IS5 in R: Understanding and Comparing Distributions (Chapter 4)

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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.

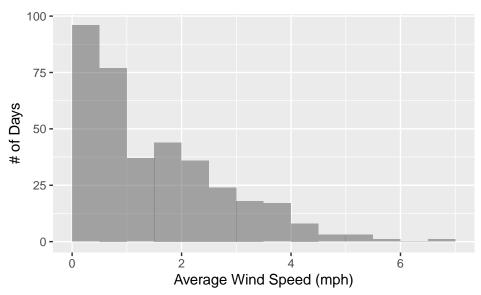
This work leverages initiatives undertaken by Project MOSAIC (http://www.mosaic-web.org), an NSF-funded effort to improve the teaching of statistics, calculus, science and computing in the undergraduate curriculum. In particular, we utilize the mosaic package, which was written to simplify the use of 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 4: Understanding and Comparing Distributions

```
library(mosaic)
library(readr)
library(janitor)
HopkinsForest <- read csv("http://nhorton.people.amherst.edu/is5/data/Hopkins Forest.csv") %>%
  janitor::clean names()
names(HopkinsForest)
   [1] "date"
                               "year"
                                                      "month"
##
    [4] "day"
                               "day_of_year"
                                                      "avg temp c"
                               "min_temp_c"
                                                      "avg_temp_f"
   [7] "max_temp_c"
## [10] "max_temp_f"
                               "min temp f"
                                                      "avg_rel_hum_percent"
  [13] "max_rel_hum_percent"
                               "min_rel_hum_percent"
                                                     "avg_sol_rad_w_m_2"
## [16] "max_sol_rad_w_m_2"
                               "min_sol_rad_w_m_2"
                                                      "total_sol_rad_w_m_2"
## [19] "avg_wind_mph"
                               "max_wind_mph"
                                                      "min_wind_mph"
## [22] "avg barom mb"
                               "max barom mb"
                                                      "min barom mb"
## [25] "precip in"
                               "deep_well_ft"
                                                      "shallow well ft"
## [28] "x80 cm soil c"
                               "x10 cm soil c"
```

By default, read_csv() prints the variable names. We suppressed these 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). You can use the names() function to check the cleaned names.

```
# Figure 4.1, page 96
gf_histogram(~avg_wind_mph,
  data = HopkinsForest,
  xlab = "Average Wind Speed (mph)", ylab = "# of Days", binwidth = 0.5, center = 0.25
)
```



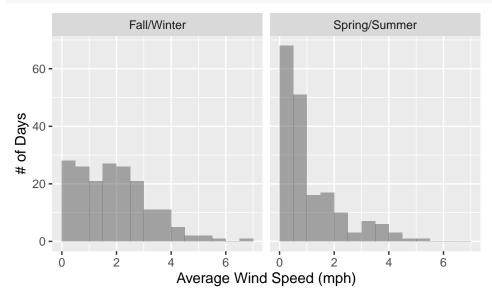
```
df_stats(~avg_wind_mph, data = HopkinsForest) # an improved version of "favstats()"

## response min Q1 median Q3 max mean sd n missing
## 1 avg_wind_mph 0 0.46 1.12 2.28 6.73 1.507808 1.260161 365 0
```

Section 4.1: Displays for Comparing Groups

Histograms We began by creating a new month to categorize the dates.

```
HopkinsForest <- HopkinsForest %>%
  mutate(catmonth = ifelse(month <= 9 & month >= 4, "Spring/Summer", "Fall/Winter"))
# Figure 4.2, page 96
gf_histogram(~avg_wind_mph,
  data = HopkinsForest, binwidth = 0.5, center = 0.25,
  xlab = "Average Wind Speed (mph)", ylab = "# of Days"
) %>%
  gf_facet_wrap(~catmonth)
```



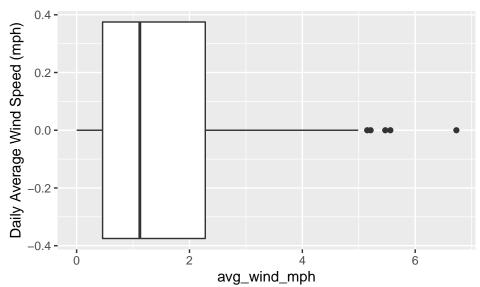
```
df_stats(avg_wind_mph ~ catmonth, data = HopkinsForest)
##
         response
                       catmonth min
                                       Q1 median
                                                      Q3
                                                                              sd
                                                          max
                                                                  mean
                                                                                   n
## 1 avg_wind_mph
                    Fall/Winter 0.02 0.84
                                             1.72 2.6575 6.73 1.904176 1.287233 182
## 2 avg_wind_mph Spring/Summer 0.00 0.35
                                            0.71 1.6150 5.47 1.113607 1.102176 183
     missing
## 1
           0
## 2
           0
```

Example 4.1: Comparing Groups with Stem-And-Leaf We begin by reading in the data.

```
# Figure 4.1, page 97
NestEgg <- read_csv("http://nhorton.people.amherst.edu/is5/data/Nest_Egg_Index.csv") %>%
    janitor::clean_names()
with(NestEgg, stem(nest_egg_index))
##
## The decimal point is 1 digit(s) to the right of the |
##
## 8 | 57789
## 9 | 0123344
```

Boxplots As noted in the book, boxplots are most useful to compare distributions. Below, we have replicated the single boxplot from page 98.

```
# Step 4 on page 98
gf_boxplot(~avg_wind_mph, data = HopkinsForest) %>% # or gf_boxplot(X ~ 1)
gf_labs(y = "Daily Average Wind Speed (mph)")
```



##

##

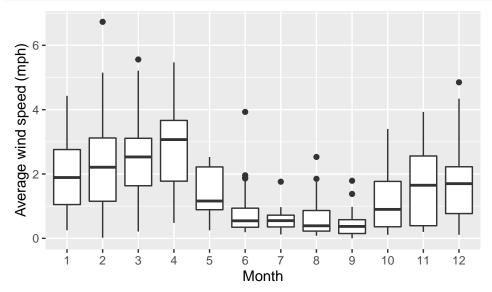
9 | 667777888899

10 | 0012233333344 10 | 5566779

11 | 122444

I don't recommend the use of single boxplots. Instead, one can make comparisons more easily by placing boxplots side by side with the following code:

```
# Figure 4.3, page 99
gf_boxplot(avg_wind_mph ~ as.factor(month), data = HopkinsForest) %>%
gf_labs(x = "Month", y = "Average wind speed (mph)")
```



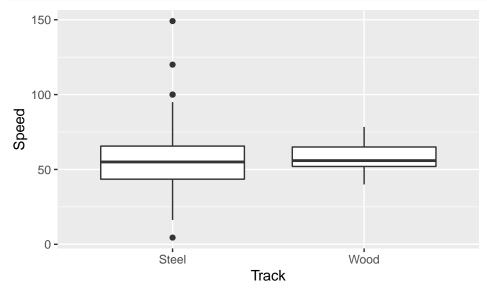
We use the as.factor() function to convert a variable into a factor.

We also use gf_labs() to clean up the code for the first line and improve readability.

Here we use the mosaic modeling language to specify the variables. As a general form, $\texttt{GOAL}(Y \sim X)$ carries out a specific goal for Y as a function of X.

Example 4.2: Comparing Groups with Boxplots We begin by reading in the data.

```
# Example 4.2, page 99
Coasters <- read_csv("http://nhorton.people.amherst.edu/is5/data/Coasters_2015.csv")
gf_boxplot(Speed ~ Track, data = Coasters)</pre>
```

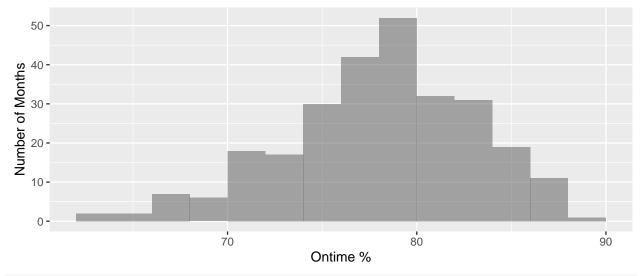


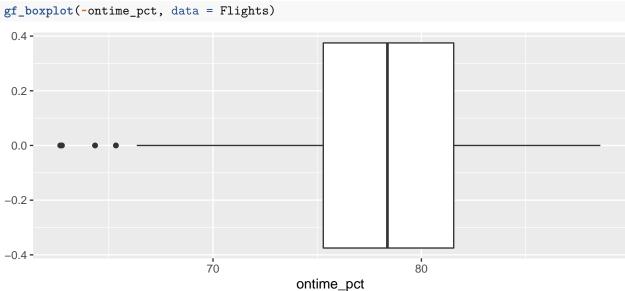
Step-By-Step Example: Comparing Groups We begin by reading in the data.

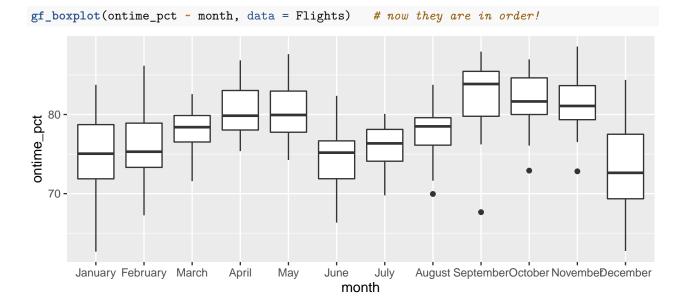
Cups <- read_csv("http://nhorton.people.amherst.edu/is5/data/Cups.csv")</pre> ## -- Column specification -----## cols(## Difference = col_double(), ## Container = col_character() ##) df_stats(Difference ~ Container, data = Cups) Q1 median response Container min Q3 max meansd n missing ## 1 Difference CUPPS 6 6.00 8.25 12.625 18.5 10.1875 5.202592 8 ## 2 Difference Nissan 0 0 1.25 2.00 3.750 7.0 2.7500 2.507133 8 ## 3 Difference SIGG 9 11.75 14.25 21.125 24.5 16.0625 5.900590 8 0 8.50 13.625 17.5 10.2500 4.551295 8 ## 4 Difference Starbucks 6 6.75 0 # Mechanics, page 101 gf_boxplot(Difference ~ Container, data = Cups, ylab = "Temp Change in F") 25 -20 ш Temp Change in 15 -10 -5 -0 -CUPPS Nissan SIGG Starbucks Container

Just Checking We begin by reading in the data.

```
Flights <- read_csv("http://nhorton.people.amherst.edu/is5/data/Flights_on_time_2016.csv") %>%
    janitor::clean_names()
# Let's improve the ordering of the months (by default they are alphabetical!)
Flights <- Flights %>%
    mutate(month = forcats::fct_relevel(
        month,
        "January", "February", "March", "April",
        "May", "June", "July", "August",
        "September", "October", "November", "December"
    )
    )
# Bureau of Transportation Statistics, page 101
gf_histogram(~ontime_pct, data = Flights, binwidth = 2, center = 1) %>%
    gf_labs(x = "Ontime %", y = "Number of Months")
```

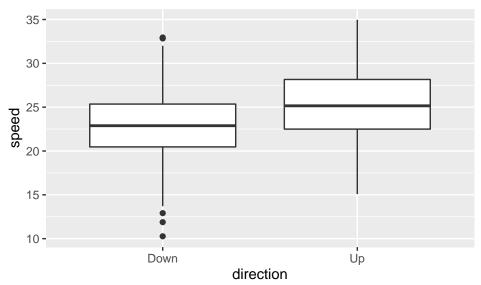






Random Matters We begin by reading in the data.

```
# Figure 4.4, page 102
CarSpeeds <- read_csv("http://nhorton.people.amherst.edu/is5/data/Car_speeds.csv")
gf_boxplot(speed ~ direction, data = CarSpeeds)</pre>
```

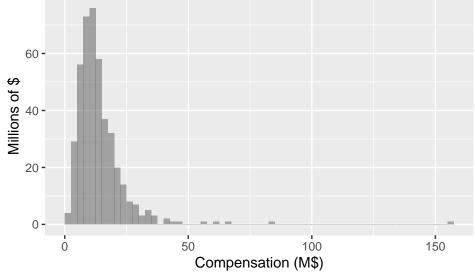


Section 4.3: Re-Expressing Data: A First Look

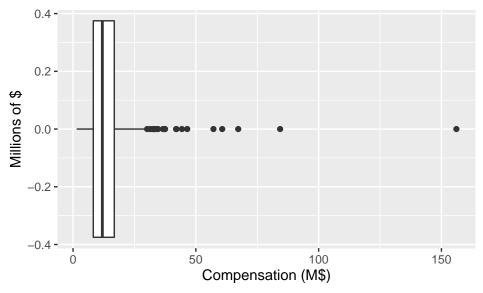
Re-Expressing to Improve Symmetry We begin by reading in the data.

```
CEOComp <- read_csv("http://nhorton.people.amherst.edu/is5/data/CEO_Compensation_2014.csv") %>%
    janitor::clean_names()
```

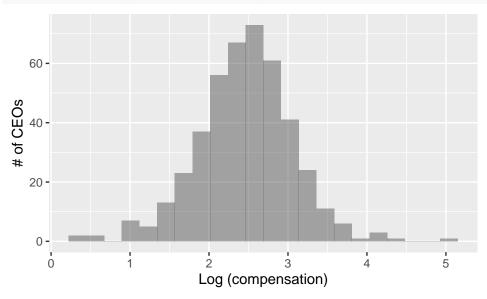
```
# Figure 4.6, page 105
gf_histogram(~ceo_compensation_m, data = CEOComp, binwidth = 2.5, center = 2.5 / 2) %>%
gf_labs(x = "Compensation (M$)", y = "Millions of $")
```



```
gf_boxplot(~ceo_compensation_m, data = CEOComp) %>%
gf_labs(x = "Compensation (M$)", y = "Millions of $")
```



```
# Figure 4.7, page 106
gf_histogram(~ log(ceo_compensation_m), data = CEOComp, binwidth = .224, center = .112) %>%
gf_labs(x = "Log (compensation)", y = "# of CEOs")
```



Re-Expression to Equalize Spread Across Groups We begin by reading in the data.

```
PassiveSmoke <- read_csv("http://nhorton.people.amherst.edu/is5/data/Passive_smoke.csv")
```

```
##
## -- Column specification -----
## cols(
## cotinine = col_double(),
## smoke_exposure = col_character()
## )

# Figure 4.8, page 107
gf_boxplot(cotinine ~ smoke_exposure, data = PassiveSmoke) %>%
    gf_labs(x = "Smoke Exposure", y = "Cotinine (ng/ml)")
```

