

```
library(dplyr)

rladies_global %>%
  filter(city == 'Austin')
```



R FOR DATA SCIENCE: Exploratory Data Analysis and Workflow



Hello!

Welcome to R-Ladies



1.

Introduction

R language, RStudio,
R4DS Workshop series



Three things you'll need to install

1. **Install R** -- this is the open-source programming language we'll use (download via CRAN -- Comprehensive R Archive Network)
2. **Install RStudio** -- this is the most popular IDE for R and will make your life a lot easier (download from rstudio.com/download)
3. **Install the tidyverse** -- this is the group of packages we'll use within R to work with data. Install with one line of code in R:
`install.packages("tidyverse")`



1b. Introduction

R for Data Science Workshop Series

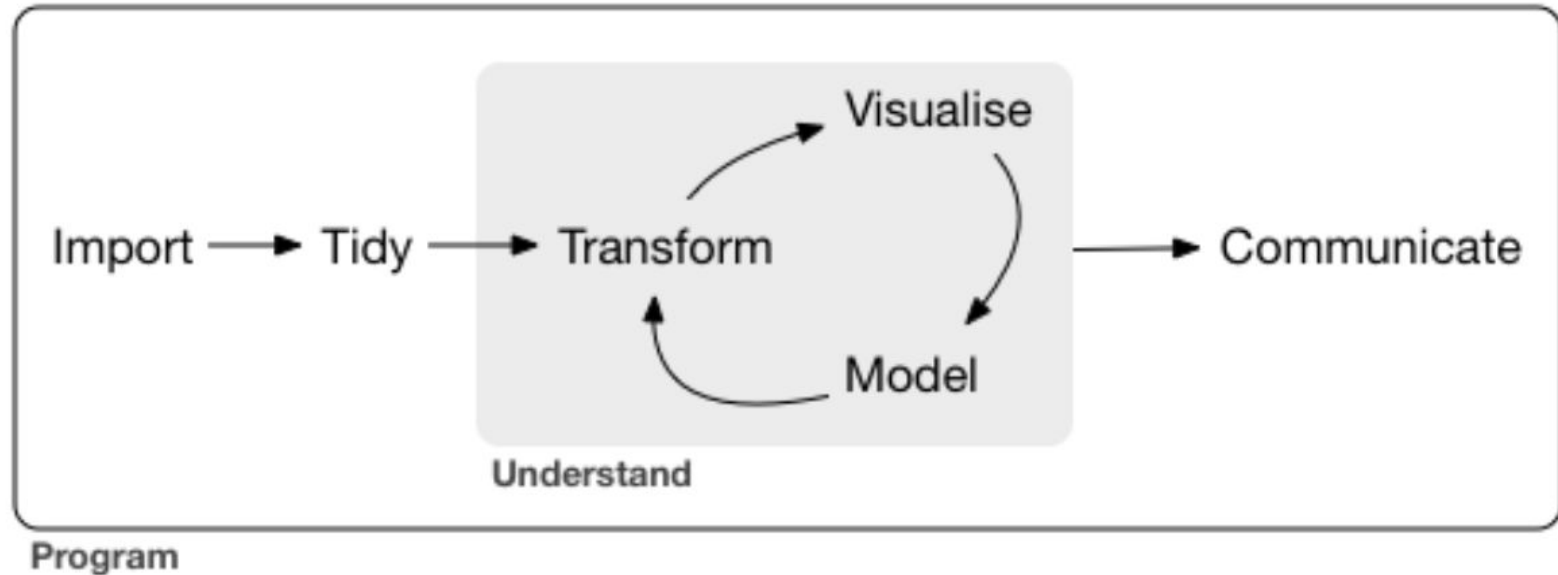


R4DS

Workshop Series

- Exploring Data with ggplot2 + dplyr [COMPLETE-see github]
- Exploratory Data Analysis and Workflow [today]
- Data Wrangling in the Tidyverse [November 29]
- Programming -- Functions, Vectors, and Iteration [December 13]
- Modeling with modelr, purrr, and broom [January 24]
- Communicating Results with rmarkdown and ggplot2 [February 21]

The data science process (tidied)





What is the tidyverse?

- Collection of R packages based on tidy data principles
- Designed to work together
- An easier way to code!
- AKA “Hadleyverse” (most packages written by Hadley Wickham)

What is the tidyverse?



What is tidy data?

- Each variable is a column
- Each observation is a row
- Each type of observational unit is a table

id	artist	track	time
1	2 Pac	Baby Don't Cry	4:22
2	2Ge+her	The Hardest Part Of ...	3:15
3	3 Doors Down	Kryptonite	3:53
4	3 Doors Down	Loser	4:24
5	504 Boyz	Wobble Wobble	3:35
6	98~0	Give Me Just One Nig...	3:24
7	A*Teens	Dancing Queen	3:44
8	Aaliyah	I Don't Wanna	4:15
9	Aaliyah	Try Again	4:03
10	Adams, Yolanda	Open My Heart	5:30
11	Adkins, Trace	More	3:05
12	Aguilera, Christina	Come On Over Baby	3:38
13	Aguilera, Christina	I Turn To You	4:00
14	Aguilera, Christina	What A Girl Wants	3:18
15	Alice DeeJay	Better Off Alone	6:50



2. Workflow



Workflow: Basics

You can use R as a calculator:

```
10 / 2 * 5
```

Create new objects with:

```
object_name <- value
```

```
x <- 3 * 4
```

<- has an RStudio keyboard shortcut!

```
Alt+- (the minus sign)]
```



Workflow: Basics

Naming

The Rules:

1. Objects must start with a letter
2. Names can only contain letters, numbers, _, and .
3. Object names are case-sensitive!

`this != THIS`

Your Preference:

`i_use_snake_case`

`otherPeopleUseCamelCase`

`Some.people.use.periods`

`And_some.People_AREweird`



Workflow: Basics

Calling Functions

Functions are called like this:

```
function_name(arg1 = val1, arg2 = val2,  
...)
```

- Text (string) arguments go in quotes
- Use tab for auto-complete (less typing!)
- If you see “+” output, you’re probably missing a parenthesis or a quote



Workflow: Scripts

Scripts are good for code that:

- You want to re-use
- Is long or complicated

Tips:

- Start script with packages to use (`library(package_name)`)
- Script editor will highlight syntax with red squiggly line; hover to see what the issue is



Workflow: Projects

“Capture all important interactions in your code”

- Your environment is hard to replicate by memory
(All packages used, the order they're loaded, your working directory, etc.)
- Important to save everything if you want to share code



Workflow: Projects

Paths and Directories

Working directory is where R looks for files to load, and where it will write out any files you want to save.

- `getwd()` prints your current working directory
- `setwd()` allows you to set a working directory

Best practices for paths:

- Forward slashes in paths are great because a backslash is a special character for R
- Avoid absolute paths because it makes it hard to share



Workflow: Projects

Project Architecture

Keep all files associated with a project together -- including input files, R scripts, analytical results and deliverables.

- Can do this using RStudio Projects
- ProjectTemplate() is my favorite package for creating folders and subfolders for organizing projects



3. Exploratory data analysis

What is exploratory data analysis (EDA)?

A state of mind--are your data what you expect them to be?

**Generate
Questions**

Investigate quality!

**Search for
answers**

Transform, visualize, and model

Refine

Adapt your questions to your results



What you need

```
library(tidyverse)
```

ggplot2

dplyr

What are we looking for?

- ❑ Creative process
 - a. No rules....
- ❑ In general:
 - a. Variation in variables
 - b. Covariation in variables (relationships)

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	666	20095360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272015272
China	2000	216766	128042583

variables

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	666	20095360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272015272
China	2000	216766	128042583

observations

country	year	cases	population
Afghanistan	1999	745	19987071
Afghanistan	2000	666	20095360
Brazil	1999	37737	172006362
Brazil	2000	80488	174504898
China	1999	212258	1272015272
China	2000	216766	128042583

values

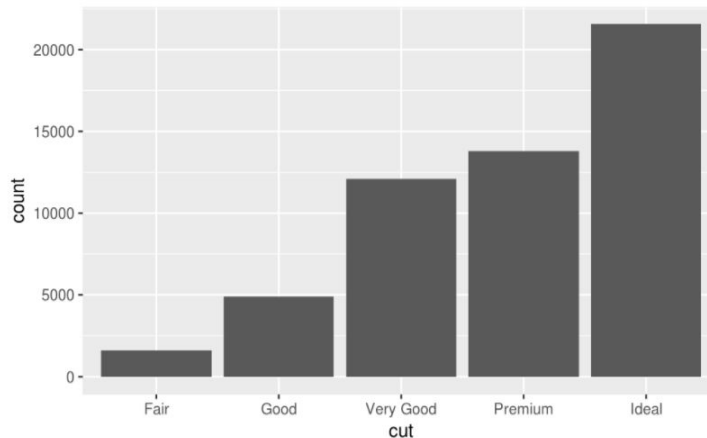
variation!

Variation- Visualizing distributions

Categorical variables = one set of values

In R, saved as character or factor

```
ggplot(data = diamonds) +  
  geom_bar(mapping = aes(x = cut))
```

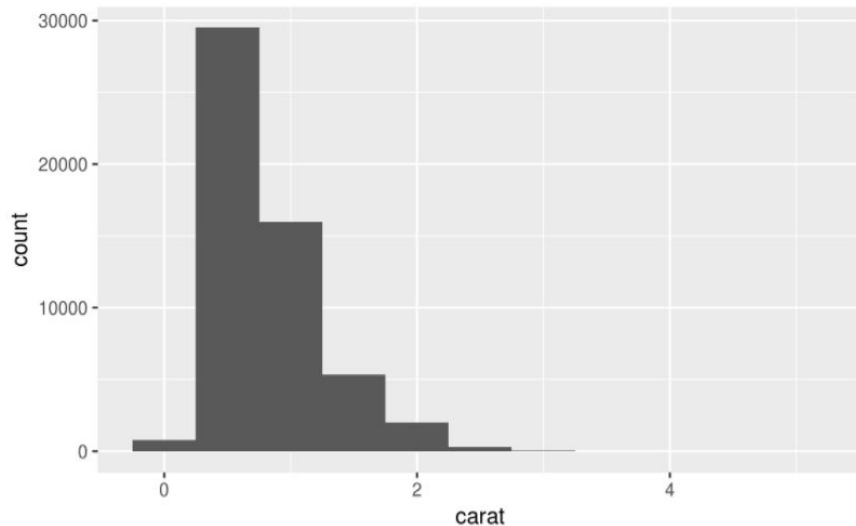


Variation-Visualizing distributions

Continuous variable = any infinite
set of ordered values

E.g. numbers, datetime

```
ggplot(data = diamonds) +  
  geom_histogram(mapping = aes(x = carat), binwidth = 0.5)
```



Variation-Typical values

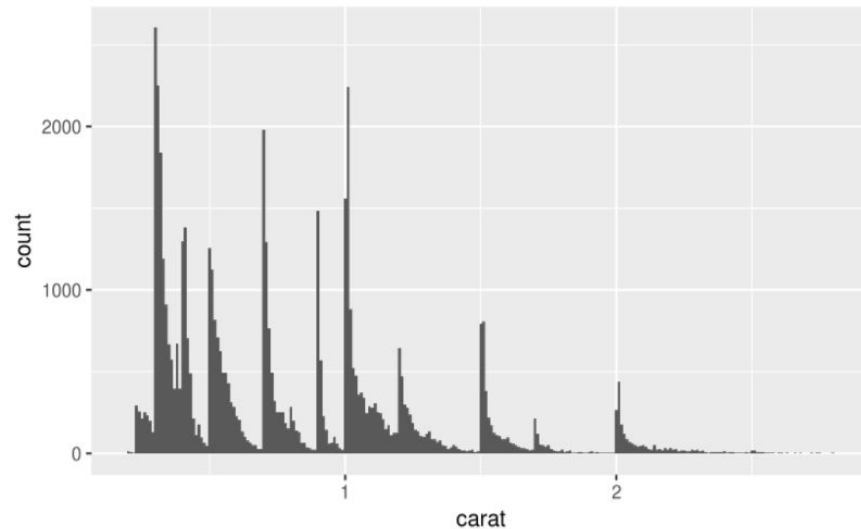
What is common?

What is rare?

Any unusual patterns?

---> likely leads to questions to explore relationship *between* vars

```
ggplot(data = smaller, mapping = aes(x = carat)) +  
  geom_histogram(binwidth = 0.01)
```



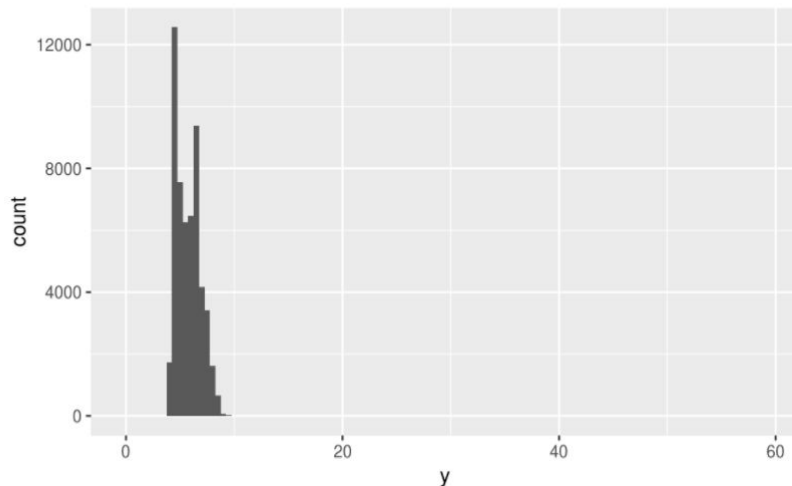
Variation-Unusual values

Outliers = unusual observations

---> can be errors, can be important

Repeat analysis without them-what happens?

```
ggplot(diamonds) +  
  geom_histogram(mapping = aes(x = y), binwidth = 0.5)
```



Variation-Missing values

What do to with unusual values?

1. Drop entire row - not recommended!
2. Replace with NA (special value type)

```
diamonds2<-diamonds%>%  
  mutate(y= ifelse(y<3 | y>20, NA, y))
```

Note: ggplot statements will not plot NA values-you will receive a warning

Covariation- a categorical var and a continuous var

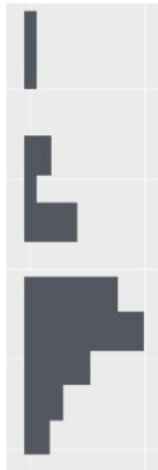
Enter the boxplot! Aka box and whisker

Display distribution of a continuous var broken down by a categorical var

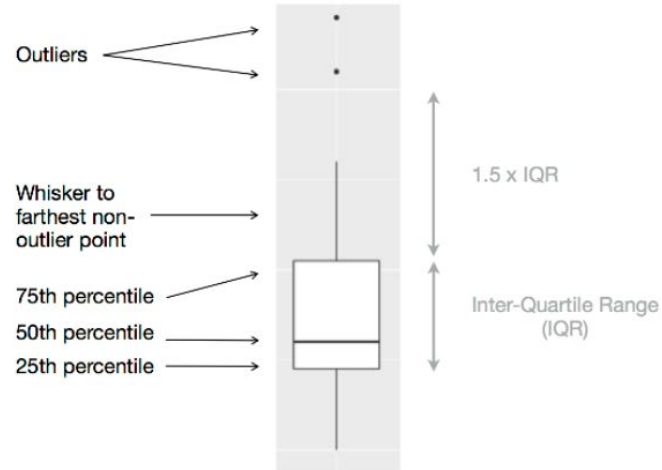
The actual values in a distribution



How a histogram would display the values (rotated)



How a boxplot would display the values





Covariation- Two categorical vars

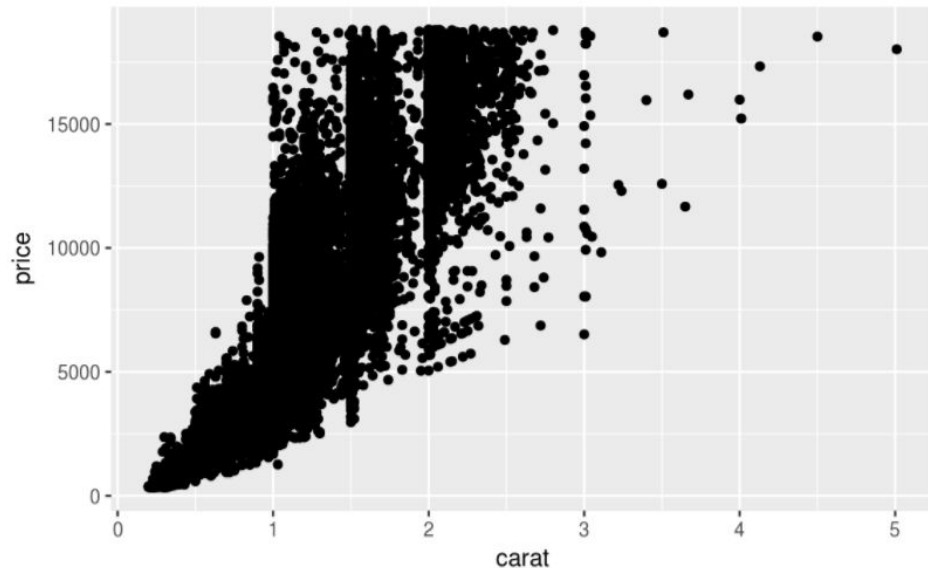
Enter the heat map!

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


Covariation- Two continuous vars

Scatterplots !
Or boxplots with
varwidth= TRUE

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



Patterns and Models

```
library(modelr)
```

- **Models** = tool to extract patterns out of data
- View relationships once effects are adjusted for
 - Residuals = residual variation after adjusting for factors included in a model

```
mod<- lm(log(price) ~ log(carat), data=diamonds)
```

- We will learn more later!



4. Wrap-up

Announcements, upcoming events, etc.



R-Ladie Austin

Upcoming Events

[Book Club: Dear Data](#) [November 8]

[R for Data Science Workshop: Data Wrangling in the Tidyverse](#) [Nov 29]

[All The Ladies in Tech Happy Hour!](#) [December 5]

Looking for presenters: Workshop on package development