You can't do data science in a GUI

March 2018

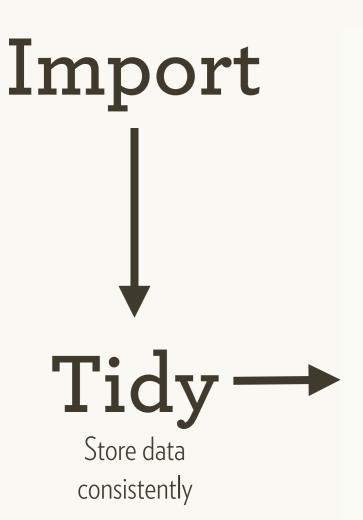
Hadley Wickham

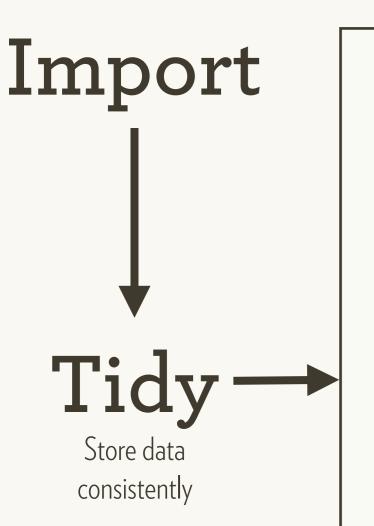
@hadleywickham
Chief Scientist, RStudio



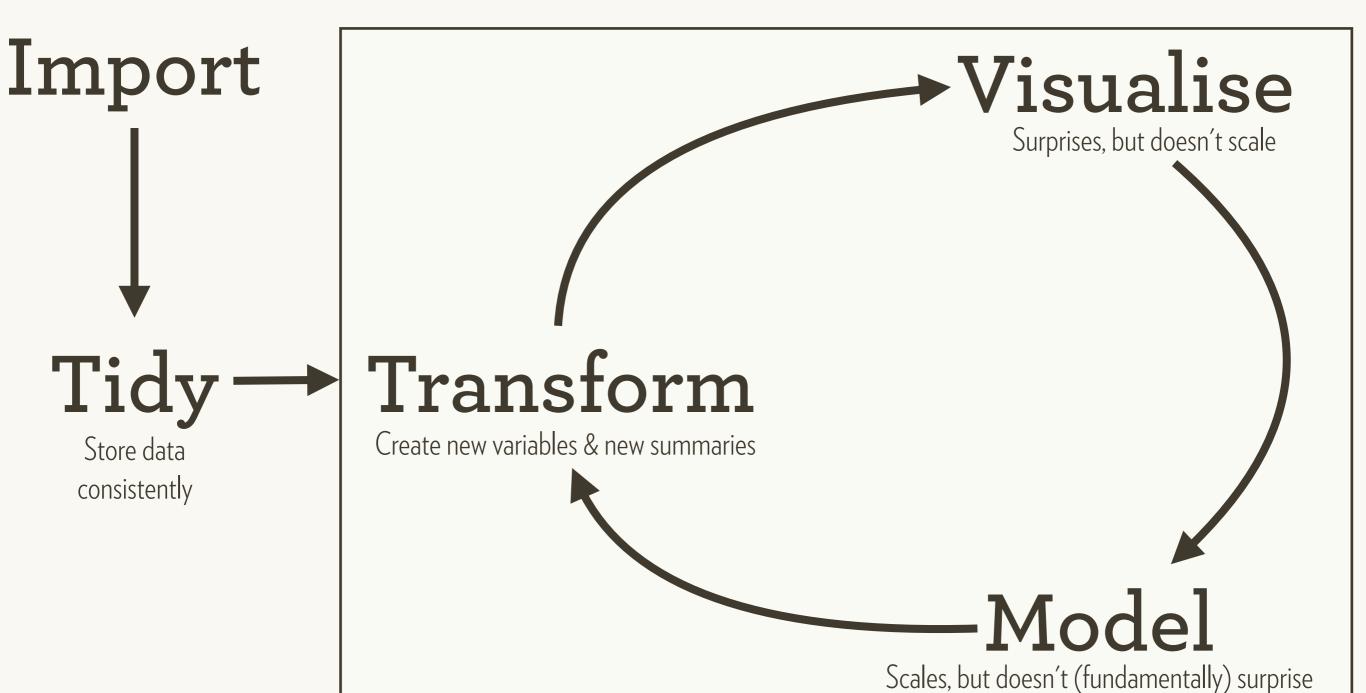
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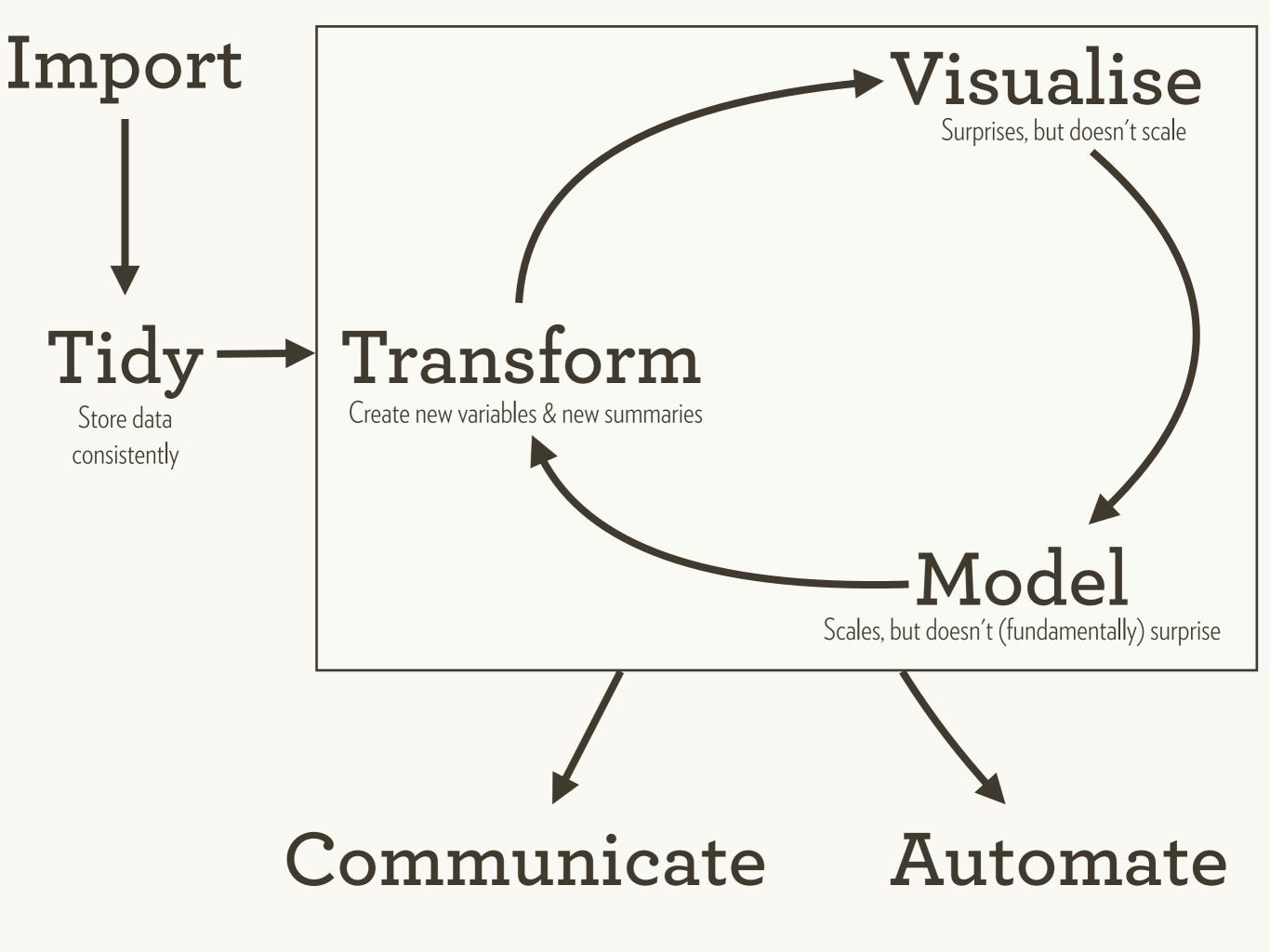
Data science is the process by which data becomes understanding, knowledge and insight

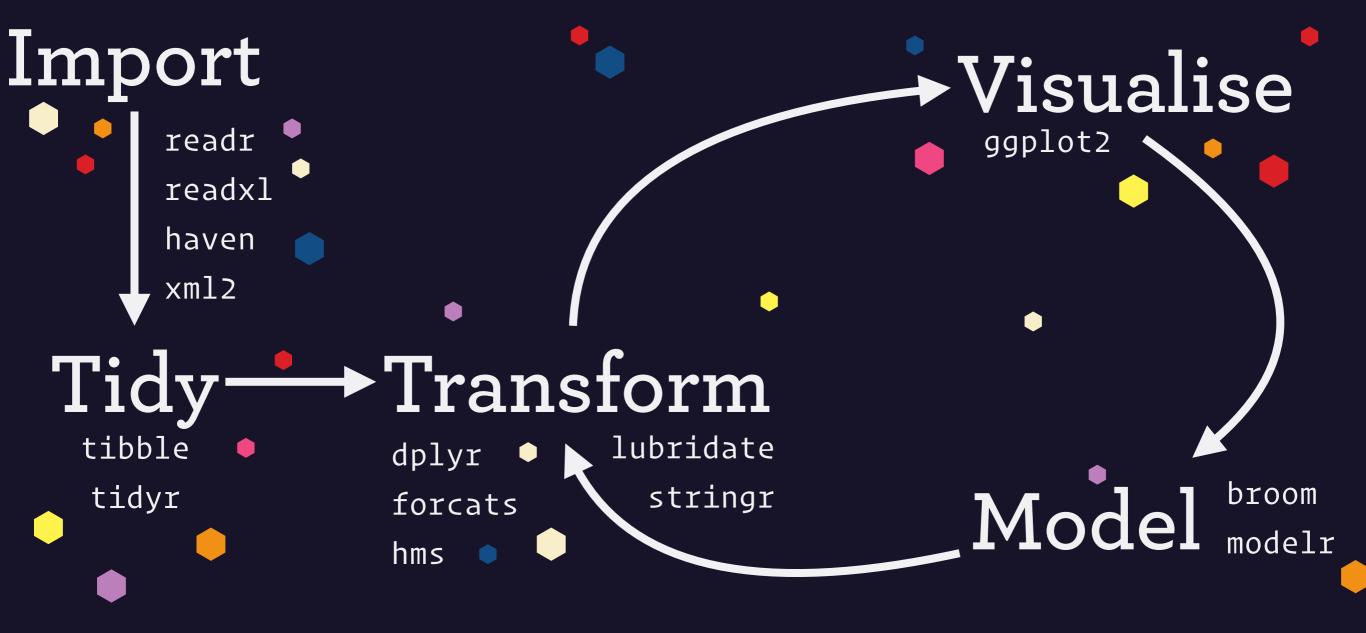




Understand





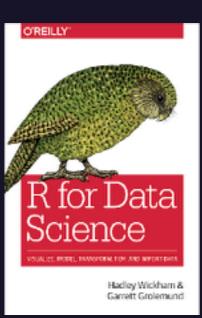






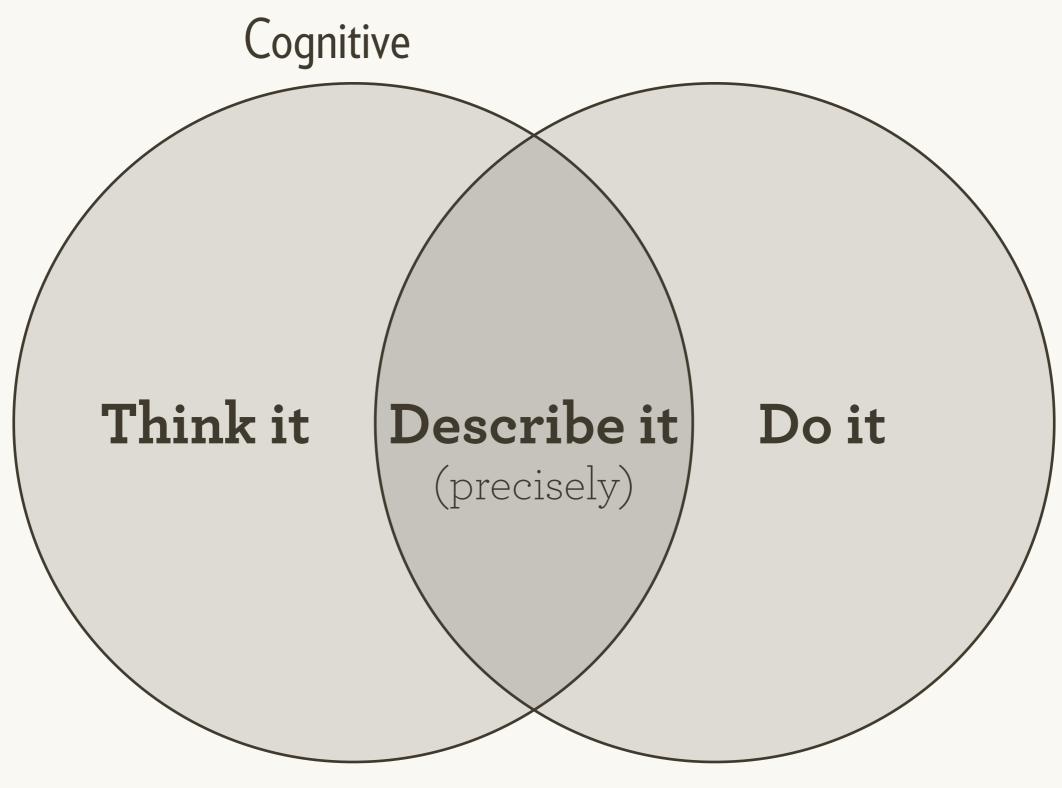


tidyverse.org

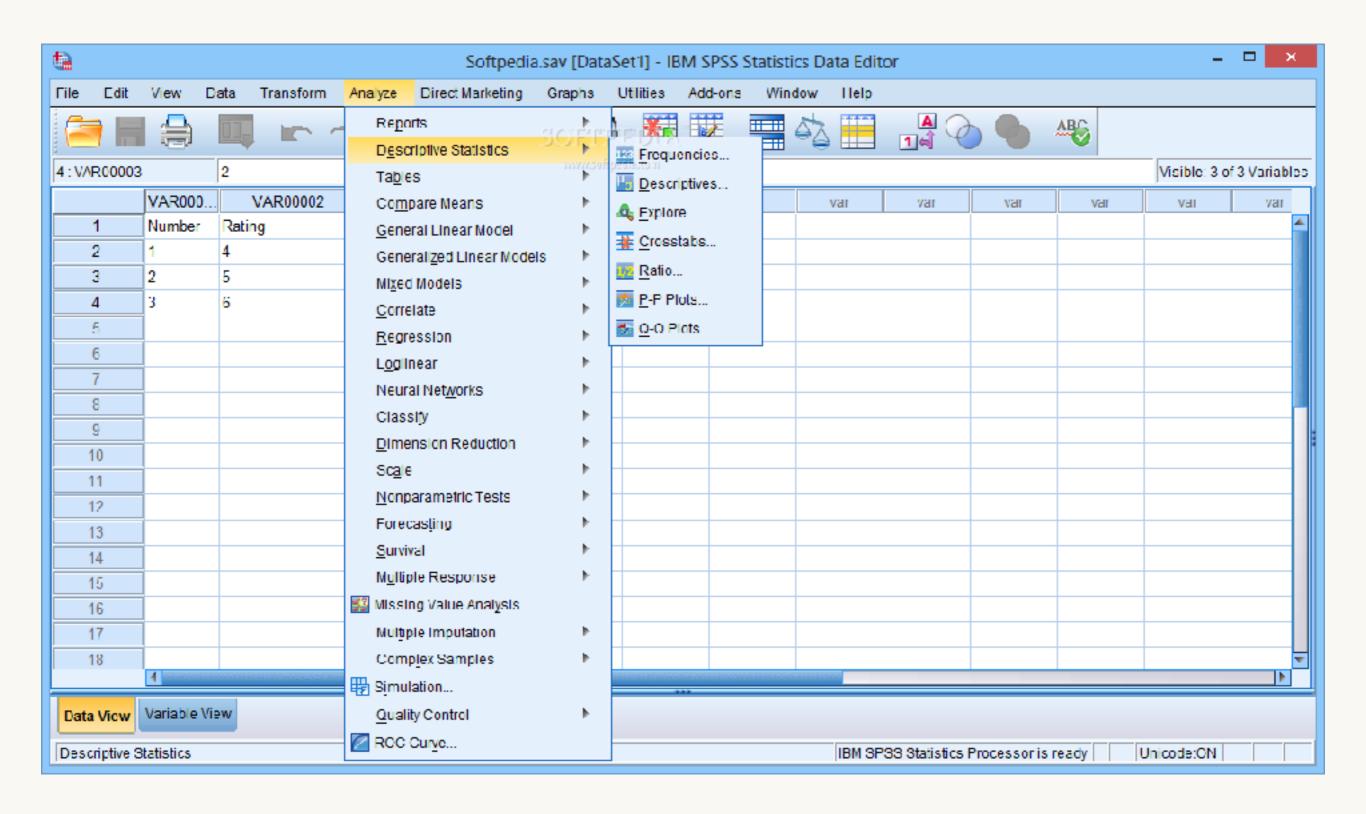


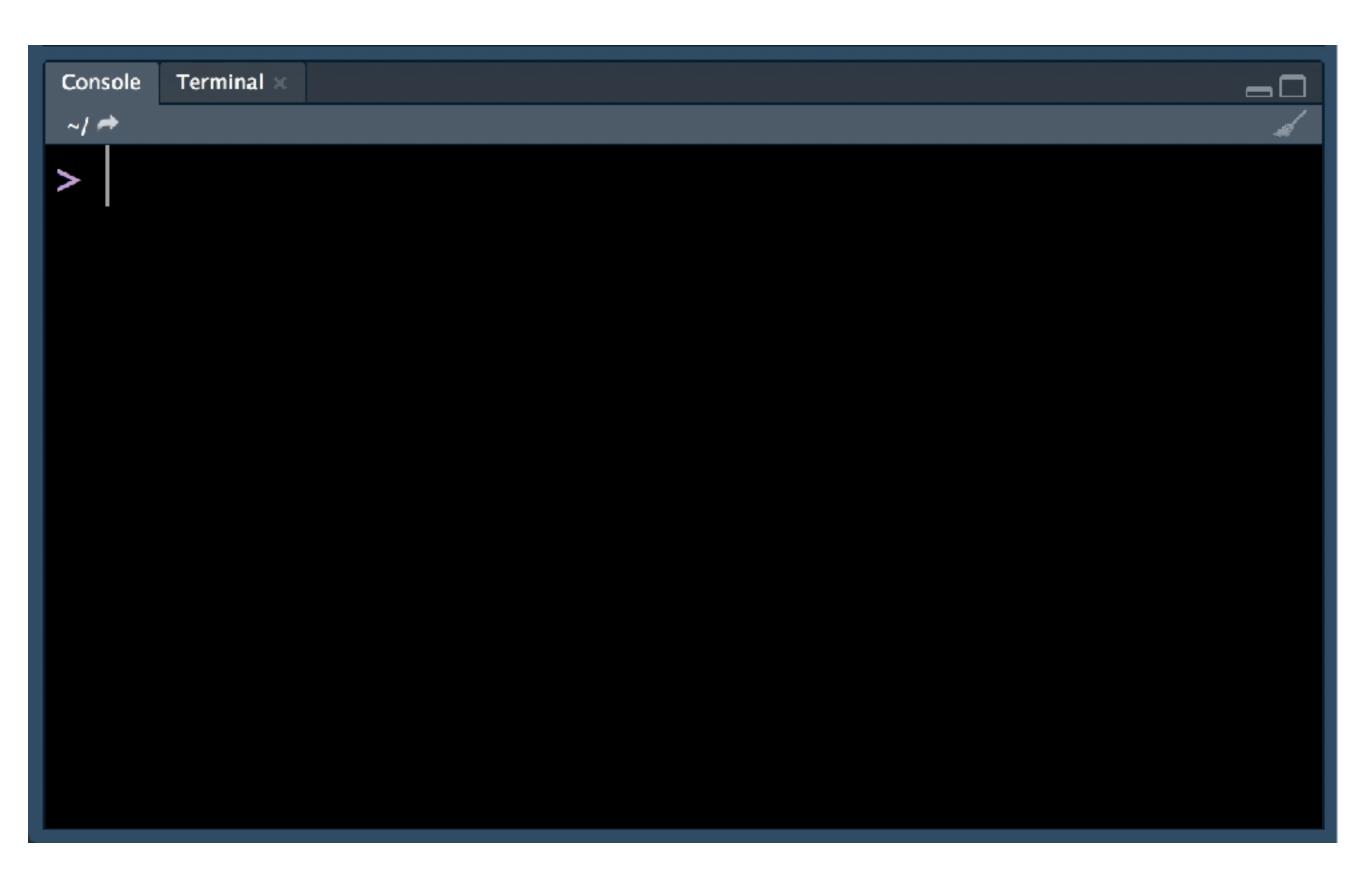
r4ds.had.co.nz

Why program?



Computational





Programming languages are languages

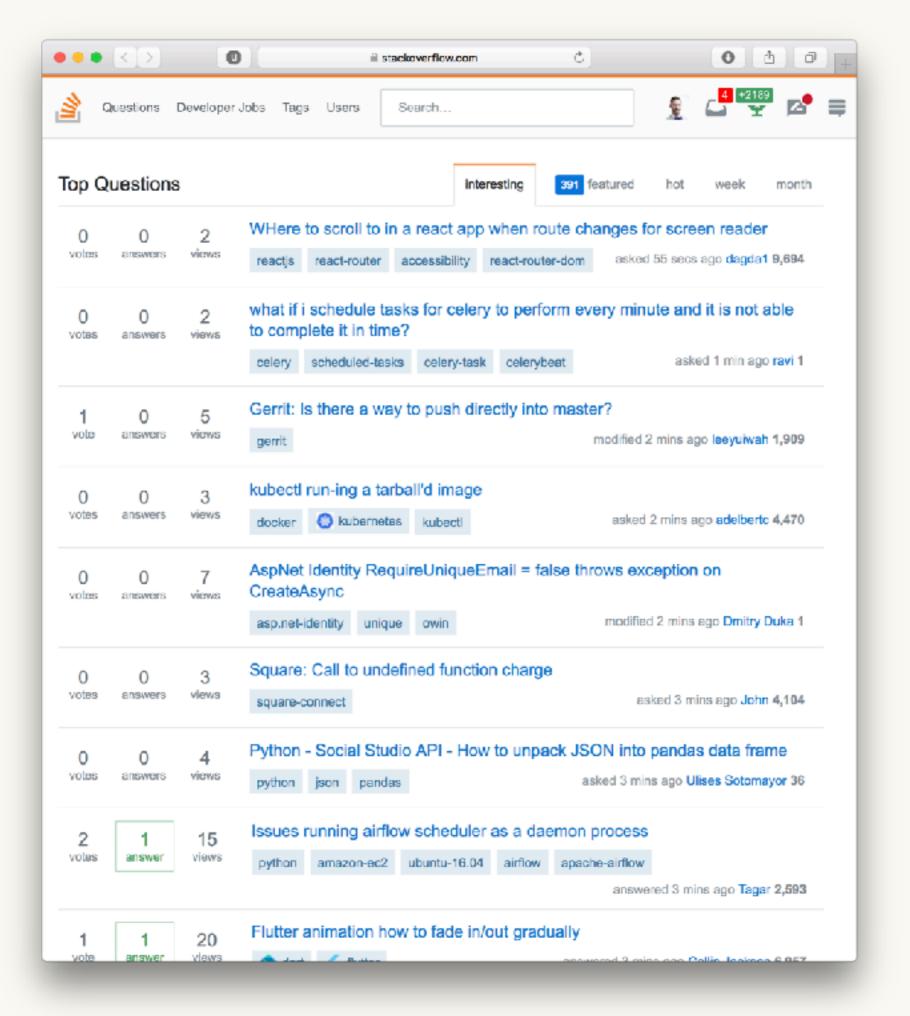
```
table %>%
  rename(player = X1, team = X2, position = X3) %>%
  filter(player != 'PLAYER') %>%
  mutate(
    college = ifelse(player == position, player, NA)
  ) %>%
  fill(college) %>%
  filter(player != college)
```

It's just text!

And this gives you access to two extremely powerful techniques

HOCO

B V

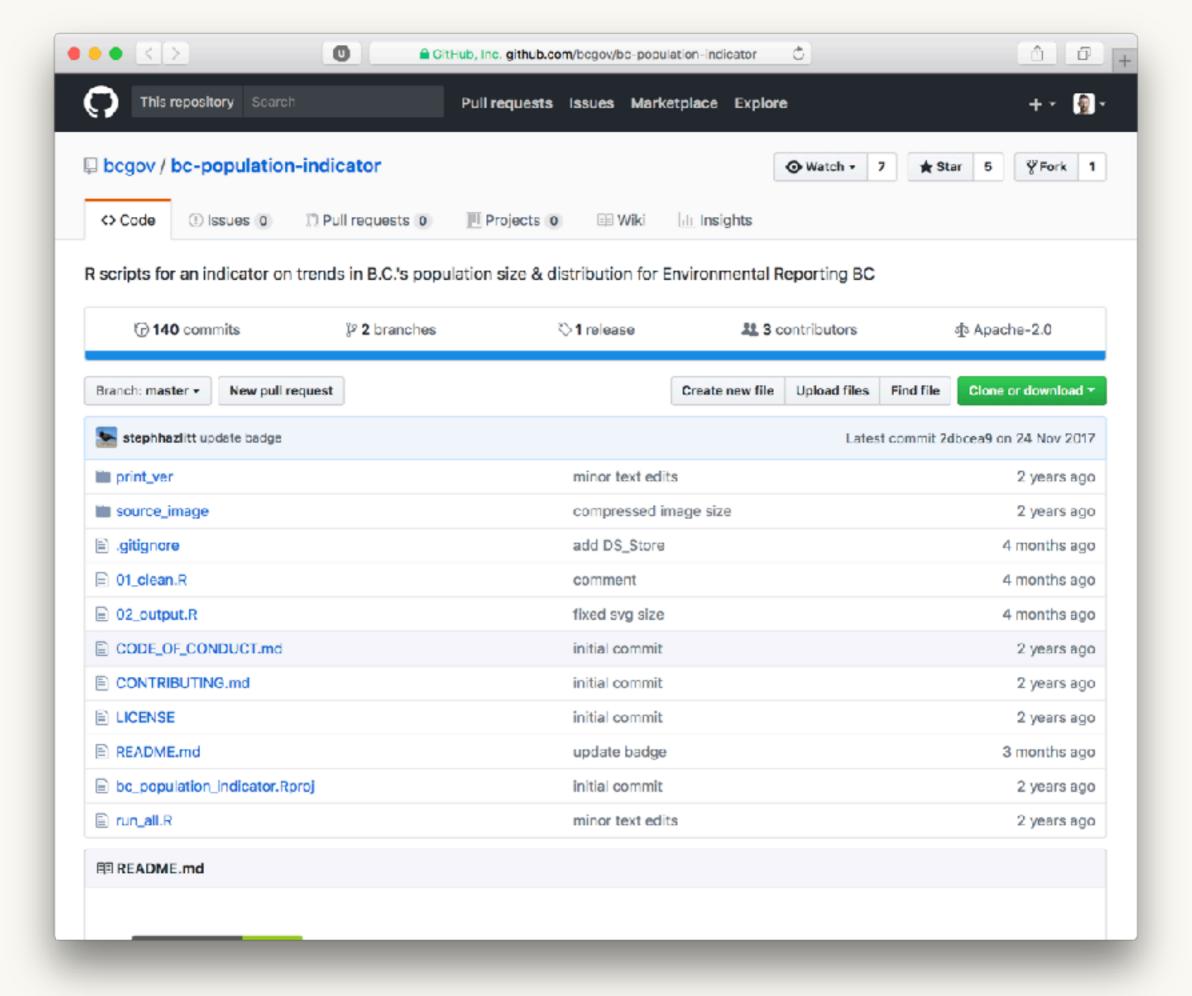


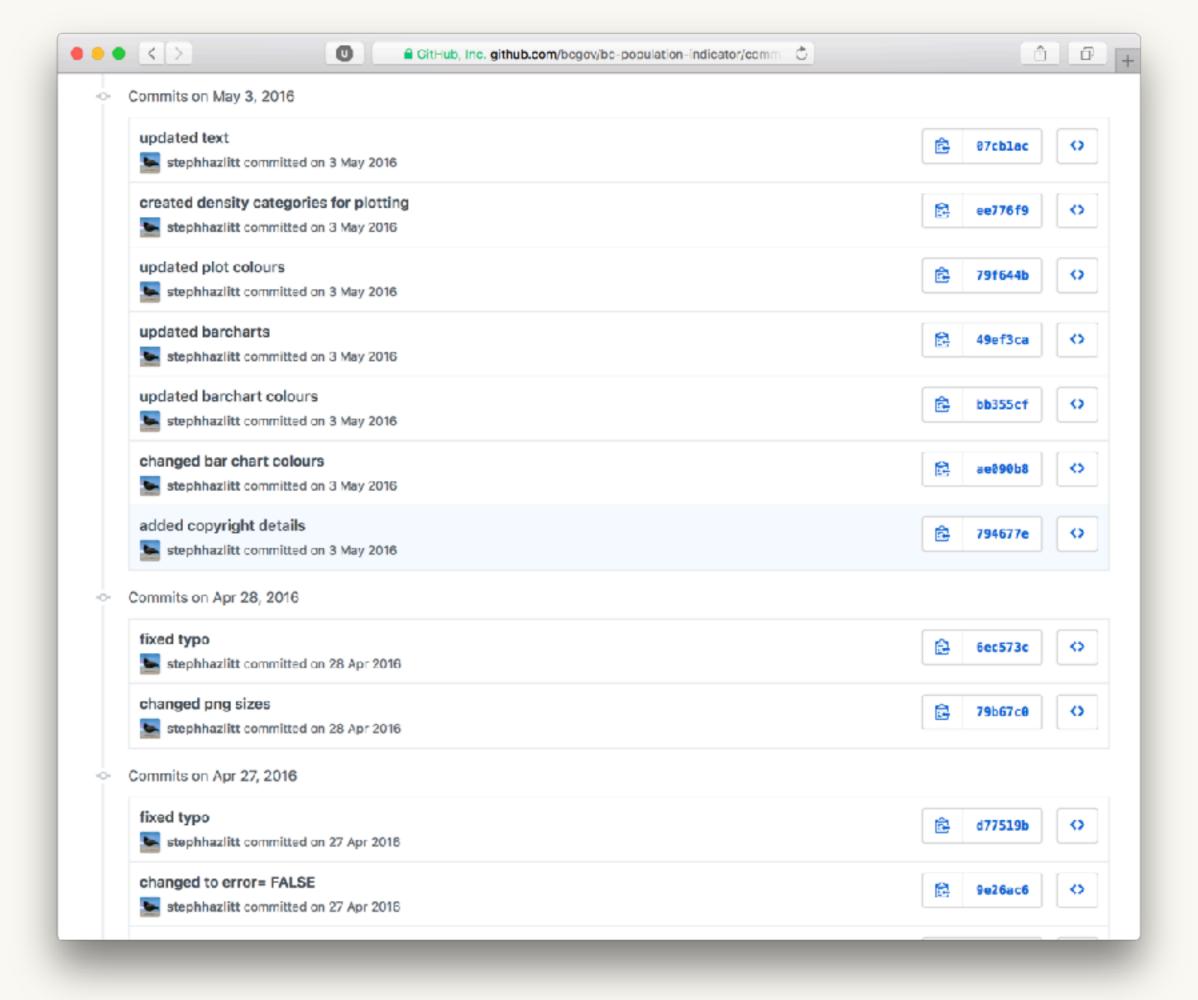
Reproducible

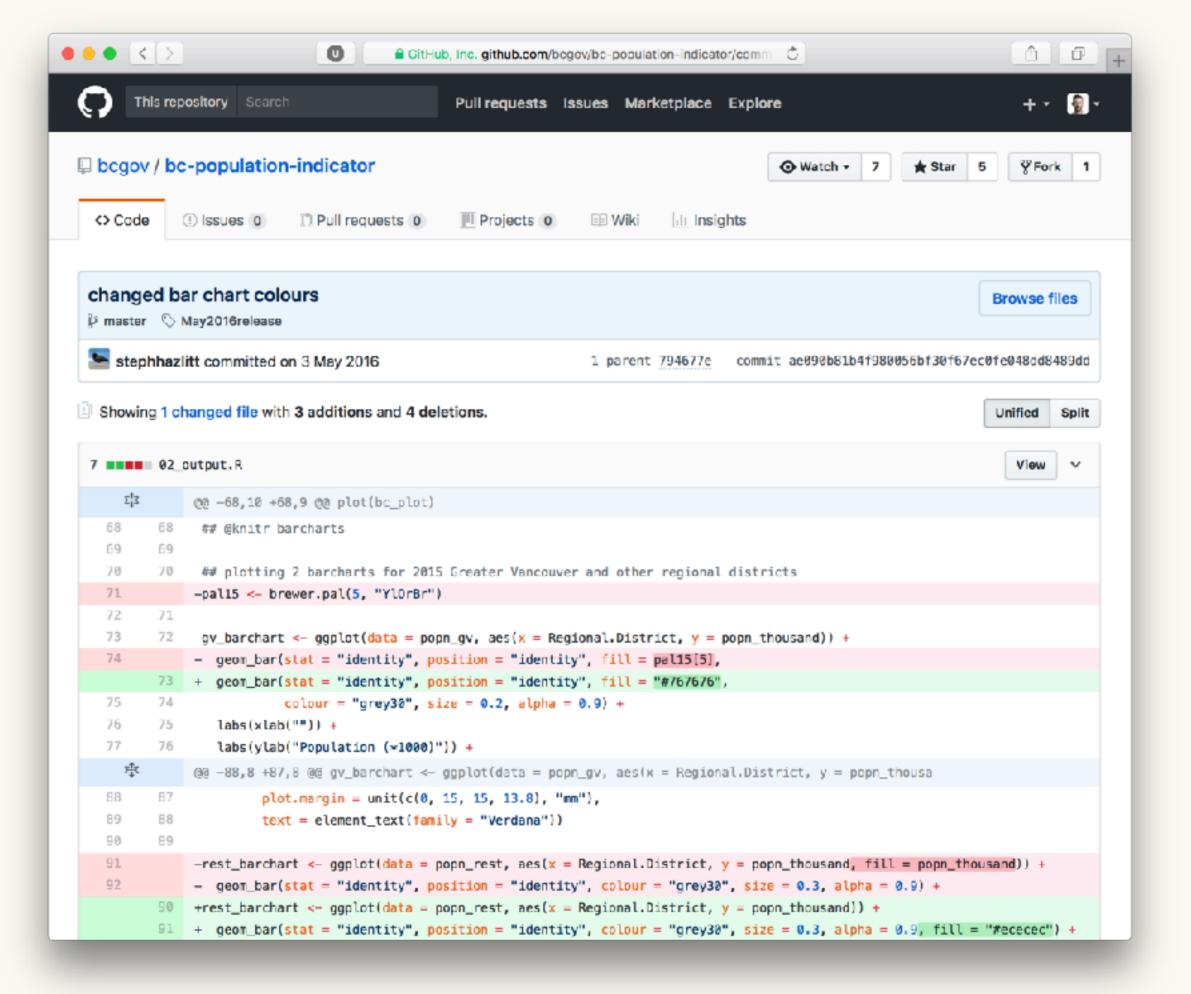
Diffable

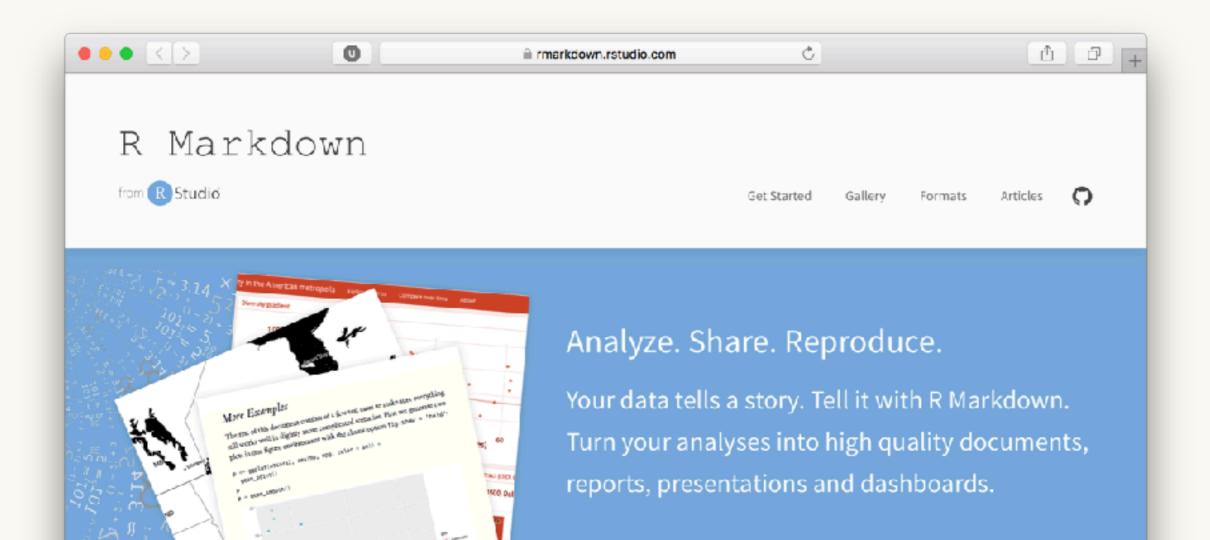
Readable

Open



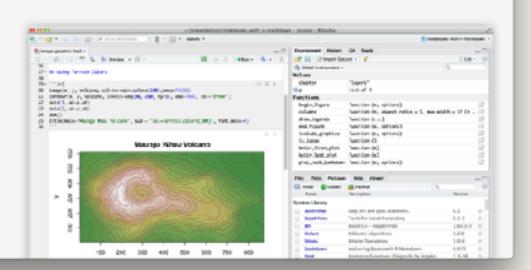


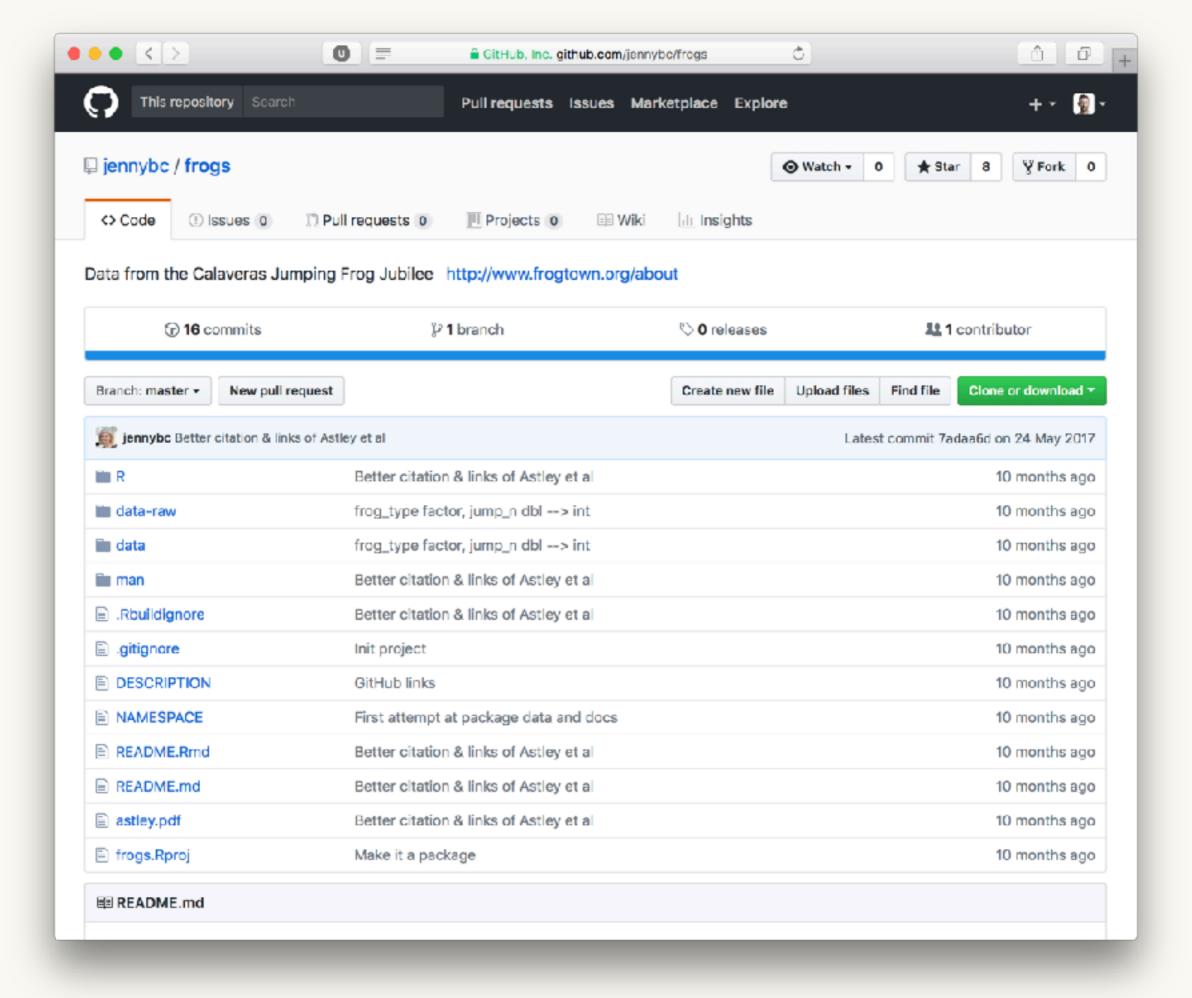




R Markdown documents are fully reproducible.

Use a productive notebook interface to weave together narrative text and code to produce





Getting to know the frogs

At this point, all we know is that each row is one frog-jump. Frog ids coming ...

```
library(frogs)
library(tidyverse)
#> + ggplot2 2.2.1
                           Date: 2017-05-24
#> + tibble 1.3.1
                            R: 3.3.2
#> + tidyr 0.6.2.9000
                           OS: OS X El Capitan 10.11.6
#> + readr 1.1.0
                            GUI: X11
#> + purrr 0.2.2.9000 Locale: en_CA.UTF-8
#> + dplyr 0.6.0
                           TZ: America/Vancouver
#> + stringr 1.2.0
\#> + forcats 0.2.0
#> Conflicts -----
#> * filter(), from dplyr, masks stats::filter()
#> * lag(), from dplyr, masks stats::lag()
frogs
#> # A tibble: 3,272 x 15
       row distance duration distance 3 jump_n frog_type distance 3 off
             <dbl>
                     <dbl>
                              <dbl> <int>
                                              <chr>
                                                           <dbl>
        1 165.950 0.58333
                                                              -1
                                               pro
        2 177.480 0.71667
                                                             -1
                                               pro
        3 0.000 0.00000
                                               pro
        4 27.158 0.43333
                                               pro
                                                             -1
      5 0.000 0.00000
                                               pro
                                                             -1
      6 0.000 0.00000
                                                             -1
                                               pro
      7 40.914 0.40000
                                                             -1
                                               pro
      8 0.000 0.00000
                                                             -1
                                               pro
                                  0 3
#> 9 0.000 0.00000
                                                             -1
                                               pro
-1
                                               pro
#> # ... with 3,262 more rows, and 8 more variables: distance_rel <dbl>,
#> # day <dbl>, angle_01 <dbl>, angle_10 <dbl>, angle_00 <dbl>,
#> # velocity_01 <dbl>, velocity_10 <dbl>, velocity_00 <dbl>
glimpse(frogs)
#> Observations: 3,272
#> Variables: 15
```

Getting to know the frogs

At this point, all we know is that each row is one frog-jump. Frog ids coming ...

```
library(frogs)
library(tidyverse)

frogs
glimpse(frogs)
```

An early figure. Do frogs need to warm up? Do they fatigue? Yes and yes.

```
frogs2 <- frogs %>%
  filter(jump_n < 7) %>%
  mutate(
    jump_n = as.factor(as.integer(jump_n))
)

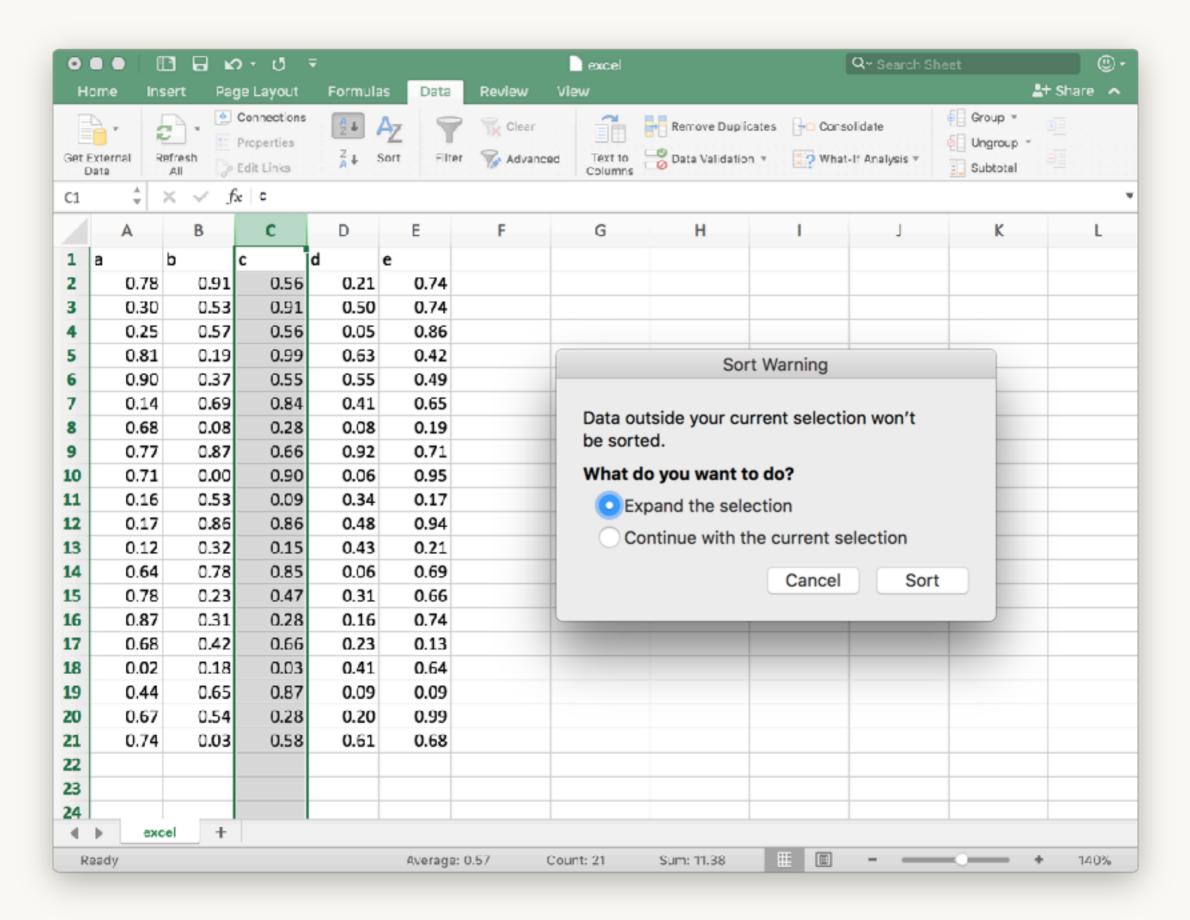
ggplot(frogs2, aes(x = distance, color = jump_n)) +
  geom_density()
```

Do professional frog jumping teams get better results? YES.

```
ggplot(frogs, aes(x = distance, color = frog_type)) +
  geom_density()
```

```
û D
                      i raw.githubusercontent.com/jennybc/frogs/master/R
# install.packages("devtools")
devtools::install_github("jennybc/frogs")
## Getting to know the frogs
At this point, all we know is that each row is one frog-jump. Frog ids coming ...
```{r}
library(frogs)
library(tidyverse)
frogs
glimpse(frogs)
An early figure. Do frogs need to warm up? Do they fatigue? Yes and yes.
```{r frog-fatigue, echo = FALSE}
frogs2 <- frogs %>%
  filter(jump n < 7) %>%
  mutate(
    jump n = as.factor(as.integer(jump n))
ggplot(frogs2, aes(x = distance, color = jump n)) +
  geom_density()
Do professional frog jumping teams get better results? YES.
```{r frog-type, echo = FALSE}
ggplot(frogs, aes(x = distance, color = frog type)) +
geom_density()
```

#### I live in fear of clicking the wrong thing



# Why program in R?

#### R is a vector language

```
x <- sample(100, 10)
x > 50
#> [1] TRUE FALSE FALSE TRUE TRUE
#> [6] TRUE TRUE FALSE FALSE TRUE
sum(x > 50)
#> [1] 6
(There are no scalars! 🚱)
```

#### Missing values are baked in

```
y < - sample(c(1:5, NA))
#> [1] 1 NA 2 3 5 4
y > 2
#> [1] FALSE NA FALSE TRUE TRUE TRUE
y == NA
#> [1] NA NA NA NA NA NA
```

#### An example makes this clearer

```
john_age <- NA
mary_age <- NA

john_age == mary_age
#> [1] NA
```

#### Missing values are baked in

```
y \leftarrow sample(c(1:5, NA))
#> [1] 1 NA 2 3 5 4
y > 2
#> [1] FALSE NA FALSE TRUE TRUE TRUE
is.na(y)
#> [1] FALSE TRUE FALSE FALSE FALSE
```

#### So are relational tables (aka data frames/tibbles)

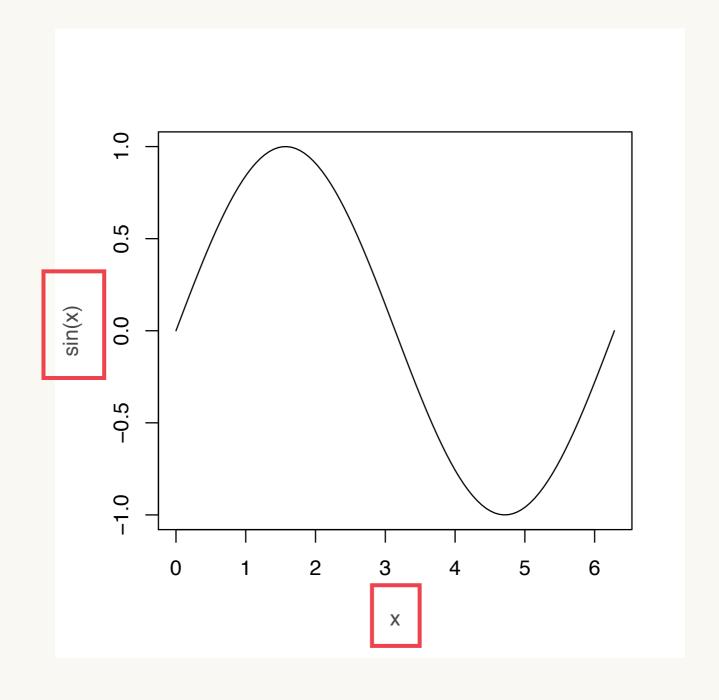
```
data.frame(
 x = 1:4,
 y = sample(letters[1:4]),
 z = runif(4)
#> x y
#> 1 1 c 0.1189635
#> 2 2 a 0.0518956
#> 3 3 b 0.4471441
#> 4 4 d 0.0818547
```

#### Functional programming

```
It's well suited to data science but I
can't (yet) articulate why
Something about having a standard
container for 80% of problems, and
needing to do something to each element
of that container
Whole object thinking?
```

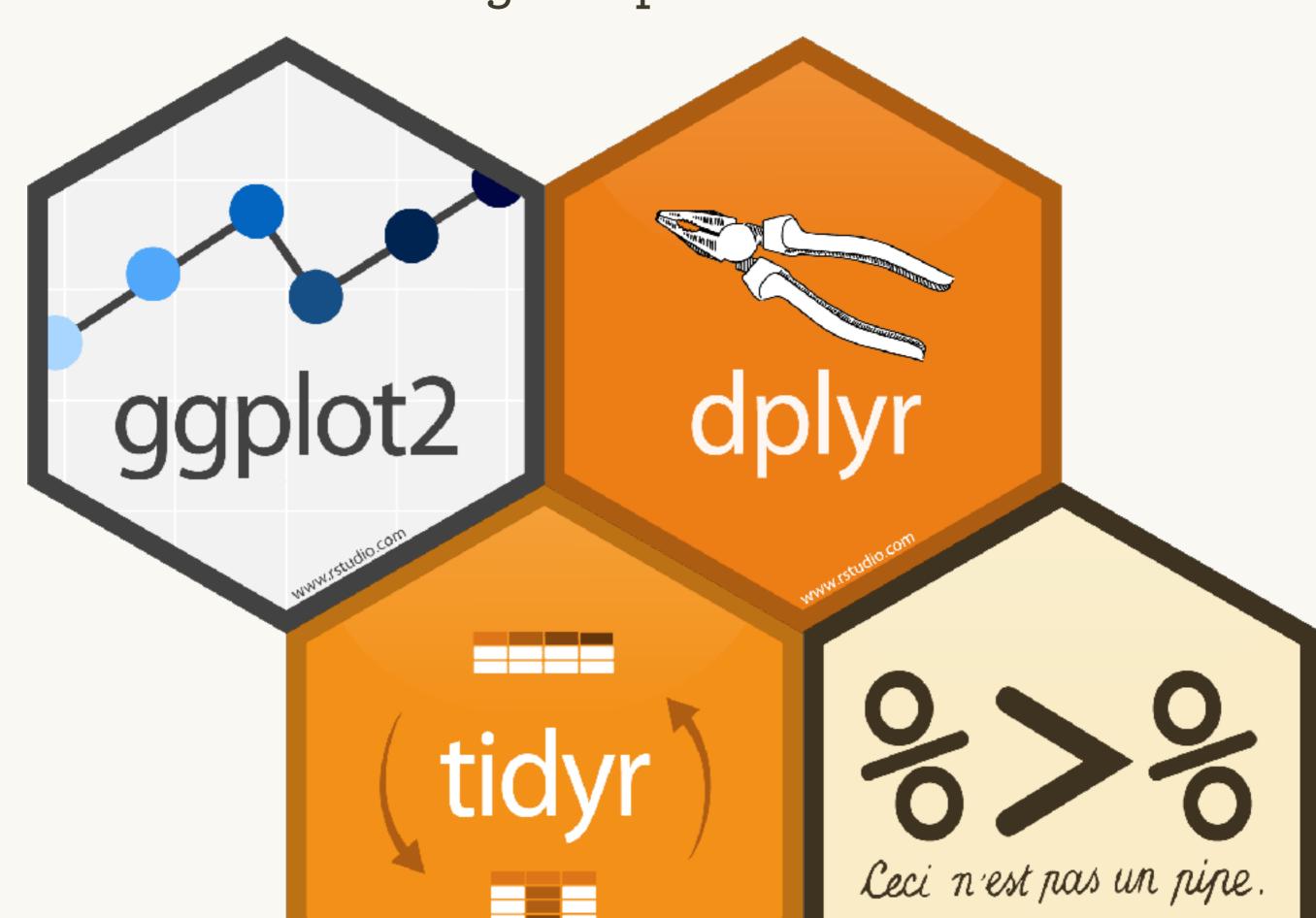
#### Metaprogramming

```
x <- seq(0, 2 * pi, length = 100)
plot(x, sin(x), type = "l")
```



```
> lobstr::ast(y <-1 + 1 + 2 * 3)
_^<-`
 | | 1
```

Which makes it a great place to write DSLs



# Why program in R with the tidyverse?



https://unsplash.com/photos/tjX\_sniNzgQ

# A small example

```
library(tidycensus)
geo <- get_acs(</pre>
 geography = "metropolitan statistical area...",
 variables = "DP03_0021PE",
 summary_var = "B01003_001",
 survey = "acs1",
 endyear = 2016
Thanks to Kyle Walker (@kyle_e_walker)
For package and example
```

# A tibble: 518 x 7

|    | GEOID           | NAME                                     | variable    | estimate    | moe         | summary_est    | summar              |
|----|-----------------|------------------------------------------|-------------|-------------|-------------|----------------|---------------------|
|    | <chr>&gt;</chr> | <chr></chr>                              | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl>    | <dbl></dbl>         |
| 1  | 10140           | "Aberdeen, WA Micro Area"                | DP03_0021P  | 1.50        | 1.10        | <b>716</b> 28  | -5.56e <sup>8</sup> |
| 2  | 10180           | "Abilene, TX Metro Area"                 | DP03_0021P  | 1.20        | 0.600       | <b>170</b> 860 | 2.84e³              |
| 3  | 10300           | "Adrian, MI Micro Area"                  | DP03_0021P  | 0.100       | 0.100       | 98504          | -5.56e <sup>8</sup> |
| 4  | 10380           | "Aguadilla-Isabela, PR Metro Area"       | DP03_0021P  | 0.400       | 0.500       | 309764         | 1.96e³              |
| 5  | 10420           | "Akron, OH Metro Area"                   | DP03_0021P  | 1.10        | 0.300       | <b>702</b> 221 | -5.56e <sup>8</sup> |
| 6  | 10460           | "Alamogordo, NM Micro Area"              | DP03_0021P  | 0.700       | 1.00        | 65410          | -5.56e <sup>8</sup> |
| 7  | 10500           | "Albany, GA Metro Area"                  | DP03_0021P  | 0.400       | 0.400       | <b>152</b> 506 | 2.13e³              |
| 8  | 10540           | "Albany, OR Metro Area"                  | DP03_0021P  | 0           | 0.100       | 122849         | -5.56e8             |
| 9  | 10580           | "Albany-Schenectady-Troy, NY Metro Area" | DP03_0021P  | 4.00        | 0.600       | <b>881</b> 839 | -5.56e <sup>8</sup> |
| 10 | 10700           | "Albertville, AL Micro Area"             | DP03_0021P  | 0.800       | 0.800       | <b>951</b> 57  | -5.56e <sup>8</sup> |
| #  | wit             | h 508 more rows                          |             |             |             |                |                     |

# Followed by data munging

```
big_metro <- geo %>%
 filter(summary_est > 2e6) %>%
 select(-variable) %>%
 mutate(
 NAME = gsub(" Metro Area", "", NAME)
) %>%
 separate(NAME, c("city", "state"), ", ") %>%
 mutate(
 city = str_extract(city, "^[A-Za-z]+"),
 state = str_extract(state, "^[A-Za-z]+"),
 name = paste0(city, ", ", state),
 summary_moe = na_if(summary_moe, -55555555)
```

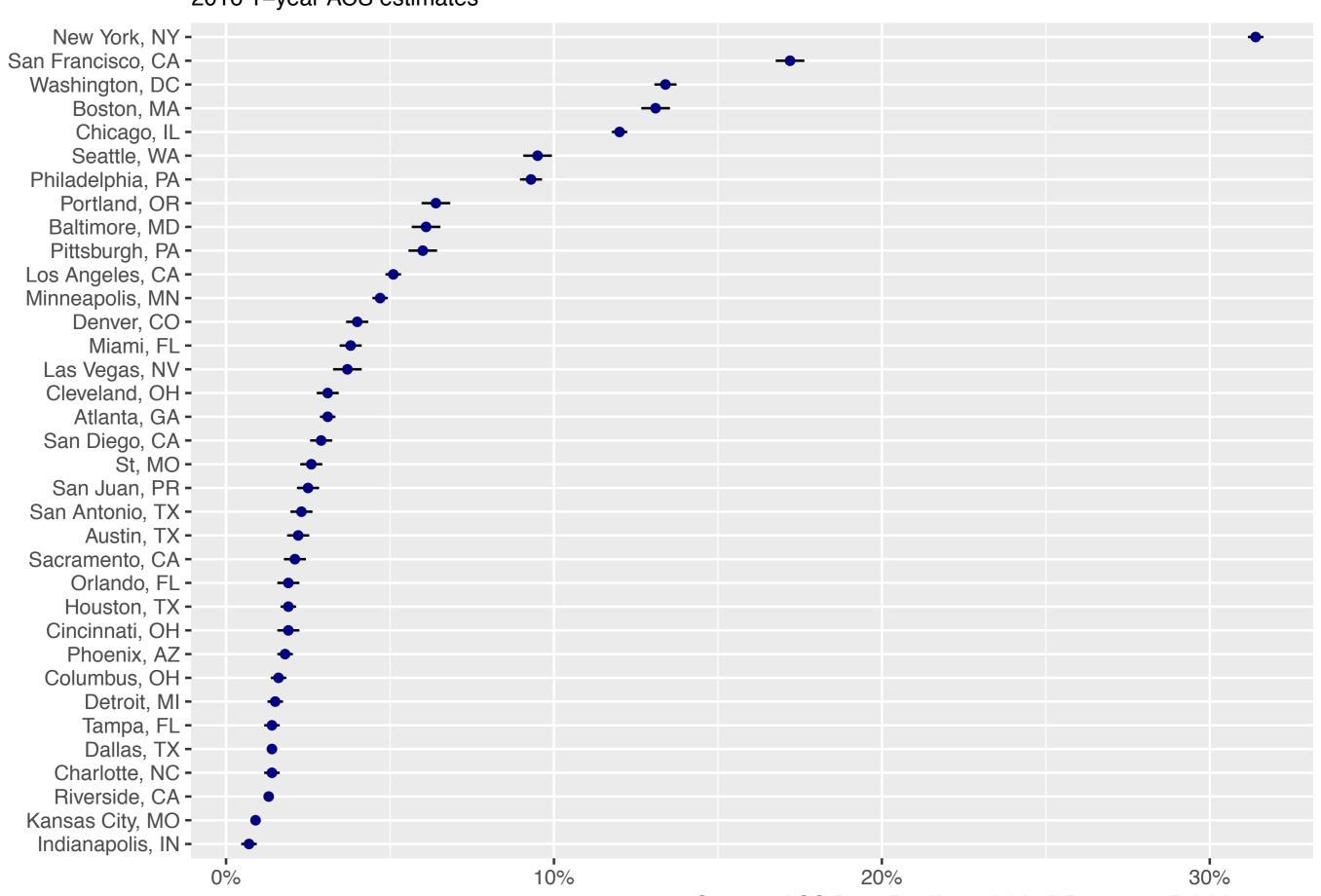
# A tibble: 35 x 8

|     | GEOID       | city         | state       | estimate    | moe         | summary_est     | summary_moe  | name             |
|-----|-------------|--------------|-------------|-------------|-------------|-----------------|--------------|------------------|
|     | <chr></chr> | <chr></chr>  | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl>     | <dbl></dbl>  | <chr></chr>      |
| 1   | 12060       | Atlanta      | GA          | 3.10        | 0.200       | <b>579</b> 0210 | 2964         | "Atlanta, GA"    |
| 2   | 12420       | Austin       | TX          | 2.20        | 0.300       | <b>205</b> 6405 | NA           | "Austin, TX"     |
| 3   | 12580       | Baltimore    | MD          | 6.10        | 0.400       | <b>279</b> 8886 | NA           | "Baltimore, MD"  |
| 4   | 14460       | Boston       | MA          | 13.1        | 0.400       | <b>479</b> 4447 | NA           | "Boston, MA"     |
| 5   | 16740       | Charlotte    | NC          | 1.40        | 0.200       | <b>247</b> 4314 | NA           | "Charlotte, NC"  |
| 6   | 16980       | Chicago      | IL          | 12.0        | 0.200       | <b>951</b> 2968 | 1542         | "Chicago, IL"    |
| 7   | 17140       | Cincinnati   | ОН          | 1.90        | 0.300       | <b>21</b> 61441 | <b>445</b> 3 | "Cincinnati, OH" |
| 8   | 17460       | Cleveland    | ОН          | 3.10        | 0.300       | <b>205</b> 5612 | NA           | "Cleveland, OH"  |
| 9   | 18140       | Columbus     | OH          | 1.60        | 0.200       | <b>204</b> 1520 | NA           | "Columbus, OH"   |
| 10  | 19100       | Dallas       | TX          | 1.40        | 0.100       | <b>723</b> 2599 | 2088         | "Dallas, TX"     |
| # . | wit         | th 25 more : | rows        |             |             |                 |              |                  |

```
big_metro %>%
 ggplot(aes(
 x = estimate,
 y = reorder(name, estimate))
) +
 geom_errorbarh(
 aes(
 xmin = estimate - moe,
 xmax = estimate + moe
 width = 0.1
) +
 geom_point(color = "navy")
```

### Residents who take public transportation to work

2016 1-year ACS estimates



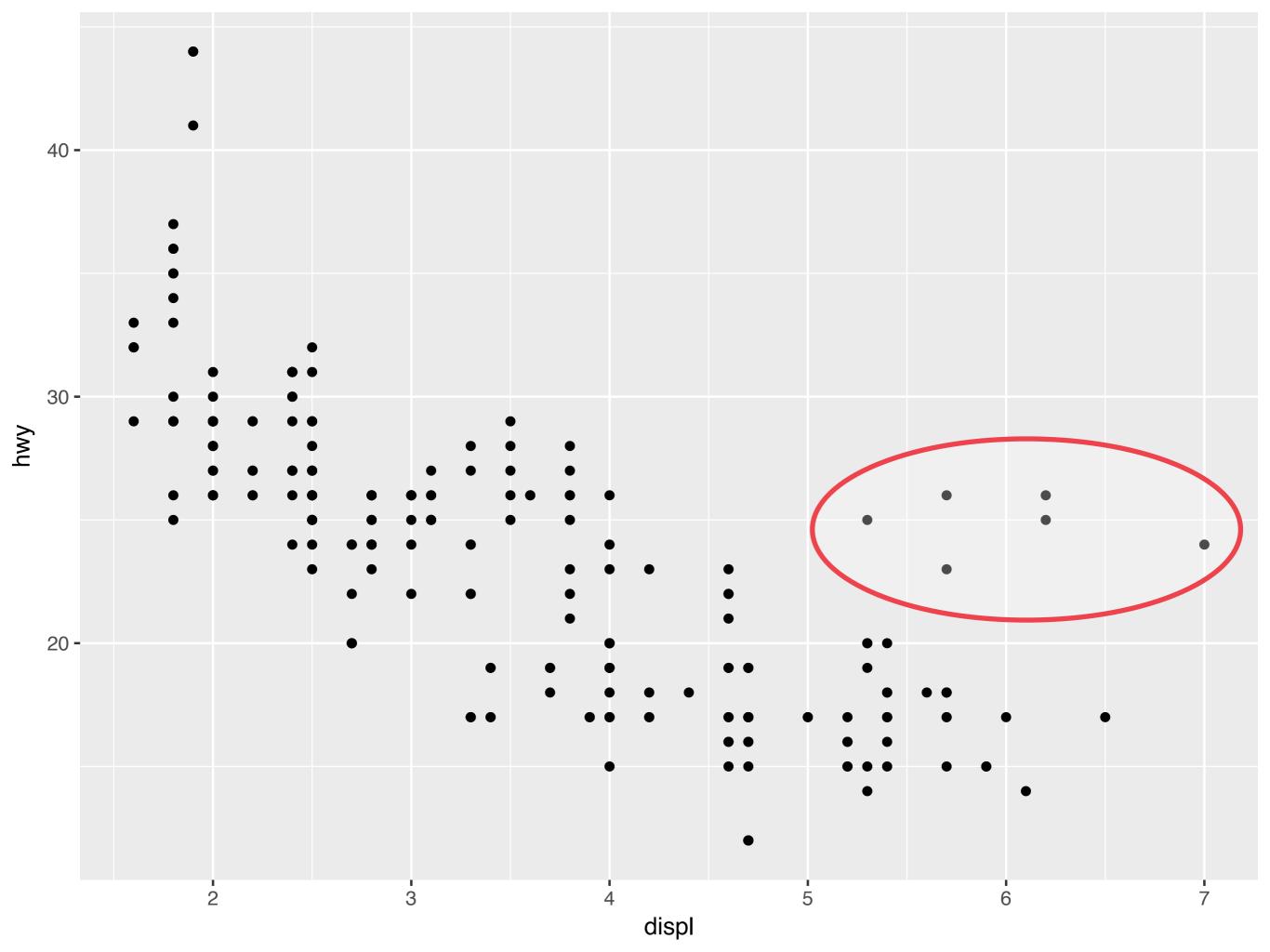
Source: ACS Data Profile variable DP03\_0021P / tidycensus

No matter how complex and polished the individual operations are, it is often the quality of the glue that most directly determines the power of the system.

— Hal Abelson

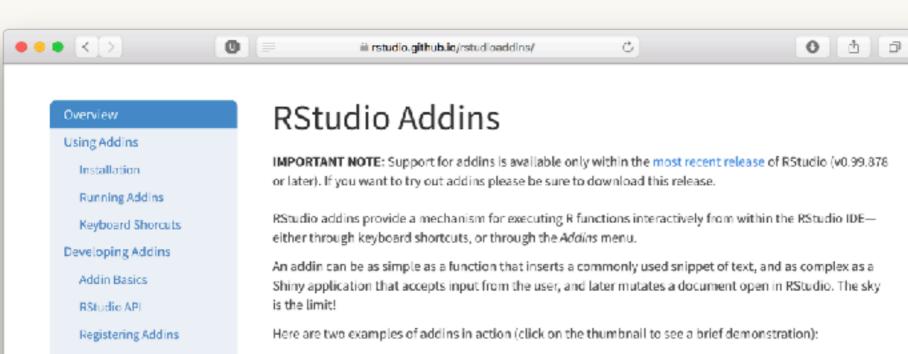


# But



# But this is painful!

```
df %>%
 select(
 date = `Date Created`,
 name = Name,
 plays = `Total Plays`,
 loads = `Total Loads`,
 apv = `Average Percent Viewed`
```



### Subset a dataset

Execution Modes

Shiny Gadgets

Gadget UI

Gadget Server

Gadget Viewer

Installation

More Examples

Putting It Together

# | The control of the

### Reformat R code



### Using Addins

This guide will walk you through the basics of installing addins, binding keyboard shorcuts to them, and finally developing your own addins.

### Installation

RStudio Addins are distributed as R packages. Once you've installed an R package that contains addins, they'll be immediately become available within RStudio.

Let's start by playing around with a couple of the example addins provided by the addinexamples package. Within RStudio, install this package (plus its requisite dependencies) with:

devtools::install\_github("rstudio/addinexamples", type = "source")

Punning Adding

# What next?

```
df %>%
 filter(n > 1e6) %>%
 mutate(x = f(y))) %>%
 ???

How predictable is next step from
previous steps?
```

# Can we do more with autocomplete?

```
abind
 acepack
 p addcol
 🔑 ash
 🔼 assertthat
 babynames
 backports
> library()
```

Where do dialogs and autocomplete intersect?

# Learning from examples

(a)

### Reported crime in Alabama

```
before:
 {'in', ''}
 'Alabama' → {'Alabama', word}
(b) selection:
 {'Alabama'}
 'in' \rightarrow {'in', word, lowercase}
 '' → {''}
 after:
 {(''), ('in', ''), (word, ''), (lowercase, '')}
 before:
(c) selection:
 {('Alabama'), (word)}
 after:
 {(),('Alabama'),()}
 \{(),(word),()\}
 {(word, ''),(),()}
 {(','),(),()}
 {(word, ''),('Alabama'),()}
 {(' '),('Alabama'),()}
(d) {(' '),(word),()}
 {(word, `'),(word),()}
 {('in', ''),(),()}
 {(lowercase, ' '),(),()}
 {('in', ''),('Alabama'),()} {(lowercase, ''),('Alabama'),()}
 {('in', ''),(word),()}
 {(lowercase, ' '),(word),()}
 \{(lowercase, '), ('Alabama'), ()\} \rightarrow /[a-z] + (Alabama)/
```

Figure 10. Regular Expression Inference. (a) The user selects text in a cell. (b) We tokenize selected and surrounding text. For clarity, the figure only includes two neighboring tokens. For each token, we generate a set of matching labels. (c) We enumerate all label sequences matching the text. (d) We then enumerate all candidate before, selection and after combinations. Patterns that do not uniquely match the selected text are filtered (indicated by strike-through). (e) Finally, we construct regular expressions for each candidate pattern.

# What about deep learning?



https://twitter.com/carroll\_jono/status/914254139873361920

# Conclusion

# I believe that:

- 1. Huge advantages to code
- 2. R provides great environment
- 3. DSLs help express your thoughts
- 4. Code should be primary artefact (but might be generated other than typing)