

Data Visualization 1

2023-10-02

R Markdown

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When you click the **Knit** button a document will be generated that includes both content as well as the output of any embedded R code chunks within the document. You can embed an R code chunk like this:

Note that the `echo = FALSE` parameter was added to the code chunk to prevent printing of the R code that generated the plot.

```
library(readr)
library(knitr)
```

```
data_description <- function(df) {
  cat("First few rows of the data:\n")
  print(head(df))
  cat("\n")

  cat("Statistical summary of the data:\n")
  print(summary(df))
  cat("\n")

  cat("Structure of the data:\n")
  str(df)
  cat("\n")

  cat("Dimensions of the data:\n")
  print(dim(df))
  cat("\n")

  cat("Column names in the data:\n")
  print(colnames(df))
  cat("\n")
}
```

```
county_logalpha <- read.csv("county_logalpha.csv")
data_description(county_logalpha)
```

```
## First few rows of the data:
##   X      county.name pop_density mean_PRE sd_PRE mean_POST sd_POST
## 1 0 Autauga, Alabama, US      91.8      0      0      0      0
## 2 1 Baldwin, Alabama, US     114.6      0      0      0      0
## 3 2 Barbour, Alabama, US      31.0      0      0      0      0
## 4 3 Bibb, Alabama, US       36.8      0      0      0      0
## 5 4 Blount, Alabama, US      88.9      0      0      0      0
## 6 5 Bullock, Alabama, US     17.5      0      0      0      0
```

```
##
## Statistical summary of the data:
##      X      county.name      pop_density      mean_PRE
## Min.   : 0.0      Length:3140      Min.    : 0.00      Min.    : -Inf
## 1st Qu.: 784.8      Class :character      1st Qu.: 16.98      1st Qu.: 0.0000
## Median :1569.5      Mode  :character      Median : 45.20      Median : 0.0000
## Mean   :1569.5                      Mean   : 259.28      Mean    : -Inf
## 3rd Qu.:2354.2                      3rd Qu.: 113.72      3rd Qu.: 0.0000
## Max.   :3139.0                      Max.    :69467.50      Max.    : 0.6931
##                                     NA's    :18
##      sd_PRE      mean_POST      sd_POST
## Min.   :0.00000      Min.   : -0.05068      Min.   :0.00000
## 1st Qu.:0.00000      1st Qu.: 0.00000      1st Qu.:0.00000
## Median :0.00000      Median : 0.00000      Median :0.00000
## Mean   :0.01364      Mean   : 0.02180      Mean   :0.01476
## 3rd Qu.:0.00000      3rd Qu.: 0.00000      3rd Qu.:0.00000
## Max.   :0.42105      Max.   : 0.48648      Max.   :0.38056
## NA's    :66                      NA's    :21
##
## Structure of the data:
## 'data.frame':    3140 obs. of  7 variables:
## $ X      : int  0 1 2 3 4 5 6 7 8 9 ...
## $ county.name: chr  "Autauga, Alabama, US" "Baldwin, Alabama, US" "Barbour, Alabama, US" "Bibb, Ala
## $ pop_density: num  91.8 114.6 31 36.8 88.9 ...
## $ mean_PRE   : num  0 0 0 0 0 ...
## $ sd_PRE     : num  0 0 0 0 0 ...
## $ mean_POST  : num  0 0 0 0 0 ...
## $ sd_POST    : num  0 0 0 0 0 ...
##
## Dimensions of the data:
## [1] 3140    7
##
## Column names in the data:
## [1] "X" "county.name" "pop_density" "mean_PRE" "sd_PRE"
## [6] "mean_POST" "sd_POST"

names(county_logalpha)[names(county_logalpha) == "county.name"] <- "long_county_name"

covid_county_census_data <- read.csv("covid_county_census_data.csv")
data_description(covid_county_census_data)

## First few rows of the data:
## X fips pst045212 pst040210 pst120212 pop010210 age135212 age295212 age775212
## 1 0 1001 55514 54571 1.7 54571 6.5 26.0 13.0
## 2 1 1003 190790 182265 4.7 182265 5.9 22.6 17.7
## 3 2 1005 27201 27457 -0.9 27457 5.6 21.2 15.2
## 4 3 1007 22597 22919 -1.4 22915 5.4 21.6 13.9
## 5 4 1009 57826 57322 0.9 57322 6.2 24.0 15.9
## 6 5 1011 10474 10915 -4.0 10914 6.8 21.2 14.6
## sex255212 rhi125212 rhi225212 rhi325212 rhi425212 rhi525212 rhi625212
## 1 51.3 78.5 18.4 0.5 1.0 0.1 1.5
## 2 51.2 87.3 9.6 0.7 0.8 0.1 1.5
## 3 46.3 50.5 47.4 0.7 0.4 0.2 0.8
## 4 46.0 76.2 22.3 0.4 0.1 0.1 0.9
## 5 50.5 96.2 1.7 0.6 0.3 0.1 1.1
```

```

## 6      45.6      27.2      69.7      0.9      0.2      0.8      1.1
## rhi725212 rhi825212 pop715211 pop645211 pop815211 edu635211 edu685211
## 1      2.6      76.4      84.8      2.0      3.8      86.5      21.6
## 2      4.6      83.2      83.3      3.7      5.4      87.9      27.2
## 3      5.1      46.4      82.1      3.0      5.0      72.4      13.9
## 4      2.0      74.6      90.7      1.3      2.0      75.9      9.8
## 5      8.6      88.1      86.6      4.5      6.8      73.2      11.3
## 6      7.6      22.6      84.4      4.5      5.1      72.0      14.4
## vet605211 lfe305211 hsg010211 hsg445211 hsg096211 hsg495211 hsd410211
## 1      5942      25.3      22460      77.7      7.2      137500      19998
## 2      20254      25.5      104701      76.2      23.3      175700      70757
## 3      2167      23.4      11692      66.4      10.3      91600      9589
## 4      1824      27.8      9014      83.0      6.3      87500      7225
## 5      4019      33.5      24144      79.9      4.9      111500      20954
## 6      830      28.3      4447      79.6      10.1      68000      3760
## hsd310211 inc910211 inc110211 pvpy020211 bza010211 bza110211 bza115211
## 1      2.68      25035      53899      10.9      835      10290      1.2
## 2      2.50      27217      51321      12.5      4624      51386      0.4
## 3      2.57      15899      34041      24.7      501      7572      -0.4
## 4      3.05      18462      40506      15.7      285      2980      1.8
## 5      2.70      21185      45404      13.7      690      6345      -1.4
## 6      2.78      20678      31955      26.0      113      0      0.0
## nes010211 sbo001207 sbo315207 sbo115207 sbo215207 sbo515207 sbo415207
## 1      3062      4067      15.2      0.0      1.3      0      0.7
## 2      16097      19035      2.7      0.4      1.0      0      1.3
## 3      1513      1667      0.0      0.0      0.0      0      0.0
## 4      1193      1385      14.9      0.0      0.0      0      0.0
## 5      3544      4458      0.0      0.0      0.0      0      0.0
## 6      478      417      0.0      0.0      0.0      0      0.0
## sbo015207 man450207 wtn220207 rtn130207 rtn131207 afn120207 bps030212
## 1      31.7      0      0      598175      12003      88157      385
## 2      27.3      1410270      0      2966490      17166      436955      1184
## 3      27.0      0      0      188337      6334      0      2
## 4      0.0      0      0      124707      5804      10757      16
## 5      23.2      341544      0      319700      5622      20941      6
## 6      38.8      0      0      43810      3995      3670      0
## lnd110210 pop060210 fips.1      date county_name      long_county_name
## 1      594.44      91.8      1001 1/22/20      Autauga Autauga, Alabama, US
## 2      1589.78      114.6      1003 1/22/20      Baldwin Baldwin, Alabama, US
## 3      884.88      31.0      1005 1/22/20      Barbour Barbour, Alabama, US
## 4      622.58      36.8      1007 1/22/20      Bibb Bibb, Alabama, US
## 5      644.78      88.9      1009 1/22/20      Blount Blount, Alabama, US
## 6      622.81      17.5      1011 1/22/20      Bullock Bullock, Alabama, US
## province_state death_count population
## 1      Alabama      0      55869
## 2      Alabama      0      223234
## 3      Alabama      0      24686
## 4      Alabama      0      22394
## 5      Alabama      0      57826
## 6      Alabama      0      10101
##
## Statistical summary of the data:
## X      fips      pst045212      pst040210
## Min. : 0      Min. : 1001      Min. : 71      Min. : 82

```

## 1st Qu.: 64370	1st Qu.:18178	1st Qu.: 11014	1st Qu.: 11118
## Median :128740	Median :29176	Median : 25834	Median : 25890
## Mean :128740	Mean :30388	Mean : 99964	Mean : 98319
## 3rd Qu.:193109	3rd Qu.:45080	3rd Qu.: 67355	3rd Qu.: 66898
## Max. :257479	Max. :56045	Max. :9962790	Max. :9818600
## pst120212	pop010210	age135212	age295212
## Min. :-18.1000	Min. : 82	Min. : 0.000	Min. : 0.00
## 1st Qu.: -1.1000	1st Qu.: 11118	1st Qu.: 5.300	1st Qu.:20.80
## Median : -0.1000	Median : 25890	Median : 5.900	Median :22.80
## Mean : 0.2127	Mean : 98318	Mean : 6.025	Mean :22.85
## 3rd Qu.: 1.3000	3rd Qu.: 66898	3rd Qu.: 6.600	3rd Qu.:24.50
## Max. : 25.6000	Max. :9818600	Max. :13.300	Max. :41.10
## age775212	sex255212	rhi125212	rhi225212
## Min. : 3.60	Min. :30.10	Min. :10.60	Min. : 0.000
## 1st Qu.:14.00	1st Qu.:49.50	1st Qu.:80.88	1st Qu.: 0.600
## Median :16.40	Median :50.40	Median :92.40	Median : 2.250
## Mean :16.75	Mean :49.96	Mean :85.51	Mean : 9.147
## 3rd Qu.:19.10	3rd Qu.:51.10	3rd Qu.:96.30	3rd Qu.:10.600
## Max. :49.30	Max. :57.00	Max. :99.20	Max. :85.400
## rhi325212	rhi425212	rhi525212	rhi625212
## Min. : 0.00	Min. : 0.000	Min. : 0.0000	Min. : 0.000
## 1st Qu.: 0.30	1st Qu.: 0.400	1st Qu.: 0.0000	1st Qu.: 1.100
## Median : 0.50	Median : 0.600	Median : 0.0000	Median : 1.400
## Mean : 2.14	Mean : 1.298	Mean : 0.1128	Mean : 1.791
## 3rd Qu.: 1.20	3rd Qu.: 1.100	3rd Qu.: 0.1000	3rd Qu.: 2.000
## Max. :86.80	Max. :43.300	Max. :48.9000	Max. :29.500
## rhi725212	rhi825212	pop715211	pop645211
## Min. : 0.200	Min. : 3.20	Min. : 49.00	Min. : 0.000
## 1st Qu.: 1.800	1st Qu.:66.20	1st Qu.: 83.70	1st Qu.: 1.200
## Median : 3.600	Median :85.20	Median : 86.60	Median : 2.450
## Mean : 8.667	Mean :77.76	Mean : 86.14	Mean : 4.446
## 3rd Qu.: 8.700	3rd Qu.:93.70	3rd Qu.: 89.20	3rd Qu.: 5.500
## Max. :95.600	Max. :98.60	Max. :100.00	Max. :63.400
## pop815211	edu635211	edu685211	vet605211
## Min. : 0.000	Min. :46.30	Min. : 4.20	Min. : 0.0
## 1st Qu.: 2.800	1st Qu.:79.20	1st Qu.:13.30	1st Qu.: 927.8
## Median : 4.900	Median :85.20	Median :17.10	Median : 2124.5
## Mean : 9.101	Mean :83.69	Mean :19.27	Mean : 7074.4
## 3rd Qu.:10.100	3rd Qu.:89.10	3rd Qu.:22.80	3rd Qu.: 5801.5
## Max. :95.900	Max. :98.60	Max. :72.00	Max. :354430.0
## lfe305211	hsg010211	hsg445211	hsg096211
## Min. : 4.30	Min. : 48	Min. : 0.00	Min. : 0.00
## 1st Qu.:19.20	1st Qu.: 5430	1st Qu.:69.20	1st Qu.: 6.20
## Median :22.60	Median : 12226	Median :74.30	Median : 9.70
## Mean :22.86	Mean : 42135	Mean :73.01	Mean :12.34
## 3rd Qu.:26.30	3rd Qu.: 30705	3rd Qu.:78.30	3rd Qu.:16.00
## Max. :42.50	Max. :3449270	Max. :93.70	Max. :98.40
## hsg495211	hsd410211	hsd310211	inc910211
## Min. : 0	Min. : 27	Min. :1.200	Min. : 9412
## 1st Qu.: 81775	1st Qu.: 4242	1st Qu.:2.370	1st Qu.:19534
## Median :108000	Median : 9821	Median :2.500	Median :22364
## Mean :133137	Mean : 36546	Mean :2.516	Mean :23127
## 3rd Qu.:154900	3rd Qu.: 25622	3rd Qu.:2.630	3rd Qu.:25516
## Max. :993900	Max. :3218520	Max. :4.070	Max. :61290

```

##      inc110211      pvy020211      bza010211      bza110211
## Min.   : 19344   Min.   : 0.00   Min.   : 0.0   Min.   : 0
## 1st Qu.: 37793   1st Qu.:11.30   1st Qu.: 222.0   1st Qu.: 2088
## Median : 43421   Median :15.10   Median : 536.5   Median : 6310
## Mean   : 45304   Mean   :15.84   Mean   : 2336.1   Mean   : 35256
## 3rd Qu.: 50178   3rd Qu.:19.40   3rd Qu.: 1440.8   3rd Qu.: 19242
## Max.   :120332   Max.   :48.40   Max.   :245261.0   Max.   :3648850
##      bza115211      nes010211      sbo001207      sbo315207
## Min.   : -40.8000   Min.   : 0.0   Min.   : 0.0   Min.   : 0.000
## 1st Qu.: -2.1000   1st Qu.: 750.8   1st Qu.: 904.8   1st Qu.: 0.000
## Median : 0.2000   Median : 1657.0   Median : 2159.0   Median : 0.000
## Mean   : 0.5151   Mean   : 7162.3   Mean   : 8788.7   Mean   : 2.393
## 3rd Qu.: 2.6000   3rd Qu.: 4259.2   3rd Qu.: 5643.0   3rd Qu.: 0.000
## Max.   :288.2000   Max.   :904398.0   Max.   :1046940.0   Max.   :66.700
##      sbo115207      sbo215207      sbo515207      sbo415207
## Min.   : 0.0000   Min.   : 0.000   Min.   : 0.00000   Min.   : 0.000
## 1st Qu.: 0.0000   1st Qu.: 0.000   1st Qu.: 0.00000   1st Qu.: 0.000
## Median : 0.0000   Median : 0.000   Median : 0.00000   Median : 0.000
## Mean   : 0.5752   Mean   : 0.801   Mean   : 0.01599   Mean   : 1.692
## 3rd Qu.: 0.0000   3rd Qu.: 0.000   3rd Qu.: 0.00000   3rd Qu.: 0.000
## Max.   :71.8000   Max.   :56.600   Max.   :10.50000   Max.   :78.000
##      sbo015207      man450207      wtn220207      rtn130207
## Min.   : 0.00   Min.   : 0   Min.   : 0   Min.   : 0
## 1st Qu.: 0.00   1st Qu.: 0   1st Qu.: 0   1st Qu.: 76040
## Median :23.20   Median : 0   Median : 44379   Median : 247782
## Mean   :17.96   Mean   : 1421028   Mean   : 1211975   Mean   : 1246527
## 3rd Qu.:27.80   3rd Qu.: 896601   3rd Qu.: 249462   3rd Qu.: 767070
## Max.   :56.20   Max.   :169275000   Max.   :205479000   Max.   :119112000
##      rtn131207      afn120207      bps030212      lnd110210
## Min.   : 0   Min.   : 0   Min.   : 0   Min.   : 2.0
## 1st Qu.: 6859   1st Qu.: 6068   1st Qu.: 6   1st Qu.: 430.7
## Median : 9729   Median : 25158   Median : 30   Median : 615.6
## Mean   :10237   Mean   : 193079   Mean   : 264   Mean   : 1118.7
## 3rd Qu.:12933   3rd Qu.: 93528   3rd Qu.: 128   3rd Qu.: 923.6
## Max.   :80800   Max.   :24857800   Max.   :27154   Max.   :145505.0
##      pop060210      fips.1      date      county_name
## Min.   : 0.00   Min.   : 1001   Length:257480   Length:257480
## 1st Qu.: 16.98   1st Qu.:18178   Class :character   Class :character
## Median : 45.20   Median :29176   Mode :character   Mode :character
## Mean   : 259.28   Mean   :30388
## 3rd Qu.: 113.72   3rd Qu.:45080
## Max.   :69467.50   Max.   :56045
##      long_county_name      province_state      death_count      population
## Length:257480   Length:257480   Min.   : 0.000   Min.   : 86
## Class :character   Class :character   1st Qu.: 0.000   1st Qu.: 10906
## Mode :character   Mode :character   Median : 0.000   Median : 25764
## Mean : 0.719   Mean : 105857
## 3rd Qu.: 0.000   3rd Qu.: 68104
## Max. :7349.000   Max. :10039107
##
## Structure of the data:
## 'data.frame': 257480 obs. of 60 variables:
## $ X : int 0 1 2 3 4 5 6 7 8 9 ...
## $ fips : int 1001 1003 1005 1007 1009 1011 1013 1015 1017 1019 ...

```

```

## $ pst045212 : num 55514 190790 27201 22597 57826 ...
## $ pst040210 : num 54571 182265 27457 22919 57322 ...
## $ pst120212 : num 1.7 4.7 -0.9 -1.4 0.9 -4 -3.1 -1.1 -0.4 0.1 ...
## $ pop010210 : num 54571 182265 27457 22915 57322 ...
## $ age135212 : num 6.5 5.9 5.6 5.4 6.2 6.8 6.1 6.1 5.8 4.8 ...
## $ age295212 : num 26 22.6 21.2 21.6 24 21.2 23.1 22.7 22 20.8 ...
## $ age775212 : num 13 17.7 15.2 13.9 15.9 14.6 17.4 15.2 17.6 19.3 ...
## $ sex255212 : num 51.3 51.2 46.3 46 50.5 45.6 53.2 51.8 52.3 50.3 ...
## $ rhi125212 : num 78.5 87.3 50.5 76.2 96.2 27.2 54.4 76 58.9 93.2 ...
## $ rhi225212 : num 18.4 9.6 47.4 22.3 1.7 69.7 43.5 21 39.1 4.6 ...
## $ rhi325212 : num 0.5 0.7 0.7 0.4 0.6 0.9 0.3 0.5 0.2 0.5 ...
## $ rhi425212 : num 1 0.8 0.4 0.1 0.3 0.2 1 0.8 0.6 0.3 ...
## $ rhi525212 : num 0.1 0.1 0.2 0.1 0.1 0.8 0 0.1 0.1 0 ...
## $ rhi625212 : num 1.5 1.5 0.8 0.9 1.1 1.1 0.7 1.6 1.1 1.4 ...
## $ rhi725212 : num 2.6 4.6 5.1 2 8.6 7.6 1.2 3.5 1.9 1.4 ...
## $ rhi825212 : num 76.4 83.2 46.4 74.6 88.1 22.6 53.7 73.2 57.6 91.9 ...
## $ pop715211 : num 84.8 83.3 82.1 90.7 86.6 84.4 93 83.5 84.8 89.3 ...
## $ pop645211 : num 2 3.7 3 1.3 4.5 4.5 1 2.3 1.1 0.8 ...
## $ pop815211 : num 3.8 5.4 5 2 6.8 5.1 1.6 4.5 1.8 1.3 ...
## $ edu635211 : num 86.5 87.9 72.4 75.9 73.2 72 74.7 77.9 74.2 73.5 ...
## $ edu685211 : num 21.6 27.2 13.9 9.8 11.3 14.4 12.4 15.8 10.5 11.3 ...
## $ vet605211 : num 5942 20254 2167 1824 4019 ...
## $ lfe305211 : num 25.3 25.5 23.4 27.8 33.5 28.3 23.9 22.2 24.1 25.8 ...
## $ hsg010211 : num 22460 104701 11692 9014 24144 ...
## $ hsg445211 : num 77.7 76.2 66.4 83 79.9 79.6 71.1 70.1 71.4 75.8 ...
## $ hsg096211 : num 7.2 23.3 10.3 6.3 4.9 10.1 13.8 13.8 10.6 4.5 ...
## $ hsg495211 : num 137500 175700 91600 87500 111500 ...
## $ hsd410211 : num 19998 70757 9589 7225 20954 ...
## $ hsd310211 : num 2.68 2.5 2.57 3.05 2.7 2.78 2.55 2.5 2.5 2.2 ...
## $ inc910211 : num 25035 27217 15899 18462 21185 ...
## $ inc110211 : num 53899 51321 34041 40506 45404 ...
## $ pvy020211 : num 10.9 12.5 24.7 15.7 13.7 26 25.7 20.4 21 20.8 ...
## $ bza010211 : num 835 4624 501 285 690 ...
## $ bza110211 : num 10290 51386 7572 2980 6345 ...
## $ bza115211 : num 1.2 0.4 -0.4 1.8 -1.4 0 -3.3 -2.3 4.6 3.6 ...
## $ nes010211 : num 3062 16097 1513 1193 3544 ...
## $ sbo001207 : num 4067 19035 1667 1385 4458 ...
## $ sbo315207 : num 15.2 2.7 0 14.9 0 0 0 7.2 0 0 ...
## $ sbo115207 : num 0 0.4 0 0 0 0 0 0 0 0 ...
## $ sbo215207 : num 1.3 1 0 0 0 0 3.3 1.6 0 0 ...
## $ sbo515207 : num 0 0 0 0 0 0 0 0 0 0 ...
## $ sbo415207 : num 0.7 1.3 0 0 0 0 0 0.5 0 0 ...
## $ sbo015207 : num 31.7 27.3 27 0 23.2 38.8 0 24.7 29.3 14.5 ...
## $ man450207 : num 0 1410270 0 0 341544 ...
## $ wtn220207 : num 0 0 0 0 0 ...
## $ rtn130207 : num 598175 2966490 188337 124707 319700 ...
## $ rtn131207 : num 12003 17166 6334 5804 5622 ...
## $ afn120207 : num 88157 436955 0 10757 20941 ...
## $ bps030212 : num 385 1184 2 16 6 ...
## $ lnd110210 : num 594 1590 885 623 645 ...
## $ pop060210 : num 91.8 114.6 31 36.8 88.9 ...
## $ fips.1 : int 1001 1003 1005 1007 1009 1011 1013 1015 1017 1019 ...
## $ date : chr "1/22/20" "1/22/20" "1/22/20" "1/22/20" ...
## $ county_name : chr "Autauga" "Baldwin" "Barbour" "Bibb" ...

```

```
## $ long_county_name: chr "Autauga, Alabama, US" "Baldwin, Alabama, US" "Barbour, Alabama, US" "Bibb
## $ province_state : chr "Alabama" "Alabama" "Alabama" "Alabama" ...
## $ death_count : int 0 0 0 0 0 0 0 0 0 ...
## $ population : int 55869 223234 24686 22394 57826 10101 19448 113605 33254 26196 ...
##
## Dimensions of the data:
## [1] 257480 60
##
## Column names in the data:
## [1] "X" "fips" "pst045212" "pst040210"
## [5] "pst120212" "pop010210" "age135212" "age295212"
## [9] "age775212" "sex255212" "rhi125212" "rhi225212"
## [13] "rhi325212" "rhi425212" "rhi525212" "rhi625212"
## [17] "rhi725212" "rhi825212" "pop715211" "pop645211"
## [21] "pop815211" "edu635211" "edu685211" "vet605211"
## [25] "lfe305211" "hsg010211" "hsg445211" "hsg096211"
## [29] "hsg495211" "hsd410211" "hsd310211" "inc910211"
## [33] "inc110211" "pvy020211" "bza010211" "bza110211"
## [37] "bza115211" "nes010211" "sbo001207" "sbo315207"
## [41] "sbo115207" "sbo215207" "sbo515207" "sbo415207"
## [45] "sbo015207" "man450207" "wtm220207" "rtn130207"
## [49] "rtn131207" "afn120207" "bps030212" "lnd110210"
## [53] "pop060210" "fips.1" "date" "county_name"
## [57] "long_county_name" "province_state" "death_count" "population"
```

```
covid_data <- merge(county_logalpha, covid_county_census_data, by = "long_county_name")
data_description(covid_data)
```

```
## First few rows of the data:
```

```
##           long_county_name X.x pop_density mean_PRE sd_PRE mean_POST
## 1 Abbeville, South Carolina, US 2315      51.8      0      0      0
## 2 Abbeville, South Carolina, US 2315      51.8      0      0      0
## 3 Abbeville, South Carolina, US 2315      51.8      0      0      0
## 4 Abbeville, South Carolina, US 2315      51.8      0      0      0
## 5 Abbeville, South Carolina, US 2315      51.8      0      0      0
## 6 Abbeville, South Carolina, US 2315      51.8      0      0      0
## sd_POST X.y fips pst045212 pst040210 pst120212 pop010210 age135212
## 1      0 171875 45001      25101      25421      -1.3      25417      5.6
## 2      0  83955 45001      25101      25421      -1.3      25417      5.6
## 3      0 253515 45001      25101      25421      -1.3      25417      5.6
## 4      0 237815 45001      25101      25421      -1.3      25417      5.6
## 5      0 121635 45001      25101      25421      -1.3      25417      5.6
## 6      0  90235 45001      25101      25421      -1.3      25417      5.6
## age295212 age775212 sex255212 rhi125212 rhi225212 rhi325212 rhi425212
## 1      22.2      17.9      51.5      69.9      28.2      0.3      0.4
## 2      22.2      17.9      51.5      69.9      28.2      0.3      0.4
## 3      22.2      17.9      51.5      69.9      28.2      0.3      0.4
## 4      22.2      17.9      51.5      69.9      28.2      0.3      0.4
## 5      22.2      17.9      51.5      69.9      28.2      0.3      0.4
## 6      22.2      17.9      51.5      69.9      28.2      0.3      0.4
## rhi525212 rhi625212 rhi725212 rhi825212 pop715211 pop645211 pop815211
## 1      0      1.2      1.2      68.9      90.1      1.5      2.3
## 2      0      1.2      1.2      68.9      90.1      1.5      2.3
## 3      0      1.2      1.2      68.9      90.1      1.5      2.3
## 4      0      1.2      1.2      68.9      90.1      1.5      2.3
```

```

## 5      0      1.2      1.2      68.9      90.1      1.5      2.3
## 6      0      1.2      1.2      68.9      90.1      1.5      2.3
##   edu635211 edu685211 vet605211 lfe305211 hsg010211 hsg445211 hsg096211
## 1      77.1      13.9      2019      26      12072      77.9      6.9
## 2      77.1      13.9      2019      26      12072      77.9      6.9
## 3      77.1      13.9      2019      26      12072      77.9      6.9
## 4      77.1      13.9      2019      26      12072      77.9      6.9
## 5      77.1      13.9      2019      26      12072      77.9      6.9
## 6      77.1      13.9      2019      26      12072      77.9      6.9
##   hsg495211 hsd410211 hsd310211 inc910211 inc110211 pvpy020211 bza010211
## 1      89100      9811      2.53      17424      34670      19.6      334
## 2      89100      9811      2.53      17424      34670      19.6      334
## 3      89100      9811      2.53      17424      34670      19.6      334
## 4      89100      9811      2.53      17424      34670      19.6      334
## 5      89100      9811      2.53      17424      34670      19.6      334
## 6      89100      9811      2.53      17424      34670      19.6      334
##   bza110211 bza115211 nes010211 sbo001207 sbo315207 sbo115207 sbo215207
## 1      4119      -4.1      1472      1385      19.1      0      0
## 2      4119      -4.1      1472      1385      19.1      0      0
## 3      4119      -4.1      1472      1385      19.1      0      0
## 4      4119      -4.1      1472      1385      19.1      0      0
## 5      4119      -4.1      1472      1385      19.1      0      0
## 6      4119      -4.1      1472      1385      19.1      0      0
##   sbo515207 sbo415207 sbo015207 man450207 wtn220207 rtn130207 rtn131207
## 1      0      0      33.4      657498      0      71936      2841
## 2      0      0      33.4      657498      0      71936      2841
## 3      0      0      33.4      657498      0      71936      2841
## 4      0      0      33.4      657498      0      71936      2841
## 5      0      0      33.4      657498      0      71936      2841
## 6      0      0      33.4      657498      0      71936      2841
##   afn120207 bps030212 lnd110210 pop060210 fips.1      date county_name
## 1      10963      23      490.48      51.8      45001 3/17/20 Abbeville
## 2      10963      23      490.48      51.8      45001 2/17/20 Abbeville
## 3      10963      23      490.48      51.8      45001 4/12/20 Abbeville
## 4      10963      23      490.48      51.8      45001 4/7/20 Abbeville
## 5      10963      23      490.48      51.8      45001 3/1/20 Abbeville
## 6      10963      23      490.48      51.8      45001 2/19/20 Abbeville
##   province_state death_count population
## 1 South Carolina      0      24527
## 2 South Carolina      0      24527
## 3 South Carolina      0      24527
## 4 South Carolina      0      24527
## 5 South Carolina      0      24527
## 6 South Carolina      0      24527
##
## Statistical summary of the data:
##   long_county_name      X.x      pop_density      mean_PRE
## Length:257480      Min.      :      0.0      Min.      :      0.00      Min.      : -Inf
## Class :character      1st Qu.: 784.8      1st Qu.:      16.98      1st Qu.:0.0000
## Mode  :character      Median :1569.5      Median :      45.20      Median :0.0000
##      Mean      :1569.5      Mean      :      259.28      Mean      : -Inf
##      3rd Qu.:2354.2      3rd Qu.:      113.72      3rd Qu.:0.0000
##      Max.      :3139.0      Max.      :69467.50      Max.      :0.6931
##      NA's      :1476

```



```

##      sd_PRE      mean_POST      sd_POST      X.y
## Min.   :0.000   Min.   :-0.05068   Min.   :0.0000   Min.   :    0
## 1st Qu.:0.000   1st Qu.: 0.00000   1st Qu.:0.0000   1st Qu.: 64370
## Median :0.000   Median : 0.00000   Median :0.0000   Median :128740
## Mean   :0.014   Mean   : 0.02180   Mean   :0.0148   Mean   :128740
## 3rd Qu.:0.000   3rd Qu.: 0.00000   3rd Qu.:0.0000   3rd Qu.:193109
## Max.   :0.421   Max.   : 0.48648   Max.   :0.3806   Max.   :257479
## NA's   :5412
##      fips      pst045212      pst040210      pst120212
## Min.   : 1001   Min.   :    71   Min.   :    82   Min.   : -18.1000
## 1st Qu.:18178   1st Qu.: 11014   1st Qu.: 11118   1st Qu.: -1.1000
## Median :29176   Median : 25834   Median : 25890   Median : -0.1000
## Mean   :30388   Mean   : 99964   Mean   : 98319   Mean   : 0.2127
## 3rd Qu.:45080   3rd Qu.: 67355   3rd Qu.: 66898   3rd Qu.: 1.3000
## Max.   :56045   Max.   :9962790   Max.   :9818600   Max.   : 25.6000
##
##      pop010210      age135212      age295212      age775212
## Min.   :    82   Min.   : 0.000   Min.   : 0.00   Min.   : 3.60
## 1st Qu.: 11118   1st Qu.: 5.300   1st Qu.:20.80   1st Qu.:14.00
## Median : 25890   Median : 5.900   Median :22.80   Median :16.40
## Mean   : 98318   Mean   : 6.025   Mean   :22.85   Mean   :16.75
## 3rd Qu.: 66898   3rd Qu.: 6.600   3rd Qu.:24.50   3rd Qu.:19.10
## Max.   :9818600   Max.   :13.300   Max.   :41.10   Max.   :49.30
##
##      sex255212      rhi125212      rhi225212      rhi325212
## Min.   :30.10   Min.   :10.60   Min.   : 0.000   Min.   : 0.00
## 1st Qu.:49.50   1st Qu.:80.88   1st Qu.: 0.600   1st Qu.: 0.30
## Median :50.40   Median :92.40   Median : 2.250   Median : 0.50
## Mean   :49.96   Mean   :85.51   Mean   : 9.147   Mean   : 2.14
## 3rd Qu.:51.10   3rd Qu.:96.30   3rd Qu.:10.600   3rd Qu.: 1.20
## Max.   :57.00   Max.   :99.20   Max.   :85.400   Max.   :86.80
##
##      rhi425212      rhi525212      rhi625212      rhi725212
## Min.   : 0.000   Min.   : 0.0000   Min.   : 0.000   Min.   : 0.200
## 1st Qu.: 0.400   1st Qu.: 0.0000   1st Qu.: 1.100   1st Qu.: 1.800
## Median : 0.600   Median : 0.0000   Median : 1.400   Median : 3.600
## Mean   : 1.298   Mean   : 0.1128   Mean   : 1.791   Mean   : 8.667
## 3rd Qu.: 1.100   3rd Qu.: 0.1000   3rd Qu.: 2.000   3rd Qu.: 8.700
## Max.   :43.300   Max.   :48.9000   Max.   :29.500   Max.   :95.600
##
##      rhi825212      pop715211      pop645211      pop815211
## Min.   : 3.20   Min.   : 49.00   Min.   : 0.000   Min.   : 0.000
## 1st Qu.:66.20   1st Qu.: 83.70   1st Qu.: 1.200   1st Qu.: 2.800
## Median :85.20   Median : 86.60   Median : 2.450   Median : 4.900
## Mean   :77.76   Mean   : 86.14   Mean   : 4.446   Mean   : 9.101
## 3rd Qu.:93.70   3rd Qu.: 89.20   3rd Qu.: 5.500   3rd Qu.:10.100
## Max.   :98.60   Max.   :100.00   Max.   :63.400   Max.   :95.900
##
##      edu635211      edu685211      vet605211      lfe305211
## Min.   :46.30   Min.   : 4.20   Min.   :    0.0   Min.   : 4.30
## 1st Qu.:79.20   1st Qu.:13.30   1st Qu.: 927.8   1st Qu.:19.20
## Median :85.20   Median :17.10   Median : 2124.5   Median :22.60
## Mean   :83.69   Mean   :19.27   Mean   : 7074.4   Mean   :22.86
## 3rd Qu.:89.10   3rd Qu.:22.80   3rd Qu.: 5801.5   3rd Qu.:26.30

```

```

## Max. :98.60 Max. :72.00 Max. :354430.0 Max. :42.50
##
## hsg010211 hsg445211 hsg096211 hsg495211
## Min. : 48 Min. : 0.00 Min. : 0.00 Min. : 0
## 1st Qu.: 5430 1st Qu.:69.20 1st Qu.: 6.20 1st Qu.: 81775
## Median : 12226 Median :74.30 Median : 9.70 Median :108000
## Mean : 42135 Mean :73.01 Mean :12.34 Mean :133137
## 3rd Qu.: 30705 3rd Qu.:78.30 3rd Qu.:16.00 3rd Qu.:154900
## Max. :3449270 Max. :93.70 Max. :98.40 Max. :993900
##
## hsd410211 hsd310211 inc910211 inc110211
## Min. : 27 Min. :1.200 Min. : 9412 Min. : 19344
## 1st Qu.: 4242 1st Qu.:2.370 1st Qu.:19534 1st Qu.: 37793
## Median : 9821 Median :2.500 Median :22364 Median : 43421
## Mean : 36546 Mean :2.516 Mean :23127 Mean : 45304
## 3rd Qu.: 25622 3rd Qu.:2.630 3rd Qu.:25516 3rd Qu.: 50178
## Max. :3218520 Max. :4.070 Max. :61290 Max. :120332
##
## pvy020211 bza010211 bza110211 bza115211
## Min. : 0.00 Min. : 0.0 Min. : 0 Min. : -40.8000
## 1st Qu.:11.30 1st Qu.: 222.0 1st Qu.: 2088 1st Qu.: -2.1000
## Median :15.10 Median : 536.5 Median : 6310 Median : 0.2000
## Mean :15.84 Mean : 2336.1 Mean : 35256 Mean : 0.5151
## 3rd Qu.:19.40 3rd Qu.: 1440.8 3rd Qu.: 19242 3rd Qu.: 2.6000
## Max. :48.40 Max. :245261.0 Max. :3648850 Max. :288.2000
##
## nes010211 sbo001207 sbo315207 sbo115207
## Min. : 0.0 Min. : 0.0 Min. : 0.000 Min. : 0.0000
## 1st Qu.: 750.8 1st Qu.: 904.8 1st Qu.: 0.000 1st Qu.: 0.0000
## Median : 1657.0 Median : 2159.0 Median : 0.000 Median : 0.0000
## Mean : 7162.3 Mean : 8788.7 Mean : 2.393 Mean : 0.5752
## 3rd Qu.: 4259.2 3rd Qu.: 5643.0 3rd Qu.: 0.000 3rd Qu.: 0.0000
## Max. :904398.0 Max. :1046940.0 Max. :66.700 Max. :71.8000
##
## sbo215207 sbo515207 sbo415207 sbo015207
## Min. : 0.000 Min. : 0.00000 Min. : 0.000 Min. : 0.00
## 1st Qu.: 0.000 1st Qu.: 0.00000 1st Qu.: 0.000 1st Qu.: 0.00
## Median : 0.000 Median : 0.00000 Median : 0.000 Median :23.20
## Mean : 0.801 Mean : 0.01599 Mean : 1.692 Mean :17.96
## 3rd Qu.: 0.000 3rd Qu.: 0.00000 3rd Qu.: 0.000 3rd Qu.:27.80
## Max. :56.600 Max. :10.50000 Max. :78.000 Max. :56.20
##
## man450207 wtn220207 rtn130207 rtn131207
## Min. : 0 Min. : 0 Min. : 0 Min. : 0
## 1st Qu.: 0 1st Qu.: 0 1st Qu.: 76040 1st Qu.: 6859
## Median : 0 Median : 44379 Median : 247782 Median : 9729
## Mean : 1421028 Mean : 1211975 Mean : 1246527 Mean :10237
## 3rd Qu.: 896601 3rd Qu.: 249462 3rd Qu.: 767070 3rd Qu.:12933
## Max. :169275000 Max. :205479000 Max. :119112000 Max. :80800
##
## afn120207 bps030212 lnd110210 pop060210
## Min. : 0 Min. : 0 Min. : 2.0 Min. : 0.00
## 1st Qu.: 6068 1st Qu.: 6 1st Qu.: 430.7 1st Qu.: 16.98
## Median : 25158 Median : 30 Median : 615.6 Median : 45.20

```

```

## Mean : 193079 Mean : 264 Mean : 1118.7 Mean : 259.28
## 3rd Qu.: 93528 3rd Qu.: 128 3rd Qu.: 923.6 3rd Qu.: 113.72
## Max. :24857800 Max. :27154 Max. :145505.0 Max. :69467.50
##
## fips.1 date county_name province_state
## Min. : 1001 Length:257480 Length:257480 Length:257480
## 1st Qu.:18178 Class :character Class :character Class :character
## Median :29176 Mode :character Mode :character Mode :character
## Mean :30388
## 3rd Qu.:45080
## Max. :56045
##
## death_count population
## Min. : 0.000 Min. : 86
## 1st Qu.: 0.000 1st Qu.: 10906
## Median : 0.000 Median : 25764
## Mean : 0.719 Mean : 105857
## 3rd Qu.: 0.000 3rd Qu.: 68104
## Max. :7349.000 Max. :10039107
##
##
## Structure of the data:
## 'data.frame': 257480 obs. of 66 variables:
## $ long_county_name: chr "Abbeville, South Carolina, US" "Abbeville, South Carolina, US" "Abbeville
## $ X.x : int 2315 2315 2315 2315 2315 2315 2315 2315 2315 2315 ...
## $ pop_density : num 51.8 51.8 51.8 51.8 51.8 51.8 51.8 51.8 51.8 51.8 ...
## $ mean_PRE : num 0 0 0 0 0 0 0 0 0 0 ...
## $ sd_PRE : num 0 0 0 0 0 0 0 0 0 0 ...
## $ mean_POST : num 0 0 0 0 0 0 0 0 0 0 ...
## $ sd_POST : num 0 0 0 0 0 0 0 0 0 0 ...
## $ X.y : int 171875 83955 253515 237815 121635 90235 140475 93375 231535 77675 ...
## $ fips : int 45001 45001 45001 45001 45001 45001 45001 45001 45001 45001 ...
## $ pst045212 : num 25101 25101 25101 25101 25101 ...
## $ pst040210 : num 25421 25421 25421 25421 25421 ...
## $ pst120212 : num -1.3 -1.3 -1.3 -1.3 -1.3 -1.3 -1.3 -1.3 -1.3 -1.3 ...
## $ pop010210 : num 25417 25417 25417 25417 25417 ...
## $ age135212 : num 5.6 5.6 5.6 5.6 5.6 5.6 5.6 5.6 5.6 5.6 ...
## $ age295212 : num 22.2 22.2 22.2 22.2 22.2 22.2 22.2 22.2 22.2 22.2 ...
## $ age775212 : num 17.9 17.9 17.9 17.9 17.9 17.9 17.9 17.9 17.9 17.9 ...
## $ sex255212 : num 51.5 51.5 51.5 51.5 51.5 51.5 51.5 51.5 51.5 51.5 ...
## $ rhi125212 : num 69.9 69.9 69.9 69.9 69.9 69.9 69.9 69.9 69.9 69.9 ...
## $ rhi225212 : num 28.2 28.2 28.2 28.2 28.2 28.2 28.2 28.2 28.2 28.2 ...
## $ rhi325212 : num 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 0.3 ...
## $ rhi425212 : num 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 0.4 ...
## $ rhi525212 : num 0 0 0 0 0 0 0 0 0 0 ...
## $ rhi625212 : num 1.2 1.2 1.2 1.2 1.2 1.2 1.2 1.2 1.2 1.2 ...
## $ rhi725212 : num 1.2 1.2 1.2 1.2 1.2 1.2 1.2 1.2 1.2 1.2 ...
## $ rhi825212 : num 68.9 68.9 68.9 68.9 68.9 68.9 68.9 68.9 68.9 68.9 ...
## $ pop715211 : num 90.1 90.1 90.1 90.1 90.1 90.1 90.1 90.1 90.1 90.1 ...
## $ pop645211 : num 1.5 1.5 1.5 1.5 1.5 1.5 1.5 1.5 1.5 1.5 ...
## $ pop815211 : num 2.3 2.3 2.3 2.3 2.3 2.3 2.3 2.3 2.3 2.3 ...
## $ edu635211 : num 77.1 77.1 77.1 77.1 77.1 77.1 77.1 77.1 77.1 77.1 ...
## $ edu685211 : num 13.9 13.9 13.9 13.9 13.9 13.9 13.9 13.9 13.9 13.9 ...
## $ vet605211 : num 2019 2019 2019 2019 2019 ...

```

```

## $ lfe305211      : num  26 26 26 26 26 26 26 26 26 26 ...
## $ hsg010211      : num  12072 12072 12072 12072 12072 ...
## $ hsg445211      : num  77.9 77.9 77.9 77.9 77.9 77.9 77.9 77.9 77.9 77.9 ...
## $ hsg096211      : num  6.9 6.9 6.9 6.9 6.9 6.9 6.9 6.9 6.9 6.9 ...
## $ hsg495211      : num  89100 89100 89100 89100 89100 89100 89100 89100 89100 89100 ...
## $ hsd410211      : num  9811 9811 9811 9811 9811 ...
## $ hsd310211      : num  2.53 2.53 2.53 2.53 2.53 2.53 2.53 2.53 2.53 2.53 ...
## $ inc910211      : num  17424 17424 17424 17424 17424 ...
## $ inc110211      : num  34670 34670 34670 34670 34670 ...
## $ pvy020211      : num  19.6 19.6 19.6 19.6 19.6 19.6 19.6 19.6 19.6 19.6 ...
## $ bza010211      : num  334 334 334 334 334 334 334 334 334 334 ...
## $ bza110211      : num  4119 4119 4119 4119 4119 ...
## $ bza115211      : num  -4.1 -4.1 -4.1 -4.1 -4.1 -4.1 -4.1 -4.1 -4.1 -4.1 ...
## $ nes010211      : num  1472 1472 1472 1472 1472 ...
## $ sbo001207      : num  1385 1385 1385 1385 1385 ...
## $ sbo315207      : num  19.1 19.1 19.1 19.1 19.1 19.1 19.1 19.1 19.1 19.1 ...
## $ sbo115207      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ sbo215207      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ sbo515207      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ sbo415207      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ sbo015207      : num  33.4 33.4 33.4 33.4 33.4 33.4 33.4 33.4 33.4 33.4 ...
## $ man450207      : num  657498 657498 657498 657498 657498 ...
## $ wtn220207      : num  0 0 0 0 0 0 0 0 0 0 ...
## $ rtn130207      : num  71936 71936 71936 71936 71936 ...
## $ rtn131207      : num  2841 2841 2841 2841 2841 ...
## $ afn120207      : num  10963 10963 10963 10963 10963 ...
## $ bps030212      : num  23 23 23 23 23 23 23 23 23 23 ...
## $ lnd110210      : num  490 490 490 490 490 ...
## $ pop060210      : num  51.8 51.8 51.8 51.8 51.8 51.8 51.8 51.8 51.8 51.8 ...
## $ fips.1         : int  45001 45001 45001 45001 45001 45001 45001 45001 45001 45001 ...
## $ date           : chr   "3/17/20" "2/17/20" "4/12/20" "4/7/20" ...
## $ county_name    : chr   "Abbeville" "Abbeville" "Abbeville" "Abbeville" ...
## $ province_state  : chr   "South Carolina" "South Carolina" "South Carolina" "South Carolina" ...
## $ death_count     : int    0 0 0 0 0 0 0 0 0 0 ...
## $ population      : int  24527 24527 24527 24527 24527 24527 24527 24527 24527 24527 ...
##
## Dimensions of the data:
## [1] 257480      66
##
## Column names in the data:
## [1] "long_county_name" "X.x" "pop_density" "mean_PRE"
## [5] "sd_PRE" "mean_POST" "sd_POST" "X.y"
## [9] "fips" "pst045212" "pst040210" "pst120212"
## [13] "pop010210" "age135212" "age295212" "age775212"
## [17] "sex255212" "rhi125212" "rhi225212" "rhi325212"
## [21] "rhi425212" "rhi525212" "rhi625212" "rhi725212"
## [25] "rhi825212" "pop715211" "pop645211" "pop815211"
## [29] "edu635211" "edu685211" "vet605211" "lfe305211"
## [33] "hsg010211" "hsg445211" "hsg096211" "hsg495211"
## [37] "hsd410211" "hsd310211" "inc910211" "inc110211"
## [41] "pvy020211" "bza010211" "bza110211" "bza115211"
## [45] "nes010211" "sbo001207" "sbo315207" "sbo115207"
## [49] "sbo215207" "sbo515207" "sbo415207" "sbo015207"
## [53] "man450207" "wtm220207" "rtn130207" "rtn131207"

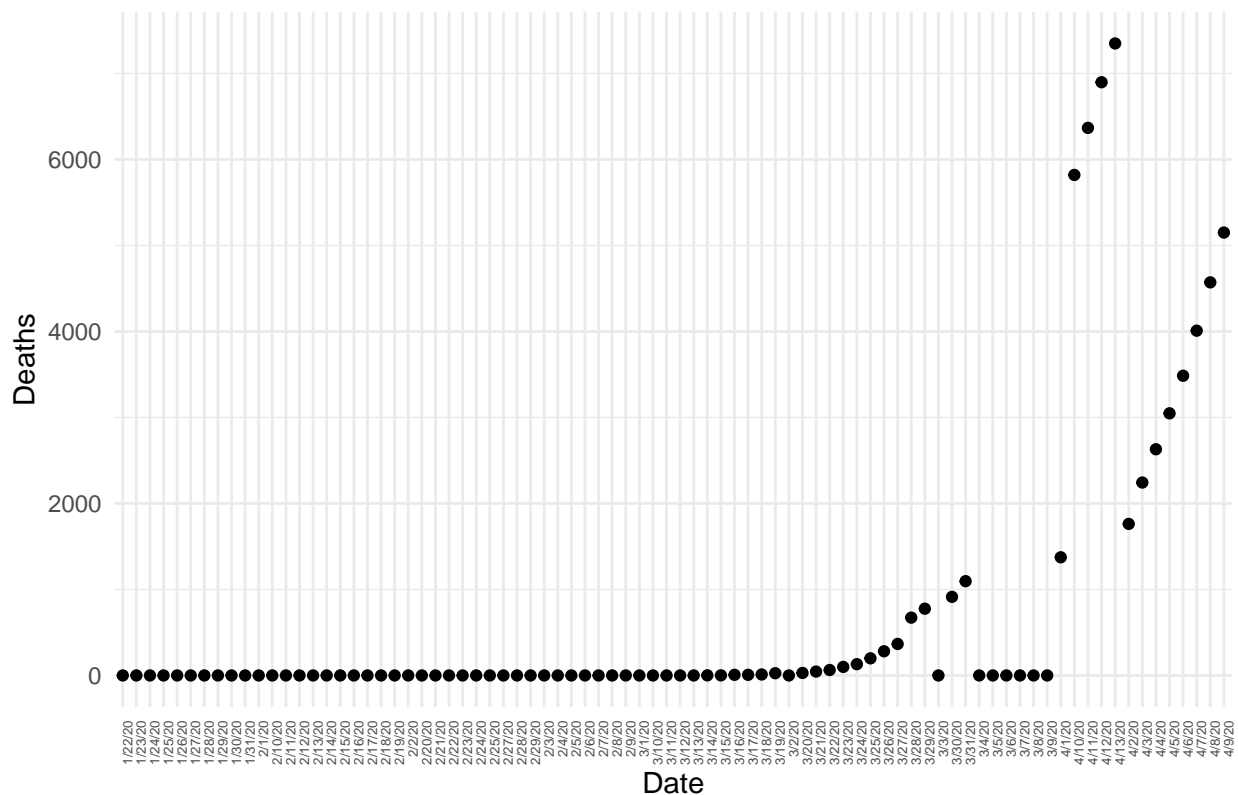
```

```
## [57] "afn120207"      "bps030212"      "lnd110210"      "pop060210"
## [61] "fips.1"         "date"           "county_name"     "province_state"
## [65] "death_count"    "population"
```

```
ny <- covid_data[covid_data$county_name == 'New York', ]
```

```
library(ggplot2)
ggplot(data = ny, aes(x = date, y = death_count)) +
  geom_line() +
  geom_point() +
  theme_minimal() +
  labs(title = "Time Trend of Deaths in New York County",
       x = "Date",
       y = "Deaths") +
  theme(axis.text.x = element_text(angle = 90, hjust = 1, size = 5))
```

Time Trend of Deaths in New York County



```
df <- read.csv('case-hosp-death.csv', header=TRUE)
data_description(df)
```

```
## First few rows of the data:
```

```
##   DATE_OF_INTEREST CASE_COUNT HOSPITALIZED_COUNT DEATH_COUNT
## 1    02/29/2020         1         11             0
## 2    03/01/2020         0          4             0
## 3    03/02/2020         0         21             0
## 4    03/03/2020         2         20             0
## 5    03/04/2020         5         22             0
## 6    03/05/2020         3         20             0
##   CASE_COUNT_7DAY_AVG INCOMPLETE
```

```

## 1          NA          NA
## 2          NA          NA
## 3          NA          NA
## 4          NA          NA
## 5          NA          NA
## 6          NA          NA
##
## Statistical summary of the data:
## DATE_OF_INTEREST CASE_COUNT HOSPITALIZED_COUNT DEATH_COUNT
## Length:173      Min.      : 0      Min.      : 0.0      Min.      : 0.0
## Class :character 1st Qu.: 260    1st Qu.: 31.0    1st Qu.: 8.0
## Mode  :character Median : 437    Median : 65.0    Median : 25.0
##                      Mean  :1316   Mean  : 320.2    Mean  :109.6
##                      3rd Qu.:2024   3rd Qu.: 359.0    3rd Qu.:131.0
##                      Max.   :6377   Max.   :1723.0    Max.   :598.0
##
## CASE_COUNT_7DAY_AVG INCOMPLETE
## Min.      : 3      Min.      :6000
## 1st Qu.: 314      1st Qu.:6000
## Median : 446      Median :6000
## Mean  :1362      Mean  :6000
## 3rd Qu.:2030      3rd Qu.:6000
## Max.   :5306      Max.   :6000
## NA's   :6         NA's   :166
##
## Structure of the data:
## 'data.frame': 173 obs. of 6 variables:
## $ DATE_OF_INTEREST : chr "02/29/2020" "03/01/2020" "03/02/2020" "03/03/2020" ...
## $ CASE_COUNT : int 1 0 0 2 5 3 8 7 21 58 ...
## $ HOSPITALIZED_COUNT : int 11 4 21 20 22 20 18 15 17 49 ...
## $ DEATH_COUNT : int 0 0 0 0 0 0 0 0 0 0 ...
## $ CASE_COUNT_7DAY_AVG: int NA NA NA NA NA NA NA 3 4 7 15 ...
## $ INCOMPLETE : int NA NA NA NA NA NA NA NA NA NA ...
##
## Dimensions of the data:
## [1] 173 6
##
## Column names in the data:
## [1] "DATE_OF_INTEREST" "CASE_COUNT" "HOSPITALIZED_COUNT"
## [4] "DEATH_COUNT" "CASE_COUNT_7DAY_AVG" "INCOMPLETE"
df$t <- c(1:dim(df)[1])
df$tsquared <- df$t^2
df$CUM_CASE_COUNT <- cumsum(df$CASE_COUNT)

m1a <- glm(CUM_CASE_COUNT ~ t + tsquared, family=poisson(link="identity"), start=c(1,1,1), data=df)
summary(m1a)

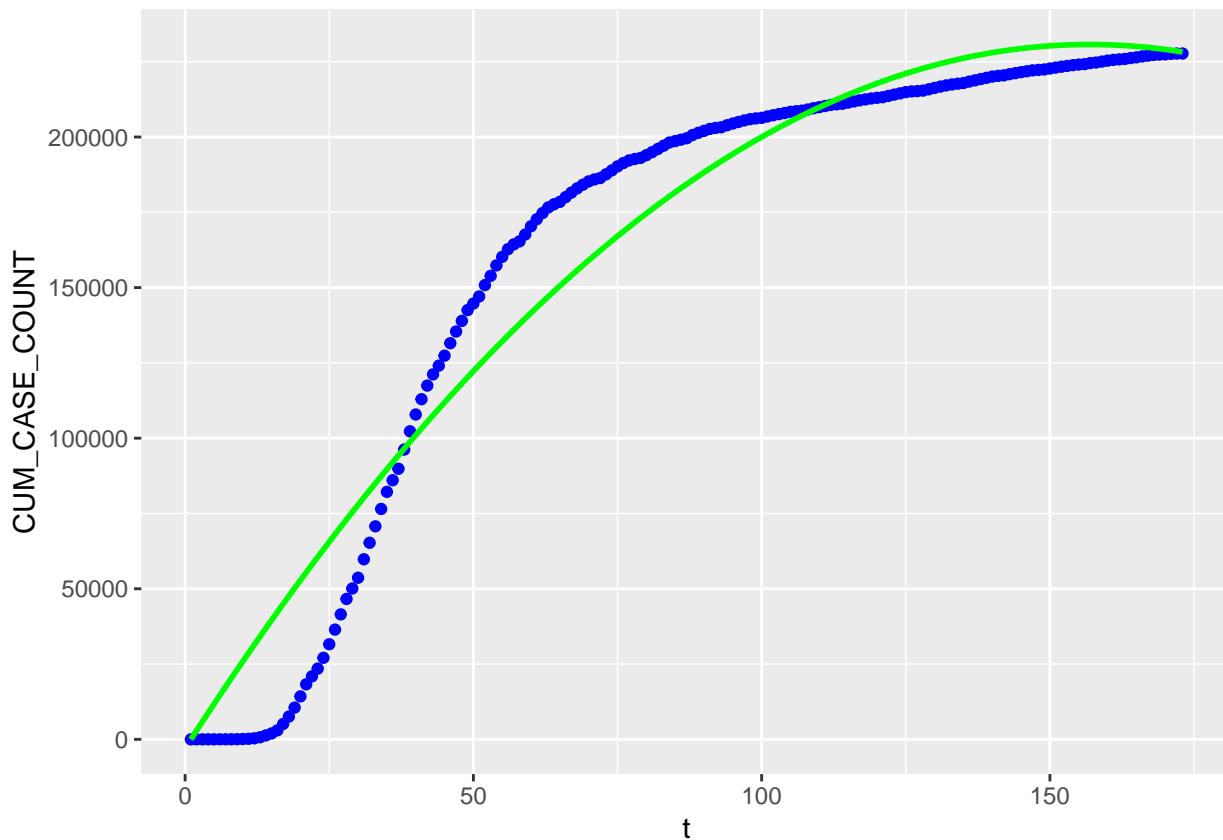
##
## Call:
## glm(formula = CUM_CASE_COUNT ~ t + tsquared, family = poisson(link = "identity"),
## data = df, start = c(1, 1, 1))
##
## Coefficients:
## Estimate Std. Error z value Pr(>|z|)

```

```
## (Intercept) -2.969e+03  1.179e+00  -2518  <2e-16 ***
## t           2.979e+03  1.180e+00   2525  <2e-16 ***
## tsquared    -9.496e+00  9.035e-03  -1051  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
## Null deviance: 9737173  on 172  degrees of freedom
## Residual deviance: 1222503  on 170  degrees of freedom
## AIC: 1224763
##
## Number of Fisher Scoring iterations: 25
```

```
ypredm1a <- predict(m1a)
```

```
ggplot(df, aes(x = t, y = CUM_CASE_COUNT)) +
  geom_point(color='blue') +
  geom_line(aes(x = t, y = ypredm1a), size = 1, color='green')
```

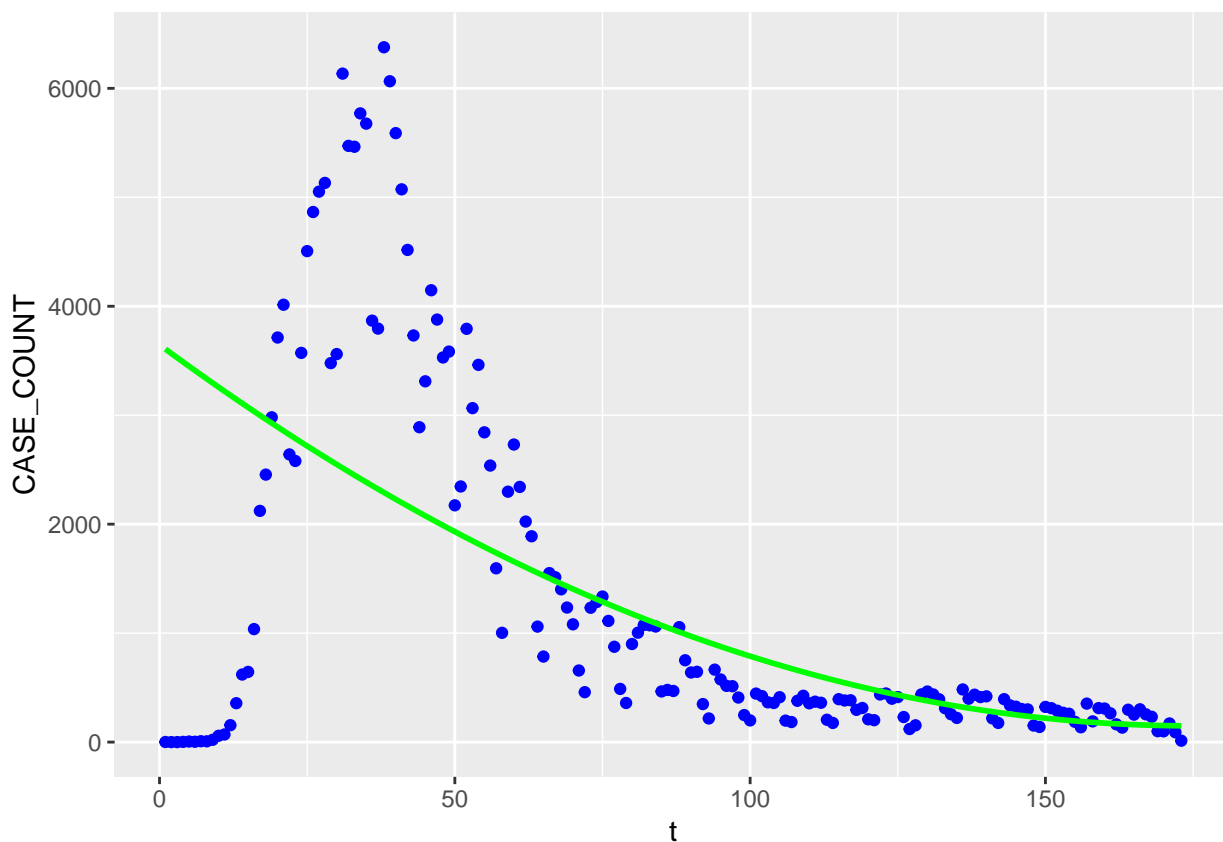


```
m1b <- glm(CASE_COUNT ~ t + tsquared, family=poisson(link="identity"), start=c(1,1,1), data=df)
summary(m1b)
```

```
##
## Call:
## glm(formula = CASE_COUNT ~ t + tsquared, family = poisson(link = "identity"),
## data = df, start = c(1, 1, 1))
##
```

```
## Coefficients:
##           Estimate Std. Error z value Pr(>|z|)
## (Intercept)  3.647e+03  1.086e+01   335.6  <2e-16 ***
## t           -4.004e+01  2.135e-01  -187.5  <2e-16 ***
## tsquared      1.146e-01  9.848e-04   116.4  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
## Null deviance: 292730  on 172  degrees of freedom
## Residual deviance: 159949  on 170  degrees of freedom
## AIC: 161352
##
## Number of Fisher Scoring iterations: 21
ypredm1b <- predict(m1b)
```

```
ggplot(df, aes(x = t, y = CASE_COUNT)) +
  geom_point(color='blue') +
  geom_line(aes(x = t, y = ypredm1b), size = 1, color='green')
```



```
shift <- function(x, n){
  c(rep(NA, n), x[seq(length(x) - n)])
}

for (k in c(1:7))
{
```



```

colname <- paste0('CASE_MINUS_', k)
df[colname]<- shift(df$CASE_COUNT, k)
}

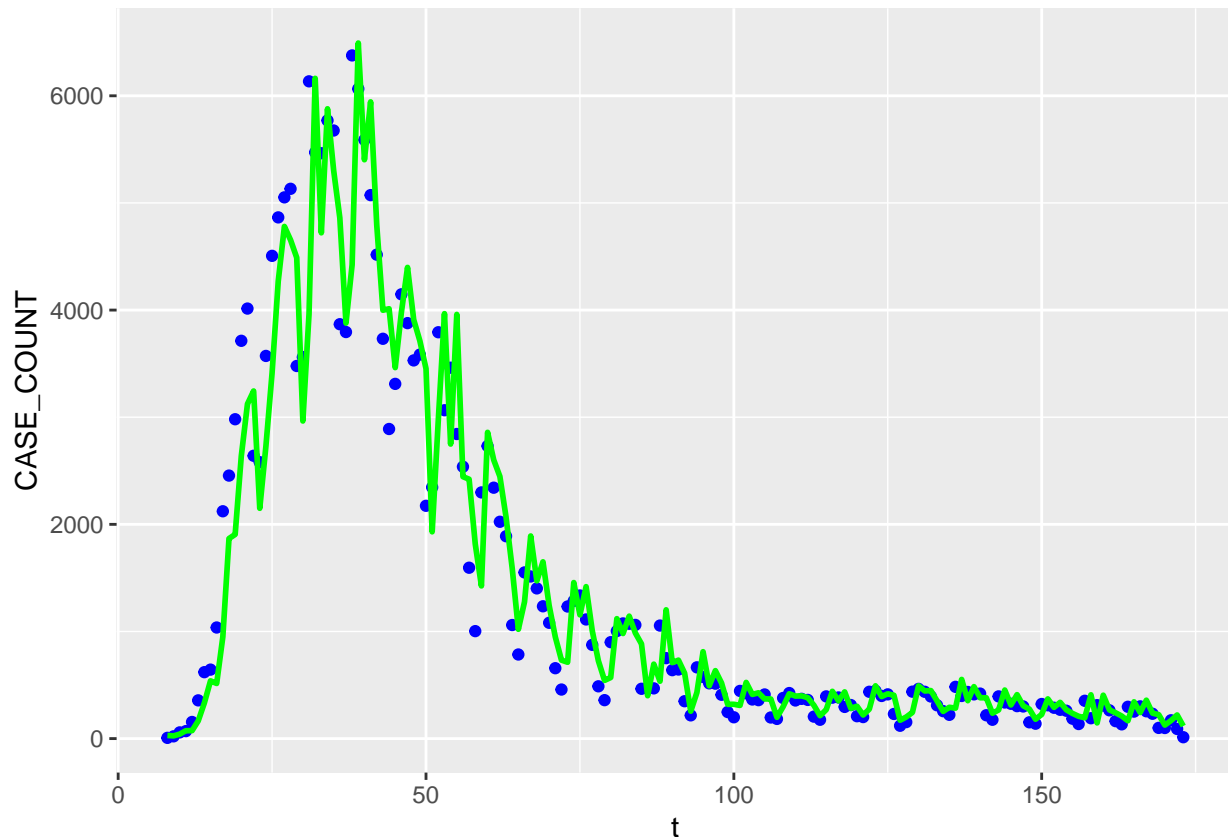
m2a <- glm(CASE_COUNT ~ CASE_MINUS_1 + CASE_MINUS_2 +
           CASE_MINUS_3 + CASE_MINUS_4 + CASE_MINUS_5 +
           CASE_MINUS_6 + CASE_MINUS_7, family=poisson(link="identity"),
           start=c(1,1,1,1,1,1,1), data=df[8:nrow(df),])
summary(m2a)

##
## Call:
## glm(formula = CASE_COUNT ~ CASE_MINUS_1 + CASE_MINUS_2 + CASE_MINUS_3 +
##      CASE_MINUS_4 + CASE_MINUS_5 + CASE_MINUS_6 + CASE_MINUS_7,
##      family = poisson(link = "identity"), data = df[8:nrow(df),
##      ], start = c(1, 1, 1, 1, 1, 1, 1))
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  21.445623   1.704655  12.581 < 2e-16 ***
## CASE_MINUS_1   1.007670   0.008443 119.344 < 2e-16 ***
## CASE_MINUS_2  -0.401347   0.011247 -35.684 < 2e-16 ***
## CASE_MINUS_3   0.282205   0.011742  24.034 < 2e-16 ***
## CASE_MINUS_4  -0.108602   0.011761  -9.234 < 2e-16 ***
## CASE_MINUS_5  -0.092772   0.010716  -8.658 < 2e-16 ***
## CASE_MINUS_6   0.251908   0.009984  25.231 < 2e-16 ***
## CASE_MINUS_7   0.046078   0.007575   6.083 1.18e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
##      Null deviance: 274172  on 165  degrees of freedom
## Residual deviance:  15619  on 158  degrees of freedom
## AIC: 17017
##
## Number of Fisher Scoring iterations: 9

ypredm2a <- predict(m2a)

ggplot(df[8:nrow(df),], aes(x = t, y = CASE_COUNT)) +
  geom_point(color='blue') +
  geom_line(aes(x = t, y = ypredm2a), size = 1, color='green')

```



```
m2a5 <- glm(CASE_COUNT ~ CASE_MINUS_1 + CASE_MINUS_2 +
CASE_MINUS_3 + CASE_MINUS_4 + CASE_MINUS_5, family=poisson(link="identity"),
start=c(1,1,1,1,1,1), data=df[6:nrow(df),])
```

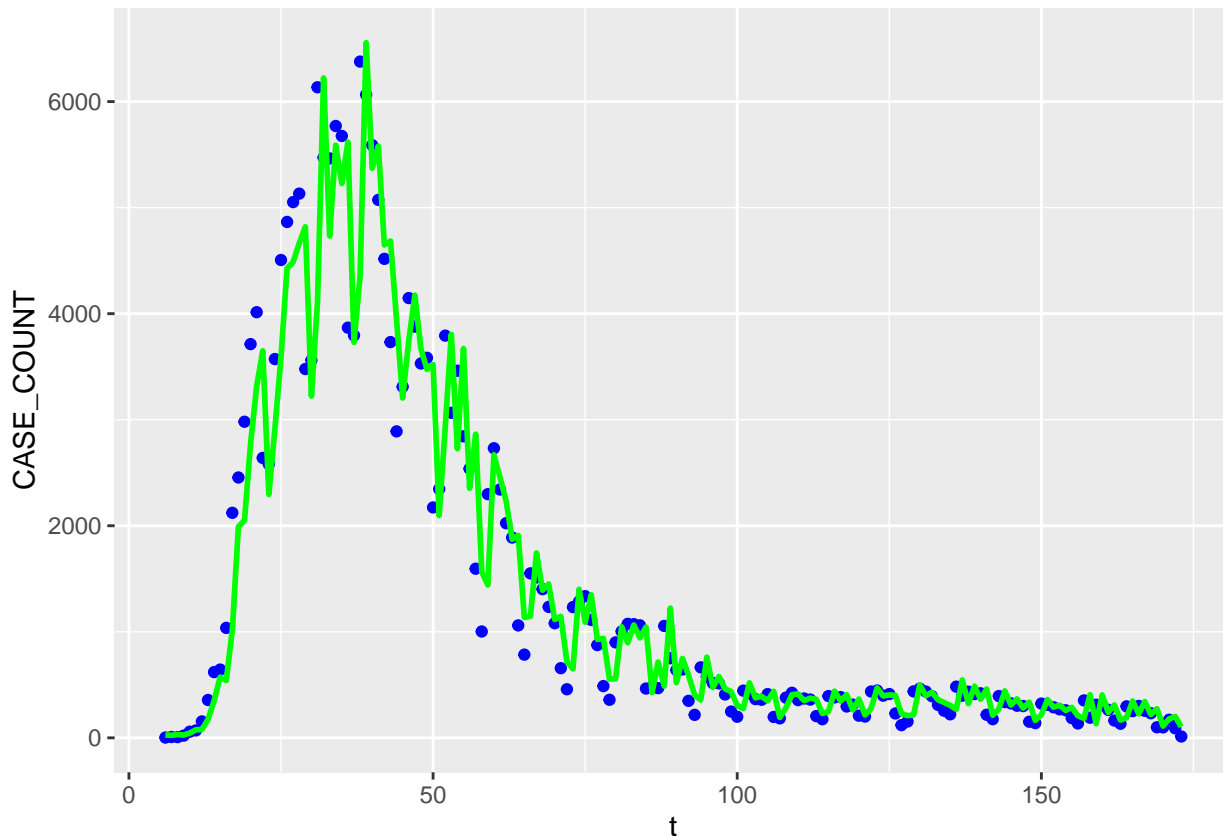
```
summary(m2a5)
```

```
##
## Call:
## glm(formula = CASE_COUNT ~ CASE_MINUS_1 + CASE_MINUS_2 + CASE_MINUS_3 +
## CASE_MINUS_4 + CASE_MINUS_5, family = poisson(link = "identity"),
## data = df[6:nrow(df), ], start = c(1, 1, 1, 1, 1, 1))
##
## Coefficients:
## Estimate Std. Error z value Pr(>|z|)
## (Intercept) 22.302889 1.543974 14.45 <2e-16 ***
## CASE_MINUS_1 1.056281 0.007836 134.80 <2e-16 ***
## CASE_MINUS_2 -0.441629 0.010965 -40.27 <2e-16 ***
## CASE_MINUS_3 0.354777 0.011369 31.20 <2e-16 ***
## CASE_MINUS_4 -0.182627 0.011309 -16.15 <2e-16 ***
## CASE_MINUS_5 0.197119 0.007736 25.48 <2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
## Null deviance: 279485 on 167 degrees of freedom
## Residual deviance: 16954 on 162 degrees of freedom
```

```
## AIC: 18355
##
## Number of Fisher Scoring iterations: 7
```

```
ypredm2a5 <- predict(m2a5)
```

```
ggplot(df[6:nrow(df),], aes(x = t, y = CASE_COUNT)) +
  geom_point(color='blue') +
  geom_line(aes(x = t, y = ypredm2a5), size = 1, color='green')
```



```
m2a3 <- glm(CASE_COUNT ~ CASE_MINUS_1 + CASE_MINUS_2 +
  CASE_MINUS_3, family=poisson(link="identity"),
  start=c(1,1,1,1), data=df[4:nrow(df),])
```

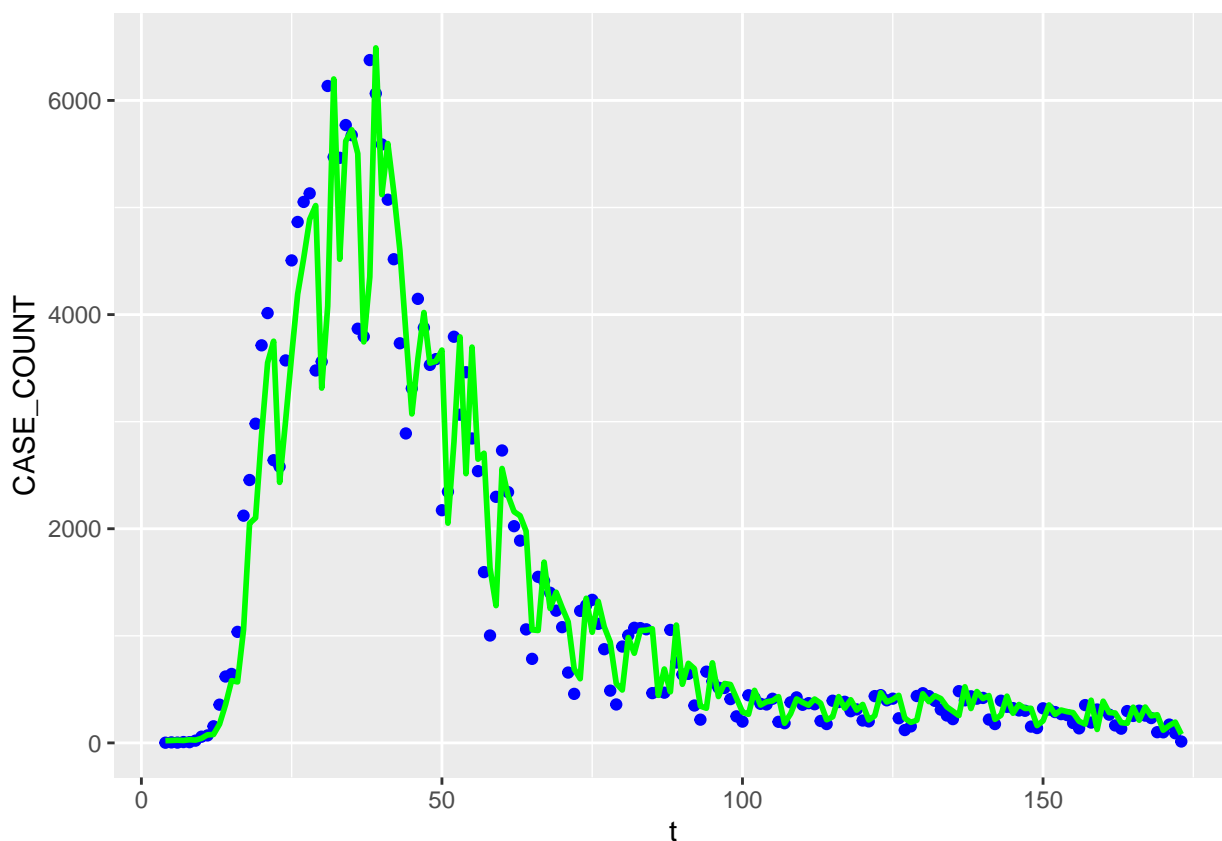
```
summary(m2a3)
```

```
##
## Call:
## glm(formula = CASE_COUNT ~ CASE_MINUS_1 + CASE_MINUS_2 + CASE_MINUS_3,
##      family = poisson(link = "identity"), data = df[4:nrow(df),
##      ], start = c(1, 1, 1, 1))
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  20.673739   1.375165   15.03  <2e-16 ***
## CASE_MINUS_1   1.049060   0.007662  136.92  <2e-16 ***
## CASE_MINUS_2  -0.394642   0.010492  -37.61  <2e-16 ***
## CASE_MINUS_3   0.330427   0.007205   45.86  <2e-16 ***
```

```
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
##      Null deviance: 284779  on 169  degrees of freedom
## Residual deviance:  17668  on 166  degrees of freedom
## AIC: 19071
##
## Number of Fisher Scoring iterations: 8
```

```
ypredm2a3 <- predict(m2a3)
```

```
ggplot(df[4:nrow(df),], aes(x = t, y = CASE_COUNT)) +
  geom_point(color='blue') +
  geom_line(aes(x = t, y = ypredm2a3), size = 1, color='green')
```



```
#install.packages("readxl")
```

```
library(readxl)
esti_asymptomatic <- read_csv('esti_asymptomatic.csv')
names(esti_asymptomatic)[names(esti_asymptomatic) == "P.Sev.1...age."] <- "P_Se_v_age"
esti_asymptomatic
```

##	Age	Male	Female	X. P_Se_v_age	Population_perc	PDF.asymp.
## 1	0-4	45106267	39934767	5.931232e-02	27%	1.590998e-02 3.162801e-02
## 2	5-10	46319087	40225712	6.036113e-02	27%	1.619132e-02 3.218728e-02
## 3	11-14	45044032	38648113	5.837153e-02	27%	1.565762e-02 3.112634e-02

```
## 4 15-19 44286173 38535391 5.776434e-02 51% 2.953473e-02 5.871312e-02
## 5 20-24 46956410 41620012 6.177809e-02 75% 4.660250e-02 9.264274e-02
## 6 25-29 54633535 49681080 7.275478e-02 75% 5.488280e-02 1.091034e-01
## 7 30-34 65544747 61670474 8.872693e-02 75% 6.693144e-02 1.330553e-01
## 8 35-39 49370823 46920301 6.715875e-02 75% 5.066142e-02 1.007116e-01
## 9 40-44 51155493 48766184 6.969090e-02 75% 5.257155e-02 1.045088e-01
## 10 45-49 62836913 60442592 8.598194e-02 48% 4.135601e-02 8.221305e-02
## 11 50-54 60946291 59710325 8.415259e-02 48% 4.047612e-02 8.046389e-02
## 12 55-59 47588180 46379016 6.553792e-02 48% 3.152274e-02 6.266516e-02
## 13 60-64 38607054 38368062 5.368670e-02 48% 2.582249e-02 5.133342e-02
## 14 65-69 35185041 36260434 4.983002e-02 13% 6.477903e-03 1.287765e-02
## 15 70-74 19746314 21631501 2.885918e-02 13% 3.751693e-03 7.458121e-03
## 16 75-79 11801212 13715744 1.779694e-02 13% 2.313602e-03 4.599289e-03
## 17 80-84 6808650 9097473 1.109381e-02 13% 1.442195e-03 2.866990e-03
## 18 85-89 2787103 4583542 5.140695e-03 13% 6.682904e-04 1.328518e-03
## 19 90-94 748881 1525990 1.586621e-03 13% 2.062607e-04 4.100328e-04
## 20 95-99 140322 386612 3.675129e-04 13% 4.777668e-05 9.497692e-05
## 21 100+ 11731 56108 4.731467e-05 13% 6.150907e-06 1.222760e-05
## CDF.asymp.
## 1 0.03162801
## 2 0.06381529
## 3 0.09494162
## 4 0.15365474
## 5 0.24629748
## 6 0.35540091
## 7 0.48845625
## 8 0.58916785
## 9 0.69367667
## 10 0.77588972
## 11 0.85635361
## 12 0.91901877
## 13 0.97035220
## 14 0.98322985
## 15 0.99068797
## 16 0.99528725
## 17 0.99815425
## 18 0.99948276
## 19 0.99989280
## 20 0.99998777
## 21 1.00000000
```

```
categorize_age <- function(age) {
  if (age %in% c("0-4", "5-10", "11-14")) return("0-17")
  if (age %in% c("15-19", "20-24", "25-29", "30-34", "35-39", "40-44")) return("18-44")
  if (age %in% c("45-49", "50-54", "55-59", "60-64")) return("45-64")
  if (age %in% c("65-69", "70-74")) return("65-74")
  return("75+")
}

esti_asymptomatic$Broad_Age_Group <- sapply(esti_asymptomatic$Age, categorize_age)

assign_severity <- function(broad_age) {
  switch(broad_age,
    "0-17" = 0.27,
```

```

    "18-44" = 0.75,
    "45-64" = 0.48,
    "65-74" = 0.13,
    "75+" = 0.13)
}

esti_asymptomatic$Assigned_Sev_Percentage <- sapply(esti_asymptomatic$Broad_Age_Group, assign_severity)

summary_data <- unique(esti_asymptomatic[, c("Broad_Age_Group", "Assigned_Sev_Percentage")])
colnames(summary_data) <- c("Age Group", "Sev 1 | age")

print(summary_data)

##      Age Group Sev 1 | age
## 1         0-17      0.27
## 4        18-44      0.75
## 10       45-64      0.48
## 14       65-74      0.13
## 16        75+      0.13

#Calculate the total population for each broad age group:
esti_asymptomatic$Total_Population <- esti_asymptomatic$Male + esti_asymptomatic$Female
age_group_population <- aggregate(Total_Population ~ Broad_Age_Group, data = esti_asymptomatic, FUN = sum)
total_population <- sum(esti_asymptomatic$Total_Population)

#Calculate the total asymptomatic individuals for each broad age group:
esti_asymptomatic$Asymptomatic_Count <- esti_asymptomatic$PDF.asymp. * esti_asymptomatic$Total_Population
age_group_asymptomatic <- aggregate(Asymptomatic_Count ~ Broad_Age_Group, data = esti_asymptomatic, FUN = sum)

#Identify which broad age group has the highest asymptomatic proportion:
age_group_asymptomatic$Asymptomatic_Proportion <- age_group_asymptomatic$Asymptomatic_Count / total_population
highest_asymptomatic_group <- age_group_asymptomatic[which.max(age_group_asymptomatic$Asymptomatic_Proportion)]

merged_data <- merge(age_group_population, age_group_asymptomatic, by = "Broad_Age_Group")

overall_data <- data.frame(
  Broad_Age_Group = "Overall",
  Total_Population = total_population,
  Asymptomatic_Count = sum(merged_data$Asymptomatic_Count),
  Asymptomatic_Proportion = sum(merged_data$Asymptomatic_Count) / total_population
)
highest_group_data <- data.frame(
  Broad_Age_Group = "Highest Asymptomatic Group",
  Total_Population = NA,
  Asymptomatic_Count = NA,
  Asymptomatic_Proportion = max(merged_data$Asymptomatic_Proportion)
)

final_data <- rbind(merged_data, overall_data, highest_group_data)

print(final_data)

##      Broad_Age_Group Total_Population Asymptomatic_Count
## 1         0-17      255277978      8080350.0
## 2        18-44      599140623      61516750.8
## 3        45-64      414878433      29683550.9
## 4        65-74      112823290      1228650.5

```

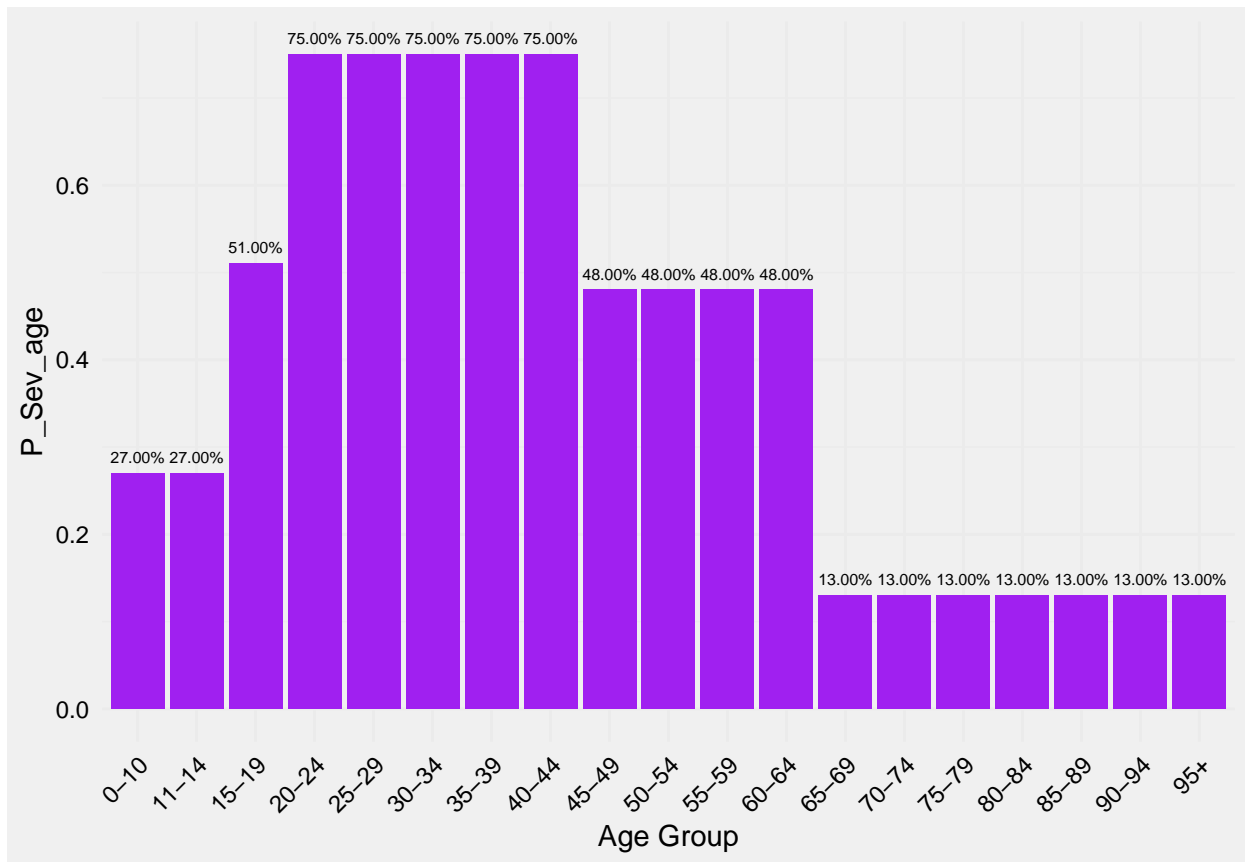
```
## 5          75+          51663368          173738.2
## 6          Overall      1433783692          100683040.4
## 7 Highest Asymptomatic Group          NA          NA
## Asymptomatic_Proportion
## 1          0.0056356828
## 2          0.0429051824
## 3          0.0207029492
## 4          0.0008569288
## 5          0.0001211746
## 6          0.0702219177
## 7          0.0429051824
```

```
library(ggplot2)
library(dplyr)
library(scales)
```

```
esti_asymptomatic$P_Sev_age <- as.numeric(gsub("%", "", esti_asymptomatic$P_Sev_age)) / 100
esti_asymptomatic <- esti_asymptomatic[!(esti_asymptomatic$Age %in% c("5-10", "100+")), ]
esti_asymptomatic$Age[esti_asymptomatic$Age == "0-4"] <- "0-10"
esti_asymptomatic$Age[esti_asymptomatic$Age == "95-99"] <- "95+"
```

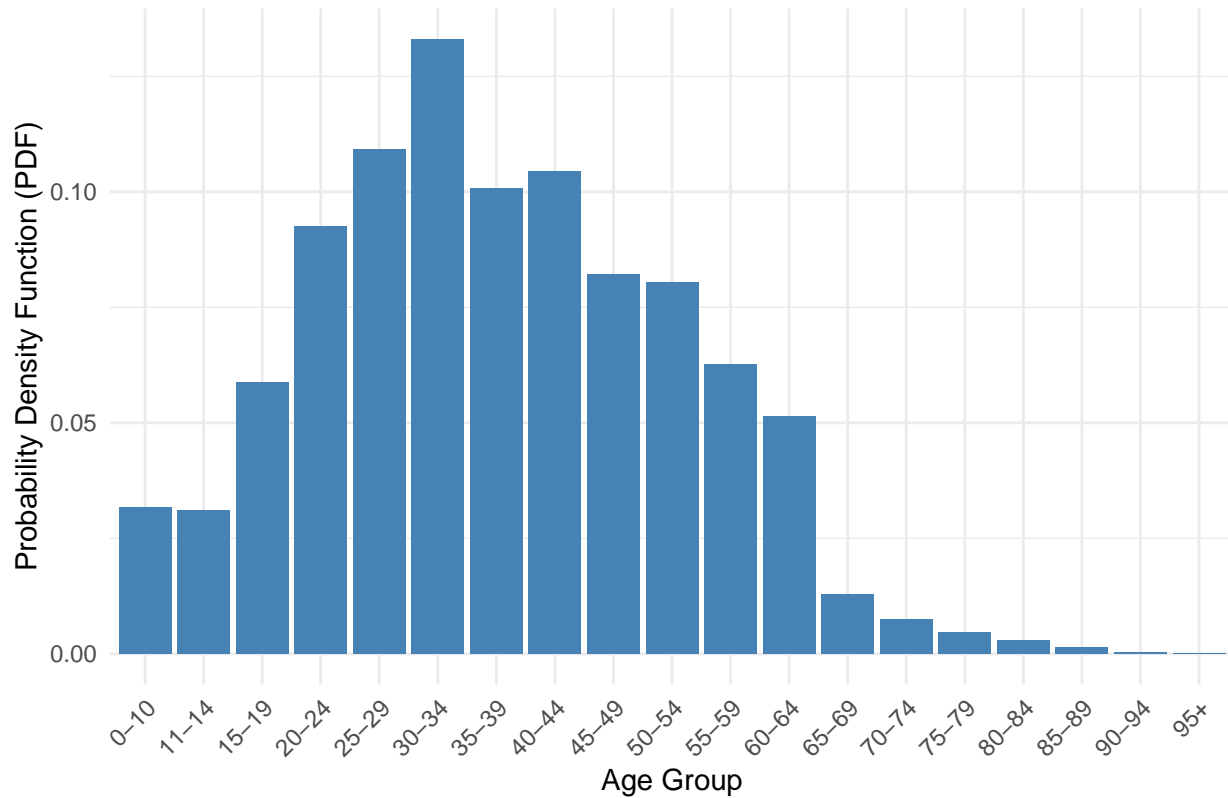
```
# Plotting the distribution of P_Sev_age
```

```
ggplot(data = esti_asymptomatic, aes(x = Age, y = P_Sev_age)) +
  geom_bar(stat = "identity", fill = "purple") +
  theme_minimal() +
  theme(
    plot.background = element_rect(fill = "#F0F0F0", color = NA), # Light purple background for the en
    axis.text.x = element_text(angle = 45, hjust = 1),
    axis.text = element_text(color = "black") # Ensuring text is visible on the light background
  ) +
  labs(
    #title = "Distribution of P_Sev_age across Age Groups",
    x = "Age Group",
    y = "P_Sev_age"
  ) +
  geom_text(
    aes(label = scales::percent(P_Sev_age, accuracy = 0.01)),
    vjust = -0.9,
    size = 2
  )
```

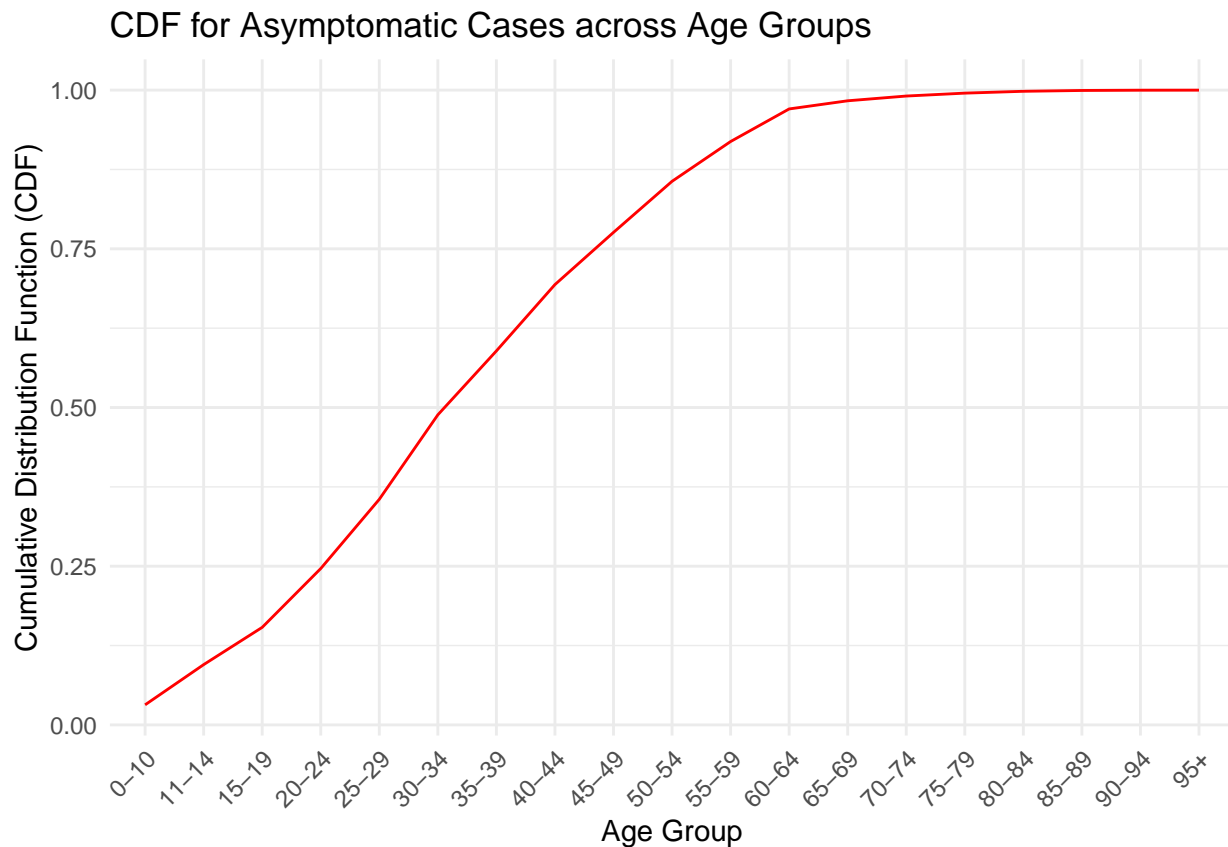


```
# Plotting PDF for asymptomatic cases
ggplot(data = esti_asymptomatic, aes(x = Age, y = PDF.asymp.)) +
  geom_bar(stat = "identity", fill = "steelblue") +
  theme_minimal() +
  labs(title = "PDF for Asymptomatic Cases across Age Groups",
       x = "Age Group",
       y = "Probability Density Function (PDF)") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```


PDF for Asymptomatic Cases across Age Groups



```
# Plotting CDF for asymptomatic cases
ggplot(data = esti_asymptomatic, aes(x = Age, y = CDF.asymp., group = 1)) +
  geom_line(color = "red") +
  theme_minimal() +
  labs(title = "CDF for Asymptomatic Cases across Age Groups",
       x = "Age Group",
       y = "Cumulative Distribution Function (CDF)") +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```



```
scaling_factor <- max(estimates$PDF.asymp.) / max(estimates$CDF.asymp.)

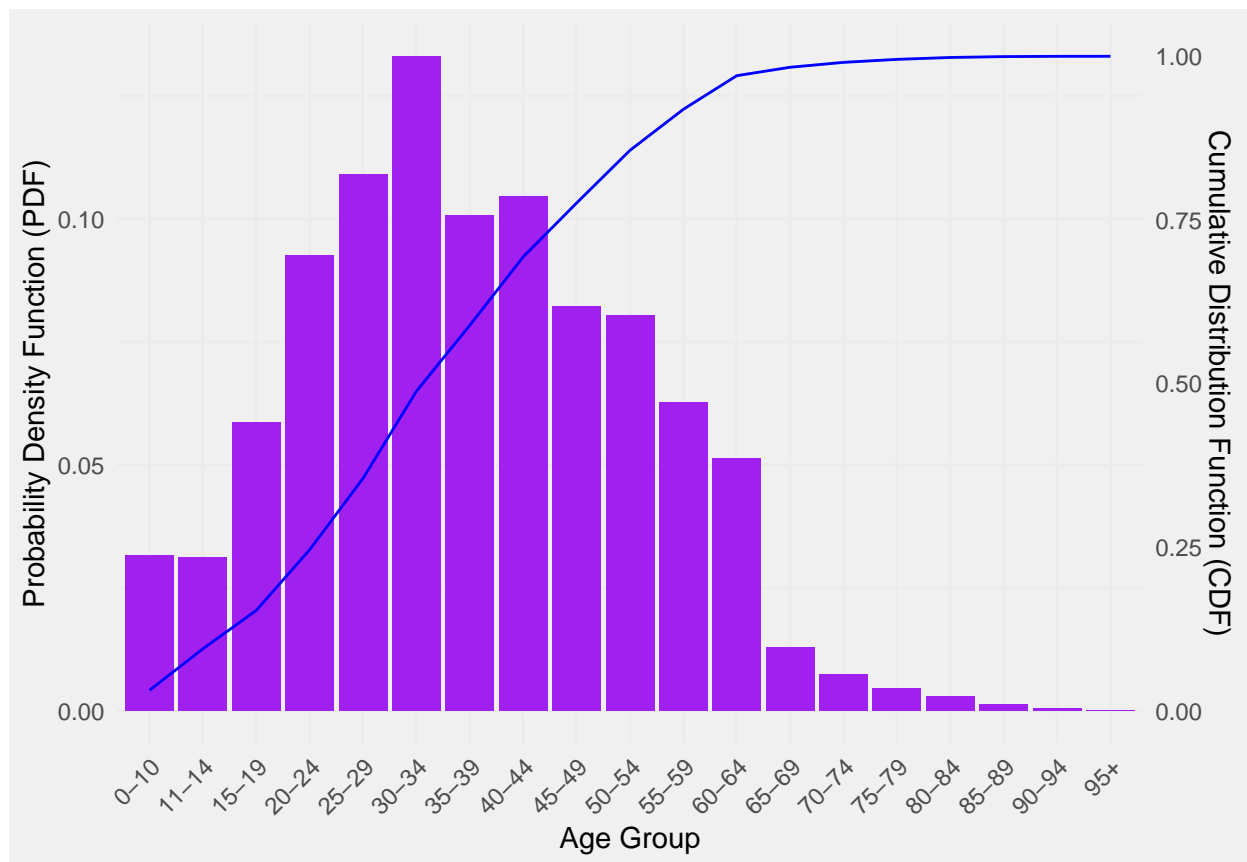
ggplot(data = estimates, aes(x = Age)) +
  # Plot PDF.asymp.
  geom_bar(aes(y = PDF.asymp.), stat = "identity", fill = "purple") +

  # Plot CDF.asymp. scaled by the factor
  geom_line(aes(y = CDF.asymp. * scaling_factor, group = 1), color = "blue") +

  labs(#title = "PDF & CDF for Asymptomatic Cases across Age Groups",
       x = "Age Group",
       y = "Probability Density Function (PDF)") +

  scale_y_continuous(sec.axis = sec_axis(~./scaling_factor, name = "Cumulative Distribution Function (CDF)")) +

  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1),
        plot.background = element_rect(fill = "#F0F0F0", color = NA),)
```



```

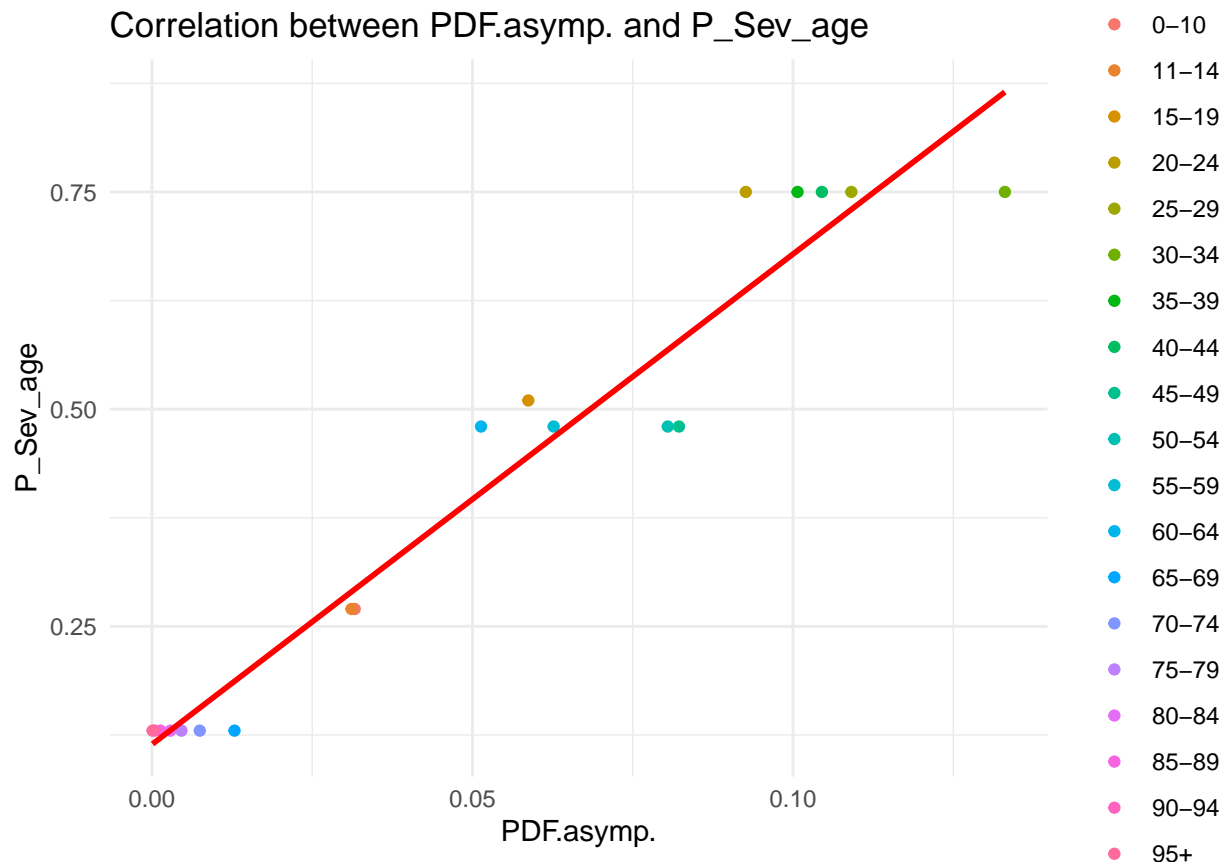
#Calculate the total probability of being asymptomatic in the entire population.
esti_asymptomatic$Weighted_PDF <- esti_asymptomatic$PDF.asymp. * esti_asymptomatic$Total_Population
total_probability_asymptomatic <- sum(esti_asymptomatic$Weighted_PDF) / total_population

#Determine the correlation between different variables, e.g., how PDF.asymp. correlates with P_Sev_age
correlation <- cor(esti_asymptomatic$PDF.asymp., esti_asymptomatic$P_Sev_age)
print(paste("The correlation between PDF.asymp. and P_Sev_age is:", round(correlation, 2)))

## [1] "The correlation between PDF.asymp. and P_Sev_age is: 0.97"

ggplot(data = esti_asymptomatic, aes(x = PDF.asymp., y = P_Sev_age)) +
  geom_point(aes(color = Age)) +
  geom_smooth(method = "lm", se = FALSE, color = "red") +
  theme_minimal() +
  labs(title = "Correlation between PDF.asymp. and P_Sev_age",
       x = "PDF.asymp.",
       y = "P_Sev_age")

```



```
library(ggplot2)
library(tidyr)

long_data <- gather(esti_asymptomatic, key = "Variable", value = "Value", -Age, -Broad_Age_Group, -Assi)

ggplot(long_data, aes(x = Age, y = Variable, fill = Value)) +
  geom_tile() +
  scale_fill_viridis_c() +
  labs(title = "Heatmap of Variables across Age Groups") +
  theme_minimal() +
  theme(axis.text.x = element_text(angle = 45, hjust = 1))
```

