

Week 11: Maps

m EMSE 4572: Exploratory Data Analysis

2 John Paul Helveston

Today's data

```
milk_production <- read_csv(here::here('data', 'milk_production.csv'))
us_coffee_shops <- read_csv(here::here('data', 'us_coffee_shops.csv'))</pre>
```

New packages:

```
install.packages('maps')
install.packages('mapproj')
install.packages('sf')
install.packages('rgeos')
install.packages('rnaturalearth')
devtools::install_github("ropensci/rnaturalearthhires")
devtools::install_github("ropensci/rnaturalearthdata")
```

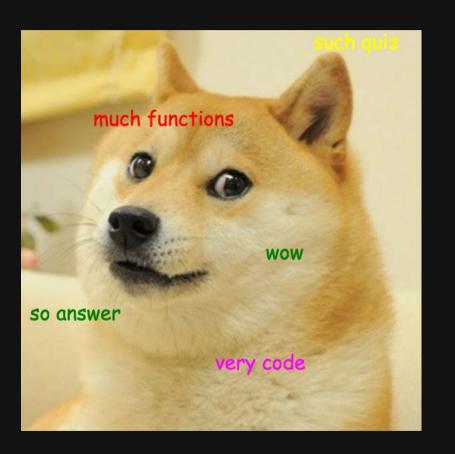
Quiz 4

Download the template from the #class channel

Make sure you unzip it!

When done, submit your quiz4 qmd on Blackboard

10:00



Week 11: Maps

- 1. Plotting maps
- 2. Adding data to maps

BREAK

3. Projections

Week 11: Maps

- 1. Plotting maps
- 2. Adding data to maps

BREAK

3. Projections

How to make a map

Step 1: Load a shape file

- a. Use a library
- b. Read in a shape file

Step 2: Plot the shape file

- a. Polygon data: geom_polygon()
- b. Simple Features data: geom_sf()

Polygon maps

Get the "World" shape file

```
library(ggplot2)
world <- map_data("world")
head(world)</pre>
```

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

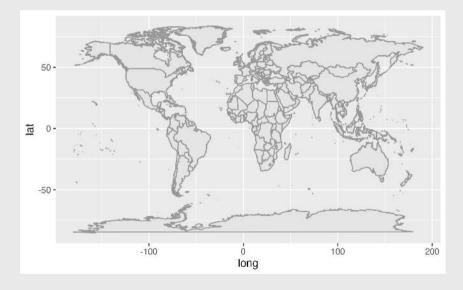
Polygon maps

Get the "World" shape file

```
library(ggplot2)
world <- map_data("world")</pre>
```

Make the plot with geom_polygon()

```
ggplot(world) +
  geom_polygon(
   aes(x = long, y = lat, group = group),
  fill = "grey90", color = "grey60"
)
```



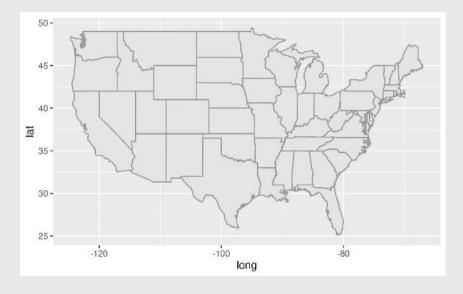
Polygon maps

Get the "US States" shape file

```
library(ggplot2)
us_states <- map_data("state")</pre>
```

Make the plot with geom_polygon()

```
ggplot(us_states) +
  geom_polygon(
   aes(x = long, y = lat, group = group),
  fill = "grey90", color = "grey60"
)
```



Library data from Natural Earth

```
library(rnaturalearth)
library(rnaturalearthdata)

world <- ne_countries(
    scale = "medium",
    returnclass = "sf"
)

world %>%
    select(name, geometry) %>%
    head()
```

```
#> Simple feature collection with 6 features and 1 field
  Geometry type: MULTIPOLYGON
  Dimension:
                  XY
  Bounding box: xmin: -70.06611 ymin: -18.01973 xmax:
  Geodetic CRS: +proj=longlat +datum=WGS84 +no defs +e
#>
                                       geometry
            name
           Aruba MULTIPOLYGON
  1 Afghanistan MULTIPOLYGON
#>
          Angola MULTIPOLYGON
        Anguilla MULTIPOLYGON
#> 3
                              (((-63.00122 1...
#> 4
         Albania MULTIPOLYGON
                              (((20.0639642...
           Aland MULTIPOLYGON (((20.61133 60...
#> 5
```

Get the "World" shape file

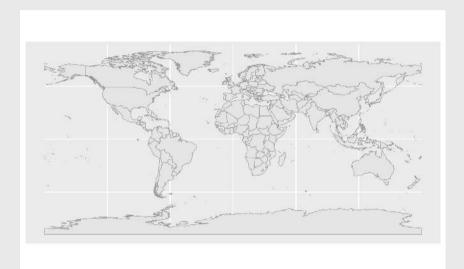
```
library(rnaturalearth)
library(rnaturalearthdata)

world <- ne_countries(
   scale = "medium",
   returnclass = "sf"
)</pre>
```

Make the plot with geom_sf()

```
library(sf)

ggplot(data = world) +
  geom_sf(fill = "grey90", color = "grey60")
```



Get the "US States" shape file

```
library(rnaturalearth)
library(rnaturalearthdata)

us_states <- ne_states(
  country = 'united states of america',
  returnclass = 'sf'
)</pre>
```

Get the "US States" shape file

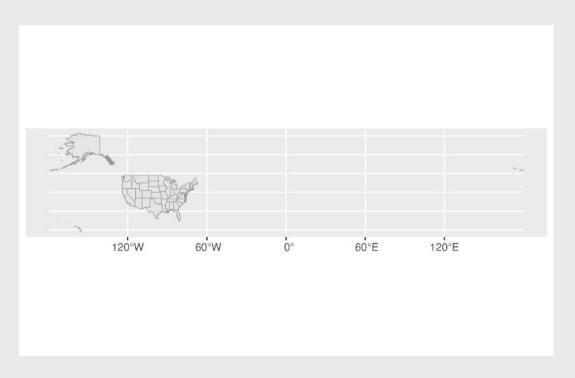
```
library(rnaturalearth)
library(rnaturalearthdata)

us_states <- ne_states(
   country = 'united states of america',
   returnclass = 'sf'
)</pre>
```

Make the plot with geom_sf()

```
library(sf)

ggplot(data = us_states) +
    geom_sf(fill = "grey90", color = "grey60")
```



Get the Continental "US States" shape file

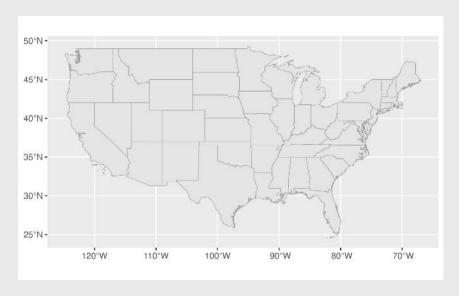
```
library(rnaturalearth)
library(rnaturalearthdata)

us_states_cont <- ne_states(
    country = 'united states of america',
    returnclass = 'sf'
    ) %>%
    filter(! name %in% c('Alaska', 'Hawaii'))
```

Make the plot with geom_sf()

```
library(sf)

ggplot(data = us_states_cont) +
   geom_sf(fill = "grey90", color = "grey60")
```



The maps package

Includes data on:

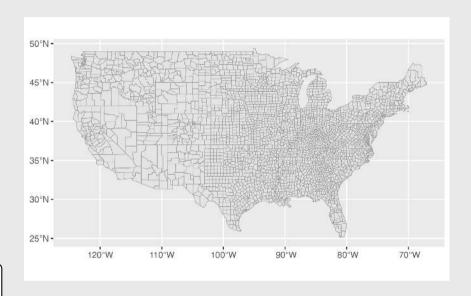
- World: world, world.cities, lakes
- US: states, county, state, usa
- France: france
- Italy: italy
- New zealand: nz

Example:

```
library(maps)

us_counties <- st_as_sf(
    map("county", plot = FALSE, fill = TRUE))

ggplot(data = us_counties) +
    geom_sf(fill = 'grey90', color = 'grey60')</pre>
```



Simple Features (sf) maps: st_read()

Read in the "World" shape file from Natural Earth

```
library(sf)
world <- st_read(here::here(
    'data', 'natural_earth_countries',
    'ne_50m_admin_0_countries.shp')) %>%
    clean_names()
```

```
#> Reading layer `ne_50m_admin_0_countries' from data so
#> Simple feature collection with 241 features and 94 fix
#> Geometry type: MULTIPOLYGON
#> Dimension: XY
#> Bounding box: xmin: -180 ymin: -89.99893 xmax: 180 ymin: -89.99893 xmax: -8
```

Simple Features (sf) maps: st_read()

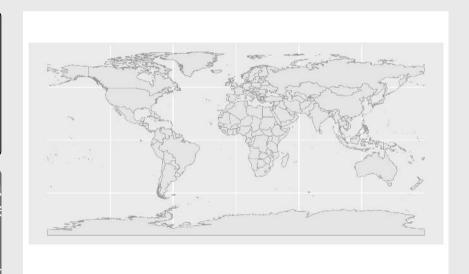
Read in the "World" shape file

```
library(sf)

world <- st_read(here::here(
   'data', 'natural_earth_countries',
   'ne_50m_admin_0_countries.shp')) %>%
   clean_names()
```

```
#> Reading layer `ne_50m_admin_0_countries' from data so
#> Simple feature collection with 241 features and 94 fix
#> Geometry type: MULTIPOLYGON
#> Dimension: XY
#> Bounding box: xmin: -180 ymin: -89.99893 xmax: 180 ymin: Geodetic CRS: WGS 84
```

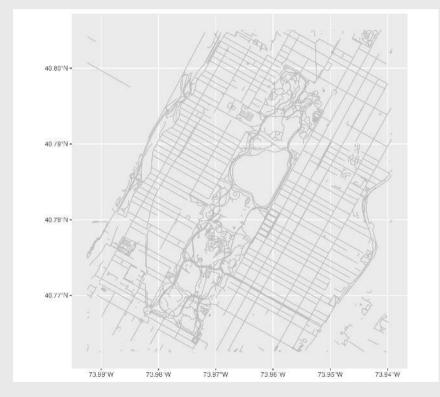
```
ggplot(data = world) +
  geom_sf(fill = "grey90", color = "grey60")
```



Simple Features (sf) maps: st_read()

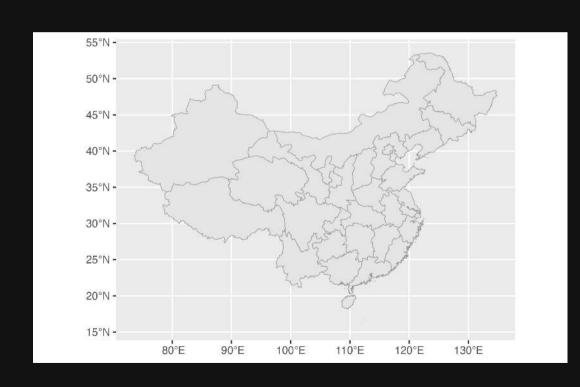
Read in the "Central Park" shape file [source]

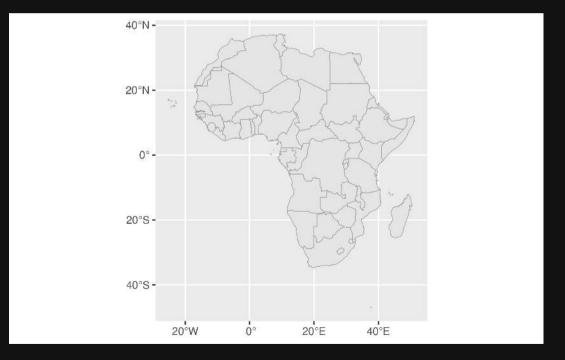
```
library(sf)
central park <- st read(here::here(</pre>
  'data', 'central_park', 'CentralPark.shp'))
#> Reading layer `CentralPark' from data source `/Users/:
#> Simple feature collection with 2550 features and 6 fig
#> Geometry type: LINESTRING
#> Dimension:
#> Bounding box: xmin: -73.99249 ymin: 40.7625 xmax: -73
#> Geodetic CRS: WGS 84
ggplot(data = central_park) +
  geom sf(color = 'grey75')
```



Your turn

Use the **rnaturalearth** library to extract and plot the shape files for China and Africa:





Week 11: Maps

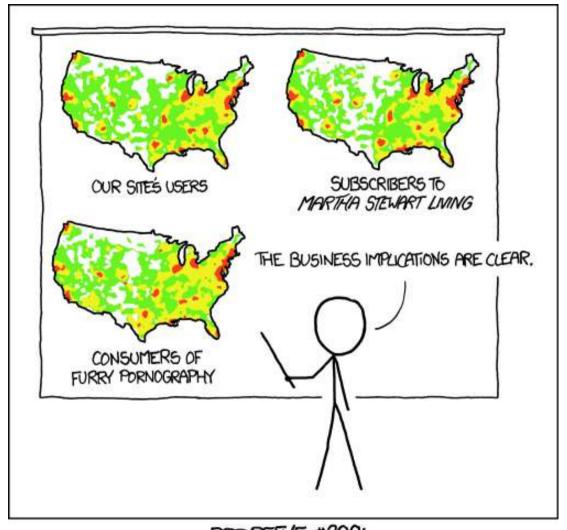
- 1. Plotting maps
- 2. Adding data to maps

BREAK

3. Projections

First rule of adding data to maps:

Do you need to make a map?



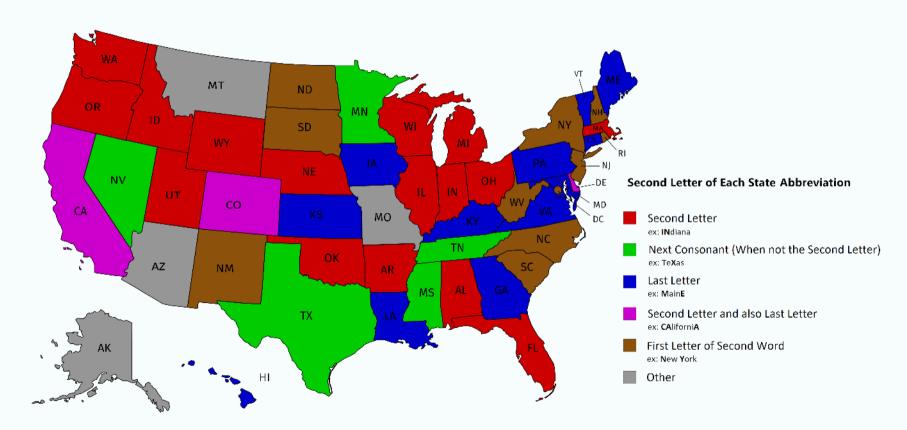
PET PEEVE #208: GEOGRAPHIC PROFILE MAPS WHICH ARE BASICALLY JUST POPULATION MAPS

1. Choropleth maps

2. Point maps

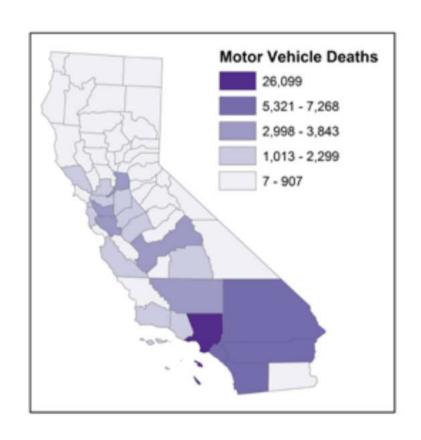
Choropleth - from Greek:

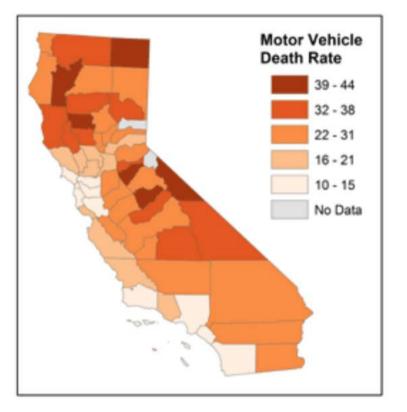
- χῶρος "choros" (area/region)
- πλῆθος "plethos" (multitude)



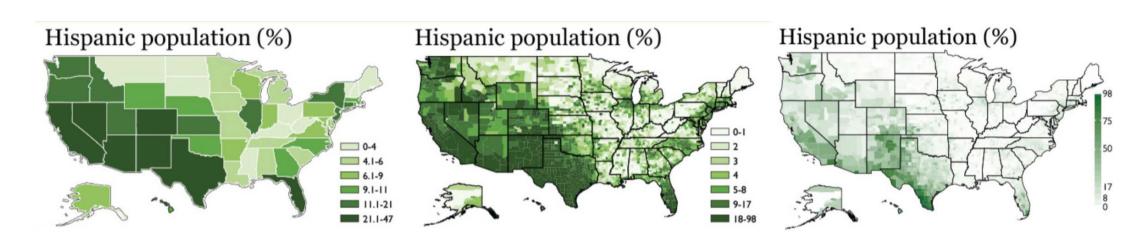
Choropleth maps are easily misleading

Number of events != Number of events per capita

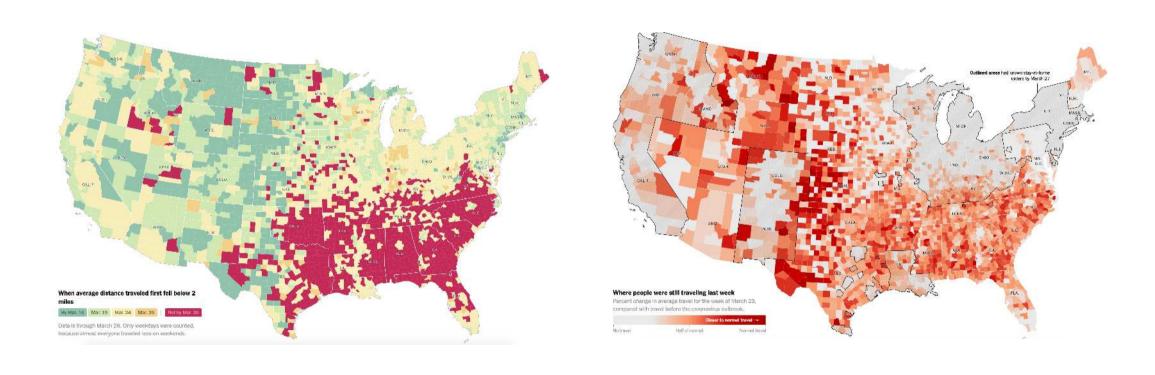




Manipulating fill scale produces wildly different maps

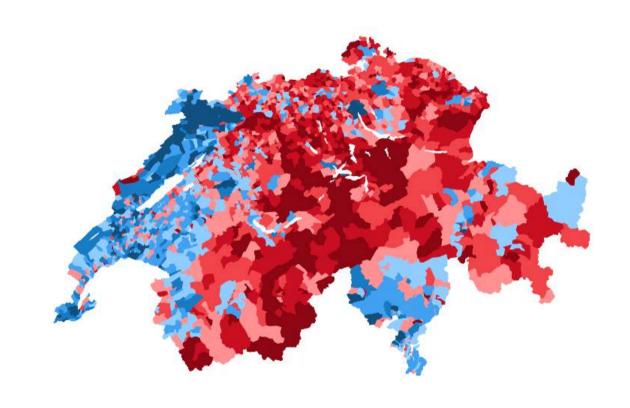


Manipulating fill scale produces wildly different maps



Source: New York Times

Land doesn't vote - people vote

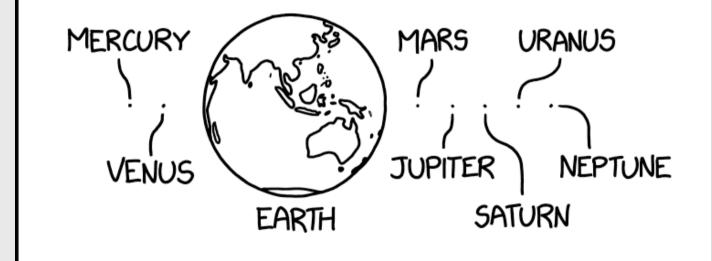


Land doesn't vote - people vote

Election maps from: http://www-personal.umich.edu/~mejn/election/2016/



MOST SOLAR SYSTEM DIAGRAMS ARE MISLEADING.
THIS CHART OFFERS A MORE ACCURATE VIEW BY
SHOWING THE PLANETS SIZED BY POPULATION.



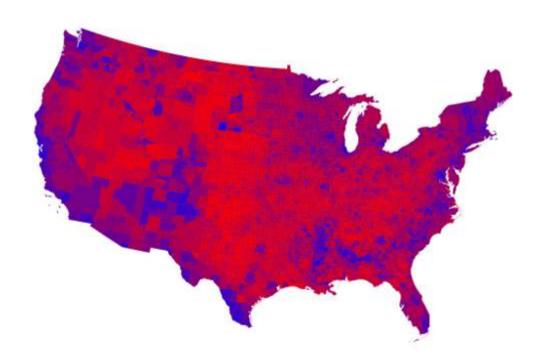
Easy to lie with fake news

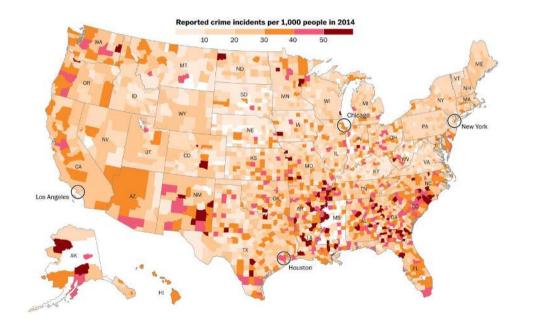


(here is what actual crime rates look like)

2016 Election map [source]

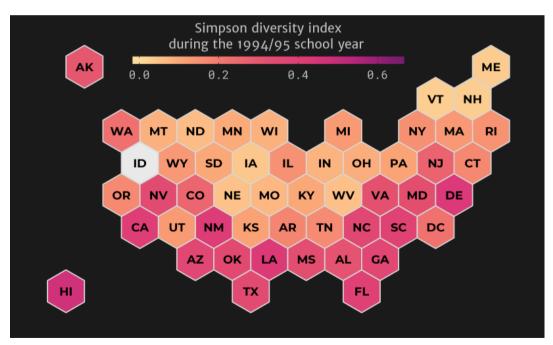
2014 Crime map [source]





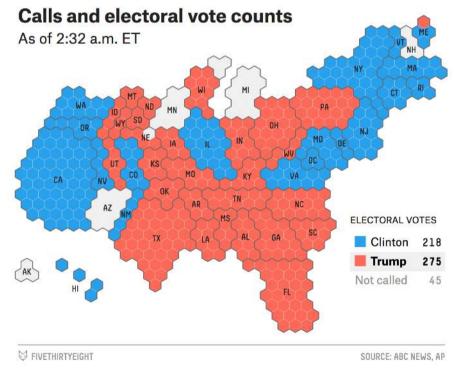
A choropleth alternative: hex maps

1994 Simpson Diversity Index in US Schools



https://github.com/malcolmbarrett/designing_ggplots

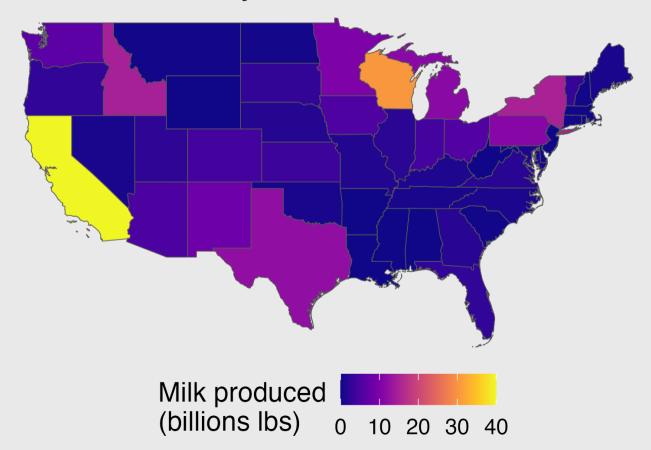
2016 Electoral College



https://fivethirtyeight.com/

How to make a choropleth map

Milk Production by State in 2017



How to make a choropleth map

Get the "fill" data

```
milk_2017 <- milk_production %>%
    filter(year == 2017) %>%
    select(name = state, milk_produced) %>%
    mutate(milk_produced = milk_produced / 10^9)
```

Get the "map" data

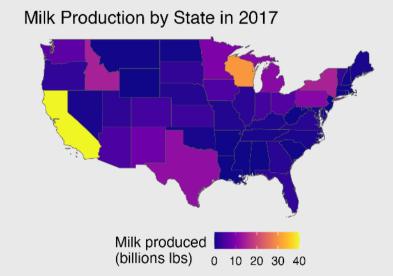
```
us_states <- ne_states(
    country = 'united states of america',
    returnclass = 'sf') %>%
    filter(! name %in% c('Alaska', 'Hawaii')) %>%
    left_join(milk_2017, by = 'name')
```

```
us_states %>%
    select(name, milk_produced) %>9
    head()
```

```
#> Simple feature collection with 6
#> Geometry type: MULTIPOLYGON
#> Dimension:
                 XY
#> Bounding box: xmin: -124.7346
#> Geodetic CRS: WGS 84
#>
            name milk produced
#> 1 Washington
                         6.526 MUL
         Idaho
                        14.627 MUL
#> 2
     Montana
                         0.288 MUL
#> 4 North Dakota
                         0.345 MUI
#> 5
       Minnesota
                         9.864 MUI
        Michigan
                        11.231 MUI
#> 6
```

How to make a choropleth map

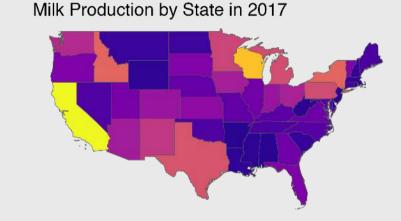
```
ggplot(us_states) +
  geom_sf(aes(fill = milk_produced)) +
  scale_fill_viridis(
    option = "plasma",
    limits = c(0, 40)) +
  theme_void(base_size = 15) +
  theme(legend.position = 'bottom') +
  labs(
    fill = 'Milk produced\n(billions lbs)',
    title = 'Milk Production by State in 2017'
)
```



How to make a choropleth map

```
ggplot(us_states) +
  geom_sf(aes(fill = milk_produced)) +
  scale_fill_viridis(
    trans = 'sqrt',
    option = "plasma",
    limits = c(0, 40)) +
  theme_void(base_size = 15) +
  theme(legend.position = 'bottom') +
  labs(
    fill = 'Milk produced\n(billions lbs)',
    title = 'Milk Production by State in 2017'
)
```

Non-linear scale:



10 203040

Milk produced (billions lbs) 0

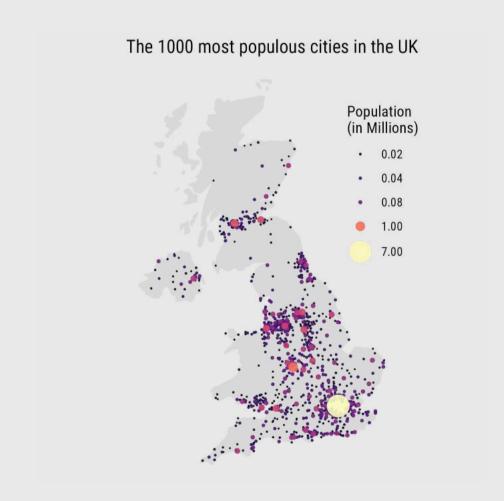
1. Choropleth maps

2. Point maps

Points as locations



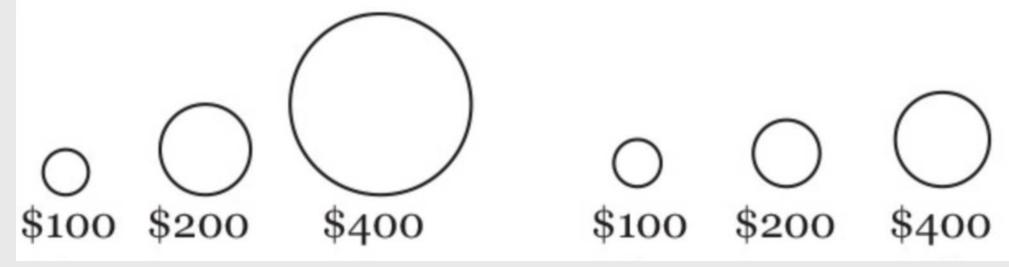
Points encoding a variable



For point size, use **area**, not radius

$$Area = \pi r^2$$

Radius Area

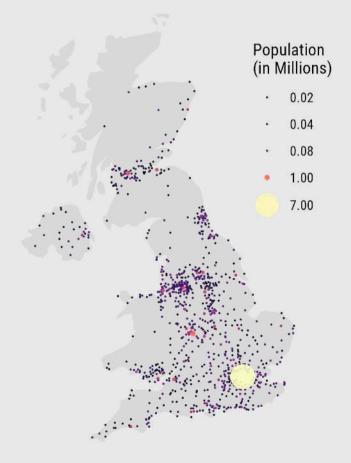


42 / 65

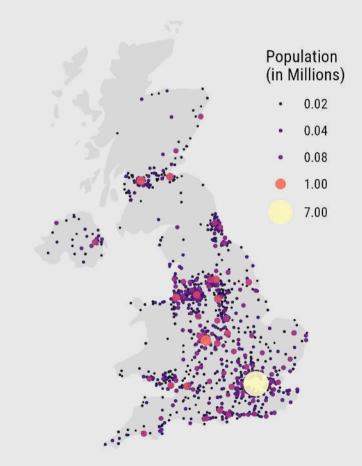
Radius

Area

The 1000 most populous cities in the UK

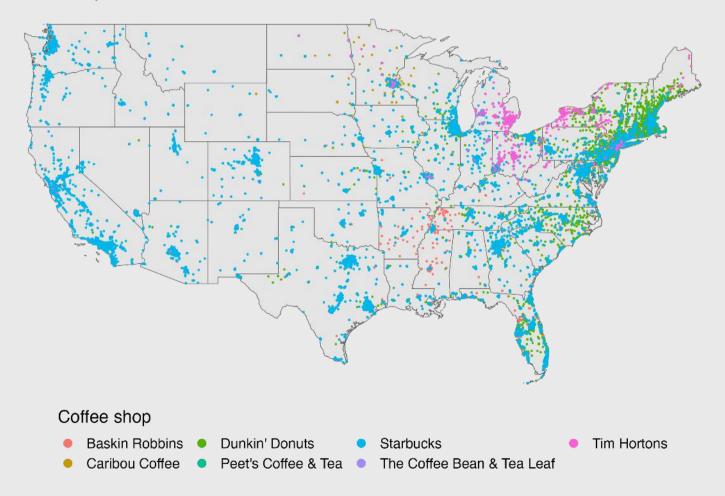


The 1000 most populous cities in the UK



How to add points to a map

Coffee Shops in the US



How to add points to a map

Load the continental US shape file

```
us_states_cont <- ne_states(
    country = 'united states of america',
    returnclass = 'sf') %>%
    filter(! name %in% c('Alaska', 'Hawaii'))
```

Read in the coffee shop data

```
us_coffee_shops <- read_csv(here::here(
   'data', 'us_coffee_shops.csv'))

# Only keep data in continental US
us_coffee_shops <- us_coffee_shops %>%
   filter(
    lat > 22, lat < 50,
    long > -150, long < -66
   )</pre>
```

```
head(us_coffee_shops)
```

```
#> # A tibble: 6 × 8
                     lat long unique id ci
    name
               <dbl> <dbl>
    <chr>
                                   <dbl> <d
  1 Baskin Robbins
                   40.8 -73.4
                                 3304448
  2 Baskin Robbins
                   42.1 -88.0
                                11342048 Rd
#> 3 Baskin Robbins 34.0 -84.5
                                 3304169 Ma
#> 4 Baskin Robbins
                   29.8 -95.6
                                 3304006 Hd
#> 5 Baskin Robbins
                   36.4 -89.5
                                 3303959 T
\#> 6 Baskin Robbins 40.7 -73.6
                                 3304507 Me
```

How to add points to a map

Plot coffee shop locations over map

```
ggplot() +
  geom sf(data = us states cont) +
  geom point(
    data = us coffee shops,
    aes(x = long, y = lat, color = name),
    size = 0.3
  theme_void(base_size = 15) +
  theme(legend.position = 'bottom') +
  guides(color = guide_legend(
    # Move legend title to top
    title.position = "top",
    # Increase legend point size
    override.aes = list(size = 3))) +
  labs(
    color = 'Coffee shop',
    title = 'Coffee Shops in the US'
```



Your turn

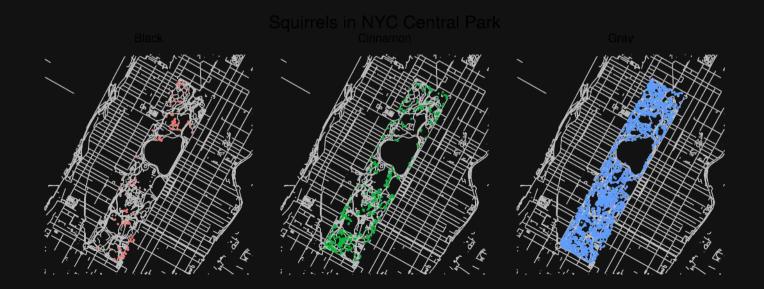


Create this map of squirrels in NYC's Central Park using this data from the Squirrel Census:

- The CentralPark.shp file in the data/central_park folder.
- The nyc_squirrels.csv file in the data folder.

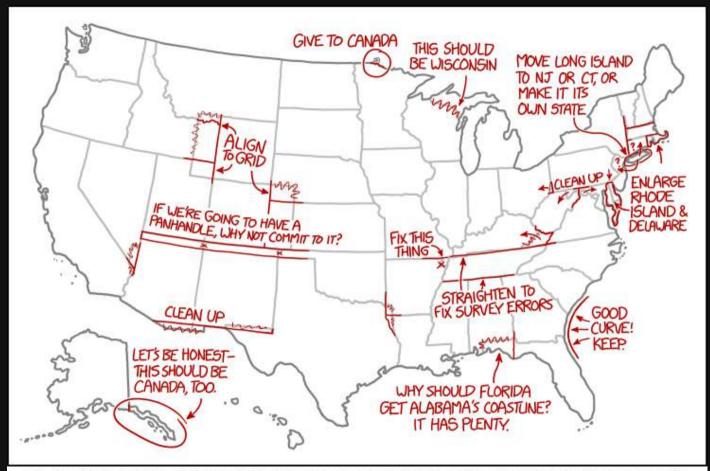
Hint: The color is mapped to the primary_fur_color variable

(More about the Squirrel Census <u>here</u>)



Intermission





IT WAS SCARY WHEN THE GRAPHIC DESIGNERS SEIZED CONTROL OF THE COUNTRY, BUT IT TURNED OUT THEY JUST WANTED TO FIX SOME THINGS ABOUT THE STATE BORDERS THAT HAD ALWAYS BOTHERED THEM.

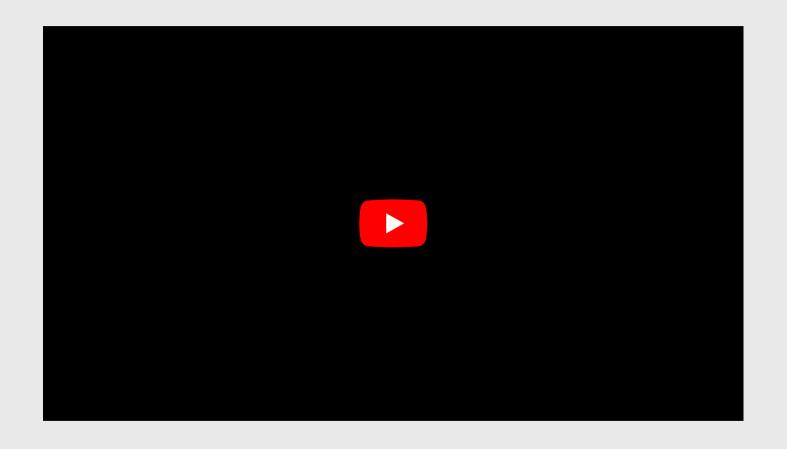
Week 11: Maps

- 1. Plotting maps
- 2. Adding data to maps

BREAK

3. Projections

What's a map projection?



What is the best projection?...it depends

1. Compare projections

2. Compare country sizes

Using projections

To modify the projection of a map, use $coord_sf(crs = st_crs(XXXX))$

```
world <- ne_countries(scale = "medium", returnclass = "sf")</pre>
```

Default (long-lat)

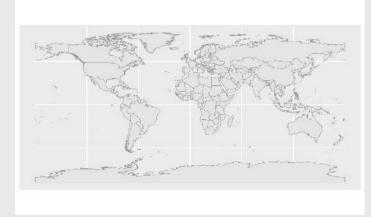
Robinson projection

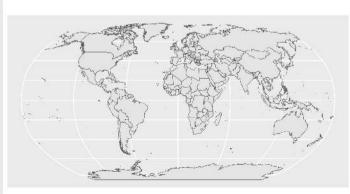
Mollweide projection

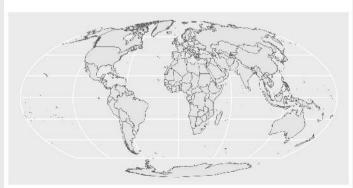
```
ggplot(data = world) +
   geom_sf()
```

```
ggplot(data = world) +
    geom_sf() +
    coord_sf(crs = "ESRI:54030")
```

```
ggplot(data = world) +
    geom_sf() +
    coord_sf(crs = "ESRI:54009")
```







Common Projections

ggplot layer:

```
coord_sf(crs = "ESRI:XXXX")
```

World

Code	Projection
"ESRI:54030"	Robinson
"ESRI:54002"	Equidistant cylindrical
"ESRI:54004"	Mercator
"ESRI:54008"	Sinusoidal
"ESRI:54009"	Mollweide

United States

Code	Projection
"ESRI:102003"	Albers
"ESRI:102004"	Lambert Conformal Conic
4269	NAD 83

US projections

```
us_states_cont <- ne_states(country = 'united states of america',
    returnclass = 'sf') %>%
    filter(! name %in% c('Alaska', 'Hawaii'))
```

NAD 83 projection

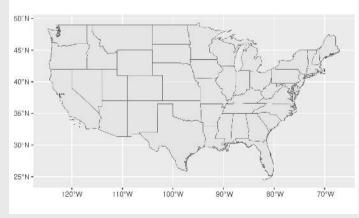
Mercator

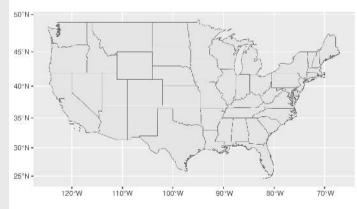
Albers

```
ggplot(data = world) +
    geom_sf() +
    coord_sf(crs = 4269)
```

```
ggplot(data = world) +
    geom_sf() +
    coord_sf(crs = "ESRI:54004")
```

```
ggplot(data = us_states_cont)
    geom_sf() +
    coord_sf(crs = "ESRI:10200")
```

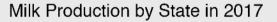


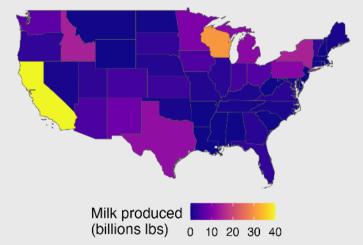




Mapping data to projections - choropleth map

```
milk 2017 <- milk production %>%
  filter(year == 2017) %>%
  select(name = state, milk produced) %>%
 mutate(milk produced = milk produced / 10^9)
us states <- ne states(
  country = 'united states of america',
  returnclass = 'sf') %>%
  filter(! name %in% c('Alaska', 'Hawaii')) %>%
  left join(milk 2017, by = 'name')
ggplot(us states) +
  geom sf(aes(fill = milk produced)) +
  scale fill viridis(
    option = "plasma",
    limits = c(0, 40)) +
  theme void(base size = 15) +
  theme(legend.position = 'bottom') +
  labs(
    fill = 'Milk produced\n(billions lbs)',
    title = 'Milk Production by State in 2017'
```



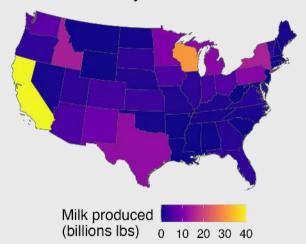


Mapping data to projections - choropleth map

```
milk 2017 <- milk production %>%
  filter(year == 2017) %>%
  select(name = state, milk produced) %>%
 mutate(milk produced = milk produced / 10^9)
us states <- ne states(
  country = 'united states of america',
  returnclass = 'sf') %>%
  filter(! name %in% c('Alaska', 'Hawaii')) %>%
  left join(milk 2017, by = 'name')
ggplot(us states) +
  geom sf(aes(fill = milk produced)) +
  scale fill viridis(
    option = "plasma",
    limits = c(0, 40)) +
  theme void(base size = 15) +
  theme(legend.position = 'bottom') +
  labs(
    fill = 'Milk produced\n(billions lbs)',
    title = 'Milk Production by State in 2017'
  coord sf(crs = "ESRI:102003")
```

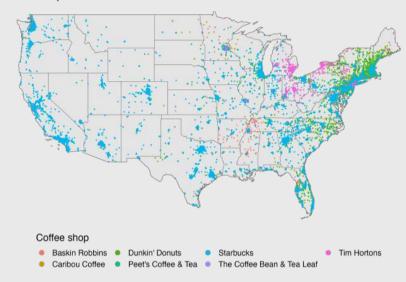
Albers Projection

Milk Production by State in 2017



```
us states cont <- ne states(
 country = 'united states of america',
 returnclass = 'sf') %>%
 filter(! name %in% c('Alaska', 'Hawaii'))
us coffee shops <- us coffee shops %>%
 filter(lat > 22, lat < 50,
         long > -150, long < -66)
qqplot() +
 geom sf(data = us states cont) +
 geom point(
   data = us coffee shops.
   aes(x = long, y = lat, color = name),
   size = 0.3) +
 theme void(base size = 15) +
 theme(legend.position = 'bottom') +
 guides(color = guide legend(
   # Move legend title to top
   title.position = "top",
   # Increase legend point size
   override.aes = list(size = 3))) +
  labs(
   color = 'Coffee shop',
   title = 'Coffee Shops in the US'
```

Coffee Shops in the US



```
us states cont <- ne states(
 country = 'united states of america',
  returnclass = 'sf') %>%
 filter(! name %in% c('Alaska', 'Hawaii'))
us coffee shops <- us coffee shops %>%
 filter(lat > 22, lat < 50,
         long > -150, long < -66)
qqplot() +
 geom sf(data = us states cont) +
 geom point(
   data = us coffee shops.
   aes(x = long, y = lat, color = name),
   size = 0.3) +
 theme void(base size = 15) +
 theme(legend.position = 'bottom') +
 guides(color = guide legend(
   # Move legend title to top
   title.position = "top",
   # Increase legend point size
   override.aes = list(size = 3))) +
  labs(
   color = 'Coffee shop',
   title = 'Coffee Shops in the US'
 coord sf(crs = "ESRI:102003")
```

Fail!





First match us_coffee_shops crs to us_states_cont

```
us_states_cont <- ne_states(
    country = 'united states of america',
    returnclass = 'sf') %>%
    filter(! name %in% c('Alaska', 'Hawaii'))

us_coffee_shops <- us_coffee_shops %>%
    filter(lat > 22, lat < 50,
        long > -150, long < -66)

us_coffee_shops_sf <- st_as_sf(us_coffee_shops,
    coords = c("long", "lat"),
    crs = st_crs(us_states_cont))</pre>
```

```
head(us_coffee_shops_sf)
```

```
#> Simple feature collection with 6
#> Geometry type: POINT
#> Dimension: XY
#> Bounding box: xmin: -95.60337
#> Geodetic CRS: WGS 84
#> # A tibble: 6 × 7
                  unique id city
  name
#>
                      <dbl> <chr>
#> <chr>
#> 1 Baskin Robbins 3304448 Hunt:
#> 2 Baskin Robbins
                   11342048 Roll:
#> 3 Baskin Robbins
                    3304169 Marie
#> 4 Baskin Robbins
                    3304006 Houst
#> 5 Baskin Robbins
                    3303959 Tipto
#> 6 Baskin Robbins
                    3304507 Merri
```

Plot coffee shop locations over map with geom_sf()

```
ggplot() +
  geom_sf(data = us_states_cont) +
  geom sf(
    data = us coffee shops sf,
    aes(color = name),
    size = 0.3) +
  theme_void(base_size = 15) +
  theme(legend.position = 'bottom') +
  guides(color = guide legend(
    # Move legend title to top
   title.position = "top",
   # Increase legend point size
   override.aes = list(size = 3))) +
  labs(
    fill = 'Coffee shop',
    title = 'Coffee Shops in the US'
```



Plot coffee shop locations over map with geom_sf()

```
ggplot() +
  geom_sf(data = us_states cont) +
  geom sf(
    data = us coffee shops sf,
    aes(color = name),
    size = 0.3) +
  theme void(base size = 15) +
  theme(legend.position = 'bottom') +
  guides(color = guide legend(
    # Move legend title to top
    title.position = "top",
    # Increase legend point size
   override.aes = list(size = 3))) +
  labs(
    fill = 'Coffee shop',
   title = 'Coffee Shops in the US'
  coord sf(crs = "ESRI:102003")
```

Albers Projection



Plot coffee shop locations over map with geom_sf()

```
ggplot() +
  geom_sf(data = us_states cont) +
  geom sf(
    data = us coffee shops sf,
    aes(color = name),
    size = 0.3) +
  theme void(base size = 15) +
  theme(legend.position = 'bottom') +
  guides(color = guide legend(
    # Move legend title to top
    title.position = "top",
    # Increase legend point size
   override.aes = list(size = 3))) +
  labs(
    fill = 'Coffee shop',
    title = 'Coffee Shops in the US'
  coord_sf(crs = "ESRI:102004")
```

LCC Projection



20:00

Your turn

Use the <u>internet_users_country.csv</u> data and the <u>world</u> data frame from the **rnaturalearth** library to create these two versions of internet access by country in 2015.

Hints:

- The iso_a3 variable in the worlds data frame corresponds with the code variable in the internet_users_country.csv data frame (use this for joining).
- Use scale_fill_gradient() to fill the color:

```
scale_fill_gradient(
    low = "#e7e1ef",
    high = "#dd1c77",
    na.value = "grey70",
    limits = c(0, 100))
```

Robinson Projection

% of population with access

Mercator Projection

Internet access by country in 2015



% of population with access

Extra practice

Your turn

Use the us_states_cont data frame and the state_abbs data frame to create a labeled map of the U.S.:

