

Module 0: Introduction

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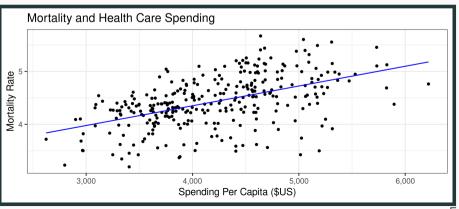
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Motivation

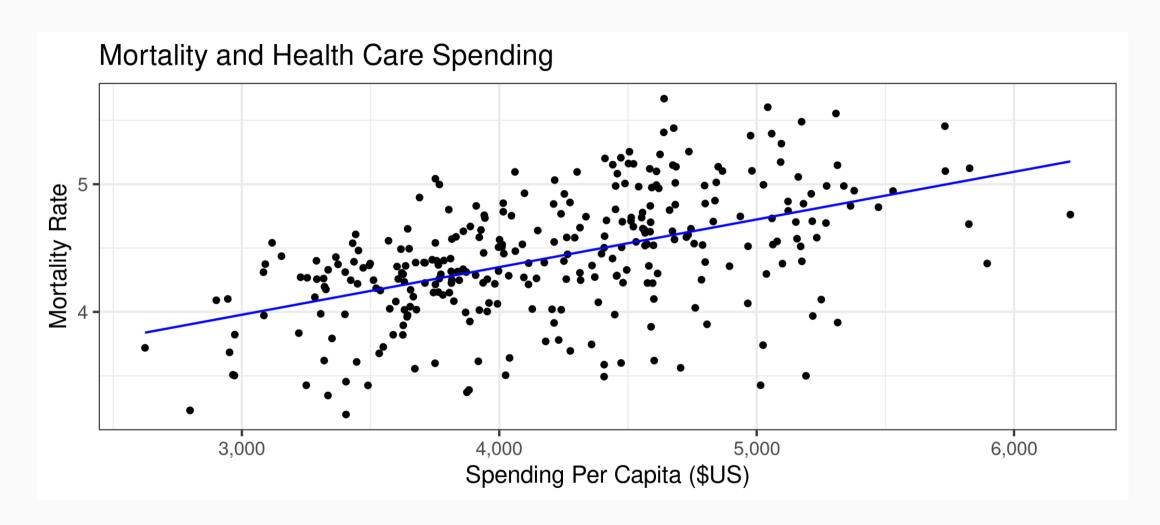
Motivating question

Does health care spending improve health?

Spending and Health

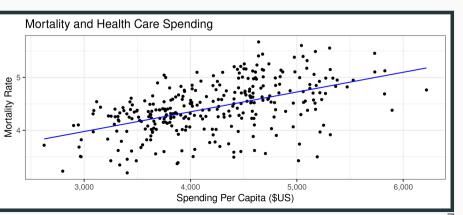


Spending and Health



Spending and Health

- Does medical spending make us sicker?
- What else might explain this relationship?



Goals of this course.

- 1. Understand and implement selected methods for causal inference
- 2. Along the way...data management and version control with real data
- 3. Summarize, visualize, and explain research results

Syllabus highlights

(Read the full document here.)

Why this course?

- 1. Major problems that need solutions
- 2. Need good, convincing empirical work for policy
- 3. Working with data is hard, particularly health care data
- 4. Your work should be transparent and reproducible

Structure

- Very applied in nature
- Methods for causal inference
 - Selection on observables (regression, re-weighting, matching, propensity scores)
 - Instrumental variables
 - Regression discontinuity
 - Difference-in-differences

Structure

- Substantive areas
 - Hospital pricing, policy, and competition
 - Cigarette taxes and demand
 - Medicare Advantage and quality disclosure
 - Medicaid expansion and health insurance

Structure

- Datasets from the real world
 - Hospital Cost Report Information System (HCRIS)
 - Centers for Disease Control (CDC)
 - Medicare Advantage data
 - Behavioral Risk Factor Surveillance System (BRFSS), Medicaid, Health Insurance Exchanges

Assignments

- Homework (x5)
- Research project
- Participation

Grading

Component	Weight
5 × homework assignments (11% each)	55%
Research project	40%
Participation	5%

Software Installation

Software Installation

- 1. Download R
- 2. Download RStudio
- 3. Download Git
- 4. Create an account on GitHub

For help and troubleshooting with Git and GitHub, take a look at Jenny Bryan's http://happygitwithr.com.

Checklist

☑ Do you have the most recent version of R?

```
version$version.string
## [1] "R version 4.1.2 (2021-11-01)"
```

☑ Do you have the most recent version of RStudio? (The preview version is fine.)

```
RStudio.Version()$version
## Requires an interactive session but should return something like "[1] '1.4.1717'"
```

☑ Have you updated all of your R packages?

```
update.packages(ask = FALSE, checkBuilt = TRUE)
```

Checklist

- Open up the shell
- Windows users, make sure that you installed a Bash-compatible version of the shell. If you installed Git for Windows, then you should be good to go.

Checklist

☑ Which version of Git have you installed?

```
git --version
```

☑ Did you introduce yourself to Git? (Substitute in your details.)

```
git config --global user.name 'Ian McCarthy'
git config --global user.email 'ian.mccarthy@emory.edu'
git config --global --list
```

☑ Did you register an account in GitHub?

Altnernative setup...

Just use the cloud!

- We have our own virtual computer via AWS
- This computer has all the space you need and the data are already available
- Downside: the computer will be "on" during designated times only
- Details of this are on Canvas

Practice with Git and RStudio

Before next class (see http://happygitwithr.com)

- 1. Download R
- 2. Download RStudio
- 3. Download Git
- 4. Create an account on GitHub
- 5. Connect RStudio to Git and GitHub
- 6. Start/clone/fork a repository for this class

Setting things up

Now we're going to clone a GitHub repository (repo) using RStudio. The video below is from Grant McDermott's class.



Some common mistakes for windows users

- Windows folders are *not* files...there is no content without a file. You can't commit or push changes without content.
- Let RStudio/GitHub create the directory (main folder) for you.
- If you're working across devices on your own repo, be sure to pull before starting and push afterward.
- Avoid spaces in file names. Avoid them at all costs. DO NOT PUT SPACES IN YOUR FILE NAMES.
 - "A space in a file name is a space in your soul."

Ideal workflow

Until you are a Git(Hub) expert...

- 1. Start project on GitHub (fork from another repo if needed)
- 2. Clone to desktop with RStudio
- 3. See http://happygitwithr.com for instructions on linking your local repo with a new upstream remote

Working with AWS

Let's do this live!

Tidy Data

The tidyverse

- Suite of packages collectively known as the tidyverse
- Different from base R in many ways
- The tidyverse with pipes¹ is more intuitive to me

¹ We'll talk about pipes very soon!

What is Tidy data?

Resources:

- Paper: Tidy Data (Hadley Wickham, 2014 JSS)
- Vignette: Tidy data (from the tidyr package)

Essentially:

- 1. Variables are columns
- 2. Observations are rows
- 3. Variables and observations make a table

Intro to Tidy data

Let's load the tidyverse package and check the output:

```
library(tidyverse)
```

Comes with lots of other packages like ggplot2, tibble, dplyr, etc.

Pipes: %>%

- The pipe operator is denoted %>% and is automatically loaded with the tidyverse.
- Pipes are awesome!

```
## These next two lines of code do exactly the same thing.
mpg %>% filter(manufacturer="audi") %>% group_by(model) %>% summarise(hwy_mean = mean(hwy summarise(group_by(filter(mpg, manufacturer="audi"), model), hwy_mean = mean(hwy))
```

The first line reads from left to right and from data to operation. The Base R version (line 2) works in the opposite order.

Pipes: %>%

2 a4 quattro

3 a6 quattro

Helps to break the pipes over several lines

25.8

24

The dplyr package:

- 1. filter(): Find or exclude certian rows
- 2. arrange(): Sort your observations
- 3. select(): Select specific variables
- 4. mutate(): Create new variables
- 5. summarise(): Collapse multiple rows into a single summary value

1) dplyr::filter()

Multiple filters separated by commas:

```
starwars %>%
  filter(
    species = "Human",
    height ≥ 190
    ) %>% head(5)
## # A tibble: 4 × 14
         height mass hair color skin color eye color birth year sex
                                                                 gender
##
    name
    <chr>
         <int> <dbl> <chr>
                                <chr> <chr>
                                                       <dbl> <chr> <chr>
###
## 1 Darth Va... 202 136 none white yellow 41.9 male mascu...
## 2 Qui-Gon ... 193 89 brown fair
                                          blue
                                                        92
                                                           male mascu…
         193 80 white
                             fair
  3 Dooku
                                          brown
                                                       102
                                                            male mascu…
## 4 Bail Pre... 191
                    NA black
                                tan
                                          brown
                                                   67
                                                            male
                                                                 mascu...
## # ... with 5 more variables: homeworld <chr>, species <chr>, films <list>,
    vehicles <list>, starships <list>
```

1) dplyr::filter()

To remove missing observations, use filter(!is.na(height)). Also try the drop_na() function from tidyr.

2) dplyr::arrange()

Arrange in ascending order:

```
starwars %>%
  arrange(birth year) %>% head(5)
## # A tibble: 5 × 14
              height mass hair color skin color eye color birth year sex
                                                                          gender
##
    name
                                                               <dbl> <chr> <chr>
          <int> <dbl> <chr>
                                                <chr>
    <chr>
                                     <chr>
##
## 1 Wicket S...
                                                brown
               88
                        20 brown
                                     brown
                                                                   8 male
                                                                          mascu...
                                     metal
## 2 IG-88
          200 140 none
                                                red
                                                                 15 none
                                                                          mascu...
  3 Luke Sky... 172 77 blond
                                     fair
                                                blue
                                                                 19 male mascu...
## 4 Leia Org... 150 49 brown
                                     light
                                                brown
                                                                 19 fema... femin...
  5 Wedge An... 170
                       77 brown
                                     fair
                                                hazel
                                                                 21 male
                                                                          mascu...
## # ... with 5 more variables: homeworld <chr>, species <chr>, films <list>,
## #
     vehicles <list>, starships <list>
```

2) dplyr::arrange()

Arrange descending order using arrange(desc()):

```
starwars %>%
  arrange(desc(birth year)) %>% head(5)
## # A tibble: 5 × 14
            height mass hair color skin color eye color birth year sex
                                                                           gender
##
    name
          <int> <dbl> <chr>
                                                <chr>
    <chr>
                                    <chr>
                                                               <dbl> <chr>
                                                                           <chr>
##
                66
## 1 Yoda
                      17 white
                                                brown
                                                                 896 male
                                    green
                                                                           mascu...
                                                                600 herma... mascu...
## 2 Jabba ... 175 1358 <NA>
                                    green-tan,... orange
                                   unknown
                                                blue
  3 Chewba...
            228
                     112 brown
                                                                 200 male
                                                                           mascu...
## 4 C-3PO
                                   gold
                                                vellow
          167 75 <NA>
                                                                112 none
                                                                           mascu...
## 5 Dooku
          193 80 white
                                    fair
                                                brown
                                                                102 male
                                                                           mascu...
## # ... with 5 more variables: homeworld <chr>, species <chr>, films <list>,
## #
     vehicles <list>, starships <list>
```

- Use commas to select multiple columns
- Use "first:last" for consecutive columns
- Deselect a column with "-"

```
starwars %>%
  select(name:skin_color, species, -height) %>% head(5)
## # A tibble: 5 × 5
                    mass hair color skin color
                                                 species
###
     name
                    <dbl> <chr>
                                     <chr>
                                                 <chr>
    <chr>
###
## 1 Luke Skywalker
                      77 blond
                                    fair
                                                 Human
                                                 Droid
## 2 C-3P0
                      75 <NA>
                                    gold
  3 R2-D2
                       32 <NA>
                                     white, blue Droid
                                    white
## 4 Darth Vader
                     136 none
                                                 Human
                                    light
                                                 Human
## 5 Leia Organa
                       49 brown
```

Rename within select():

```
starwars %>%
  select(alias=name, crib=homeworld, sex=gender) %>% head(5)
## # A tibble: 5 × 3
   alias
          crib
###
                          sex
   <chr>
          <chr> <chr>
###
## 1 Luke Skywalker Tatooine masculine
## 2 C-3P0
                 Tatooine masculine
                 Naboo masculine
## 3 R2-D2
## 4 Darth Vader Tatooine masculine
## 5 Leia Organa Alderaan feminine
```

Use select(contains(PATTERN)) to find rows that contain some strings of interest

```
starwars %>%
  select(name, contains("color")) %>% head(5)
## # A tibble: 5 × 4
                  hair color skin color eye color
##
    name
    <chr>
                  <chr>
                            <chr>
                                       <chr>
###
## 1 Luke Skywalker blond fair blue
## 2 C-3P0
                  <NA>
                            gold yellow
                            white, blue red
## 3 R2-D2
                  <NA>
## 4 Darth Vader
                            white
                                  yellow
                  none
## 5 Leia Organa
                  brown
                            light
                                       brown
```

Also look into the stringr package.

Create new variables with mutate()

```
starwars %>%
  select(name, birth year) %>%
  mutate(dog years = birth year * 7) %>%
  mutate(comment = paste0(name, " is ", dog years, " in dog years.")) %>% head(5)
## # A tibble: 5 × 4
                  birth year dog years comment
###
    name
  <chr>
                       <dbl> <dbl> <chr>
###
                    19
## 1 Luke Skywalker
                                 133 Luke Skywalker is 133 in dog years.
## 2 C-3P0
                       112
                                 784 C-3PO is 784 in dog years.
                        33
## 3 R2-D2
                                 231 R2-D2 is 231 in dog years.
                   41.9
## 4 Darth Vader
                                 293. Darth Vader is 293.3 in dog years.
## 5 Leia Organa
                        19
                                     Leia Organa is 133 in dog years.
```

mutate() processes in order, so you can put dependent mutates one after another:

```
## # A tibble: 5 × 4
                birth year dog years comment
###
    name
  <chr>
                      <dbl> <dbl> <chr>
###
                   19
## 1 Luke Skywalker
                                133 Luke Skywalker is 133 in dog years.
## 2 C-3P0
                      112
                                784 C-3PO is 784 in dog years.
                       33
## 3 R2-D2
                                231 R2-D2 is 231 in dog years.
                  41.9
## 4 Darth Vader
                                293. Darth Vader is 293.3 in dog years.
## 5 Leia Organa
                       19
                                   Leia Organa is 133 in dog years.
```

Other handy ways to use mutate():

```
starwars %>%
  select(name, height) %>%
  filter(name %in% c("Luke Skywalker", "Anakin Skywalker")) %>%
  mutate(tall1 = height > 180) %>%
  mutate(tall2 = ifelse(height > 180, "Tall", "Short")) ## Same effect, but can choose lab
## # A tibble: 2 × 4
                    height tall1 tall2
###
    name
    <chr>
          <int> <lgl> <chr>
###
## 1 Luke Skywalker 172 FALSE Short
## 2 Anakin Skywalker 188 TRUE Tall
```

Note the "scoped" variants of mutate() that work on a subset of variables:

- mutate_all() affects every variable
- mutate_at() affects named or selected variables
- mutate_if() affects variables that meet some criteria (e.g. are numeric)

These have since been replaced with <code>across()</code>. So you would use <code>mutate(across())</code> instead.

5) dplyr::summarise() with group_by()

```
starwars %>%
  group by(species, gender) %>%
  summarise(mean_height = mean(height, na.rm = T)) %>% head(5)
## # A tibble: 5 × 3
## # Groups: species [5]
   species gender mean height
##
   <chr> <chr> <dbl>
###
## 1 Aleena masculine
                           79
## 2 Besalisk masculine
                             198
## 3 Cerean
            masculine
                             198
## 4 Chagrian masculine
                       196
## 5 Clawdite feminine
                             168
```

Note: na.rm = T is usually a good idea, otherwise your summary will be NA too.

5) dplyr::summarise()

starwars %>% group_by(species, gender) %>% summarise(across(where(is.numeric), mean, na.rm

```
# A tibble: 5 \times 5
## # Groups:
             species [5]
             gender height mass birth year
    species
###
             <chr> <dbl> <dbl>
                                          <dbl>
    <chr>
###
## 1 Aleena
             masculine
                                            NaN
                            79
                                  15
  2 Besalisk masculine
                                 102
                                            NaN
                          198
  3 Cerean
             masculine
                           198
                                82
                                            92
  4 Chagrian masculine
                           196
                                            NaN
                                 NaN
## 5 Clawdite feminine
                           168
                                  55
                                            NaN
```

Joining operations

Central feature of the dplyr package invovles merging data from multiple tables with join operations.

- inner_join(df1, df2)
- left_join(df1, df2)
- right_join(df1, df2)
- full_join(df1, df2)
- semi_join(df1, df2)
- anti_join(df1, df2)

Joining operations

- For some simple examples, we'll need some data sets that come bundled with the nycflights13 package.
- Load it now and then inspect these data frames in your own console.

```
library(nycflights13)
flights
planes
```

Let's perform a left join on the flights and planes datasets.

```
left join(flights, planes) %>%
  select(year, month, day, dep time, arr time, carrier, flight, tailnum, type, model) %>%
## # A tibble: 5 × 10
                  day dep time arr time carrier flight tailnum type model
###
     vear month
     <int> <int> <int>
                         <int>
                                  <int> <chr>
                                              <int> <chr>
                                                              <chr> <chr>
##
## 1
     2013
                           517
                                    830 UA
                                                  1545 N14228 <NA>
                                                                    <NA>
## 2
     2013
                           533
                                    850 UA
                                                  1714 N24211 <NA>
                                                                    <NA>
                           542
                                923 AA
                                                  1141 N619AA <NA>
## 3
     2013
                                                                    <NA>
                           544
                                   1004 B6
                                                   725 N804JB
## 4
     2013
                                                              <NA>
                                                                    <NA>
## 5
     2013
                           554
                                    812 DL
                                                   461 N668DN
                                                              <NA>
                                                                    <NA>
```

dplyr guessed about which columns to join on (i.e. columns that share the same name). It also told us its choices:

```
## Joining, by = c("year", "tailnum")
```

Problem: the variable "year" does not have a consistent meaning across our joining datasets!

year of flight versus year of construction

Luckily, there's an easy way to avoid this problem.

- See if you can figure it out before turning to the next slide.
- Try ?dplyr::join.

Let's be more explicit with the by = argument:

```
left_join(
  flights,
  planes %>% rename(year_built = year), ## Not necessary w/ below line, but helpful
  by = "tailnum" ## Be specific about the joining column
  ) %>%
  select(year, month, day, dep_time, arr_time, carrier, flight, tailnum, year_built, type,
  head(5) ## Just to save vertical space on the slide
```

```
## # A tibble: 5 × 11
     year month day dep time arr time carrier flight tailnum year built type
##
    <int> <int> <int> <int> <int> <chr>
                                                           <int> <chr>
###
                                           1545 N14228
                                                            1999 Fixed w...
     2013
                        517
                               830 UA
## 1
                                                            1998 Fixed w...
## 2
    2013
                       533
                           850 UA 1714 N24211
                        542
                                                            1990 Fixed w...
## 3
    2013
                           923 AA
                                           1141 N619AA
                        544
                              1004 B6
                                            725 N804JB
                                                            2012 Fixed w...
     2013
## 4
```

Be specific

What happens if we again specify the join column but don't rename the ambiguous "year"?

```
left_join(flights, planes, ## Not renaming "year" to "year_built" this time
by = "tailnum") %>%
select(contains("year"), month, day, dep_time, arr_time, carrier, flight, tailnum, type,
head(5)
```

```
## # A tibble: 5 × 11
                         day dep time arr time carrier flight tailnum type model
##
     vear.x year.y month
      <int> <int> <int> <int>
                               <int>
                                           <int> <chr>
                                                        <int> <chr> <chr> <chr>
##
                                             830 UA
## 1
      2013
             1999
                                    517
                                                           1545 N14228 Fixe... 737-...
## 2
      2013
              1998
                                    533
                                             850 UA
                                                           1714 N24211 Fixe... 737-...
                                    542
                                                           1141 N619AA Fixe... 757-...
      2013
              1990
                                        923 AA
## 3
      2013
              2012
                                    544
                                            1004 B6
                                                            725 N804JB Fixe... A320...
## 4
      2013
              1991
                                    554
                                             812 DL
                                                            461 N668DN
                                                                        Fixe... 757-...
## 5
```

Other dplyr goodies

pull(): Extract a column from a data frame as a vector or scalar.

• e.g. starwars %>% filter(gender="female") %>% pull(height)

count() and distinct(): Number and isolate unique observations.

- e.g. starwars %>% count(species), Of starwars %>% distinct(species)
- You could also use a combination of mutate(), group_by(), and n(), e.g. starwars
 %>% group_by(species) %>% mutate(num = n())
- Built-in combination using add_count().

Other dplyr goodies

There is also a whole class of window functions for getting leads and lags, ranking, creating cumulative aggregates, etc.

See vignette("window-functions") for more.

Some dplyr tips

- Any group_by() statement stays until ungroup()
- Look out for plyr package. Do not use plyr and dplyr together. Just don't do it.



tidyr

Key tidyr verbs

- 1. pivot_wider() and pivot_longer() to reshape data between wide and long format
- 2. separate(): Split one column into multiple columns
- 3. unite(): Combine multiple columns into one

1) tidyr::pivot_longer()

```
stocks ← tibble(
  time = as.Date('2009-01-01') + 0:1,
  X = rnorm(2, 0, 1),
  Y = rnorm(2, 0, 2),
  Z = rnorm(2, 0, 4)
  )
stocks
```

1) tidyr::pivot_longer()

```
## # A tibble: 6 × 3
    time stock price
###
   <date> <chr> <dbl>
##
## 1 2009-01-01 X
                 0.192
## 2 2009-01-01 Y -1.62
## 3 2009-01-01 Z
                    -0.780
  4 2009-01-02 X
                    -0.907
  5 2009-01-02 Y
                    -0.332
  6 2009-01-02 Z
                     6.00
```

1) tidyr::pivot_longer()

Aside: Remembering the syntax

There's a long-running joke about no-one being able to remember Stata's "reshape" command. (Exhibit A.)

It's easy to see this happening with pivot_wider() and pivot_longer() too.

1) tidyr::pivot_wider()

2 2009-01-02 -0.907 -0.332 6.00

2) tidyr::separate()

<chr>

1 Abhijit Banerjee

2 Esther Duflo

3 Michael Kremer

##

```
economists 		 tibble(
  name = c("Abhijit Banerjee", "Esther Duflo", "Michael Kremer")
)
economists
### # A tibble: 3 × 1
### name
```

2) tidyr::separate()

Should also specify the separation character with separate(..., sep=" ").

3) tidyr::separate_rows()

A related function is separate_rows() for splitting into new rows

```
jobs ← tibble(
  name = c("Jack", "Jill"),
  occupation = c("Homemaker", "Philosopher, Philanthropist, Troublemaker")
  )
jobs
```

```
## # A tibble: 2 × 2
## name occupation
## <chr> <chr>
## 1 Jack Homemaker
## 2 Jill Philosopher, Philanthropist, Troublemaker
```

3) tidyr::separate_rows()

```
## Now split out Jill's various occupations into different rows
jobs %>% separate_rows(occupation)
```

```
## # A tibble: 4 × 2
## name occupation
## <chr> <chr> <hr> ## 1 Jack Homemaker
## 2 Jill Philosopher
## 3 Jill Philanthropist
## 4 Jill Troublemaker
```

```
gdp ← data.frame(
    yr = rep(2016, times = 4),
    mnth = rep(1, times = 4),
    dy = 1:4,
    gdp = rnorm(4, mean = 100, sd = 2)
    )
gdp
```

```
## Combine "yr", "mnth", and "dy" into one "date" column
gdp %>% unite(date, c("yr", "mnth", "dy"), sep = "-")
```

```
## date gdp
## 1 2016-1-1 98.72684
## 2 2016-1-2 98.91063
## 3 2016-1-3 99.23657
## 4 2016-1-4 105.45198
```

unite() automatically creates character variable:

Use mutate() with lubridate date functions to change the variable type.

```
library(lubridate)
gdp_u %>% mutate(date = ymd(date))

## # A tibble: 4 × 2

## date gdp

## <date> <dbl>
## 1 2016-01-01 98.7

## 2 2016-01-02 98.9

## 3 2016-01-03 99.2

## 4 2016-01-04 105.
```

Other tidyr goodies

- drop_na() to drop missing values among specified columns
- fill() to impute missing values from past/future values
- replace_na() to replace missing values with known value

Summary

dplyr

- 1. filter()
- 2. arrange()
- 3. select()
- 4. mutate()
- 5. summarise()

tidyr

- 1. pivot_longer()
- 2. pivot_wider()
- 3. separate()
- 4. unite()

Other useful items include: pipes (%>%), grouping (group_by()), joining functions (left_join(), inner_join, etc.).

Real World

Practice data versus the real world



Advice 1: Be patient and careful in your coding



Advice 2: Comment, comment, comment

You don't want to end up like this guy...



Medicare Advantage

Medicare Advantage

Let's work with the Medicare Advantage GitHub repository

Download the data

First step is to download the raw data that we'll be using.

- Monthly Enrollment
- Plan Characteristics
- Service Areas

Lots more out there, but this is enough for now.

```
for (y in 2006:2015) {
  monthlist=get(paste0("monthlist ",y))
  for (m in monthlist) {
    ## Basic contract/plan information
    ma.path=paste0("...CPSC Contract Info ",y," ",m,".csv")
    contract.info=read csv(ma.path,
                            skip=1,
                            col names = c("contractid", "planid", "org type", "plan type",
                                          "partd", "snp", "eghp", "org name", "org marketing na
                                          "plan name", "parent org", "contract date"),
                            col types = cols(
                              contractid = col character(),
                              planid = col double(),
                              org type = col character(),
                              plan type = col character(),
                              partd = col character(),
                              snp = col character(),
```

```
## Clean the contract level data
contract.info = contract.info %>%
    group_by(contractid, planid) %>%
    mutate(id_count=row_number())

contract.info = contract.info %>%
    filter(id_count=1) %>%
    select(-id_count)
```

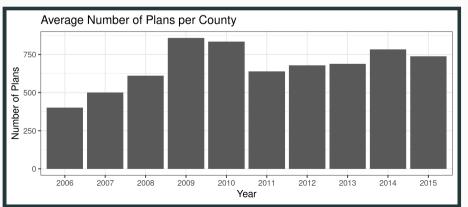
```
## Enrollments per plan
ma.path=paste0("...CPSC_Enrollment_Info_",y,"_",m,".csv")
enroll.info=read csv(ma.path,
                     skip=1,
                     col_names = c("contractid", "planid", "ssa", "fips", "state", "county"
                     col types = cols(
                       contractid = col character(),
                       planid = col double(),
                       ssa = col double(),
                       fips = col double(),
                       state = col character(),
                       county = col character(),
                       enrollment = col double()
                     ),na="*")
```

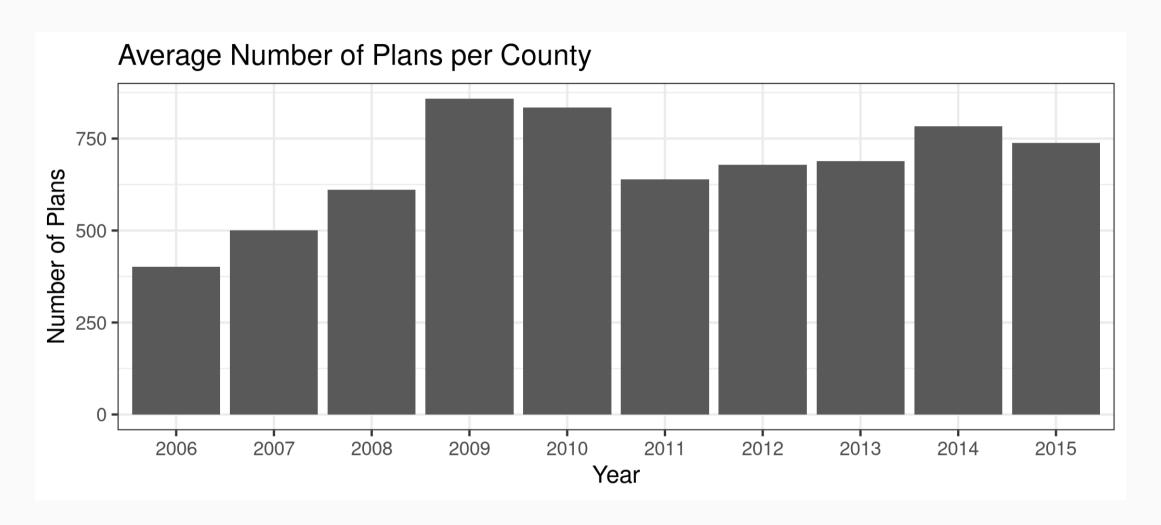
```
## Merge contract info with enrollment info
plan.data = contract.info %>%
   left_join(enroll.info, by=c("contractid", "planid")) %>%
   mutate(month=as.numeric(m),year=y)
assign(paste0("plan.data.",y,".",m),plan.data)
}
```

```
## Fill in missing fips codes (by state and county)
plan.month = plan.month %>%
  group by(state, county) %>%
  fill(fips)
## Fill in missing plan characteristics by contract and plan id
plan.month = plan.month %>%
  group by(contractid, planid) %>%
  fill(plan type, partd, snp, eghp, plan name)
## Fill in missing contract characteristics by contractid
plan.month = plan.month %>%
  group by(contractid) %>%
 fill(org_type,org_name,org_marketing_name,parent_org)
```

```
## Collapse from monthly data to yearly
  plan.year = plan.month %>%
    group by(contractid, planid, fips) %>%
    arrange(contractid,planid,fips,month) %>%
    summarize(avg enrollment=mean(enrollment),sd enrollment=sd(enrollment),
              min_enrollment=min(enrollment), max_enrollment=max(enrollment),
              first enrollment=first(enrollment), last enrollment=last(enrollment),
              state=last(state), county=last(county), org type=last(org type),
              plan type=last(plan type),partd=last(partd),snp=last(snp),
              eghp=last(eghp),org name=last(org name),org marketing name=last(org marketin
              plan name=last(plan name), parent org=last(parent org), contract date=last(con
              vear=last(vear))
  write_rds(plan.year,paste0(path.data.final,"/ma_data_",y,".rds"))
full.ma.data ← readRDS(paste0(path.data.final,"/ma_data_2006.rds"))
for (y in 2007:2015) {
                                                                                          88 / 95
  full.ma.data ← rbind(full.ma.data, paste0(path.data.final,"/ma_data_",y,".rds"))
```

```
full.ma.data %>% group_by(fips, year) %>% select(fips, year) %>% summarize(plan_count=n())
    ggplot(aes(x=as.factor(year),y=plan_count)) +
    stat_summary(fun.y="mean", geom="bar") +
    labs(
         x="Year",
         y="Number of Plans",
         title="Average Number of Plans per County"
    ) + scale_y_continuous(labels=comma) +
    theme_bw()
```





```
full.ma.data %>%
  group by(fips, year) %>%
  select(fips, year) %>%
  summarize(plan count=n()) %>%
  ggplot(aes(x=as.factor(year),y=plan_count)) +
  stat summary(fun.y="mean", geom="bar") +
  labs(
   x="Year",
    y="Number of Plans",
    title="Average Number of Plans per County"
  ) + scale_y_continuous(labels=comma) +
  theme bw()
```

```
full.ma.data %>%
  filter(snp="No" & eghp="No") %>%
  group by(fips, year) %>%
  select(fips, year) %>%
  summarize(plan_count=n()) %>%
  ggplot(aes(x=as.factor(year),y=plan count)) +
  stat summary(fun.y="mean", geom="bar") +
  labs(
   x="Year",
    y="Number of Plans",
    title="Average Number of Plans per County"
  ) + scale y continuous(labels=comma) +
  theme bw()
```

```
full.ma.data %>%
  filter(snp="No" & eghp="No") %>%
  filter(planid < 800 | planid ≥ 900) %>%
  filter(!is.na(planid)) %>%
  group_by(fips, year) %>%
  select(fips, year) %>%
  summarize(plan count=n()) %>%
  ggplot(aes(x=as.factor(year),y=plan count)) +
  stat summary(fun.y="mean", geom="bar") +
  labs(
   x="Year",
    y="Number of Plans",
    title="Average Number of Plans per County"
  ) + scale y continuous(labels=comma) +
  theme bw()
```

```
full.ma.data %>%
  filter(snp="No" & eghp="No") %>%
  filter(planid < 800 | planid ≥ 900) %>%
  filter(!is.na(planid)) %>%
  inner join(service.area %>%
               select(contractid, fips, year),
             by=c("contractid", "fips", "year")) %>%
  group_by(fips, year) %>%
  select(fips, year) %>%
  summarize(plan count=n()) %>%
  ggplot(aes(x=as.factor(year),y=plan count)) +
  stat summary(fun.y="mean", geom="bar") +
  labs(
   x="Year",
    y="Number of Plans",
    title="Average Number of Plans per County"
  ) + scale y continuous(labels=comma) +
  theme bw()
```

Interactive plot

