Shaoqian_Chen_599Final

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```
library(dplyr)       # for data manipulation
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##     filter, lag
```

```
## The following objects are masked from 'package:base':
##
##     intersect, setdiff, setequal, union
```

```
library(ggplot2)     # for data visualization
library(stringr)     # for string functionality
library(cluster)     # for general clustering algorithms
library(factoextra)  # for visualizing cluster results
```

```
## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WB
a
```

```
library(yardstick)
```

```
## For binary classification, the first factor level is assumed to be the event.
## Set the global option `yardstick.event_first` to `FALSE` to change this.
```

```
library(lubridate)
```

```
##
## Attaching package: 'lubridate'
```

```
## The following objects are masked from 'package:dplyr':
##
##     intersect, setdiff, union
```

```
## The following objects are masked from 'package:base':
##
##     date, intersect, setdiff, union
```

```
library(ggplot2)
library(ggmap)
```

```
## Google's Terms of Service: https://cloud.google.com/maps-platform/terms/.
```

```
## Please cite ggmap if you use it! See citation("ggmap") for details.
```

```
library(dplyr)
library(readr)
```

```
##
## Attaching package: 'readr'
```

```
## The following object is masked from 'package:yardstick':
##
##     spec
```

```
ggmap::register_google(key = "AIzaSyDjdKw8npxWHoH5eL85_OrfxTUSyf8Wxcg")
```

Question 1

a)

```
mnist <- dslabs::read_mnist()
```

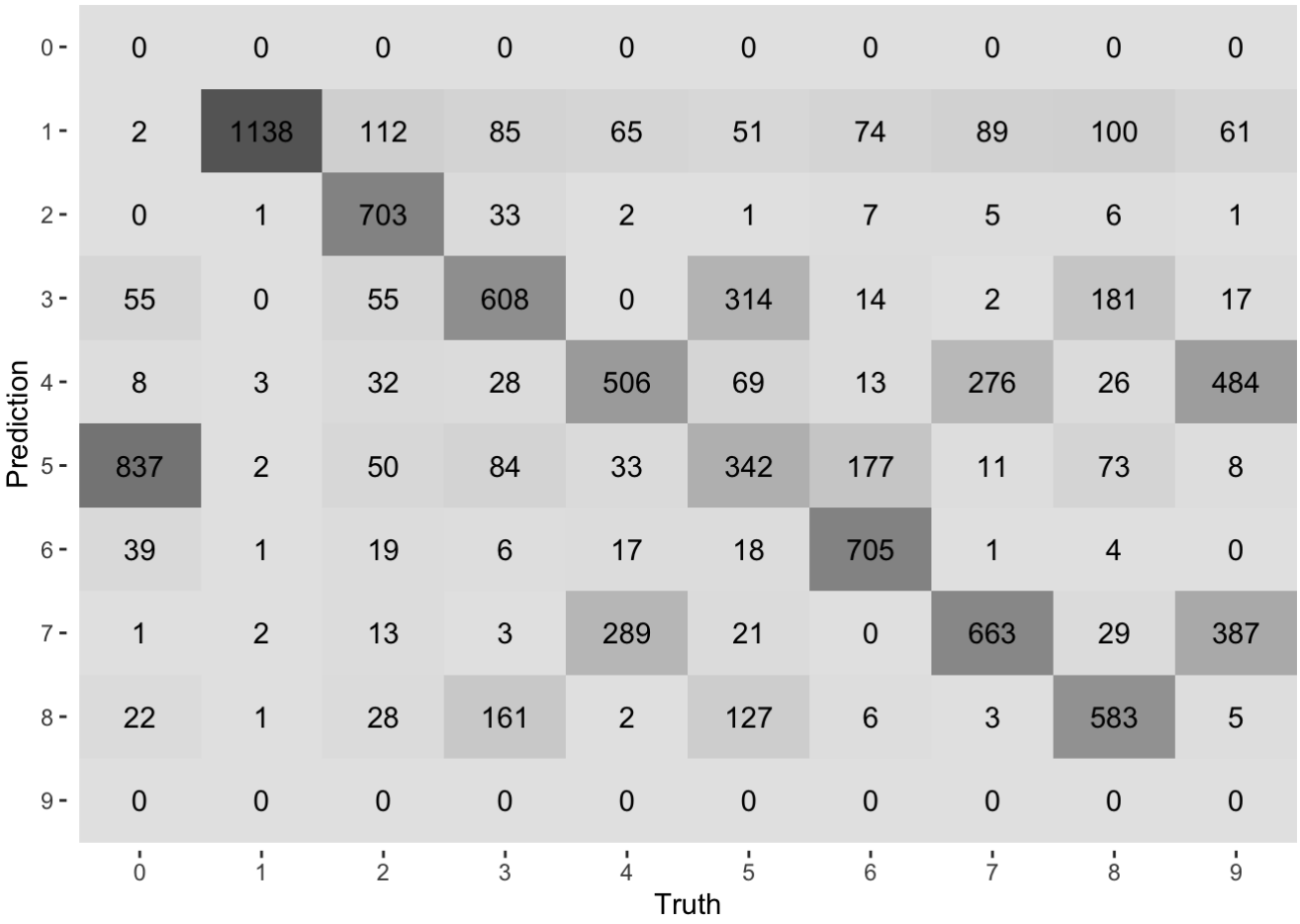
```
set.seed(1)
sample_1 = sample(60000,10000)
features <- mnist$train$images[sample_1,]

# Use k-means model with 10 centers and 10 random starts
mnist_clustering <- kmeans(features, centers = 10, nstart = 10)
```

```
## Warning: did not converge in 10 iterations

## Warning: did not converge in 10 iterations

## Warning: did not converge in 10 iterations

## Warning: did not converge in 10 iterations
```

```
# Print contents of the model output
str(mnist_clustering)
```

```
## List of 9
##  $ cluster     : int [1:10000] 7 7 6 9 1 10 5 9 5 10 ...
##  $ centers     : num [1:10, 1:784] 0 0 0 0 0 0 0 0 0 0 ...
##   ..- attr(*, "dimnames")=List of 2
##   .. ..$ : chr [1:10] "1" "2" "3" "4" ...
##   .. ..$ : NULL
##  $ totss       : num 3.42e+10
##  $ withinss    : num [1:10] 1.02e+09 2.25e+09 2.42e+09 2.59e+09 3.70e+09 ...
##  $ tot.withinss: num 2.54e+10
##  $ betweenss   : num 8.78e+09
##  $ size        : int [1:10] 734 810 759 914 1246 1043 1408 703 938 1445
##  $ iter        : int 11
##  $ ifault      : int 2
##  - attr(*, "class")= chr "kmeans"
```

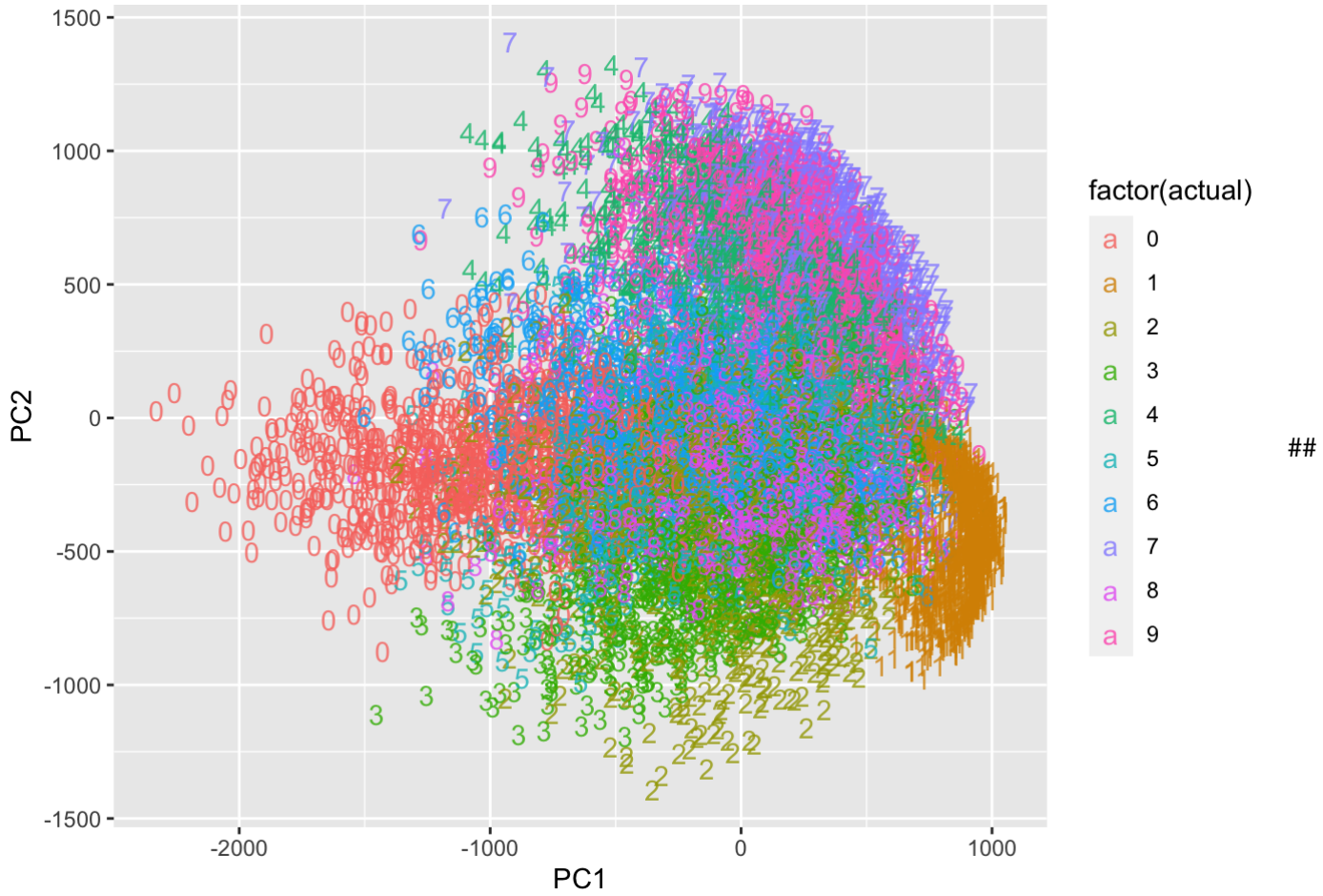
```
# Extract cluster centers
mnist_centers <- mnist_clustering$centers
# Create mode function
mode_fun <- function(x){ which.max(tabulate(x))
}
mnist_comparison <- data.frame(
  cluster = mnist_clustering$cluster,
  actual = mnist$train$labels[sample_1]
) %>%
  group_by(cluster) %>%
  mutate(mode = mode_fun(actual)) %>%
  ungroup() %>%
  mutate_all(factor, levels = 0:9)
# Create confusion matrix and plot results
yardstick::conf_mat(
  mnist_comparison,
  truth = actual,
  estimate = mode
) %>%
  autoplot(type = 'heatmap')
```



b)

```
m1 = prcomp(features, scale = F)
```

```
newdata = m1$x
actual = mnist$train$labels[sample_1]
ggplot(data.frame(newdata),aes(x=PC1,y=PC2))+geom_text(aes(label=actual,color = factor(a
ctual)),alpha=0.8)
```



From the plot above we can identidy that digit 0 are well seperated and digit 1 are not

Questioin 2

a)

```
df = crime
#head(df)
```

```
dfselected <- df$offense=="robbery"|df$offense=="aggravated assault"|df$offense=="rape"|
df$offense=="murder"
df2 <- df[dfselected,]
```

```
geocode("Houston")
```

```
## Source : https://maps.googleapis.com/maps/api/geocode/json?address=Houston&key=xxx
```

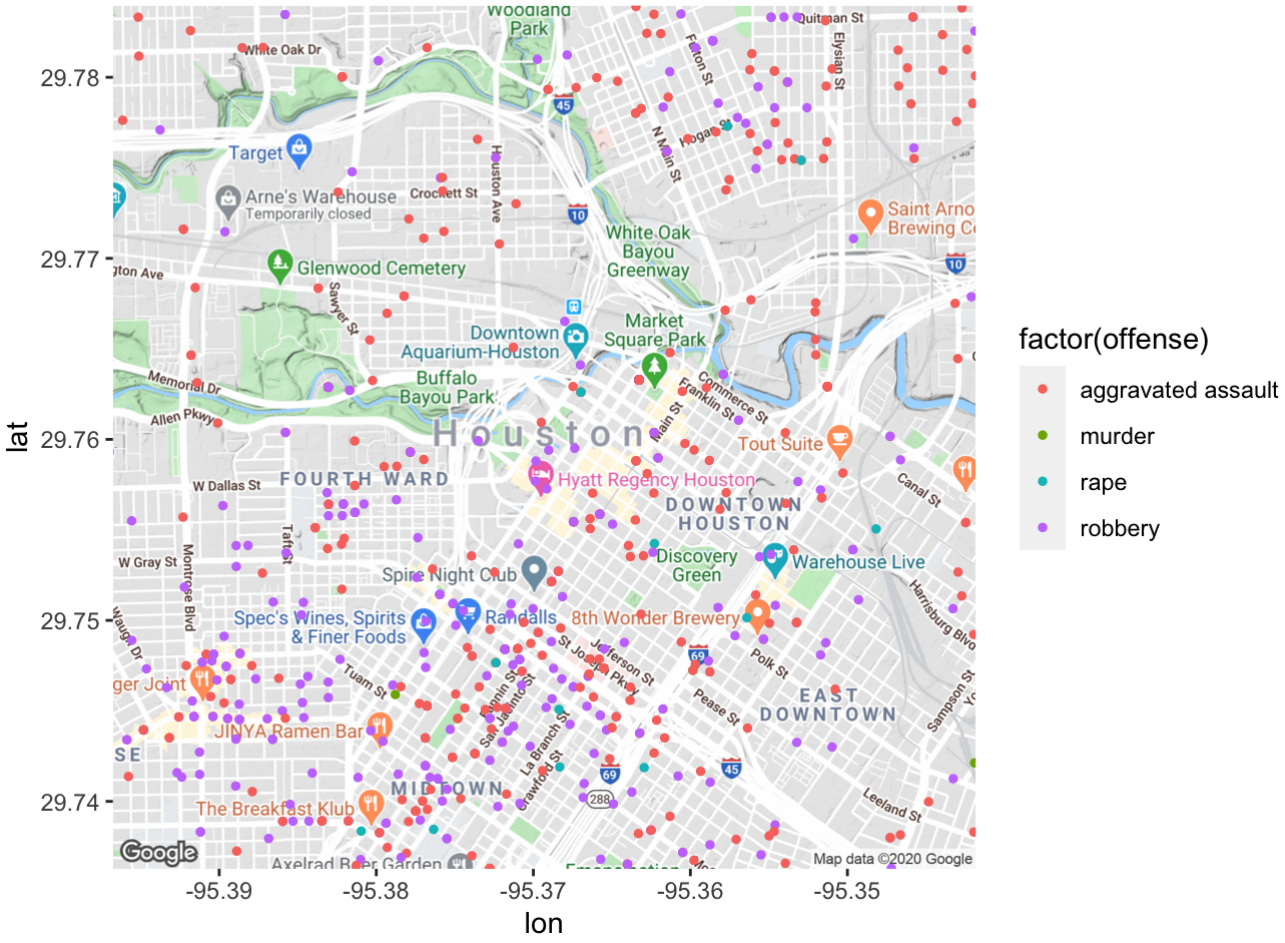
```
## # A tibble: 1 x 2
##     lon   lat
##   <dbl> <dbl>
## 1 -95.4  29.8
```

```
Houston.map = get_googlemap(center=c(lon = -95.369345,lat = 29.760155),lon=c(-95.39681,-95.34188),lat = c(29.73631,29.78400),zoom = calc_zoom(c(-95.39681,-95.34188),c(29.73631,29.78400)), scale = 2)
```

```
## Source : https://maps.googleapis.com/maps/api/staticmap?center=29.760155,-95.369345&zoom=14&size=640x640&scale=2&maptype=terrain&key=xxx
```

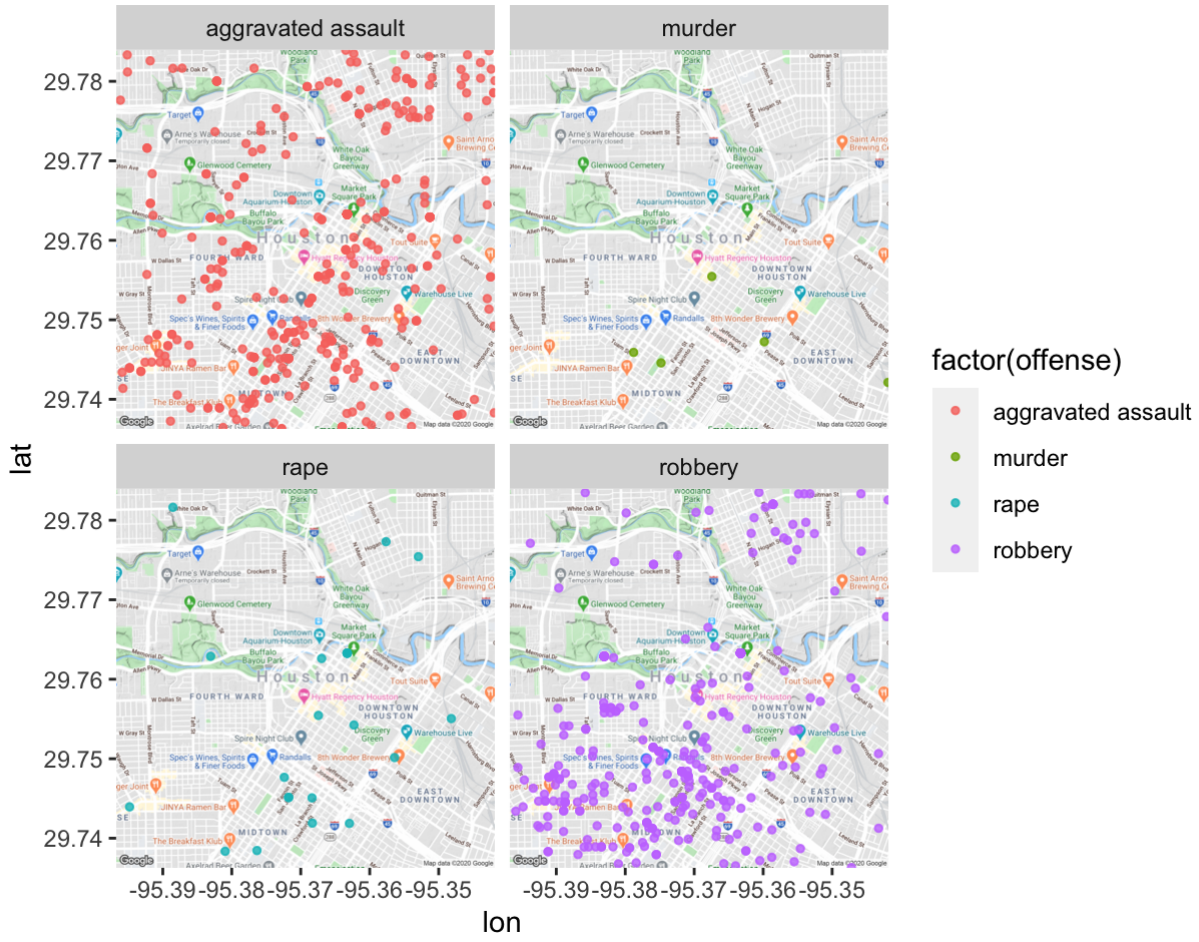
```
p = ggmap(Houston.map)
```

```
p+ geom_point(data = df2, aes(x=lon,y=lat,colour = factor(offense)),size = 1,alpha=1,na.rm = TRUE)
```



b)

```
p+ geom_point(data = df2, aes(x=lon,y=lat,colour = factor(offense)),size = 1,alpha=0.8,na.rm = TRUE)+facet_wrap(~offense,nrow = 2)
```



Question 3

```
data = read_csv("question3.csv")
```

```
## Warning: Missing column names filled in: 'X1' [1]
```

```
## Parsed with column specification:
## cols(
##   X1 = col_double(),
##   FIPS = col_double(),
##   County = col_character(),
##   population = col_double(),
##   cases = col_double(),
##   lat = col_double(),
##   lon = col_double()
## )
```

```
data$log <- log10(data$cases)
head(data)
```

```
## # A tibble: 6 x 8
##      X1  FIPS County            population cases   lat   lon   log
##   <dbl> <dbl> <chr>                  <dbl> <dbl> <dbl> <dbl> <dbl>
## 1     1 10001 Kent County           180786   503  39.1 -75.5  2.70
## 2     2 10003 New Castle County     558753  1352  39.6 -75.6  3.13
## 3     3 10005 Sussex County         234225  1317  38.7 -75.3  3.12
## 4     4  1001 Autauga County          55869    32  32.5 -86.6  1.51
## 5     5  1003 Baldwin County         223234   132  30.7 -87.7  2.12
## 6     6  1005 Barbour County          24686    29  31.9 -85.4  1.46
```

```
us <- c(left = -125, bottom = 25.75, right = -67, top = 49)
US.map = get_stamenmap(us, zoom = 5, maptype = "toner-lite")
```

```
## Source : http://tile.stamen.com/toner-lite/5/4/10.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/5/10.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/6/10.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/7/10.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/8/10.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/9/10.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/10/10.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/4/11.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/5/11.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/6/11.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/7/11.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/8/11.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/9/11.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/10/11.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/4/12.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/5/12.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/6/12.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/7/12.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/8/12.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/9/12.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/10/12.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/4/13.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/5/13.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/6/13.png
```

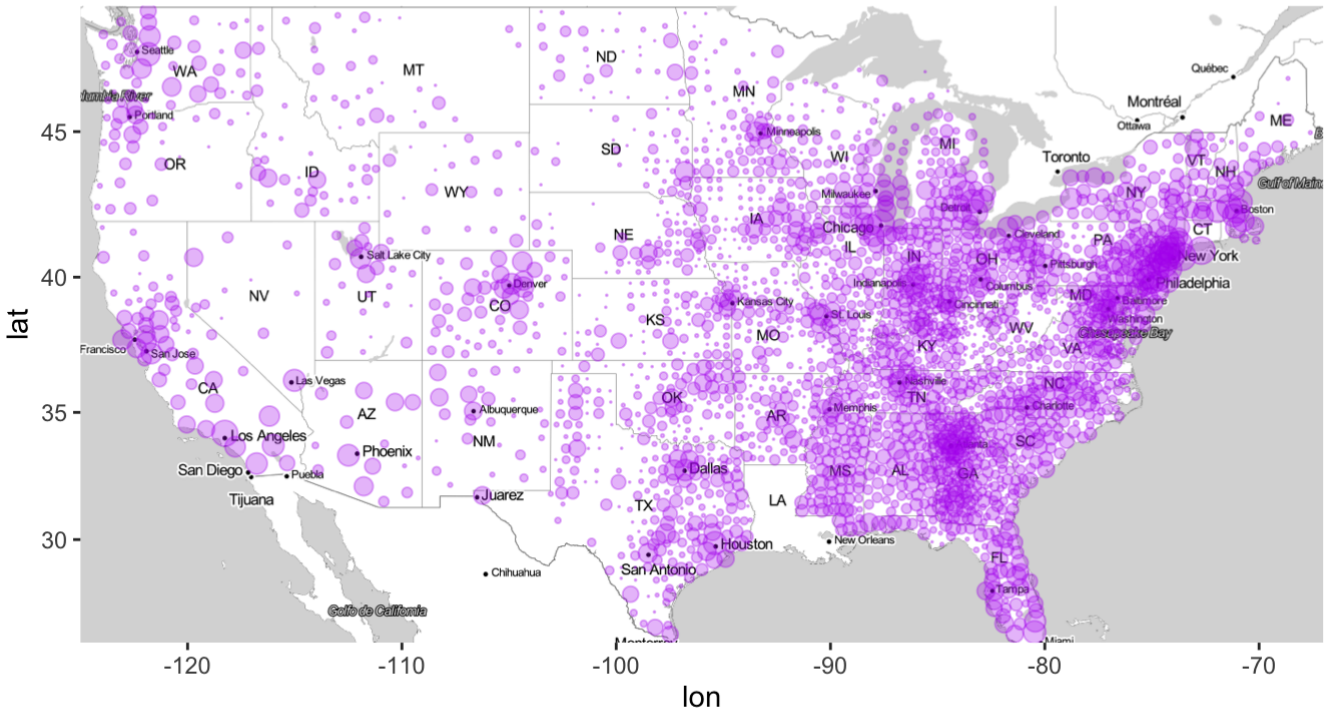
```
## Source : http://tile.stamen.com/toner-lite/5/7/13.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/8/13.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/9/13.png
```

```
## Source : http://tile.stamen.com/toner-lite/5/10/13.png
```

```
p = ggmap(US.map)
p+ geom_point(data=data,aes(x=lon,y=lat),colour = "#b207f0",size = data$log,alpha=0.3,n
a.rm = TRUE)
```


lat
lon
45
40
35
30
-120
-110
-100
-90
-80
-70
Seattle
Portland
WA
OR
MT
ID
WY
ND
SD
NE
MN
IA
WI
MI
IL
IN
OH
NY
PA
CT
MD
WV
VA
KY
TN
NC
SC
GA
FL
AL
MS
LA
AR
MO
KS
OK
TX
NM
AZ
NV
UT
CO
CA
Québec
Montréal
Ottawa
Toronto
ME
VT
NH
Boston
New York
Philadelphia
Baltimore
Washington
Chesapeake Bay
Gulf of Maine
Minneapolis
Milwaukee
Chicago
Detroit
Cleveland
Pittsburgh
Indianapolis
Columbus
Cincinnati
Kansas City
St. Louis
Nashville
Memphis
Charlotte
Atlanta
Dallas
Houston
San Antonio
New Orleans
Tampa
Miami
Salt Lake City
Denver
Las Vegas
Albuquerque
Phoenix
Los Angeles
San Diego
Tijuana
Juarez
Chihuahua
Francisco
San Jose
Puebla
Golfo de California