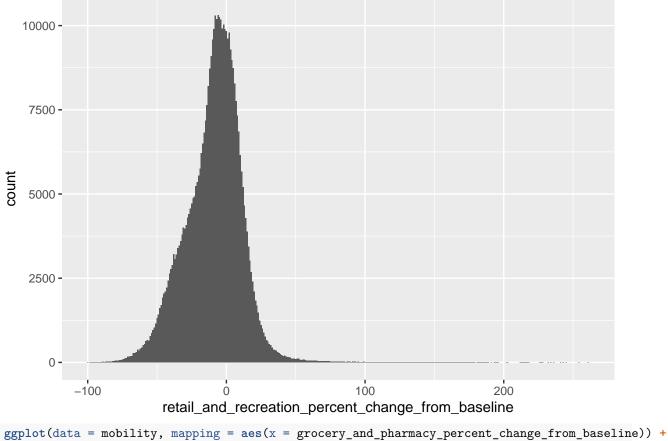
Google Movement Data Analysis

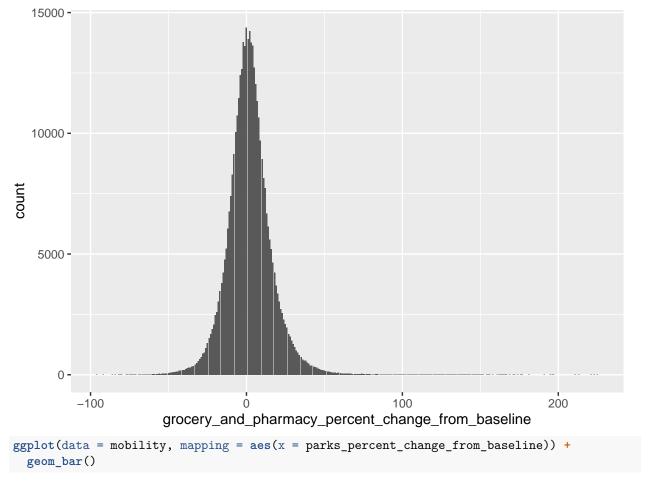
Albert Sun, Lily Zhu

```
library(tidyverse)
## -- Attaching packages ------ tidyverse 1.3.0 -
## v tibble 3.0.3
                     v purrr
                               0.3.4
## v tidyr
            1.1.1
                      v dplyr
                               1.0.1
## v readr
            1.3.1
                     v forcats 0.5.0
## -- Conflicts ----- tidyverse_conflicts() -
## x lubridate::as.difftime() masks base::as.difftime()
## x lubridate::date()
                      masks base::date()
## x dplyr::filter() masks stats::filter()
## x readr::guess_encoding() masks rvest::guess_encoding()
## x lubridate::intersect() masks base::intersect()
## x dplyr::lag()
                          masks stats::lag()
                   masks rvest::pluck()
iff() masks base::setdiff()
## x purrr::pluck()
## x lubridate::setdiff()
## x lubridate::union()
                            masks base::union()
#library(hrbrthemes)
mobility <- read.csv(".../data/US-Mobility-Report.csv")</pre>
mobility <- mobility %>%
  mutate(date_num = as.Date(date))
ggplot(data = mobility, mapping = aes(x = retail_and_recreation_percent_change_from_baseline)) +
  geom_bar()
## Warning: Removed 210137 rows containing non-finite values (stat_count).
```

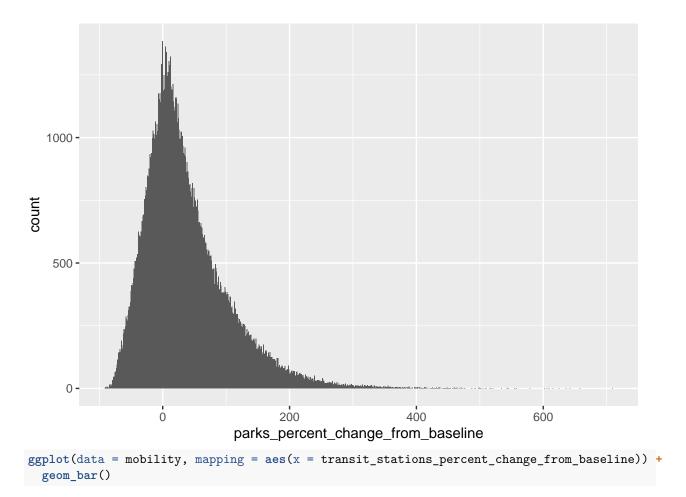


ggplot(data = mobility, mapping = aes(x = grocery_and_pharmacy_percent_change_from_baseline)) +
geom_bar()

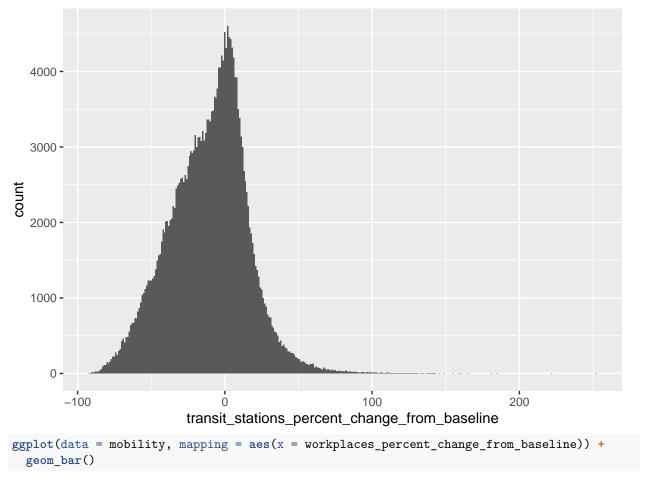
Warning: Removed 247746 rows containing non-finite values (stat_count).



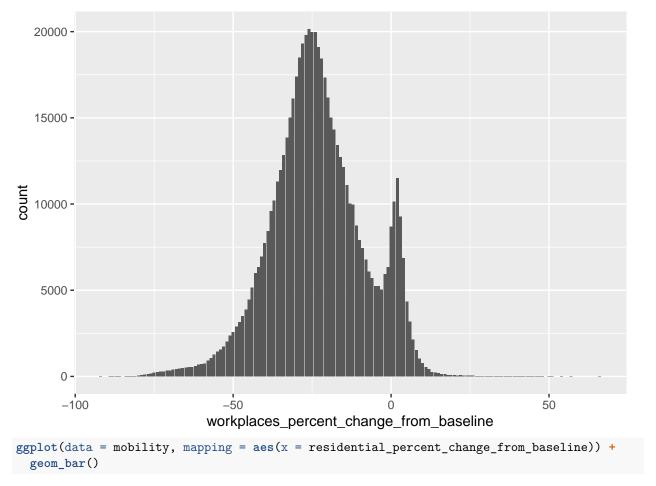
Warning: Removed 494324 rows containing non-finite values (stat_count).



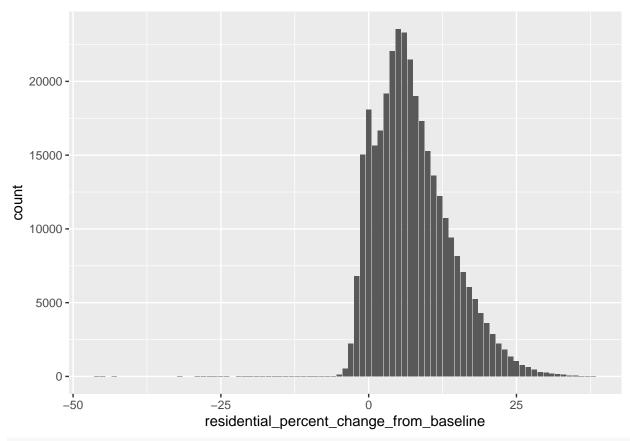
Warning: Removed 399459 rows containing non-finite values (stat_count).



Warning: Removed 22327 rows containing non-finite values (stat_count).



Warning: Removed 312892 rows containing non-finite values (stat_count).

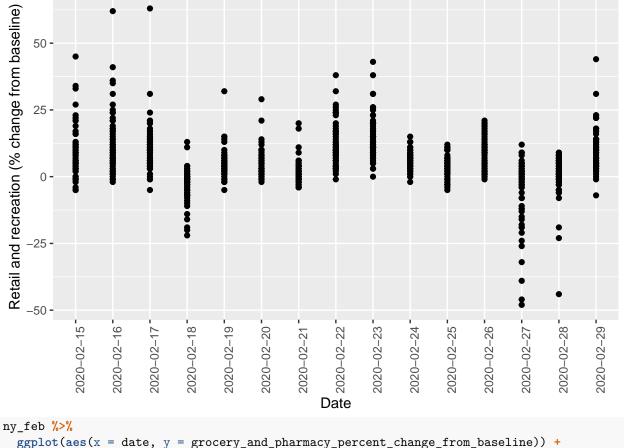


sd iqr min max

##

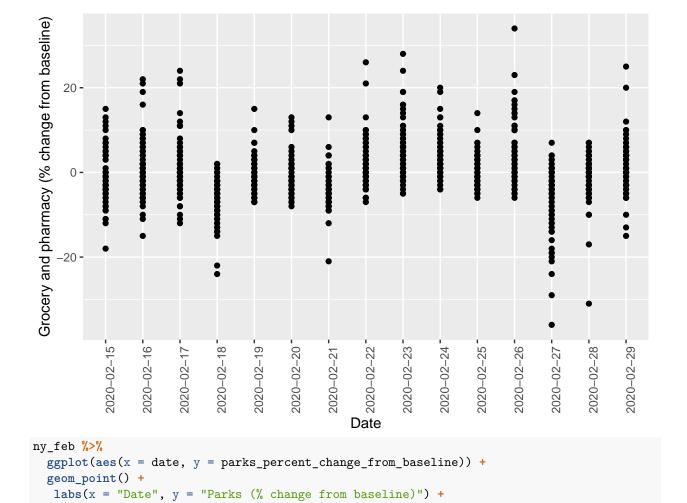
mean median

```
sd = sd(parks_percent_change_from_baseline),
          iqr = IQR(parks_percent_change_from_baseline),
          min = min(parks_percent_change_from_baseline),
          max = max(parks_percent_change_from_baseline))
         mean median
                           sd iqr min max
## 1 37.33111
                  23 66.49339 76 -91 709
mobility %>%
  filter(!is.na(transit_stations_percent_change_from_baseline)) %>%
  summarise(mean = mean(transit_stations_percent_change_from_baseline),
          median = median(transit stations percent change from baseline),
          sd = sd(transit_stations_percent_change_from_baseline),
          iqr = IQR(transit_stations_percent_change_from_baseline),
          min = min(transit_stations_percent_change_from_baseline),
          max = max(transit_stations_percent_change_from_baseline))
          mean median
                            sd iqr min max
## 1 -10.91589
                   -9 26.18124 35 -92 252
mobility %>%
  filter(!is.na(workplaces_percent_change_from_baseline)) %>%
  summarise(mean = mean(workplaces_percent_change_from_baseline),
          median = median(workplaces_percent_change_from_baseline),
          sd = sd(workplaces_percent_change_from_baseline),
          iqr = IQR(workplaces_percent_change_from_baseline),
          min = min(workplaces_percent_change_from_baseline),
          max = max(workplaces_percent_change_from_baseline))
                           sd iqr min max
         mean median
## 1 -23.0098
                 -24 14.96862 18 -92 66
mobility %>%
  filter(!is.na(residential_percent_change_from_baseline)) %>%
  summarise(mean = mean(residential_percent_change_from_baseline),
          median = median(residential_percent_change_from_baseline),
          sd = sd(residential_percent_change_from_baseline),
          iqr = IQR(residential_percent_change_from_baseline),
          min = min(residential_percent_change_from_baseline),
          max = max(residential_percent_change_from_baseline))
                           sd iqr min max
         mean median
## 1 7.505763
                   7 6.300294
                              8 -46 38
ny_feb <- mobility %>%
 filter(sub_region_1 == "New York", date_num >= as.Date("2020-02-15"), date_num < as.Date("2020-03-01"
ny_feb %>%
  ggplot(aes(x = date, y = retail_and_recreation_percent_change_from_baseline)) +
  geom point() +
 labs(x = "Date", y = "Retail and recreation (% change from baseline)") +
  theme(axis.text.x = element_text(angle = 90))
```



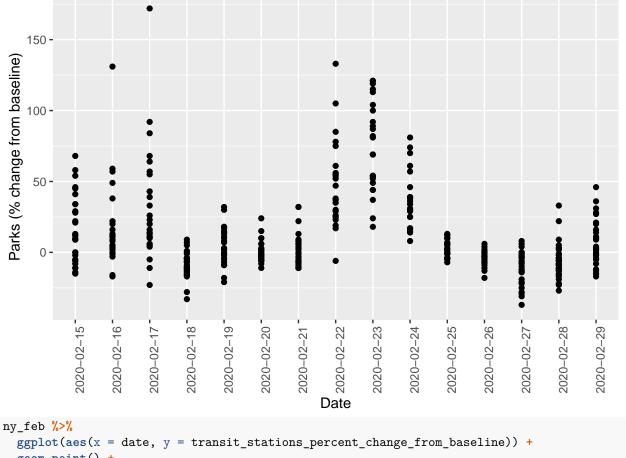
```
ggplot(aes(x = date, y = grocery_and_pharmacy_percent_change_from_baseline)) +
geom_point() +
labs(x = "Date", y = "Grocery and pharmacy (% change from baseline)") +
theme(axis.text.x = element_text(angle = 90))
```

Warning: Removed 1 rows containing missing values (geom_point).



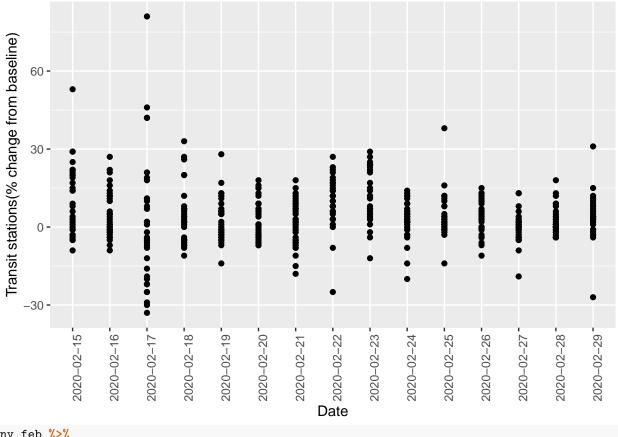
Warning: Removed 530 rows containing missing values (geom_point).

theme(axis.text.x = element_text(angle = 90))

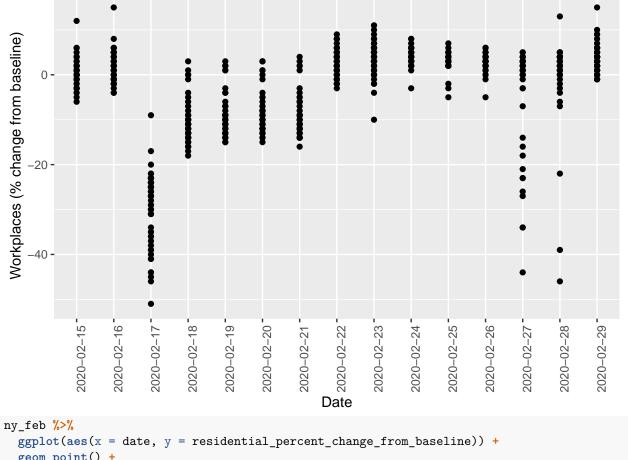


```
ggplot(aes(x = date, y = transit_stations_percent_change_from_baseline)) +
geom_point() +
labs(x = "Date", y = "Transit stations(% change from baseline)") +
theme(axis.text.x = element_text(angle = 90))
```

Warning: Removed 435 rows containing missing values (geom_point).

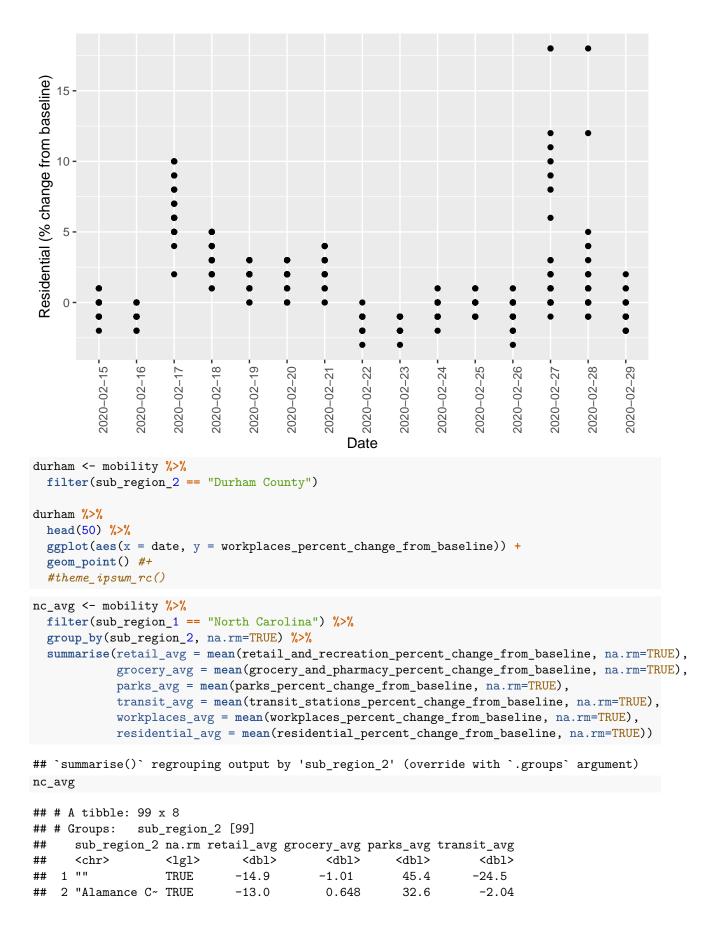


```
ny_feb %>%
  ggplot(aes(x = date, y = workplaces_percent_change_from_baseline)) +
  geom_point() +
  labs(x = "Date", y = "Workplaces (% change from baseline)") +
  theme(axis.text.x = element_text(angle = 90))
```



```
ggplot(aes(x = date, y = residential_percent_change_from_baseline)) +
geom_point() +
labs(x = "Date", y = "Residential (% change from baseline)") +
theme(axis.text.x = element_text(angle = 90))
```

Warning: Removed 159 rows containing missing values (geom_point).



```
## 3 "Alexander ~ TRUE
                                          11.7
                              -7.44
                                                     {\tt NaN}
                                                                NaN
                                          -9.75
                                                                NaN
## 4 "Alleghany ~ TRUE
                             -12.5
                                                     NaN
## 5 "Anson Coun~ TRUE
                              -7.99
                                          10.5
                                                                NaN
                                                     {\tt NaN}
## 6 "Ashe Count~ TRUE
                               4.41
                                          8.2
                                                     {\tt NaN}
                                                                NaN
## 7 "Avery Coun~ TRUE
                             -12.6
                                          13.1
                                                     NaN
                                                                NaN
## 8 "Beaufort C~ TRUE
                              -8.22
                                          3.65
                                                      17.8
                                                                NaN
## 9 "Bertie Cou~ TRUE
                              18.8
                                          7.02
                                                     {\tt NaN}
                                                                NaN
## 10 "Bladen Cou~ TRUE
                                           5.86
                                                     -24.7
                              -3.61
                                                                NaN
## # ... with 89 more rows, and 2 more variables: workplaces_avg <dbl>,
## # residential_avg <dbl>
#test_plot <- ggplot(mtcars, aes(mpg, wt)) +</pre>
  #geom_point() +
  #labs(x="Fuel efficiency (mpg)", y="Weight (tons)",
       #title="Seminal ggplot2 scatterplot example",
       #subtitle="A plot that is only useful for demonstration purposes",
       #caption="Brought to you by the letter 'g'") +
  #theme_ipsum_rc()
\#ggsave("../output/test-plot.png", test_plot, units = "in", width = 850/100, height = 600/100)
```

EDA

Linear Regression Code