

Introduction to the course (and R) and data visualization

Seung-Ho An, University of Arizona

- ① Introduction to the course
- ② Course details
- ③ R and Rmarkdown introduction
- ④ Data visualization

1. Introduction to the course

- Why do we need to learn statistics?
- Applied statistics? Data science?

Why do we need to learn statistics?

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1. To test ideas empirically

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2. To make inferences from sample to population

Why do we need to learn statistics?

1. To test ideas empirically
2. To make inferences from sample to population
3. To make evidence-based decisions/policy

Applied statistics? Data science?

- Applied statistics includes planning for the collection of data, managing data, analyzing, interpreting and drawing conclusions from data, and identifying problems, solutions and opportunities using the analysis (UM-Dearborn)

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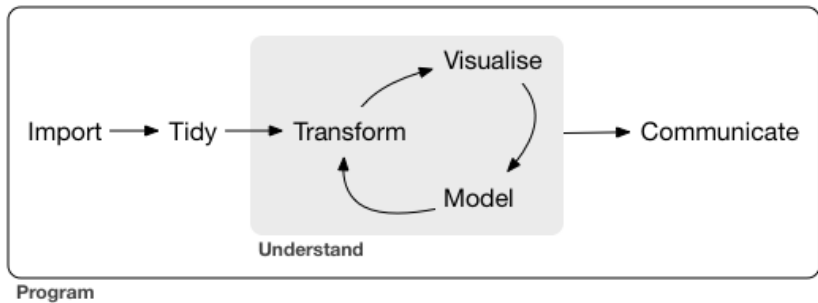


Figure 1: Data science process (Hadley)

The mix of applied statistics and data science, slightly leaning more toward data science

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E.g., calculating a sample variance of a set $\{2, 7, 7, 4, 5, 1, 3\}$

The mix of applied statistics and data science, slightly leaning more toward data science

E.g., calculating a sample variance of a set $\{2, 7, 7, 4, 5, 1, 3\}$

Instead of this,

$$\begin{aligned}
 s^2 &= \frac{\sum_{i=1}^n (x_i - \bar{x})^2}{n - 1} \\
 &= \frac{\sum_{i=1}^n x^2 - \frac{(\sum_{i=1}^n x)^2}{n}}{n - 1} \\
 &= \frac{(4+49+49+16+25+1+9) - \frac{(2+7+7+4+5+1+3)^2}{7}}{7-1} = \frac{153 - \frac{841}{7}}{6} \\
 &= 5.47619
 \end{aligned}$$

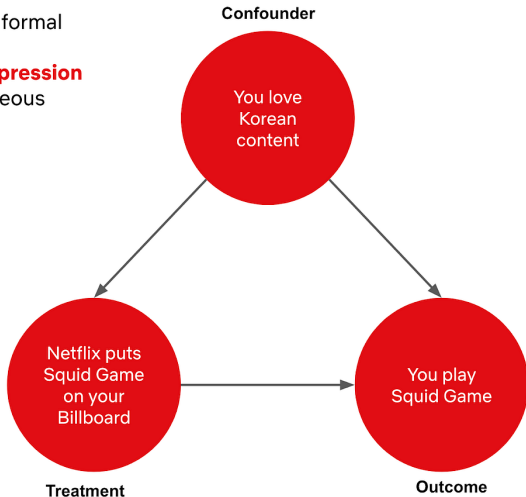
We will focus more on

```
x<-c(2, 7, 7, 4, 5, 1, 3)  
var(x)
```

```
## [1] 5.47619
```

What problems are applied statisticians/data scientists working on?

Causal Inference provides formal tools to tease out the true **incremental** value of an **impression** for each profile: Heterogeneous Treatment Effect (**HTE**)



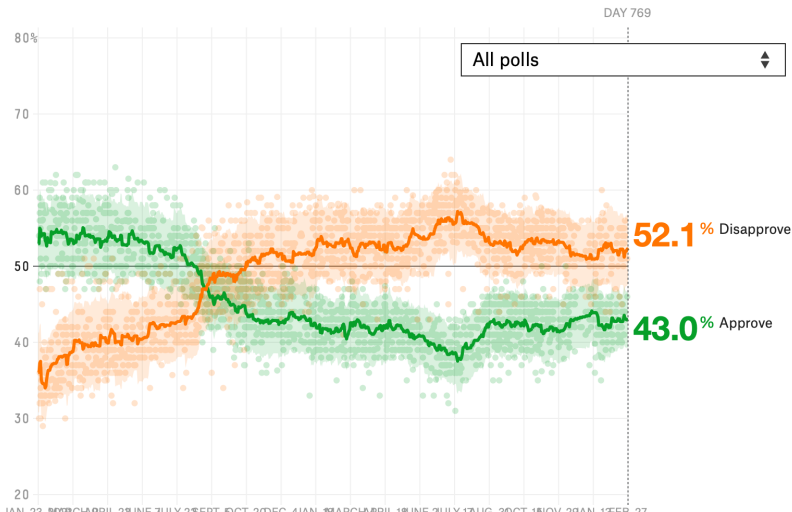
A new computer model uses publicly available data to predict crime accurately in eight U.S. cities, while revealing increased police response in wealthy neighborhoods at the expense of less advantaged areas.

June 30, 2022



How **popular** is Joe Biden?

An updating calculation of the president's approval rating, accounting for each poll's quality, recency, sample size and partisan lean. [How this works »](#)



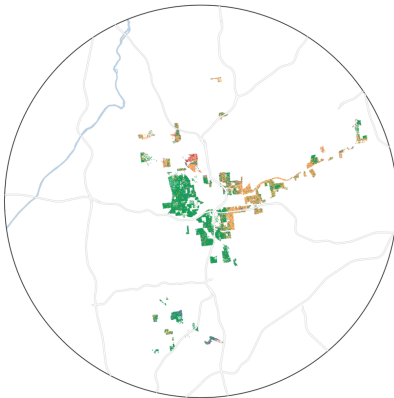


Understanding how the past matters

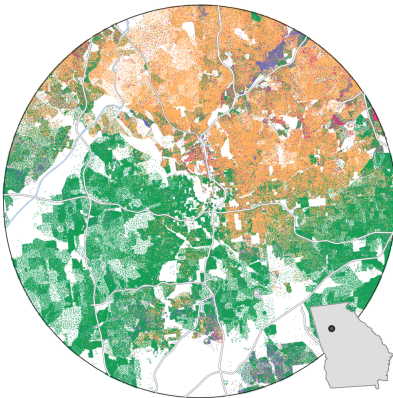
Atlanta, GA

WHITE BLACK LATINO ASIAN OTHER

ATLANTA'S "HAZARDOUS" ZONES:



ATLANTA'S SURROUNDING AREA:



2. Course details

Name, N of previous stat/programming classes, main statistical program/package (if any), etc.

What will you learn in this class?

- Summertime and visualize data

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- Summarize and visualize data
- Wrangle messy data into tidy forms

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- Understand uncertainty in data analysis and how to quantify it

What will you learn in this class?

- Summarize and visualize data
- Wrangle messy data into tidy forms
- Evaluate claims about causality
- Be able to use linear regression to analyze data
- Understand uncertainty in data analysis and how to quantify it
- Use professional tools like R and RStudio

- Deliberate pacing and **tons** of support

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- Emphasize intuition and computational approaches over mathematical equations

- Deliberate pacing and **tons** of support
- Emphasize intuition and computational approaches over mathematical equations
- Practice, practice, practice

How to become a better programmer



Hadley Wickham 

@hadleywickham

...

The only way to write good code is to write tons of shitty code first.
Feeling shame about bad code stops you from getting to good code

7:11 AM · Apr 17, 2015

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@hadleywickham



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7:11 AM · Apr 17, 2015



Allison Horst

@allison_horst



Troubleshooting lessons I guess I'll just relearn forever:

- take a break
- it's almost certainly not a bug
- extra eyes are awesome
- spelling

- **None** (no prior programming or statistics knowledge)
 - I am excited to teach this class with a room full of individuals who are just enthusiastic about data and analyses as much as I am!

- Three text book:
 - Modern Dive (MD) (free online)
 - Quantitative Social Science (QSS): An Introduction in tidyverse by Kosuke Imai (not free)
 - Introduction to Modern Statistics (IDS) (free online)
- Sometimes same materials in two/three different books. Choose which helps most.

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- Lectures:
 - Broad coverage of the course materials
 - Coding demonstrations (follow along with your laptop) (if any)
 - Slides will be posted to the website shortly (or the night) before lecture

- Roughly weekly homework
 - It will be posted on Thursday morning, and due following Wednesday
 - Your assignments will not be graded but highly encourage finishing those!
- Final project (optional), though I highly encourage everyone to do so!
 - Especially if you find assignments too easy, this would be an opportunity to practice your data/analysis skills (of course with tons of my support/assistance)! At the end, I will provide brief comments on your project.

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 - Especially if you find assignments too easy, this would be an opportunity to practice your data/analysis skills (of course with tons of my support/assistance)! At the end, I will provide brief comments on your project.
- Another option for advanced individuals: after a couple of assignments, if the assignments are still easy, let me know; I will assign relevant exercises from the QSS book.

- We will use the R statistical environment to analyze data
 - It's free
 - A popular option for data analysis
 - Academics, 538, NYT, Facebook, Google, Twitter, nonprofits, governments all use R
- Interface with R via a program called R studio

3. R and Rmarkdown introduction

Two computer revolutions



The frontier of computing

- Touch-based interfaces



Where statistical computing lives

Two computer revolutions



The frontier of computing

- Touch-based interfaces
- Single app at a time



Where statistical computing lives

Two computer revolutions



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- Touch-based interfaces
- Single app at a time
- Little multitasking between apps



Where statistical computing lives

Two computer revolutions



The frontier of computing

- Touch-based interfaces
- Single app at a time
- Little multitasking between apps
- Hides the file system



Where statistical computing lives

- Windows and pointers

Two computer revolutions



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- Windows and pointers
- Multi-tasking, multiple windows

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Where statistical computing lives

- Windows and pointers
- Multi-tasking, multiple windows
- Works heavily with the file system
- Underneath it's UNIX and the command line



- often free, open-sourced, and powerful

For more info, visit
<https://plain-text.co/>



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- So why use them?

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**The process of applied
statistics/data science is
intrinsically messy**

Office vs. engineering model of computing

What's real in the project? How are changes managed?

In the Office model

- Formatted documents are real

In the Engineering model

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- Plain-text files are real
- Intermediate outputs are produced via code, often inside documents

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In the Engineering model

- Plain-text files are real
- Intermediate outputs are produced via code, often inside documents
- Changes are tracked outside files
- Final outputs are assembled programmatically and converted to desired output format

Pros and cons to each approach

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 - Everyone knows Word, Excel, Google Docs, etc.

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 - Why won't R just do what I want!
 - Version control is a pain

We'll tend toward the Engineering model because it's better suited to keep the mess in check

Let's take a tou**R**



R version 4.2.2 (2022-10-31) -- "Innocent and Trusting"
Copyright (C) 2022 The R Foundation for Statistical Computing
Platform: aarch64-apple-darwin20 (64-bit)

R is free software and comes with ABSOLUTELY NO WARRANTY.
You are welcome to redistribute it under certain conditions.
Type 'license()' or 'licence()' for distribution details.

Natural language support but running in an English locale

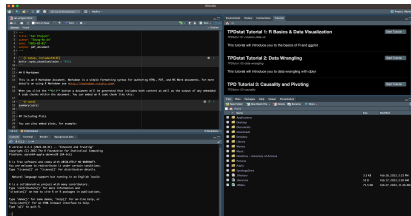
R is a collaborative project with many contributors.
Type 'contributors()' for more information and
'citation()' on how to cite R or R packages in publications.

Type 'demo()' for some demos, 'help()' for on-line help, or
'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

[R.app GUI 1.79 (8160) aarch64-apple-darwin20]

[Workspace restored from /Users/seunghoan/.RData]

> |



R vs. RStudio (cont)



The screenshot shows the RStudio IDE interface. The main editor displays a new R Markdown document titled "car-project.Rmd". The document content includes a YAML header, a title, author, date, and output format, followed by an R code chunk for setting up chunk options and a paragraph of text about R Markdown. The console at the bottom shows the R version and copyright information. The sidebar on the right contains a list of tutorials and a file explorer.

```

1 ---
2 title: "car Project"
3 author: "Sergio M. R."
4 date: "2023-01-01"
5 output: pdf_document
6 ---
7
8 <!--[r setup: includeRMarkdown]-->
9 knitr::opts_chunk$set(echo = TRUE)
10 ---
11
12 ## R Markdown
13
14 This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more
15 details on using R Markdown see <http://rmarkdown.rstudio.com>.
16
17 When you click the "Knit" button a document will be generated that includes both content as well as the output of any embedded
18 R code chunks within the document. You can embed an R code chunk like this:
19
20 <pre># R code chunk</pre>
21
22 ## Including Plots
23
24 You can also embed plots, for example:
25
26 <pre># R Markdown</pre>

```

Console Terminal Render Background Jobs

R 4.2.2 -- R

R version 4.2.2 (2022-10-31) -- "Innocent and Trusting"
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TPDstat Tutorial 1: R Basics & Data Visualization
TPDstat 01-r-basics-data-viz
This tutorial will introduce you to the basics of R and ggplot

TPDstat Tutorial 2: Data Wrangling
TPDstat 02-data-wrangling
This tutorial will introduce you to data wrangling with dplyr

TPD Tutorial 3: Causality and Pivoting
TPDstat 03-causality

File Plots Packages Help Viewer Presentation

New Folder New Blank File Delete Rename More

Name	Size	Modified
Applications		
Desktop		
Documents		
Downloads		
Dropbox		
Library		
Movies		
Music		
OneDrive - University of Arizona		
Pictures		
Public		
SynologyDrive		
Thumbnails	3.3 KB	Feb 28, 2023, 5:33 PM
Workspaces	52 B	Feb 17, 2023, 2:30 AM
Workspaces	71.5 KB	Feb 27, 2023, 11:26 AM

The screenshot displays the RStudio interface with a new R Markdown document. The editor window shows the following R code:

```
1 # ---
2 title: "New Project"
3 author: "Sergio M. R."
4 date: "2023-01-01"
5 output: pdf_document
6 # ---
7
8 #>>>[r setup::includeRMarkdown]
9 knitr::opts_chunk$set(echo = TRUE)
10 #>>>
11
12 ## R Markdown
13
14 This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more
15 details on using R Markdown see <https://rmarkdown.rstudio.com>.
16
17 When you click the "Knit" button a document will be generated that includes both content as well as the output of any embedded
18 R code chunks within the document. You can embed an R code chunk like this:
19
20 ```{r cars}
21 summary(cars)
22 ```
23
24 ## Including Plots
25
26 You can also embed plots, for example:
27
28 ```{r}
29 # R Markdown
30 ```
```

The right pane shows the R Markdown output, which includes a title "TPDstat Tutorial 1: R Basics & Data Visualization" and a brief introduction to R and ggplot. The bottom pane shows the R console with the R version and copyright information.

The screenshot displays the RStudio IDE interface. The top-left pane shows a Markdown document titled "car-project.Rmd" with the following content:

```
1 ---
2 title: "Car Project"
3 author: "Sergio M. R."
4 date: "2023-01-01"
5 output: pdf_document
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7
8 knitr::opts_chunk$set(echo = TRUE)
9
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28 ## R Markdown
```

The bottom-left pane shows the R console with the following output:

```
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Platform: x86_64-apple-darwin20 (64-bit)

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'help.start()' for an HTML browser interface to help.
Type 'q()' to quit R.

-|
```

The bottom-right pane shows the file explorer with the following content:

TPDetat Tutorial 1: R Basics & Data Visualization
 TPDetat: 01-r-basics-data-viz
 This tutorial will introduce you to the basics of R and ggplot

TPDetat Tutorial 2: Data Wrangling
 TPDetat: 02-data-wrangling
 This tutorial will introduce you to data wrangling with dplyr

TPD Tutorial 3: Causality and Pivoting
 TPDat: 03-causality

The top-right pane shows the environment, history, connections, and tutorial tabs. The bottom-right pane shows the file explorer with the following content:

Name	Size	Modified
Applications		
Desktop		
Documents		
Downloads		
Dropbox		
Library		
Music		
Music		
OneDrive - University of Arizona		
Pictures		
Public		
SynologyDrive		
Thumbnails		
Workspaces		
Workspaces		
Workspaces		

Console: run code,
send code to here,
inspect output

The screenshot displays the RStudio interface with three main panels:

- Source Panel (Left):** Contains an R Markdown document titled "car-project.Rmd". The document includes a YAML header with title, author, date, and output settings. The body text describes R Markdown and includes an R code chunk for summarizing cars.
- Environment Panel (Top Right):** Shows the current environment with variables like "TPDstat Tutorial 1: R Basics & Data Visualization" and "TPDstat Tutorial 2: Data Wrangling".
- File Explorer Panel (Bottom Right):** Displays a list of files and folders, including "Applications", "Desktop", "Documents", "Downloads", "Dropbox", "Library", "Movies", "Music", "OneDrive - University of Arizona", "Pictures", "Public", "SynologyDrive", "iPhoto", "iMovie", and "iData".

Overlaid on the File Explorer panel is the text: **Project files, plots, and help**.

The screenshot shows the RStudio IDE interface. The top toolbar includes buttons for 'Save', 'Run', and 'Environment'. The main editor window displays a Markdown file with the following content:

```
1 ---
2 title: "My Project"
3 author: "Benjamin R"
4 date: "2023-01-01"
5 output: pdf_document
6 ---
7
8 ## [r setup: includeWALSH]
9 knitr::opts_chunk$set(echo = TRUE)
10
11
12 ## R Markdown
13
14 This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more
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20 ## [r code]
21 summary(cars)
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24 ## Including Plots
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26 You can also embed plots, for example:
27
28 ## [r plot]
```

The console window at the bottom shows the R version 4.2.2 and the R Foundation for Statistical Computing logo. The right sidebar contains a 'Tutorial' pane with the following content:

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 TPDstat 03-causality

A red box highlights the tutorial links, and a red arrow points to the text "Interacting with R objects, working with git, running local tutorials".

3. Using RMarkdown

```
1 library(ggplot2)
2 ggplot(mtcars, aes(x = wt, y = mpg)) +
3   geom_point()
```

Figure 2: Writing codes

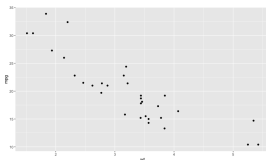


Figure 3: Looking at output

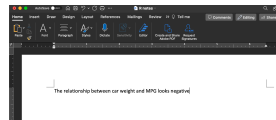


Figure 4: Taking notes

Rmarkdown files to the rescue

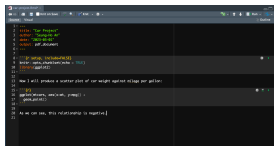


Figure 5: Rmarkdown file

Keep code and notes together in plain text

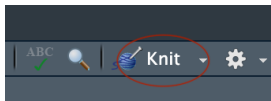


Figure 6: Knit in R

Rmarkdown files to the rescue

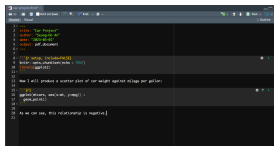


Figure 5: Rmarkdown file

Keep code and notes together in plain text

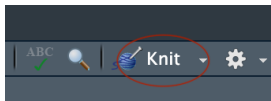


Figure 6: Knit in R

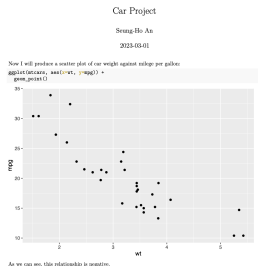


Figure 7: PDF output

Produce nice-looking outputs in different formats

Markdown: formatting in plain text

Non-code text in Rmd files is plain text with formatting instructions

syntax

Plain text
End a line with two spaces to start a new paragraph.
italics and _italics_
bold and __bold__
superscript^2^
--strikethrough--
[link](www.rstudio.com)

Header 1

Header 2

Header 3

Header 4

Header 5

Header 6

endash: --

emdash: ---

ellipsis: ...

inline equation: $A = \pi r^2$

image:

horizontal rule (or slide break):

> block quote

* unordered list

* item 2

+ sub-item 1

+ sub-item 2

1. ordered list

2. item 2

+ sub-item 1

becomes

Plain text

End a line with two spaces to start a new paragraph.

italics and *italics*

bold and **bold**

superscript²

~~strikethrough~~

[link](#)

Header 1

Header 2

Header 3

Header 4

Header 5

Header 6

endash: --

emdash: ---

ellipsis: ...

inline equation: $A = \pi r^2$

image: 

horizontal rule (or slide break):

block quote

- unordered list

- item 2

- sub-item 1

- sub-item 2

1. ordered list

2. item 2

```
---  
title: "Rmarkdown example"  
author: "Seung-Ho An"  
date: "2023-03-01"  
output: html_document  
---
```

Header contains metadata and sets options about the whole document

```
```{r setup, include=FALSE}  
knitr::opts_chunk$set(echo = TRUE)
```
```

Code chunk

R Markdown

Plain text with markdown formatting

This is an R Markdown document. Markdown is a simple formatting syntax for authoring HTML, PDF, and MS Word documents. For more details on using R Markdown see <<http://rmarkdown.rstudio.com>>.

When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

```
```{r cars}  
summary(cars)
```
```

Can "play" chunks interactively



Including Plots

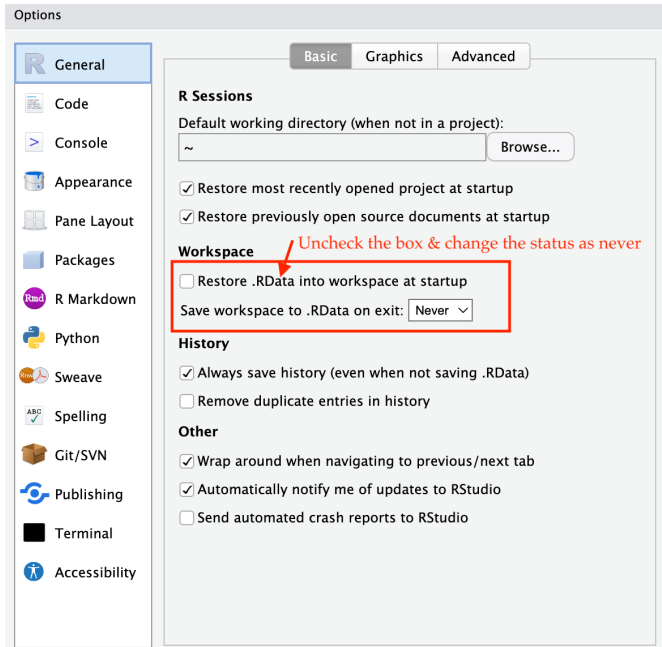
Chunks can have names and options

You can also embed plots, for example:

```
```{r pressure, echo=FALSE}  
plot(pressure)
```
```

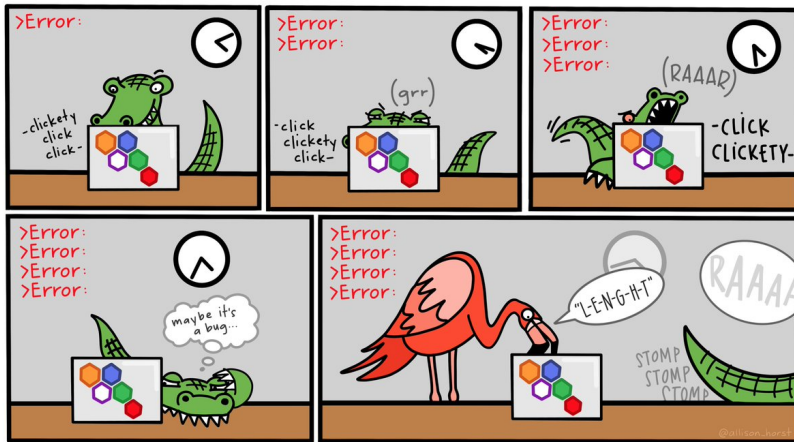
Code chunks replaced with output when knitted

Remember what's real



Try to type your code by hand

Typing speeds up the try-fail cycle



Physically typing the code is best way to familiarize yourself with R and the try-fail-try-fail-try-succeed cycle

Code that you can type and run:

```
## Any R code that begins with the # character is a comment  
## Comments are ignored by R
```

```
my_numbers <- c(4, 8, 15, 16, 23, 42) # Anything after # is also a comment
```

Code that you can type and run:

```
## Any R code that begins with the # character is a comment  
## Comments are ignored by R  
  
my_numbers <- c(4, 8, 15, 16, 23, 42) # Anything after # is also a comment
```

Output from code prefixed by ## by convention:

```
my_numbers
```

```
## [1]  4  8 15 16 23 42
```

Code that you can type and run:

```
## Any R code that begins with the # character is a comment
## Comments are ignored by R

my_numbers <- c(4, 8, 15, 16, 23, 42) # Anything after # is also a comment
```

Output from code prefixed by ## by convention:

```
my_numbers
```

```
## [1]  4  8 15 16 23 42
```

Output also has a counter in brackets when over one line:

```
options(width = 45)
letters
```

```
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j"
## [11] "k" "l" "m" "n" "o" "p" "q" "r" "s" "t"
## [21] "u" "v" "w" "x" "y" "z"
```

Everything in R has a name

```
my_numbers # just created this
```

```
## [1] 4 8 15 16 23 42
```

```
letters # this is build into R
```

```
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j"
```

```
## [11] "k" "l" "m" "n" "o" "p" "q" "r" "s" "t"
```

```
## [21] "u" "v" "w" "x" "y" "z"
```

```
pi # also built in
```

```
## [1] 3.141593
```

Everything in R has a name

```
my_numbers # just created this
```

```
## [1] 4 8 15 16 23 42
```

```
letters # this is build into R
```

```
## [1] "a" "b" "c" "d" "e" "f" "g" "h" "i" "j"
```

```
## [11] "k" "l" "m" "n" "o" "p" "q" "r" "s" "t"
```

```
## [21] "u" "v" "w" "x" "y" "z"
```

```
pi # also built in
```

```
## [1] 3.141593
```

Some names are forbidden (NA, TRUE, FALSE, etc) or strongly not recommended (c, mean, table)

We do things in R with functions

Functions take in objects, perform actions, and return outputs:

```
mean(x = my_numbers)
```

```
## [1] 18
```

- x is the argument name,

We do things in R with functions

Functions take in objects, perform actions, and return outputs:

```
mean(x = my_numbers)
```

```
## [1] 18
```

- `x` is the argument name,
- `my_numbers` is what we're passing to that argument

We do things in R with functions

Functions take in objects, perform actions, and return outputs:

```
mean(x = my_numbers)
```

```
## [1] 18
```

- x is the argument name,
- my_numbers is what we're passing to that argument

If you omit the argument name, R will assume the default argument order:

```
mean(my_numbers)
```

```
## [1] 18
```

How do we know the default argument order? Look to help files:

```
help(mean)
```

```
?mean #shorter
```

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 - Google, StackOverflow, Twitter, RStudio Community
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 - Google, StackOverflow, Twitter, RStudio Community
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- Get help **early** before becoming too frustrated!

How do we know the default argument order? Look to help files:

```
help(mean)
```

```
?mean #shorter
```

- Sometimes inscrutable, so look elsewhere:
 - Google, StackOverflow, Twitter, RStudio Community
 - Ask on the Teams Chat
- Get help **early** before becoming too frustrated! -Easy to overlook small issues like missing commas, typos, etc.

Packages are bundles of functions written by other users that we can use

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Install packages using `install.packages()` to have them on your machine:

```
install.packages("ggplot2")
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```
install.packages("ggplot2")
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Load them into your R session with `library()`:

```
library(ggplot2)
```

Functions live in packages

Packages are bundles of functions written by other users that we can use

Install packages using `install.packages()` to have them on your machine:

```
install.packages("ggplot2")
```

Load them into your R session with `library()`:

```
library(ggplot2)
```

Now we can use any function provided by `ggplot2`

We can also use the `mypackage::` prefix to access package functions without loading:

```
knitr::kable(head(mtcars))
```

| | mpg | cyl | disp | hp | drat | wt | qsec | vs | am | gear | carb |
|-------------------|------|-----|------|-----|------|-------|-------|----|----|------|------|
| Mazda RX4 | 21.0 | 6 | 160 | 110 | 3.90 | 2.620 | 16.46 | 0 | 1 | 4 | 4 |
| Mazda RX4 Wag | 21.0 | 6 | 160 | 110 | 3.90 | 2.875 | 17.02 | 0 | 1 | 4 | 4 |
| Datsun 710 | 22.8 | 4 | 108 | 93 | 3.85 | 2.320 | 18.61 | 1 | 1 | 4 | 1 |
| Hornet 4 Drive | 21.4 | 6 | 258 | 110 | 3.08 | 3.215 | 19.44 | 1 | 0 | 3 | 1 |
| Hornet Sportabout | 18.7 | 8 | 360 | 175 | 3.15 | 3.440 | 17.02 | 0 | 0 | 3 | 2 |
| Valiant | 18.1 | 6 | 225 | 105 | 2.76 | 3.460 | 20.22 | 1 | 0 | 3 | 1 |

R tutorials & course packages

```
# installing course packages
my_packages <- c("tidyverse", "usethis", "devtools", "learnr",
                 "tinytex", "gitcreds", "broom")
install.packages(my_packages, repos = "http://cran.rstudio.com")

remotes::install_github("kosukeimai/qss-package", build_vignettes = TRUE)

# installing R tutorials locally
remotes::install_github("rstudio/learnr")

remotes::install_github("rstudio-education/gradethis")

remotes::install_github("seungho-an/TPDtutor")

# installing tiny tex
tinytex::install_tinytex()
```

4.

Data visualizations

4.1

Building plots by layers

```
midwest
```

```
## # A tibble: 437 x 28
##       PID county    state  area poppto~1 popde~2
##   <int> <chr>    <chr> <dbl>   <int>   <dbl>
## 1   561 ADAMS    IL    0.052   66090   1271.
## 2   562 ALEXAND~ IL    0.014   10626    759
## 3   563 BOND     IL    0.022   14991    681.
## 4   564 BOONE    IL    0.017   30806   1812.
## 5   565 BROWN    IL    0.018    5836    324.
## 6   566 BUREAU   IL    0.05    35688    714.
## 7   567 CALHOUN  IL    0.017    5322    313.
## 8   568 CARROLL  IL    0.027   16805    622.
## 9   569 CASS     IL    0.024   13437    560.
## 10  570 CHAMPAI~ IL    0.058  173025   2983.
## # ... with 427 more rows, 22 more variables:
## #   popwhite <int>, popblack <int>,
## #   popamerindian <int>, popasian <int>,
## #   popother <int>, percwhite <dbl>,
## #   percblack <dbl>, percamerindan <dbl>,
## #   percasian <dbl>, percother <dbl>,
## #   popadults <int>, perchsd <dbl>, ...
```

Create ggplot object and direct it to the correct data:

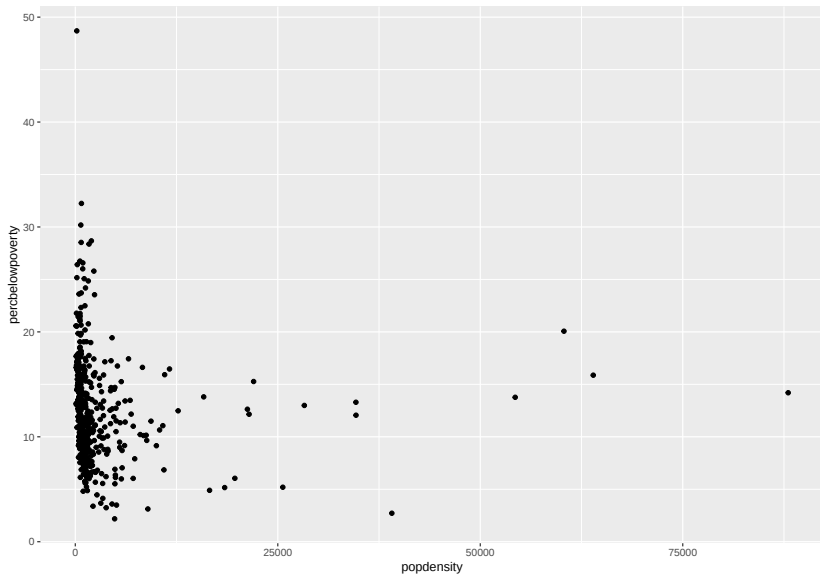
```
p <- ggplot(data = midwest)
```

Mapping: tell ggplot what visual aesthetics correspond to which variables

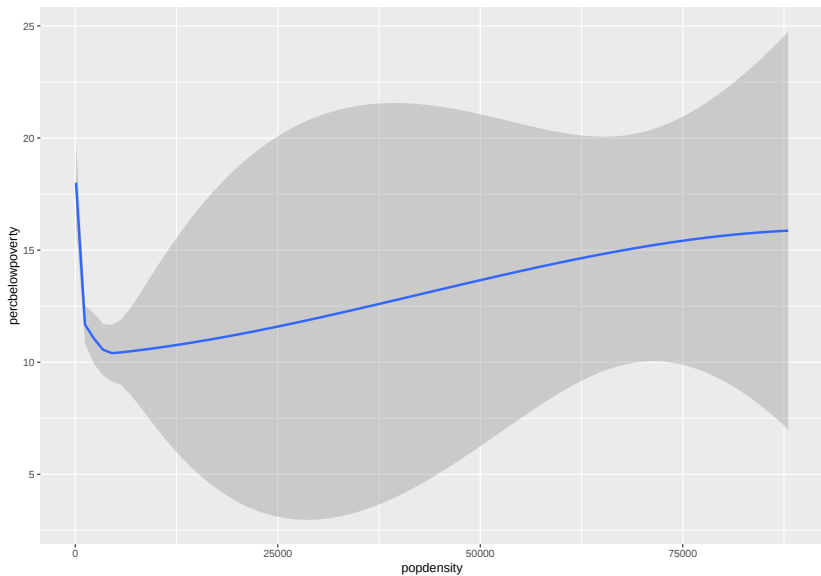
```
p <- ggplot(data = midwest,  
            mapping = aes(x = popdensity,  
                          y = percbelowpoverty))
```

Other aesthetic mappings: color, shape, size, etc.

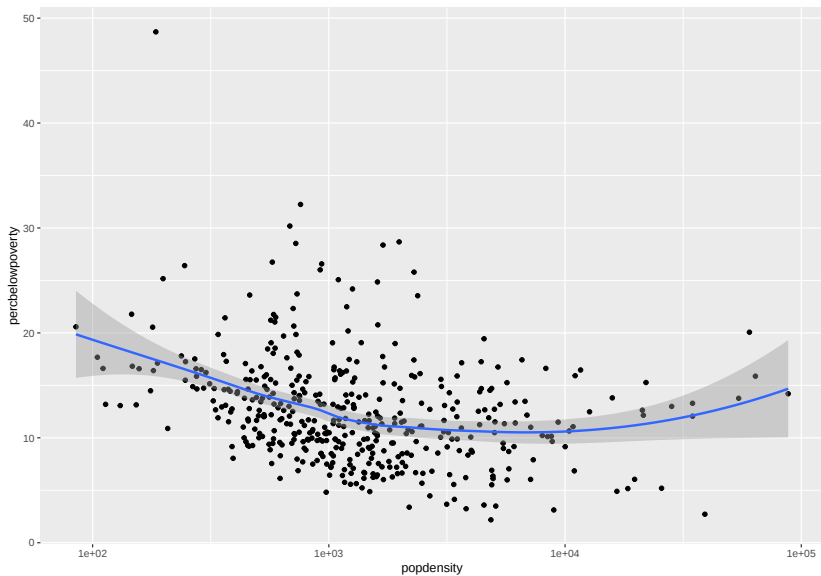
```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty)) +  
  geom_point()
```



```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                      y = percbelowpoverty)) +  
  geom_smooth()
```



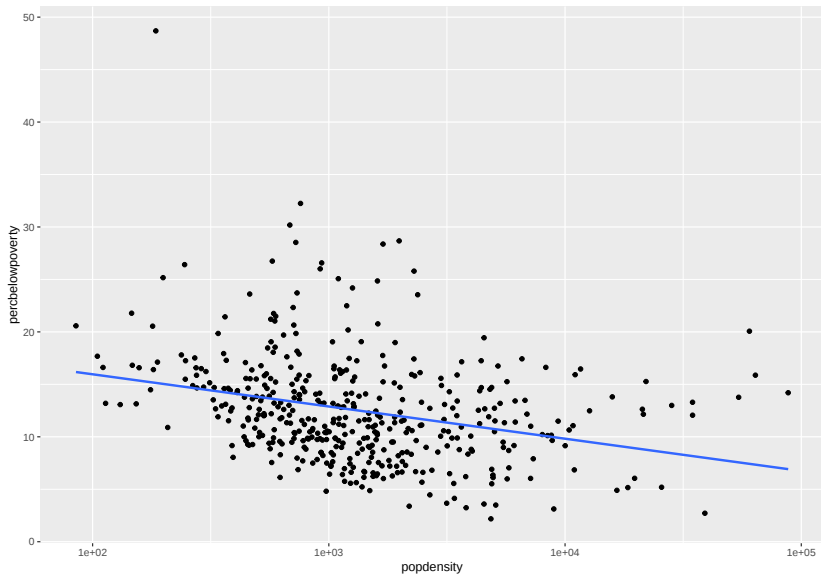
```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty)) +  
  geom_point() +  
  geom_smooth() +  
  scale_x_log10()
```



Geoms can take arguments:

```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty)) +  
  geom_point() +  
  geom_smooth(method = "lm", se = FALSE) +  
  scale_x_log10()
```

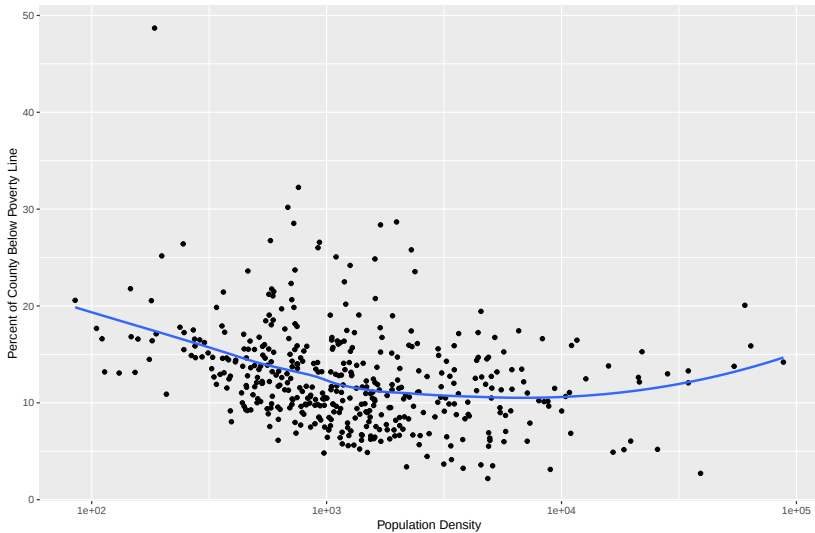
Tells `geom_smooth` to do a linear fit with no error region



Adding informative labels

```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty)) +  
geom_point() +  
geom_smooth(method = "loess", se = FALSE) + scale_x_log10() +  
labs(x = "Population Density",  
     y = "Percent of County Below Poverty Line",  
     title = "Poverty and Population Density",  
     subtitle = "Among Counties in the Midwest",  
     source = "US Census, 2000")
```

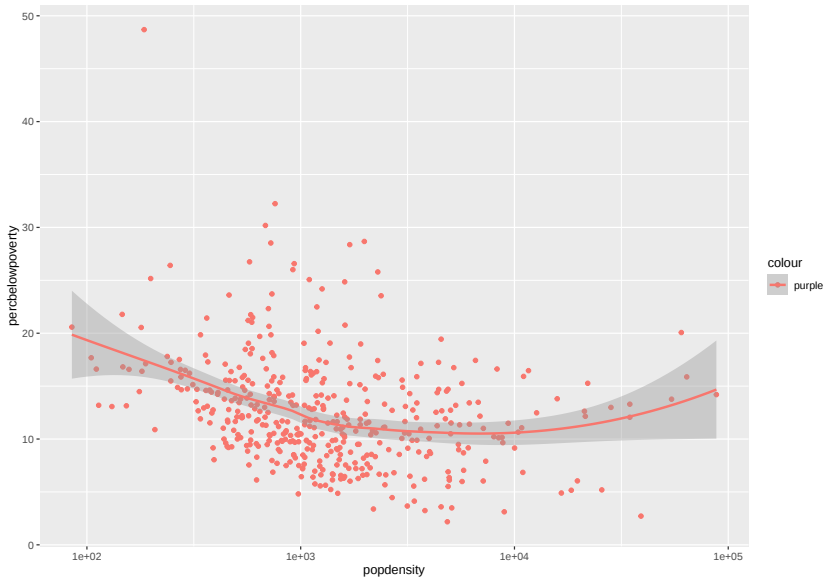
Poverty and Population Density
Among Counties in the Midwest



Mapping vs. setting aesthetics

```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty,  
                     color = "purple")) +  
geom_point() +  
geom_smooth() +  
scale_x_log10()
```

Wait what?



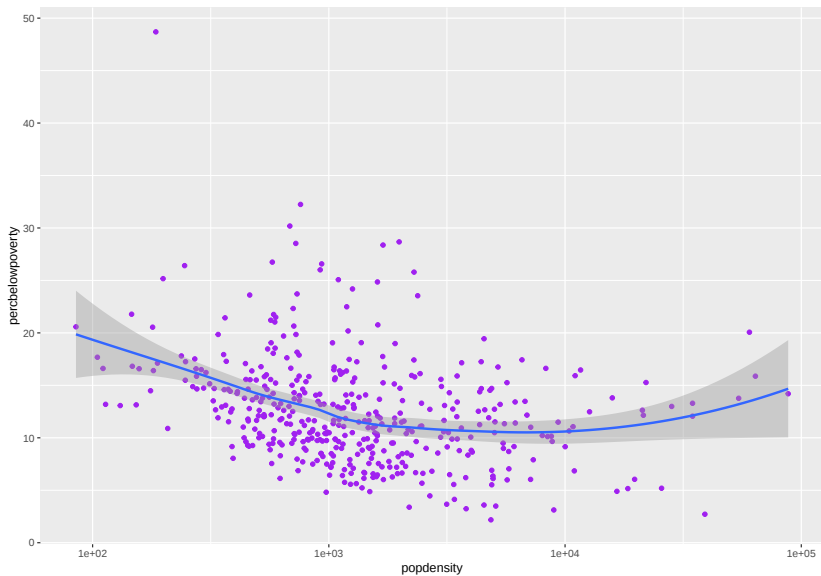
Mapping always refers to variables

If passed a value other than a variable name, ggplot will implicitly create a variable with that value (in this case "purple" that is constant)

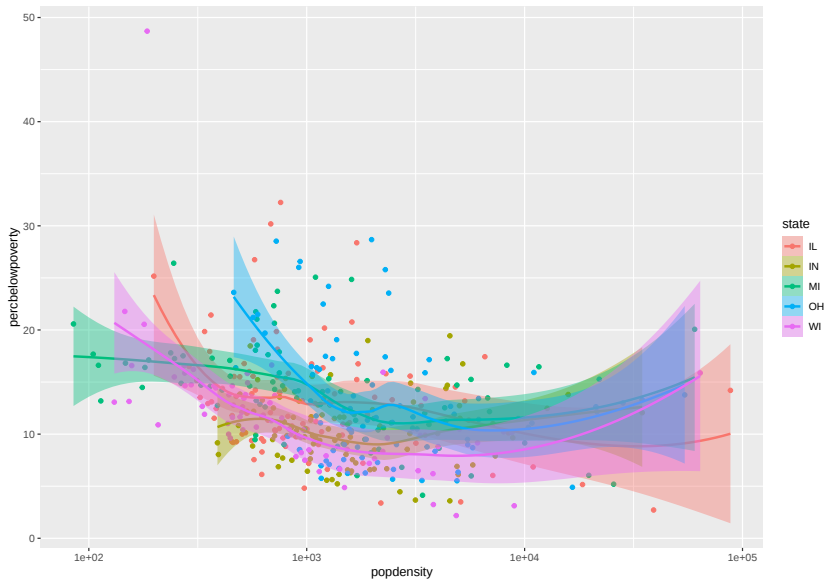
```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty,  
                     color = "purple")) +  
geom_point() +  
geom_smooth() +  
scale_x_log10()
```

Set the color outside the `mapping = aes()` format

```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty)) +  
geom_point(color = "purple") +  
geom_smooth() +  
scale_x_log10()
```

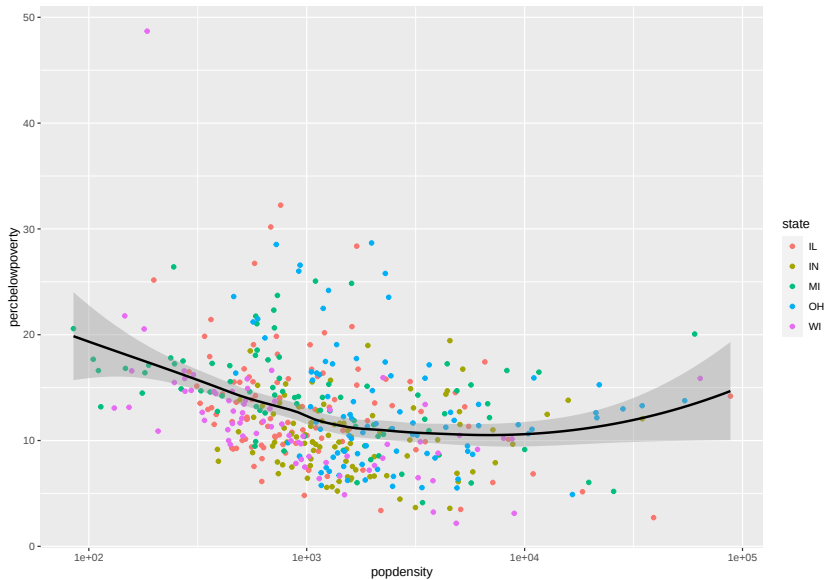


```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty,  
                     color = state,  
                     fill = state)) +  
geom_point() +  
geom_smooth() +  
scale_x_log10()
```



Mappings can be done on a per geom basis

```
ggplot(data = midwest,  
       mapping = aes(x = popdensity,  
                     y = percbelowpoverty)) +  
geom_point(mapping = aes(color = state)) +  
geom_smooth(color = "black") +  
scale_x_log10()
```



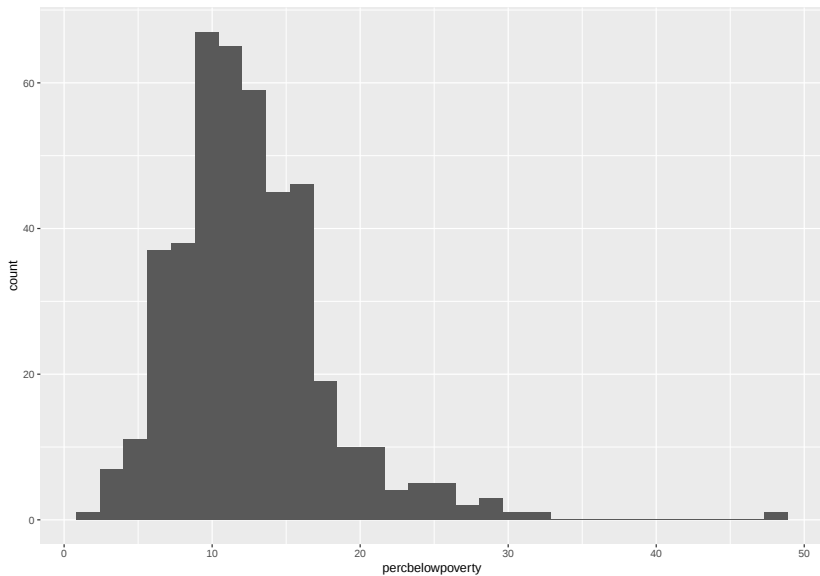
4.2

Histograms and boxplots

Histograms show where there are more or fewer observations of a numeric variable

```
ggplot(data = midwest,  
       mapping = aes(x = percbelowpoverty)) +  
  geom_histogram()
```

Split up range of variable into bins, count how many are in each bin
y aesthetic calculated automatically

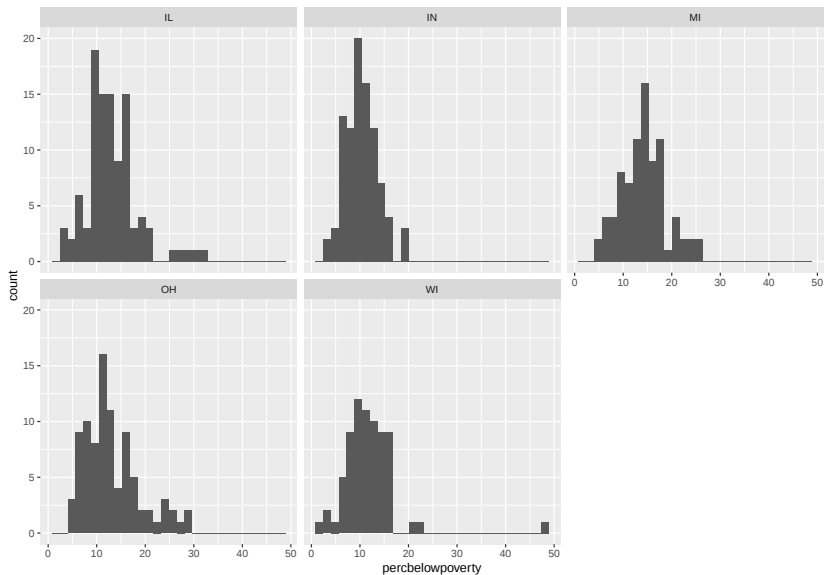


Creating small multiples with facets

Small multiples: a series of similar graphs with the same scale/axes to help with comparing different participation of a dataset

```
ggplot(data = midwest,  
       mapping = aes(x = percbelowpoverty)) +  
  geom_histogram() +  
  facet_wrap(~ state)
```

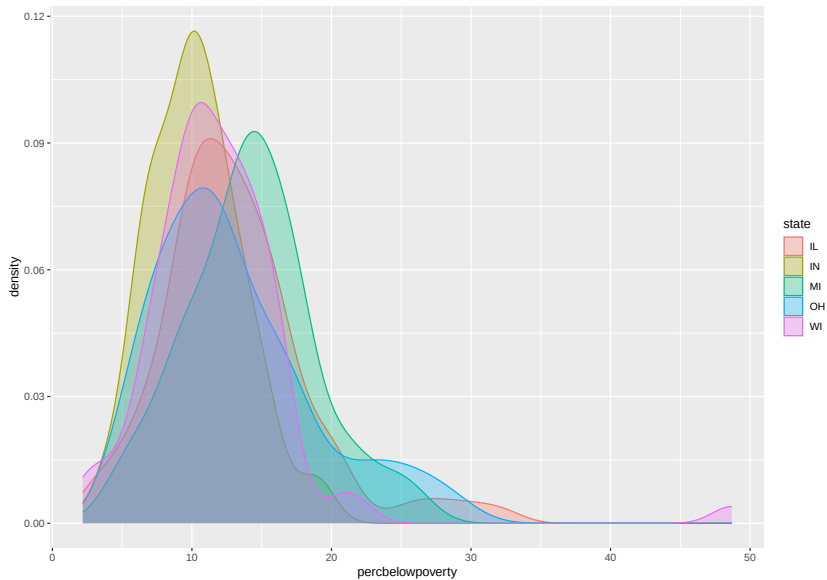
We'll see more of the ~ variable syntax (called a formula)



Density as alternative to histograms

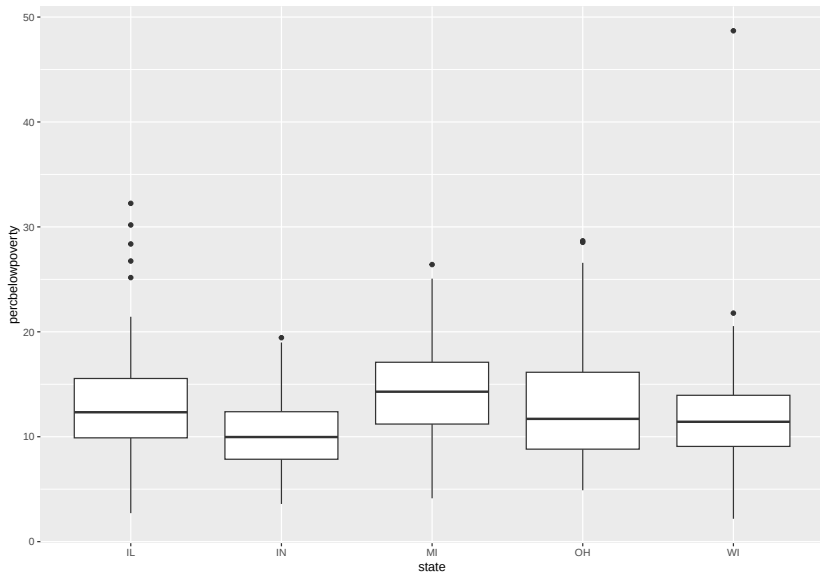
A **kernel density** plot is a smoothed version of a histogram and slightly easier to overlay

```
ggplot(data = midwest,  
       mapping = aes(x = percbelowpoverty,  
                     fill = state, color = state)) +  
  geom_density(alpha = 0.3)
```



Boxplots are another way to compare distributions across discrete groups.

```
ggplot(data = midwest,  
       mapping = aes(x = state,  
                     y = percbelowpoverty)) +  
geom_boxplot()
```



- “Box” represents middle 50% of the data

- “Box” represents middle 50% of the data
 - 25% of the data above the box, 25% below

- “Box” represents middle 50% of the data
 - 25% of the data above the box, 25% below
 - Width of the box is called the inter quartile range (IQR)

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- “Whiskers” represents either:

- “Box” represents middle 50% of the data
 - 25% of the data above the box, 25% below
 - Width of the box is called the inter quartile range (IQR)
- Horizontal line in the box is the median
 - 50% of the data above the median, 50% below
- “Whiskers” represents either:
 - $1.5 \times \text{IQR}$ or max/min of the data, whichever is smaller

- “Box” represents middle 50% of the data
 - 25% of the data above the box, 25% below
 - Width of the box is called the inter quartile range (IQR)
- Horizontal line in the box is the median
 - 50% of the data above the median, 50% below
- “Whiskers” represents either:
 - $1.5 \times \text{IQR}$ or max/min of the data, whichever is smaller
 - Points beyond whiskers are outliers

4.3

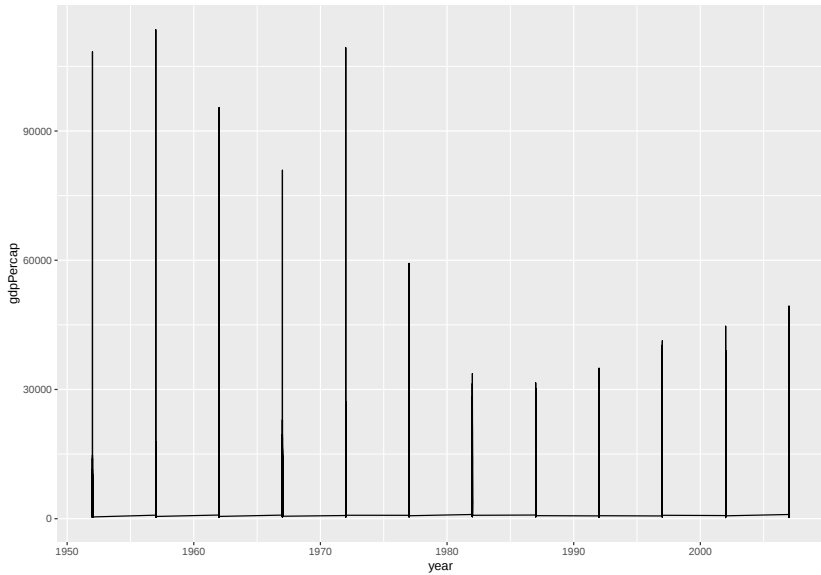
Grouped data

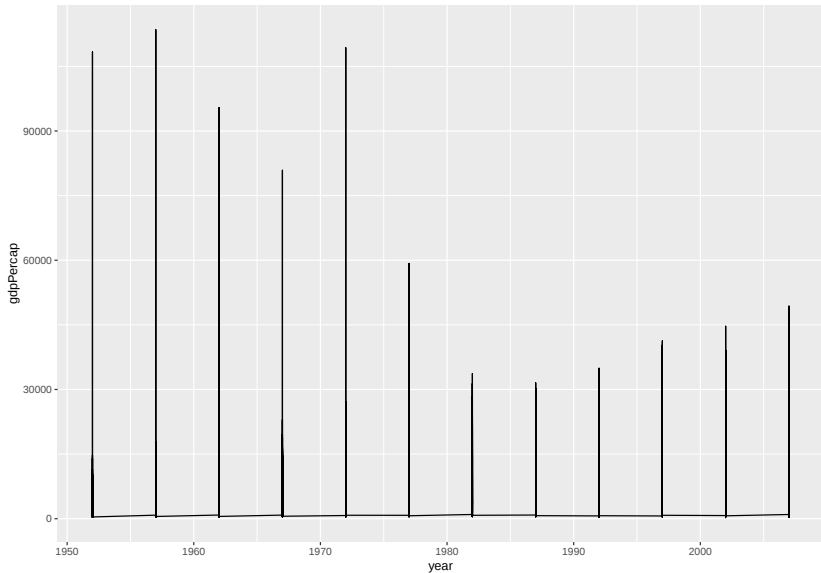
```
library(dplyr)
library(gapminder)
glimpse(gapminder)
```

```
## Rows: 1,704
## Columns: 6
## $ country    <fct> "Afghanistan", "Afghanist~
## $ continent  <fct> Asia, Asia, Asia, Asia, A~
## $ year       <int> 1952, 1957, 1962, 1967, 1~
## $ lifeExp    <dbl> 28.801, 30.332, 31.997, 3~
## $ pop        <int> 8425333, 9240934, 1026708~
## $ gdpPercap  <dbl> 779.4453, 820.8530, 853.1~
```

Let's plot the trend in income

```
ggplot(data = gapminder,  
       mapping = aes(x = year,  
                     y = gdpPercap)) +  
geom_line()
```

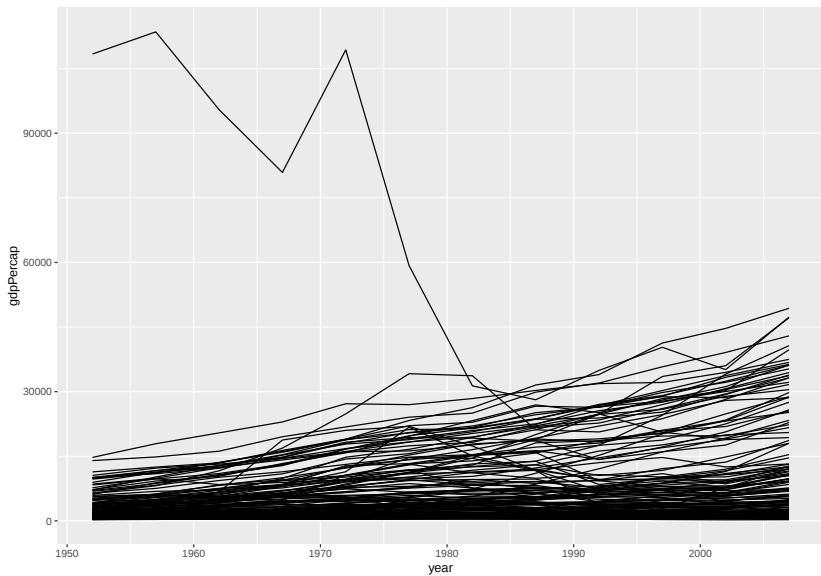




`geom_line` connects points from different countries in the same year.

Tell geom_line how to group the lines

```
ggplot(data = gapminder,  
       mapping = aes(x = year,  
                     y = gdpPercap)) +  
geom_line(mapping = aes(group = country))
```



```
ggplot(data = gapminder,  
       mapping = aes(x = year,  
                     y = gdpPercap)) +  
  geom_line(mapping = aes(group = country), color = "grey70") +  
  geom_smooth(method = "loess") +  
  scale_y_log10(labels = scales::dollar)
```

