# Visualization for Data Science in R

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Data Matters 2022

https://www.angelazoss.com/RVis-2Day/

## Schedule, Day 2

Session	Topics	Duration
Session 1	ggplot2 review, advanced techniques	9:30 a.m. – 10:35 a.m.
Morning break		10:35 a.m. – 10:50 a.m.
Session 2	Working with text variables	10:50 a.m. – 11:55 a.m.
Lunch		11:55 a.m. – 1:10 p.m.
Session 3	Simple interactive plots	1:10 p.m. – 2:15 p.m.
Afternoon break		2:15 p.m. – 2:30 p.m.
Session 4	Building visualizations into layouts	2:30 p.m. – 3:35 p.m.
Q&A		3:35 p.m. – 3:40 p.m.

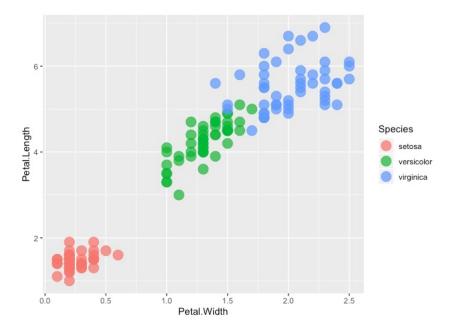
## Day 1 Review

## Example plot

"iris"

Petal.Width	Petal.Length	Species
0.3	1.4	setosa
1.3	4.0	versicolor
2.1	5.7	virginica

```
ggplot(data=iris) +
geom_point(
mapping=aes(x=Petal.Width,
y=Petal.Length,
color=Species),
size=5, alpha=.75)
```



#### General pattern

data and aesthetics will carry through from main function to shape layers

```
ggplot(data = data frame,
main plot
                   mapping = aes(...)
function
          geom ... (data = data frame,
  shape
                      mapping = aes(...),
   layer
                      non-variable adjustments)
          geom ... (data = data frame,
  shape
                      mapping = aes(...),
   layer
                      non-variable adjustments)
```

#### geom vs. scale vs. theme

Adding something that will appear inside the **chart coordinate space**?

You will (almost always) be adding a **geom**!

Changing the way a **variable is displayed**? (e.g., different axis breaks, different color mapping)

You will be adding a scale!

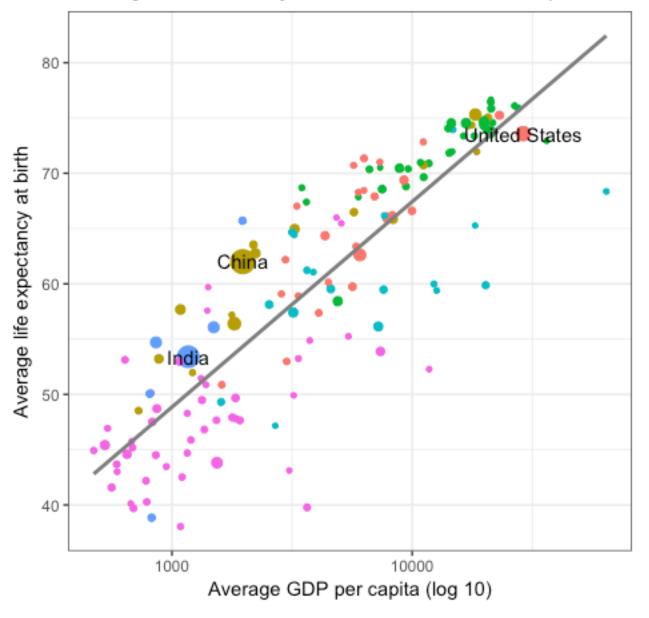
Changing the **look and feel** of the chart?

You will be adding or making changes to a **theme!** 

# Exercise 1: Gapminder Data

http://www.gapminder.org/

#### Averages across all years of the traditional Gapminder dataset



#### Average total population

- 7.5 million
- 75 million
- 750 million

#### Region

- America
- East Asia & Pacific
- Europe & Central Asia
- Middle East & North Africa
- South Asia
- Sub-Saharan Africa

# Saving charts out

# Morning Break

## Working with text variables

#### Text variables

In R, "character" variables

Gender	Age	Household Income	Education
Response	Response	Response	Response
Male	18-29		High school degree
Male	18-29	\$0 - \$24,999	Bachelor degree
Male	18-29	\$0 - \$24,999	High school degree
Male	18-29	\$100,000 - \$149,999	Some college or Associate degree
Male	18-29	\$100,000 - \$149,999	Some college or Associate degree
Male	18-29	\$25,000 - \$49,999	Bachelor degree
Male	18-29		High school degree
Male	18-29		High school degree
Male	18-29	\$0 - \$24,999	Some college or Associate degree
Male	18-29	\$25,000 - \$49,999	Some college or Associate degree
Male	18-29	\$25,000 - \$49,999	Bachelor degree
Male	30-44	\$50,000 - \$99,999	Graduate degree
Male	18-29		High school degree
Male	18-29	\$0 - \$24,999	Some college or Associate degree
Male	18-29	\$50,000 - \$99,999	Bachelor degree

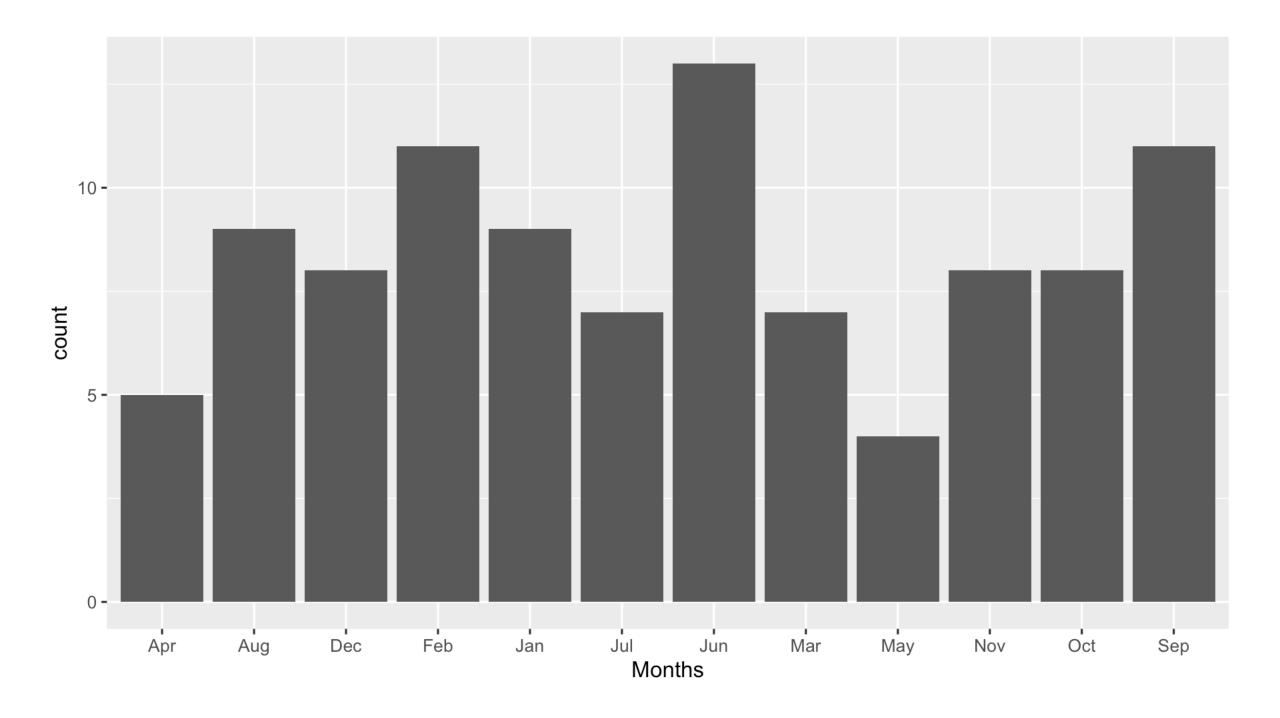
# Problems with text variables: Ordering

#### **Factors**

- Default ordering for categories:
   alphabetical
- Converting to factor allows you to:
  - Specify "levels" for a categorical variable
  - Specify the order of those levels
  - Specify whether the factor is "ordered"

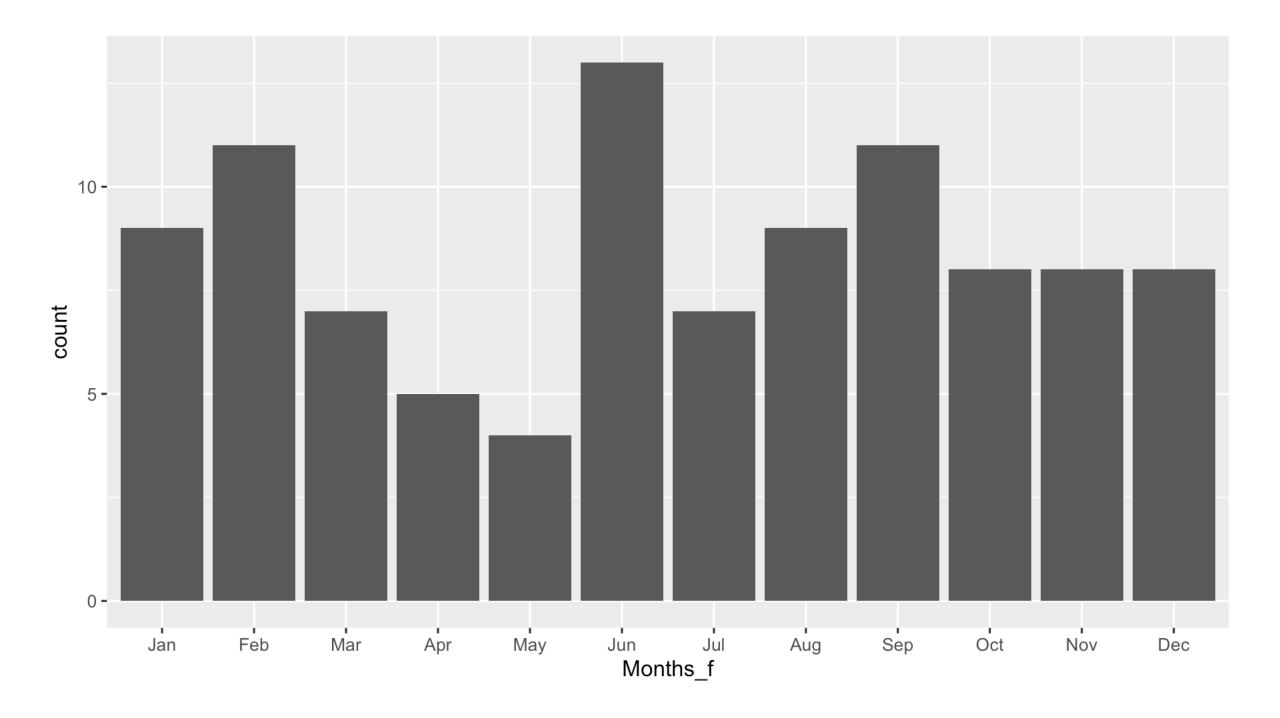
https://r4ds.had.co.nz/factors.html

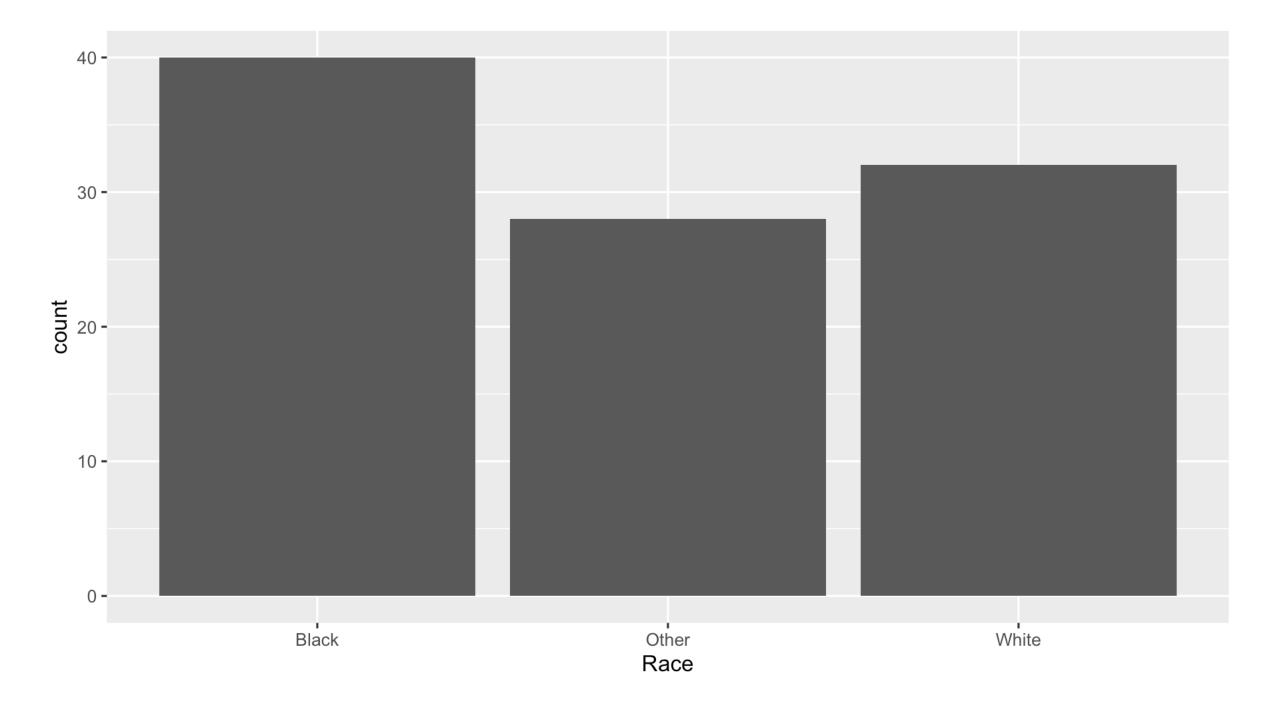
```
> x1 <- c("Dec", "Apr", "Jan",
"Mar")
> factor(x1)
   Dec Apr Jan Mar
Levels: Apr Dec Jan Mar
> month levels <- c( "Jan", "Feb",</pre>
"Mar", "Apr", "May", "Jun", "Jul",
"Aug", "Sep", "Oct", "Nov", "Dec")
> y1 <- factor(x1,
            levels = month levels)
> y1
<u>[1] Dec Apr Jan Mar</u>
Levels: Jan Feb Mar Apr May Jun Jul
Aug Sep Oct Nov Dec
```



### Order by meaning

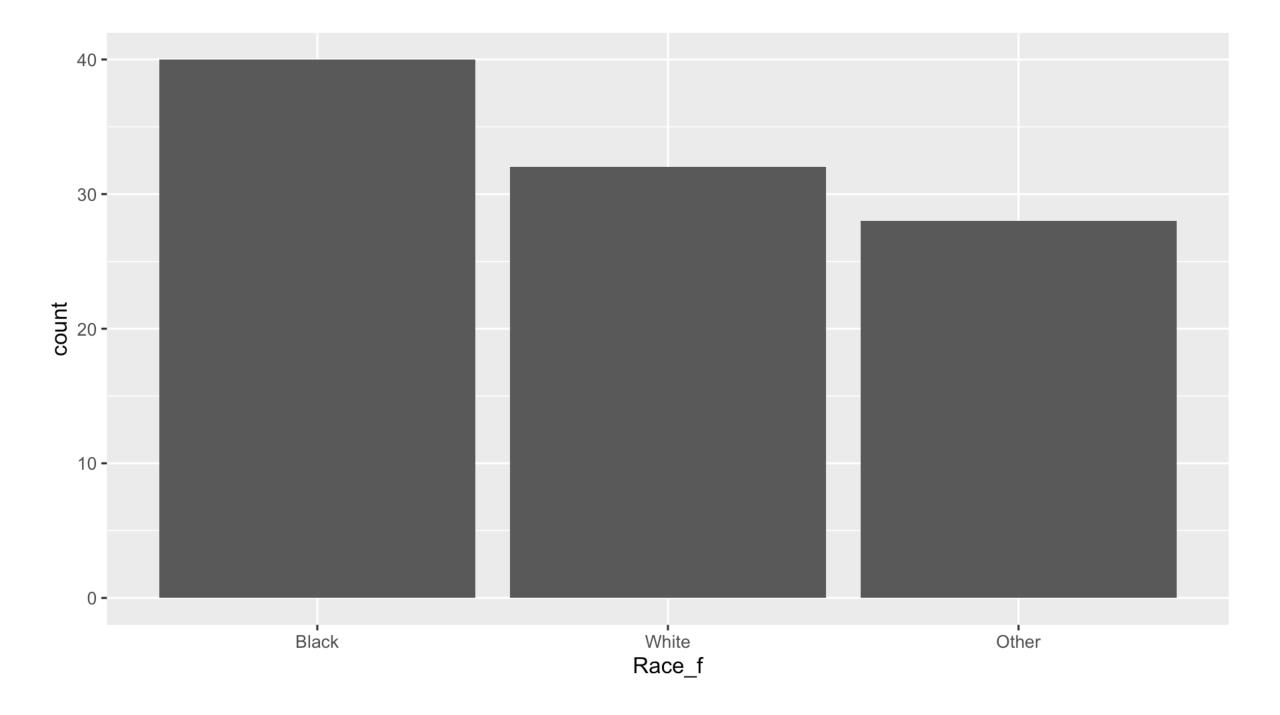
```
month levels <- c( "Jan", "Feb", "Mar", "Apr",
"May", "Jun", "Jul", "Aug", "Sep", "Oct", "Nov",
"Dec" )
data <- data %>%
    mutate (Months f = Months %>%
                       as factor() %>%
                       fct relevel(month levels))
```





## Order by value (using forcats)

```
demo <- data %>%
    mutate(Race f = Race %>%
                      as factor() %>%
                      fct infreq())
ggplot (data,
        aes (Race %>%
            as factor() %>%
            fct infreq())) +
    geom bar()
```



### forcats package: helpful functions

- as\_factor(char\_var):convert a character variable to a factor
- fct\_infreq(factor):
   take factor levels and set the order according to
   (inverse) category frequency
- fct\_reorder(factor, num\_var):
   sort factor levels by a second, numerical variable
   (like a pre-calculated count or average)

### Note about read.csv (base R)

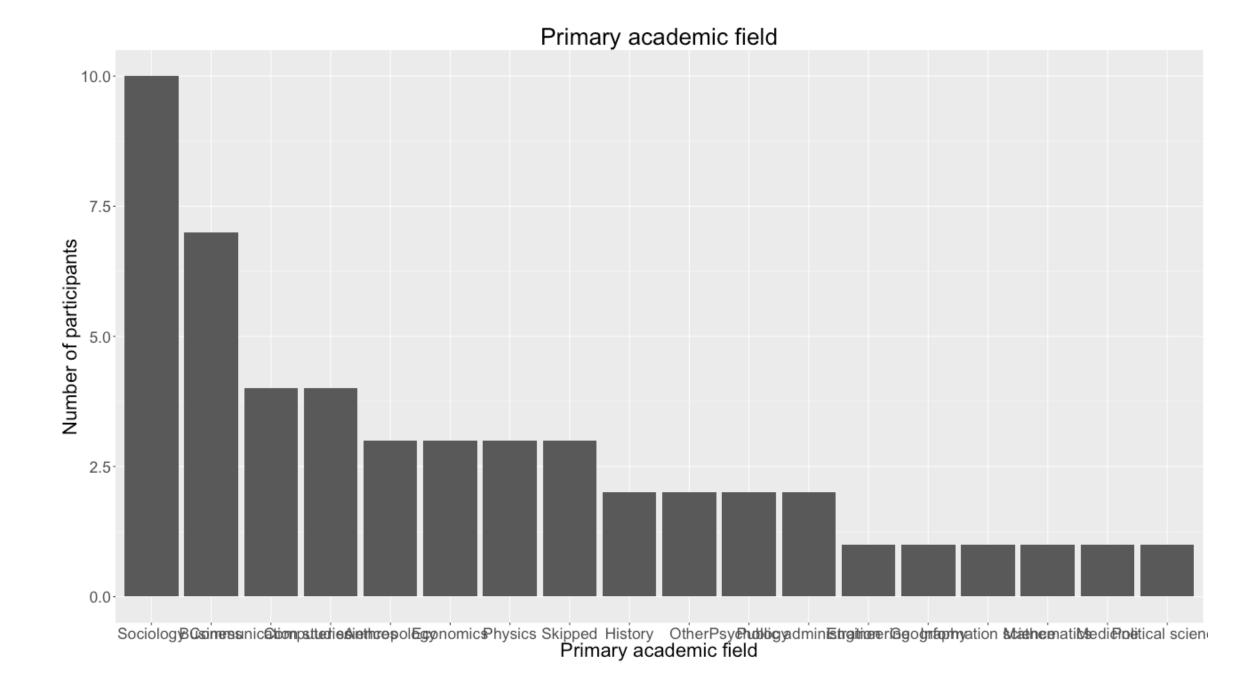
- Converts string variables to factors by default
- Can either:
  - Include stringsAsFactors=FALSE
  - Use read csv() instead

#### Factoring resources

#### From Amelia McNamara:

- RStudioConf 2019 slides:
   Working with Categorical Data in R Without Losing Your Mind
- Wrangling Categorical Data in R article
- Wrangling Categorical Data in R repository

# Problems with text variables: Long category names



### In ggplot2, have to flip the axes

```
+ coord flip()
Or
ggplot(df, aes(y=cat variable)) +
    geom bar()
```

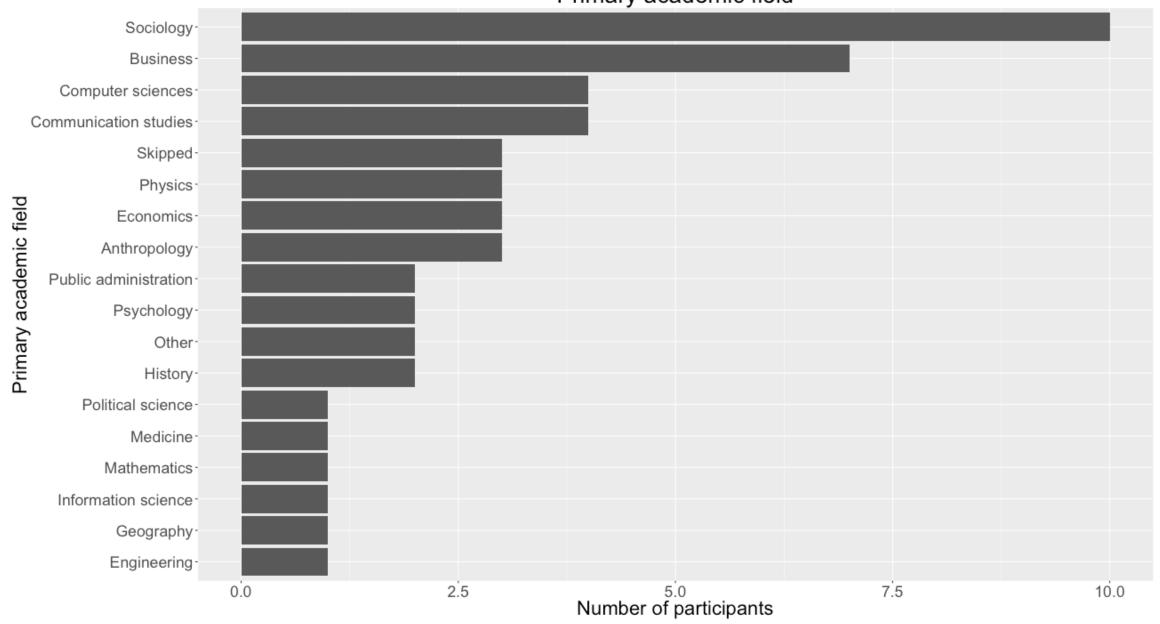
Primary academic field Political science Medicine-Mathematics -Information science Geography-Engineering-Primary academic field Public administration-Psychology-Other-History-Skipped-Physics-Economics-Anthropology -Computer sciences-Communication studies Business-Sociology-7.5 2.5 Number of participants 0.0 10.0

## When you flip axes, you sort the other way

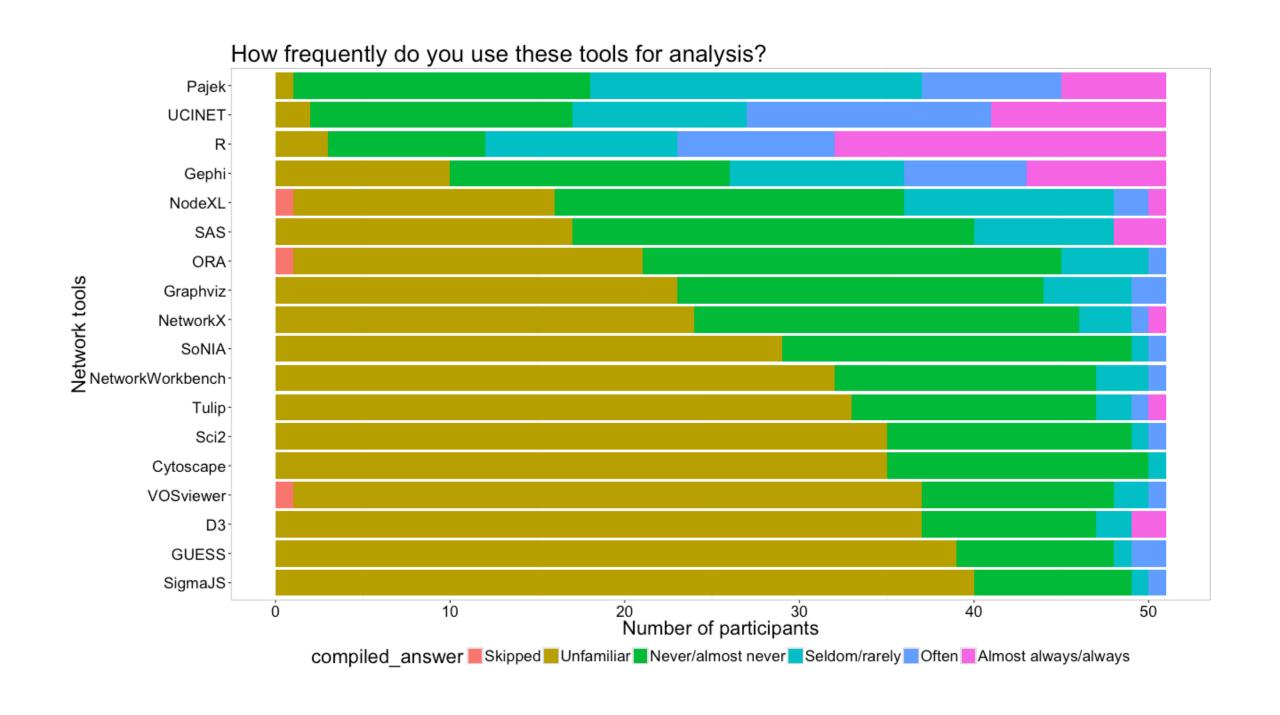
```
academic_field %>%
  as_factor() %>%
  fct_infreq() %>%
  fct_rev()
```

Have to reverse the order of the levels

Primary academic field

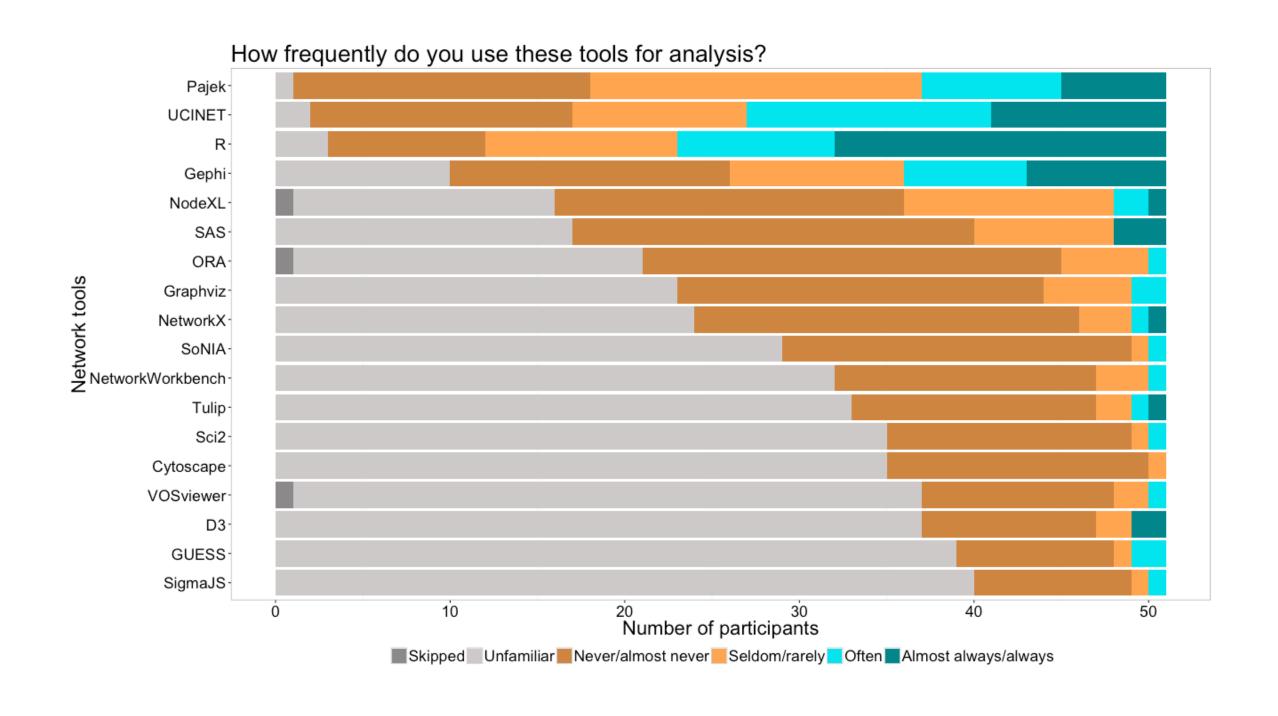


# Problems with text variables: Arbitrary colors



#### Select colors manually, or use alternate palette

```
scale fill manual (
    values=c("snow4", "snow3",
             "tan3", "tan1",
              "turquoise2", "turquoise4"))
# Also see package RColorBrewer
scale fill brewer (palette="BrBG")
```



## Lunch

# Designing tools for data exploration

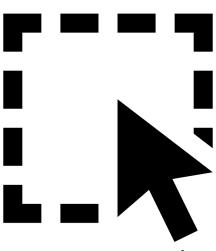
### Supporting data exploration

#### **Output**



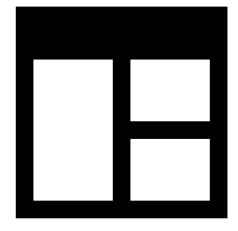
Picking the right visual elements

#### Input



Giving users the right controls

#### Layout



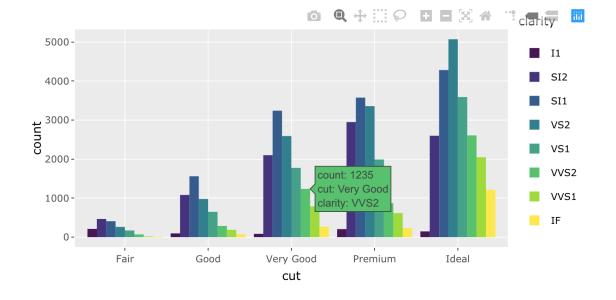
Arranging everything in the right place

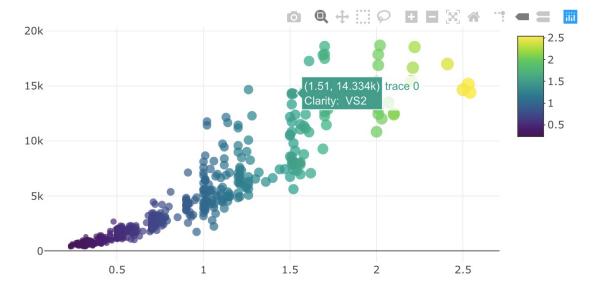
#### Interactive components

- Start with the basic info
- Show more or less on demand



Next





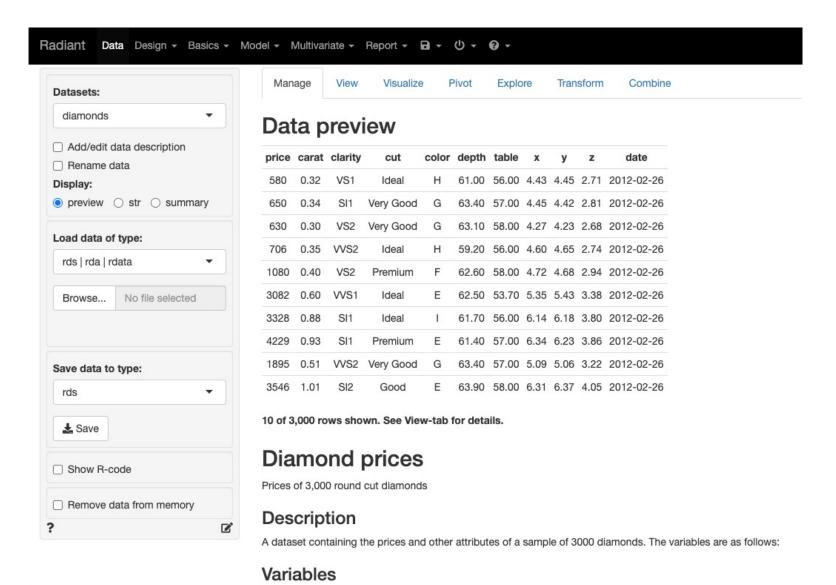
https://www.htmlwidgets.org/

Previous

http://gallery.htmlwidgets.org/

# Responding to user input

- Generalized workflows
- Custom subsetting
- Changing parameters
- Personalizing output



https://shiny.rstudio.com/

https://shiny.rstudio.com/gallery/radiant.html

# Interactive components

## Why make charts interactive?

- Easier for data exploration
  - Drill-down to data subsets of interest
  - Details on demand
  - Customize look-and-feel of chart
- Can be more engaging for users

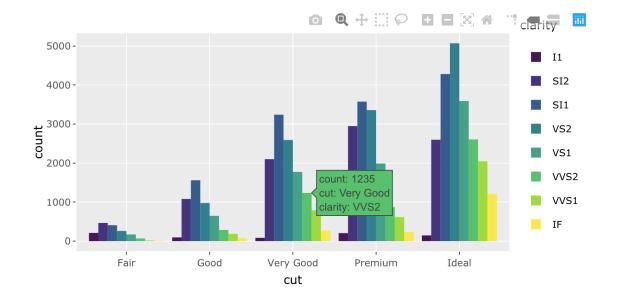
# Visual information seeking mantra

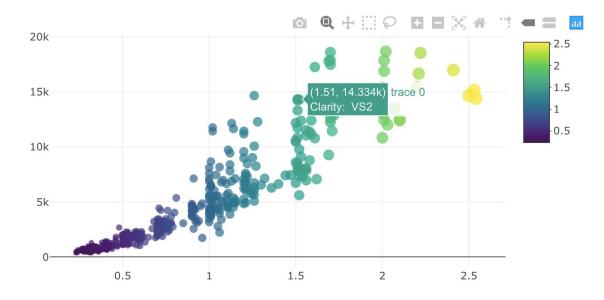
Overview first, zoom and filter, then details-on-demand

Shneiderman, B. (1996). The eyes have it: A task by data type taxonomy for information visualization. In VL '96 Proceedings of the 1996 IEEE Symposium on Visual Languages.

### Interactivity in R Markdown

- R Markdown gets compiled into HTML
- Some R packages can create interactive elements by converting R output to HTML/JavaScript code in the final document
- We will use the **plotly** package to create interactive charts





http://www.htmlwidgets.org/

### Other interactive chart packages

- ggiraph for extending ggplot2 with interactive geoms
- rCharts for an R version of Polycharts, NVD3, and MorrisJS
- rBokeh for an R version of Bokeh
- altair for an R version of Altair
- <u>leaflet</u> for interactive maps

# Exercise 4: Make ggplot2 charts interactive

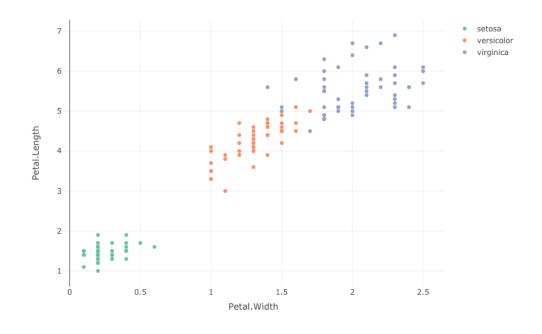
## plotly

- Create plots that are interactive right away, either in R Markdown or in a website version
- Can either convert ggplot2 charts to plotly or build natively with plotly syntax

### Basic plotly syntax

- Main plot function: plot\_ly()
- Set the data: data = [data name]
- No aesthetics function, just list aesthetics pairings
- For each variable name, need
   "~" in front
- Default plot type is scatter; for others, add: type = "[plot type]"

```
plot_ly(data = iris,
    x = ~Petal.Width,
    y = ~Petal.Length,
    color = ~Species,
    type="scatter")
```



### Publishing interactive plotly charts

- Write R Markdown in RStudio
- Make sure "output" at top is "html\_document"
- Use knitr to knit to HTML
- Publish HTML to:
  - RPubs (click the "Publish" button in RStudio)
  - GitHub (setup a <u>GitHub Pages repository</u> and add the HTML files)
  - Any website you already have that can publish HTML

# Afternoon Break

# DT for interactive data tables

# Coordinated Views

#### Views that share data

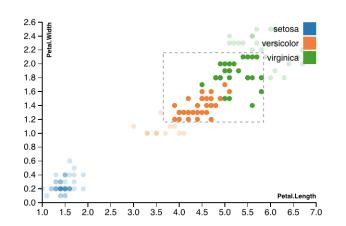
- Each view should be relatively simple, have a specific purpose
- Views can work together to explore complex interactions
- The Crosstalk package connects interactive components together

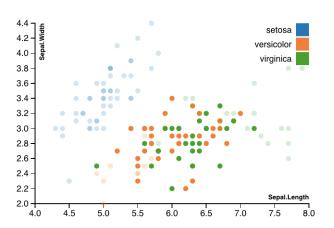
https://rstudio.github.io/crosstalk/



	lat↓↑	long ↑	depth  ↑	mag 🖹	stations.
308	-22	180.53	583	4.9	20
873	-11.02	167.01	62	4.9	36
277	-23.33	180.18	528	5	59
752	-21.29	185.77	57	5.3	69
352	-12.01	166.29	59	4.9	27
354	-30.17	182.02	56	5.5	68
168	-19.89	183.84	244	5.3	73
474	-10.79	166.06	142	5	40
338	-27.19	182.18	69	5.4	68

Showing 1 to 10 of 32 entries (filtered from 100 total entries





# Combining interactive components with Crosstalk

# Dashboards in R Markdown

#### What is a dashboard?



#### "Normal" R Markdown

• R Markdown elements like headings, text

```
# Heading 1
## Heading 2
Regular text
* Bulleted text
```

Code chunks

```
```{r}
...
```

Note: Comments work like HTML <!--HTML Comment style -->

#### Markdown for flexdashboards

Page

\_\_\_\_\_

Column (or Row)

-----

### Chart titles

Regular text

\* Bulleted text

```{r}

\*\*\*

# Arrange various elements in flexdashboard

# Thanks for your time this week!

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