

Wrap-up

Loose ends

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Week 9

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

Agenda

- Wrap up maps and loose ends with HTML

Let's get some census data

Note

To do this, you need to first register an API key with the US Census, which you can do [here](#). Then use `census_api_key("YOUR API KEY")`.

Alternatively, you can specify `CENSUS_API_KEY = "YOUR API KEY"` in **.Renviron**. You can do this by using `usethis::edit_r_environ()`

Getting the data

```
library(tidycensus)
# Find variable code
# v <- load_variables(2019, "acs5")
# View(v)

census_vals <- get_acs(
  geography = "tract",
  state = "OR",
  variables = c(med_income = "B06011_001",
                ed_attain = "B15003_001"),
  year = 2019,
  geometry = TRUE
)
```

##

```
|
|
|
|==
|
|====
|
|=====
```

Look at the data

```
census_vals
```

```
## Simple feature collection with 1668 features and 5 fields (with 12 geometries e
## Geometry type: MULTIPOLYGON
## Dimension:      XY
## Bounding box:   xmin: -124.5662 ymin: 41.99179 xmax: -116.4635 ymax: 46.29204
## Geodetic CRS:   NAD83
## First 10 features:
```

##	GEOID	NAME	variable	estima
## 1	41031960302	Census Tract 9603.02, Jefferson County, Oregon	med_income	300
## 2	41031960302	Census Tract 9603.02, Jefferson County, Oregon	ed_attain	30
## 3	41057960100	Census Tract 9601, Tillamook County, Oregon	med_income	302
## 4	41057960100	Census Tract 9601, Tillamook County, Oregon	ed_attain	27
## 5	41015950100	Census Tract 9501, Curry County, Oregon	med_income	189
## 6	41015950100	Census Tract 9501, Curry County, Oregon	ed_attain	23
## 7	41039001902	Census Tract 19.02, Lane County, Oregon	med_income	248
## 8	41039001902	Census Tract 19.02, Lane County, Oregon	ed_attain	40
## 9	41029000300	Census Tract 3, Jackson County, Oregon	med_income	250
## 10	41029000300	Census Tract 3, Jackson County, Oregon	ed_attain	43

Remove missing geometry rows

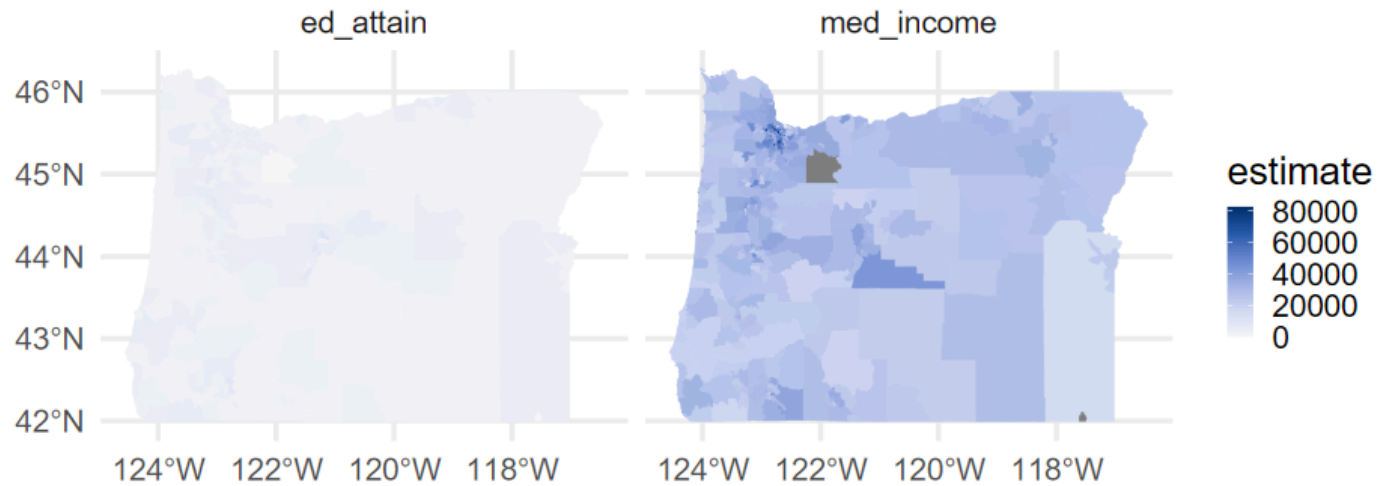
- Tidycensus is (currently) bringing in some rows with missing geometries
- This is not a big deal for ggplot, but is for other plotting systems
- Let's remove those rows

```
census_vals <- census_vals[!st_is_empty(census_vals$geometry), ,
```

Plot it

```
library(colorspace)
ggplot(census_vals) +
  geom_sf(aes(fill = estimate, color = estimate)) +
  facet_wrap(~variable) +
  guides(color = "none") +
  scale_fill_continuous_diverging("Blue-Red 3", rev = TRUE) +
  scale_color_continuous_diverging("Blue-Red 3", rev = TRUE)
```

hmm...

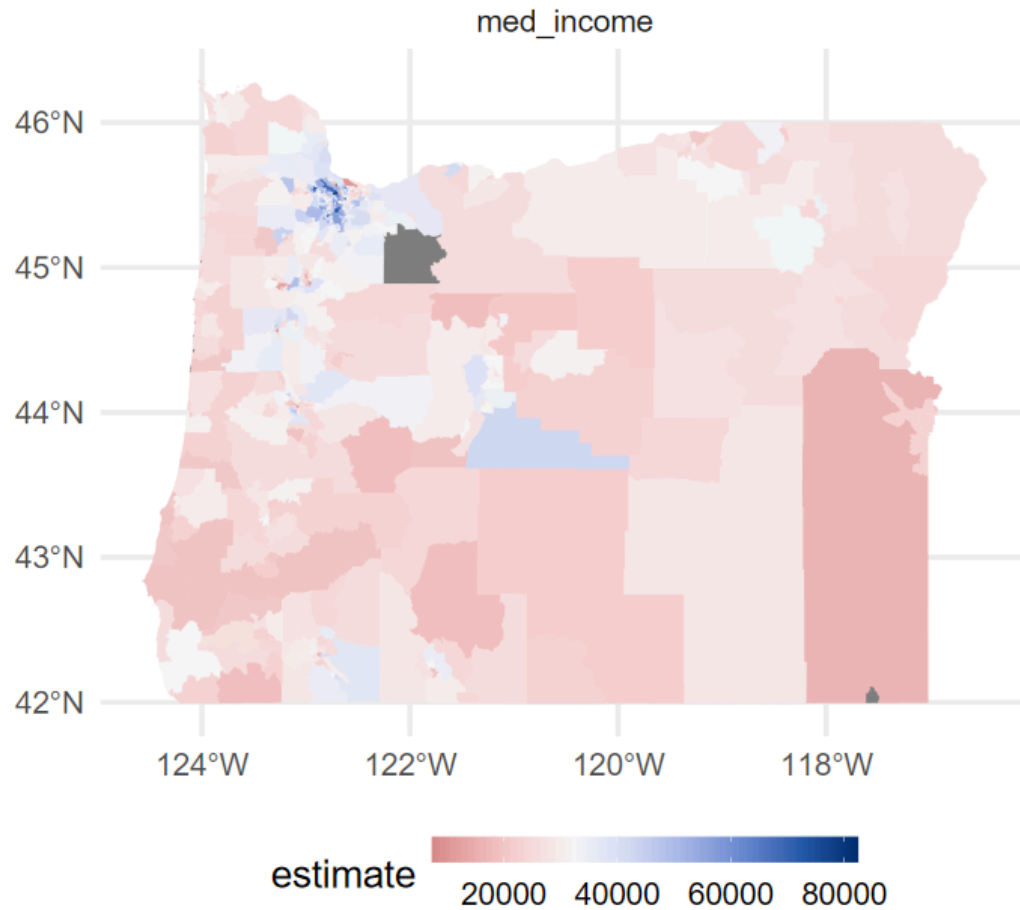


Try again

```
library(colorspace)
income <- filter(census_vals, variable == "med_income")

income_plot <- ggplot(income) +
  geom_sf(aes(fill = estimate, color = estimate)) +
  facet_wrap(~variable) +
  guides(color = "none") +
  scale_fill_continuous_diverging(
    "Blue-Red 3",
    rev = TRUE,
    mid = mean(income$estimate, na.rm = TRUE)
  ) +
  scale_color_continuous_diverging(
    "Blue-Red 3",
    rev = TRUE,
    mid = mean(income$estimate, na.rm = TRUE)
  ) +
  theme(legend.position = "bottom",
        legend.key.width = unit(2, "cm"))
```

income_plot

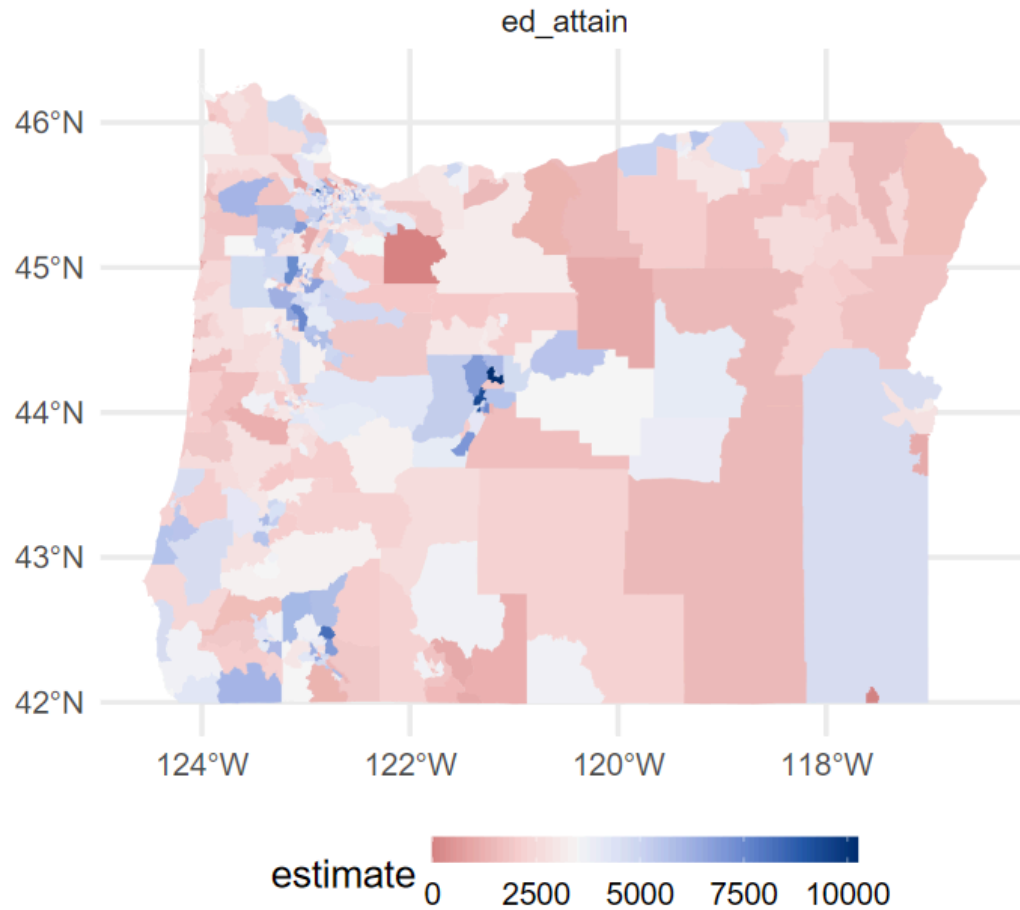


Same thing for education

```
ed <- filter(census_vals, variable == "ed_attain")

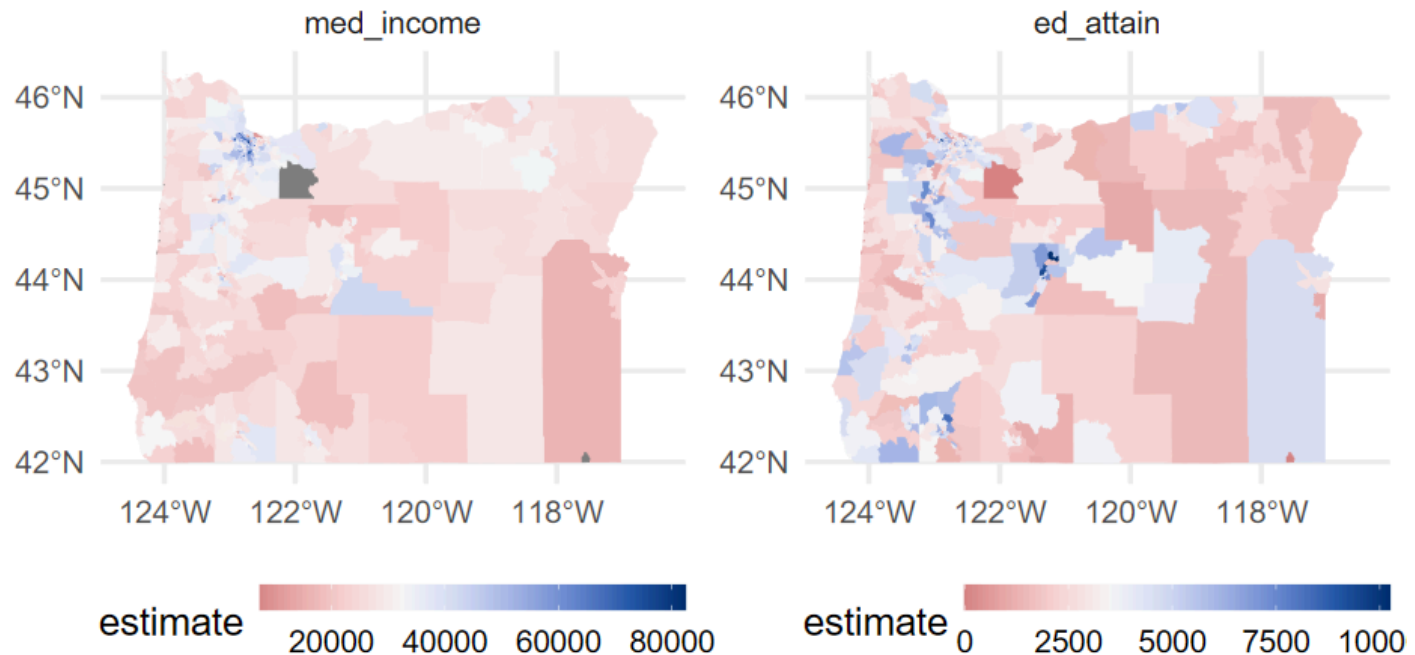
ed_plot <- ggplot(ed) +
  geom_sf(aes(fill = estimate, color = estimate)) +
  facet_wrap(~variable) +
  guides(color = "none") +
  scale_fill_continuous_diverging(
    "Blue-Red 3",
    rev = TRUE,
    mid = mean(ed$estimate, na.rm = TRUE)
  ) +
  scale_color_continuous_diverging(
    "Blue-Red 3",
    rev = TRUE,
    mid = mean(ed$estimate, na.rm = TRUE)
  ) +
  theme(legend.position = "bottom",
        legend.key.width = unit(2, "cm"))
```

ed_plot



Put them together

```
gridExtra::grid.arrange(income_plot, ed_plot, ncol = 2)
```

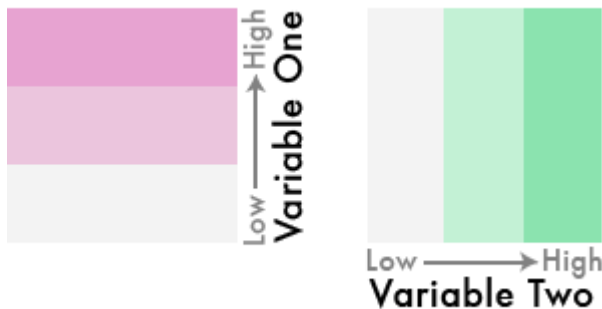


Bivariate color scales

How?

There are a few different ways. Here's one:

- Break continuous variable into categorical values
- Assign each combination of values between categorical vars a color
- Make sure the combinations of the colors make sense



Do it

Note - this will be fairly quick. I'm not expecting you to know how to do this, but I want to show you the idea and give you the breadcrumbs for the code you may need.

First - move it to wider

```
wider <- get_acs(  
  geography = "tract",  
  state = "OR",  
  variables = c(med_income = "B06011_001",  
                ed_attain = "B15003_001"),  
  year = 2019,  
  geometry = TRUE,  
  output = "wide"  
)  
  
# remove missing geometry rows  
wider <- wider[!st_is_empty(wider$geometry), , drop = FALSE]
```

Find the quartiles

```
ed_quartiles <- quantile(  
  wider$ed_attainE,  
  probs = seq(0, 1, length.out = 4),  
  na.rm = TRUE  
)
```

```
inc_quartiles <- quantile(  
  wider$med_incomeE,  
  probs = seq(0, 1, length.out = 4),  
  na.rm = TRUE  
)
```

```
ed_quartiles
```

```
##           0% 33.33333% 66.66667%      100%  
##      0.000  2715.000  3949.333 10245.000
```

```
inc_quartiles
```

```
##           0% 33.33333% 66.66667%      100%  
##    7449.00  26718.33  34375.00  82375.00
```

Create the cut variable

```
wider <- wider %>%  
  mutate(cat_ed = cut(ed_attainE, ed_quartiles),  
         cat_inc = cut(med_incomeE, inc_quartiles))  
wider %>%  
  select(starts_with("cat"))
```

Simple feature collection with 828 features and 2 fields

Geometry type: MULTIPOLYGON

Dimension: XY

Bounding box: xmin: -124.5662 ymin: 41.99179 xmax: -116.4635 ymax: 46.29204

Geodetic CRS: NAD83

First 10 features:

##	cat_ed	cat_inc	geometry
## 1	(2.71e+03,3.95e+03]	(2.67e+04,3.44e+04]	MULTIPOLYGON (((-121.8495 4...
## 2	(0,2.71e+03]	(2.67e+04,3.44e+04]	MULTIPOLYGON (((-123.983 45...
## 3	(0,2.71e+03]	(7.45e+03,2.67e+04]	MULTIPOLYGON (((-124.5646 4...
## 4	(3.95e+03,1.02e+04]	(7.45e+03,2.67e+04]	MULTIPOLYGON (((-122.9855 4...
## 5	(3.95e+03,1.02e+04]	(7.45e+03,2.67e+04]	MULTIPOLYGON (((-122.9079 4...
## 6	(2.71e+03,3.95e+03]	(3.44e+04,8.24e+04]	MULTIPOLYGON (((-123.5016 4...
## 7	(0,2.71e+03]	(7.45e+03,2.67e+04]	MULTIPOLYGON (((-123.0927 4...
## 8	(2.71e+03,3.95e+03]	(7.45e+03,2.67e+04]	MULTIPOLYGON (((-120.8811 4...
## 9	(2.71e+03,3.95e+03]	(7.45e+03,2.67e+04]	MULTIPOLYGON (((-123.5093 4...
## 10	(2.71e+03,3.95e+03]	(7.45e+03,2.67e+04]	MULTIPOLYGON (((-123.8179 4...

Set palette

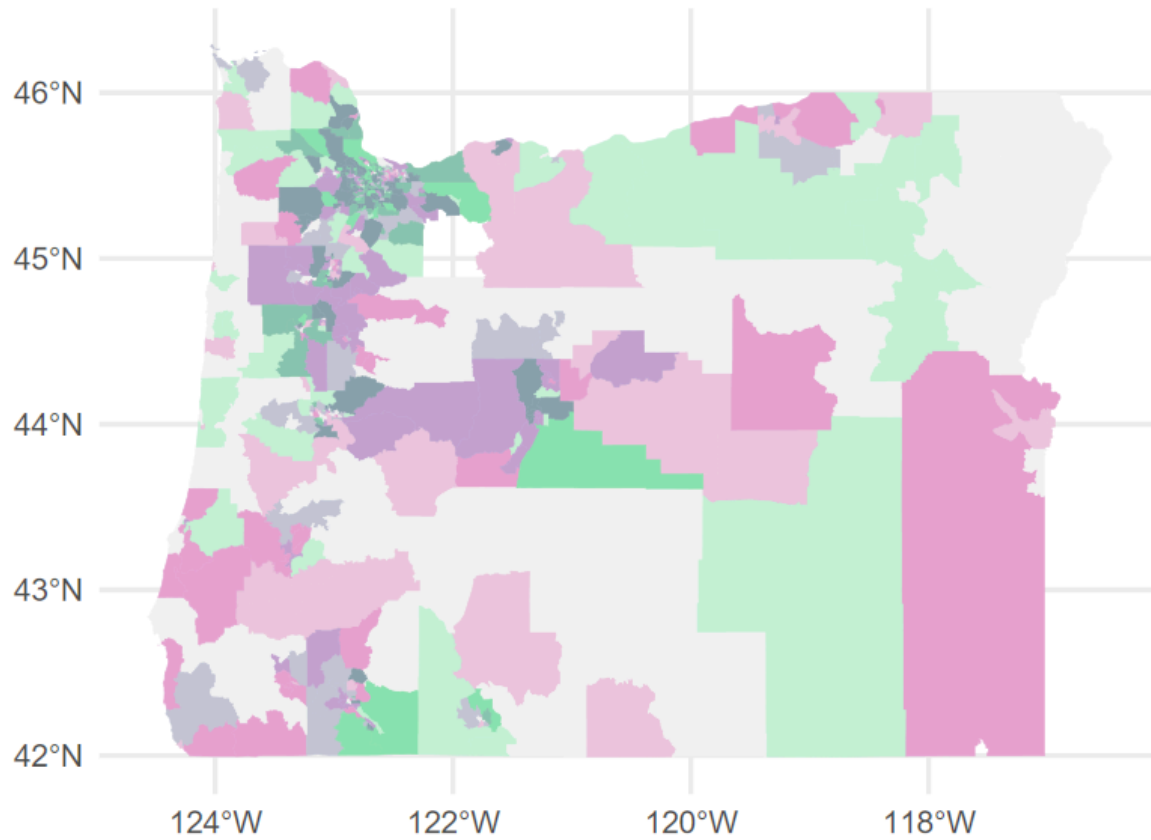
```
# First drop geo column
pal <- st_drop_geometry(wider) %>%
  count(cat_ed, cat_inc) %>%
  arrange(cat_ed, cat_inc) %>%
  drop_na(cat_ed, cat_inc) %>%
  mutate(pal = c("#F3F3F3", "#C3F1D5", "#8BE3AF",
                  "#EBC5DD", "#C3C5D5", "#8BC5AF",
                  "#E7A3D1", "#C3A3D1", "#8BA3AE"))

pal
```

##	cat_ed	cat_inc	n	pal
## 1	(0,2.71e+03]	(7.45e+03,2.67e+04]	116	#F3F3F3
## 2	(0,2.71e+03]	(2.67e+04,3.44e+04]	85	#C3F1D5
## 3	(0,2.71e+03]	(3.44e+04,8.24e+04]	70	#8BE3AF
## 4	(2.71e+03,3.95e+03]	(7.45e+03,2.67e+04]	87	#EBC5DD
## 5	(2.71e+03,3.95e+03]	(2.67e+04,3.44e+04]	97	#C3C5D5
## 6	(2.71e+03,3.95e+03]	(3.44e+04,8.24e+04]	92	#8BC5AF
## 7	(3.95e+03,1.02e+04]	(7.45e+03,2.67e+04]	71	#E7A3D1
## 8	(3.95e+03,1.02e+04]	(2.67e+04,3.44e+04]	92	#C3A3D1
## 9	(3.95e+03,1.02e+04]	(3.44e+04,8.24e+04]	113	#8BA3AE

Join & plot

```
bivar_map <- left_join(wider, pal) %>%  
  ggplot() +  
  geom_sf(aes(fill = pal, color = pal)) +  
  guides(fill = "none", color = "none") +  
  scale_fill_identity() +  
  scale_color_identity()
```

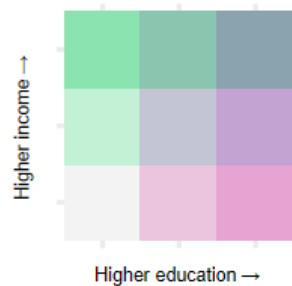


Add in legend

First create it

```
leg <- ggplot(pal, aes(cat_ed, cat_inc)) +  
  geom_tile(aes(fill = pal)) +  
  scale_fill_identity() +  
  coord_fixed() +  
  labs(x = expression("Higher education" %>% ""),  
       y = expression("Higher income" %>% "")) +  
  theme(axis.text = element_blank(),  
        axis.title = element_text(size = 12))
```

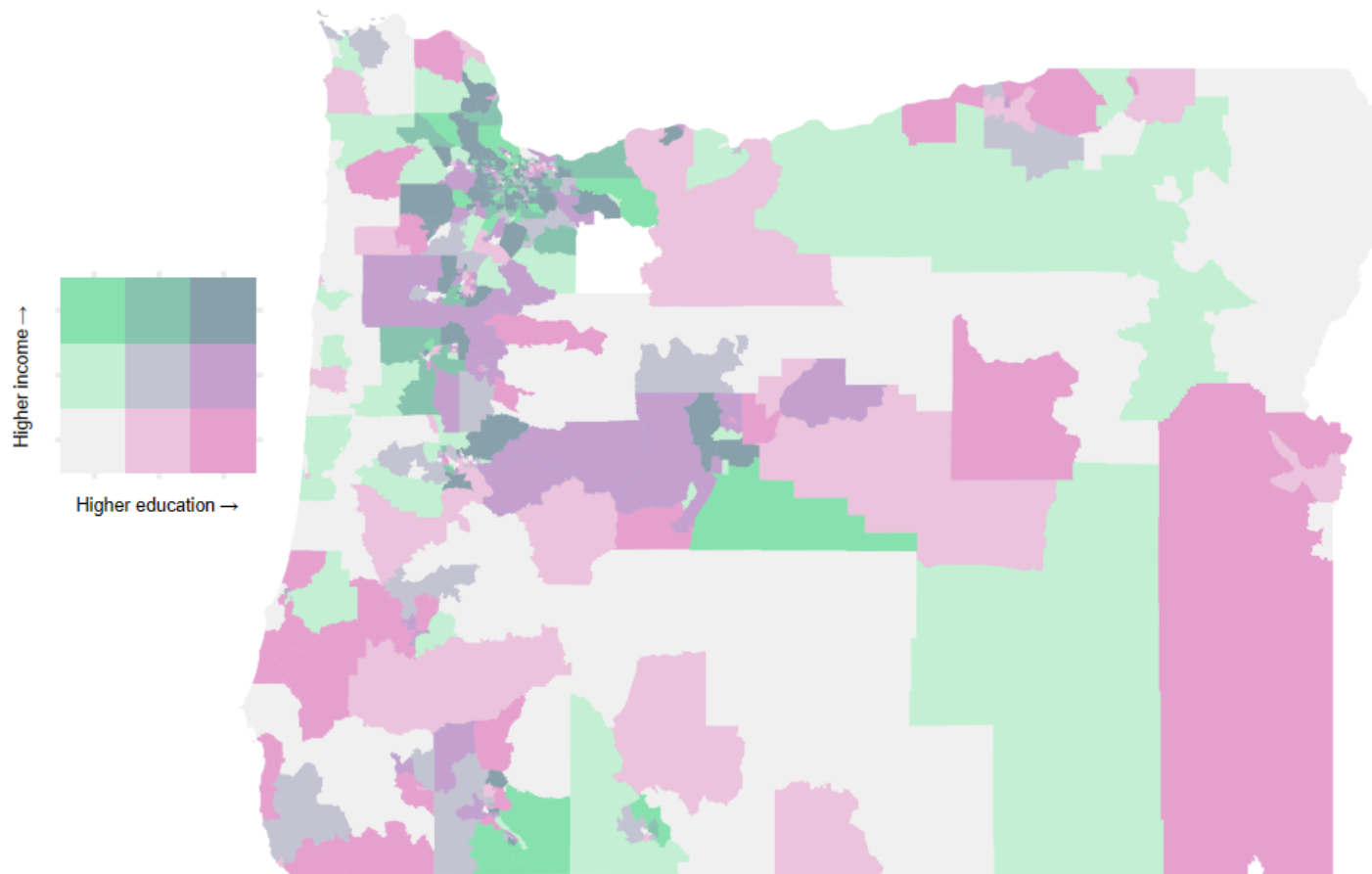
leg



Put together

```
library(cowplot)
ggdraw() +
  draw_plot(bivar_map + theme_void(), 0.1, 0.1, 1, 1) +
  draw_plot(leg, -0.05, 0, 0.3, 0.3)
```

Coordinates are mostly guess/check depending on aspect ratio



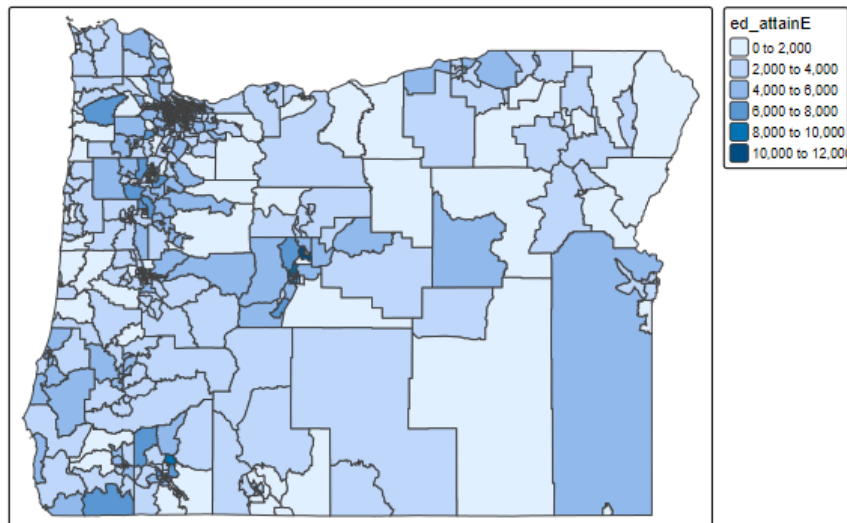
{tmap}

Back to just one variable

I mostly use `ggplot()`, but the **{tmap}** package is really powerful and the syntax is pretty straightforward, so let's have a quick overview.

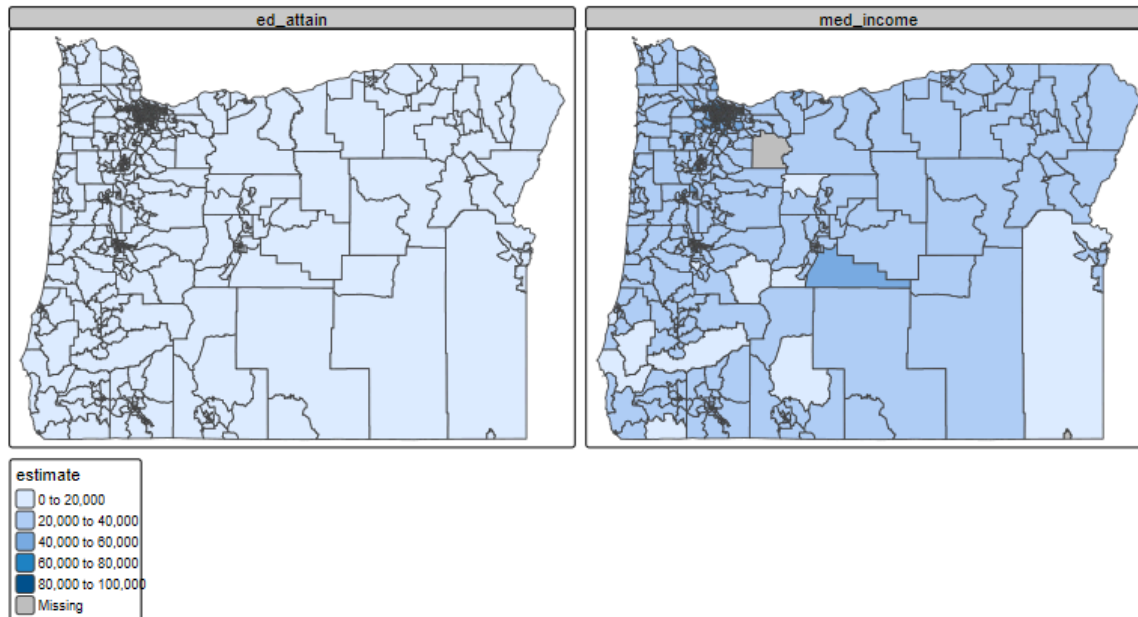
Education map with {tmap}.

```
library(tmap)
tm_shape(wider) +
  tm_polygons("ed_atainE") +
  tm_layout(legend.outside = TRUE)
```



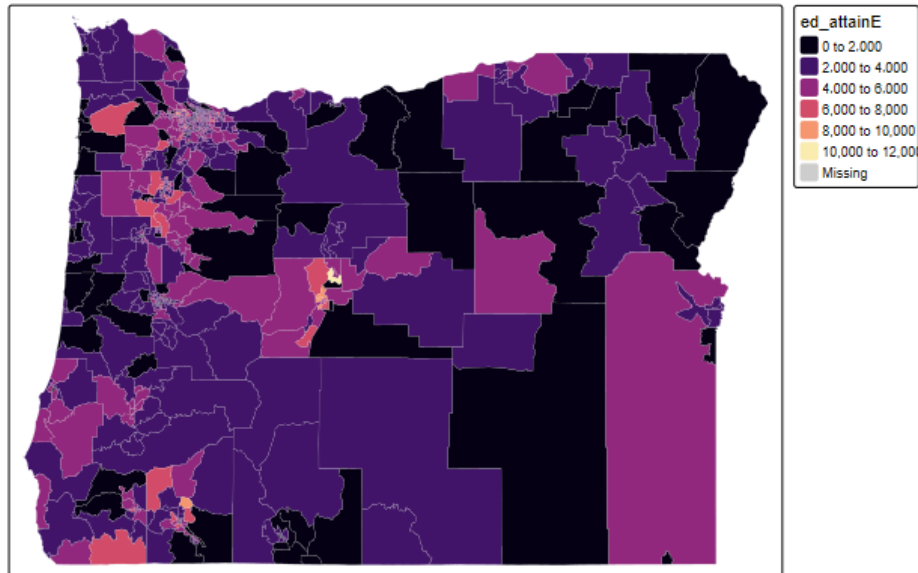
Facet

```
tm_shape(census_vals) +  
  tm_polygons("estimate") +  
  tm_facets("variable") +  
  tm_layout(legend.outside = TRUE)
```



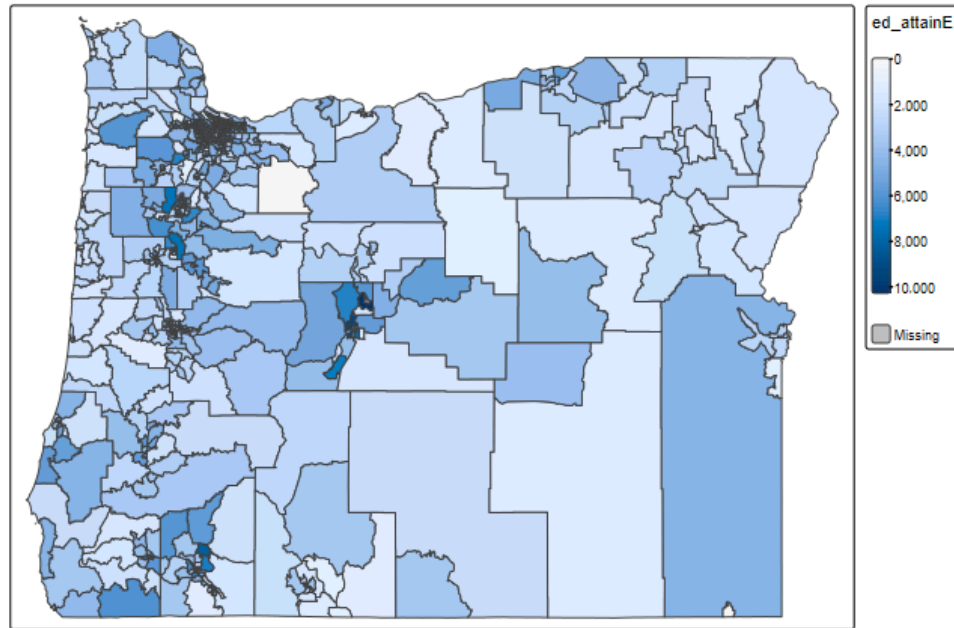
Change colors

```
tm_shape(wider) +  
  tm_polygons("ed_attainE",  
              palette = "magma",  
              border.col = "gray90",  
              lwd = 0.1) +  
  tm_layout(legend.outside = TRUE)
```



Continuous legend

```
tm_shape(wider) +  
  tm_polygons("ed_atainE",  
              style = "cont") +  
  tm_layout(legend.outside = TRUE)
```



Add text

- First, let's get data at the county level, instead of census tract level

```
cnty <- get_acs(  
  geography = "county",  
  state = "OR",  
  variables = c(ed_attain = "B15003_001"),  
  year = 2019,  
  geometry = TRUE  
)
```

##

```
|  
|  
|=====|  
|=====|  
|=====|  
|=====|
```

cnty

Simple feature collection with 36 features and 5 fields

Geometry type: MULTIPOLYGON

Dimension: XY

Bounding box: xmin: -124.5662 ymin: 41.99179 xmax: -116.4635 ymax: 46.29204

Geodetic CRS: NAD83

First 10 features:

##	GEOID	NAME	variable	estimate	moe	
## 1	41017	Deschutes County, Oregon	ed_attain	135615	192	MULTIPOLYGON (((-122.0
## 2	41003	Benton County, Oregon	ed_attain	55359	143	MULTIPOLYGON (((-123.8
## 3	41015	Curry County, Oregon	ed_attain	18304	139	MULTIPOLYGON (((-124.3
## 4	41061	Union County, Oregon	ed_attain	17539	78	MULTIPOLYGON (((-118.6
## 5	41055	Sherman County, Oregon	ed_attain	1251	84	MULTIPOLYGON (((-121.0
## 6	41051	Multnomah County, Oregon	ed_attain	587290	134	MULTIPOLYGON (((-122.9
## 7	41007	Clatsop County, Oregon	ed_attain	28475	165	MULTIPOLYGON (((-123.5
## 8	41033	Josephine County, Oregon	ed_attain	63802	148	MULTIPOLYGON (((-124.0
## 9	41031	Jefferson County, Oregon	ed_attain	16197	24	MULTIPOLYGON (((-121.8
## 10	41039	Lane County, Oregon	ed_attain	256373	147	MULTIPOLYGON (((-124.1

Estimate polygon centroid

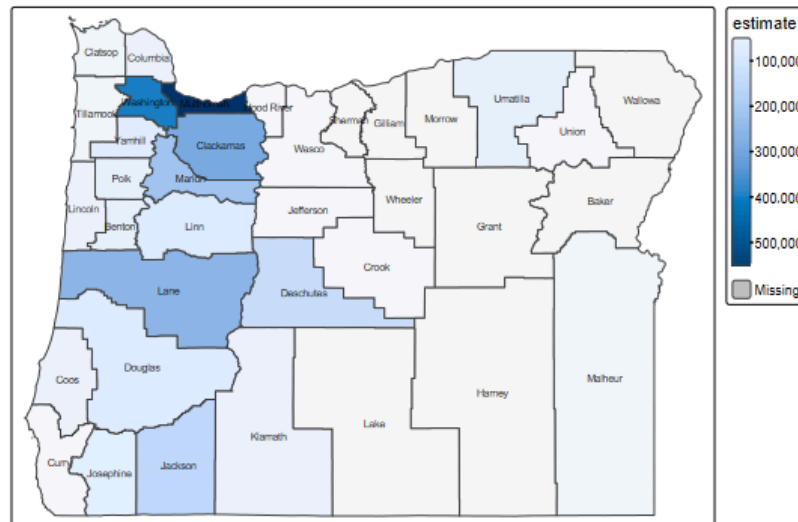
```
centroids <- st_centroid(cnty)
```

Extract just county name

```
centroids <- centroids %>%  
  mutate(county = str_replace_all(NAME, " County, Oregon", ""))
```

Plot

```
tm_shape(cnty) +  
  tm_polygons("estimate",  
              style = "cont") +  
tm_shape(centroids) +  
  tm_text("county", size = 0.5) +  
  tm_layout(legend.outside = TRUE)
```



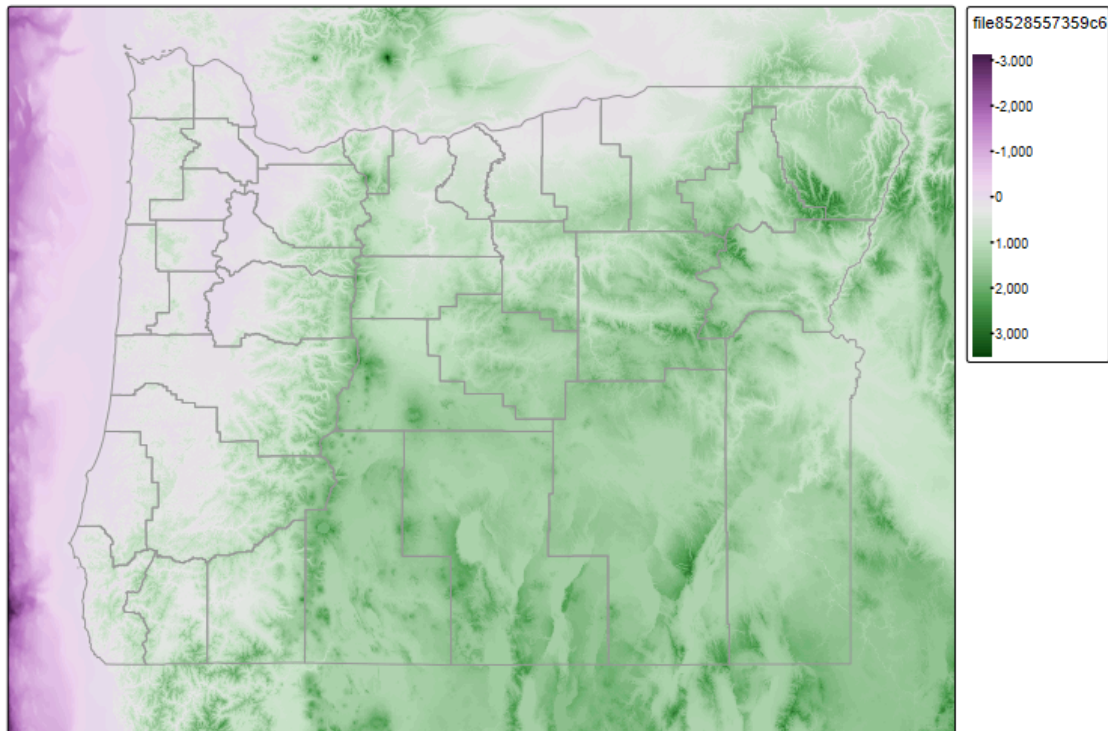
Doesn't work for me on the slides. Not sure why, but should work for you locally.

Add raster elevation data

```
states <- get_acs(  
  "state",  
  variables = c(ed_attain = "B15003_001"),  
  year = 2019,  
  geometry = TRUE  
)  
  
or <- filter(states, NAME == "Oregon")  
  
# convert to spatial data frame  
#sp <- as(or, "Spatial")  
  
# use elevatr library to pull data  
library(elevatr)  
or_elev <- get_elev_raster(or, z = 9)  
lane_elev <- get_elev_raster(or, z = 9)
```

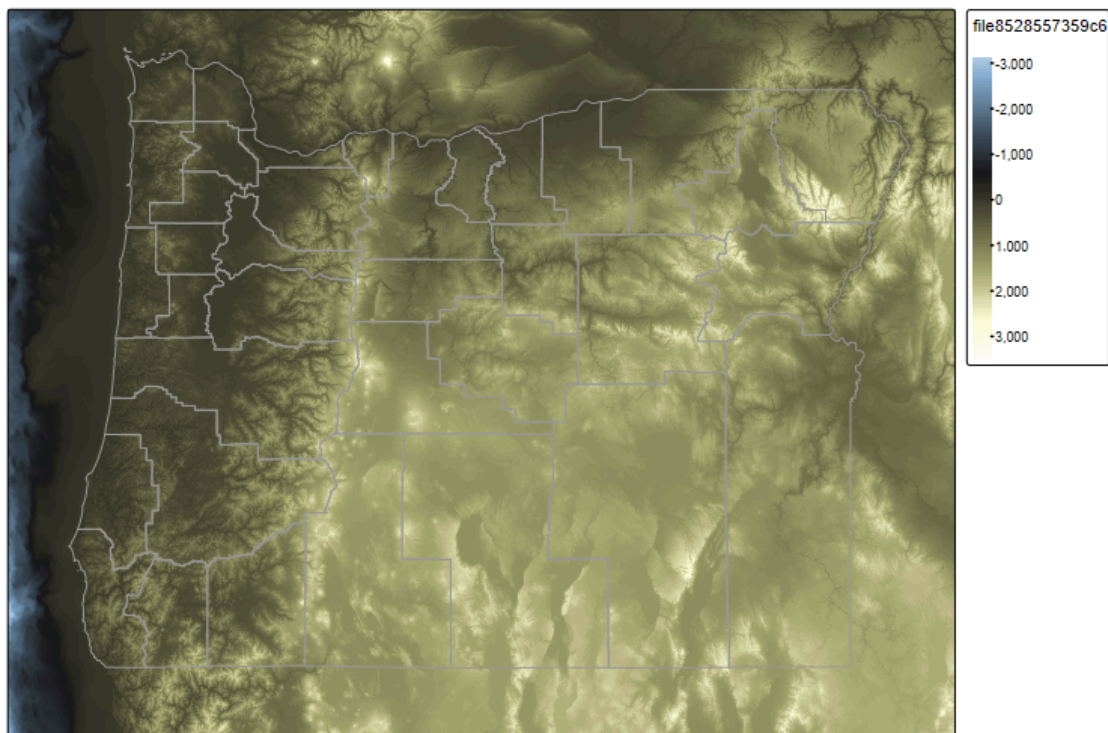
Plot

```
tm_shape(or_elev) +  
  tm_raster(midpoint = NA,  
            style = "cont") +  
  tm_layout(legend.outside = TRUE) +  
tm_shape(cnty) +  
  tm_borders(col = "gray60")
```

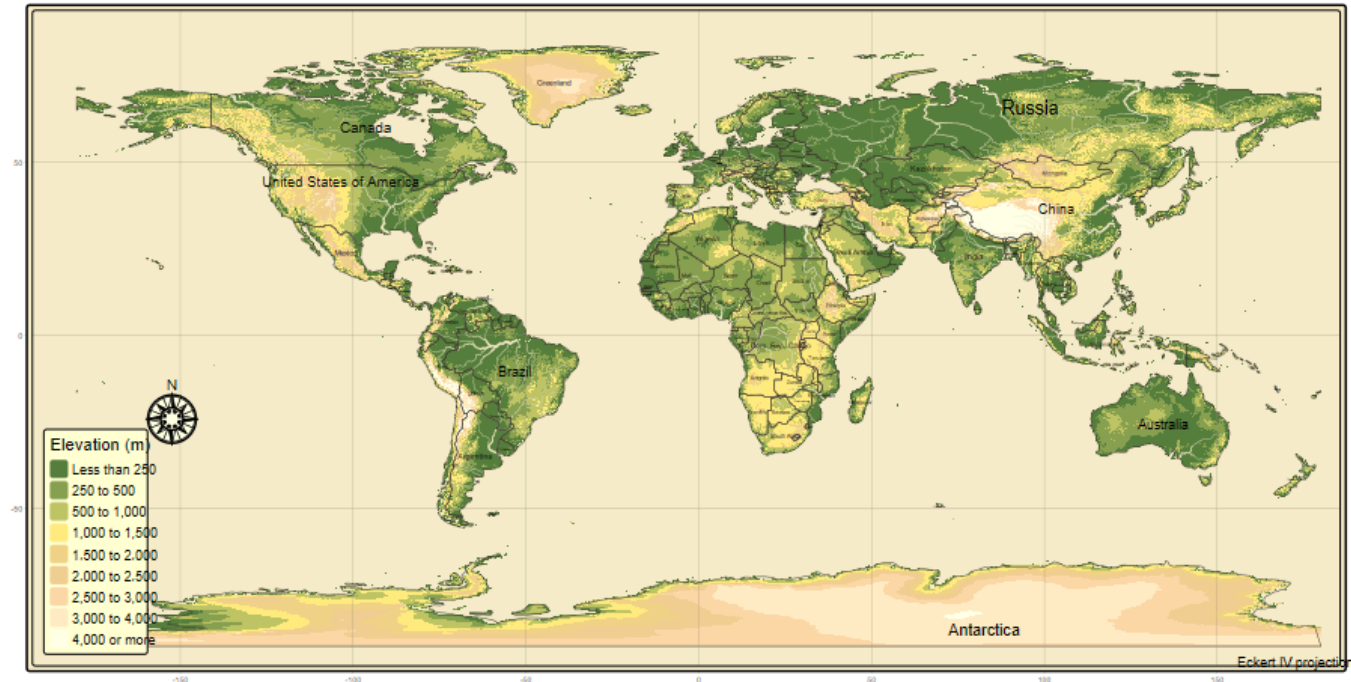


Add custom palette

```
tm_shape(or_elev) +  
  tm_raster(midpoint = NA,  
            style = "cont",  
            palette = c("#E2FCFF", "#83A9CE", "#485C6E",  
                        "#181818", "#5C5B3E", "#AAA971",  
                        "#FCFCD3", "#ffffff")) +  
  tm_layout(legend.outside = TRUE) +  
tm_shape(cnty) +  
  tm_borders(col = "gray60")
```



You can do some amazing things!

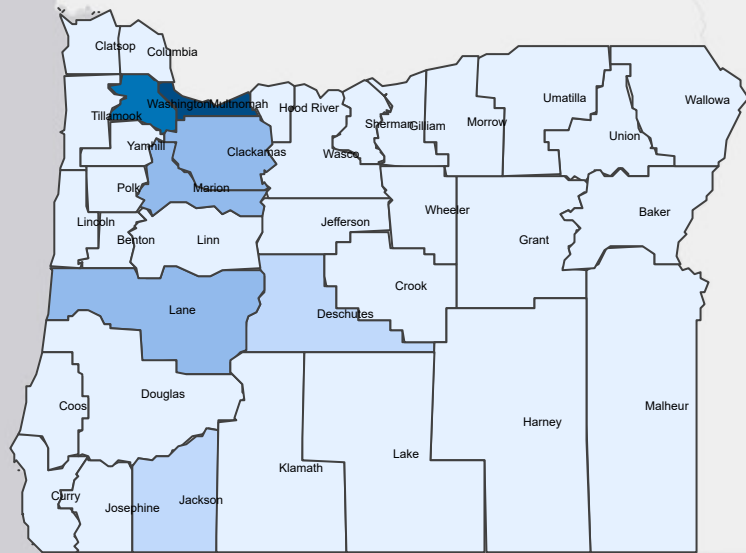


Create interactive maps

Just change run `tmap_mode("view")` then run the same code as before

```
tmap_mode("view")

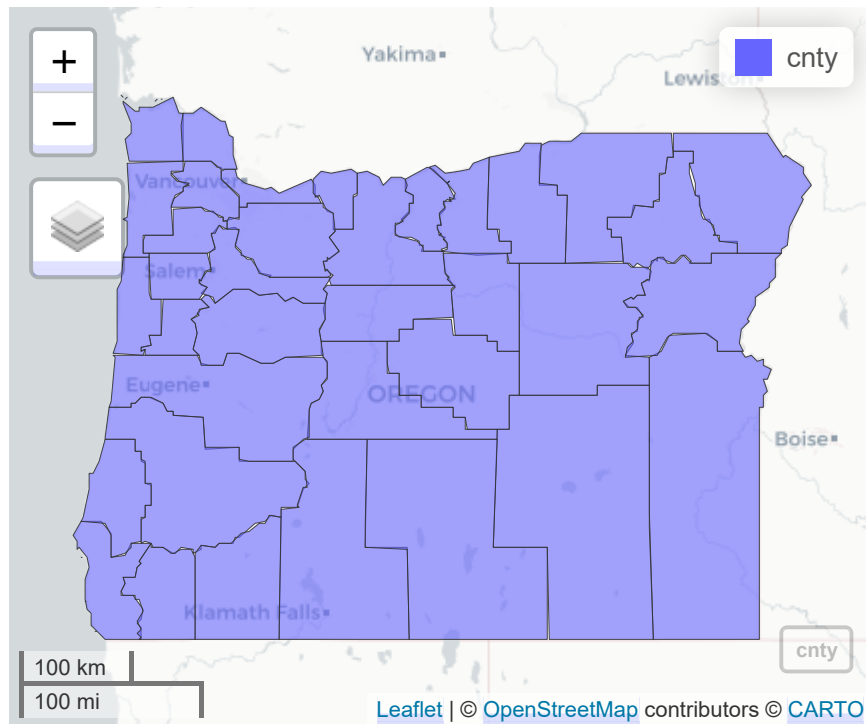
tm_shape(cnty) +
  tm_polygons("estimate") +
tm_shape(centroids) +
  tm_text("county", size = 0.5)
```

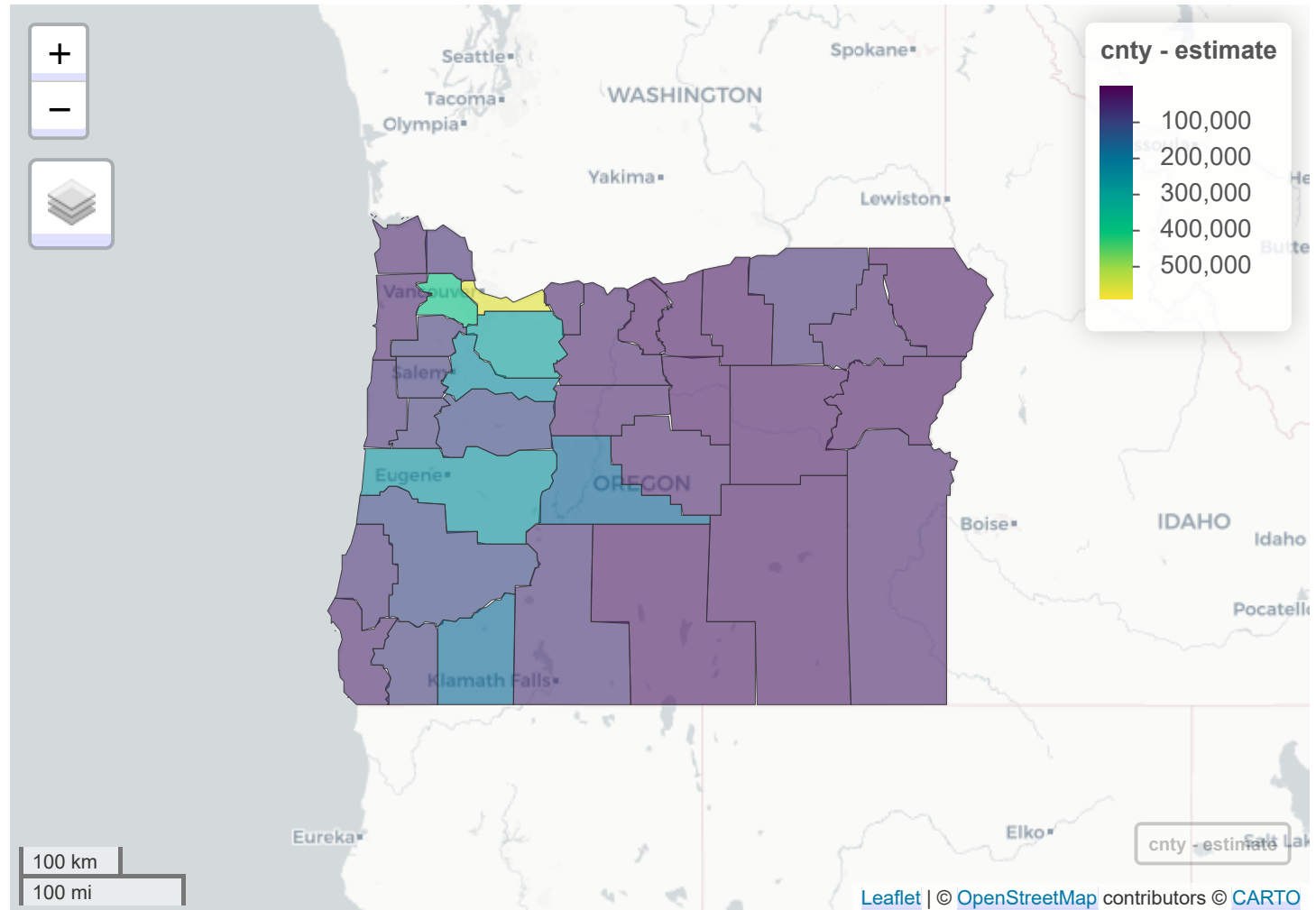
mapview

- Really quick easy interactive maps

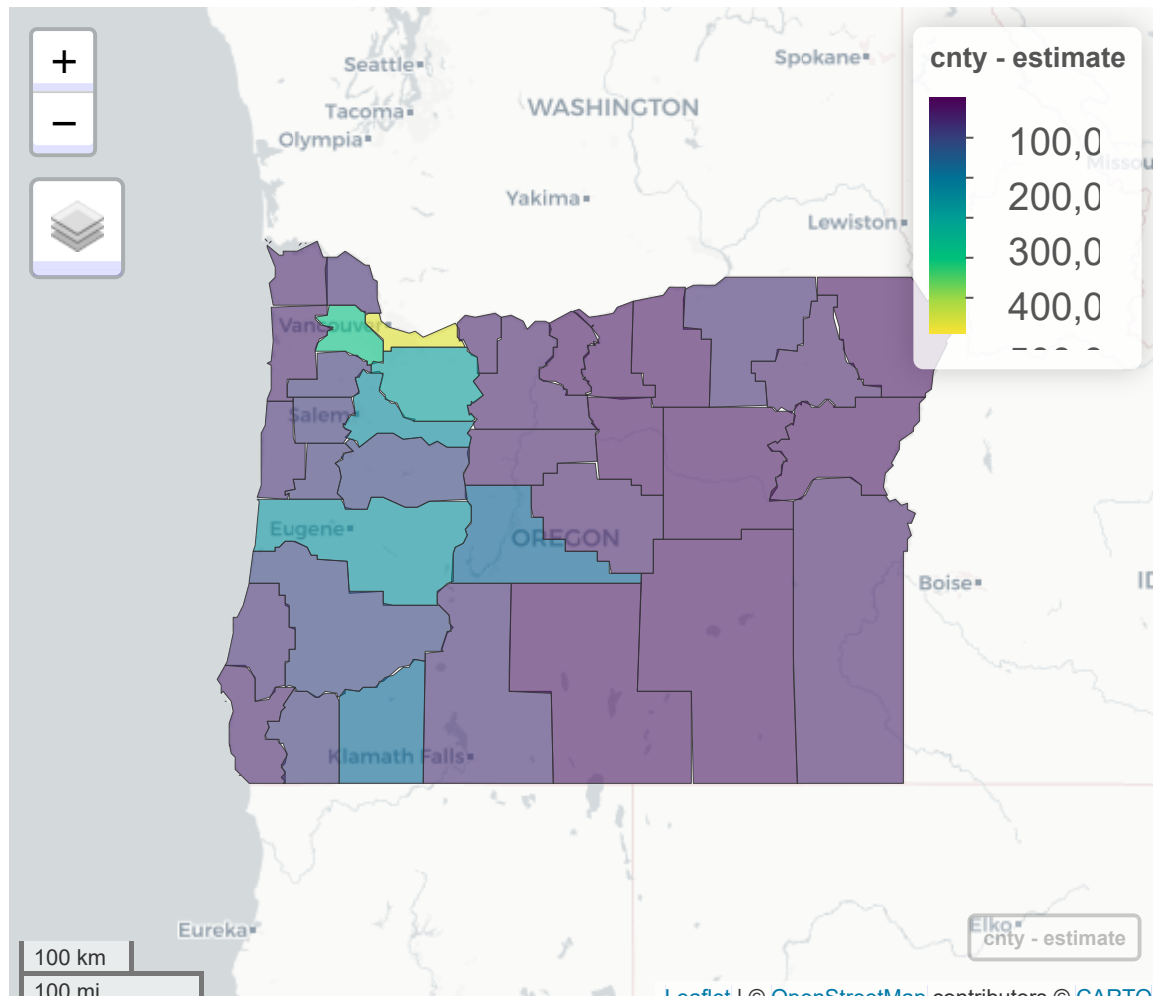
```
library(mapview)  
mapview(cnty)
```



```
mapview(cnty, zcol = "estimate")
```



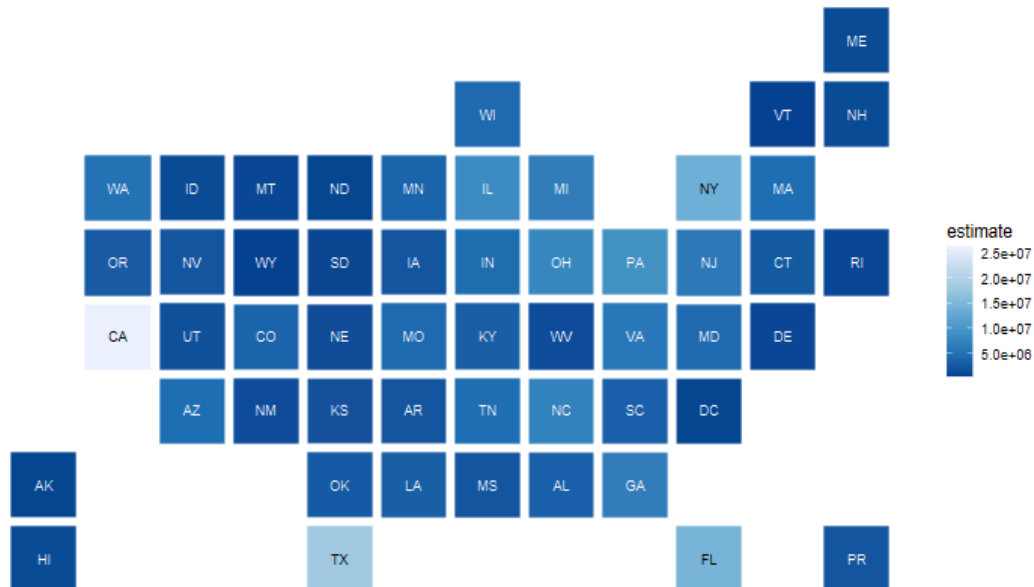
```
mapview(cnty,
  zcol = "estimate",
  popup = leafpop::popupTable(cnty,
    zcol = c("NAME", "estimate"))))
```



A few other
things of
note

statebins

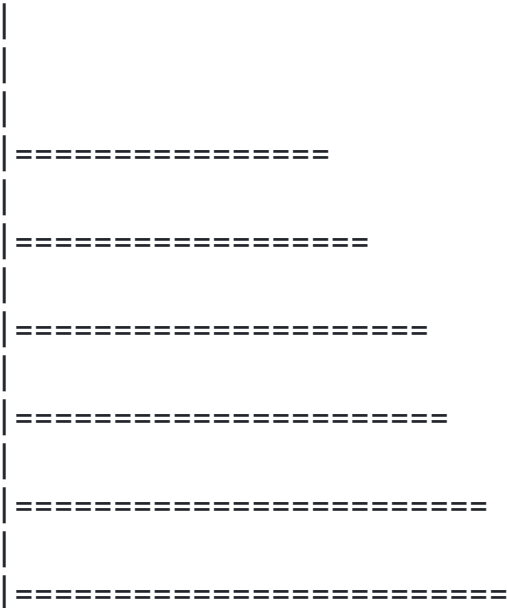
```
library(statebins)
statebins(states,
          state_col = "NAME",
          value_col = "estimate") +
  theme_void()
```



Cartograms

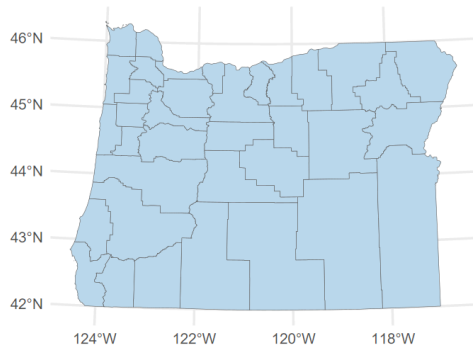
```
library(cartogram)
or_county_pop <- get_acs(
  geography = "county",
  state = "OR",
  variables = "B01001E_001",
  year = 2018,
  geometry = TRUE
)
```

##

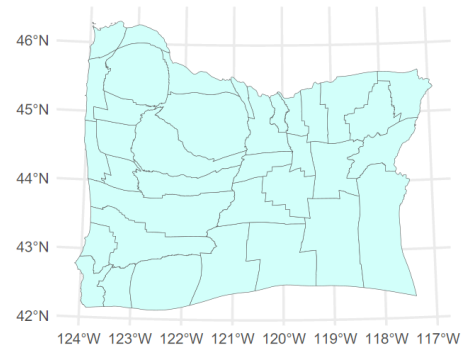


Compare

```
ggplot(or_county_pop) +  
  geom_sf(fill = "#BCD8EB")
```



```
ggplot(carto_counties) +  
  geom_sf(fill = "#D5FFFA")
```

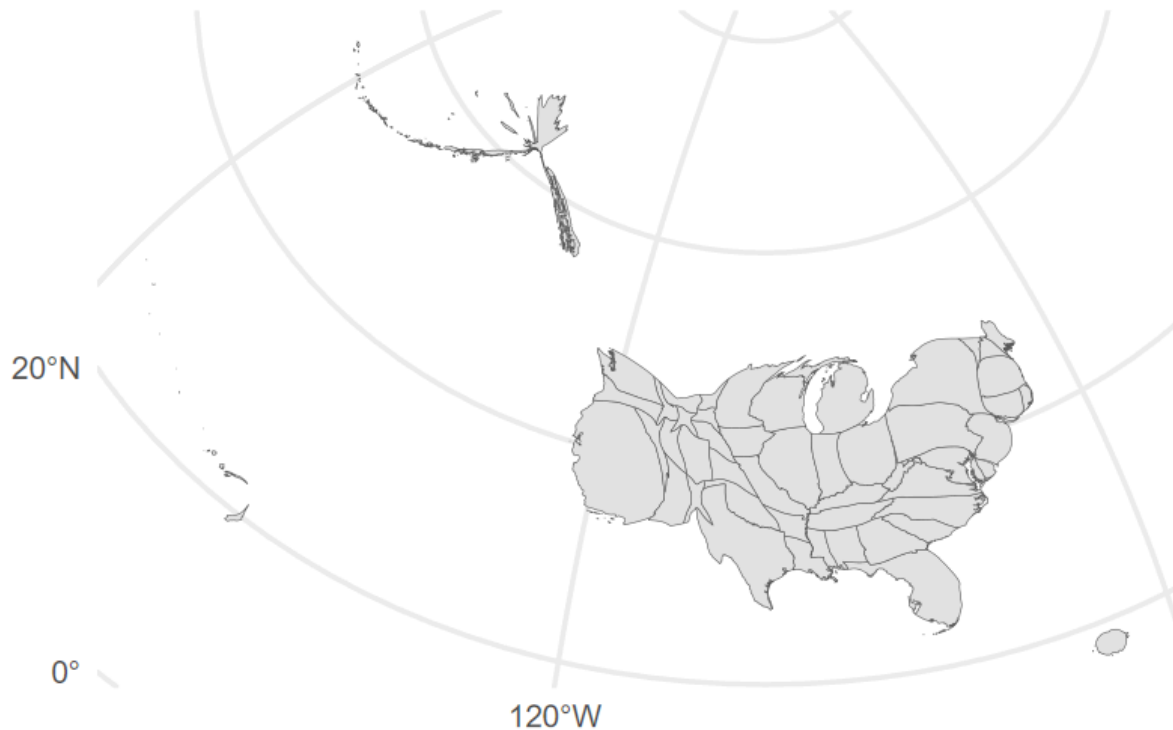


State

```
state_pop <- get_acs(  
  geography = "state",  
  variables = "B00001_001",  
  year = 2018,  
  geometry = TRUE  
)  
  
# Set projection  
state_pop <- st_transform(state_pop, crs = 2163)  
  
# found the CRS here: https://epsg.io/transform#s\_srs=3969&t\_srs=  
carto_states <- cartogram_cont(state_pop, "estimate")
```

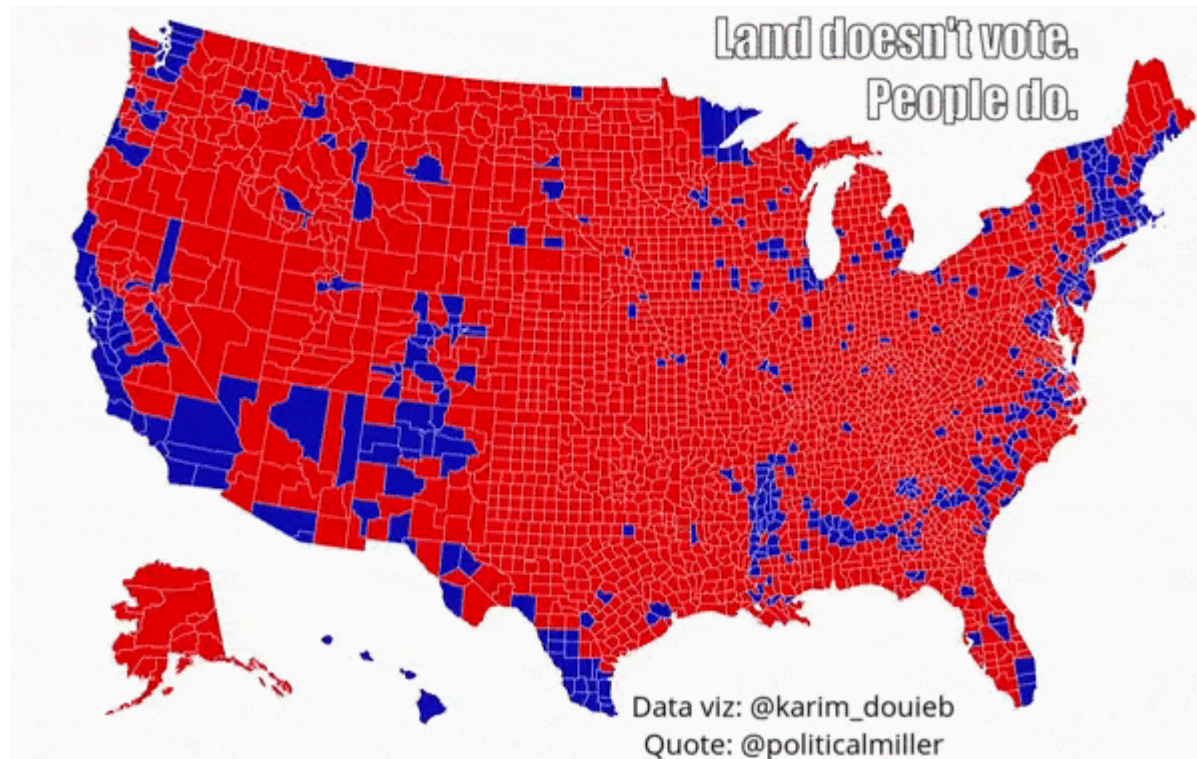
Cartogram of USA by population

```
ggplot(carto_states) +  
  geom_sf()
```



Last note

You may or may not like cartograms. Just be careful not to lie with maps.



Loose ends with HTML/Websites

- Can I use Quatro blog instead of a Distill blog?
- Best way to learn more about websites - Fork, clone, mess on your own
- Github pages deployment is the easiest and a good entry point
- But you can use others like Netlify and such for more complicated sites
- Troubleshooting and persistence through that is the key for any skill but especially so for programming!
- So, keep at it!

Presentations
for next
week - let's
quickly look
at rubric

Presentations

- Each person/group gets 10 minutes
- Followed by Q & A and a quick 2-minute discussion of plans to finalize and what finishing touches you will add!