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	Simple interactive plots	10:50 a.m. – 11:55 a.m.
Lunch		11:55 a.m. – 1:10 p.m.
Session 3	Intro to Shiny	1:10 p.m. – 2:15 p.m.
Afternoon break		2:15 p.m. – 2:30 p.m.
Session 4	Shiny examples and practice	2:30 p.m. – 3:35 p.m.
Q&A		3:35 p.m. – 3:40 p.m.

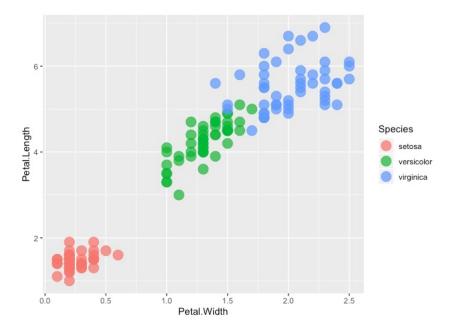
Day 1 Review

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!

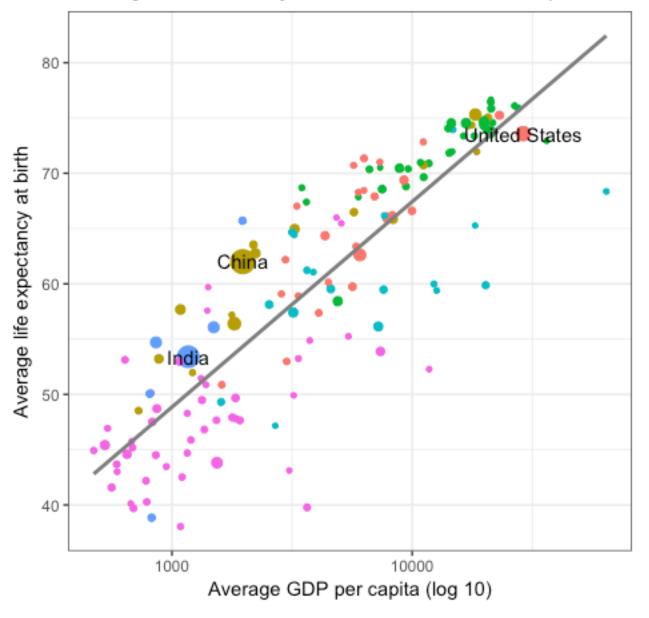
Changing the **look and feel** of the chart?

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

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

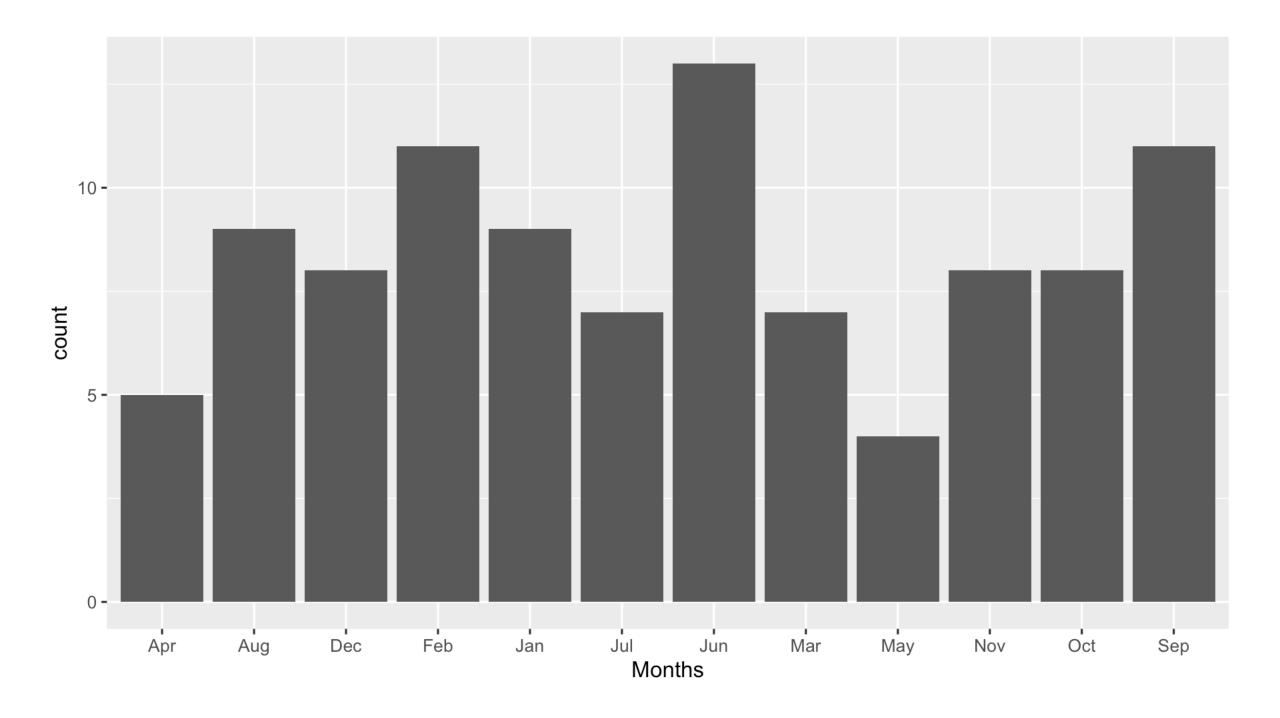
Problems with text variables: Ordering

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"

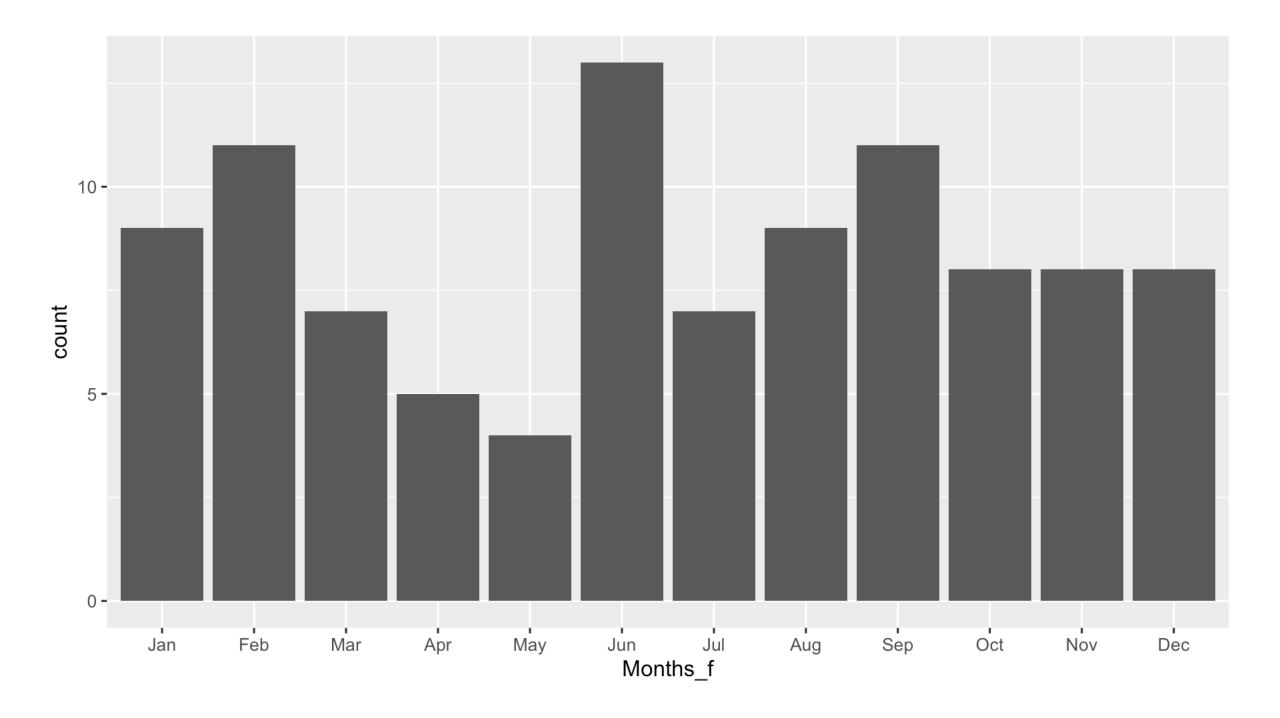
https://r4ds.had.co.nz/factors.html

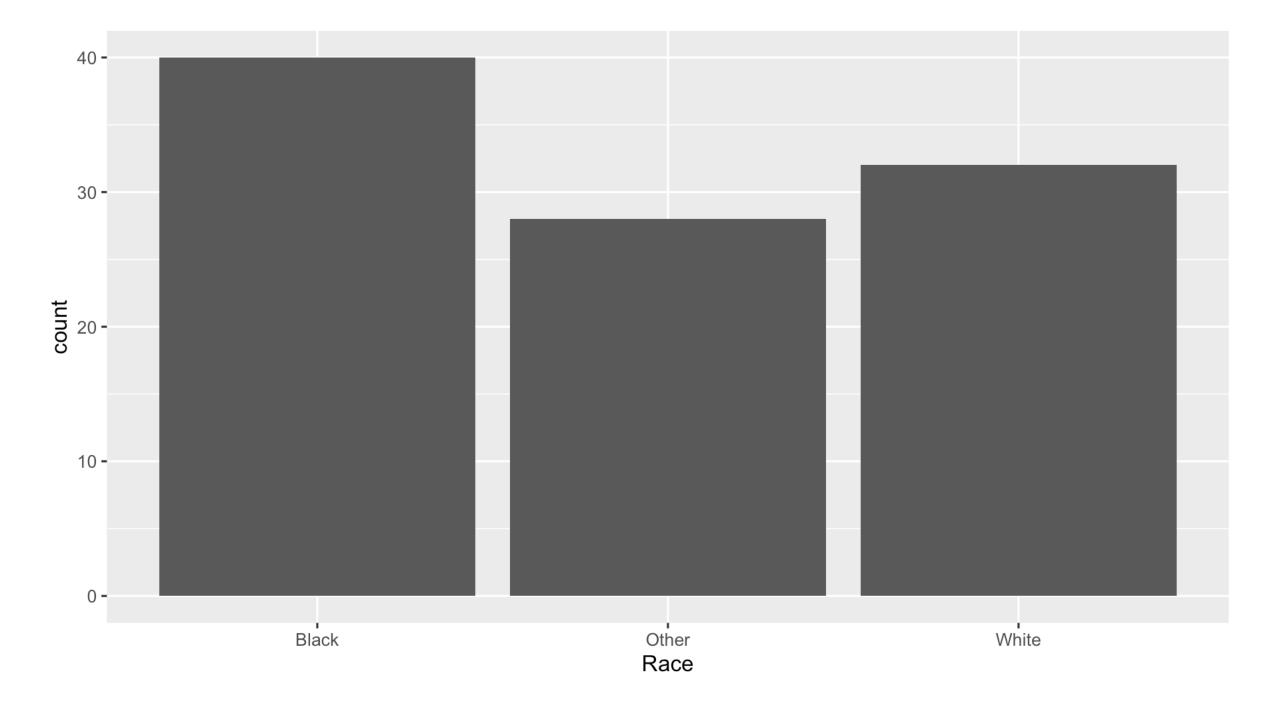
```
> x1 <- c("Dec", "Apr", "Jan",
"Mar")
> factor(x1)
   Dec Apr Jan Mar
Levels: Apr Dec Jan Mar
> month levels <- c( "Jan", "Feb",</pre>
"Mar", "Apr", "May", "Jun", "Jul",
"Aug", "Sep", "Oct", "Nov", "Dec")
> y1 <- factor(x1,
            levels = month levels)
> y1
<u>[1] Dec Apr Jan Mar</u>
Levels: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



Order by meaning

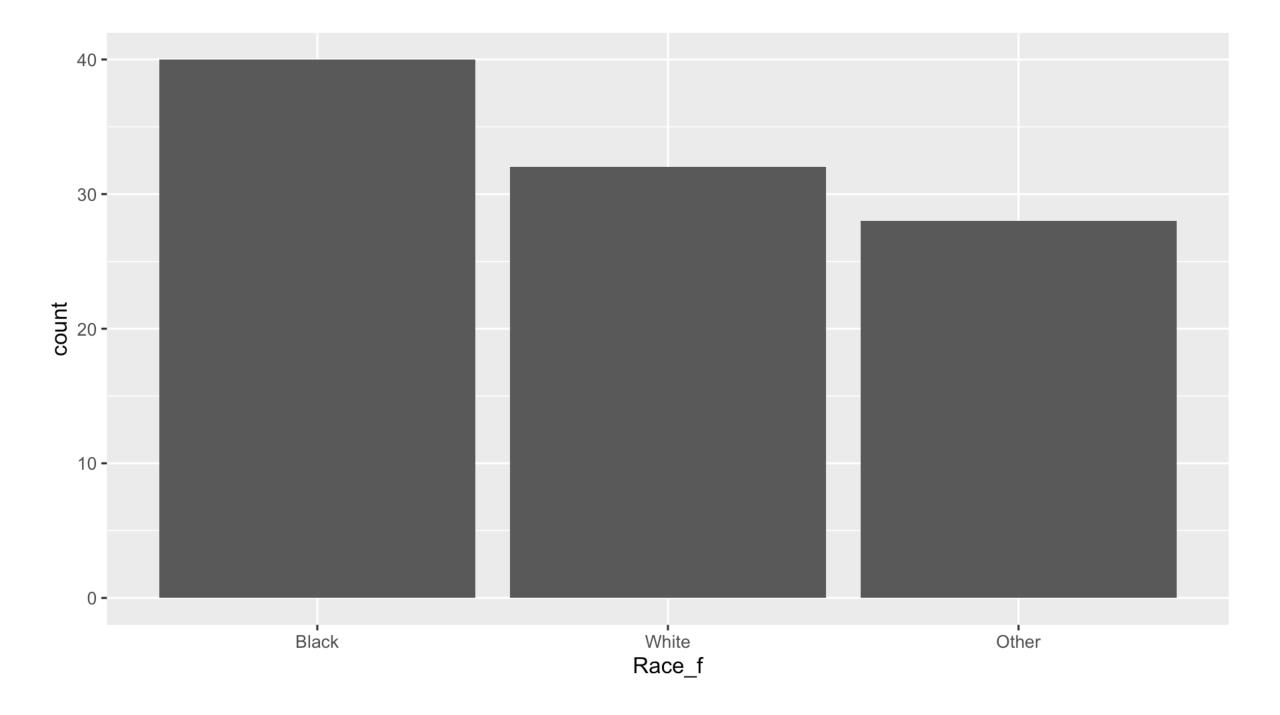
```
month levels <- c( "Jan", "Feb", "Mar", "Apr",
"May", "Jun", "Jul", "Aug", "Sep", "Oct", "Nov",
"Dec" )
data <- data %>%
    mutate (Months f = Months %>%
                       as factor() %>%
                       fct relevel(month levels))
```





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)

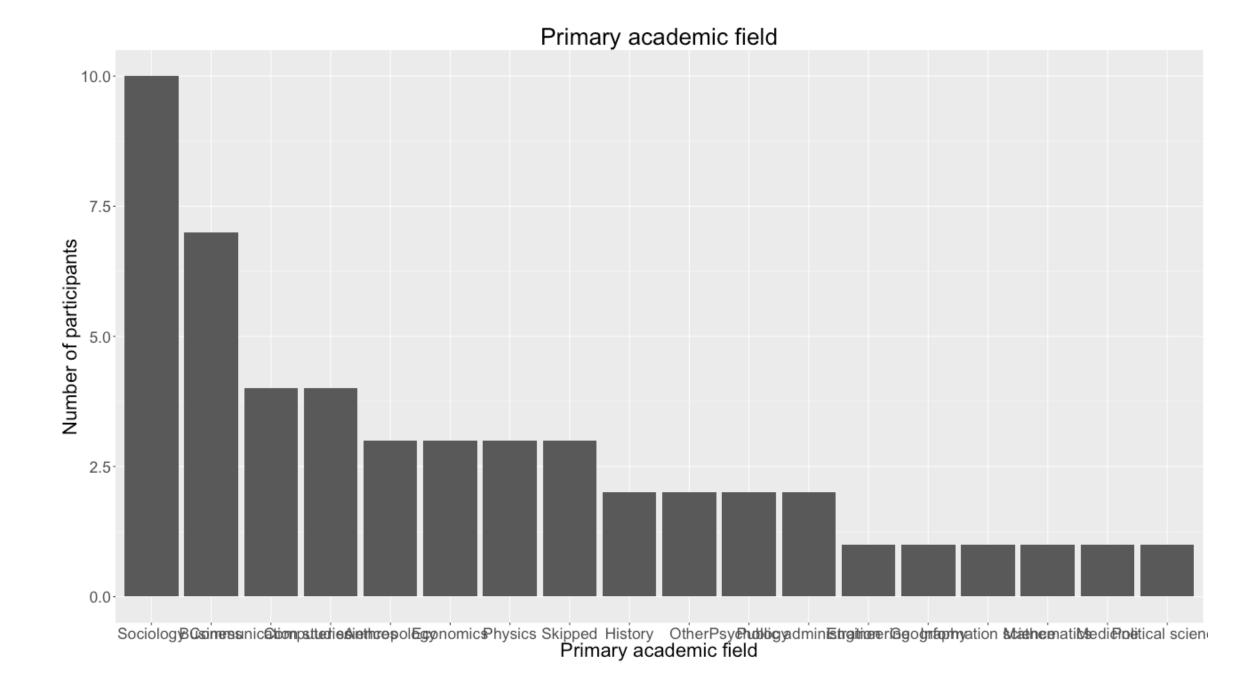
- Converts string variables to factors by default
- Can either:
 - Include stringsAsFactors=FALSE
 - Use read csv() instead

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()
```

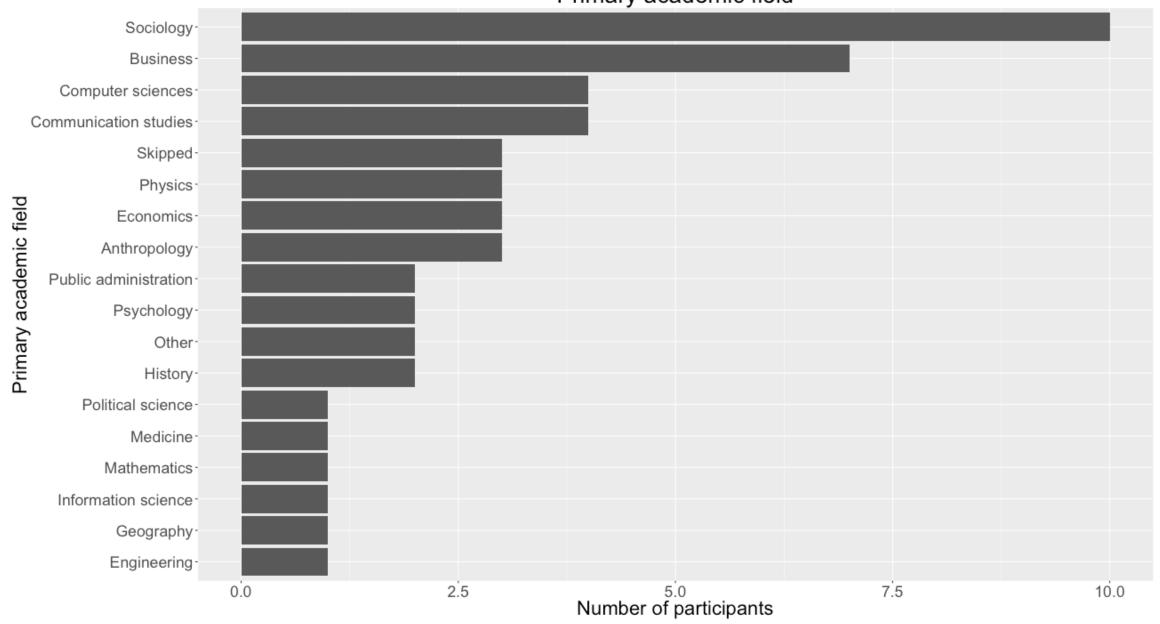
Primary academic field Political science Medicine-Mathematics -Information science Geography-Engineering-Primary academic field Public administration-Psychology-Other-History-Skipped-Physics-Economics-Anthropology -Computer sciences-Communication studies Business-Sociology-7.5 2.5 Number of participants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

When you flip axes, you sort the other way

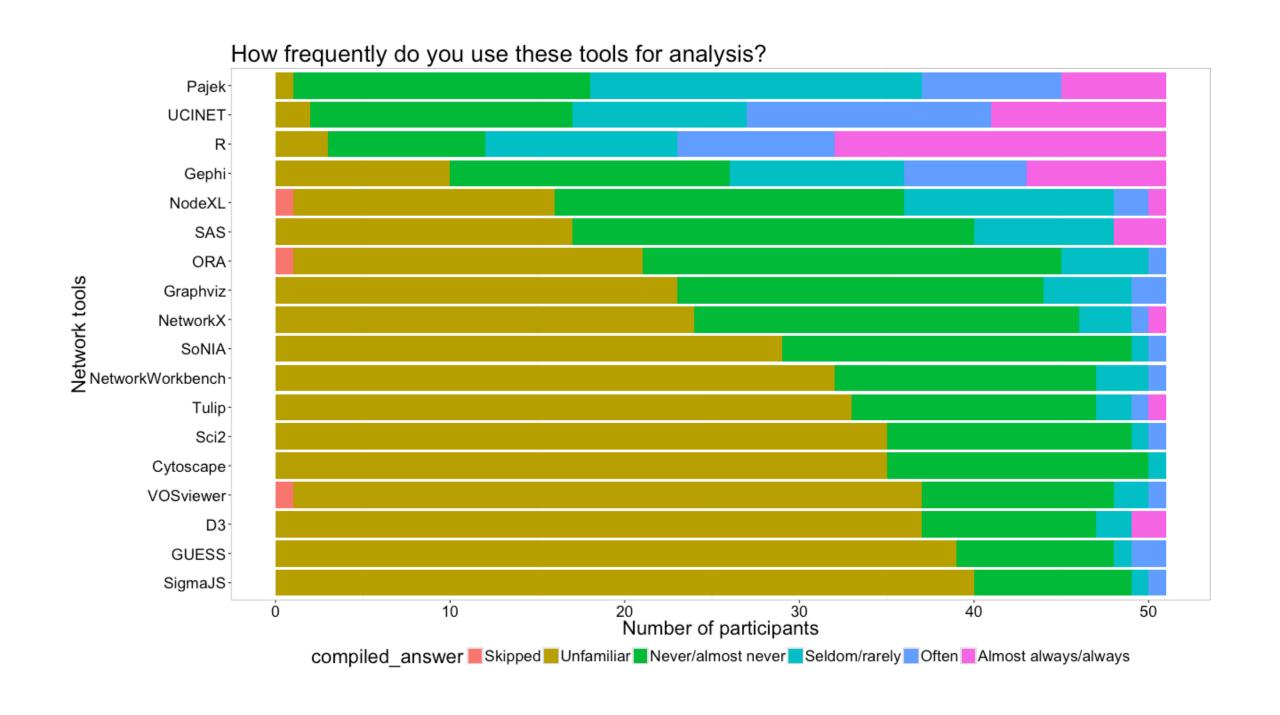
```
academic_field %>%
  as_factor() %>%
  fct_infreq() %>%
  fct_rev()
```

Have to reverse the order of the levels

Primary academic field

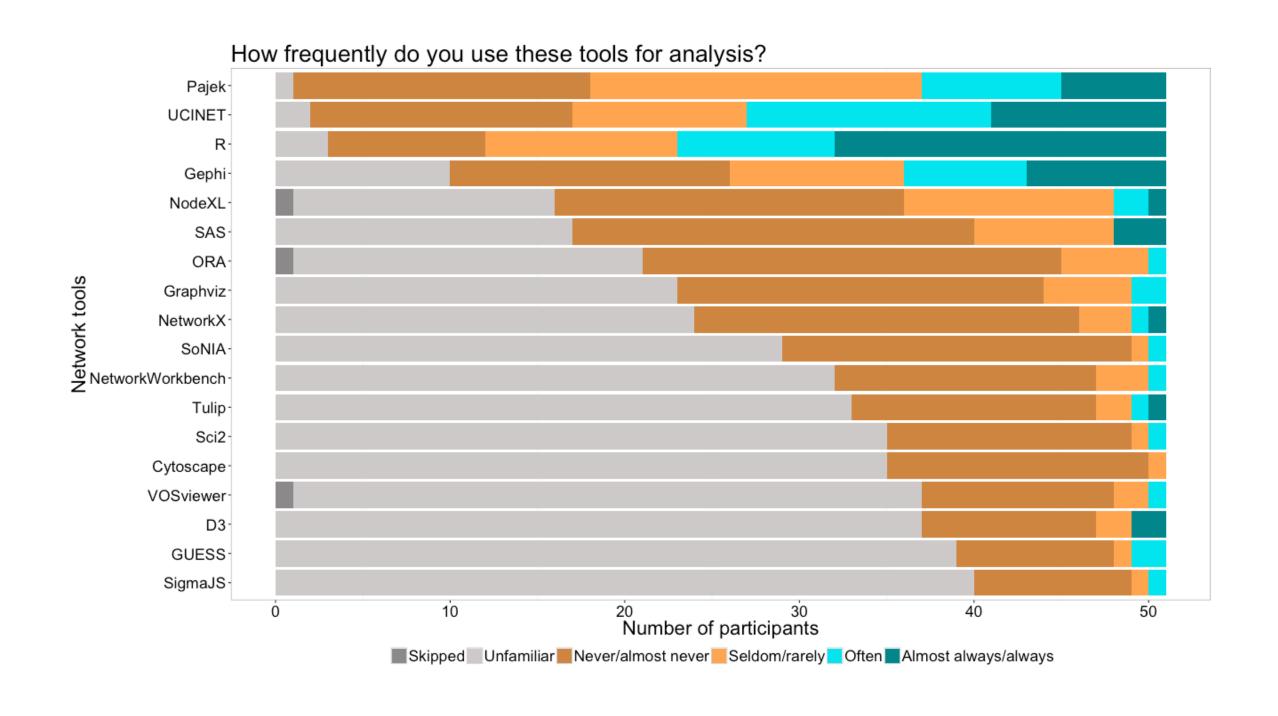


Problems with text variables: Arbitrary colors



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")
```



Sample Projects

Morning Break

Designing tools for data exploration

Supporting data exploration

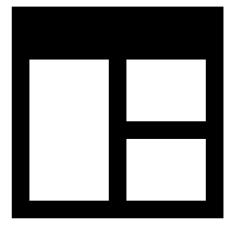
Output



visual elements

Input Giving users the right controls

Layout



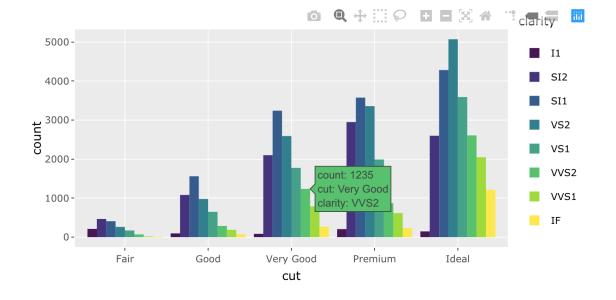
Arranging everything in the right place

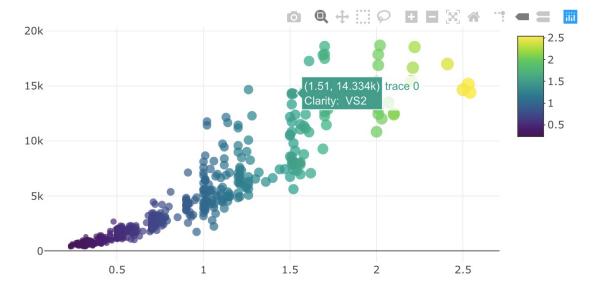
Interactive components

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



Next





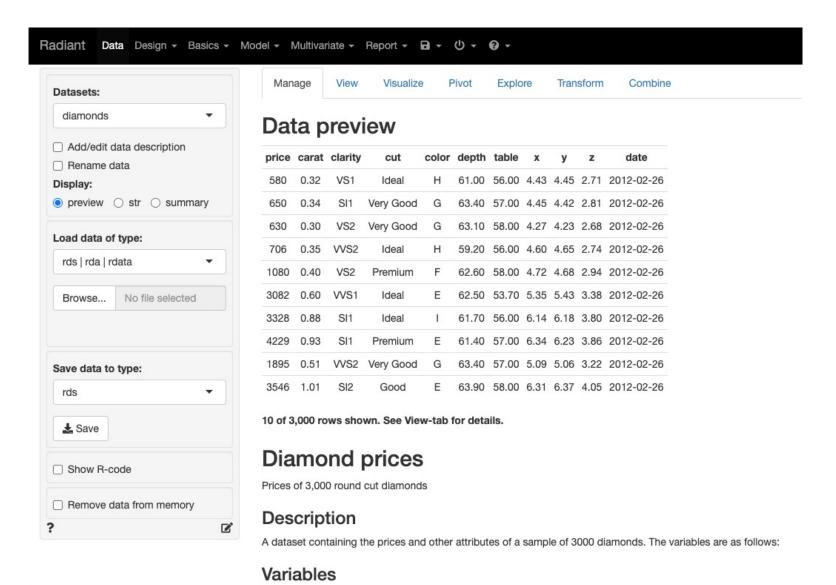
https://www.htmlwidgets.org/

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

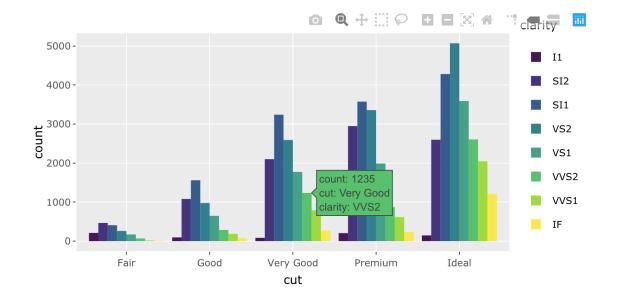
Visual information seeking mantra

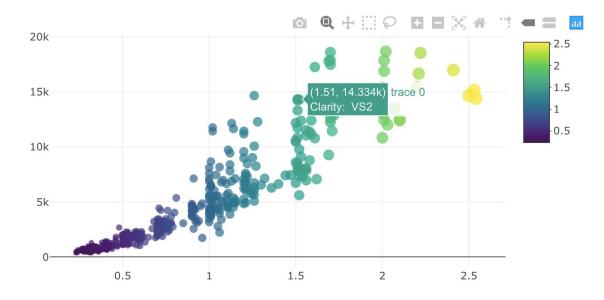
Overview first, zoom and filter, then details-on-demand

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 2: Make yesterday's charts interactive

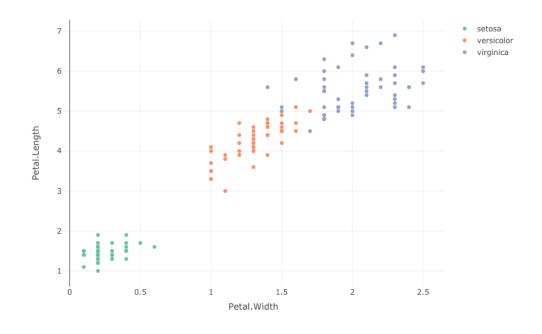
plotly

- 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

Lunch

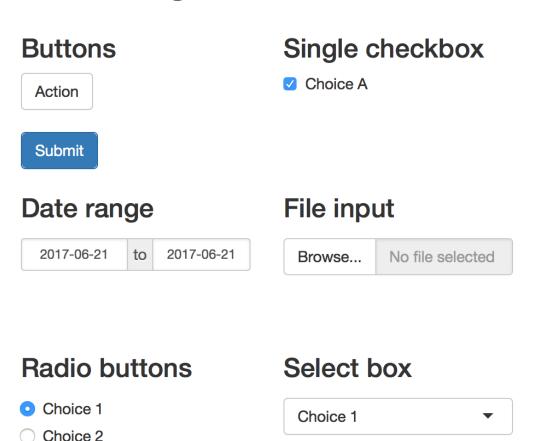
Responding to user input

Input controls to guide exploration

- For more complex data exploration, you may need input from the user
- Input controls can gather different kinds of information from the user, from free text to buttons to date ranges
- Simple input processing can happen within a standalone website, but for complex data processing, the input may need to feed back into a real R calculation (Shiny)

Basic widgets

Choice 3

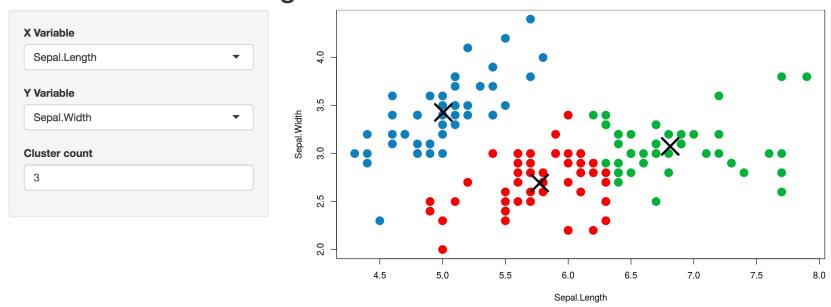


Shiny

What is Shiny?

An interactive interface onto an R program (requires a special server to publish)

Iris k-means clustering



http://shiny.rstudio.com/

How can you use Shiny for visualization?

- Use Shiny to control some kind of simulation interactively, then visualize the results
- Use Shiny to change components within the chart (e.g., switch the mappings)
- Use Shiny to filter data to subsets to highlight patterns
- Change type of regression, plot results

Shiny examples

Interactive visualizations

Shiny is designed for fully interactive visualization, using JavaScript libraries like d3, Leaflet, and Google Charts.



SuperZip example

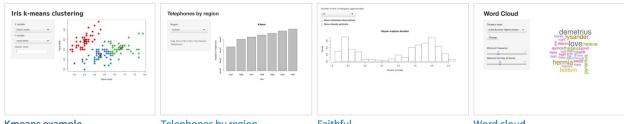
Bus dashboard

Movie explorer

Google Charts

Start simple

If you're new to Shiny, these simple but complete applications are designed for you to study.

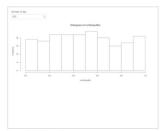


Kmeans example

Telephones by region

Faithful

Word cloud



Single-file shiny app

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

Shiny Apps

How do you build a Shiny app?

User Interface (UI)

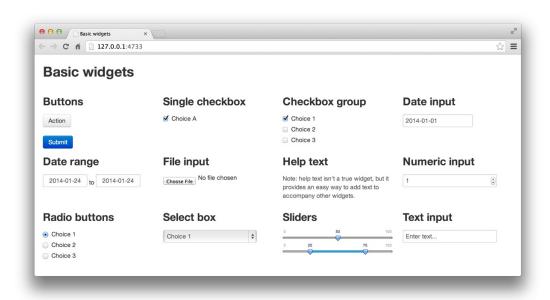
the website people will see and interact with, including inputs and (placeholders for) outputs

Server

takes values from the inputs, does some calculations, and fills in the outputs

Components

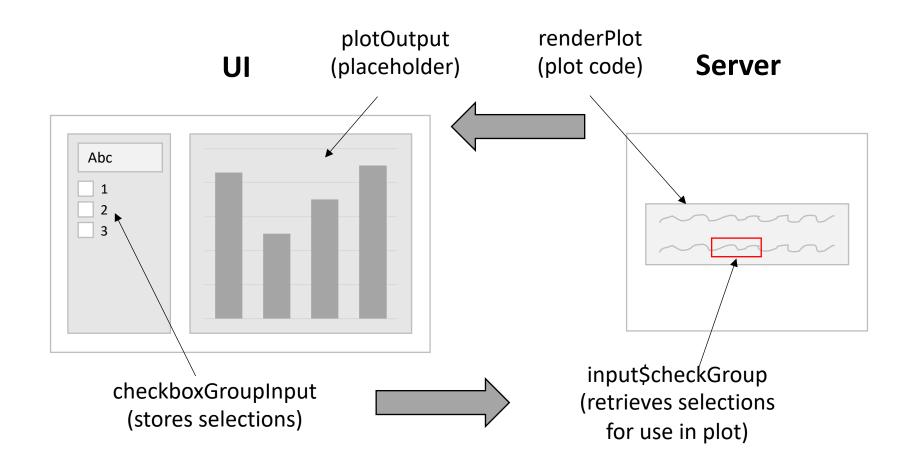
Some kind of **input widget** (e.g., selectInput, sliderInput)



Some kind of **render object** (e.g., renderPlot, renderTable)

renderPlot wraps around something like a ggplot() plot

a plot can read data from the input widget using input\$inputId



Step 1: Create the interface

What to put in the UI?

- Layout containers
- Input widgets
- Placeholders for reactive output
- Extra text/HTML elements

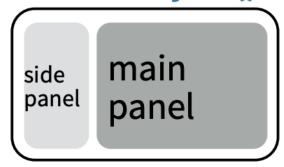


Page layout containers

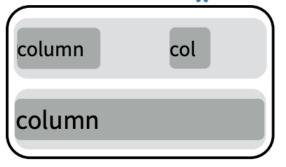
- 1. fluidPage
 - titlePanel
 - sidebarLayout
 - sidebarPanel
 - mainPanel
 - fluidRow
 - column
 - wellPanel
 - tabsetPanel
 - navlistPanel

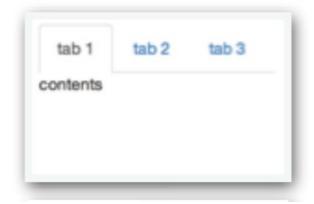
- 2. fixedPage
 - fixedRow
- navbarPage
 - tabPanel
 - navbarMenu
 - tabPanel

sidebarLayout()



fluidRow()



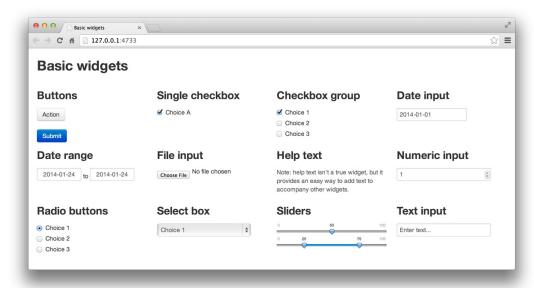


https://shiny.rstudio.com/articles/layout-guide.html

https://www.rstudio.com/resources/cheatsheets/ (Shiny)

Input widgets

- Button
- Checkboxes
- Date, date range input
- File input
- Numeric input
- Radio buttons
- Drop-down (select) box
- Slider bar
- Text input
- Text



http://shiny.rstudio.com/tutorial/written-tutorial/lesson3/http://shiny.rstudio.com/gallery/widget-gallery.html

Anatomy of an input widget

- inputId for the widget (internal only)
- label (will be visible)
- Check documentation for other required arguments (e.g., selectInput requires choices)

Select box

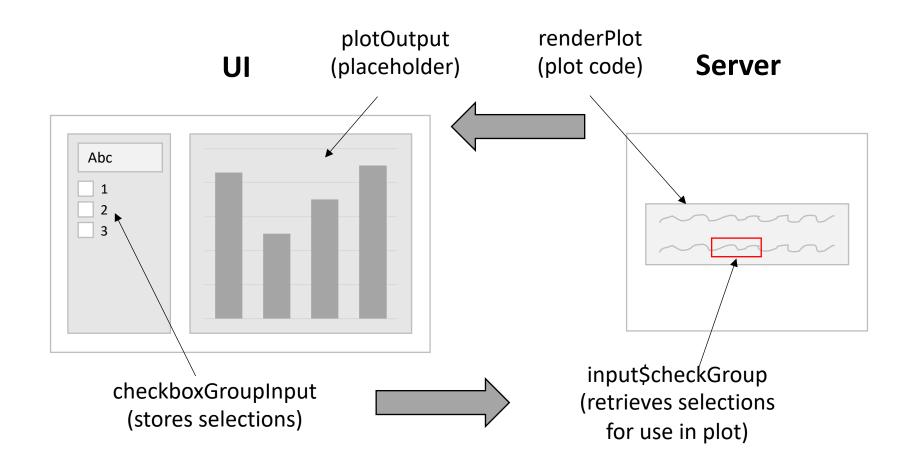
Choice 1

Reactive output objects

UI	Server
htmlOutput	renderUI
imageOutput	renderImage
plotOutput	renderPlot
tableOutput	renderTable
textOutput	renderText
uiOutput	renderUI
verbatimTextOutput	renderPrint

http://shiny.rstudio.com/tutorial/written-tutorial/lesson4/

Step 2: Set up server to create dynamic objects



What to put in the server

- R code
- Render objects with same names and types as the ones listed in UI
- Input objects with the same names as the control widgets

UI:

```
sliderInput("slider1", ...)
textOutput("text1")
```

Server:

```
output$text1 <- renderText({
    input$slider1
})</pre>
```

Step 3: Test

Running the app

Set options in RStudio:

- Window
- Viewer
- External

Sharing an app

- Shiny Apps http://www.shinyapps.io/
- Shiny Server (free host on your own server)
 https://github.com/rstudio/shiny-server/blob/master/README.md
- Shiny Server Pro (fee)
 https://www.rstudio.com/products/shiny/shiny-server/

Exercise 3: Explore the default Shiny app

Create a new app

- File → New File → Shiny Web App...
- Set a name
- Use "Single File" application type
- Look through the code
- Click "Run App" to see the default app

Afternoon Break

Exercise 4: App from scratch

Tips for building your first app

- Start with basic layout elements and static content, like plain text
- Add one output and connect it to something in the server (e.g., plotOutput/renderPlot), but don't use input\$ in the plot yet
- Now create a control and add input\$ to the server code
- You can save individual components as variables and then just use the variable names in your layout, if it gets confusing

Shiny resources

- Shiny Gallery
- Shiny Tutorial
- Shiny Articles
- Shiny function reference
- Shinyapps.io
- RStudio::conf 2019 workshop: <u>Introduction to Shiny and R Markdown</u>
- Shiny in Production (slides), Shiny in Production (book)
- Interactive web-based data visualization with R, plotly, and shiny
- Accessing and responding to plotly events in shiny

Thanks for your time this week!

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