# plotting numeric data using scales, labels, jitter

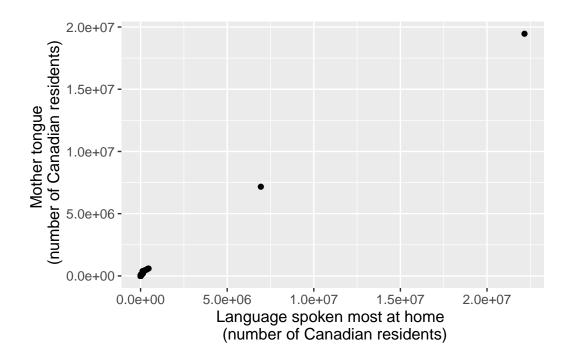
Emily Malcolm-White

```
library(tidyverse)

#Import the can_lang dataset
can_lang <- read.csv("https://raw.githubusercontent.com/ttimbers/canlang/master/inst/extda</pre>
```

## scatterplot of can\_lang

```
ggplot(can_lang, aes(x=most_at_home, y=mother_tongue)) +
   geom_point() +
   xlab("Language spoken most at home \n (number of Canadian residents)") +
   ylab("Mother tongue \n (number of Canadian residents)") +
   theme(text = element_text(size = 12))
```

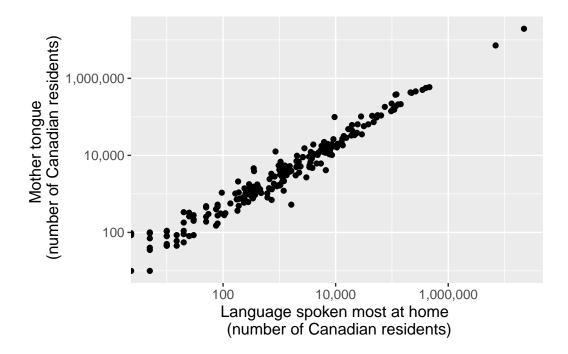


Notice anything weird about this plot?

# Axis transformations using the scales package

```
library(scales)

ggplot(can_lang, aes(x = most_at_home, y = mother_tongue)) +
    geom_point() +
    xlab("Language spoken most at home \n (number of Canadian residents)") +
    ylab("Mother tongue \n (number of Canadian residents)") +
    theme(text = element_text(size = 12)) +
    scale_x_log10(labels = label_comma()) +
    scale_y_log10(labels = label_comma())
```

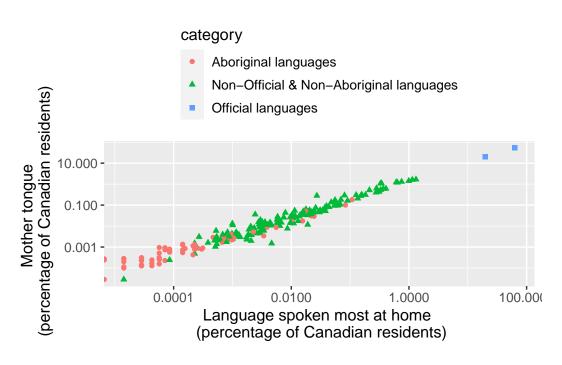


#### mutate to create new columns

```
can_lang <- can_lang %>%
  mutate(
    mother_tongue_percent = (mother_tongue / 35151728) * 100,
    most_at_home_percent = (most_at_home / 35151728) * 100
)
```

#### **Scatterplot with Percents and Colors**

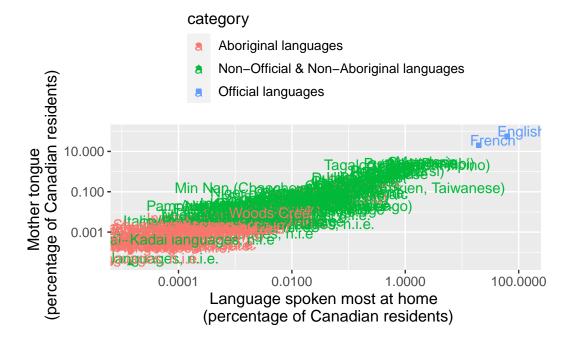
```
scale_y_log10(labels = comma)
```



## **Labels**

Adding text to a plot is one of the most common forms of annotation. Most plots will not benefit from adding text to every single observation on the plot, but labeling outliers and other important points is very useful.

#### $nudge_y=0.25$ )



Yikes! This is way too much going on in one plot. A few options to try when this happens:

- Decrease the font size of the labels (using the size= argument inside geom\_text).
- Use the ggrepel package to spread out the labels a bit more (more on this later
- Pick out only a subset of the points to label

#### Size of labels

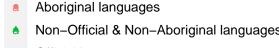
```
nudge_x = 0.25,
               nudge_y=0.25, size=3)
                            category
                                 Aboriginal languages
                                 Non-Official & Non-Aboriginal languages
   (percentage of Canadian residents)
                                 Official languages
                                                                              English
Mother tongue
      10.000 -
                                                                jen, Taiwanese)
       0.100 -
       0.001 -
               rdaageages.en.i.e.
                       0.0001
                                           0.0100
                                                                                  100.0000
                                                               1.0000
                                Language spoken most at home
                               (percentage of Canadian residents)
```

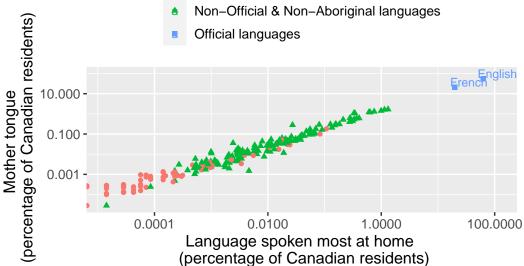
#### Subset the labels

Create a new column for the labels. Use case\_when (or ifelse) to only use the official language names and not to put a label for other language categories.

```
scale_y_log10(labels = comma) +
geom_text(aes(label=official_languages),
            nudge_x = 0.25,
            nudge_y=0.25, size = 3)
```

#### category





## Using ggrepel

```
library(ggrepel)
ggplot(can_lang, aes(x = most_at_home_percent,
                     y = mother_tongue_percent,
                     color = category, shape=category)) +
 geom_point() +
 xlab("Language spoken most at home \n (percentage of Canadian residents)") +
 ylab("Mother tongue \n (percentage of Canadian residents)") +
 theme(text = element_text(size = 12),
        legend.position = "top",
        legend.direction = "vertical") +
 scale_x_log10(labels = comma) +
 scale_y_log10(labels = comma) +
  geom_text_repel(aes(label=official_languages), min.segment.length=0, box.padding=1)
```

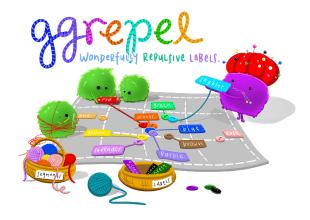
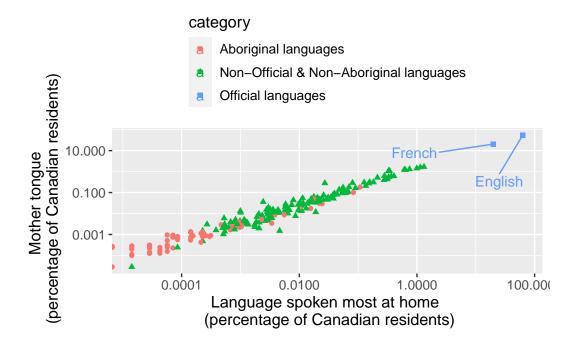


Figure 1: Artwork by @allisonhorst



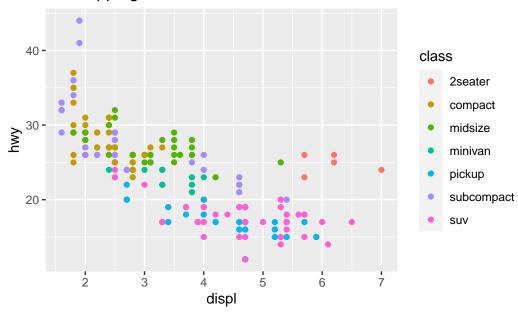
## (Optional) Jitter

This dataset contains a subset of the fuel economy data that the EPA makes available on https://fueleconomy.gov/. It contains only models which had a new release every year between 1999 and 2008 - this was used as a proxy for the popularity of the car.

data("mpg")

```
ggplot(data = mpg, aes(x = displ, y = hwy)) +
geom_point(aes(color = class)) +
ggtitle("Overlapping Points")
```

## **Overlapping Points**

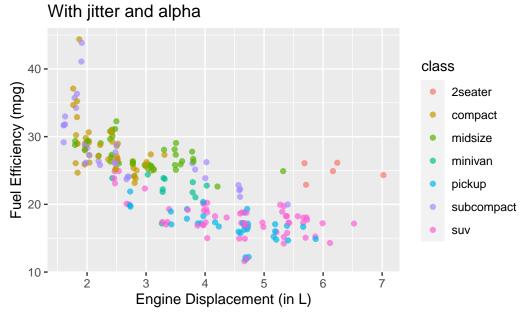


Note how there is only a fixed number of values for the x-axis and a fixed number of values for the y-axis. It's possible that some points might be overlapping others!

**Jittering** is a technique for adding random noise to data points that have identical values in a plot. It is a useful method for avoiding overplotting and making it easier to visualize the density of data points that are otherwise hidden behind each other.

We add position="jitter" inside the geom\_point layer:

```
ggplot(data = mpg, aes(x = displ, y = hwy)) +
  geom_point(aes(color = class), position = "jitter", alpha=0.7) +
  ggtitle("With jitter and alpha") +
  xlab("Engine Displacement (in L)") +
  ylab("Fuel Efficiency (mpg)") +
  labs(caption = "Source: Fuel Economy Data 1999 - 2008 www.fueleconomy.gov")
```



Source: Fuel Economy Data 1999 - 2008 www.fueleconomy.gov

# **Brain Break**

Hans Rosling's famous lectures combine enormous quantities of public data with a sport's commentator's style to reveal the story of the world's past, present and future development.