IS5 in R: Confidence Intervals for Means (Chapter 14)

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Introduction and background

This document is intended to help describe how to undertake analyses introduced as examples in the Fifth Edition of *Intro Stats* (2018) by De Veaux, Velleman, and Bock. More information about the book can be found at http://wps.aw.com/aw_deveaux_stats_series. This file as well as the associated R Markdown reproducible analysis source file used to create it can be found at http://nhorton.people.amherst.edu/is5.

This work leverages initiatives undertaken by Project MOSAIC (http://www.mosaic-web.org), an NSF-funded effort to improve the teaching of statistics, calculus, science and computing in the undergraduate curriculum. In particular, we utilize the mosaic package, which was written to simplify the use of R for introductory statistics courses. A short summary of the R needed to teach introductory statistics can be found in the mosaic package vignettes (http://cran.r-project.org/web/packages/mosaic). A paper describing the mosaic approach was published in the R Journal: https://journal.r-project.org/archive/2017/RJ-2017-024.

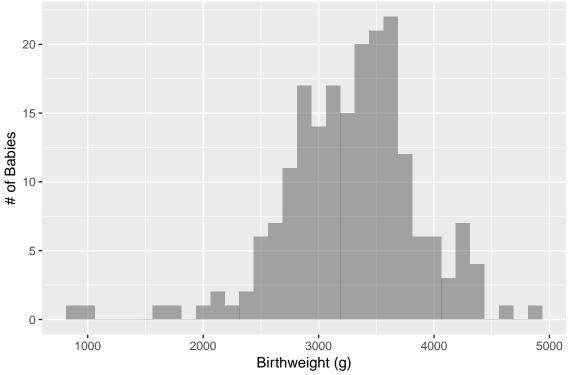
Chapter 14: Confidence Intervals for Means

```
library(mosaic)
library(readr)
library(janitor)
Babies <- read_csv("http://nhorton.people.amherst.edu/is5/data/Babysamp_98.csv") %>%
  clean_names()
## Parsed with column specification:
## cols(
##
     MomAge = col_integer(),
##
     DadAge = col_integer(),
##
     MomEduc = col_integer(),
##
     MomMarital = col_integer(),
     numlive = col_integer(),
##
##
     dobmm = col integer(),
##
     gestation = col_integer(),
##
     sex = col_character(),
     weight = col_integer(),
##
##
     prenatalstart = col integer(),
     orig.id = col integer(),
##
##
     preemie = col_logical()
## )
```

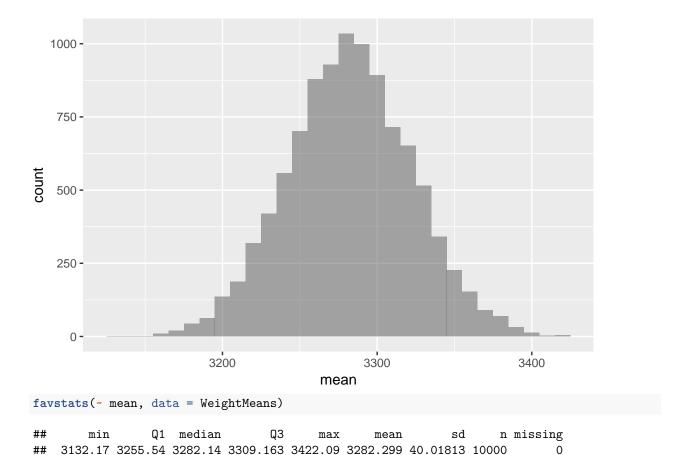
By default, read_csv() prints the variable names. These messages can be suppressed using the message = FALSE code chunk option to save space and improve readability.

Here we use the clean_names() function from the janitor package to sanitize the names of the columns (which would otherwise contain special characters or whitespace).

```
# Figure 14.1, page 441
gf_histogram(~ weight, data = Babies, binwidth = 125) %>%
gf_labs(x = "Birthweight (g)", y = "# of Babies")
```

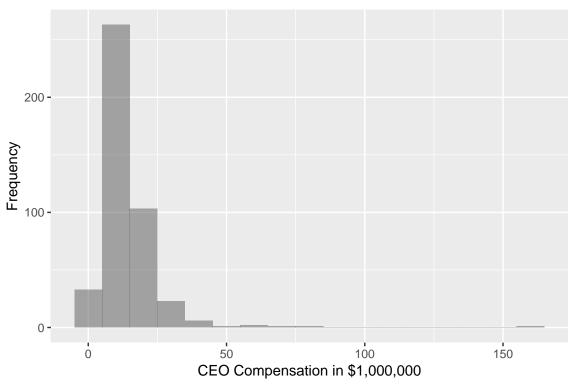


```
# Figure 14.2
set.seed(12346) # For reproducibility
numsim <- 10000 # Number of simulations
# What does do() do?
mean(~ weight, data = sample(Babies, size = 100)) # Mean of a random sample of 100
## [1] 3214.53
mean(~ weight, data = sample(Babies, size = 100)) # Mean of another random sample
## [1] 3290.34
do(2) * mean(~ weight, data = sample(Babies, size = 100)) # Calculates the mean twice
##
        mean
## 1 3297.84
## 2 3286.70
# For the visualization, we need 10,000 means
WeightMeans <- do(numsim) * mean(~ weight, data = sample(Babies, size = 100))</pre>
Here, the do() function calculates, 10,000 times, the mean of a random sample of 100 Babies weights.
gf_histogram(~ mean, data = WeightMeans, binwidth = 10)
```

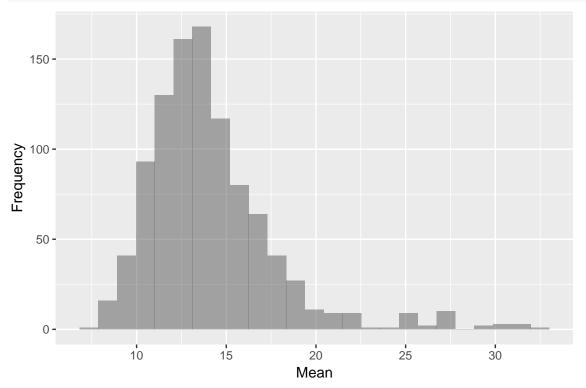


Section 14.1: The Central Limit Theorem

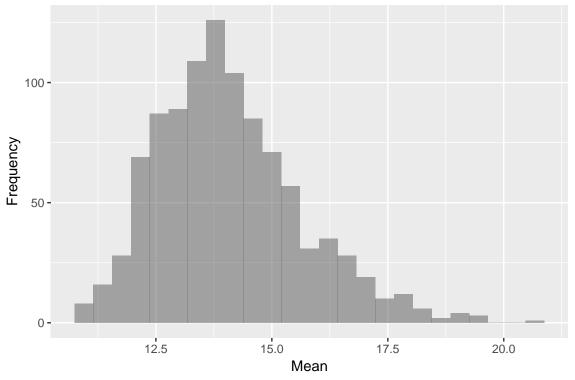
```
CEOComp <- read_csv("http://nhorton.people.amherst.edu/is5/data/CEO_Compensation_2014.csv") %>%
  clean_names()
## Parsed with column specification:
## cols(
##
     Employer = col_character(),
     CEO = col_character(),
##
     CEO_Compensation = col_integer(),
##
##
    Median_Worker_Comp = col_integer(),
##
    Ratio = col integer(),
##
     Company_Rating = col_double(),
     `CEO_Compensation_($M)` = col_double()
##
## )
# Figure 14.3
gf_histogram(~ ceo_compensation_m, data = CEOComp, binwidth = 10) %>%
  gf_labs(x = "CEO Compensation in $1,000,000", y = "Frequency")
```



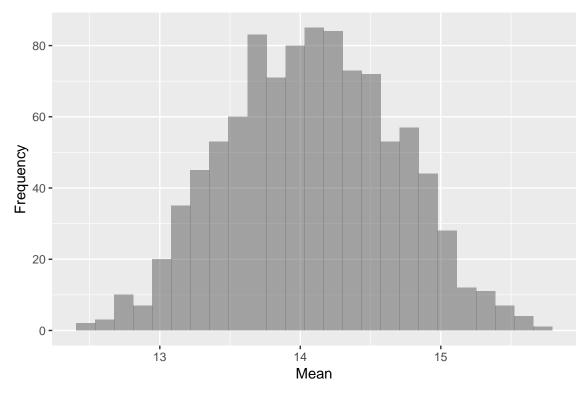
```
# Figure 14.4
set.seed(14354)
numsim <- 1000 # Here the number of simulations is 1,000
CEOMeans <- do(numsim) * mean(~ ceo_compensation_m, data = sample(CEOComp, 10))
gf_histogram(~ mean, data = CEOMeans) %>%
gf_labs(x = "Mean", y = "Frequency")
```



```
# Figure 14.5
CEOMeans2 <- do(numsim) * mean(~ ceo_compensation_m, data = sample(CEOComp, 50))
gf_histogram(~ mean, data = CEOMeans2) %>%
gf_labs(x = "Mean", y = "Frequency")
```



```
# Figure 14.7 (skipped 14.6 because it's similar)
CEOMeans3 <- do(numsim) * mean(~ ceo_compensation_m, data = sample(CEOComp, 200))
gf_histogram(~ mean, data = CEOMeans3) %>%
gf_labs(x = "Mean", y = "Frequency")
```

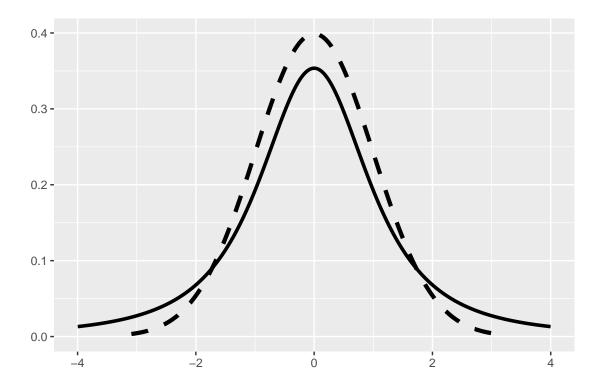


Here, in each example, the do() function calculates the mean of random samples of 10, 50, and 200 CEO compensations.

Section 14.2: A Confidence Interval for the Mean

```
# Figure 14.9, page 446
gf_dist(dist = "norm", linetype = 2, lwd = 1.5) %>%
gf_labs(x = "", y = "") %>%
gf_dist(dist = "t", df = 2, lwd = 1.25) +
xlim(-4, 4)
```

Warning: Removed 4104 rows containing missing values (geom_path).



Example 14.1: A One-Sample t-Interval for the Mean

clean_names()

confint(salmonlm, data = Salmon)

2.5 %

97.5 %

##

```
## Parsed with column specification:
## cols(
##
     Kind = col_character(),
##
     Location = col_character(),
##
     Mirex = col_double(),
##
     Hexachlorobenzene = col_double(),
     HCH_gamma = col_double(),
##
##
     `Heptachlor Epoxide` = col_double(),
##
     Dieldrin = col_double(),
##
     Endrin = col_double(),
##
     `Total Chlordane` = col_double(),
##
     `Total DDT` = col_double(),
##
     Dioxin = col_double(),
##
     `Total Pesticides` = col_double(),
     `Total PCBs` = col_integer()
##
## )
Salmon <- Salmon %>%
  filter(mirex != "NA")
favstats(~ mirex, data = Salmon)
                           Q3
##
           Q1 median
                                max
                                       mean
                                                          n missing
                                                     sd
      0 0.056  0.079  0.13475  0.194  0.09134  0.04952388  150
salmonlm <- lm(mirex ~ 1, data = Salmon)</pre>
```

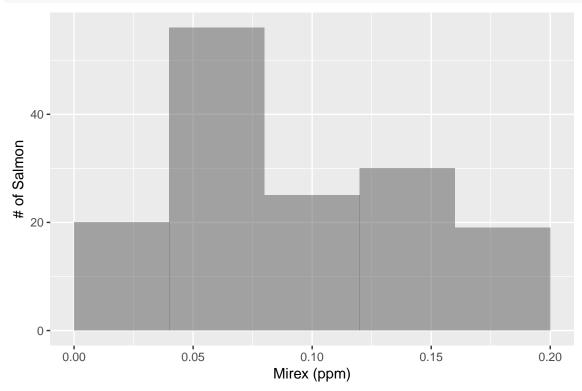
Salmon <- read_csv("http://nhorton.people.amherst.edu/is5/data/Farmed_salmon.csv") %>%

(Intercept) 0.08334978 0.09933022

The confint() function takes an object, such as a linear model, as an argument.

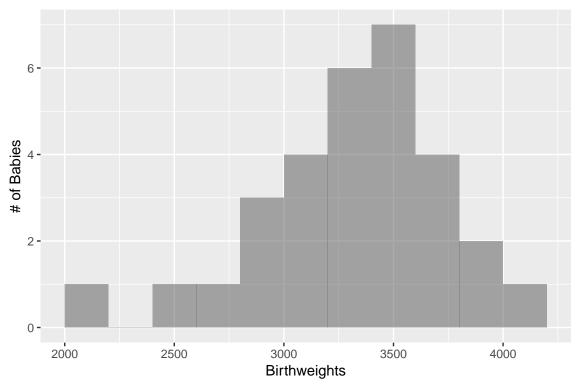
Example 14.2: Checking Assumptions and Conditions for Student's t

```
gf_histogram(~ mirex, data = Salmon, binwidth = 0.04, center = 0.02) %>%
gf_labs(x = "Mirex (ppm)", y = "# of Salmon")
```



Step-By-Step Example: A One-Sample t-Interval for the Mean

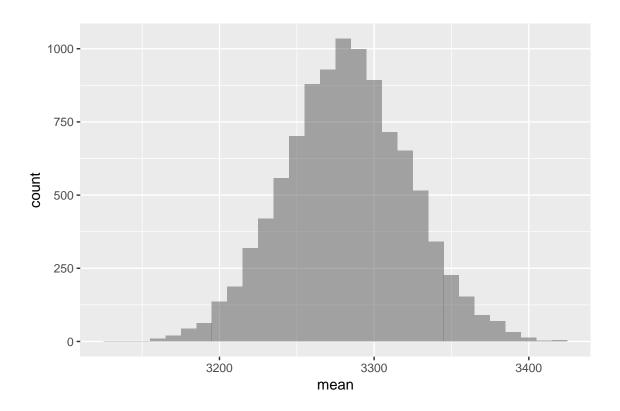
```
set.seed(34)
BabiesSample <- sample(Babies, 30)
gf_histogram(~ weight, data = BabiesSample, binwidth = 200, center = 100) %>%
gf_labs(x = "Birthweights", y = "# of Babies")
```



Section 14.3: Interpreting Confidence Intervals

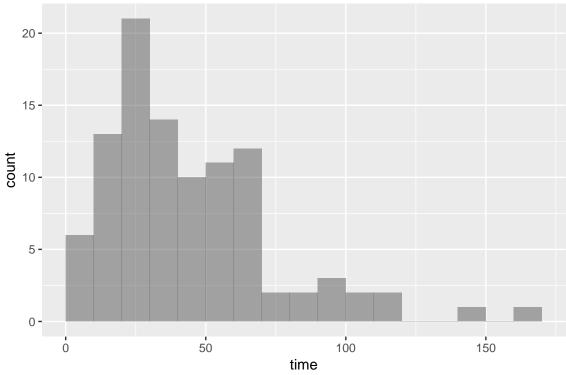
Section 14.4: Picking Our Interval up by Our Bootstraps

```
gf_histogram(~ mean, data = WeightMeans, binwidth = 10)
```



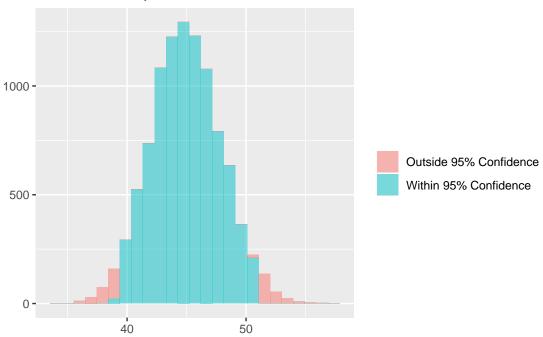
Step-By-Step Example: A Bootstrap Confidence Interval for the Mean

```
CommuteSample <- read_csv("http://nhorton.people.amherst.edu/is5/data/Commuter_sample.csv")
## Parsed with column specification:
## cols(
## time = col_integer()
## )
gf_histogram(~ time, data = CommuteSample, binwidth = 10, center = 5)</pre>
```



```
# Bootstrap
set.seed(134) # For reproducibility
numsim <- 10000
commutebootstrap <- do(numsim) * mean(~ time, data = resample(CommuteSample))</pre>
qdata(~ mean, p = c(.025, .975), data = commutebootstrap) # grab the percentiles with qdata
##
         quantile
## 2.5%
            39.28 0.025
## 97.5%
            50.98 0.975
confint(commutebootstrap, method = "quantile") # confint of data frame
     name lower upper level
                                method estimate
## 1 mean 39.28 50.98 0.95 percentile
commutebootstrap <- commutebootstrap %>%
  mutate(interval = ifelse(mean > 39.28 & mean < 50.98, "Within 95% Confidence",</pre>
                           "Outside 95% Confidence")) # for fill
gf_histogram(~ mean, fill = ~ interval, data = commutebootstrap,
             title = "Time Bootstrap Set Means") %>%
  gf_labs(x = "", y = "", fill = "")
```

Time Bootstrap Set Means



Section 14.5: Thoughts About Confidence Intervals