

# Data Visualization

## Cohen Chapter 2

EDUC/PSY 6600

# Always plot your data first!

"Always." - Severus Snape

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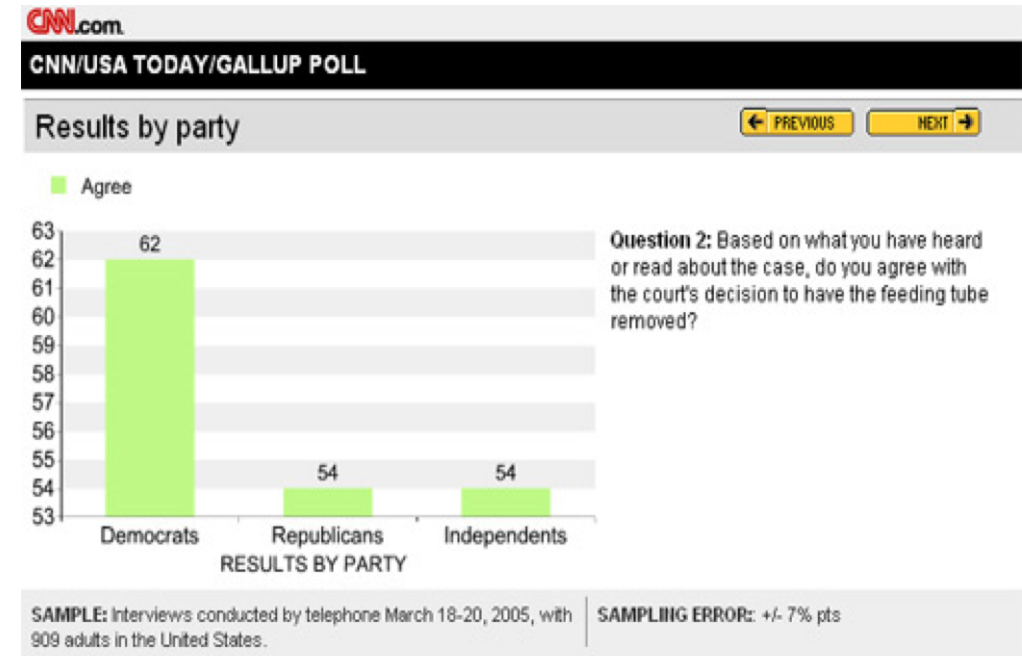
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Why?

- **Outliers** and impossible values
- Determine correct **statistical approach**
- **Assumptions** and diagnostics
- Discover new **relationships**

# The Visualization Paradox

- Often the **most informative** aspect of analysis
- **Communicates** the "data story" the best
- Most abused area of quantitative science
- Figures can be *very* **misleading**

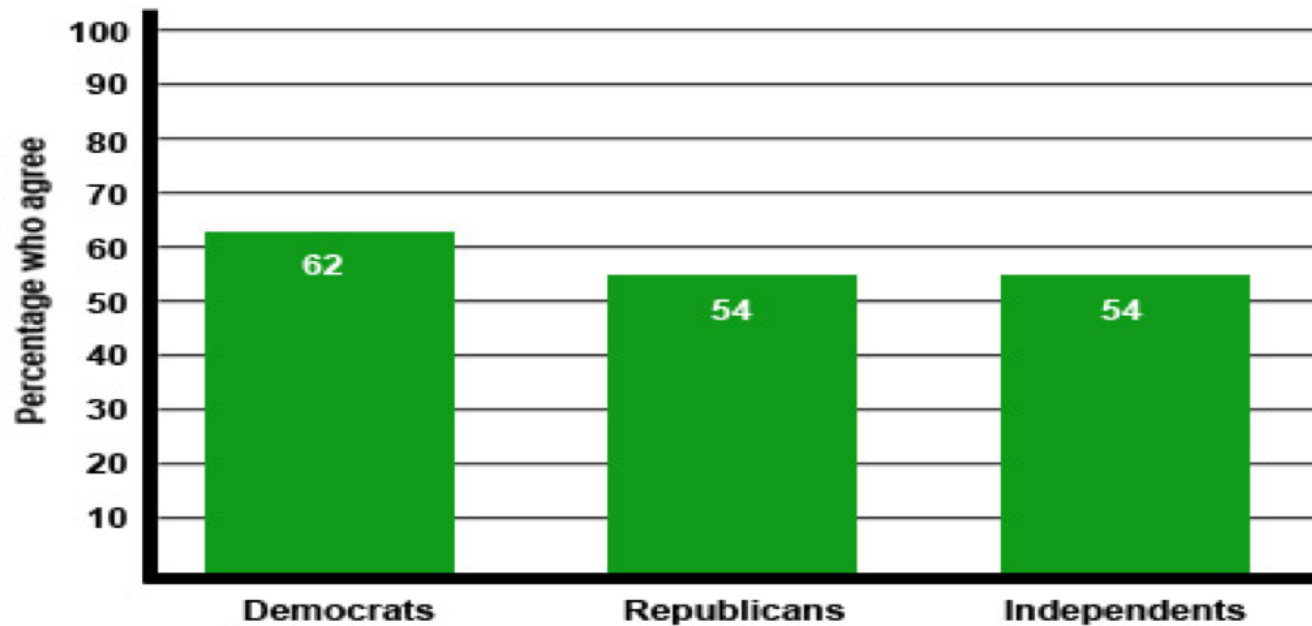


Misleading Graphs

# Much better

**RESULTS BY PARTY: CNN/USA Today/Gallup Poll**  
Margin of error: +/- 7%

**Question 2:** Based on what you have heard or read about the case, do you agree with the court's decision to have the feeding tube removed?



# Keys to Good Viz's

- Graphical method should match level of measurement
- Label all axes and include figure caption
- Simplicity and clarity
- Avoid of 'chartjunk'

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- Graphical method should match level of measurement
- Label all axes and include figure caption
- Simplicity and clarity
- Avoid of 'chartjunk'
- Unless there are 3 or more variables, avoid 3D figures (and even then, avoid it)
- Black & white, grayscale/pattern fine for most simple figures

# Data Visualizations

Takes practice -- try a bunch of stuff



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## Resources

- Edward Tufte's books
- "R for Data Science" by Grolemund and Wickham
- "Data Visualization for Social Science" by Healy

# Frequency Distributions

## Counting the number of occurrences of unique events

- Categorical or continuous
- just like with `tableF()` and `table1()`

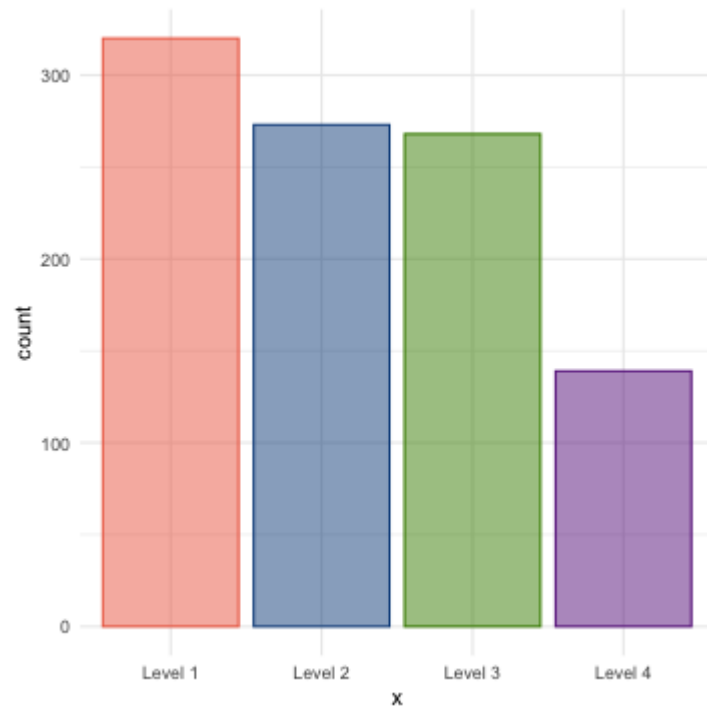
Can see **central tendency** (continuous data) or **most common value** (categorical data)

Can see **range and extremes**

x	Freq	CumFreq	Percent	CumPerc	Valid	CumValid
1	248	248	24.80%	24.80%	25.46%	25.46%
2	254	502	25.40%	50.20%	26.08%	51.54%
3	244	746	24.40%	74.60%	25.05%	76.59%
4	228	974	22.80%	97.40%	23.41%	100.00%
Missing	26	1000	2.60%	100.00%		

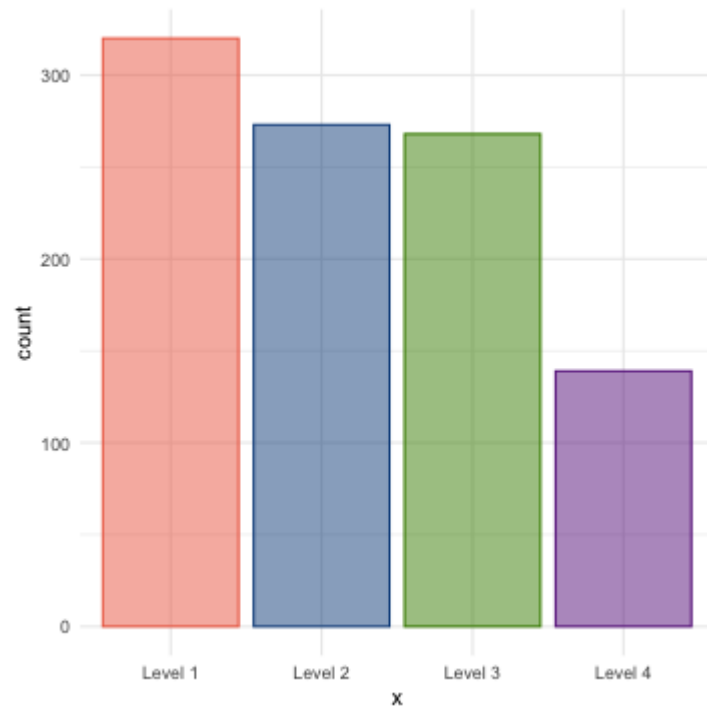
# Frequencies and Viz's Together ❤️

## Bar Graph

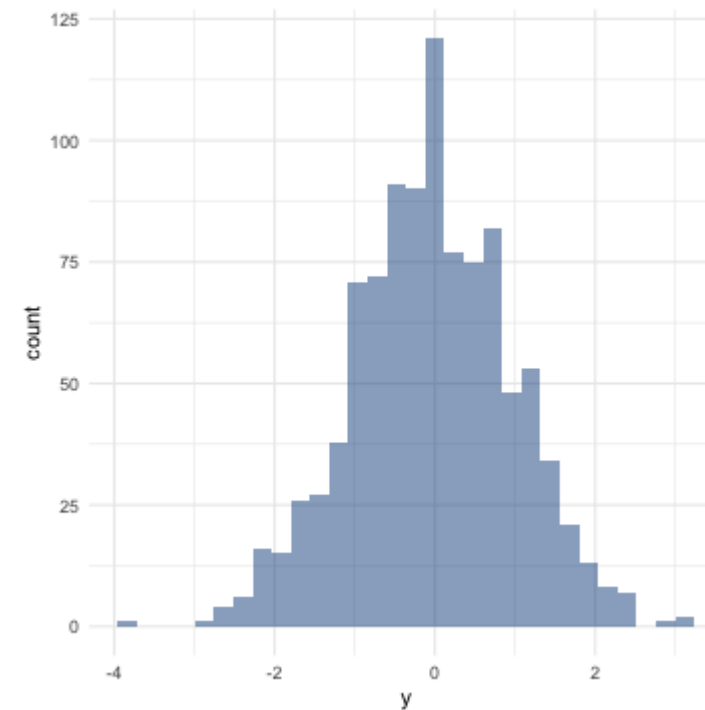


# Frequencies and Viz's Together ❤️

## Bar Graph



## Histogram



# What does DISTRIBUTION mean?

The way that the data points are scattered

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The way that the data points are scattered

## For Continuous

- General shape
- Exceptions (outliers)
- Modes (peaks)
- Center & spread (chap 3)
- Histogram

## For Categorical

- Counts of each
- Percent or Rate (adjusts for an 'out of' to compare)
- Bar chart
- Pie chart - avoid!

Let's Apply This To the Inho Dataset

# Reminder

## Key

Sub\_num: arbitrary ID number for each participant.

Gender: 1 = Female; 2 = Male.

Major: 1 = Psychology; 2 = Premed; 3 = Biology; 4 = Sociology; 5 = Economics.

Reason: 1 = Program requirement; 2 = Personal interest; 3 = Advisor recommendation.

Exp\_cond: 1 = Easy; 2 = Moderate; 3 = Difficult; 4 = Impossible.

Coffee: 0 = not a regular coffee drinker; 1 = regularly drinks coffee.

Num\_cups = number of cups of coffee drunk prior to the experiment on the same day.

Phobia: 0 = No phobia to 10 = Extreme phobia.

Prevmath = Number of math courses taken prior to statistics course.

Mathquiz = Score on Math Background Quiz (a blank for this value indicates that a student did not take the quiz).

Statquiz = Score on 10-question stats quiz given one week before the experiment.

Exp\_sqz = Score on stats quiz given as part of the experiment (number correct, including the 11th question).

HR\_base = Baseline heart rate (in beats per minute).

HR\_pre = Prequiz heart rate.

HR\_post = Postquiz heart rate.

Anx\_base = Baseline anxiety score.

Anx\_pre = Prequiz anxiety score.

Anx\_post = Postquiz anxiety score.



# Read in the Data

```
library(tidyverse)  # the easy button
library(rio)        # read in Excel files
library(furniture)  # nice tables

data_raw <- rio::import("Ihno_dataset.xls") %>%
  dplyr::rename_all(tolower)           # converts all variable names to lower case
```

# Read in the Data

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```

## And Clean It

```
data_clean <- data_raw %>%
  dplyr::mutate(majorF = factor(major,
                                levels= c(1, 2, 3, 4, 5),
                                labels = c("Psychology", "Premed",
                                           "Biology", "Sociology",
                                           "Economics")))) %>%
  dplyr::mutate(coffeeF = factor(coffee,
                                  levels = c(0, 1),
                                  labels = c("Not a regular coffee drinker",
                                             "Regularly drinks coffee"))))
```

# Frequency Distributions

```
data_clean %>%  
  furniture::tableF(majorF)
```

```
##  
##  
## majorF      Freq CumFreq Percent CumPerc  
## Psychology  29    29      29.00%  29.00%  
## Premed      25    54      25.00%  54.00%  
## Biology     21    75      21.00%  75.00%  
## Sociology   15    90      15.00%  90.00%  
## Economics  10   100      10.00% 100.00%  
##
```

```
data_clean %>%  
  furniture::tableF(phobia)
```

```
##  
##  
## phobia  Freq CumFreq Percent CumPerc  
## 0       12    12      12.00%  12.00%  
## 1       15    27      15.00%  27.00%  
## 2       12    39      12.00%  39.00%  
## 3       16    55      16.00%  55.00%  
## 4       21    76      21.00%  76.00%  
## 5       11    87      11.00%  87.00%  
## 6        1    88       1.00%  88.00%  
## 7        4    92       4.00%  92.00%  
## 8        4    96       4.00%  96.00%  
## 9        1    97       1.00%  97.00%  
## 10       3   100       3.00% 100.00%  
##
```

# Frequency Viz's

For viz's, we will use `ggplot2`

This provides the most powerful, beautiful framework for data visualizations

# Frequency Viz's

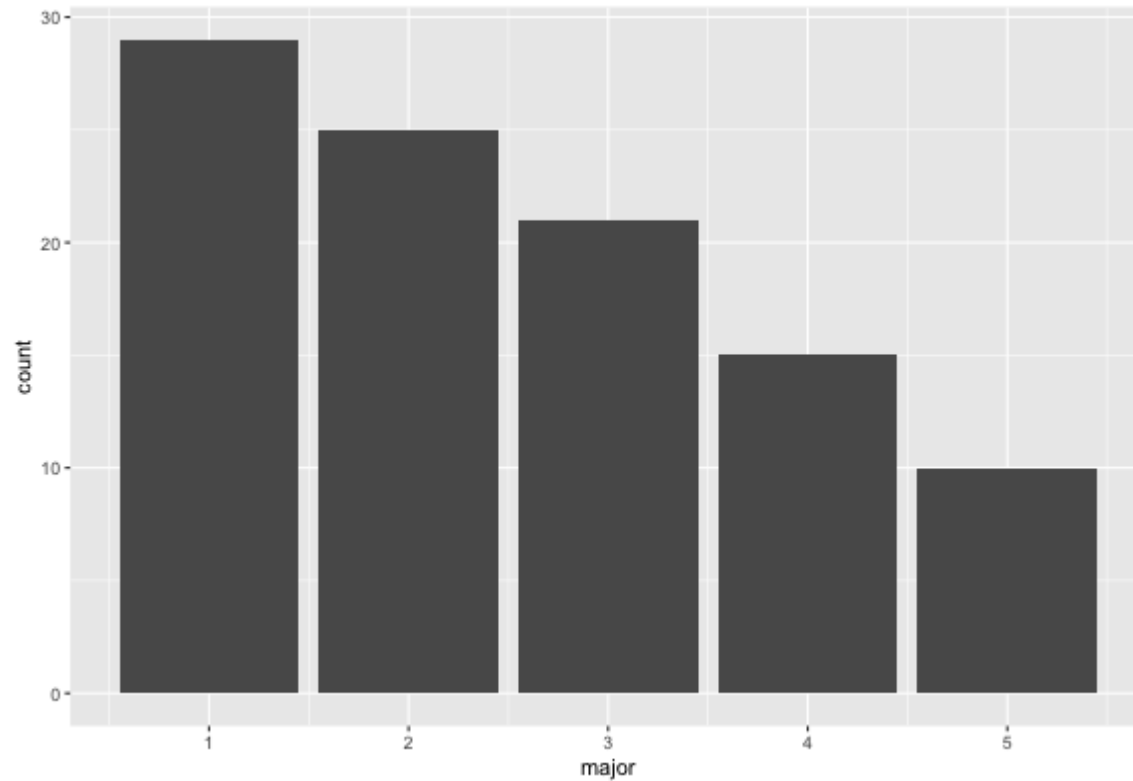
For viz's, we will use `ggplot2`

This provides the most powerful, beautiful framework for data visualizations

- It is built on making **layers**
- Each plot has a **"geom"** function
  - e.g. `geom_bar()` for bar charts, `geom_histogram()` for histograms, etc.

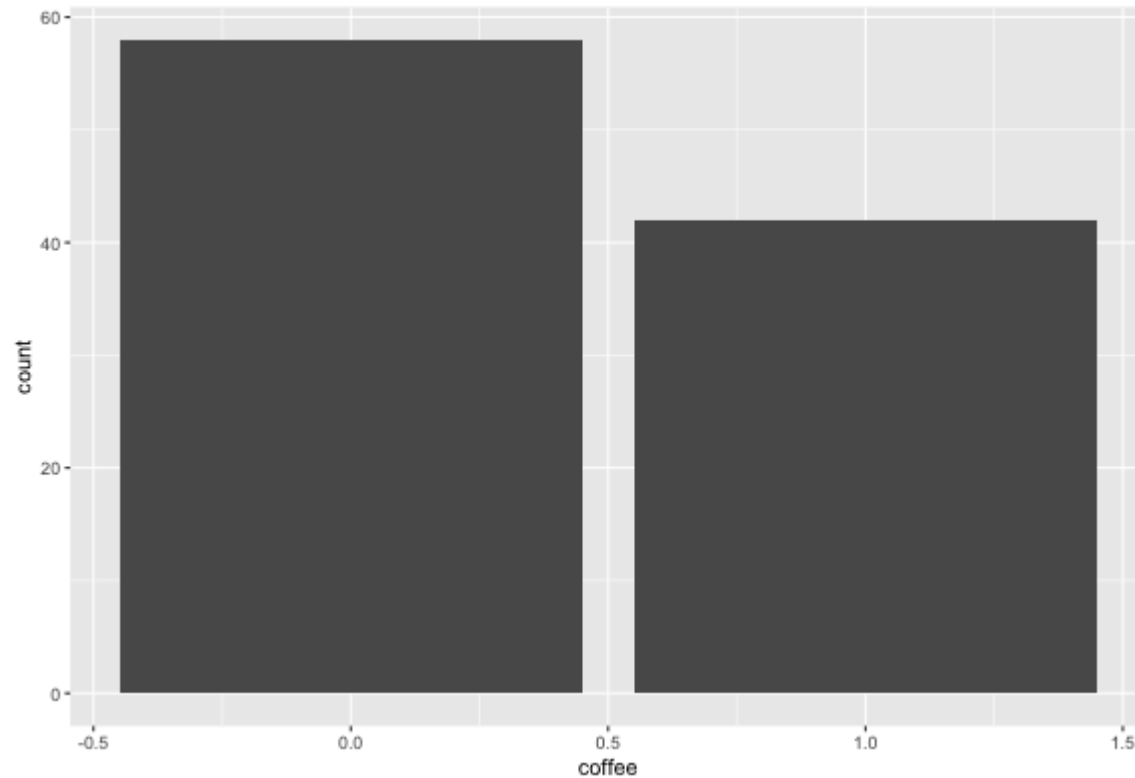
# Bar Charts

```
data_clean %>%  
  ggplot(aes(major)) +  
  geom_bar()
```



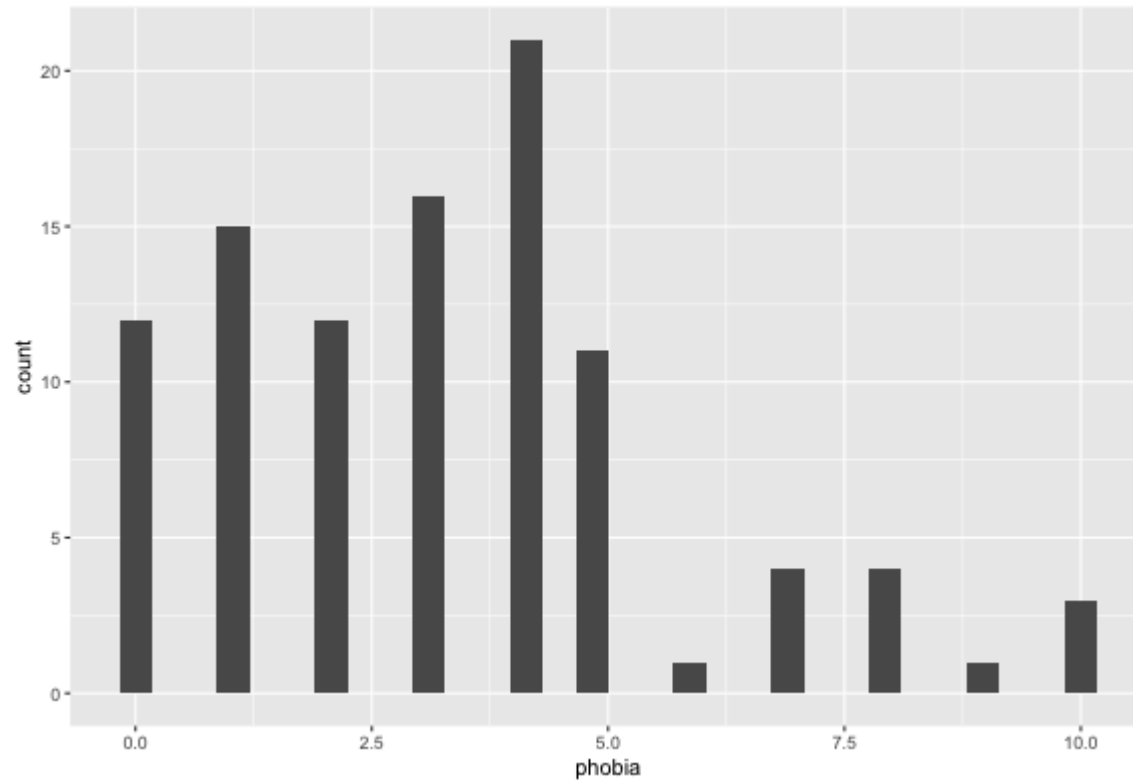
# Bar Charts

```
data_clean %>%  
  ggplot(aes(coffee)) +  
  geom_bar()
```



# Histograms

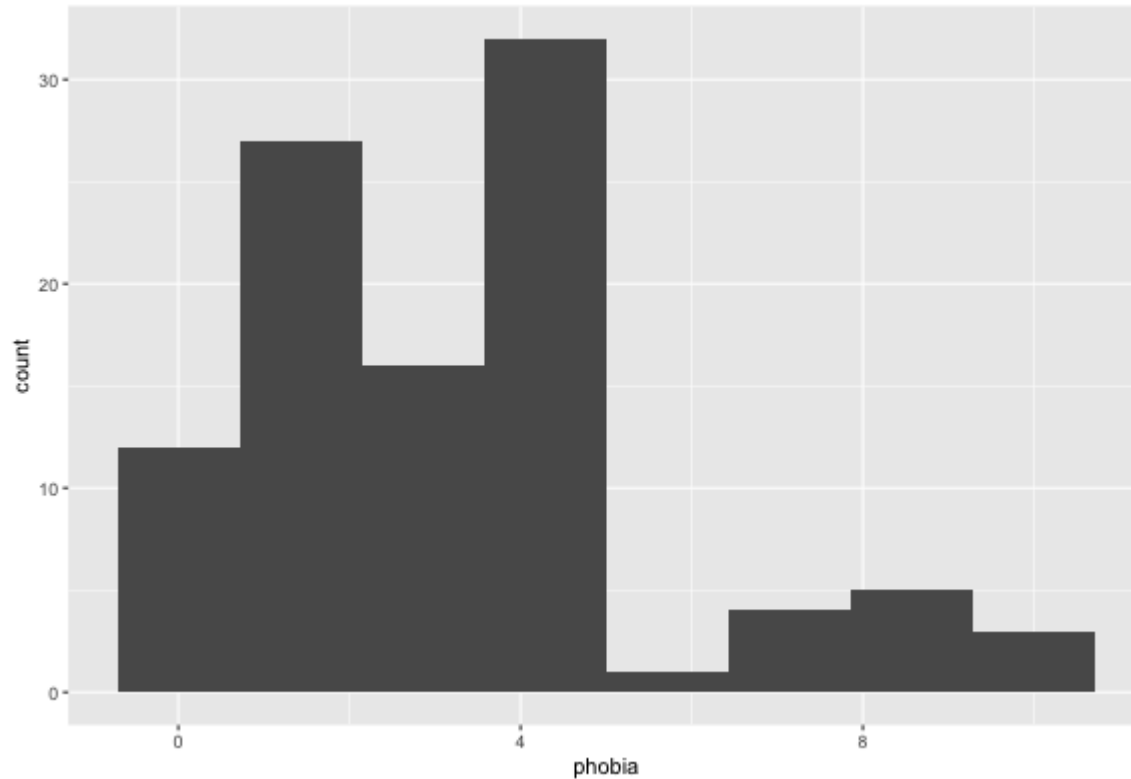
```
data_clean %>%  
  ggplot(aes(phobia)) +  
  geom_histogram()
```





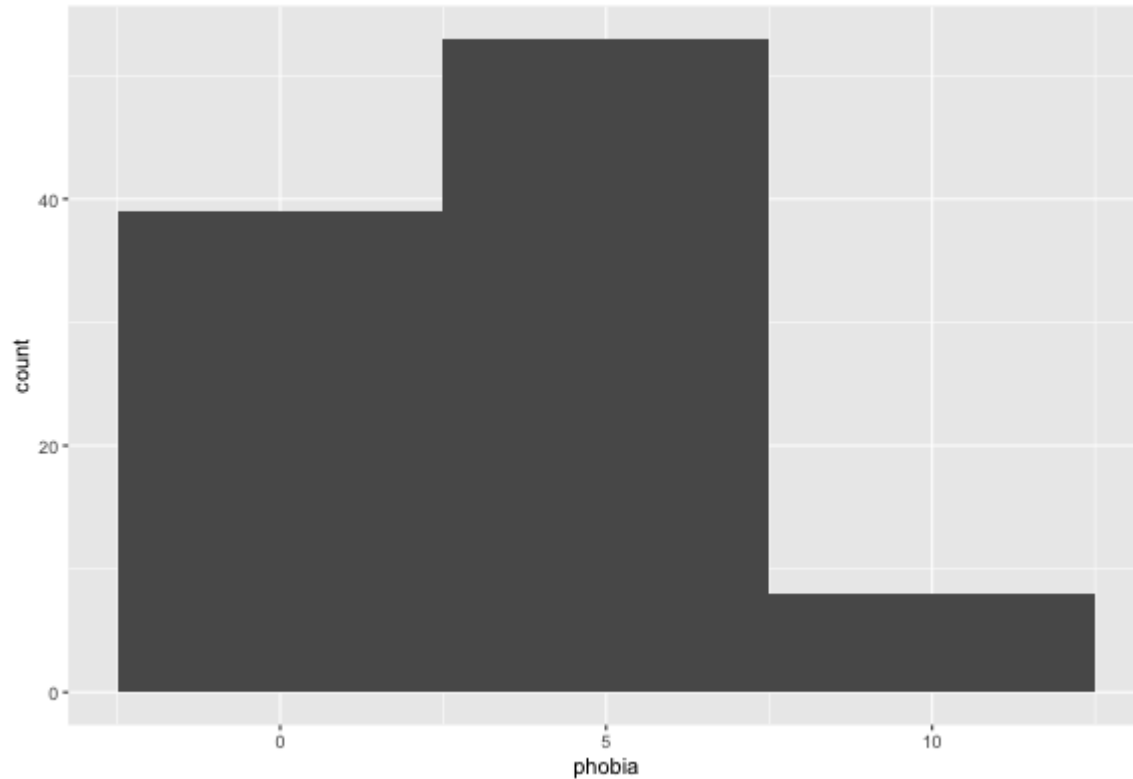
# Histograms (change number of bins)

```
data_clean %>%  
  ggplot(aes(phobia)) +  
  geom_histogram(bins = 8)
```



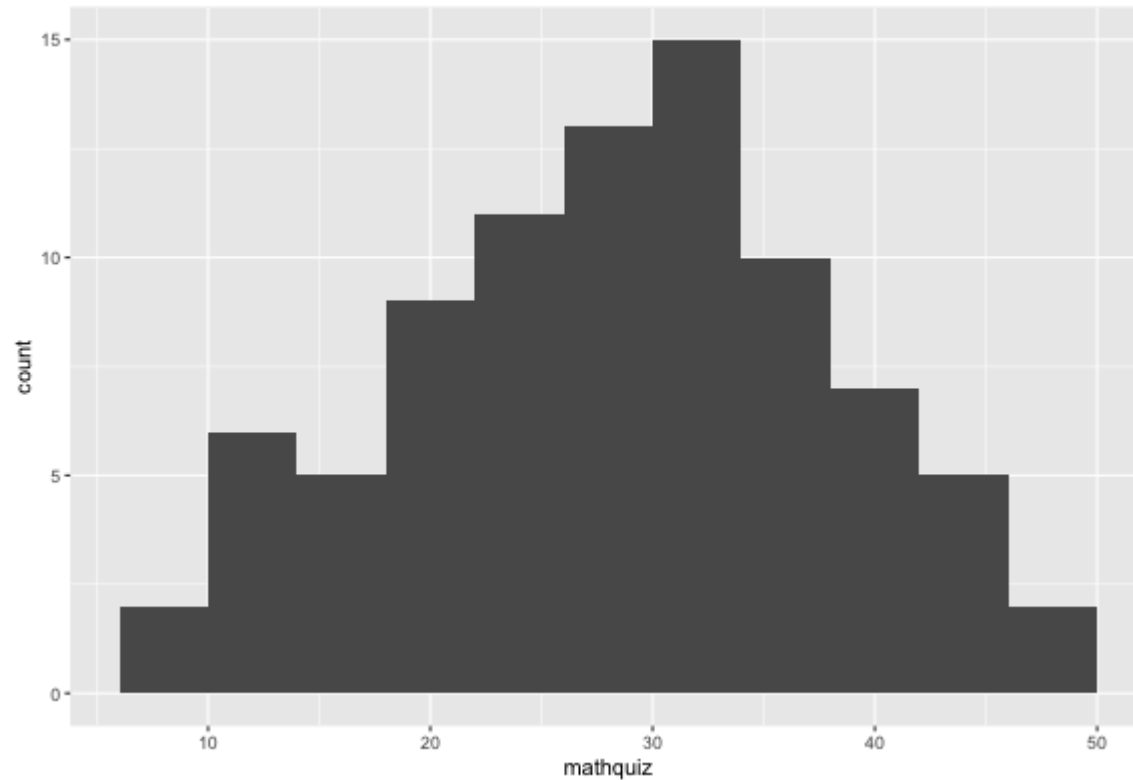
# Histograms (change bins to size 5)

```
data_clean %>%  
  ggplot(aes(phobia)) +  
  geom_histogram(binwidth = 5)
```



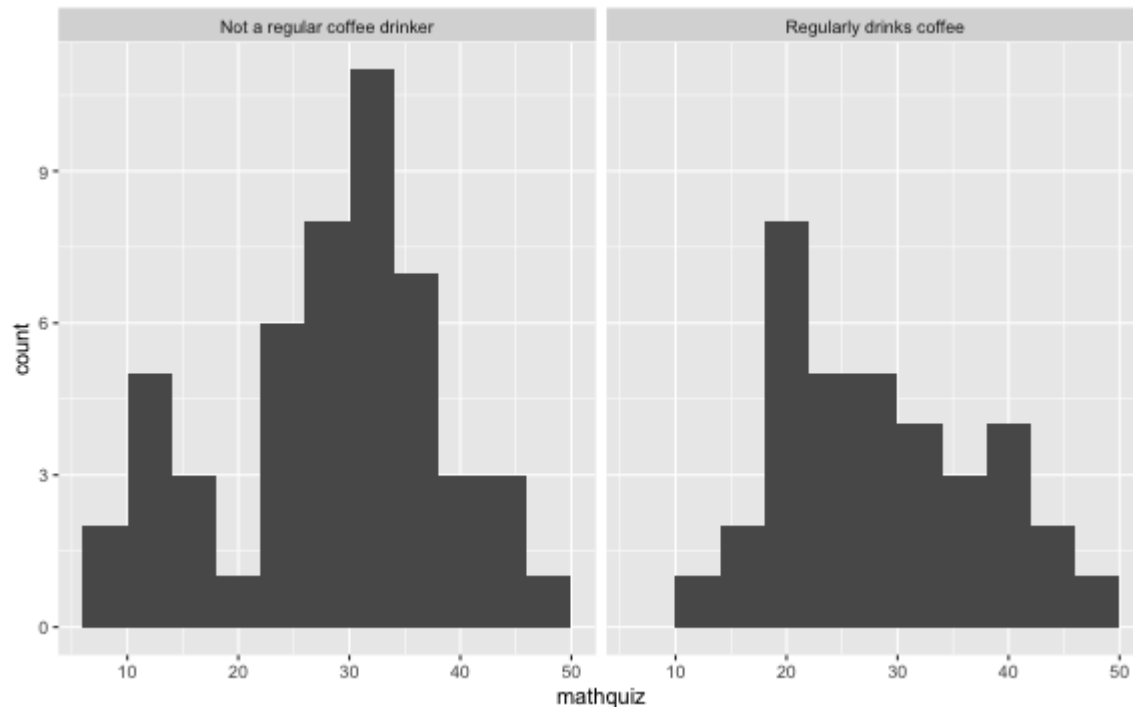
# Histograms

```
data_clean %>%  
  ggplot(aes(mathquiz)) +  
  geom_histogram(binwidth = 4)
```



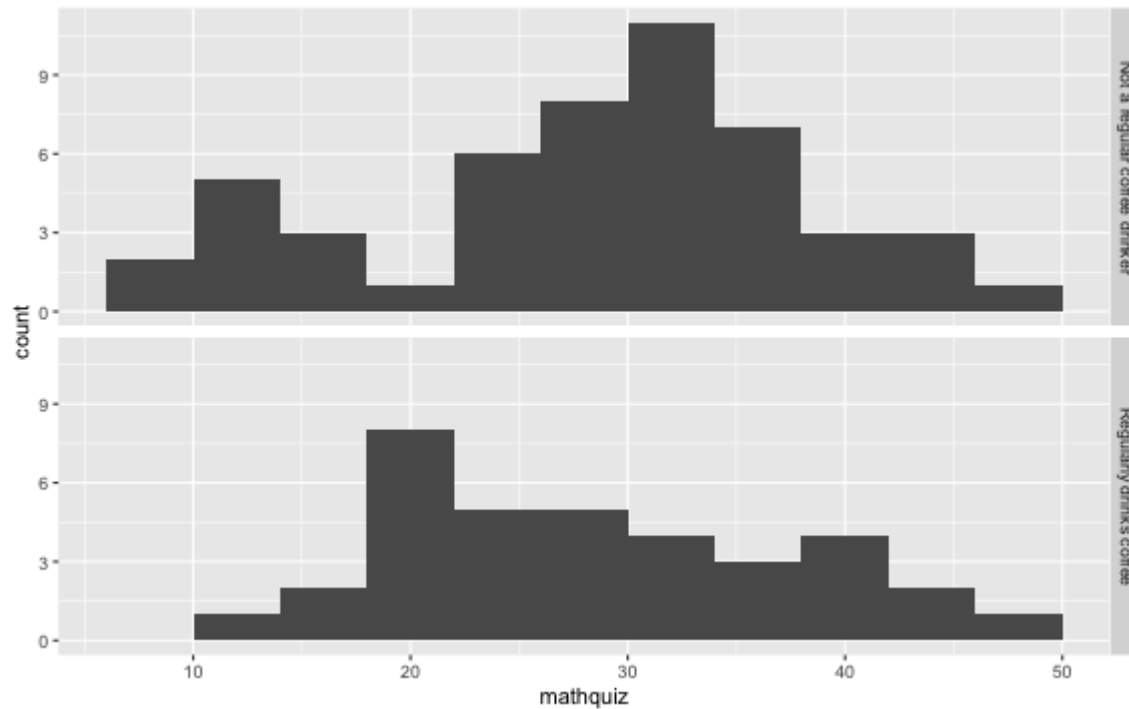
# Histograms -by- a Factor (columns)

```
data_clean %>%  
  ggplot(aes(mathquiz)) +  
  geom_histogram(binwidth = 4) +  
  facet_grid(. ~ coffeeF)
```



# Histograms -by- a Factor (rows)

```
data_clean %>%  
  ggplot(aes(mathquiz)) +  
  geom_histogram(binwidth = 4) +  
  facet_grid(coffeeF ~ .)
```



# Deciles (break into 10% chunks)

```
data_clean %>%  
  dplyr::pull(statquiz) %>%  
  quantile(probs = c(.10, .20, .30, .40, .50, .60, .70, .80, .90))
```

```
## 10% 20% 30% 40% 50% 60% 70% 80% 90%  
## 4.0 6.0 6.0 7.0 7.0 8.0 8.0 8.0 8.1
```

# Deciles - with missing values

```
data_clean %>%  
  dplyr::pull(mathquiz) %>%  
  quantile(probs = c(.10, .20, .30, .40, .50, .60, .70, .80, .90))
```

```
Error in quantile.default(., probs = c(0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, : missing  
values and NaN's not allowed if 'na.rm' is FALSE
```

# Deciles - `na.rm = TRUE`

```
data_clean %>%  
  dplyr::pull(mathquiz) %>%  
  quantile(probs = c(.10, .20, .30, .40, .50, .60, .70, .80, .90),  
           na.rm = TRUE)
```

```
## 10% 20% 30% 40% 50% 60% 70% 80% 90%  
## 15.0 21.0 25.2 28.0 30.0 32.0 33.8 37.2 41.0
```



# Quartiles (break into 4 chunks)

```
data_clean %>%  
  dplyr::pull(statquiz) %>%  
  quantile(probs = c(0, .25, .50, .75, 1))
```

```
##    0%   25%   50%   75%  100%  
##     1     6     7     8    10
```

# Percentiles

```
data_clean %>%  
  dplyr::pull(statquiz) %>%  
  quantile(probs = c(.01, .05, .173, .90))
```

```
##      1%      5% 17.3%   90%  
##  2.98  3.00  5.00  8.10
```

# Questions?

# Next Topic

## Center and Spread