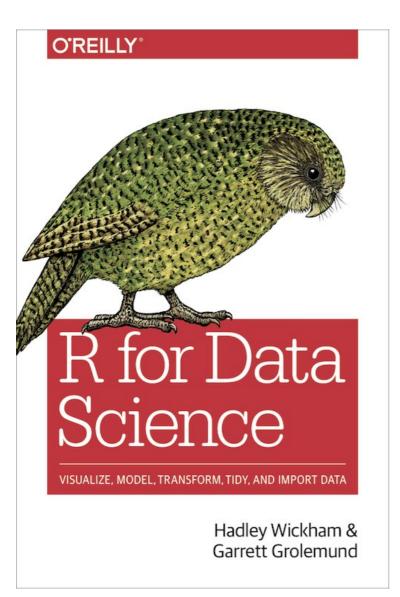
Data Analytics CS301 Exploratory Data Analysis (Continued)

Week 5: 12th Feb
Spring 2020
Oliver BONHAM-CARTER

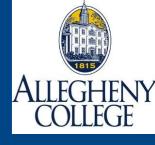
Where in the Web? Where in the Book?





- Note the chapter differences!
- Book:
 - Chap 5: Exploratory DataAnalysis
- Web:
 - http://r4ds.had.co.nz/ exploratory-dataanalysis.html
 - Chap 7: Exploratory Data Analysis













Missing Data Entries

- Missing data in R appears as NA.
- NA is not a string or a numeric value, but an indicator of missing data.
- Let's create vectors with missing values to test

```
library(tidyverse)
library(tibble)
x1 <- c(1, 4, 3, NA, 7)
x2 <- c("a", "B", NA, "NA")
is.na(x1)
is.na(x2)
```

Spot missing data

Missing Data Entries



- What to do when elements of your data go missing?
- Why not just DROP the ENTIRE ROW, as well as to drop all the value contained by its other variables as well??

diamonds2 <- diamonds %>% filter(between(y, 3, 20))

This is a shortcut for $y \ge 3 \& y \le 20$

View(diamonds2)

compare to the the size of original dataset

View(diamonds)

Note: Good data may have been lost by dropping rows.



IfElse(): Condition Statement

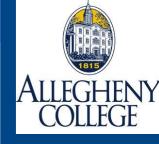
$$y = ifelse(y < 3 | y > 20, NA, y)$$

- Function
- Test Condition
- If True, then assign this
- If False, then assign this



Data: Diamond

```
# The book recommends to mark the data as
bad or missing.
diamonds2 <- diamonds %>%
 mutate(y = ifelse(y < 3 | y > 20, NA, y))
# syntax: ifelse(test, yes, no)
# Inspect each value of y. If the y is not
between 3 and 20, then y = NA, else y = y
```



We Plot All Non-NA Values

```
# Missing, outliers values marked as NA
ggplot(data = diamonds2, mapping =
aes(x = x, y = y)) + geom point()
# compared to, no removed missing or outlier
values
ggplot(data = diamonds, mapping =
aes(x = x, y = y)) + geom point()
```

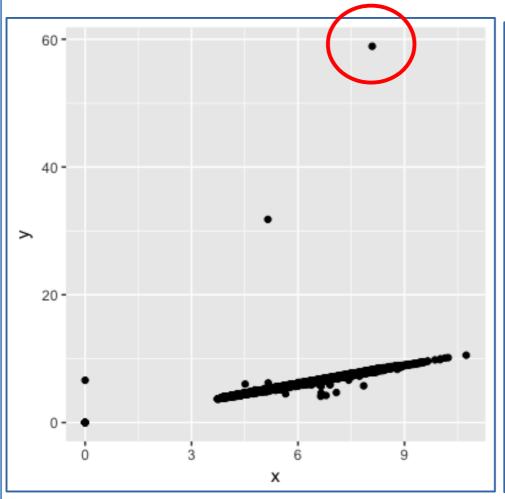


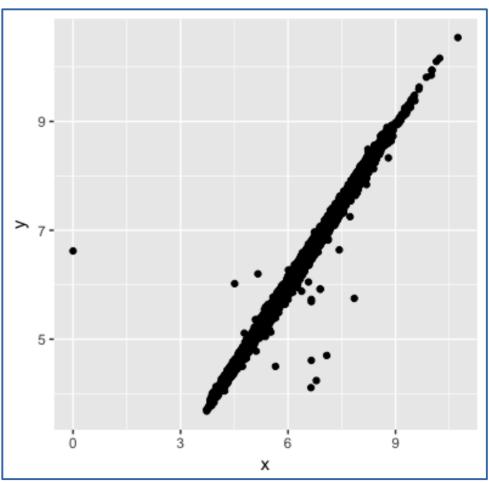
Missing Values, continued

```
# remove the outliers for y (I.e., y<3 and y >20)
library(tidyverse)
diamonds2 <- diamonds %>% mutate(y = ifelse(y <
3 | y > 20, NA, y))
ggplot(data = diamonds2, mapping = aes(x = x, y = y)) + geom_point()
```

Add the ifelse() code directly to your other code.

Trimmed Data, Slightly Different Plot...





Left: WITH outliers

Above: NO outliers



Data: Diamond

Can you use the below code to further trim outliers or missing data?

Plot your new graphic after using ifelse()

diamonds3 <- diamonds %>%

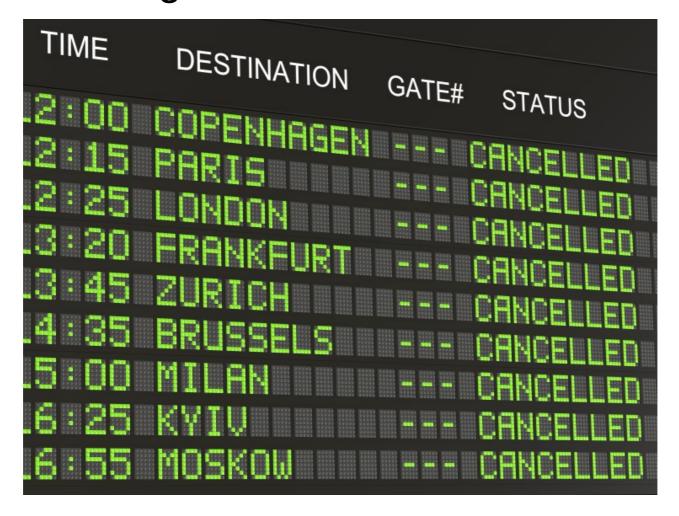
mutate(y = ifelse(y < ## | y > ##, NA, y))

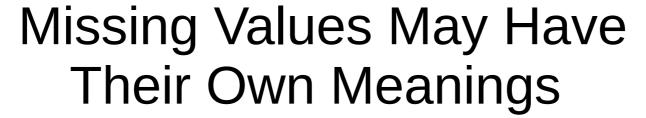






 Does missing flight arrival-time data indicate canceled flights?







```
# install the flights data, if necessary.
install.packages("nycflights13")
library(tidyverse, nycflights13)
flights <- nycflights13::flights
View(flights)
# Where are the missing values
flights$dep time
```

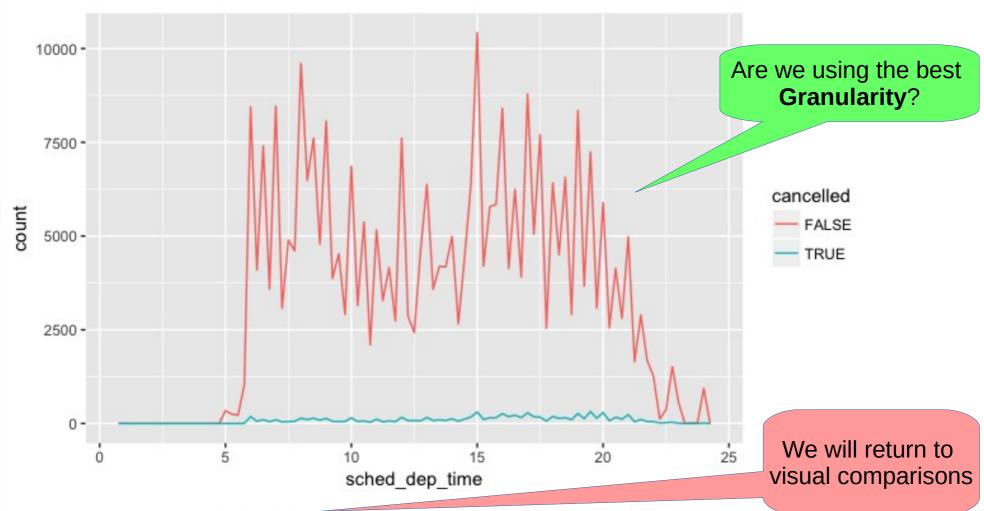


The Distribution of a Continuous Variable, Aggregated By a Categorical variable

```
# compare the scheduled departure times for cancelled and non-
cancelled times
flights %>%
 mutate(
  cancelled = is.na(dep time),
  sched hour = sched dep time %/% 100,
  sched min = sched dep time %% 100,
  sched dep time = sched hour + sched min / 60
 ) %>%
 ggplot(mapping = aes(sched_dep_time)) +
 geom freqpoly(mapping = aes(colour = cancelled),
 binwidth = 1/4)
```

ALLEGHENY COLLEGE

Potential Pitfalls in Theory



- We get an slight idea of when cancellations happen
- Many more non-cancelled flights than cancelled flights: does the business side of flying introduce a bias for not-canceling flights?



Covariation

covariance



co·var·i·ance

/ˌkōˈverēəns/ ••)

noun

- 1. MATHEMATICS the property of a function of retaining its form when the variables are linearly transformed.
- 2. STATISTICS the mean value of the product of the deviations of two variates from their respective means.
- **Covariation** is the tendency for the values of two or more variables to vary together in a related way.
- Study covariation by visualizing relationships between two or more variables.
- Pay attention to your variables to known how best to visualize these variables



Covariation

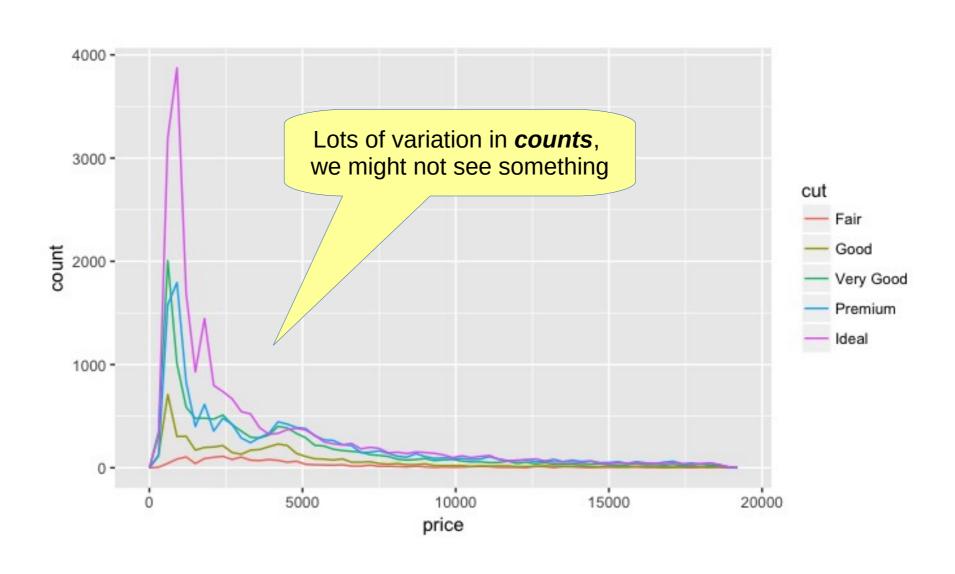
 How do the prices of diamonds vary with quality?

```
# Plot the count of each each cut quality according to price.
```

```
ggplot(data = diamonds, mapping = aes(x = price)) + geom_freqpoly(mapping = aes(colour = cut), binwidth = 500)
```



The Plot of the Diamond Counts

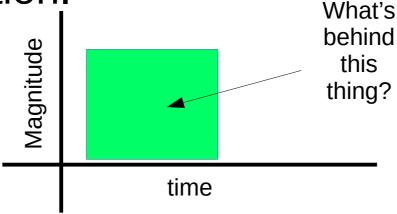


This Plot May Make It Hard To See The Phenomenon



- The counts variable seems to have values from all over the range.
- This is noise in our plot

 If one group is much smaller than the others, then it is hard to see the differences in its distribution.





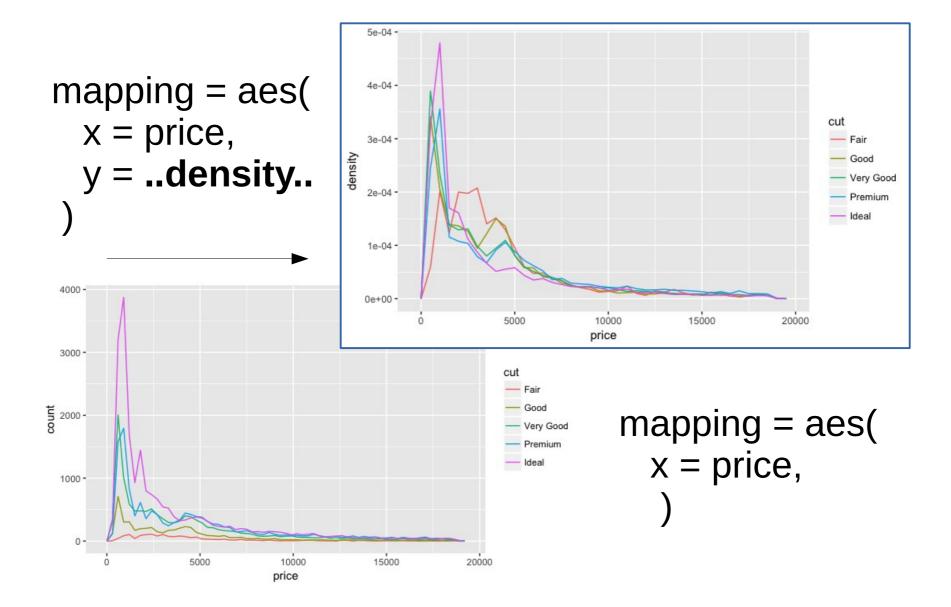
Let's Change Our Plotting

```
# Does a histogram help?
ggplot(diamonds) + geom_bar(mapping = aes(x = cut))
#Note: Density, is the count standardised so that the
area under each frequency polygon is one.
# change the axis to see behind them.
ggplot(data = diamonds, mapping = aes(x = price, y)
= ..density..)) + geom freqpoly(mapping = aes(colour
= cut), binwidth = 500)
```

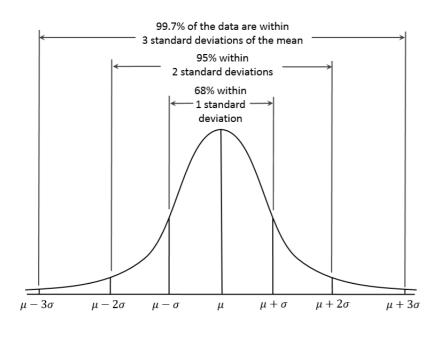
Normalize Your View!



Different Plots



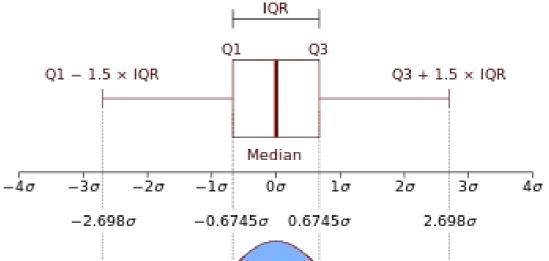
ALLEGHENY COLLEGE



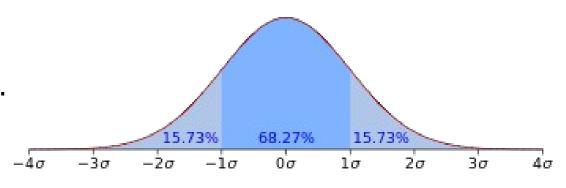
Box Plots

24.65%

 -1σ



For the normal distribution, the values less than one standard deviation away from the mean account for 68.27% of the set; while two standard deviations from the mean account for 95.45%; and three standard deviations account for 99.73%.



50%

24.65%

 2σ

 3σ

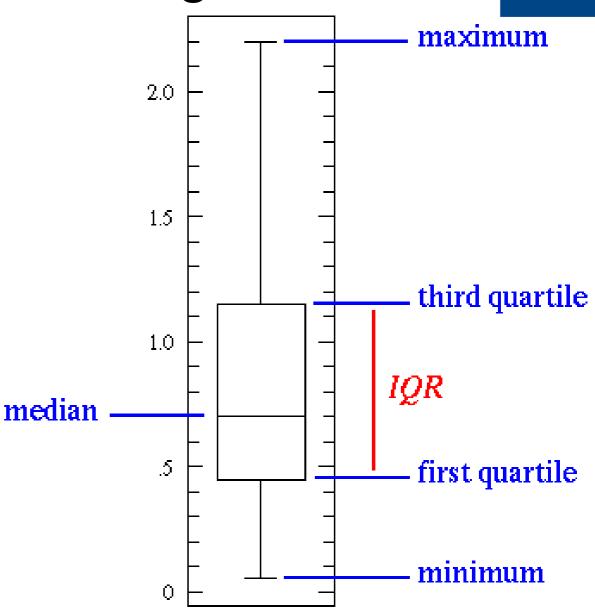
 4σ

 1σ



Explore Data Using Box Plots

Standardized way of displaying the distribution of data based on the five number summary: minimum, first quartile, median, third quartile, and maximum





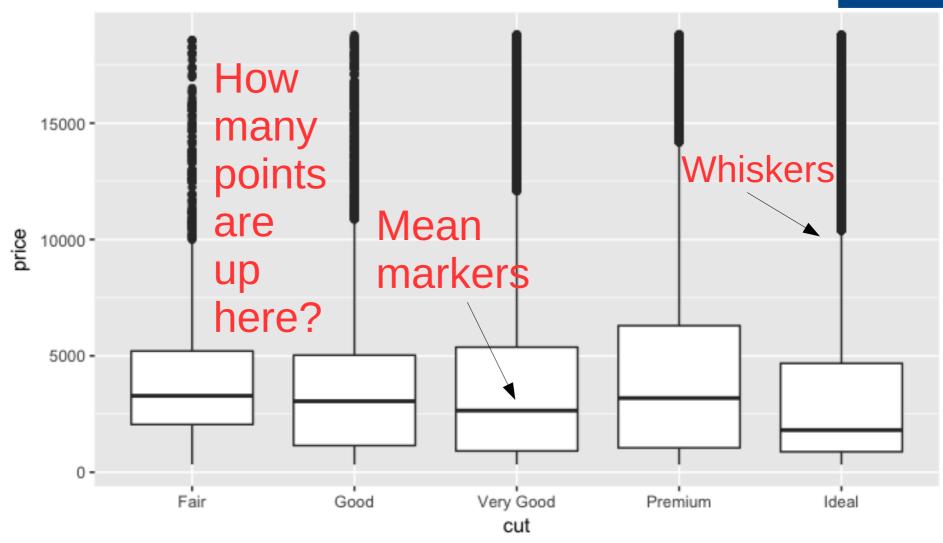
Explore Data Using Box Plots

Make a box plot to describe covariance between cut and price.

```
ggplot(data = diamonds, mapping = aes(x = cut, y = price)) + geom_boxplot()
```



Explore Data Using Box Plots





Box Plots: Pros and Cons

- Much less information about the *cut* distribution.
- Boxplots much more compact for convenient comparison
- Be careful, we could incorrectly conclude that better quality diamonds are cheaper on average.





Two Categorical Variables

Visualize the covariation between categorical variables with a "Plot of Dots" to determine observations.

```
ggplot(data = diamonds) +
geom_count(mapping = aes(x = cut, y = color))
```

- # Note: The size of each circle in the plot displays how many observations occurred at each combination of values
- # Get exact text details of the plot diamonds %>% count(color, cut)



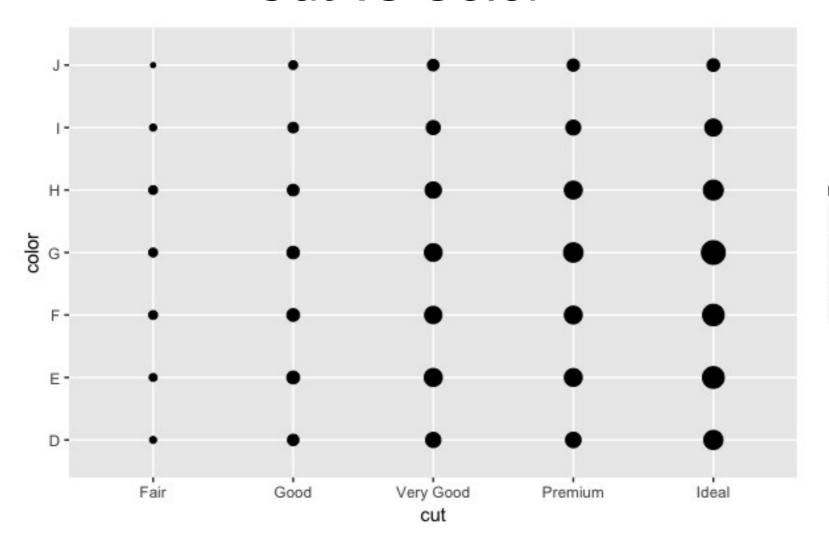
1000

2000

3000

4000

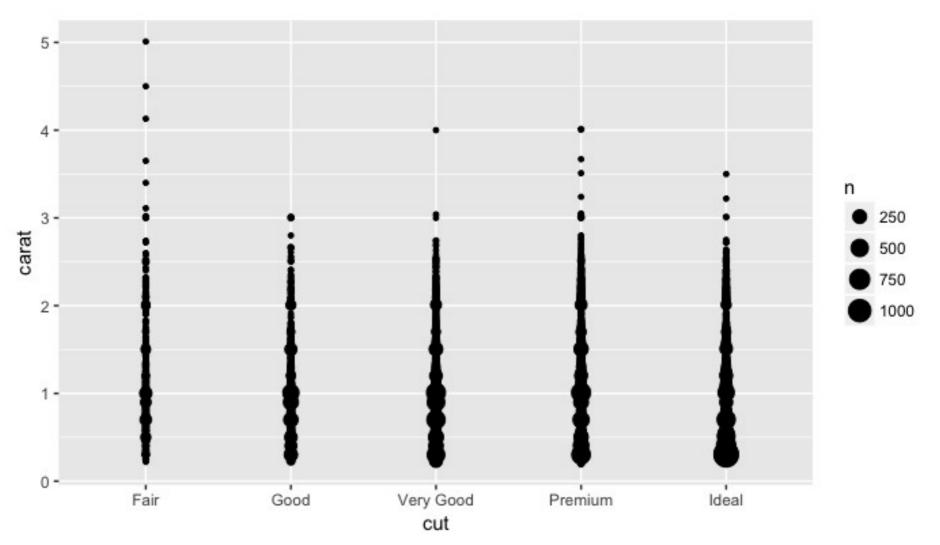
Mini Distributions: Cut vs Color



ggplot(data = diamonds) +
 geom_count(mapping = aes(x = cut, y = color))



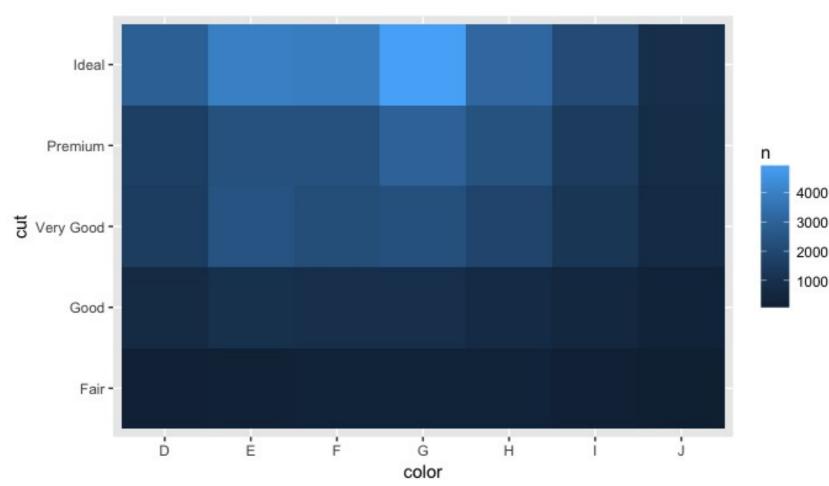
Mini Distributions: Cut vs Carat



ggplot(data = diamonds) + geom_count(mapping = aes(x = cut, y = carat))



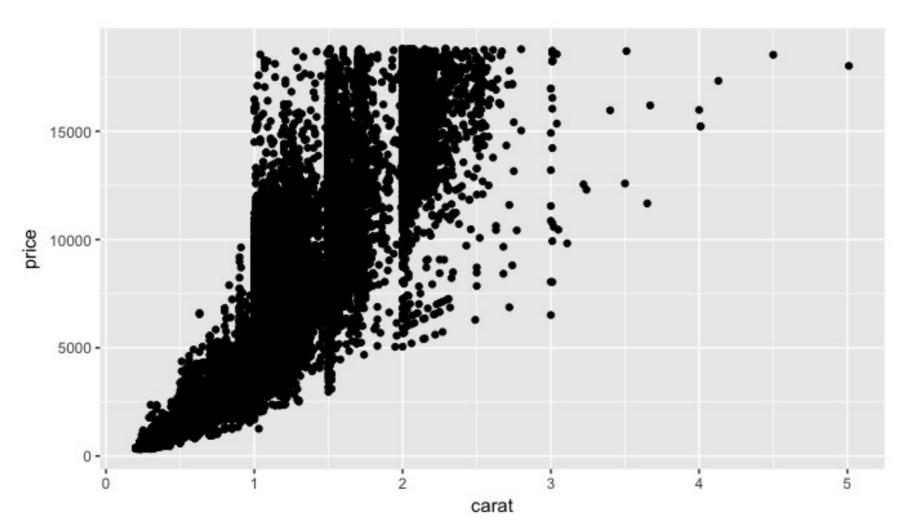




diamonds %>%
 count(color, cut) %>%
 ggplot(mapping = aes(x = color, y = cut)) +
 geom_tile(mapping = aes(fill = n))



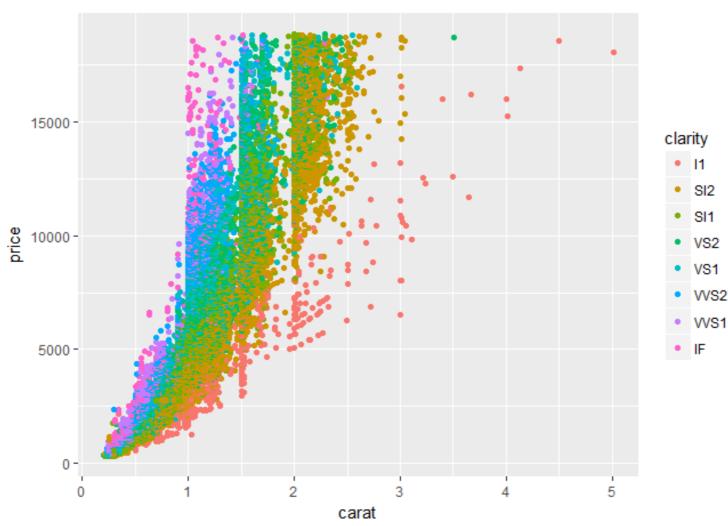
Mini Distributions: Carat vs Price



ggplot(data = diamonds) +
 geom_point(mapping = aes(x = carat, y = price))



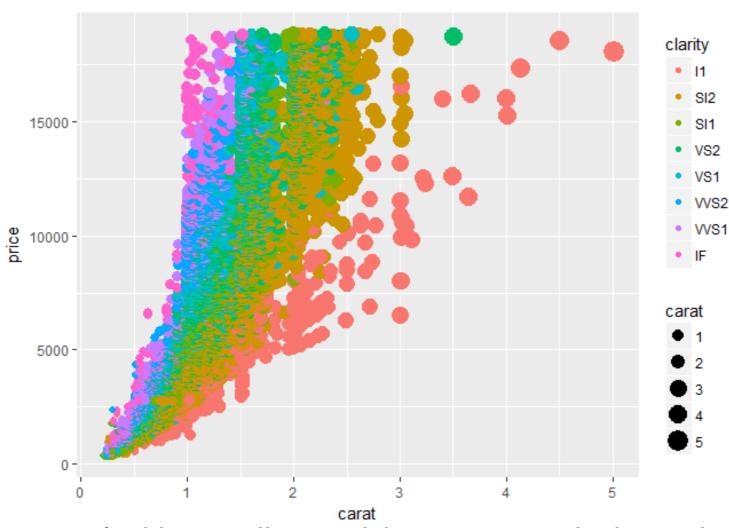




ggplot(data = diamonds) + geom_point(mapping = aes(x = carat, y = price, color= clarity))







ggplot(data = diamonds) + geom_point(mapping =
aes(x = carat, y = price,color= clarity, size = carat))



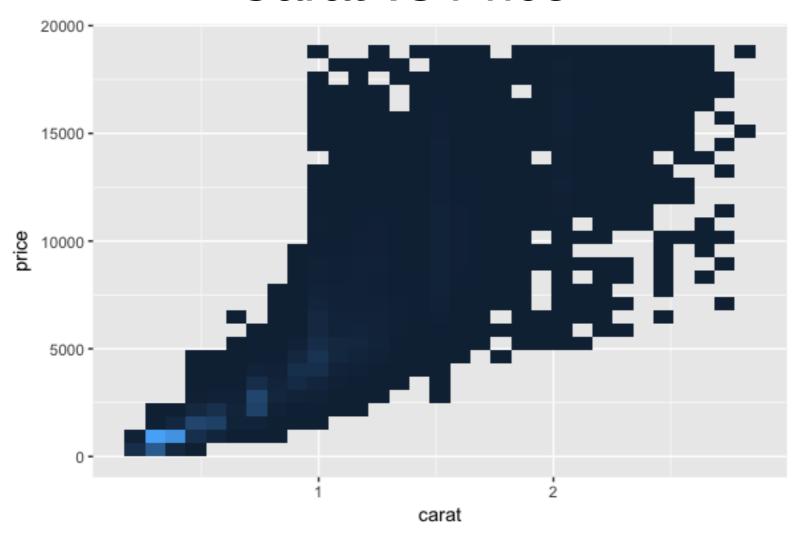
count

6000

4000

2000

Mini Distributions: Carat vs Price



ggplot(data = smaller) + geom_bin2d(mapping = aes(x = carat, y = price))