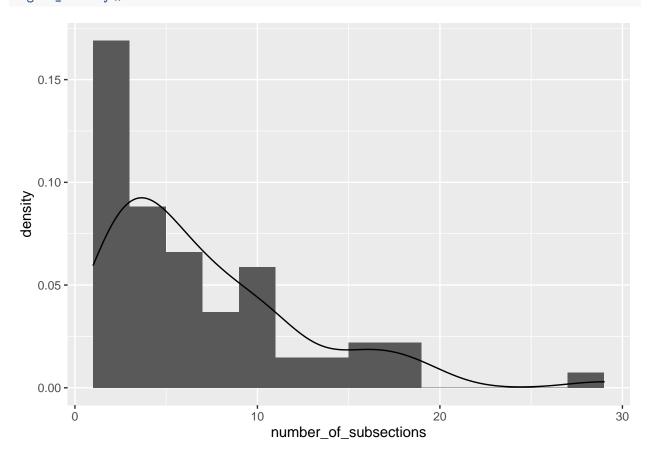
Exploring Similarities and Differences in Ecosubsections

```
# Import data
spatial <- read_csv("../data/plot_level/plt_spatial.csv")</pre>
response <- read_csv("../data/plot_level/plot_response.csv")</pre>
# Join data
## Keep only observations in both `spatial` and `response`
dat <- inner_join(spatial, response,</pre>
                  by = c("PLT_CN" = "PLT_CN",
                         "INVYR" = "INVYR"))
# Create columns for province, sections, and subsections
dat <- dat %>%
  mutate(
    subsection = ECOSUBCD.x,
    section = str_remove_all(ECOSUBCD.x, "[:lower:]"),
    province = str_sub(section, end = -2)
  )
# Select small subset of columns to work with for this EDA
dat_small <- dat %>%
  select(PLT CN, INVYR, PLOT.x, LON PUBLIC.x, LAT PUBLIC.x, LON PUBLIC.y, LAT PUBLIC.y,
         ELEV_PUBLIC.x, ELEV_PUBLIC.y, forgrp, forprob, nlcd11, demLF, evtLF, forbio,
         BALIVE_TPA, CNTLIVE_TPA, BIOLIVE_TPA, VOLNLIVE_TPA, subsection, section, province)
# Remove redundent columns, rename columns for ease of use
dat small <- dat small %>%
  select(-LON_PUBLIC.y, -LAT_PUBLIC.y, -ELEV_PUBLIC.y) %>%
  rename(PLOT = PLOT.x,
         LON_PUBLIC = LON_PUBLIC.x,
         LAT_PUBLIC = LAT_PUBLIC.x,
         ELEV_PUBLIC = ELEV_PUBLIC.x)
n subsections <- dat small %>%
  group_by(section, subsection) %>%
  summarize(n()) %>%
  group_by(section) %>%
  summarize(number_of_subsections = n())
## `summarise()` regrouping output by 'section' (override with `.groups` argument)
## `summarise()` ungrouping output (override with `.groups` argument)
head(n_subsections)
## # A tibble: 6 x 2
   section number_of_subsections
     <chr>
                              <int>
## 1 313A
                                 19
## 2 313B
                                  7
## 3 313C
                                  4
## 4 313D
                                  5
## 5 315A
                                  3
                                  4
## 6 315B
```

```
## Does this make sense?
sum(n_subsections$number_of_subsections)
## [1] 480
length(unique(dat_small$subsection))
## [1] 480
## Yes!
## Let's look at this distribution:
ggplot(n_subsections, aes(x = number_of_subsections)) +
  geom_boxplot() +
  theme_bw()
  0.4
  0.2
  0.0
 -0.2
 -0.4
                                 10
                                                            20
                                    number_of_subsections
mean(n_subsections$number_of_subsections)
## [1] 7.058824
sd(n_subsections$number_of_subsections)
## [1] 5.5068
median(n_subsections$number_of_subsections)
## [1] 5
ggplot(n_subsections, aes(x = number_of_subsections)) +
  geom_histogram(bins = 15, aes(y = ..density..)) +
```

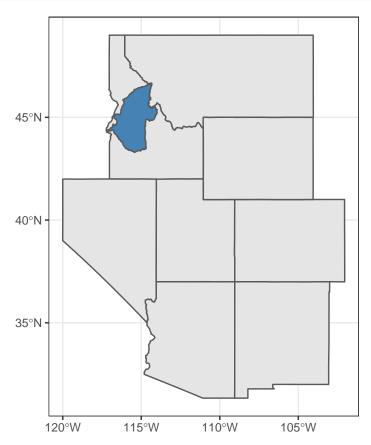
geom_density()



From this, we see that the average number of subsections in a section is about 7, with a median of 5 giving us a right-skewed distribution. There is one outlying section which I will investigate now:

```
# M332A: Idaho Batholith, "The batholith section is a large, contiguous uplifted area of granitic pluto
library(concaveman)
library(sf)
```

Linking to GEOS 3.7.2, GDAL 2.4.2, PROJ 5.2.0



0.02157692 [1]

1 / length(unique(dat_small\$section)) # This is the proportion of total area an "average" section would

```
## [1] 0.01470588
```

M332A takes up more area than average but not *way* more. This means that it likely has some small su ### this doesn't seem right based on the picture. revisit this with fresh eyes tomorrow.

Quantifying Homogeneity in Ecosubsections