IS5 in R: Regression Wisdom (Chapter 8)

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July 25, 2018

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 8: Regression Wisdom

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

Section 8.1: Examining Residuals

Getting the "Bends": When the Residuals Aren't Straight

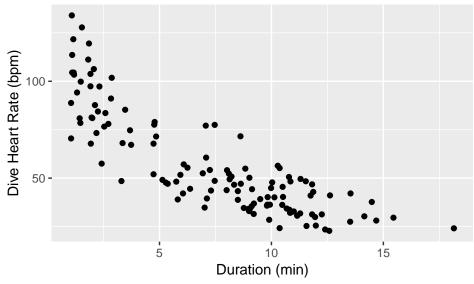
```
Penguins <- read_csv("http://nhorton.people.amherst.edu/is5/data/Penguins.csv") %>%
    clean_names()

## Parsed with column specification:
## cols(
## `Dive Heart Rate` = col_double(),
## `Depth(m)` = col_double(),
## `Duration(min)` = col_double(),
## `Bird#` = col_character()
## )
```

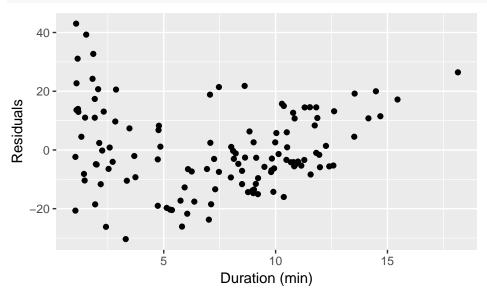
By default, read_csv() prints the variable names. These messages can be suppressed using the message=FALSE code chunk option to save space and improve readability.

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 8.1, page 234
gf_point(dive_heart_rate ~ duration_min, data = Penguins) %>%
gf_labs(x = "Duration (min)", y = "Dive Heart Rate (bpm)")
```



```
penguinlm <- lm(dive_heart_rate ~ duration_min, data = Penguins)
# Figure 8.2
gf_point(resid(penguinlm) ~ duration_min, data = Penguins) %>%
gf_labs(x = "Duration (min)", y = "Residuals")
```

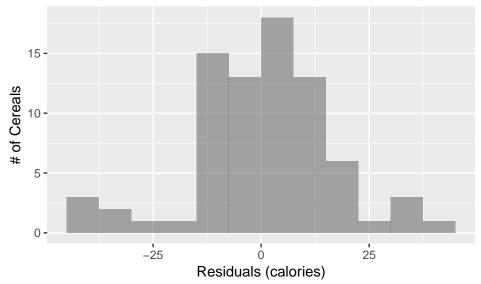


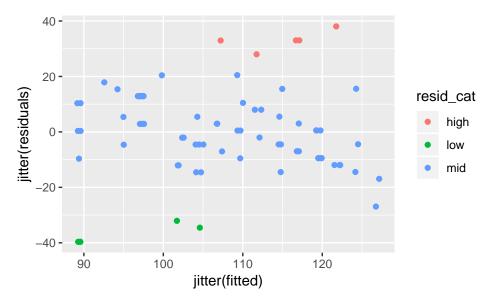
Sifting Residuals for Groups

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

```
## Parsed with column specification:
## cols(
## name = col_character(),
## mfr = col_character(),
## calories = col_integer(),
## sugars = col_integer(),
## carbo = col_double(),
## protein = col_integer(),
## fat = col_integer(),
```

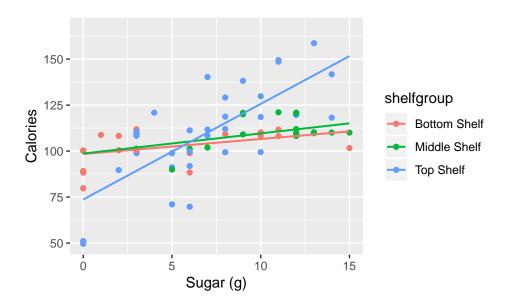
```
sodium = col_integer(),
##
     fiber = col_double(),
##
     potass = col_integer(),
##
##
     shelf = col_integer(),
##
     Middle = col_character(),
##
     shelf_1 = col_integer(),
##
     shelf_2 = col_integer(),
     shelf_3 = col_integer()
##
## )
cereallm <- lm(calories ~ sugars, data = Cereal)</pre>
# Figure 8.3, page 235
gf_histogram(~ resid(cereallm), binwidth = 7.5, center = 7.5/2) %>%
  gf_labs(x = "Residuals (calories)", y = "# of Cereals")
```





Jitter adds some random noise to allow easier observation of values that are shared by more than one type of breakfast cereal.

The recode() function allows for efficient mutation of levels.



Section 8.2: Extrapolation: Reaching Beyond the Data

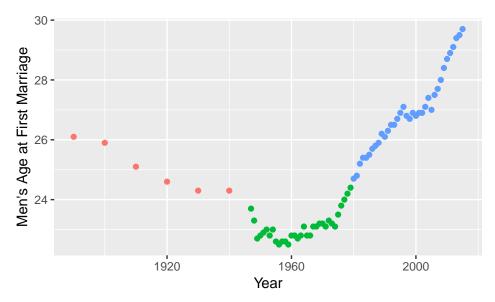
See displays on page 237 and 238.

Example 8.1: Extrapolation: Reaching Beyond the Data

```
## Parsed with column specification:
## cols(
## Year = col_integer(),
## Men = col_double(),
## Women = col_double()
## )

MarriageAge <- MarriageAge %>%
    mutate(timeperiod = ifelse(Year <= 1940, "Section1", ifelse(Year >= 1980, "Section3", "Section2")))
gf_point(Men ~ Year, color = ~ timeperiod, data = MarriageAge, ylab = "Men's Age at First Marriage") +
    guides(color = FALSE)
```

MarriageAge <- read_csv("http://nhorton.people.amherst.edu/is5/data/Marriage_age_2015.csv")

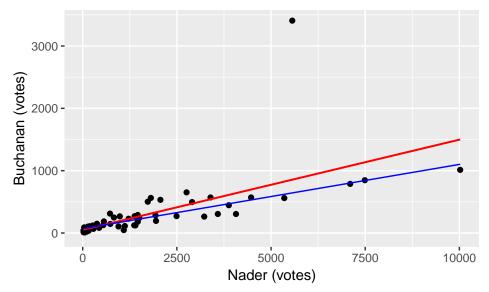


ifelse() works similarly to recode().

Section 8.3: Outliers, Leverage, and Influence

```
Election 2000 <- read csv("http://nhorton.people.amherst.edu/is5/data/Election 2000.csv")
## Parsed with column specification:
  cols(
##
     County = col_character(),
##
     Gore = col_integer(),
##
     Bush = col_integer(),
##
     Buchanan = col_integer(),
     Nader = col_integer()
##
withlm <- lm(Buchanan ~ Nader, data = Election2000)
msummary(withlm)
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 50.25627
                          51.63965
                                     0.973
## Nader
                0.14472
                           0.02076
                                     6.971 1.95e-09 ***
## Residual standard error: 343 on 65 degrees of freedom
## Multiple R-squared: 0.4278, Adjusted R-squared: 0.419
## F-statistic: 48.59 on 1 and 65 DF, p-value: 1.954e-09
withoutlm <- lm(Buchanan ~ Nader, data = filter(Election2000, Buchanan <= 3000))
msummary(withoutlm)
               Estimate Std. Error t value Pr(>|t|)
                                     4.791 1.02e-05 ***
## (Intercept) 69.52787
                          14.51176
## Nader
                0.10309
                           0.00602 17.126 < 2e-16 ***
##
## Residual standard error: 96.27 on 64 degrees of freedom
## Multiple R-squared: 0.8209, Adjusted R-squared: 0.8181
## F-statistic: 293.3 on 1 and 64 DF, p-value: < 2.2e-16
```

```
# Figure 8.10, page 241
gf_point(Buchanan ~ Nader, data = Election2000) %>%
gf_lm(color = "red") %>%
gf_labs(x = "Nader (votes)", y = "Buchanan (votes)") %>%
gf_fun(withoutlm, color = "blue") # adds line for model without outlier
```



See page 242 for example of high-leverage point.

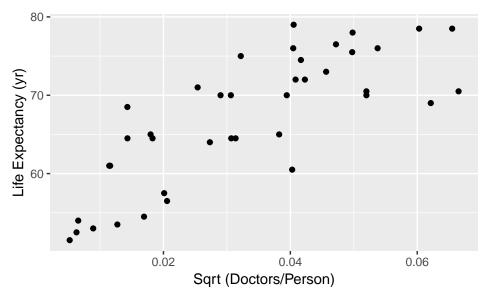
Section 8.4: Lurking Variables with Causation

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

## Parsed with column specification:
## cols(
## country = col_character(),
## `life exp` = col_double(),
## `sqrtTV/person` = col_double(),
## `sqrtDoctors/person` = col_double()
## )

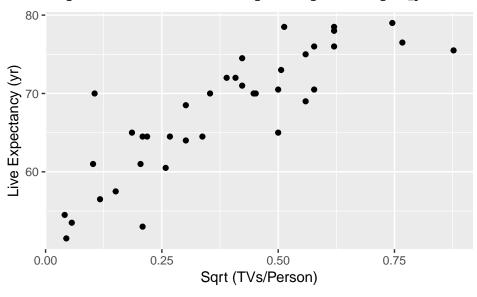
# Figure 8.13, page 243

gf_point(life_exp ~ sqrt_doctors_person, data = Doctors) %>%
    gf_labs(x = "Sqrt (Doctors/Person)", y = "Life Expectancy (yr)")
```



```
# Figure 8.14
gf_point(life_exp ~ sqrt_tv_person, data = Doctors) %>%
gf_labs(x = "Sqrt (TVs/Person)", y = "Live Expectancy (yr)")
```

Warning: Removed 2 rows containing missing values (geom_point).



Example 8.2: Using Several of These Methods Together

```
# page 244
DirtBikes <- read_csv("http://nhorton.people.amherst.edu/is5/data/Dirt_bikes_2014.csv")

## Parsed with column specification:
## cols(
## .default = col_character(),
## Year = col_integer(),
## MSRP = col_integer(),
## Displacement = col_double(),
## `Wheel Base` = col_double(),</pre>
```

```
Bore = col_double(),
##
     Stroke = col_double(),
##
     Ratio = col_double(),
##
##
     Weight = col_double(),
     Rake = col_double(),
##
##
     Trail = col_double(),
     Tank = col_double(),
##
      `Engine cooling` = col_integer()
##
## )
## See spec(...) for full column specifications.
gf_point(MSRP ~ Displacement, data = DirtBikes)
   10000 -
    7500 -
MSRP
    5000 -
    2500 -
                 100
                             200
                                          300
                                                      400
                                                                   500
                                  Displacement
bikeslm <- lm(MSRP ~ Displacement, data = DirtBikes)</pre>
gf_point(resid(bikeslm) ~ fitted(bikeslm)) %>%
  gf_labs(x = "Predicted Values", y = "Residuals")
    2000 -
    1000 -
Residuals
   -1000 -
   -2000 -
                                     7000
                                                      9000
                     5000
                                Predicted Values
```

DirtBikes %>%
filter(Cooling != "NA") %>%

```
mutate(Cooling = ifelse(Cooling == "Air-Cooled", "Air-Cooled", "LiquidCooled")) %>%
gf_point(MSRP ~ (Displacement)^(1/3), color = ~ Cooling) %>%
gf_lm()

10000-
7500-
2500-
4 5 6 7 8

(Displacement)^(1/3)
Cooling
LiquidCooled
```

Section 8.5: Working with Summary Values

65

150 -

100 -

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

## Parsed with column specification:

## cols(

## Weight = col_integer(),

## Height = col_double()

## )

# Figure 8.15, page 246

gf_point(Weight ~ Height, data = HeightsWeights) %>%

gf_lm()

250-
```

75

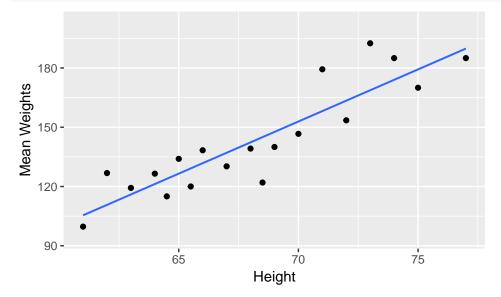
70

Height

```
# Figure 8.16
HeightWeightSum <- favstats(~ Weight | Height, data = HeightsWeights)
head(HeightWeightSum)
## Height min Q1 median Q3 max mean sd n missing</pre>
```

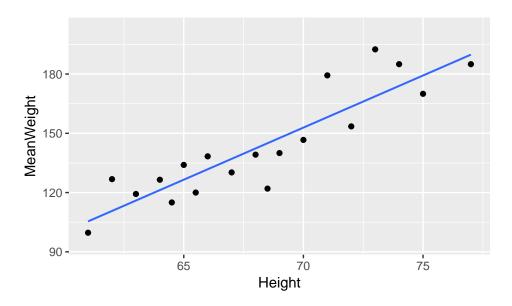
```
## 1
        61 88 89.5
                       91.0 105.50 120 99.66667 17.672955
                                                                    0
## 2
        62 107 118.0 129.0 130.00 150 126.80000 15.990622
                                                                    0
## 3
        63 105 118.5 120.0 121.50 130 119.28571 7.521398
                                                           7
                                                                    0
        64 110 112.0 122.5 129.75 180 126.50000 20.855322 10
                                                                    0
## 5
      64.5 115 115.0 115.0 115.00 115 115.00000
                                                                    0
        65 123 128.5 134.0 139.50 145 134.00000 15.556349
## 6
                                                                    0
```

```
gf_point(mean ~ as.numeric(Height), data = HeightWeightSum) %>%
    gf_lm() %>%
    gf_labs(x = "Height", y = "Mean Weights")
```



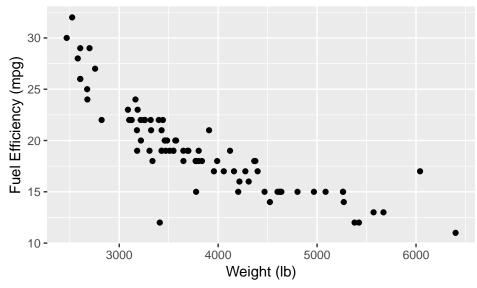
Alternately, we can use group_by() and summarise() together to find summary values of data:

```
HeightsWeights %>%
group_by(Height) %>%
summarise(MeanWeight = mean(Weight)) %>%
gf_point(MeanWeight ~ Height) %>%
gf_lm()
```

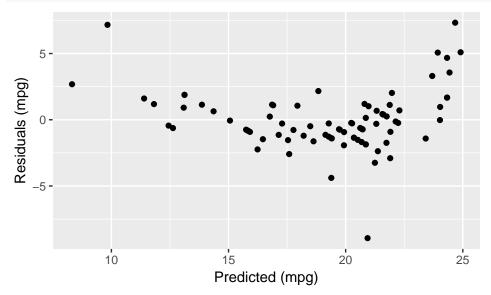


Section 8.6: Straightening Scatterplots–The Three Goals

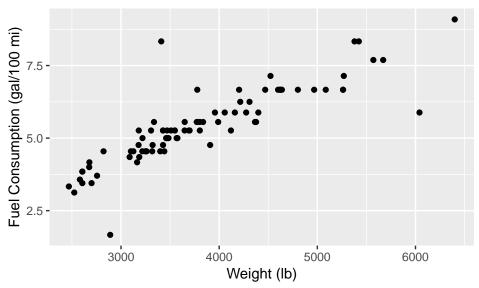
```
FuelEfficiency <- read_csv("http://nhorton.people.amherst.edu/is5/data/Fuel_efficiency.csv") %>%
  clean_names()
## Parsed with column specification:
## cols(
##
    Model = col_character(),
##
     `Eng Size` = col_double(),
##
     Cylinders = col_integer(),
     MSRP = col_integer(),
##
##
     `City Mpg` = col_integer(),
     `Highway Mpg` = col_integer(),
##
##
     Weight = col_integer(),
##
     Type = col_character(),
##
     Country = col_character()
## )
# Figure 8.17
gf_point(city_mpg ~ weight, data = filter(FuelEfficiency, city_mpg <= 40)) %>%
 gf_labs(x = "Weight (lb)", y = "Fuel Efficiency (mpg)")
```



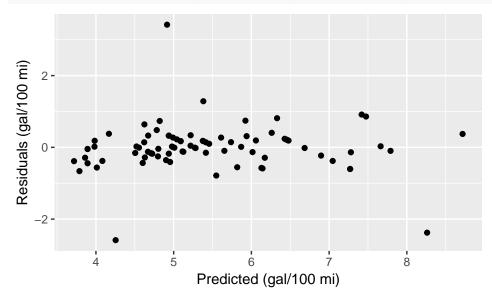
```
fuellm <- lm(city_mpg ~ weight, data = filter(FuelEfficiency, city_mpg <= 40))
gf_point(resid(fuellm) ~ fitted(fuellm)) %>%
gf_labs(x = "Predicted (mpg)", y = "Residuals (mpg)")
```



```
FuelEfficiency <- FuelEfficiency %>%
  mutate(fuel_consumption = (1/city_mpg) * 100)
# Figure 8.19, page 247
gf_point(fuel_consumption ~ weight, data = FuelEfficiency) %>%
  gf_labs(x = "Weight (1b)", y = "Fuel Consumption (gal/100 mi)")
```



```
fuellm2 <- lm(fuel_consumption ~ weight, data = FuelEfficiency)
gf_point(resid(fuellm2) ~ fitted(fuellm2)) %>%
gf_labs(x = "Predicted (gal/100 mi)", y = "Residuals (gal/100 mi)")
```



Goals of Re-Expression for Regression

```
HandDexterity <- read_csv("http://nhorton.people.amherst.edu/is5/data/Hand_dexterity.csv") %>%
   clean_names() %>%
   mutate(dominant = ifelse(dominant == 0, "Dominant hand", "Non-dominant hand")) %>%
   mutate(dominant = as.factor(dominant))

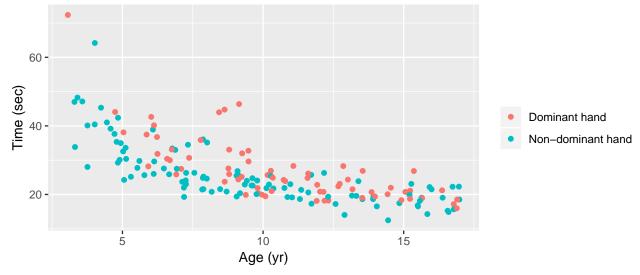
## Parsed with column specification:
## cols(
## `Time(sec)` = col_double(),
## Speed = col_double(),
## Speed = col_double(),
## Age(yr)` = col_double(),
## Dominant = col_integer(),
## Gender#` = col_integer(),
```

```
## HD = col_character(),
## `hand used` = col_character()
## )

# Figure 8.20, page 248

gf_point(time_sec ~ age_yr, color = ~ dominant, data = HandDexterity) %>%
    gf_labs(x = "Age (yr)", y = "Time (sec)", color = "")
```

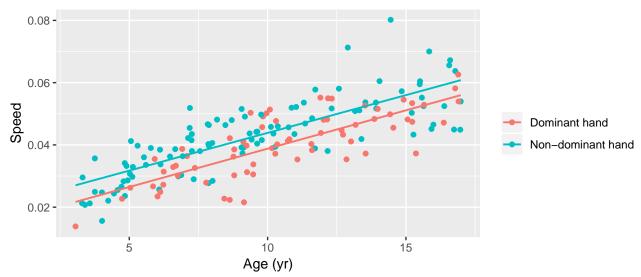
Warning: Removed 1 rows containing missing values (geom_point).



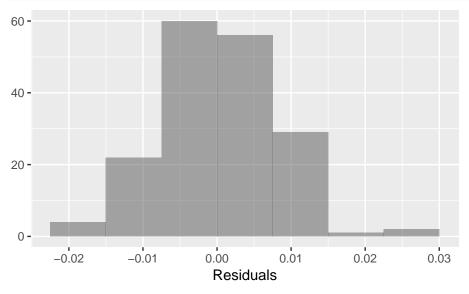
```
HandDexterity <- HandDexterity %>%
  mutate(speed = 1/time_sec)
# Figure 8.21
gf_point(speed ~ age_yr, color = ~ dominant, data = HandDexterity) %>%
  gf_lm() %>%
  gf_labs(x = "Age (yr)", y = "Speed", color = "")
```

Warning: Removed 1 rows containing non-finite values (stat_lm).

Warning: Removed 1 rows containing missing values (geom_point).



```
handlm <- lm(speed ~ age_yr, data = HandDexterity)
# Figure 8.22
gf_histogram(~ resid(handlm), binwidth = .0075, center = .0075/2) %>%
gf_labs(x = "Residuals", y = "")
```

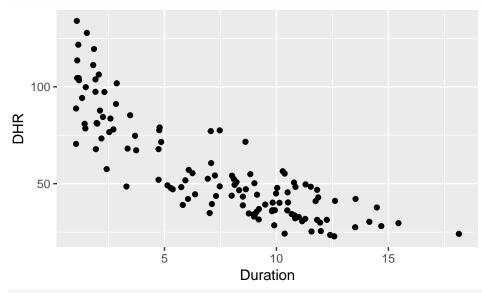


Section 8.7: Finding a Good Re-expression

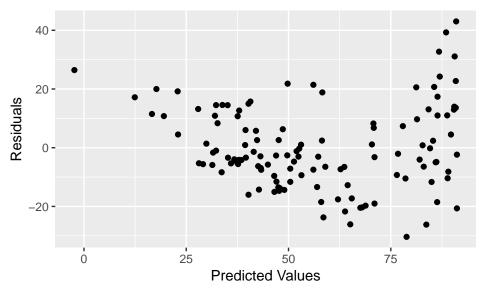
See table and Figure 8.23 on page 250.

Step-By-Step Example: Re-Expressing to Straighten a Scatterplot

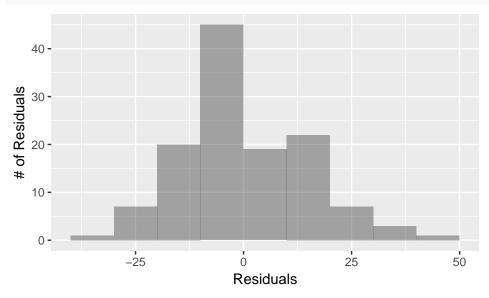
gf_point(dive_heart_rate ~ duration_min, data = Penguins, xlab = "Duration", ylab = "DHR")



```
gf_point(resid(penguinlm) ~ fitted(penguinlm), data = Penguins) %>%
gf_labs(x = "Predicted Values", y = "Residuals")
```

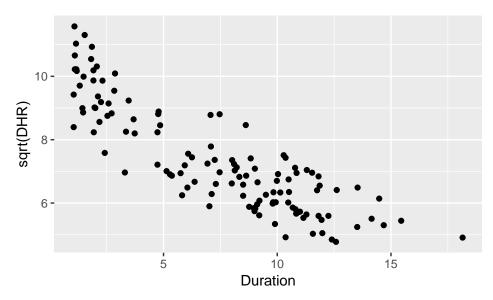


```
gf_histogram(~ resid(penguinlm), binwidth = 10, center = 5) %>%
gf_labs(x = "Residuals", y = "# of Residuals")
```



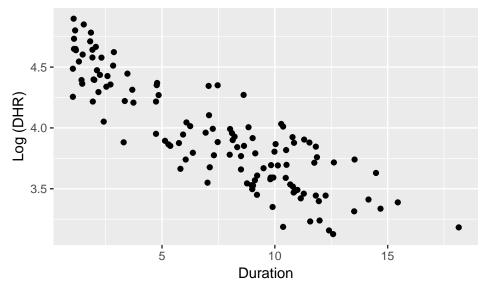
Mutating with the square root:

```
gf_point((dive_heart_rate)^(1/2) ~ duration_min, data = Penguins) %>%
gf_labs(x = "Duration", y = "sqrt(DHR)")
```

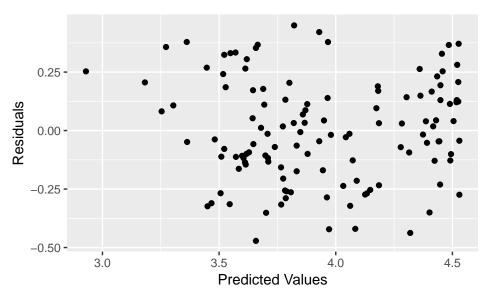


Mutating with a log:

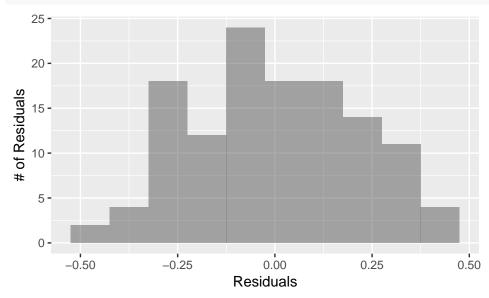
```
gf_point(log(dive_heart_rate) ~ duration_min, data = Penguins) %>%
gf_labs(x = "Duration", y = "Log (DHR)")
```



```
penguinlm2 <- lm(log(dive_heart_rate) ~ duration_min, data = Penguins)
gf_point(resid(penguinlm2) ~ fitted (penguinlm2)) %>%
gf_labs(x = "Predicted Values", y = "Residuals")
```

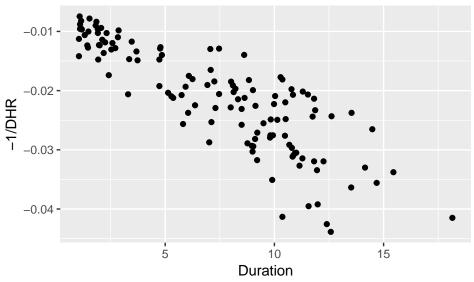


```
gf_histogram(~ resid(penguinlm2), binwidth = 0.1, center = .025) %>%
gf_labs(x = "Residuals", y = "# of Residuals")
```

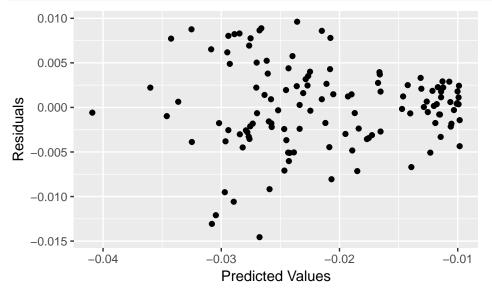


Mutating with a (negative) reciprocal:

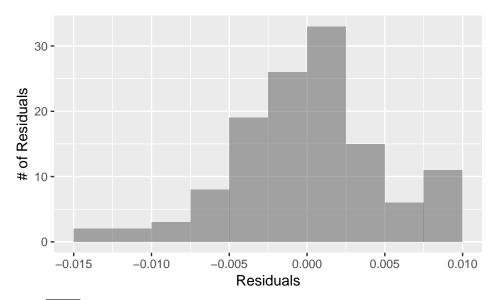
```
gf_point(-1/(dive_heart_rate) ~ duration_min, data = Penguins) %>%
gf_labs(x = "Duration", y = "-1/DHR")
```



```
penguinlm3 <- lm(-1/(dive_heart_rate) ~ duration_min, data = Penguins)
gf_point(resid(penguinlm3) ~ fitted(penguinlm3)) %>%
gf_labs(x = "Predicted Values", y = "Residuals")
```



```
gf_histogram(~ resid(penguinlm3), binwidth = 0.0025, center = 0.00125) %>%
gf_labs(x = "Residuals", y = "# of Residuals")
```



 $-1/\sqrt{DHR}$ follows the same process on pages 253-254.