

# **Exploratory Data Analysis with R**

## **Visualizing Geospatial Data**

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# Outline

- Introduction
- Displaying maps with R packages
- Displaying maps using shapefiles
- Plotting locations
- Plotting symbols
- Plotting choropleth maps
- You can disable/enable mouse click advance by pressing key 'k', when viewing the presentation.

# Introduction

- For analysis of spatial/geographical data, R can provide maps of data in the same computing environment that the data analysis is being performed in.
- The main place to go to get an overview of the kinds and capabilities of the **spatial packages** in R is the [Spatial Task Views on CRAN](http://cran.us.r-project.org/web/views/Spatial.html) <http://cran.us.r-project.org/web/views/Spatial.html>.
  - *R has been acquiring much of the functionality of traditional GIS packages such as `sp`, `rgeos`, `spData`, `shapefiles`, `maptools` and `raster`.*
- The maps can be *static* or *interactive*.
- Another classification of maps:
  - *Locations only*
  - *Choropleth map: coloring/sizing regions/locations by data values.*

# Introduction

- Packages for US maps:
  - *usmap, fiftystater, ...*
- Packages for static visualization:
  - *ggmap; maps, ...*
  - *mapdata provides additional map databases for the maps package*
- Packages for interactive visualization:
  - *leaflet, mapview, tmap, ...*
  - *tmap is an actively maintained open-source R-library for drawing thematic maps. The API is based on A Layered Grammar of Graphics and resembles the syntax of ggplot2.*
- Packages for reading and writing spatial data
  - *shapefiles, maptools, sf, ...*
- We use several R packages to illustrate how to draw static and interactive choropleth maps.

# Introduction - points

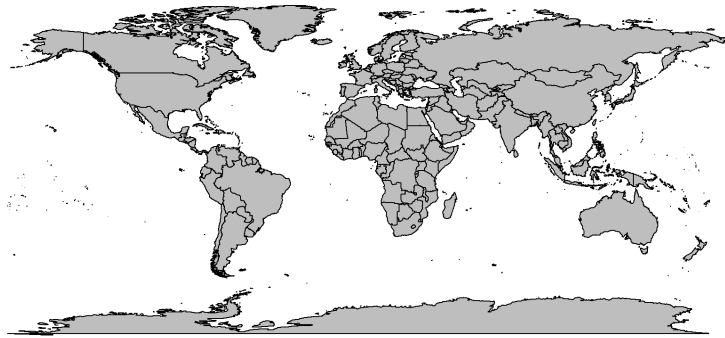
- Geographical points are on a sphere while maps are plotted on a flat surface.
- There are two different types of Coordinate Reference Systems in GIS (Geographic Information System ).
  - *Geographic coordinate systems: coordinate systems that span the entire globe (e.g. latitude / longitude).*
  - *Projected coordinate systems: coordinate systems that are localized to minimize visual distortion in a particular region (e.g. Robinson, **UTM**, State Plane)*
- For example,
  - *The package maps and fiftystater use geographic coordinate systems.*
  - *The package usmap uses UTM Coordinates.*
- For more information, see [Penn State E-education UTM Coordinate System](#).
- The function `convUL()` in the package `PBSmapping` can be used to convert coordinates between UTM and Lon/Lat.

# Introduction - choropleth map

- The **choropleth map** colors/sizes regions/points by data values.
  - *It's a widely used mapping method that a good proportion of people seem to understand, so it's often a good choice if you want to show or look at regional patterns.*
- The package `usmap` plot US map including Alaska and Hawaii. The function `usmap::plot_usmap` returns a `ggplot` object, which means we can add `ggplot` layers to the plot right out of the box.
- The package `tmap` offers a flexible, layer-based, and easy to use approach to create thematic maps. It resembles the syntax of `ggplot2`.
- The package `leaflet` <https://rstudio.github.io/leaflet/> is one of the most popular open-source JavaScript libraries for mobile-friendly interactive maps.

# Displaying maps: World map

```
library(maps)  
maps::map('world', fill=TRUE, border='black', col='gray', bg='white')
```



# Displaying maps: World map

- To map individual countries, we use the regions parameter.

```
par(mfrow=c(1,3));  
map("world", "italy");  
map("world", "france");  
map("world", "spain");
```





# Displaying maps: World map

- `map_data()` in `ggplot2` can easily turn data from the `maps` package in to a data frame suitable for plotting with `ggplot2`.
- See [ggplot2 themes](#) for more information about themes.

```
library(ggplot2)
worldmap = ggplot2::map_data('world')
str(worldmap)
```

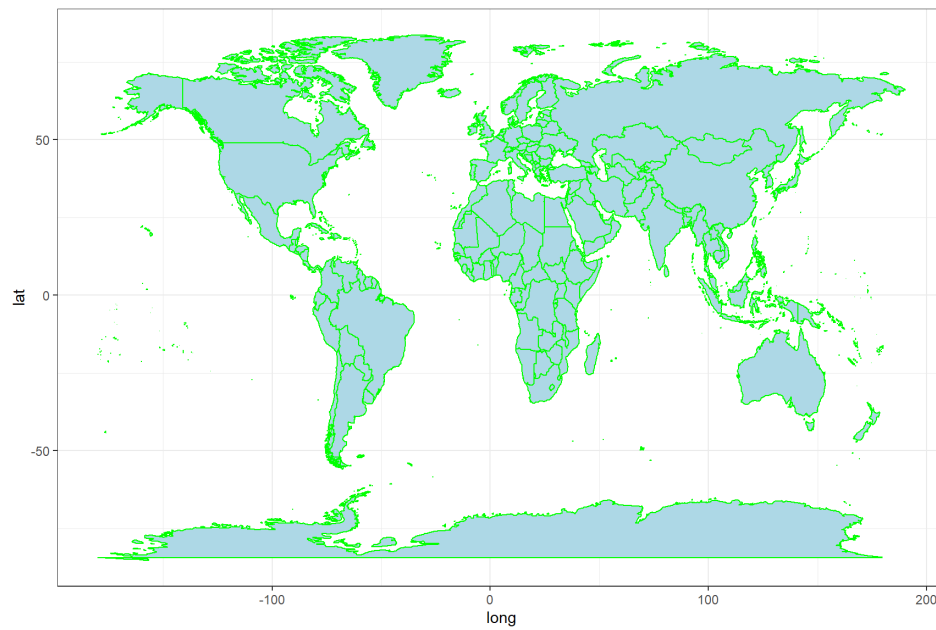
```
## 'data.frame':  99338 obs. of  6 variables:
## $ long      : num  -69.9 -69.9 -69.9 -70 -70.1 ...
## $ lat       : num   12.5 12.4 12.4 12.5 12.5 ...
## $ group     : num    1 1 1 1 1 1 1 1 1 1 ...
## $ order     : int    1 2 3 4 5 6 7 8 9 10 ...
## $ region    : chr   "Aruba" "Aruba" "Aruba" "Aruba" ...
## $ subregion: chr    NA NA NA NA ...
```

```
head(worldmap)
```

```
##      long      lat group order region subregion
## 1 -69.89912 12.45200     1     1  Aruba      <NA>
## 2 -69.89571 12.42300     1     2  Aruba      <NA>
## 3 -69.94219 12.43853     1     3  Aruba      <NA>
## 4 -70.00415 12.50049     1     4  Aruba      <NA>
## 5 -70.06612 12.54697     1     5  Aruba      <NA>
## 6 -70.05088 12.59707     1     6  Aruba      <NA>
```

# Displaying maps: World map

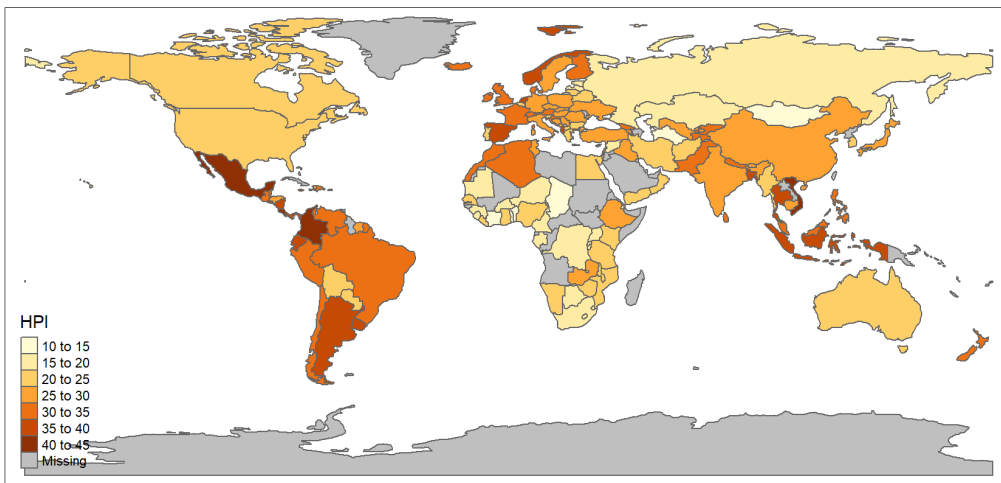
```
ggplot(worldmap, aes(long, lat, group=group)) +  
  geom_polygon(fill='lightblue', colour="green") + theme_bw()
```



# Displaying maps: World map

- Let's use the package `tmap` <https://cran.r-project.org/web/packages/tmap/vignettes/tmap-getstarted.html>

```
library(tmap)
data("World")
#str(World)
#str(World$HPI)
tm_shape(World) +tm_polygons("HPI")
```



*#It is a choropleth map*

# Displaying maps: World map

- Each map can be plotted as a static image or viewed interactively using “plot” and “view” modes, respectively.
- Toggling between the modes can be done with the ‘switch’ `ttm()` (which stands for toggle thematic map).

```
tmap_mode("view")  
tm_shape(World) + tm_polygons("HPI")
```



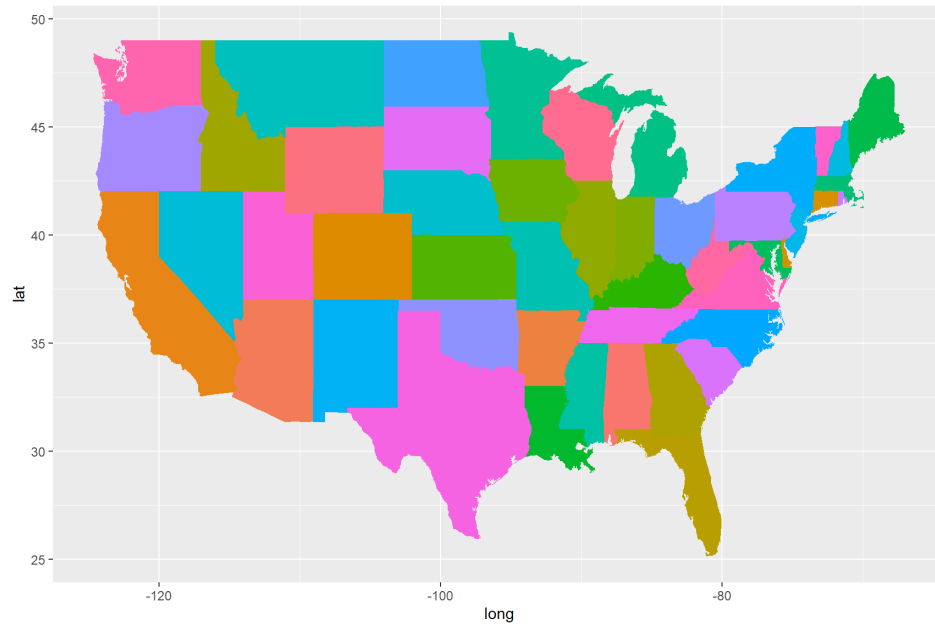
HPI

10 to 15
15 to 20
20 to 25
25 to 30
30 to 35
35 to 40
40 to 45
Missing

# Displaying maps: US map

- maps package can be used to plot US map.

```
usmap = ggplot2::map_data("state") # US map
ggplot(usmap, aes(long, lat, group=group, fill=region)) +
  geom_polygon(show.legend = F);
```



# Displaying maps: US map

- You have to know the names of the regions in the database though. To find that out, enter the following in the console.
- “fips” is for each state’s FIPS code. FIPS stands for “Federal Information Processing Standard”, and pretty much any geography-based data from the government uses it.

```
maps::state.fips # See state region names
```

```
##      fips ssa region division abb      polynome
## 1      1  1      3          6 AL      alabama
## 2      4  3      4          8 AZ      arizona
## 3      5  4      3          7 AR      arkansas
## 4      6  5      4          9 CA      california
## 5      8  6      4          8 CO      colorado
## 6      9  7      1          1 CT      connecticut
## 7     10  8      3          5 DE      delaware
## 8     11  9      3          5 DC      district of columbia
## 9     12 10      3          5 FL      florida
## 10     13 11      3          5 GA      georgia
## 11     16 13      4          8 ID      idaho
## 12     17 14      2          3 IL      illinois
## 13     18 15      2          3 IN      indiana
## 14     19 16      2          4 IA      iowa
## 15     20 17      2          4 KS      kansas
## 16     21 18      3          6 KY      kentucky
## 17     22 19      3          7 LA      louisiana
## 18     23 20      1          1 ME      maine
## 19     24 21      3          5 MD      maryland
## 20     25 22      1          1 MA      massachusetts:martha's vineyard
## 21     25 22      1          1 MA      massachusetts:main
## 22     25 22      1          1 MA      massachusetts:nantucket
## 23     26 23      2          3 MI      michigan:north
## 24     26 23      2          3 MI      michigan:south
## 25     27 24      2          4 MN      minnesota
## 26     28 25      3          6 MS      mississippi
## 27     29 26      2          4 MO      missouri
## 28     30 27      4          8 MT      montana
## 29     31 28      2          4 NE      nebraska
## 30     32 29      4          8 NV      nevada
## 31     33 30      1          1 NH      new hampshire
## 32     34 31      1          2 NJ      new jersey
## 33     35 32      4          8 NM      new mexico
## 34     36 33      1          2 NY      new york:manhattan
## 35     36 33      1          2 NY      new york:main
## 36     36 33      1          2 NY      new york:staten island
## 37     36 33      1          2 NY      new york:long island
## 38     37 34      3          5 NC      north carolina:knotts
## 39     37 34      3          5 NC      north carolina:main
## 40     37 34      3          5 NC      north carolina:spit
## 41     38 35      2          4 ND      north dakota
## 42     39 36      2          3 OH      ohio
## 43     40 37      3          7 OK      oklahoma
## 44     41 38      4          9 OR      oregon
## 45     42 39      1          2 PA      pennsylvania
## 46     44 41      1          1 RI      rhode island
## 47     45 42      3          5 SC      south carolina
## 48     46 43      2          4 SD      south dakota
```

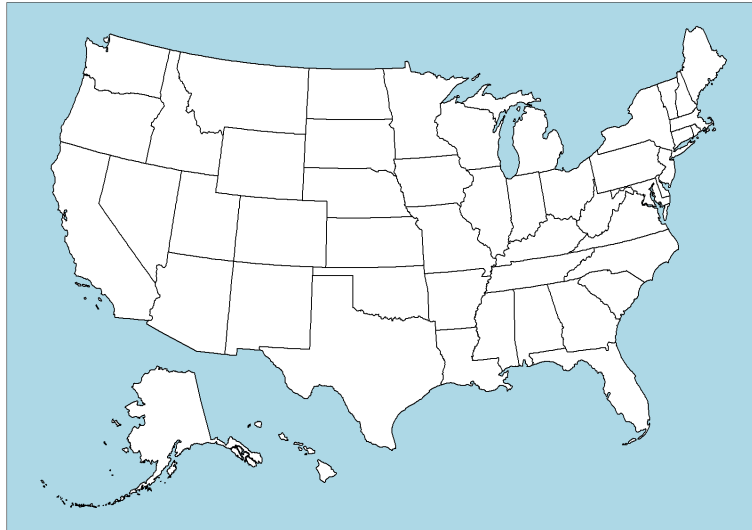
## 49	47	44	3	6	TN	tennessee
## 50	48	45	3	7	TX	texas
## 51	49	46	4	8	UT	utah
## 52	50	47	1	1	VT	vermont
## 53	51	49	3	5	VA	virginia:chesapeake
## 54	51	49	3	5	VA	virginia:chincoteague
## 55	51	49	3	5	VA	virginia:main
## 56	53	50	4	9	WA	washington:san juan island
## 57	53	50	4	9	WA	washington:lopez island
## 58	53	50	4	9	WA	washington:orcas island
## 59	53	50	4	9	WA	washington:whidbey island
## 60	53	50	4	9	WA	washington:main
## 61	54	51	3	5	WV	west virginia
## 62	55	52	2	3	WI	wisconsin
## 63	56	53	4	8	WY	wyoming

# Displaying maps: US map

- Package usmap includes Alaska and Hawaii

```
library(usmap)
usmap::plot_usmap(regions = "states", lines = "brown")+
  theme(panel.background = element_rect(colour = "black",
                                         fill = "lightblue"))
```

```
## Warning: Ignoring unknown parameters: lines
```



- The function `theme()` in `ggplot2` is used to modify a plot's theme.

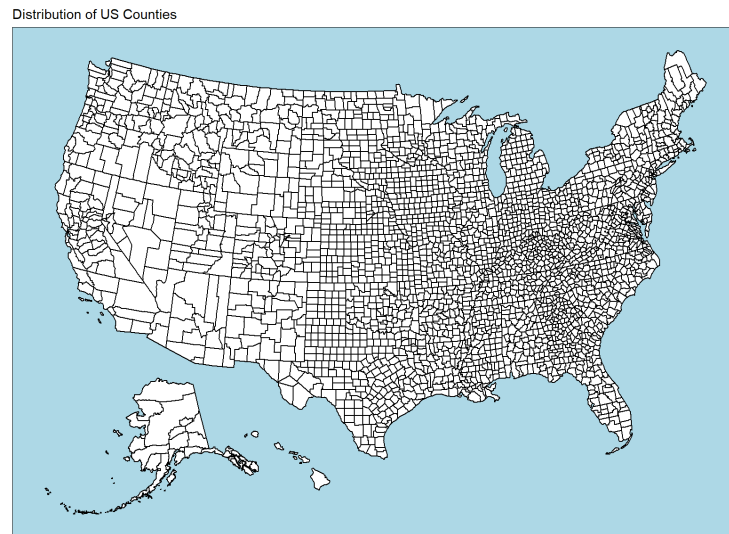


# Displaying maps: US map

- Package usmap includes Alaska and Hawaii

```
usmap::plot_usmap(regions = "counties", lines = "brown") +  
  labs(title = "Distribution of US Counties") +  
  theme(panel.background = element_rect(colour = "black",  
                                         fill = "lightblue"));
```

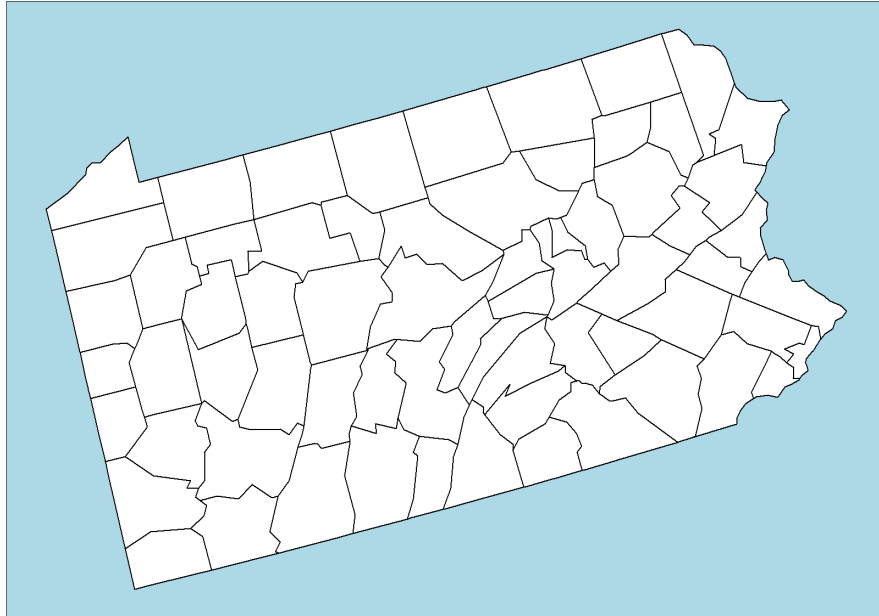
```
## Warning: Ignoring unknown parameters: lines
```



# Displaying maps: US map

- We can plot US map with specified states

```
usmap::plot_usmap(regions = "counties", include = c("PA")) +  
theme(panel.background = element_rect(colour = "black",  
fill = "lightblue"))
```



# Displaying maps: US map

- We can also specify multiple regions to map them all at once.

```
usmap::plot_usmap(include = c("CA", "ID", "NV", "OR", "WA")) +  
labs(title = "Western US States",  
      subtitle = "These are the states in the Pacific Timezone.")
```

Western US States  
These are the states in the Pacific Timezone.



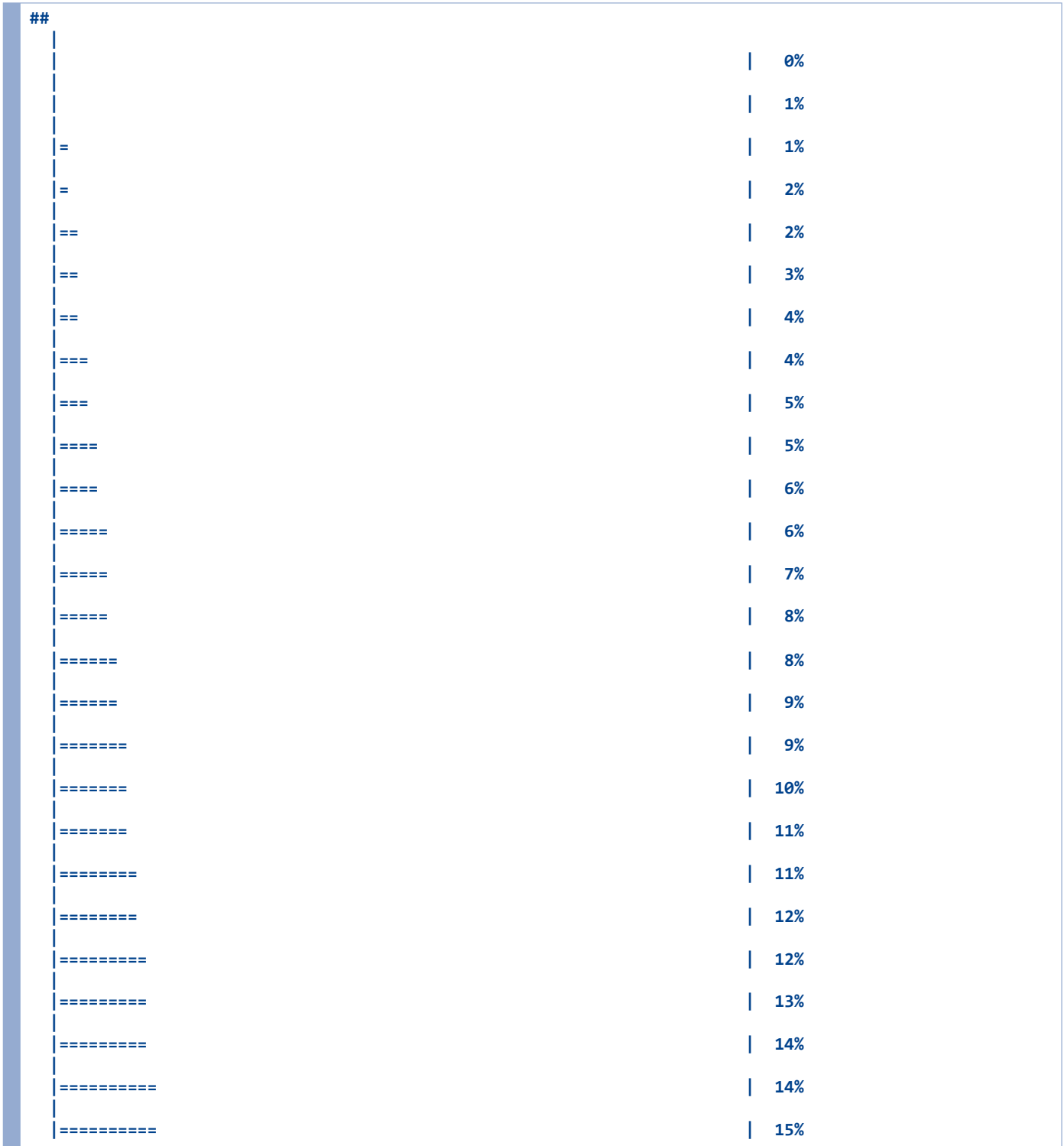
# Displaying maps using shapefiles

- The disadvantage of using R packages is that you're limited to the regions that the package offers.
- Using shapefiles provides more flexibility.
- Shapefiles are specifically for geographic data. The file format encodes points, lines, and polygons in geographic space and is a common way to distribute spatial data. The file extension is .shp.
- The `rgdal` package was very easy to use, but not recommended any more because it will be retired by the end of 2023.

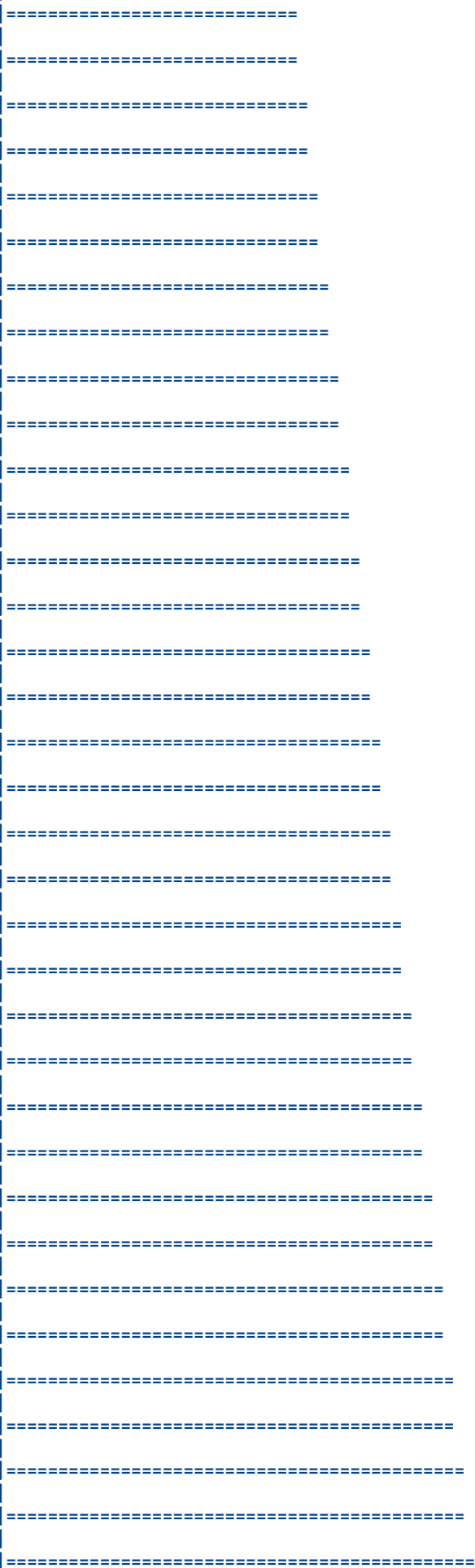
# Displaying maps using shapefiles

Using the tigris package, get Census Tiger shapefiles for US census geographies. Tigris will return the shapefile in the sf, or simple features, format. ([https://map-rfun.library.duke.edu/o31\\_thematic\\_mapping.html](https://map-rfun.library.duke.edu/o31_thematic_mapping.html))

```
library(tigris)
us_geo <- tigris::states(class = "sf", year=2020)
```



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# Displaying maps using shapefiles

```
library(dplyr)
class(us_geo)
```

```
## [1] "sf"          "data.frame"
```

```
dplyr::glimpse(us_geo)
```

```
## Rows: 56
## Columns: 15
## $ REGION    <chr> "3", "3", "2", "2", "3", "1", "4", "1", "3", "1", "1", "3", "..."
## $ DIVISION  <chr> "5", "5", "3", "4", "5", "1", "8", "1", "5", "1", "1", "5", "..."
## $ STATEFP   <chr> "54", "12", "17", "27", "24", "44", "16", "33", "37", "50", "..."
## $ STATENS   <chr> "01779805", "00294478", "01779784", "00662849", "01714934", "..."
## $ GEOID     <chr> "54", "12", "17", "27", "24", "44", "16", "33", "37", "50", "..."
## $ STUSPS    <chr> "WV", "FL", "IL", "MN", "MD", "RI", "ID", "NH", "NC", "VT", "..."
## $ NAME      <chr> "West Virginia", "Florida", "Illinois", "Minnesota", "Marylan..."
## $ LSAD      <chr> "00", "00", "00", "00", "00", "00", "00", "00", "00", "00", "..."
## $ MTFCC     <chr> "G4000", "G4000", "G4000", "G4000", "G4000", "G4000", "G4000", "..."
## $ FUNCSTAT  <chr> "A", "A", "A", "A", "A", "A", "A", "A", "A", "A", "A", "A", "..."
## $ ALAND     <dbl> 62266296765, 138958484319, 143778461053, 206232157570, 251518...
## $ AWATER    <dbl> 489206049, 45975808217, 6216594318, 18949864226, 6979171386, ...
## $ INTPTLAT  <chr> "+38.6472854", "+28.3989775", "+40.1028754", "+46.3159573", "..."
## $ INTPTLON  <chr> "-080.6183274", "-082.5143005", "-089.1526108", "-094.1996043..."
## $ geometry  <MULTIPOLYGON [°]> MULTIPOLYGON (((-81.74725 3..., MULTIPOLYGON (((...
```

# Displaying maps using shapefiles

- Data frame view:

```
as_tibble(us_geo)
```

```
## # A tibble: 56 × 15
##   REGION DIVISION STATEFP STATENS GEOID STUSPS NAME      LSAD MTFCC FUNCSTAT
##   <chr>  <chr>    <chr>  <chr>  <chr> <chr>  <chr>    <chr> <chr> <chr>
##  1 3      5      54    01779805 54    WV    West Virg... 00    G4000 A
##  2 3      5      12    00294478 12    FL    Florida      00    G4000 A
##  3 2      3      17    01779784 17    IL    Illinois     00    G4000 A
##  4 2      4      27    00662849 27    MN    Minnesota    00    G4000 A
##  5 3      5      24    01714934 24    MD    Maryland     00    G4000 A
##  6 1      1      44    01219835 44    RI    Rhode Isl... 00    G4000 A
##  7 4      8      16    01779783 16    ID    Idaho        00    G4000 A
##  8 1      1      33    01779794 33    NH    New Hamps... 00    G4000 A
##  9 3      5      37    01027616 37    NC    North Car... 00    G4000 A
## 10 1      1      50    01779802 50    VT    Vermont      00    G4000 A
## # ... with 46 more rows, and 5 more variables: ALAND <dbl>, AWATER <dbl>,
## #   INTPTLAT <chr>, INTPTLON <chr>, geometry <MULTIPOLYGON [°]>
```

# Displaying maps using shapefiles

- Quick Plot using the geometry data

```
plot(sf::st_geometry(us_geo))
```

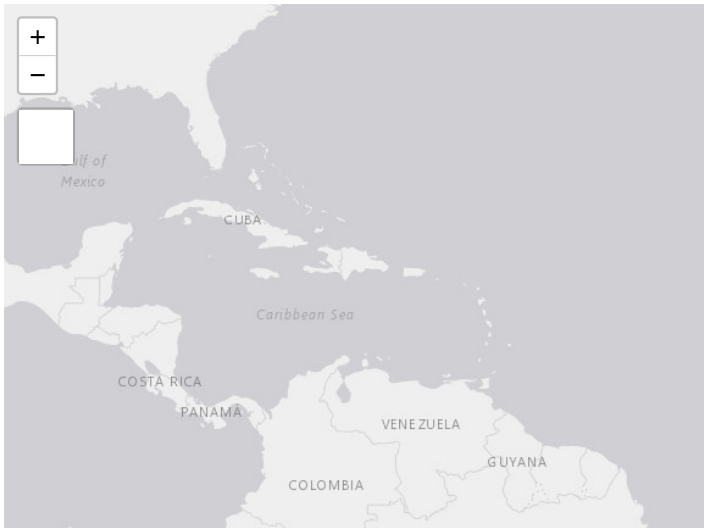


# Displaying maps using shapefiles

- `tigris::shift_geometry()` will shift and resize Alaska and Hawaii.

```
contiguous_states <- us_geo%>%  
  filter(REGION != 9)%>%  
  shift_geometry()
```

```
tm_shape(contiguous_states) + tm_borders()+  
  tm_fill("darkolivegreen3")
```



# Projections

- Longitude/latitude position points on a sphere; maps are drawn on a flat surface.
- Consider the maps package.
- The default uses a **rectangular projection** with the aspect ratio chosen so that longitude and latitude scales are equivalent at the center of the map.

```
maps::map("state");
```



# Projections

- Other projections try to preserve angles or areas.

```
library(mapproj)  
maps::map("state", project = "mercator");
```



# Projections

```
maps::map("state", project = "bonne", param = 45);
```





# Plotting locations

- Region boundaries essentially serve as your backdrop.
- For **point-based data**, you draw the map, and then you add things (locations and weighted locations) on top.
- We will use an **airports** data frame with eight variables from the package `nycflights13`.

```
library(nycflights13)
dim(airports)
```

```
## [1] 1458    8
```

```
head(airports)
```

```
## # A tibble: 6 × 8
##   faa   name                lat   lon   alt    tz dst  tzone
##   <chr> <chr>                <dbl> <dbl> <dbl> <dbl> <chr> <chr>
## 1 04G   Lansdowne Airport      41.1 -80.6  1044   -5 A   America/Ne...
## 2 06A   Moton Field Municipal Airport 32.5 -85.7   264   -6 A   America/Ch...
## 3 06C   Schaumburg Regional    42.0 -88.1   801   -6 A   America/Ch...
## 4 06N   Randall Airport        41.4 -74.4   523   -5 A   America/Ne...
## 5 09J   Jekyll Island Airport   31.1 -81.4    11   -5 A   America/Ne...
## 6 0A9   Elizabethton Municipal Airport 36.4 -82.2  1593   -5 A   America/Ne...
```

# Plotting locations with maps

## Approximate Centroids

- In mathematics, the centroid of a plane figure is the arithmetic **mean position** of all the points in the figure.

```
usmap = ggplot2::map_data("state") # US map
str(usmap)
```

```
## 'data.frame':   15537 obs. of  6 variables:
## $ long      : num  -87.5 -87.5 -87.5 -87.5 -87.6 ...
## $ lat       : num   30.4 30.4 30.4 30.3 30.3 ...
## $ group     : num    1 1 1 1 1 1 1 1 1 1 ...
## $ order     : int    1 2 3 4 5 6 7 8 9 10 ...
## $ region    : chr   "alabama" "alabama" "alabama" "alabama" ...
## $ subregion: chr    NA NA NA NA ...
```

# Plotting locations with maps

## Approximate Centroids

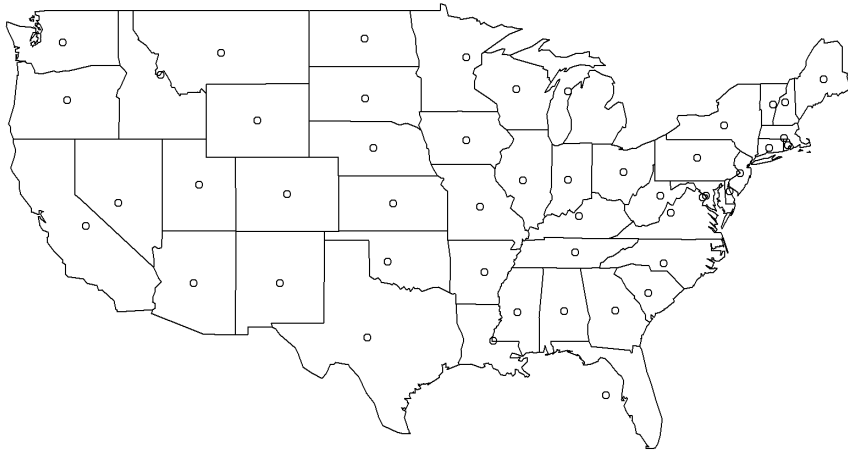
```
state_centroids = dplyr::summarize(group_by(usmap, region),  
  x = mean(range(long)), y = mean(range(lat)));  
names(state_centroids)[1] = "state"  
head(state_centroids)
```

```
## # A tibble: 6 × 3  
##   state      x      y  
##   <chr>    <dbl> <dbl>  
## 1 alabama  -86.7  32.6  
## 2 arizona  -112.  34.2  
## 3 arkansas -92.1  34.8  
## 4 california -119.  37.3  
## 5 colorado -106.  39.0  
## 6 connecticut -72.8  41.5
```

# Plotting locations with maps

- Adding points to a map drawn by the `maps::map()` function with the default projection can be added using the function `points()`, treating longitude as your x-coordinate and latitude as your y-coordinate.

```
maps::map("state");  
with(state_centroids, points(x, y));
```

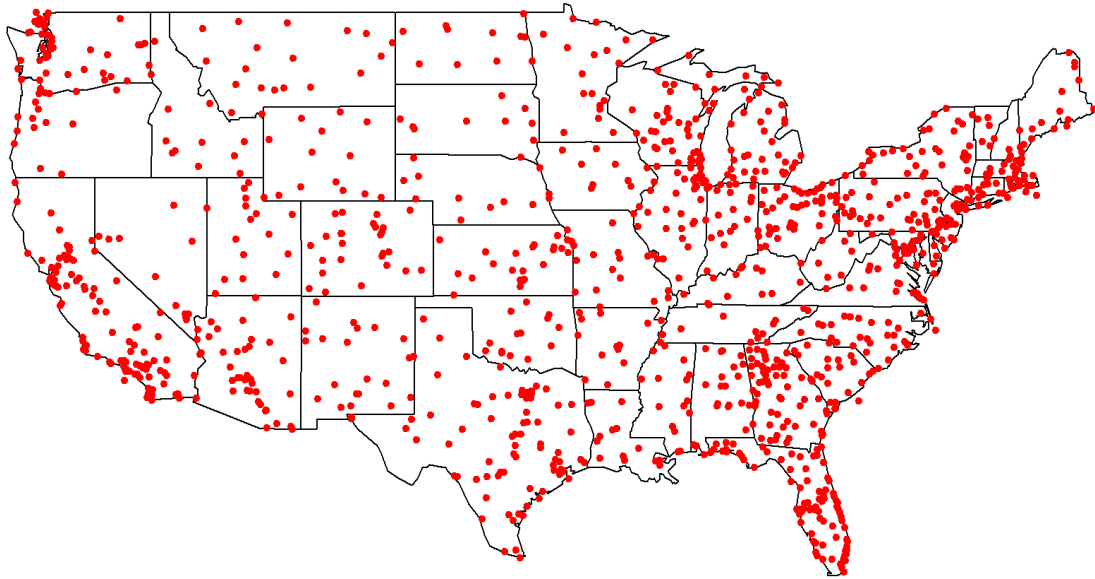


```
# with() function evaluates an R expression in an  
# environment constructed from data, original or modified  
# or: points(state_centroids$x, state_centroids$y)
```

# Plotting locations with maps

- Example: Plot all airports in the **airports** data.

```
maps::map("state");  
with(airports, points(lon, lat, pch=20, col='red'));
```



# Plotting locations with tmap

```
library(tigris)
us_geo <- tigris::states(class = "sf")
contiguous_states <- us_geo %>% filter(REGION != 9) %>% shift_geometry()
nrow(contiguous_states) #51 rows
```

- We first convert the data to an sf object using the function `st_as_sf`

```
library(sf)
nrow(airports) # 1458 rows
```

```
## [1] 1458
```

```
airports2=sf::st_as_sf(airports, coords = c("lon", "lat"), crs =4326 ) #Convert  
foreign object to an sf object
```

# Plotting locations with tmap

- Note that some airports are not in US

```
tm_shape(airports2) +  
tm_dots(size=0.01, col="red")
```



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- We can consider airports with tzone starting with America only

```
dplyr::filter(airports, grepl('America', tzone))  
#grepl returns a logical vector (match or not for each element).
```

# Plotting locations with tmap

```
airports_copy=dplyr::filter(airports, grepl('America', tzone))  
airports3=sf::st_as_sf(airports_copy, coords = c("lon", "lat"), crs =4326 )  
tm_shape(airports3) +  
  tm_dots(size=0.01, col="red")
```



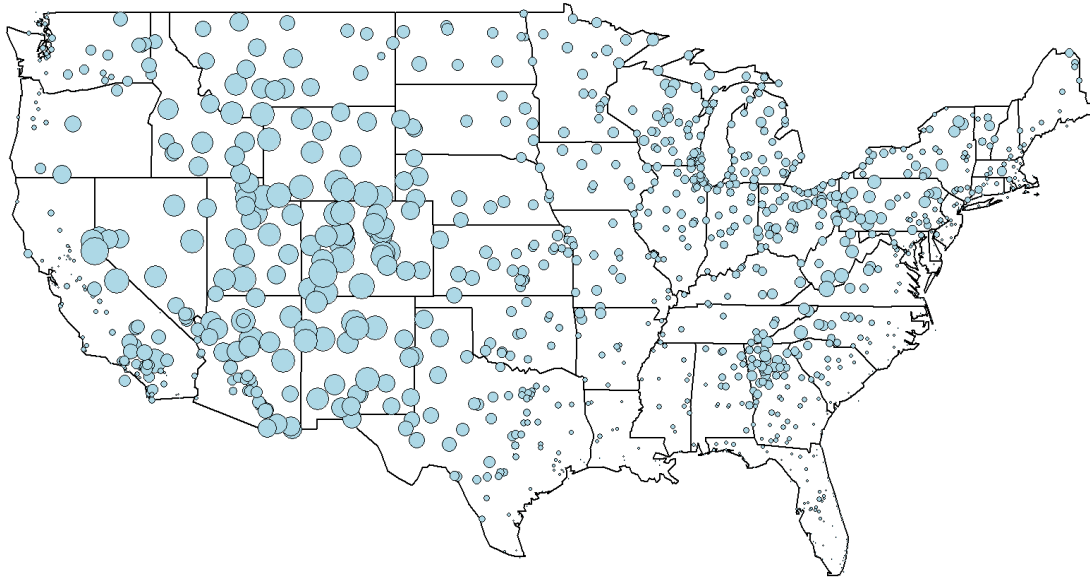


# Plotting symbols

- Points only represent location.
- Symbols (scaled shapes) both represent location and a second metric that corresponds to the location.
- For example, the function `symbols()` can be used to draw symbols.
- In the following example,
  - Set *add* to *TRUE*, so that the circles are added to the current map instead of drawing a new plot.
  - Set *inches* to *FALSE* so that you can size circles by the current coordinate system instead of by inches.
  - Set *bg* to whatever *color* you want.
  - *circles* is a vector giving the radii of the circles.
  - Circle size is set to the absolute value of *alt* (altitude) of each airport.
  - The *0.008* multiplier sizes all the circles to fit how you want on the screen.

# Plotting symbols

```
maps::map("state")  
with(airports, symbols(lon, lat, circles=.008*sqrt(abs(alt)),  
  add=TRUE, inches=FALSE, bg="lightblue", lwd=0.5))
```



# Plotting symbols with tmap

```
airports_copy$alt2=0.001*sqrt(abs(airports_copy$alt))  
airports4=sf::st_as_sf(airports_copy, coords = c("lon", "lat"), crs =4326 )  
tm_shape(airports4) +  
  tm_dots(size='alt2', col="red")
```



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*#Legend for symbol sizes not available in view mode*

# Plotting choropleth maps

## US Population Data:

- The [census bureau](#) provides estimates of populations of US counties.
- Estimates are available in several formats, including CSV.
- The CSV file is available at <https://www2.census.gov/programs-surveys/popest/datasets/2020-2021/counties/totals/>
- The data can be read from the web directly:

```
library(readr)
url="https://www2.census.gov/programs-surveys/popest/datasets/2020-
    2021/counties/totals/co-est2021-alldata.csv";
data = read_csv(url)
#read.csv(url, stringsAsFactors = FALSE)
```

# Plotting choropleth maps

- If you already downloaded the data, import from your local disk:

```
library(readr)
data = read_csv("../data/co-est2021-alldata.csv")
dim(data)
```

```
## [1] 3194 35
```

```
str(data)
```

```
## spec_tbl_df [3,194 × 35] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ SUMLEV      : chr [1:3194] "040" "050" "050" "050" ...
## $ REGION      : num [1:3194] 3 3 3 3 3 3 3 3 3 3 ...
## $ DIVISION     : num [1:3194] 6 6 6 6 6 6 6 6 6 6 ...
## $ STATE       : chr [1:3194] "01" "01" "01" "01" ...
## $ COUNTY      : chr [1:3194] "000" "001" "003" "005" ...
## $ STNAME      : chr [1:3194] "Alabama" "Alabama" "Alabama" "Alabama" ...
## $ CTYNAME     : chr [1:3194] "Alabama" "Autauga County" "Baldwin County" "Barbour County"
## ...
## $ ESTIMATESBASE2020 : num [1:3194] 5024279 58805 231767 25223 22293 ...
## $ POPESTIMATE2020  : num [1:3194] 5024803 58877 233140 25180 22223 ...
## $ POPESTIMATE2021  : num [1:3194] 5039877 59095 239294 24964 22477 ...
## $ NPOPCHG2020      : num [1:3194] 524 72 1373 -43 -70 ...
## $ NPOPCHG2021      : num [1:3194] 15074 218 6154 -216 254 ...
## $ BIRTHS2020       : num [1:3194] 13410 143 527 64 62 ...
## $ BIRTHS2021       : num [1:3194] 56320 649 2260 274 226 ...
## $ DEATHS2020        : num [1:3194] 16148 168 661 109 90 ...
## $ DEATHS2021        : num [1:3194] 64868 681 2867 394 282 ...
## $ NATURALCHG2020    : num [1:3194] -2738 -25 -134 -45 -28 ...
## $ NATURALCHG2021    : num [1:3194] -8548 -32 -607 -120 -56 ...
## $ INTERNATIONALMIG2020 : num [1:3194] -2 0 -1 0 0 0 0 1 0 ...
## $ INTERNATIONALMIG2021 : num [1:3194] 1244 5 63 1 2 ...
## $ DOMESTICMIG2020    : num [1:3194] 3339 97 1516 3 -42 ...
## $ DOMESTICMIG2021    : num [1:3194] 22136 237 6780 -98 309 ...
## $ NETMIG2020         : num [1:3194] 3337 97 1515 3 -42 ...
## $ NETMIG2021         : num [1:3194] 23380 242 6843 -97 311 ...
## $ RESIDUAL2020       : num [1:3194] -75 0 -8 -1 0 -2 0 -2 -4 -3 ...
## $ RESIDUAL2021       : num [1:3194] 242 8 -82 1 -1 9 2 0 24 6 ...
## $ GQESTIMATESBASE2020 : num [1:3194] 114572 442 2177 2789 2062 ...
## $ GQESTIMATES2020    : num [1:3194] 114572 442 2177 2789 2062 ...
## $ GQESTIMATES2021    : num [1:3194] 114572 442 2177 2789 2062 ...
## $ RBIRTH2021         : num [1:3194] 11.19 11 9.57 10.93 10.11 ...
## $ RDEATH2021         : num [1:3194] 12.9 11.5 12.1 15.7 12.6 ...
## $ RNATURALCHG2021    : num [1:3194] -1.699 -0.543 -2.57 -4.786 -2.506 ...
## $ RINTERNATIONALMIG2021 : num [1:3194] 0.2472 0.0848 0.2667 0.0399 0.0895 ...
## $ RDOMESTICMIG2021   : num [1:3194] 4.4 4.02 28.7 -3.91 13.83 ...
## $ RNETMIG2021        : num [1:3194] 4.65 4.1 28.97 -3.87 13.91 ...
## - attr(*, "spec")=
## .. cols(
## ..   SUMLEV = col_character(),
## ..   REGION = col_double(),
## ..   DIVISION = col_double(),
## ..   STATE = col_character(),
## ..   COUNTY = col_character(),
## ..   STNAME = col_character(),
## ..   CTYNAME = col_character(),
## ..   ESTIMATESBASE2020 = col_double(),
## ..   POPESTIMATE2020 = col_double(),
## ..   POPESTIMATE2021 = col_double(),
## ..   NPOPCHG2020 = col_double(),
```

```
## .. NPOPCHG2021 = col_double(),
## .. BIRTHS2020 = col_double(),
## .. BIRTHS2021 = col_double(),
## .. DEATHS2020 = col_double(),
## .. DEATHS2021 = col_double(),
## .. NATURALCHG2020 = col_double(),
## .. NATURALCHG2021 = col_double(),
## .. INTERNATIONALMIG2020 = col_double(),
## .. INTERNATIONALMIG2021 = col_double(),
## .. DOMESTICMIG2020 = col_double(),
## .. DOMESTICMIG2021 = col_double(),
## .. NETMIG2020 = col_double(),
## .. NETMIG2021 = col_double(),
## .. RESIDUAL2020 = col_double(),
## .. RESIDUAL2021 = col_double(),
## .. GQESTIMATESBASE2020 = col_double(),
## .. GQESTIMATES2020 = col_double(),
## .. GQESTIMATES2021 = col_double(),
## .. RBIRTH2021 = col_double(),
## .. RDEATH2021 = col_double(),
## .. RNATURALCHG2021 = col_double(),
## .. RINTERNATIONALMIG2021 = col_double(),
## .. RDOMESTICMIG2021 = col_double(),
## .. RNETMIG2021 = col_double()
## .. )
## - attr(*, "problems")=<externalptr>
```

# Plotting choropleth maps

- We first consider the **state** population total data.

```
library(dplyr);  
datapop0=data[,1:17]; #select the first 17 variables  
datapop0=datapop0%>%filter(CTYNAME%in%STNAME);  
#choose the state pop total data  
dim(datapop0); # 52 by 17
```

```
## [1] 52 17
```

```
str(datapop0);
```

```
## tibble [52 × 17] (S3: tbl_df/tbl/data.frame)  
## $ SUMLEV      : chr [1:52] "040" "040" "040" "040" ...  
## $ REGION      : num [1:52] 3 4 4 3 4 4 1 3 3 3 ...  
## $ DIVISION     : num [1:52] 6 9 8 7 9 8 1 5 5 5 ...  
## $ STATE       : chr [1:52] "01" "02" "04" "05" ...  
## $ COUNTY      : chr [1:52] "000" "000" "000" "000" ...  
## $ STNAME      : chr [1:52] "Alabama" "Alaska" "Arizona" "Arkansas" ...  
## $ CTYNAME     : chr [1:52] "Alabama" "Alaska" "Arizona" "Arkansas" ...  
## $ ESTIMATESBASE2020 : num [1:52] 5024279 733391 7151502 3011524 39538223 ...  
## $ POPESTIMATE2020  : num [1:52] 5024803 732441 7177986 3012232 39499738 ...  
## $ POPESTIMATE2021  : num [1:52] 5039877 732673 7276316 3025891 39237836 ...  
## $ NPOPCHG2020     : num [1:52] 524 -950 26484 708 -38485 ...  
## $ NPOPCHG2021     : num [1:52] 15074 232 98330 13659 -261902 ...  
## $ BIRTHS2020      : num [1:52] 13410 2407 18036 8554 104957 ...  
## $ BIRTHS2021      : num [1:52] 56320 9280 76497 35021 424333 ...  
## $ DEATHS2020      : num [1:52] 16148 1384 18340 9449 81738 ...  
## $ DEATHS2021      : num [1:52] 64868 5641 75665 38257 332337 ...  
## $ NATURALCHG2020  : num [1:52] -2738 1023 -304 -895 23219 ...
```

# Plotting choropleth maps

- The data is wide in terms of years. We change the names of the variables and then convert the data to long form.

```
library(tidyr)
vars=c("POPESTIMATE2020", "POPESTIMATE2021")
datapop0%>%select(STATE,STNAME,vars)%>%
pivot_longer(cols=vars, names_to = "Year", values_to = "Pop") -> datapop
datapop
```

```
## # A tibble: 104 × 4
##   STATE STNAME      Year      Pop
##   <chr> <chr>    <chr>    <dbl>
## 1 01 Alabama POPESTIMATE2020 5024803
## 2 01 Alabama POPESTIMATE2021 5039877
## 3 02 Alaska POPESTIMATE2020 732441
## 4 02 Alaska POPESTIMATE2021 732673
## 5 04 Arizona POPESTIMATE2020 7177986
## 6 04 Arizona POPESTIMATE2021 7276316
## 7 05 Arkansas POPESTIMATE2020 3012232
## 8 05 Arkansas POPESTIMATE2021 3025891
## 9 06 California POPESTIMATE2020 39499738
## 10 06 California POPESTIMATE2021 39237836
## # ... with 94 more rows
```



# Plotting choropleth maps

```
library(stringr)
datapop$Year= str_sub(datapop$Year, start=-4, end=-1)
datapop=dplyr::rename(datapop, fips=STATE)
#Change the variable 'STATE' to fips
datapop
```

```
## # A tibble: 104 × 4
##   fips STNAME      Year      Pop
##   <chr> <chr>    <chr>    <dbl>
## 1 01 Alabama    2020    5024803
## 2 01 Alabama    2021    5039877
## 3 02 Alaska     2020     732441
## 4 02 Alaska     2021     732673
## 5 04 Arizona    2020    7177986
## 6 04 Arizona    2021    7276316
## 7 05 Arkansas   2020    3012232
## 8 05 Arkansas   2021    3025891
## 9 06 California 2020   39499738
## 10 06 California 2021   39237836
## # ... with 94 more rows
```

# Plotting choropleth maps

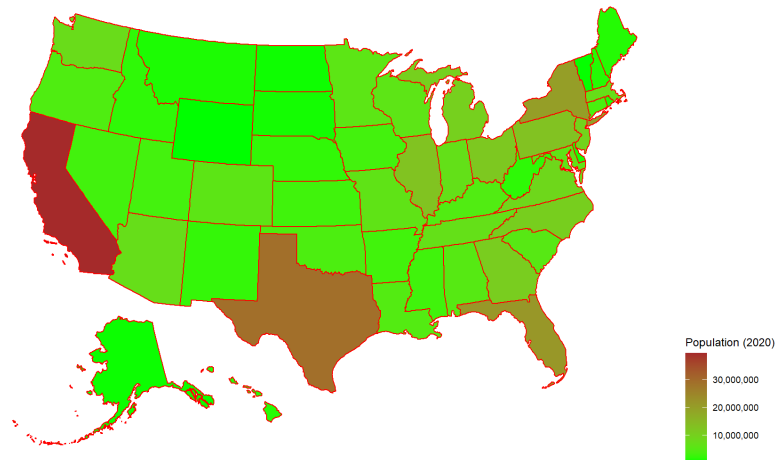
- A choropleth map needs to have the information for coloring all the pieces of a region. We use the Pop variable.
- First, we'll visualize the population data for the year 2020.

```
datapop_2020=dplyr::filter(datapop, Year == 2020)  
str(datapop_2020)
```

```
## tibble [52 × 4] (S3: tbl_df/tbl/data.frame)  
## $ fips : chr [1:52] "01" "02" "04" "05" ...  
## $ STNAME: chr [1:52] "Alabama" "Alaska" "Arizona" "Arkansas" ...  
## $ Year : chr [1:52] "2020" "2020" "2020" "2020" ...  
## $ Pop : num [1:52] 5024803 732441 7177986 3012232 39499738 ...
```

# Plotting choropleth maps

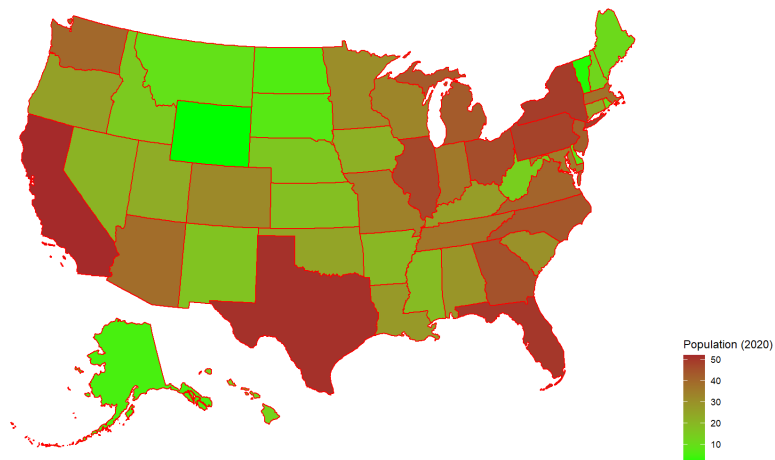
```
library(usmap)
library(ggplot2)
plot_usmap(data = datapop_2020, values = "Pop", color = "red") +
  scale_fill_continuous(low = "green", high = "brown",
    name = "Population (2020)", label = scales::comma)+
  theme(legend.position = "right")
```



# Plotting choropleth maps

- This image is dominated by the fact that most state populations are small.
- First, showing population **ranks** can help see the variation a bit better.

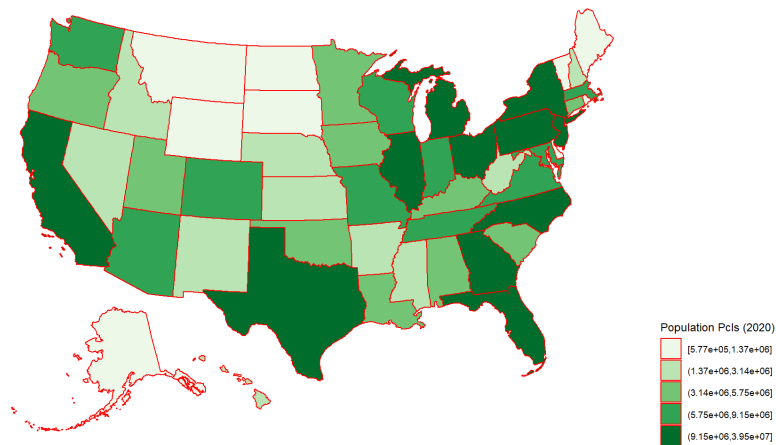
```
datapop_2020 = dplyr::mutate(datapop_2020, Rpop = rank(Pop))  
plot_usmap(data = datapop_2020, values = "Rpop", color = "red") +  
scale_fill_continuous(low = "green", high = "brown",  
  name = "Population (2020)", label = scales::comma)+  
theme(legend.position = "right")
```



# Plotting choropleth maps

- Second, using quantile bins instead of a continuous scale can help see the variation better.

```
bins=6 # consider 6 percentiles
datapop_2020 = dplyr::mutate(datapop_2020,
pcls = cut(Pop, quantile(Pop, seq(0, 1, len = bins)),
            include.lowest = TRUE));
plot_usmap(data = datapop_2020, values = "pcls", color = "red") +
scale_fill_brewer(palette = "Greens",
                  name = "Population Pcls (2020)") +
theme(legend.position = "right")
```

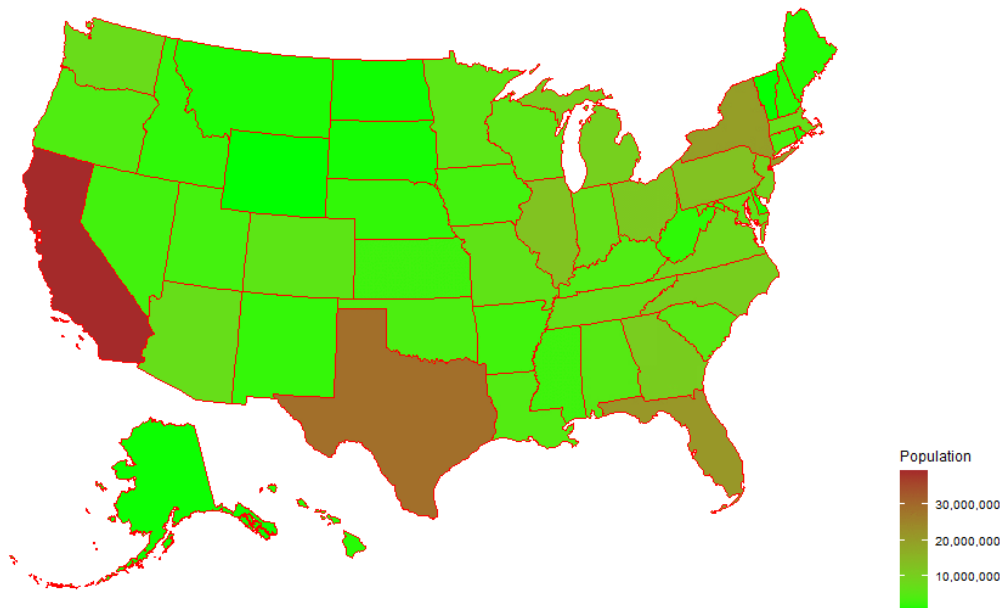


# Plotting choropleth maps

- Animated Maps by Year using gganimate
  - *You can update the values following the above two data sets*

```
library(gganimate)
statepop_ani=plot_usmap(data = datapop, values = "Pop", color = "red") +
scale_fill_continuous(low = "green", high = "brown",
name = "Population", label = scales::comma)+
transition_manual(Year, cumulative = FALSE)+
ggtitle(paste("Distribution of US Population in Year", '{current_frame}'))+
theme(legend.position = "right");
animate(statepop_ani,fps = 100)
```

Distribution of US Population in Year 2020



```
#anim_save("Popanimation.gif")
```

# Plotting choropleth maps

## Visualization of County Population Data

- First, remove the rows with state population total.

```
datapop1=data[,1:17]; #select the first 17 variables
datapop1 = datapop1%>% filter(!CTYNAME %in% STNAME);
dim(datapop1) #52 rows are removed
```

```
## [1] 3142 17
```

```
str(datapop1)
```

```
## tibble [3,142 × 17] (S3: tbl_df/tbl/data.frame)
## $ SUMLEV      : chr [1:3142] "050" "050" "050" "050" ...
## $ REGION      : num [1:3142] 3 3 3 3 3 3 3 3 3 ...
## $ DIVISION     : num [1:3142] 6 6 6 6 6 6 6 6 6 ...
## $ STATE       : chr [1:3142] "01" "01" "01" "01" ...
## $ COUNTY      : chr [1:3142] "001" "003" "005" "007" ...
## $ STNAME      : chr [1:3142] "Alabama" "Alabama" "Alabama" "Alabama" ...
## $ CTYNAME     : chr [1:3142] "Autauga County" "Baldwin County" "Barbour County" "Bibb County"
## ...
## $ ESTIMATESBASE2020: num [1:3142] 58805 231767 25223 22293 59134 ...
## $ POPESTIMATE2020  : num [1:3142] 58877 233140 25180 22223 59081 ...
## $ POPESTIMATE2021  : num [1:3142] 59095 239294 24964 22477 59041 ...
## $ NPOPCHG2020      : num [1:3142] 72 1373 -43 -70 -53 ...
## $ NPOPCHG2021      : num [1:3142] 218 6154 -216 254 -40 ...
## $ BIRTHS2020       : num [1:3142] 143 527 64 62 152 33 40 311 85 55 ...
## $ BIRTHS2021       : num [1:3142] 649 2260 274 226 629 130 214 1320 356 238 ...
## $ DEATHS2020       : num [1:3142] 168 661 109 90 220 37 58 409 121 114 ...
## $ DEATHS2021       : num [1:3142] 681 2867 394 282 820 ...
## $ NATURALCHG2020   : num [1:3142] -25 -134 -45 -28 -68 -4 -18 -98 -36 -59 ...
```

```
head(datapop1)
```

```
## # A tibble: 6 × 17
##   SUMLEV REGION DIVISION STATE COUNTY STNAME CTYNAME ESTIMATESBASE2020
##   <chr>   <dbl>   <dbl> <chr> <chr> <chr>   <chr>           <dbl>
## 1 050     3       6 01    001    Alabama Autauga County      58805
## 2 050     3       6 01    003    Alabama Baldwin County    231767
## 3 050     3       6 01    005    Alabama Barbour County    25223
## 4 050     3       6 01    007    Alabama Bibb County      22293
## 5 050     3       6 01    009    Alabama Blount County    59134
## 6 050     3       6 01    011    Alabama Bullock County   10357
## # ... with 9 more variables: POPESTIMATE2020 <dbl>, POPESTIMATE2021 <dbl>,
## #   NPOPCHG2020 <dbl>, NPOPCHG2021 <dbl>, BIRTHS2020 <dbl>, BIRTHS2021 <dbl>,
## #   DEATHS2020 <dbl>, DEATHS2021 <dbl>, NATURALCHG2020 <dbl>
```

# Plotting choropleth maps

## Visualization of County Population Data

- Again, the data is wide in terms of years. We change the names of the variables and then convert the data to long form.

```
library(tidyr)
vars=c("POPESTIMATE2020","POPESTIMATE2021")
datapop1%>%select(STATE,COUNTY,vars)%>%
pivot_longer(cols=vars, names_to = "Year", values_to = "Pop") -> CountyPop
head(CountyPop)
```

```
## # A tibble: 6 × 4
##   STATE COUNTY Year      Pop
##   <chr> <chr> <chr>   <dbl>
## 1 01     001 POPESTIMATE2020 58877
## 2 01     001 POPESTIMATE2021 59095
## 3 01     003 POPESTIMATE2020 233140
## 4 01     003 POPESTIMATE2021 239294
## 5 01     005 POPESTIMATE2020 25180
## 6 01     005 POPESTIMATE2021 24964
```

```
library(stringr)
CountyPop$Year= str_sub(CountyPop$Year, start=-4, end=-1)
head(CountyPop)
```

```
## # A tibble: 6 × 4
##   STATE COUNTY Year      Pop
##   <chr> <chr> <chr>   <dbl>
## 1 01     001 2020    58877
## 2 01     001 2021    59095
## 3 01     003 2020   233140
## 4 01     003 2021   239294
## 5 01     005 2020    25180
## 6 01     005 2021    24964
```



# Plotting choropleth maps

## Visualization of County Population Data

- To use the `usmap::plot_usmap()` function, we need the county fips information.

```
str(usmap::us_map(regions = "counties"))
```

```
## 'data.frame':   55211 obs. of  10 variables:
## $ x      : num  1225889 1244873 1244129 1272010 1276797 ...
## $ y      : num  -1275020 -1272331 -1267515 -1262889 -1295514 ...
## $ order  : int   1 2 3 4 5 6 7 8 9 10 ...
## $ hole   : logi  FALSE FALSE FALSE FALSE FALSE ...
## $ piece  : int   1 1 1 1 1 1 1 1 1 1 ...
## $ group  : chr   "01001.1" "01001.1" "01001.1" "01001.1" ...
## $ fips   : chr   "01001" "01001" "01001" "01001" ...
## $ abbr   : chr   "AL" "AL" "AL" "AL" ...
## $ full   : chr   "Alabama" "Alabama" "Alabama" "Alabama" ...
## $ county : chr   "Autauga County" "Autauga County" "Autauga County" "Autauga County" ...
```

# Plotting choropleth maps

## Visualization of County Population Data

- Next, add the variable 'fips' for county fips.
- The function `nchar()` counts the number of characters in a string.

```
CountyPop=mutate(CountyPop, fips=
case_when(
  nchar(COUNTY)==3 ~ paste(STATE, COUNTY, sep=""),
  nchar(COUNTY)==2 ~ paste(STATE, COUNTY, sep="0"),
  nchar(COUNTY)==1 ~ paste(STATE, COUNTY, sep="00")
)
); #this is more human readable compared to ifelse()
str(CountyPop)
```

```
## tibble [6,284 × 5] (S3: tbl_df/tbl/data.frame)
## $ STATE : chr [1:6284] "01" "01" "01" "01" ...
## $ COUNTY: chr [1:6284] "001" "001" "003" "003" ...
## $ Year : chr [1:6284] "2020" "2021" "2020" "2021" ...
## $ Pop : num [1:6284] 58877 59095 233140 239294 25180 ...
## $ fips : chr [1:6284] "01001" "01001" "01003" "01003" ...
```

```
head(CountyPop)
```

```
## # A tibble: 6 × 5
## STATE COUNTY Year Pop fips
## <chr> <chr> <chr> <dbl> <chr>
## 1 01 001 2020 58877 01001
## 2 01 001 2021 59095 01001
## 3 01 003 2020 233140 01003
## 4 01 003 2021 239294 01003
## 5 01 005 2020 25180 01005
## 6 01 005 2021 24964 01005
```

# Plotting choropleth maps

## Visualization of County Population Data

- Check if there are any missing values.

```
dim(CountyPop)
```

```
## [1] 6284    5
```

```
anyNA(CountyPop)
```

```
## [1] FALSE
```

```
any(complete.cases(CountyPop)==F)
```

```
## [1] FALSE
```

```
#check if there are missing values  
str(CountyPop)
```

```
## tibble [6,284 × 5] (S3: tbl_df/tbl/data.frame)  
## $ STATE : chr [1:6284] "01" "01" "01" "01" ...  
## $ COUNTY: chr [1:6284] "001" "001" "003" "003" ...  
## $ Year : chr [1:6284] "2020" "2021" "2020" "2021" ...  
## $ Pop : num [1:6284] 58877 59095 233140 239294 25180 ...  
## $ fips : chr [1:6284] "01001" "01001" "01003" "01003" ...
```

# Plotting choropleth maps

## Visualization of County Population Data

- Again, the fact that some counties with large population sizes will make other county population sizes very small. We consider quantile bins.
- Let's add a variable `pcls` to the data in each year and then row bind the two years' data

```
bins=6
CountyYears=list() #creat a list
Years=c(2020,2021)
for (i in 1:2)
{
  subdata=CountyPop%>%filter(Year==Years[i])%>%
    mutate( pcls = cut(Pop, quantile(Pop, seq(0, 1, len = bins)),
include.lowest = TRUE))
  CountyYears[[i]]=subdata
}
```

# Plotting choropleth maps

## Visualization of County Population Data

*##Then row combine the 2 data sets*

```
CountyPop2=dplyr::bind_rows(CountyYears[[1]], CountyYears[[2]])  
str(CountyPop2)
```

```
## tibble [6,284 × 6] (S3: tbl_df/tbl/data.frame)  
## $ STATE : chr [1:6284] "01" "01" "01" "01" ...  
## $ COUNTY: chr [1:6284] "001" "003" "005" "007" ...  
## $ Year : chr [1:6284] "2020" "2020" "2020" "2020" ...  
## $ Pop : num [1:6284] 58877 233140 25180 22223 59081 ...  
## $ fips : chr [1:6284] "01001" "01003" "01005" "01007" ...  
## $ pcls : Factor w/ 10 levels "[67,8.66e+03]",...: 4 5 3 3 4 2 3 5 3 3 ...
```

```
head(CountyPop2)
```

```
## # A tibble: 6 × 6  
## STATE COUNTY Year Pop fips pcls  
## <chr> <chr> <chr> <dbl> <chr> <fct>  
## 1 01 001 2020 58877 01001 (3.67e+04,9.53e+04]  
## 2 01 003 2020 233140 01003 (9.53e+04,9.99e+06]  
## 3 01 005 2020 25180 01005 (1.86e+04,3.67e+04]  
## 4 01 007 2020 22223 01007 (1.86e+04,3.67e+04]  
## 5 01 009 2020 59081 01009 (3.67e+04,9.53e+04]  
## 6 01 011 2020 10309 01011 (8.66e+03,1.86e+04]
```

# Plotting choropleth maps

## Visualization of County Population Data

- Now we plot the population data by Year.
  - *For this, we can define a function of year.*

```
PlotCountyPop= function(x)
{
  CountyPop_year=dplyr::filter(CountyPop2, Year == x);
  plot_usmap(regions="counties",data = CountyPop_year,
             values ="pcls") +
  ggplot2::scale_fill_brewer(palette = "Greens",
                             name = "Population Pcls")+
  ggtitle(paste("US County Population in Year",x))+
  theme(legend.position = "right");
}
```

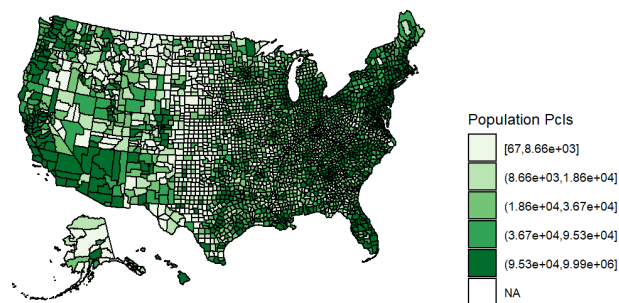
# Plotting choropleth maps

## Visualization of County Population Data

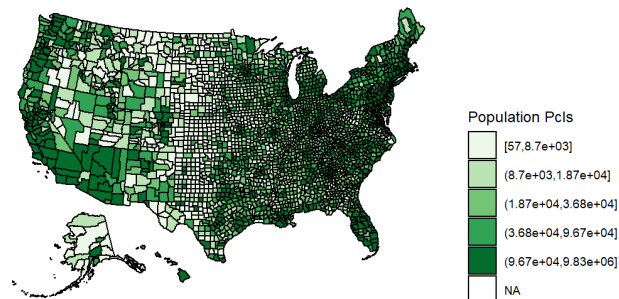
- Show two maps

```
PopCountyMaps=list();  
for( i in 1:length(Years)) #Length(Years)=2  
{  
  PopCountyMaps[[i]]=PlotCountyPop(i+2019);  
}  
gridExtra::grid.arrange(grobs = PopCountyMaps,nrow=2,ncol=1);
```

US County Population in Year 2020



US County Population in Year 2021



# Plotting choropleth maps

## Visualization of County Population Data

- Now let's plot the population distribution in certain states.

```
PA=c("PA"); #You can add more states
PACountyPop= function(x)
{
CountyPop_year=dplyr::filter(CountyPop2, Year == x);
plot_usmap(regions="counties",data = CountyPop_year,
            include = PA, values = "pcls", color= "grey") +
ggplot2::scale_fill_brewer(palette = "Greens",
                           name = "Population Pcls")+
ggtitle(paste("PA County Population in Year",x))+
theme(legend.position = "right");
}
```



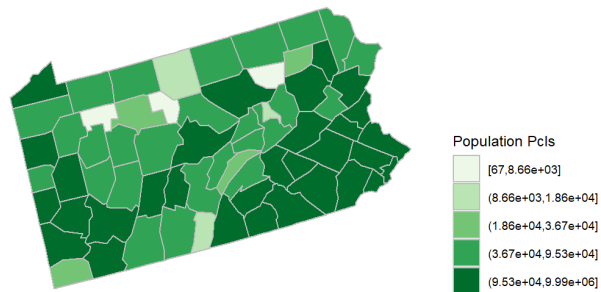
# Plotting choropleth maps

## Visualization of County Population Data

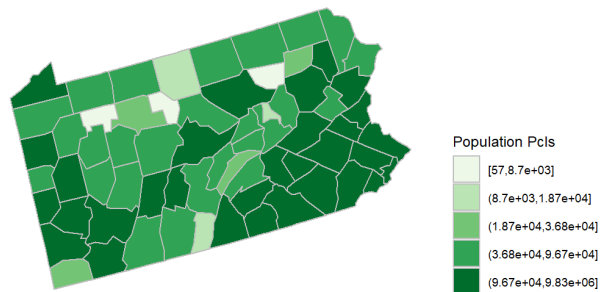
- Show the two maps

```
PACountyMaps=list();  
for( i in 1:length(Years))  
{  
  PACountyMaps[[i]]=PACountyPop(i+2019);  
}  
gridExtra::grid.arrange(grobs = PACountyMaps,nrow=2,ncol=1);
```

PA County Population in Year 2020



PA County Population in Year 2021



# Plotting choropleth maps

## Visualization of County Population Data

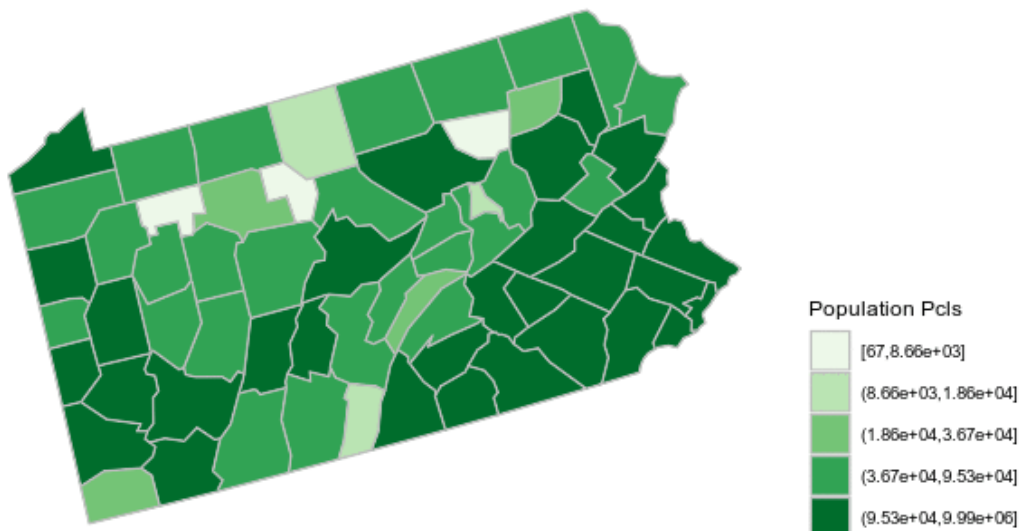
- Animated Maps using magick

```
library(magick);  
img=image_graph(width=600,height=400,res=96);  
for( i in Years) print(PACountyPop(i));  
dev.off();
```

```
## png  
## 2
```

```
PACountyanimation = image_animate(img, fps = 1);  
PACountyanimation
```

PA County Population in Year 2020



```
image_write(PACountyanimation, "PACountyanimation.gif")
```

- You can open the animated map PACountyanimation.gif in any browser.

# Plotting choropleth maps with tmap

- Let's use package tmap plot the above state population data datapop
  - For more, see [https://map-rfun.library.duke.edu/o31\\_thematic\\_mapping.html](https://map-rfun.library.duke.edu/o31_thematic_mapping.html)

```
str(datapop)
```

```
## tibble [104 × 4] (S3: tbl_df/tbl/data.frame)
## $ fips : chr [1:104] "01" "01" "02" "02" ...
## $ STNAME : chr [1:104] "Alabama" "Alabama" "Alaska" "Alaska" ...
## $ Year : chr [1:104] "2020" "2021" "2020" "2021" ...
## $ Pop : num [1:104] 5024803 5039877 732441 732673 7177986 ...
```

```
library(tigris)
us_geo <- tigris::states(class = "sf")
contiguous_states <- us_geo %>% filter(REGION != 9) %>% shift_geometry()
str(contiguous_states)
```

```
## Classes 'sf' and 'data.frame': 51 obs. of 15 variables:
## $ REGION : chr "3" "3" "2" "2" ...
## $ DIVISION : chr "5" "5" "3" "4" ...
## $ STATEFP : chr "54" "12" "17" "27" ...
## $ STATENS : chr "01779805" "00294478" "01779784" "00662849" ...
## $ GEOID : chr "54" "12" "17" "27" ...
## $ STUSPS : chr "WV" "FL" "IL" "MN" ...
## $ NAME : chr "West Virginia" "Florida" "Illinois" "Minnesota" ...
## $ LSAD : chr "00" "00" "00" "00" ...
## $ MTFCC : chr "G4000" "G4000" "G4000" "G4000" ...
## $ FUNCSTAT : chr "A" "A" "A" "A" ...
## $ ALAND : num 6.23e+10 1.39e+11 1.44e+11 2.06e+11 2.52e+10 ...
## $ AWATER : num 4.89e+08 4.60e+10 6.22e+09 1.89e+10 6.98e+09 ...
## $ INTPTLAT : chr "+38.6472854" "+28.3989775" "+40.1028754" "+46.3159573" ...
## $ INTPTLON : chr "-080.6183274" "-082.5143005" "-089.1526108" "-094.1996043" ...
## $ geometry:sfc_MULTIPOLYGON of length 51; first list element: List of 1
## ..$ :List of 1
## .. ..$ : num [1:35239, 1:2] 1216736 1216792 1216795 1216811 1216823 ...
## ..- attr(*, "class")= chr [1:3] "XY" "MULTIPOLYGON" "sfg"
## - attr(*, "sf_column")= chr "geometry"
## - attr(*, "agr")= Factor w/ 3 levels "constant","aggregate",... NA NA NA NA NA NA NA NA NA ...
## ..- attr(*, "names")= chr [1:14] "REGION" "DIVISION" "STATEFP" "STATENS" ...
```

- To use the tmap package, we need to merge the two data sets by the state names and convert to sf data type

```
datapop$NAME =datapop$STNAME
pop_2020=datapop%>%filter(Year==2020)
state_pop_2020=left_join(pop_2020, contiguous_states, by='NAME')
dim(state_pop_2020)
```

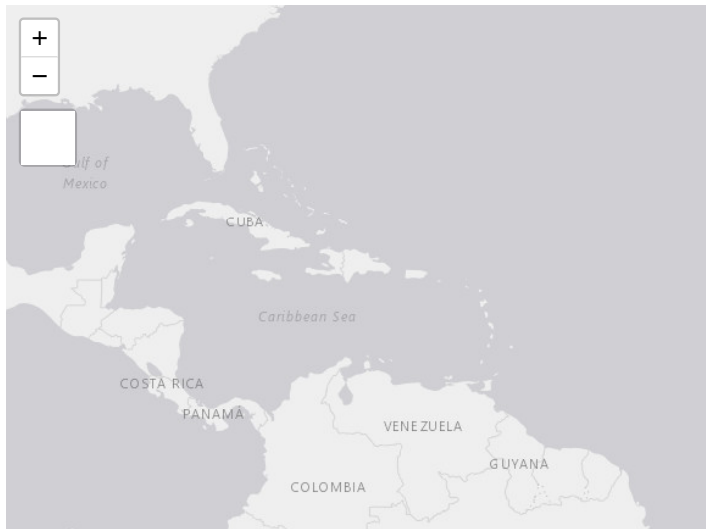
```
## [1] 52 19
```

```
state_pop_2020=st_sf(state_pop_2020)
```



# Plotting choropleth maps with tmap

```
tm_shape(state_pop_2020) + tm_polygons("Pop", id = "NAME")
```



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# Resources

- **Analyzing US Census Data: Methods, Maps, and Models in R**  
<https://walker-data.com/census-r/index.html>
- **Geocomputation with R**  
<https://geocompr.robinlovelace.net/index.html>
- **Geospatial Health Data: Modeling and Visualization with R-INLA and Shiny** <https://www.paulamoraga.com/book-geospatial/index.html>
- **Interactive Maps in R**  
[https://bhaskarvk.github.io/user2017.geodataviz/notebooks/03-Interactive-Maps.nb.html#using\\_tmap](https://bhaskarvk.github.io/user2017.geodataviz/notebooks/03-Interactive-Maps.nb.html#using_tmap)
- **Making Maps with R**  
[https://bookdown.org/nicohahn/making\\_maps\\_with\\_r5/docs/introduction.html](https://bookdown.org/nicohahn/making_maps_with_r5/docs/introduction.html)
- **Mapping in R** <https://map-rfun.library.duke.edu/>
- **Maps in R**  
[https://www.emilyburchfield.org/courses/eds/making\\_maps\\_in\\_r](https://www.emilyburchfield.org/courses/eds/making_maps_in_r)
- **Mastering Software Development in R**  
<https://bookdown.org/rdpeng/RProgDA/>
- **tmap: get started!** <https://cran.r-project.org/web/packages/tmap/vignettes/tmap-getstarted.html>
- **tmap : the easy way to plot thematic maps and show them interactively in R**  
[https://tlorusso.github.io/geodata\\_workshop/tmap\\_package](https://tlorusso.github.io/geodata_workshop/tmap_package)
- **Mapping the US with usmap** <https://cran.r-project.org/web/packages/usmap/vignettes/mapping.html>



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