Analysis of bootcamp survey

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2017-08-17 16:10:02

Contents

Goals

- Download and clean data from 2017 R Bootcamp Survey
- Visualize data
- Prepare reports in `ioslides_presentation`, `pdf_document`, and `word_document` formats

Preliminaries

Load required packages.

```r
library(tidyverse)
library(googlesheets)
```

Load data and examine

The survey data are stored in a Google Sheet. We'll use the `googlesheets` package to open it and create a data frame. Documentation about the package can be found here.

There are some idiosyncrasies in using the `googlesheets` package in an R Markdown document because it requires interaction with the console, so I created a separate R script, `Get_bootcamp_googlesheet.R` to extract the survey data. If you try to execute the next chunk, it may give you an error, or it may ask you to allow `googlesheets` to access information in your Google profile.

```r
# Set eval=FALSE so I can render non-notebook formats
source("../R/Get_bootcamp_googlesheet.R")
```

This script downloads the data file saves it to a CSV under `data/survey.csv`.We can then load this file.

I also created a test data file, `data/survey-test.csv` so I could see how everything worked before y'all filled out your responses. The `R/Make_test_survey.R` file shows how I did this. It's a great, reproducible practice to **simulate the data you expect**, then run it through your pipeline.

```r
# Created test data set for testing.
# survey <- read_csv("../data/survey-test.csv")
# Or choose data from respondents
survey <- read_csv("../data/survey.csv")
```

```
## Parsed with column specification:
## cols(
##   Timestamp = col_character(),
##   `Your current level of experience/expertise with R` = col_character(),
##   `Your enthusiasm for Game of Thrones` = col_integer(),
##   `Age in years` = col_integer(),
##   `Preferred number of hours spent sleeping/day` = col_character(),
##   `Favorite day of the week` = col_character(),
##   `Are your data tidy?` = col_character()
## )
```

```r
survey
```

```
## # A tibble: 39 x 7
##             Timestamp `Your current level of experience/expertise with R`
##                 <chr>                                               <chr>
##  1               <NA>                                                <NA>
##  2 8/13/2017 23:29:24                                                some
##  3 8/14/2017 12:01:12                                                some
##  4 8/15/2017 12:42:09                                                some
##  5 8/15/2017 17:13:08                                                none
##  6 8/15/2017 19:03:40                                             limited
##  7 8/15/2017 23:36:07                                                some
##  8 8/15/2017 23:45:05                                             limited
##  9  8/16/2017 0:26:01                                                 pro
## 10  8/16/2017 1:09:44                                                none
## # ... with 29 more rows, and 5 more variables: `Your enthusiasm for Game
## #   of Thrones` <int>, `Age in years` <int>, `Preferred number of hours
## #   spent sleeping/day` <chr>, `Favorite day of the week` <chr>, `Are your
## #   data tidy?` <chr>
```

The str() or ‘structure’ command is also a great way to see what you’ve got.

```r
str(survey)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame':    39 obs. of  7 variables:
##  $ Timestamp                                         : chr  NA "8/13/2017 23:29:24" "8/14/2017 12:01:
##  $ Your current level of experience/expertise with R: chr  NA "some" "some" "some" ...
##  $ Your enthusiasm for Game of Thrones               : int  NA 10 10 10 10 10 10 3 9 10 ...
##  $ Age in years                                      : int  NA 28 22 24 28 24 23 25 37 25 ...
##  $ Preferred number of hours spent sleeping/day      : chr  NA "8!!!" "7" "10" ...
##  $ Favorite day of the week                          : chr  NA "Friday" "Friday" "Saturday" ...
##  $ Are your data tidy?                               : chr  NA "Yes" "That's a personal question" "No
##  - attr(*, "spec")=List of 2
##   ..$ cols   :List of 7
##   .. ..$ Timestamp                                         : list()
##   .. .. ..- attr(*, "class")= chr  "collector_character" "collector"
##   .. ..$ Your current level of experience/expertise with R: list()
##   .. .. ..- attr(*, "class")= chr  "collector_character" "collector"
##   .. ..$ Your enthusiasm for Game of Thrones               : list()
##   .. .. ..- attr(*, "class")= chr  "collector_integer" "collector"
##   .. ..$ Age in years                                      : list()
##   .. .. ..- attr(*, "class")= chr  "collector_integer" "collector"
##   .. ..$ Preferred number of hours spent sleeping/day     : list()
##   .. .. ..- attr(*, "class")= chr  "collector_character" "collector"
##   .. ..$ Favorite day of the week                         : list()
##   .. .. ..- attr(*, "class")= chr  "collector_character" "collector"
##   .. ..$ Are your data tidy?                              : list()
##   .. .. ..- attr(*, "class")= chr  "collector_character" "collector"
##   ..$ default: list()
##   .. ..- attr(*, "class")= chr  "collector_guess" "collector"
##   ..- attr(*, "class")= chr "col_spec"
```

Clearly, we need to do some cleaning before we can do anything with this.

Let's start by renaming variables.

```
names(survey) <- c("Timestamp",
                  "R_exp",
                  "GoT",
                  "Age_yrs",
                  "Sleep_hrs",
                  "Fav_day",
                  "Tidy_data")
```

```
# complete.cases() drops NAs
survey <- survey[complete.cases(survey),]
survey
```

```
## # A tibble: 38 x 7
##              Timestamp   R_exp   GoT Age_yrs Sleep_hrs   Fav_day
##                  <chr>   <chr> <int>   <int>     <chr>     <chr>
##   1 8/13/2017 23:29:24    some    10      28      8!!!    Friday
##   2 8/14/2017 12:01:12    some    10      22         7    Friday
##   3 8/15/2017 12:42:09    some    10      24        10  Saturday
##   4 8/15/2017 17:13:08    none    10      28         9  Saturday
##   5 8/15/2017 19:03:40 limited    10      24         9  Saturday
##   6 8/15/2017 23:36:07    some    10      23       6-7    Friday
##   7 8/15/2017 23:45:05 limited     3      25         8    Friday
##   8  8/16/2017 0:26:01     pro     9      37         7    Friday
##   9  8/16/2017 1:09:44    none    10      25         9  Saturday
##  10  8/16/2017 8:51:05 limited     1      23       7.5  Thursday
## # ... with 28 more rows, and 1 more variables: Tidy_data <chr>
```

Now, lets make sure we have numbers where we expect them. That person who really likes 8 hours ("8!!!") is a problem (for me, not them).

```
survey$Sleep_hrs <- readr::parse_number(survey$Sleep_hrs)
survey
```

```
## # A tibble: 38 x 7
##              Timestamp   R_exp   GoT Age_yrs Sleep_hrs   Fav_day
##                  <chr>   <chr> <int>   <int>     <dbl>     <chr>
##   1 8/13/2017 23:29:24    some    10      28       8.0    Friday
##   2 8/14/2017 12:01:12    some    10      22       7.0    Friday
##   3 8/15/2017 12:42:09    some    10      24      10.0  Saturday
##   4 8/15/2017 17:13:08    none    10      28       9.0  Saturday
##   5 8/15/2017 19:03:40 limited    10      24       9.0  Saturday
##   6 8/15/2017 23:36:07    some    10      23       6.0    Friday
##   7 8/15/2017 23:45:05 limited     3      25       8.0    Friday
## 8  8/16/2017 0:26:01     pro     9      37        7.0   Friday
## 9  8/16/2017 1:09:44    none    10      25        9.0 Saturday
## 10 8/16/2017 8:51:05 limited     1      23        7.5 Thursday
## # ... with 28 more rows, and 1 more variables: Tidy_data <chr>
```

Looks good. Let's save that cleaned file so we don't have to do this again.

```
write_csv(survey, path="../data/survey_clean.csv")
```

We may want to make the `R_exp` variable ordered.

```
(survey_responses <- unique(survey$R_exp))
```

```
## [1] "some"    "none"    "limited" "pro"
```

This shows us the different survey response values.

```
survey$R_exp <- ordered(survey$R_exp, levels=c("none",
                                               "limited",
                                               "some",
                                               "lots",
                                               "pro"))
```

Visualization

Now, we follow Mike Meyer's advice: "Plot your data!"

Descriptive plots

```
R_exp_hist <- survey %>%
  ggplot() +
  aes(x=R_exp) +
  geom_histogram(stat = "count") # R_exp is discrete
```

```
## Warning: Ignoring unknown parameters: binwidth, bins, pad
```

```
R_exp_hist
```

```
Sleep_hrs_hist <- survey %>%
  ggplot() +
  aes(x=Sleep_hrs) +
  geom_histogram() # Sleep_hrs is continuous
Sleep_hrs_hist
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

```
Got_hist <- survey %>%
  ggplot() +
  aes(x=GoT) +
  geom_histogram()
Got_hist
```

```
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```

Looks like we are of two minds about GoT.

Does R experience have any relation to GoT enthusiasm?

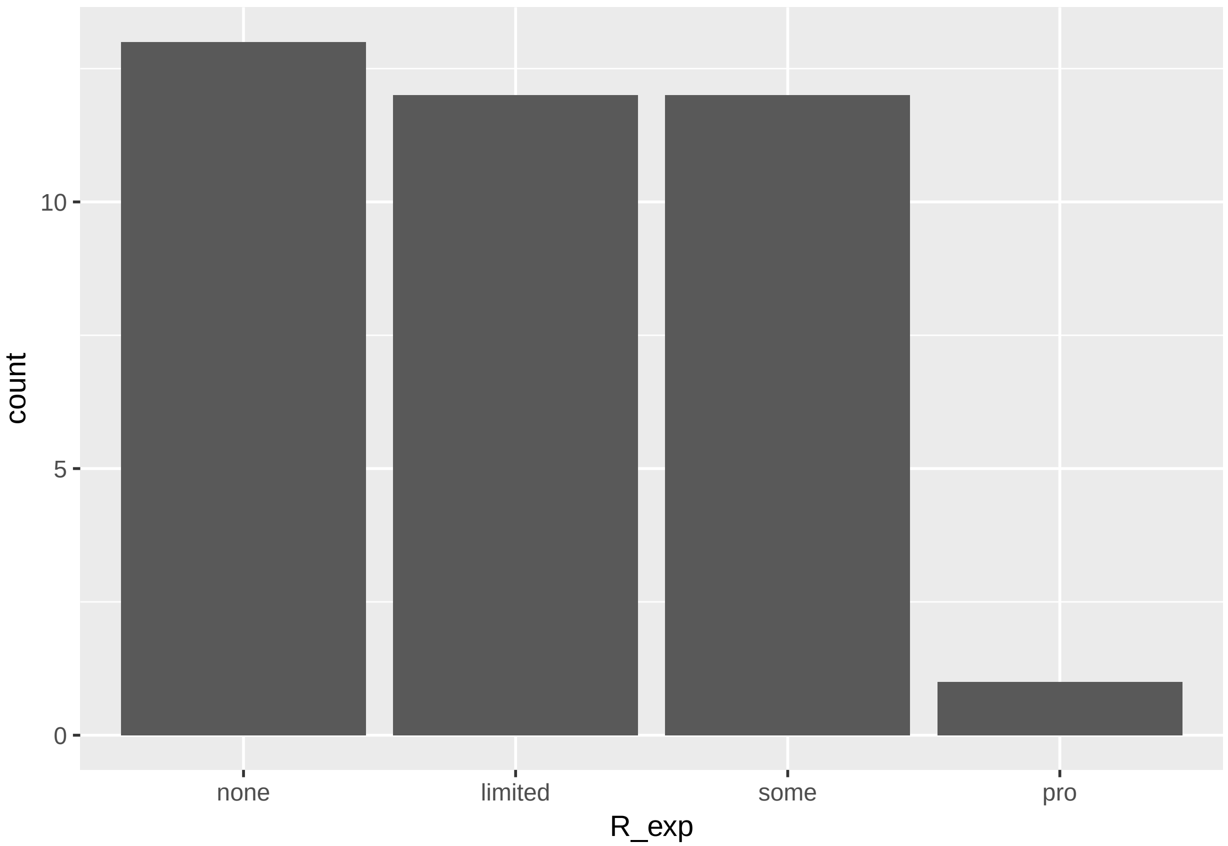


Figure 1: Distribution of prior R experience

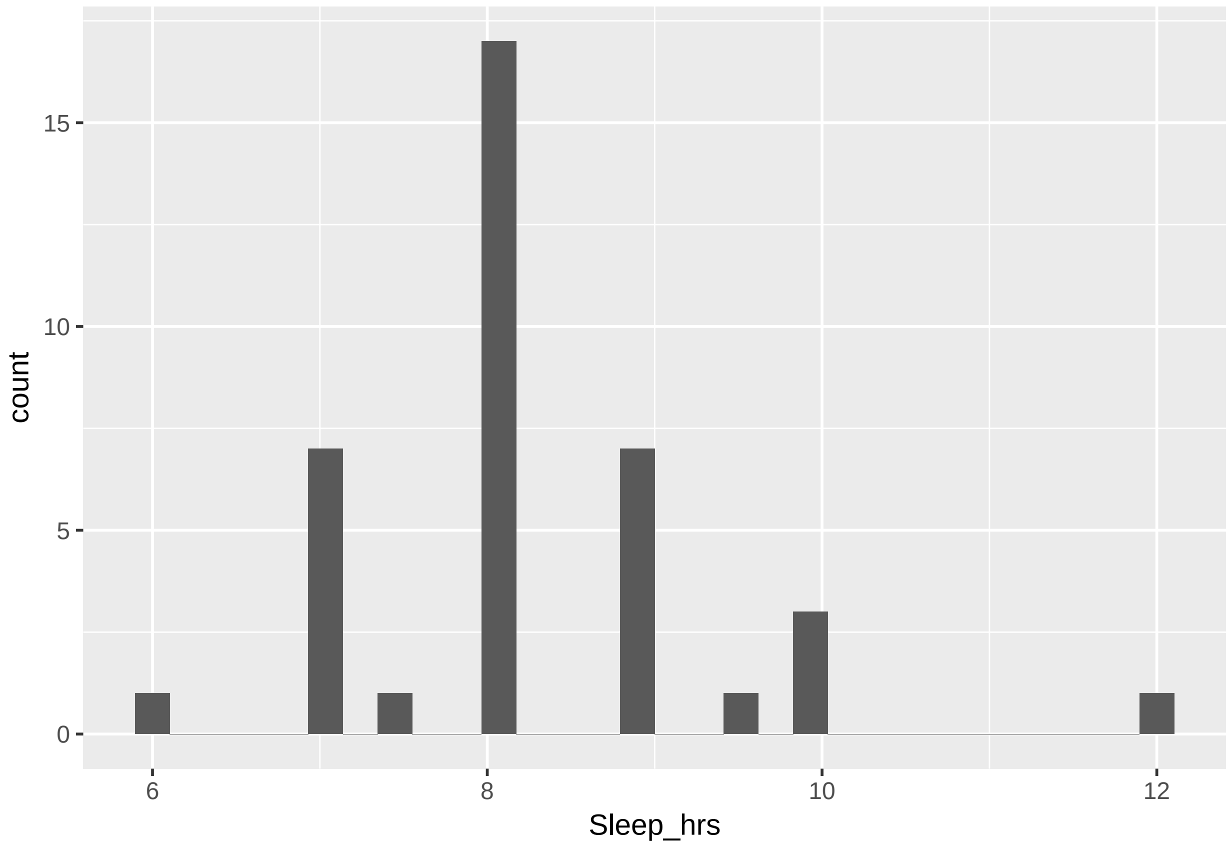


Figure 2: Distribution of preferred sleep hrs/day

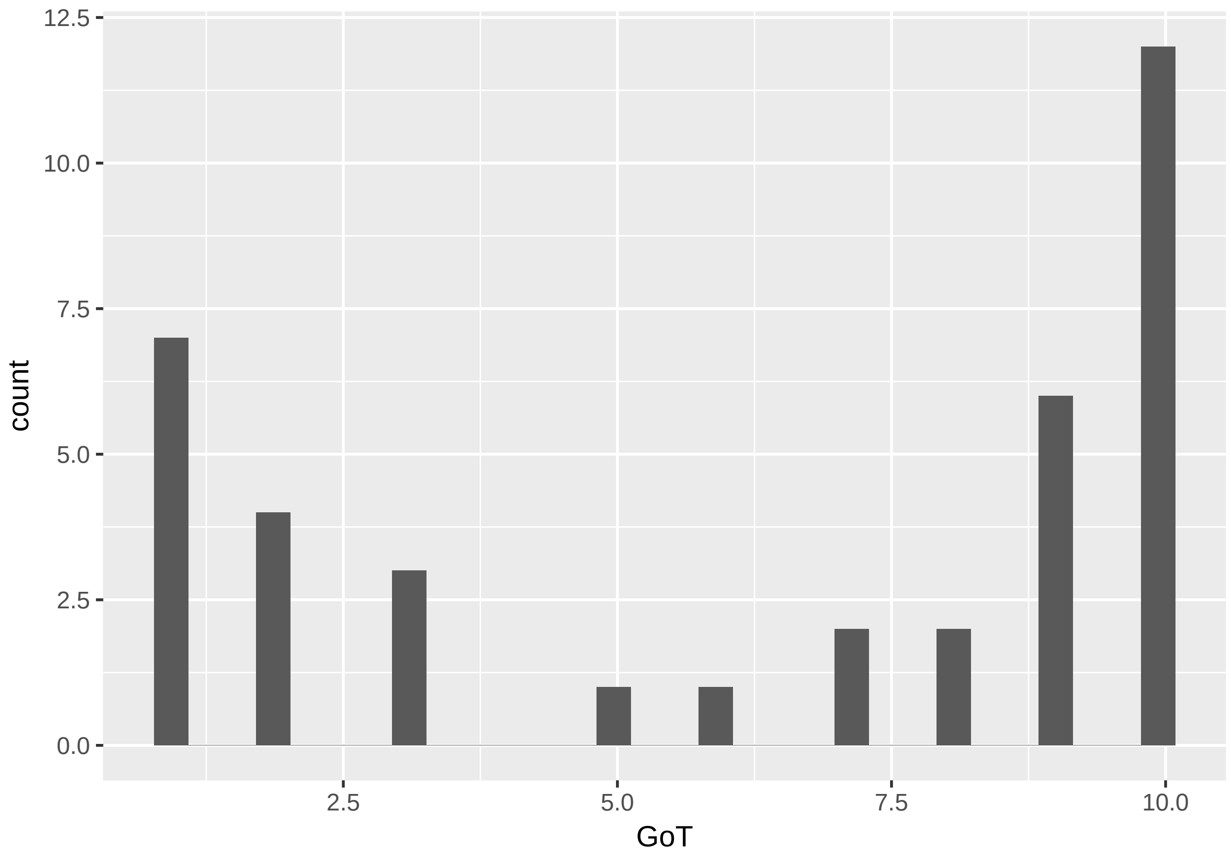
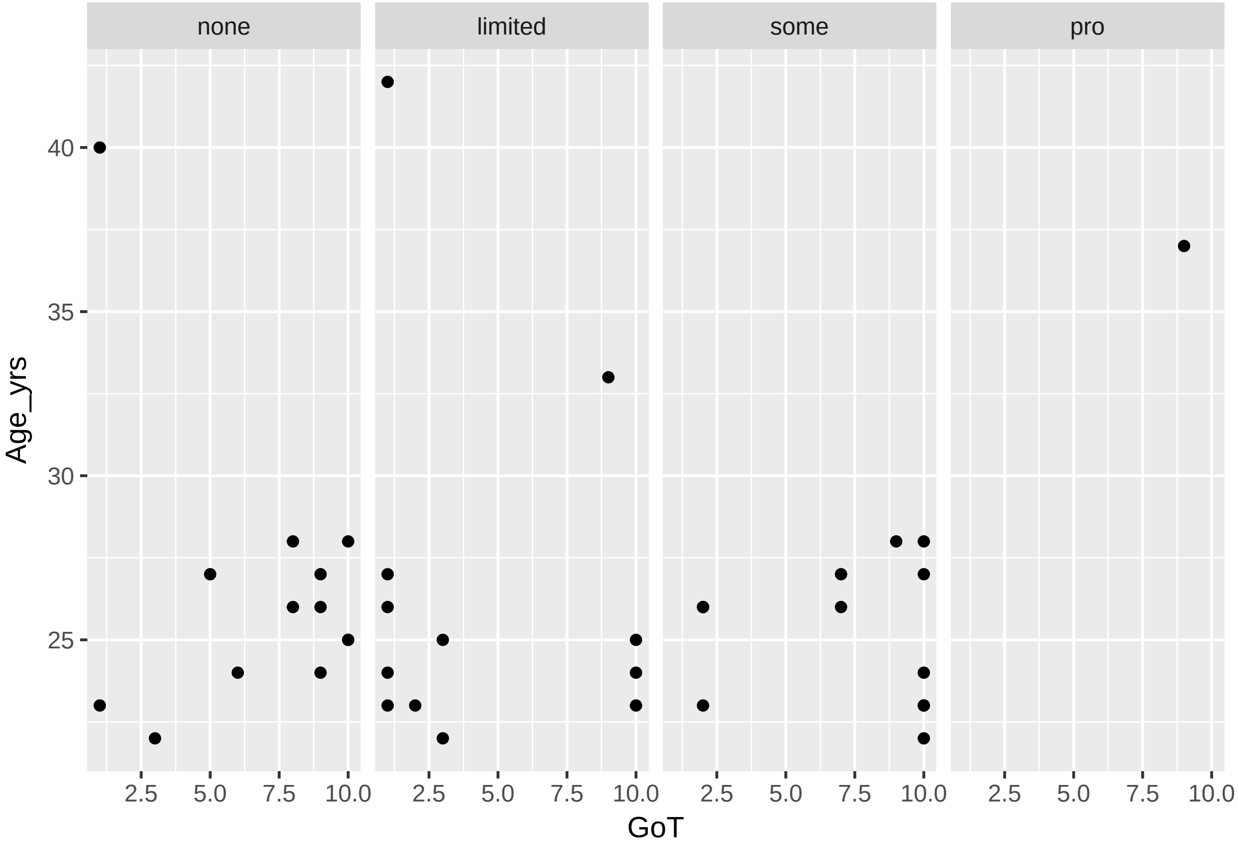


Figure 3: Distribution of GoT Enthusiasm

Figure 4:

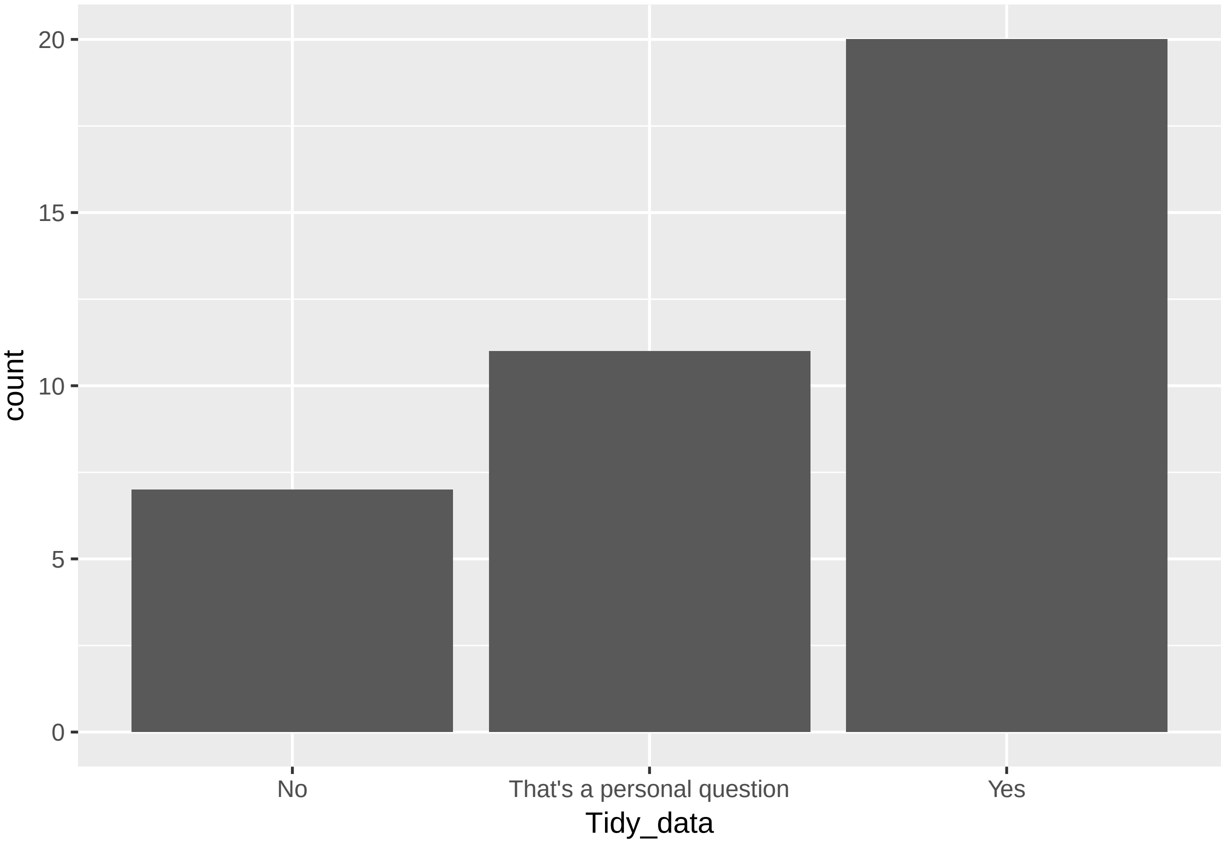
```
GoT_vs_r_exp <- survey %>%
  ggplot() +
  aes(x=GoT, y=Age_yrs) +
  facet_grid(. ~ R_exp) +
  geom_point()
  # + stat_smooth()
GoT_vs_r_exp
```



```
tidy_hist <- survey %>%
  ggplot() +
  aes(x=Tidy_data) +
  geom_histogram(stat = "count")
```

```
## Warning: Ignoring unknown parameters: binwidth, bins, pad
```

```
tidy_hist
```



Analysis

I could use a document like this to plan out my analysis plan **before** I conduct it. If I used simulated data, I could make sure that my workflow will run when I get real (cleaned) data. I could even preregister my analysis plan before I conduct it. That doesn't preclude later exploratory analyses, but it does hold me and my collaborators accountable for what I predicted in advance.

Notes

Notice that I sometimes put a label like `got-vs-r-exp` in the brackets for a given 'chunk' of R code. The main reasons to do this are:

- It sometimes makes it easier to debug your code.
- In some cases, you can have this 'chunk' name serve as the file name for a figure you generate within a chunk.
- In a bit, we'll see how these chunk names are useful for making tables, figures, and equations that generate their own numbers.