
All following plots are based off this base one. Having no geoms, it just stores the data and some aesthetics that we'll pass on to plots later.

```
bootPlot <- CPUEdf %>%
  ggplot(aes(x = factor(Year), y = CPUE, color = factor(Year))) +
  plotTheme() +
  scale_x_discrete('Year', breaks = seq(min(CPUEdf$Year), max(CPUEdf$Year))) +
  scale_color_manual(values = myPalette(CPUEdf$Year))
```

Bootstrapping via boot in ggplot2

To have better control of bootstrapping within ggplot2 plots, the following function allows you to use the boot package for bootstrapping and CI calculations. The ciMethod parameter takes the following inputs, inherited from the boot.ci function: "norm", "basic", "stud", "perc", or "bca". If parallel bootstrapping is desired, on Mac OSX, add parallel = "multicore" and on Windows add parallel = "snow" to the boot function call; also add ncpus = x, where x is the number of available cores.

```
ggBootCI <- function(inputData, numSims = 1e4, ciMethod = 'bca'){</pre>
  # List names in the `boot.ci` output for these methods differ from their input names
  if (ciMethod %in% c('norm', 'stud', 'perc')){
    outListMethod <- gsub('norm', 'normal', ciMethod) %>%
      gsub('stud', 'student', .) %>%
      gsub('perc', 'percent', .)
  } else {
    outListMethod <- ciMethod
  # Inner function to compute mean and variance of the sampling distribution of the mean
  # for each bootstrap replication
  bootMean <- function(inputData, ind){</pre>
    m <- mean(inputData[ind])</pre>
    n <- length(ind)</pre>
    v <- var(inputData[ind]) / n</pre>
    return(c(m, v))
  bootstraps <- boot(inputData, statistic = bootMean, R = numSims)</pre>
  # If all resamples return the same value, output df will simply be that value x3:
  if (diff(range(bootstraps$t)) == 0){
    result <- range(bootstraps$t)[1] %>%
      data.frame(ymin = ., y = ., ymax = .)
  } else { # If not, then run `boot.ci`:
    boot.ciOutput <- boot.ci(bootstraps, type = ciMethod)</pre>
    bootCI <- rev(boot.ciOutput[[outListMethod]])[c(2,1)]</pre>
    result <- data.frame(ymin = bootCI[1],</pre>
                          y = median(bootstraps$t),
                          ymax = bootCI[2])
  }
 return(result)
}
Now we'll employ the above function, using 1,000 simulations and calculating the 95% CI via the "bca" method: bias-corrected
and accelerated bootstrapping.
bootPlot +
  scale_y_continuous(expression(Mean ~ Catch ~ 100 ~ m^{-2})) +
  geom_point(position = position_jitter(width = 0.3, height = 0),
             alpha = 0.25, shape = 16) +
  stat_summary(fun.data = ggBootCI,
                fun.args = list(numSims = 1e3, ciMethod = 'bca'),
                geom = "errorbar", width = 0.25, size = 0.75) +
  stat_summary(fun.y = "mean", geom = "point", size = 3, shape = 23)
```



Plot on log₁₀ scale

To create a plot on the log_{10} scale, we'll use the bootstrapping we did earlier via the stratBootstrap function, located in the CPUEbootSumm data frame. Trying to transform the y-axis while calculating bootstrapped CI on the fly proved a fruitless endeavor.

