

Introduction to R Markdown, ggplot, & tidyverse

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Today's goals

- Introduce:
 - R Markdown file type
 - Key functions in the tidyverse package
 - The ggplot2 package
- Example code & plot output

What is R Markdown?

- A combination of text and code saved as a .Rmd file
- Uses plain text (with some formatting instructions) plus chunks of code
- Customizable to include text, code, and/or output
- Can render to PDF, HTML, Word, etc.
- Many, many [applications](#)

RMD workflow

- Open a new .Rmd file
- Select the output you want to render, add title, and save file to your working directory
- Write text, titles, section headings, etc.
- Embed code
- Choose output formatting options
- Save & render (to PDF, HTML, etc.)

R Markdown Basics

The screenshot shows the RStudio R Markdown editor. The top toolbar contains icons for navigation, saving, and running. The 'Knit' button (a blue circle with a white 'K') is circled in red. The 'Run' button (a green play icon) is also circled in red. The 'Visual' tab is selected, indicated by a red arrow. The 'Source' tab is also visible. The code editor shows the following content:

```
1 ---  
2 title: "Intro to RMD, ggplot, and tidyverse"  
3 author: "Amanda Loehrke"  
4 date: "06-13-2025"  
5 output: pdf_document  
6 ---  
7  
8 ```{r setup, include=FALSE}  
9 knitr::opts_chunk$set(echo = TRUE)  
10  
11 setwd("C:/Users/amloe/OneDrive/grad school/UC Davis/ICPSR/Intro to R")  
12  
13 library(stargazer)  
14 library(tidyverse)  
15 ```  
16  
17 ## R Markdown Overview  
18  
19 This is a sample of text. Below, you'll see an R code chunk & its  
20 results.  
21 ```{r}  
22 # Set seed for replicability of random sampling  
23 set.seed(1234)  
24  
25 # Set up fake data  
26 x <- rnorm(100)  
27 y <- 2*x + rnorm(100)  
28
```

Red arrows point to the 'Visual' tab, the 'output: pdf_document' line, and the first R code chunk. The second R code chunk is also circled in red, along with its execution icons (gear, play, and a green arrow) in the bottom right corner.

Example output (PDF)

Intro to RMD, ggplot, and tidyverse

Amanda Loehrke

06-13-2025

R Markdown Overview

This is a sample of text. Below, you'll see an R code chunk & its results.

```
# Set seed for replicability of random sampling
set.seed(1234)

# Set up fake data
x <- rnorm(100)
y <- 2*x + rnorm(100)

# Make a dataframe
df <- data.frame(x, y)

# Run a model
model <- lm(y ~ x, df)
summary(model)
```

```
##
## Call:
## lm(formula = y ~ x, data = df)
##
## Residuals:
```

Table 1:

<i>Dependent variable:</i>	
	<i>y</i>
x	1.974*** (0.104)
Constant	0.037 (0.105)
Observations	100
R ²	0.787
Adjusted R ²	0.785
Residual Std. Error	1.037 (df = 98)
F Statistic	361.796*** (df = 1; 98)
<i>Note:</i>	*p<0.1; **p<0.05; ***p<0.01

Including Plots

You can also embed plots, for example:

```
ggplot(df, aes(x, y)) +  
  geom_point()
```



Text formatting (basics)

Code:

```
50
51 ▾ # Main title
52 ▾ ## Section title
53 ▾ ### Subsection title
54
55 *italic text*
56
57 **bold text**
58
59 * This
60 * Is
61 * A List
62
63 1. This
64 2. Is A
65 3. Numbered list
66
67
```

Output:

Main title

Section title

Subsection title

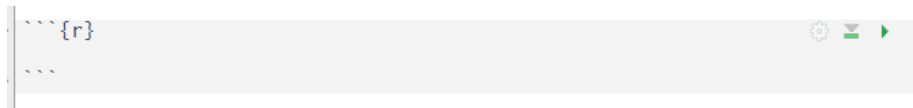
italic text

bold text

- This
- Is
- A List

1. This
2. Is A
3. Numbered list

Code chunks (basics)



There are default options for a code chunk; some common options to change:

- `include` = `FALSE` no code or results will render
- `echo` = `FALSE` results will render, code will not
- `message` = `FALSE` no code generated messages will render
- `warning` = `FALSE` no code generated messages will render
- `error` = `TRUE` display error message in rendered doc

Other options

- Equations with TeX math mode

```
$Y = \beta_1 X_1 + \beta_2 X_2 + \alpha + \epsilon$
```

- ▶ $Y = \beta_1 X_1 + \beta_2 X_2 + \alpha + \epsilon$

- Bibliographies

- ▶ Can add BibTex, Zotero, DOI citations and bibliography

- Output tables

- ▶ Packages like knitr, stargazer, flextable, kableExtra

- Insert a url: `< url >`

- And much, much more!

Introduction to the tidyverse

- This big package includes many important packages like `ggplot2`, `tidyr`, and `dplyr`
- Useful for data wrangling, combining functions, and visualization

Data wrangling with dplyr

- `filter()`
 - ▶ Subset the data, select rows that meet certain criteria
- `select()`
 - ▶ Choose certain variables to keep
- `group_by()`
 - ▶ Prep data for grouped operations/summaries

Wrangling, continued

- `case_when()`
 - ▶ Create conditional values (like nested `ifelse()`)
- `mutate()`
 - ▶ Add or modify variables
- `summarize()`
 - ▶ Calculate summary statistics for one or more variables
- Pipes `%>%`
 - ▶ For a sequence of operations
 - ▶ A useful Base R function used here, too: `%in%`

Example of filter(), select(), and pipes

- Say we want to subset the ANES data to be four feeling thermometers for people in the South:

```
## Filter and select functions
# Without pipes
anes.filtered <- filter(anes.data, region == "South")

anes.filtered <- filter(anes.filtered, pid7 != 8)

anes.filtered <- select(anes.filtered, cand_biden,
                        cand_trump, group_nra, group_blm, pid7)

variable.names(anes.filtered)
head(anes.filtered)
```

Example continued

```
# Now, with pipes
anes.filtered.pipes <- anes.data %>%
  filter(region == "South") %>%
  filter(pid7!= 8) %>%
  select(cand_biden, cand_trump, group_nra, group_blm, pid7)
```

```
# Or this way (just combine two filter arguments into one)
anes.filtered.pipes <- anes.data %>%
  filter(region == "South", pid7!= 8) %>%
  select(cand_biden, cand_trump, group_nra, group_blm, pid7)

variable.names(anes.filtered.pipes)
head(anes.filtered.pipes)
```

Example of `group_by()`, `mutate()`, and `summarize()`

- Say we want the average Trump and Biden Feeling thermometer rating by region (Northeast, South, Midwest, West)
- We need to clean the data (get rid of "No Answer" and "not asked" rows), break up the data by region, and calculate the average feeling thermometer ratings for each region
- To do this with Base R:

Base R Example

```
# In base R:
anes.data$cand_biden_clean <- ifelse(anes.data$cand_biden == "No Answer", NA,
                                     ifelse(anes.data$cand_biden == "not asked", NA,
                                             anes.data$cand_biden))

anes.data$cand_biden_clean <- as.numeric(anes.data$cand_biden_clean)

anes.data$cand_trump_clean <- ifelse(anes.data$cand_trump == "No Answer", NA,
                                     ifelse(anes.data$cand_trump == "not asked", NA,
                                             anes.data$cand_trump))

anes.data$cand_trump_clean <- as.numeric(anes.data$cand_trump_clean)

# Check cleaned data
table(anes.data$cand_biden_clean)
table(anes.data$cand_trump_clean)

# Find average ft score by region
aggregate(cbind(cand_biden_clean, cand_trump_clean) ~ region,
          data = anes.data,
          FUN = mean,
          na.rm = TRUE)
```

tidyverse example

```
# With dplyr:
anes.clean <- anes.data %>%
  filter(!cand_biden %in% c("No Answer", "not asked")) %>%
  filter(!cand_trump %in% c("No Answer", "not asked")) %>%
  mutate(cand_biden = as.numeric(cand_biden),
         cand_trump = as.numeric(cand_trump))

# Check cleaned data
table(anes.clean$cand_biden)
table(anes.clean$cand_trump)

# Find average ft score by region
anes.clean %>%
  group_by(region) %>%
  summarize(avg.bidenft = mean(cand_biden),
           avg.trumpft = mean(cand_trump))
```

R output from each example

- With Base R:

region <chr>	cand_biden_clean <dbl>	cand_trump_clean <dbl>
Midwest	35.00522	45.06005
Northeast	46.01767	43.95053
South	40.81639	46.39302
West	45.70833	40.72135

- With tidyverse:

region <chr>	avg.bidenft <dbl>	avg.trumpft <dbl>
Midwest	35.00522	45.06005
Northeast	46.01767	43.95053
South	40.81639	46.39302
West	45.70833	40.72135

The ggplot2 package

- "Grammar of graphics"
- Think about grammar in language: it is a set of rules that govern how language is structured (words, phrases, clauses)
- Used to produce layered, statistical graphics
- Uses a "grammar" to build graphs layer-by-layer
- For more package information, click [here](#).

Grammar of graphics (basics)

- **Data:** what information to map
- **Aesthetics:** which data is on your x-axis? y-axis? color? size? fill?
- **Geometries:** the type of plot you want to create (scatterplot, bar plot, time series, etc.)
- **Scales:** mappings of aesthetic values to data values
- **Stats:** transformations with statistics to summarize data
- **Faceting:** splitting data to create multiple variations of the same graph

Using ggplot()

- All graphics begin with the `ggplot()` function
- In this function, we specify the dataset and variables to map to the aesthetics

```
ggplot(data, aes(x = variable1, y = variable2))
```

- **Data**: name of the data frame object that holds the variables we want to plot
- **X and Y**: aesthetics that position objects on the graph
- Running this will produce a plot with x and y axes, no data will be plotted

Aesthetics

- Commonly used aesthetics:
 - ▶ **x**: positioning on the x-axis
 - ▶ **y**: positioning on the y-axis
 - ▶ **color**: color of the data objects
 - ▶ **fill**: fill color of points, boxes, etc.
 - ▶ **linetype**: solid, dashed, dotted, etc.
 - ▶ **shape**: circle, square, solid, etc.
 - ▶ **size**: how large objects appear
 - ▶ **alpha**: transparency of objects

Using `ggplot()` + `geom_*`

- We add layers with `+` and our next layer will be the geometry
`ggplot(data, aes(x)) + geom_*()`
- You'll specify the type of geometry you want, based on the type of plot you want to create (scatterplot, histogram, barplot, etc.)
- The geometry will take the mapping from `ggplot()` but can also be overwritten in `geom_*(aes(variable))`
 - ▶ Can specify line type, colors, sizes, etc.

Geometries

- * There are many geometries available, some common ones are:
 - ▶ `geom_histogram()`
 - ▶ `geom_density()`
 - ▶ `geom_boxplot()`
 - ▶ `geom_bar()`
 - ▶ `geom_point()`
 - ▶ `geom_line()`

Using the stats layer

- Can map on statistical functions to transform data, just add a layer with `+`
- Each stat function is associated with a geometry, so they can be used without specifying a geometry
- Ex) `stat_summary()` will give a point for the mean and error bars
- Many ways to use this layer depending on what you are trying to do!

Using the scales layer

- Scales define which aesthetic values are mapped to the data values
- This layer is added to `ggplot()` and the geometry layer

```
ggplot(data, aes(x)) +  
  geom_bar() +  
  scale_*()
```
- Many different scaling layers depending on what you are trying to accomplish

Scales

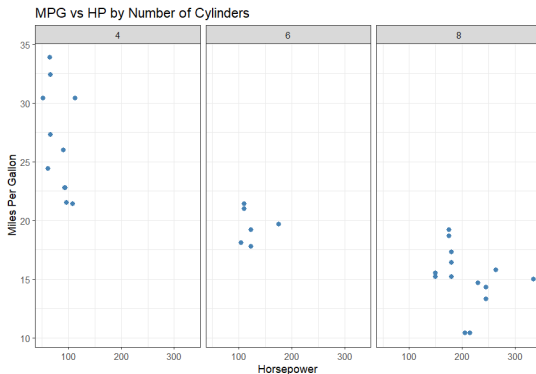
- Specify that you are using scale, then name the primary aesthetic (x, y, color, etc.), then name the scale you want to use (continuous, discrete, etc.)
- Some common scale specifications
 - ▶ `scale_x_continuous()`
 - ▶ `scale_y_continuous()`
 - ▶ `scale_color_manual()`
- Axis scales specify the breaks, labels, and names

More on axes, limits, and labeling

- Can also control the limits and titles of the axes using shortcut functions instead of scaling
- Some examples:
 - ▶ `lims()`: set axis limits
 - ▶ `xlim()`: set x axis limit
 - ▶ `ylim()`: set y axis limit
 - ▶ `xlab()`: label x axis
 - ▶ `ylab()`: label y axis
 - ▶ `ggtitle()`: label title
 - ▶ Or, use `lab()` to specify each axis label, title, subtitle, etc.
`lab(title = "Insert your title here", x = "X
axis label", y = "Y axis label")`

Faceting

- Used to split plots into multiple panels
- `facet_wrap()` and `facet_grid()`



Themes and colors

- Themes control the parts of the graph not related to the data (background color, font size, gridlines, label colors, etc.)
- Can set one theme for the entire plot ([complete themes](#))
- Can modify specific components of a theme ([modify components](#))
- There are various default colors in R, for a full list, click [here](#)

Interactive plots

- You can make interactive charts with ggplot
- Also uses the package plotly
- Zoom, hover, etc.
- This [website](#) is a great introduction to interactive plots in R
- Or [this website](#) that introduces dynamic, interactive plots with the ggiraph package

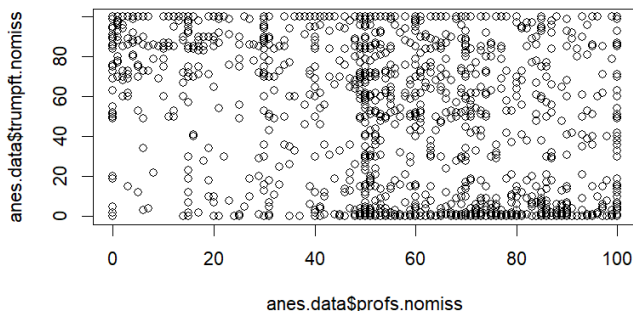
Saving plots to files

- 1 Export from the Plots tab in RStudio
- 2 Right-click on an image and save it
- 3 Use `ggsave()` in your code to save the plot object to your working directory
 - Easy to change the default width, height, filename, etc.

```
myplot <- ggplot(data, aes(x)) + geom_hist()  
ggsave("myplot.png", plot = myplot, width = 10,  
height = 6)
```

Example scatterplot

- Remember when we made a scatterplot with `profs.nomiss` and `trumpft.nomiss`? It looked like this in Base R:

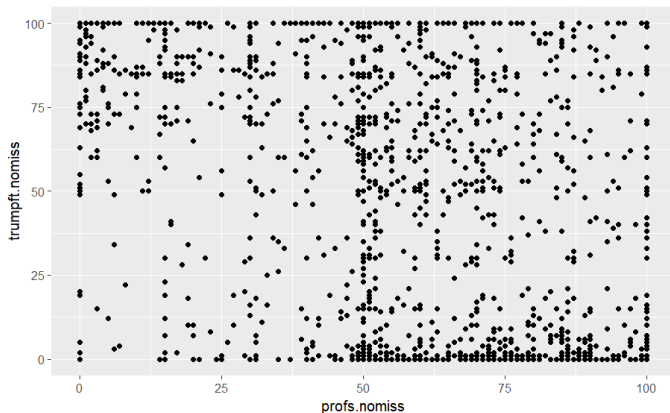


- We can make a similar plot using `ggplot`

Scatterplot example default

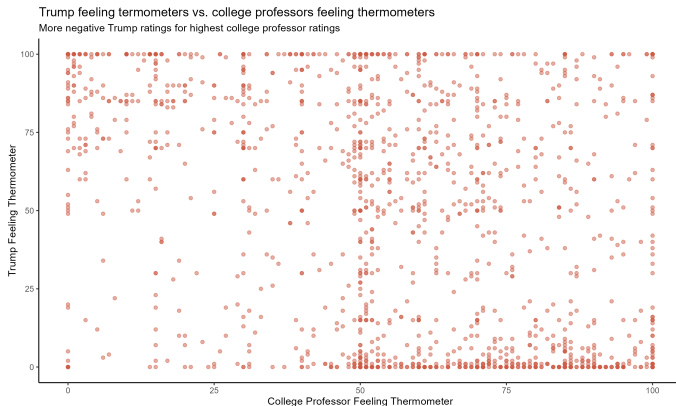
```
# Plot with tidyverse
anes.data <- anes.data %>%
  mutate(profs.nomiss = ifelse(group_colprofs %in% c(999, -7), NA, group_colprofs),
         trumpft.nomiss = ifelse(cand_trump %in% c(999, -7), NA, cand_trump))

ggplot(anes.data, aes(x = profs.nomiss, y = trumpft.nomiss)) +
  geom_point()
```



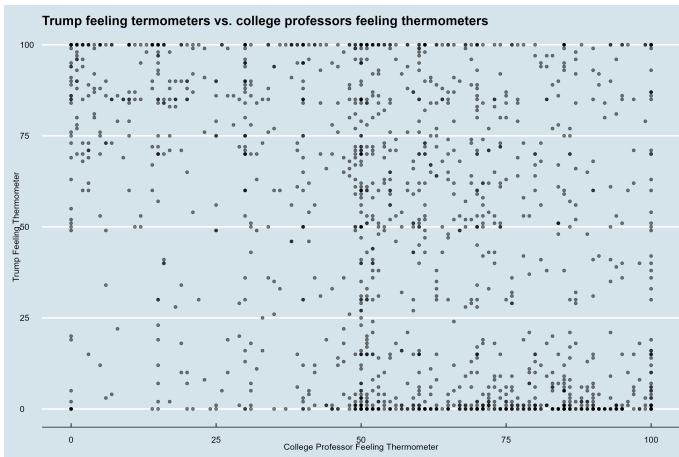
Scatterplot example stylized

```
prof.trump.plot <- ggplot(anes.data, aes(x = profs.nomiss, y = trumpft.nomiss)) +  
  geom_point(alpha = 0.5, color = "coral3") +  
  labs(title = "Trump feeling thermometers vs. college professors feeling thermometers",  
        subtitle = "More negative Trump ratings for highest college professor ratings",  
        x = "College Professor Feeling Thermometer",  
        y = "Trump Feeling Thermometer") +  
  theme_classic()  
  
ggsave("ggplot_ft.png", plot = prof.trump.plot, width = 10, height = 6)
```



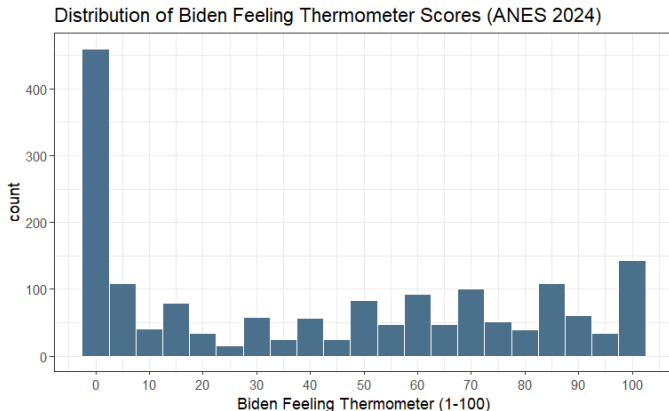
Scatterplot example, continued

```
econ.plot <- ggplot(anes.data, aes(x = profs.nomiss, y = trumpft.nomiss)) +  
  geom_point(alpha = 0.5) +  
  labs(title = "Trump feeling thermometers vs. college professors feeling thermometers",  
       x = "College Professor Feeling Thermometer",  
       y = "Trump Feeling Thermometer") +  
  theme_economist() +  
  scale_color_economist()
```



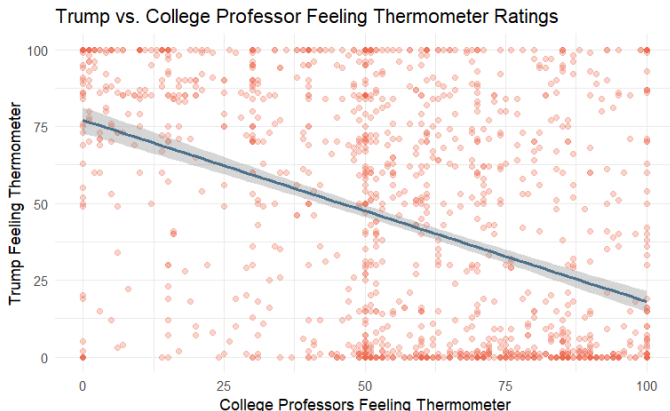
Histogram example

```
# Histogram of Biden feeling thermometer
ggplot(anes.clean, aes(cand_biden_clean)) +
  geom_histogram(binwidth = 5, fill = "skyblue4", color = "white") +
  labs(title = "Distribution of Biden Feeling Thermometer Scores (ANES 2024)",
       x = "Biden Feeling Thermometer (1-100)") +
  scale_x_continuous(breaks = seq(0, 100, 10)) +
  theme_bw()
```



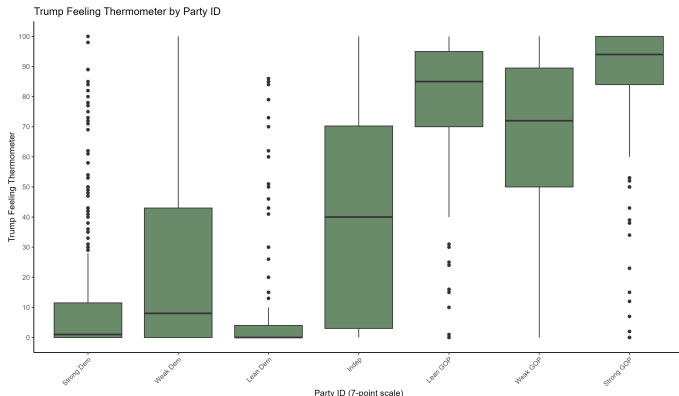
Regression line scatter plot example

```
ggplot(anes.clean, aes(x = group_colprofs, y = cand_trump)) +  
  geom_point(alpha = 0.3, color = "coral2") +  
  geom_smooth(method = "lm", se = TRUE, color = "skyblue4", size = 1) +  
  labs(title = "Trump vs. College Professor Feeling Thermometer Ratings",  
        x = "College Professors Feeling Thermometer",  
        y = "Trump Feeling Thermometer") +  
  theme_minimal()
```



Box plot example

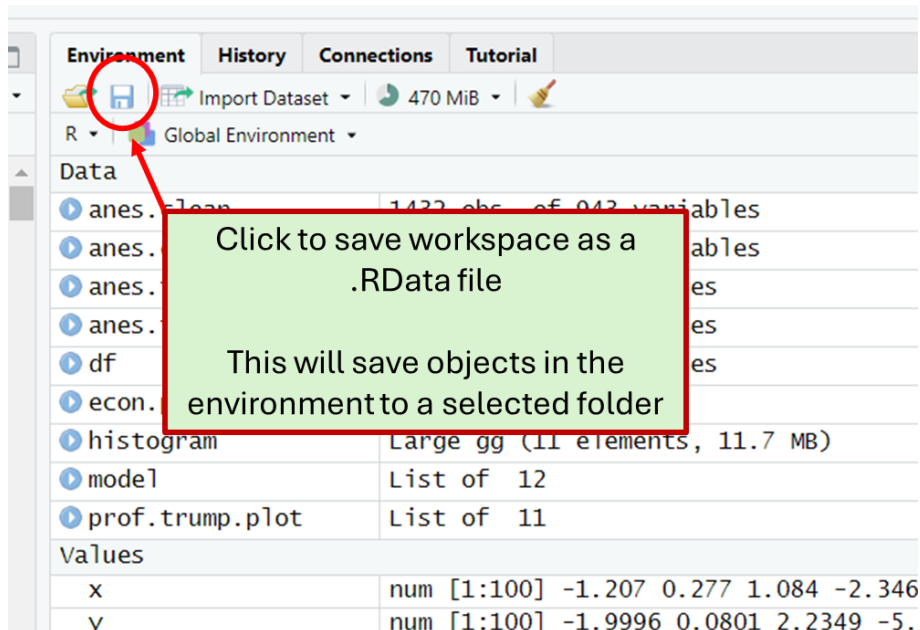
```
histogram <- ggplot(anes.clean, aes(x = pid7, y = cand_trump)) +  
  geom_boxplot(fill = "darkseagreen4") +  
  labs(title = "Trump Feeling Thermometer by Party ID",  
        x = "Party ID (7-point scale)",  
        y = "Trump Feeling Thermometer") +  
  scale_y_continuous(breaks = seq(0, 100, 10)) +  
  scale_x_discrete(labels = c("Strong Dem", "Weak Dem", "Lean Dem", "Indep",  
                              "Lean GOP", "Weak GOP", "Strong GOP")) +  
  theme_classic() +  
  theme(axis.text.x = element_text(angle = 45, hjust = 1))  
qsave("histogram_example.png", plot = histogram, width = 12)
```



Some ggplot considerations

- You can make a plot do (probably) anything you want it to
- Think:
 - ▶ What dataset am I plotting from?
 - ▶ What variables do I want to visualize?
 - ▶ What type of plot do I want to create?
 - ▶ What I can do to make the plot look nicer?
 - ★ Labels, colors, scaling, etc.
- Try different styles and layers until you produce a clear, professional-style plot that presents the data accurately & meaningfully

Saving environment (.RData)



The screenshot shows the RStudio interface with the 'Environment' tab selected. The 'Save Workspace To...' button (represented by a floppy disk icon) is circled in red. A red arrow points from this button to a green text box with the following text:

Click to save workspace as a .RData file

This will save objects in the environment to a selected folder

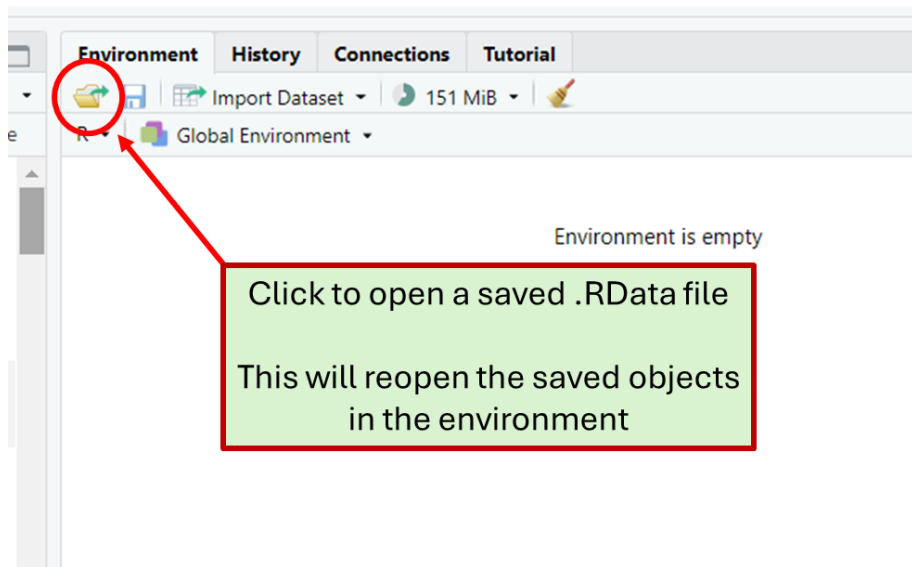
The Environment pane lists the following objects:

Object	Details
anes.lean	1422 obs. of 942 variables
anes.	ables
anes.	es
anes.	es
df	
econ.	
histogram	Large gg (11 elements, 11.7 MB)
model	List of 12
prof.trump.plot	List of 11

The Values pane shows the following data:

Variable	Value
x	num [1:100] -1.207 0.277 1.084 -2.346
y	num [1:100] -1.9996 0.0801 2.2349 -5.

Opening environment (.RData)



The screenshot shows the RStudio interface with the 'Environment' tab selected. The toolbar contains icons for 'Open' (a folder with a green arrow), 'Save' (a floppy disk), and 'Import Dataset' (a grid with a green arrow). The 'Open' icon is circled in red. Below the toolbar, the 'Global Environment' is listed. A red arrow points from the 'Open' icon to a text box.

Environment is empty

Click to open a saved .RData file

This will reopen the saved objects
in the environment

Opening RMD file in R Studio

- Open the file on Canvas, save in folder, open in R Studio
- Set your working directory (Session → Set Working Directory → Choose Directory)
- Download tinytex package in order to render to a PDF file
`install.packages("tinytex")`
`tinytex::install_tinytex()`
- Make sure `anes_pilot_2024_20240319.csv` and `anes_pilot_2024_20240319.dta` are in your working directory

Final thoughts

- When in doubt, ask questions!
- There are endless resources online and in print about R, R Studio, R Markdown, specific packages, etc.
- Talk to colleagues, instructors, TAs, etc.
- Coding in R is like learning a new language, so it takes practice

(Some) Resources

- Data Analysis for Social Science (Llaudet & Imai)
- R Workflow
- R Markdown, RMD Cheatsheet
- ggplot2
- Tidyverse & more
- DataCamp

Thank you! Feel free to send me an email with any questions/comments!

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Appendix

- (1) Example plots
- (2) Intro to R and R Studio

Examples

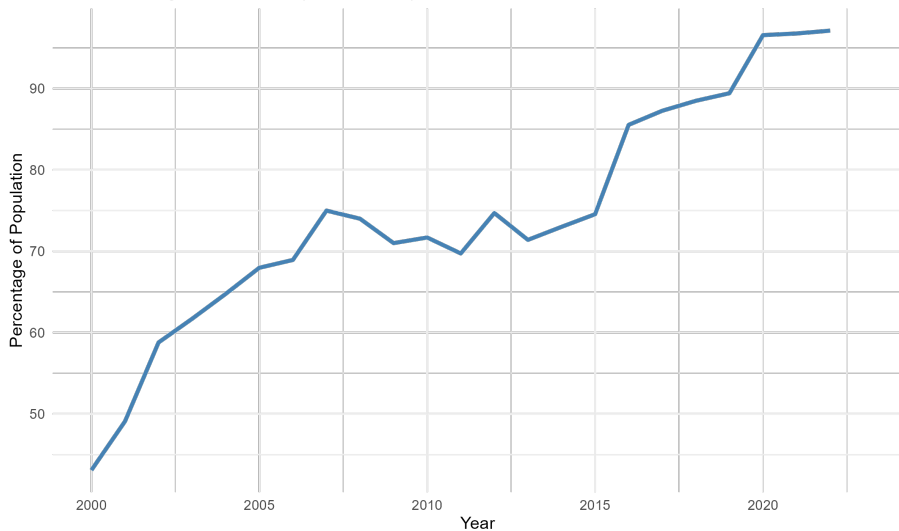
```
# Time series for US internet usage
us <- df %>% filter(Country.Name == "United States")

timeseries <- ggplot(us, aes(x = Year, y = Percentage)) +
  geom_line(color = "steelblue", size = 1.2) +
  labs(
    title = "Internet Usage in the U.S. (2000-2015)",
    x = "Year",
    y = "Percentage of Population"
  ) +
  theme_minimal()

timeseries

ggsave("timeseries.png", width = 8, height = 5, plot = timeseries)
```

Internet Usage in the U.S. (2000–2015)

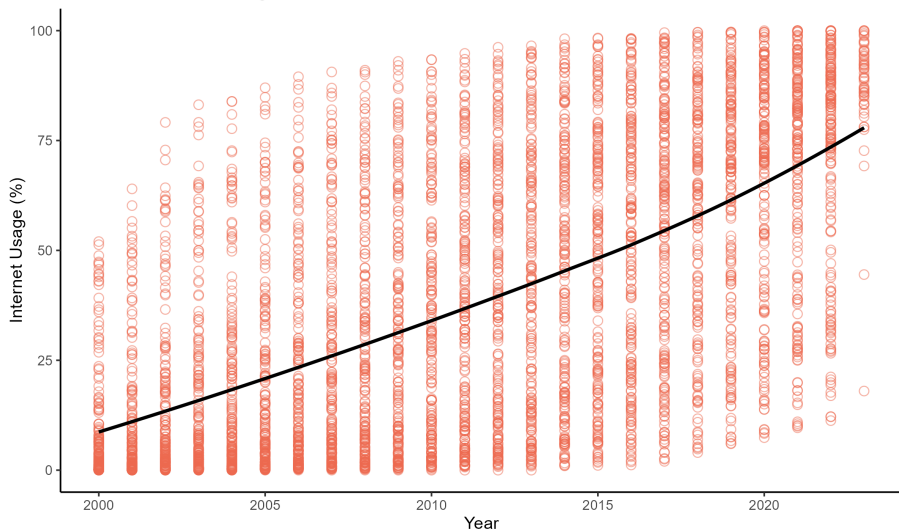


```
# Scatterplot with trendline
scatter <- ggplot(df, aes(x = Year, y = Percentage)) +
  geom_point(color = "coral2", size = 2.5, shape = 1, alpha = 0.5) +
  geom_smooth(method = "loess", color = "black", se = FALSE) +
  labs(
    title = "Trend of Internet Usage",
    x = "Year",
    y = "Internet Usage (%)"
  ) +
  theme_classic()

scatter

ggsave("scatter.png", width = 8, height = 5, plot = scatter)
```

Trend of Internet Usage



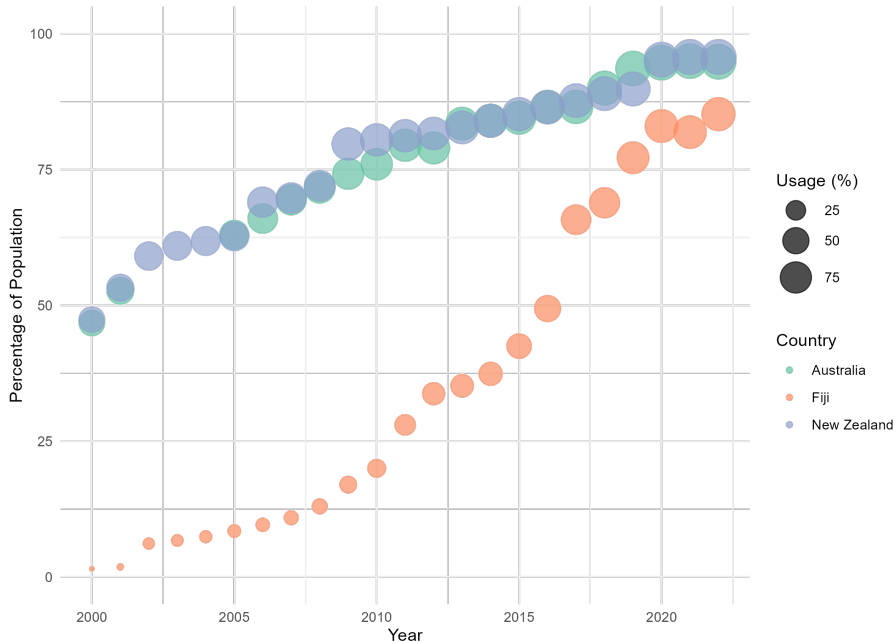
```
# Bubble plot for usage over time, three countries
clean <- df %>%
  filter(Country.Name %in% c("Australia", "New Zealand", "Fiji")) %>%
  na.omit()

bubble <- ggplot(clean, aes(x = Year, y = Percentage,
                           color = Country.Name, size = Percentage)) +
  geom_point(alpha = 0.7) +
  scale_size(range = c(1, 10), name = "Usage (%)") +
  scale_x_continuous(by = 5) +
  scale_color_brewer(palette = "Set2") +
  labs(
    title = "Internet Usage Over Time by Country",
    x = "Year",
    y = "Percentage of Population",
    color = "Country"
  ) +
  theme_minimal()

bubble

ggsave("bubble.png", width = 8, height = 6, plot = bubble)
```

Internet Usage Over Time by Country



R and RStudio

- RStudio is a user-friendly graphical interface for R
- Open source, packages for almost everything (data processing, cleaning, visualization, etc.)
- For help, click [here](#).

Downloading R and RStudio

- Downloading R
 - ▶ R for Windows: <https://cran.r-project.org/bin/windows/base/>
 - ▶ R for macOS: <https://cran.r-project.org/bin/macosx/>
- Downloading R Studio
 - ▶ RStudio: <https://posit.co/downloads/>
- Open the downloaded file (check your **Downloads** folder), click yes through all of the prompts to install like any other program

Downloading R and RStudio

- Downloading R
 - ▶ R for Windows: <https://cran.r-project.org/bin/windows/base/>
 - ▶ R for macOS: <https://cran.r-project.org/bin/macosx/>
- Downloading R Studio
 - ▶ RStudio: <https://posit.co/downloads/>
- Open the downloaded file (check your **Downloads** folder), click yes through all of the prompts to install like any other program
- *If* your computer is an older model, or you are using an iPad or similar, you can access RStudio online:
 - ▶ Go to <https://posit.cloud/>
 - ▶ Create an account using your UC Davis email
 - ▶ Click on "New Project" and select "New Project from Template"
 - ▶ Select "Data Analysis in R with the Tidyverse and click "OK"

Getting Started With RStudio

- Create a folder on your computer where you want to save your files
 - ▶ This is where you will store your R files, download data to use in R, saved visualization, tables, etc.
- Open R Studio
 - ▶ Create a new R script and save to your folder
 - ▶ Set working directory to your folder in R script
 - ★ It will look something like this:

```
setwd("C:/Desktop/R Folder")  
getwd()
```
- For a helpful RStudio tutorial, click [here](#).

More on R Studio

1. script/source
This is where you write and edit code.
To evaluate your code, you have to click "Run" to evaluate them in the console.

2. environment
In here you can see the objects in your working space (environment) or view your command history (history)

3. console
This is where the script code is evaluated by R. You can also perform calculations, etc. in here that you don't need to save in your script.

4. files/plots/packages/help
Here you can see file directories, view plots you produce, install and update packages, and access R help.
R help is very useful tool! If you need help you can search for a function or type "?functionname" in the console to see that function's help page.

```
library(tidyverse)
library(unvotes)
library(lubridate)
library(scales)

# Make a plot --
# calculate % votes yes each year by country by issue
un_yes = un_votes %>%
  filter(country != "United States") %>%
  inner_join(un_votes, by = "year") %>%
  inner_join(un_votes, by = "year") %>%
  group_by(country, year) %>%
  summarize(
    votes = n(),
    percent_yes = sum(issue == "Yes") / votes
  ) %>%
  filter(votes > 5) # only use points where there are more than 5 votes

# make plot
ggplot(un_yes, aes(x = year, y = percent_yes, color = country)) +
  geom_point() +
  geom_smooth(method = "lm", se = FALSE) +
  facet_wrap(~ issue, scales = "y") +
  labs(
    title = "Percent Yes by Year and Issue",
    subtitle = "1948-1994",
    y = "% Yes",
    x = "Year",
    color = "Country"
  ) +
  scale_y_continuous(labels = percent) +
  geom_smooth() using formula = 'y ~ x'
```