# IS5 in R: Testing Hypotheses (Chapter 15)

Nicholas Horton (nhorton@amherst.edu)

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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. 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 (https://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 15: Testing Hypotheses

```
library(mosaic)
library(readr)
library(janitor)
```

Section 15.1: Hypotheses

Section 15.2: P-Values

Section 15.3: The Reasoning of Hypothesis Testing

Example 15.5: Finding A P-Value It is straightforward to find p-values using summary statistics.

```
n <- 90
x <- 61
p <- .8
phat <- x / n
sdphat <- ((p * (1 - p)) / n)^.5
z <- (phat - p) / sdphat
pnorm(z)</pre>
```

## [1] 0.00187324

```
# Or, without calculating the z-score:
pnorm(q = phat, mean = p, sd = sdphat)
```

## [1] 0.00187324

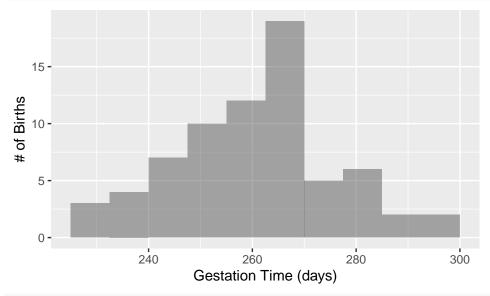
#### Section 15.4: A Hypothesis Test for the Mean

We begin by reading the data.

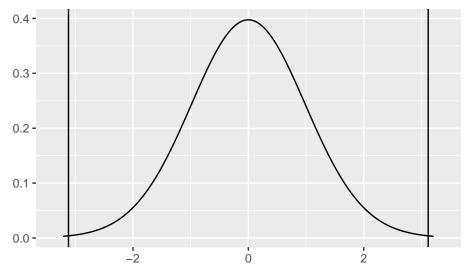
```
GestationTime <- read_csv("http://nhorton.people.amherst.edu/is5/data/Nashville.csv")</pre>
```

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.

```
# 2. Model (page 482)
gf_histogram(~Gestation, data = GestationTime, binwidth = 7.5, center = 3.75) %>%
gf_labs(x = "Gestation Time (days)", y = "# of Births")
```



```
# 3. Mechanics
gf_dist(dist = "t", df = 69) %>%
    gf_vline(xintercept = -3.118) %>%
    gf_vline(xintercept = 3.118) %>%
    gf_labs(x = "", y = "") +
    xlim(-3.347, 3.347)
```



Step-By-Step Example: A One-Sample t-Test for the Mean We begin by reading in the data.

```
# page 485
Sleep <- read_csv("http://nhorton.people.amherst.edu/is5/data/Sleep.csv")</pre>
```

```
# Plan
df_stats(~Sleep, data = Sleep)
## response min Q1 median Q3 max mean sd n missing
## 1
               4 6
                         7 7 8 6.64 1.075484 25
        Sleep
gf_histogram(~Sleep, data = Sleep, binwidth = 1) %>%
gf_labs(x = "Hours of Sleep", y = "")
8 -
6 -
4 -
2 -
                          Hours of Sleep
gf_dist(dist = "t", df = 24) %>%
 gf_vline(xintercept = -1.67) %>%
 gf_{labs}(x = "", y = "") +
xlim(-3, 3)
0.4 -
0.3 -
0.2 -
0.1 -
0.0 -
                                  0
# Mechanics
n <- 25
mean \leftarrow 7.0
df <- 24
y <- 6.64
s <- 1.075
```

```
sey <- s / (n^.5)
t <- (y - mean) / sey # t-statistic
pt(q = t, df = df) # p-value</pre>
```

## Section 15.5: Intervals and Tests

## [1] -2.675722 2.675722

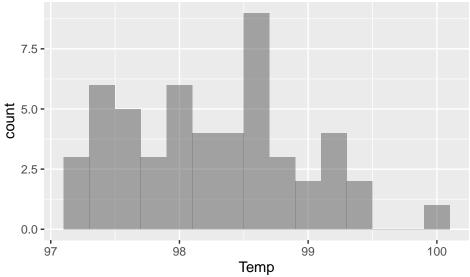
## [1] 0.05351625

It is straightforward to calculate confidence intervals and carry out hypothesis tests.

```
# page 487
Temperatures <- read_csv("http://nhorton.people.amherst.edu/is5/data/Normal_temperature.csv")
df_stats(~Temp, data = Temperatures)

## response min Q1 median Q3 max mean sd n missing
## 1 Temp 97.2 97.675 98.2 98.7 100 98.28462 0.6823789 52 0

gf_histogram(~Temp, data = Temperatures, binwidth = .2)</pre>
```



```
# Confidence interval
y <- mean(~Temp, data = Temperatures)
y

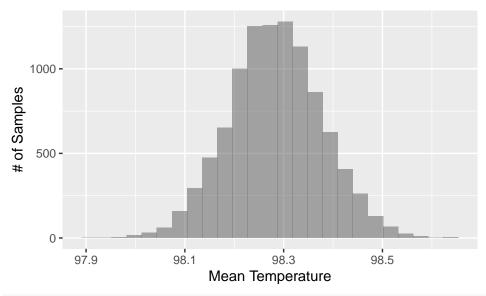
## [1] 98.28462
s <- sd(~Temp, data = Temperatures)
s

## [1] 0.6823789
n <- nrow(Temperatures)
n

## [1] 52
tstats <- qt(df = n - 1, p = c(.005, .995))
tstats
```

```
y + (tstats * (s / (n^{.5})))
## [1] 98.03141 98.53782
# Hypothesis test
mu <- 98.6
t \leftarrow (y - mu) / (s / (n^{.5}))
## [1] -3.332856
2 * pt(q = t, df = n - 1) # two sided test
## [1] 0.001605849
Random Matters: Bootstrap Hypothesis Tests and Intervals The boostrap is a flexible alternative
approach to inference.
numsamp <- 10000
# What does do() do?
mean(~Temp, data = resample(Temperatures)) # Mean of one random resample
## [1] 98.41538
mean(~Temp, data = resample(Temperatures)) # Mean of another random resample
## [1] 98.17115
do(2) * mean(~Temp, data = resample(Temperatures)) # Calculates means of two resamples
##
         mean
## 1 98.30385
## 2 98.27692
# We will use do() a numsamp number of times
resampletemps <- do(numsamp) * mean(~Temp, data = resample(Temperatures))</pre>
For more information about resample(), refer to the resample vignette in mosaic.
```

```
gf_histogram(~mean, data = resampletemps) %>%
gf_labs(x = "Mean Temperature", y = "# of Samples")
```



```
qdata(~mean, p = c(.005, .995), data = resampletemps) # reject null hypothesis

## 0.5% 99.5%

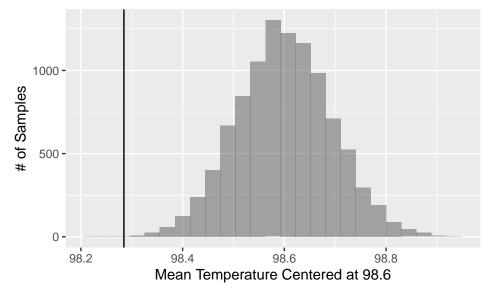
## 98.04422 98.52692

# Making a model-centric distribution

Temperatures2 <- Temperatures %>%
    mutate(Temp = Temp + .315)

resampletemps2 <- do(numsamp) * mean(~Temp, data = resample(Temperatures2))

gf_histogram(~mean, data = resampletemps2) %>%
    gf_vline(xintercept = mean(~Temp, data = Temperatures)) %>%
    gf_labs(x = "Mean Temperature Centered at 98.6", y = "# of Samples")
```



Step-By-Step Example: Tests and Intervals We begin by creating the dataset.

```
# Creating the data set
Baseball <- rbind(
  do(1308) * (winner <- "HOME"),
  do(2431 - 1308) * (winner <- "AWAY")</pre>
```

```
) %>%
 rename(winner = result)
# Mechanics (page 490)
n <- nrow(Baseball)</pre>
p < -.5
phat <- Baseball %>%
 filter(winner == "HOME") %>%
 nrow() / n
phat
## [1] 0.5380502
sdphat <- ((p * (1 - p)) / n)^.5
sdphat
## [1] 0.01014092
z \leftarrow (phat - p) / sdphat # z-value
## [1] 3.752142
1 - pnorm(z) \# p-value
## [1] 8.76651e-05
# Or, without calculating the z-score:
1 - pnorm(q = phat, mean = p, sd = sdphat)
## [1] 8.76651e-05
# Mechanics (page 491)
sep <- ((phat * (1 - phat)) / n)^.5
## [1] 0.01011152
me <- 1.96 * sep
phat - me # lower bound of 95% confidence
## [1] 0.5182316
phat + me # upper bound of 95% confidence
## [1] 0.5578688
```

Section 15.6: P-Values and Decisions: What to Tell About a Hypothesis Test