

Engaging and Effective ggplot (2)

VCS, Rice 2025

Kieran Healy
Duke University

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Polishing your plots
and Presenting them

Load our packages

```
library(here)      # manage file paths
library(tidyverse) # your friend and mine
library(socviz)    # data and some useful functions
library(ggrepel)   # Text and labels
library(colorspace) # luminance-balanced palettes
library(scales)     # scale adjustments and enhancements
library(ggforce)   # useful enhancements to ggplot
```

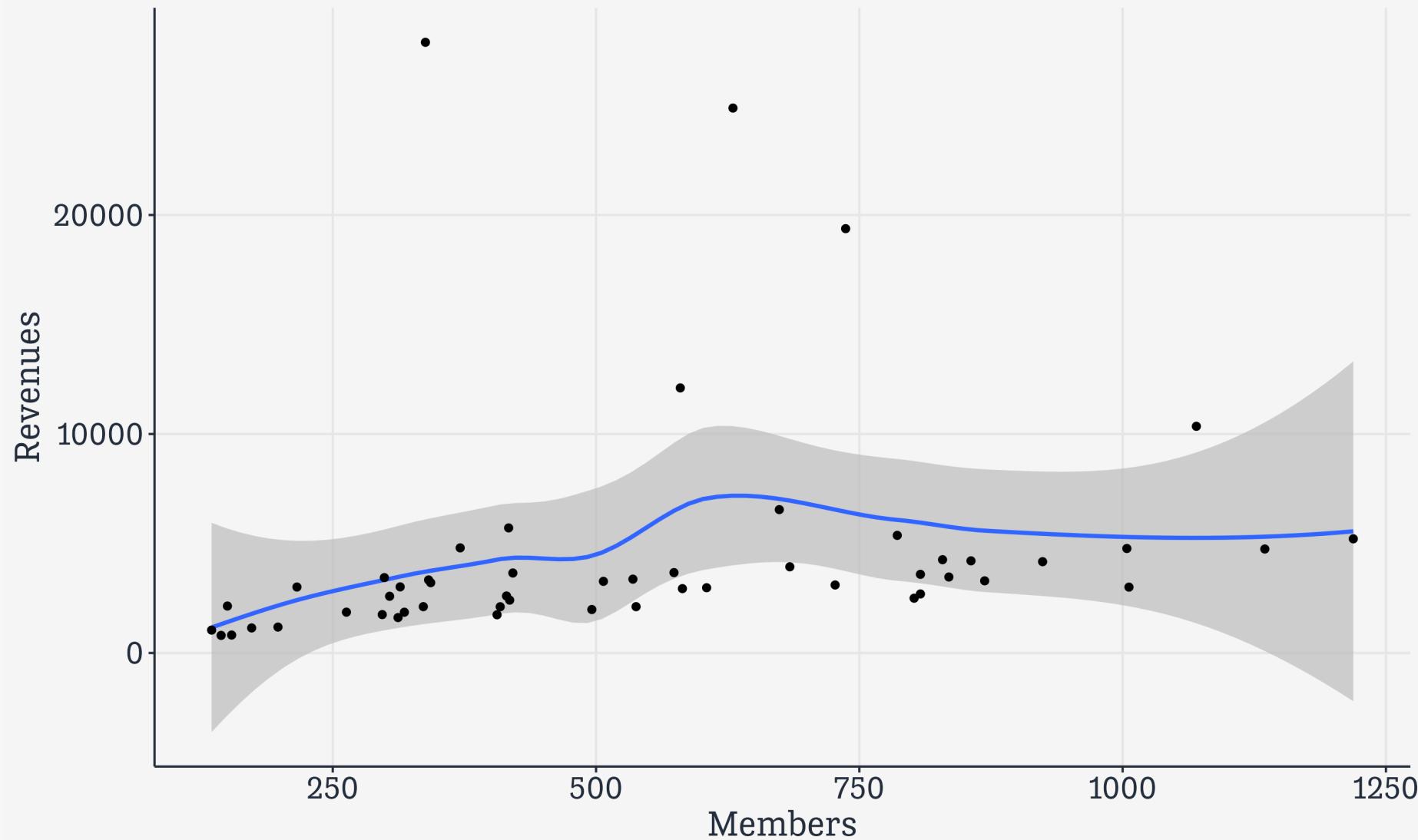
**Piece by piece,
Layer by layer**

Build your plots a piece at a time

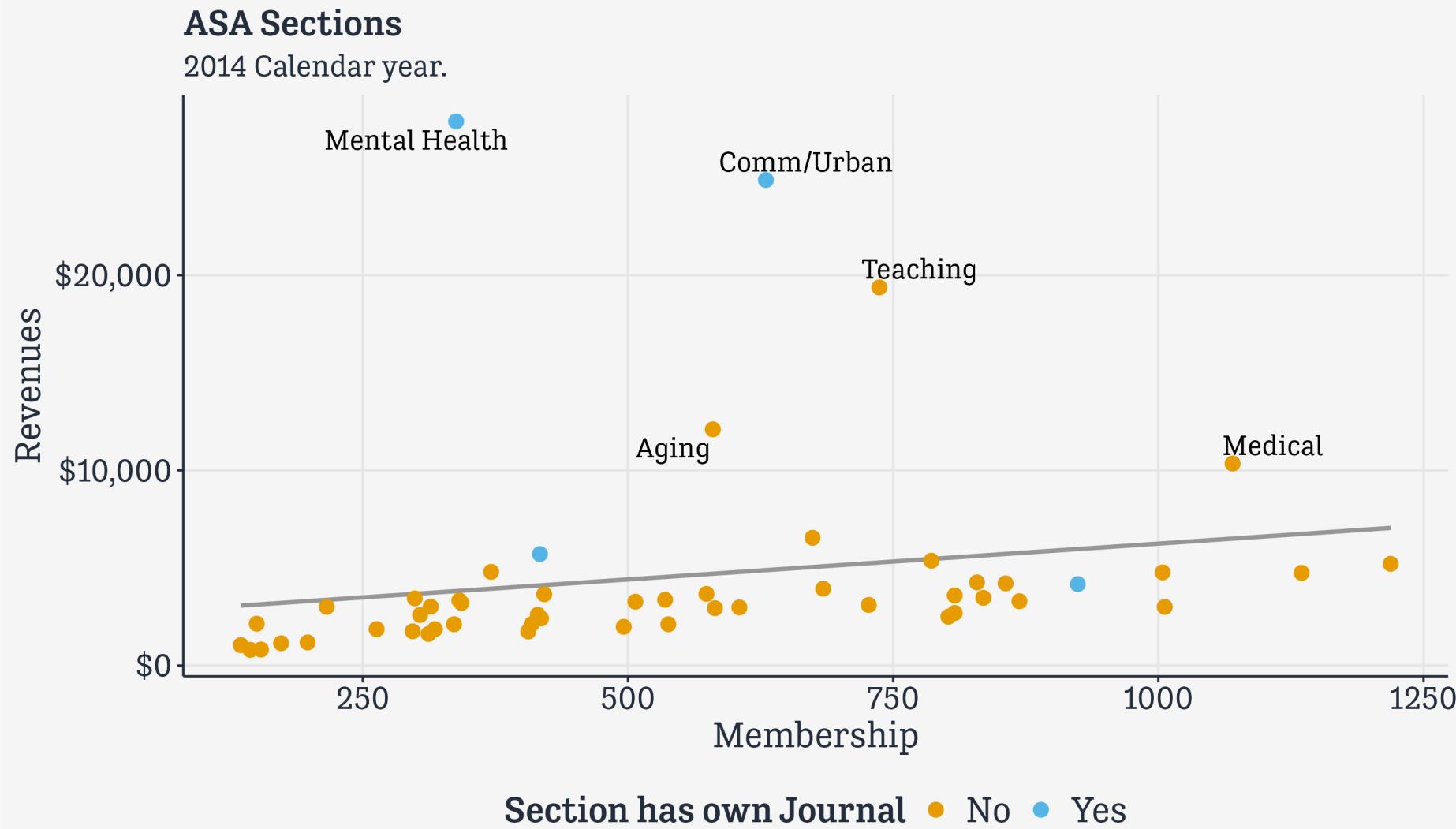
```
asasec ← as_tibble(asasec)
asasec
```

```
# A tibble: 572 × 9
  Section      Sname Beginning Revenues Expenses Ending Journal Year Members
  <fct>       <fct>    <int>     <int>    <int>   <int> <fct>    <int>    <int>
1 Aging and the... Aging     12752    12104    12007   12849 No        2005      598
2 Alcohol, Drug... Alco...    11933     1144      400    12677 No        2005      301
3 Altruism and ... Altr...    1139      1862     1875    1126 No        2005      NA
4 Animals and S... Anim...     473       820     1116     177 No        2005      209
5 Asia/Asian Am... Asia      9056      2116     1710    9462 No        2005      365
6 Body and Embo... Body     3408      1618     1920    3106 No        2005      NA
7 Children and ... Chil...    3692      3653     3713    3632 No        2005      418
8 Coll Behavior... CBSM     8127      3470     2704    8893 No        2005      708
9 Communication... CITA...   17093     4800     4804   17089 No        2005      301
10 Community and... Comm...   26598    24883    23379   28102 Yes       2005      721
# i 562 more rows
```

Build your plots a piece at a time



Build your plots a piece at a time



Source: ASA annual report.

Build your plots a piece at a time

```
df_rich ← asasec ▷ filter(Year == 2014 &  
                           Revenues > 7000)
```

Build your plots a piece at a time

```
df_rich ← asasec ▷ filter(Year = 2014 &  
                           Revenues > 7000)
```

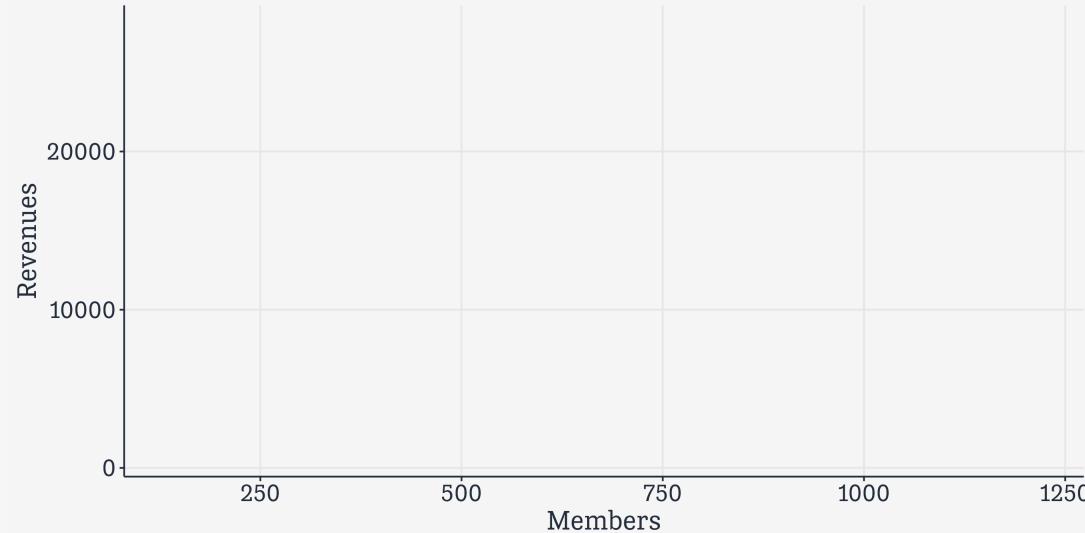
Build your plots a piece at a time

```
df_rich ← asasec ▷ filter(Year = 2014 &  
                           Revenues > 7000)  
  
asasec ▷  
  filter(Year = 2014)
```

	Section	Sname	Beginning	Revenues	Expenses	Ending	Journal	Year	Members
	<fct>	<fct>	<int>	<int>	<int>	<int>	<fct>	<int>	<int>
1	Aging and the... Aging		12752	12104	12007	12849	No	2014	580
2	Alcohol, Drug... Alco...		11933	1144	400	12677	No	2014	173
3	Altruism and ... Altr...		1139	1862	1875	1126	No	2014	318
4	Animals and S... Anim...		473	820	1116	177	No	2014	154
5	Asia/Asian Am... Asia		9056	2116	1710	9462	No	2014	336
6	Body and Embo... Body		3408	1618	1920	3106	No	2014	312
7	Children and ... Chil...		3692	3653	3713	3632	No	2014	421
8	Coll Behavior... CBSM		8127	3470	2704	8893	No	2014	835
9	Communication... CITA...		17093	4800	4804	17089	No	2014	371
10	Community and... Comm...		26598	24883	23379	28102	Yes	2014	630

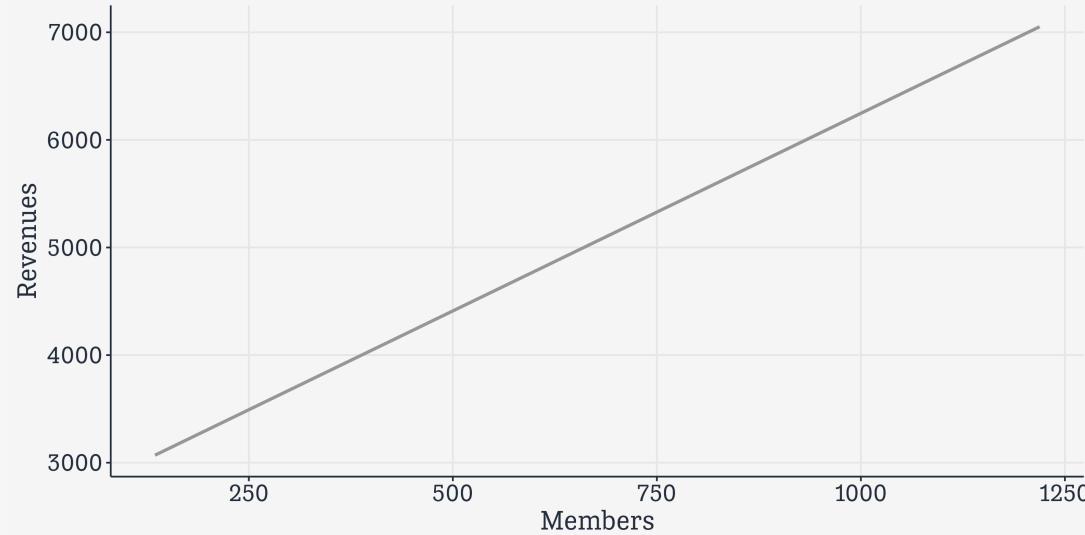
Build your plots a piece at a time

```
df_rich ← asasec ▷ filter(Year = 2014 &  
                           Revenues > 7000)  
  
asasec ▷  
  filter(Year = 2014) ▷  
  ggplot(mapping = aes(x = Members,  
                        y = Revenues,  
                        label = Sname))
```



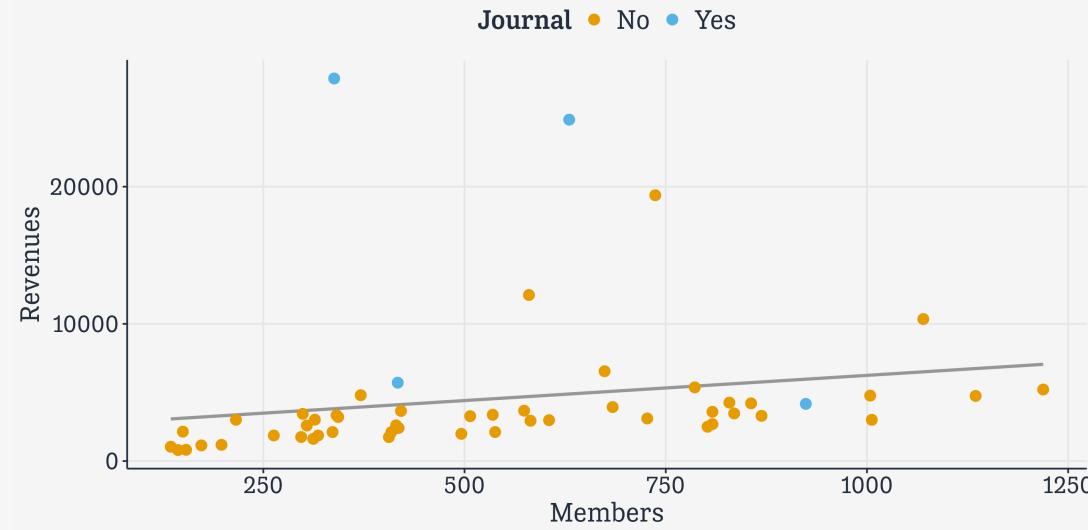
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```
df_rich ← asasec ▷ filter(Year = 2014 &  
                           Revenues > 7000)  
  
asasec ▷  
  filter(Year = 2014) ▷  
  ggplot(mapping = aes(x = Members,  
                        y = Revenues,  
                        label = Sname)) +  
  geom_smooth(method = "lm",  
              se = FALSE,  
              color = "gray60")
```



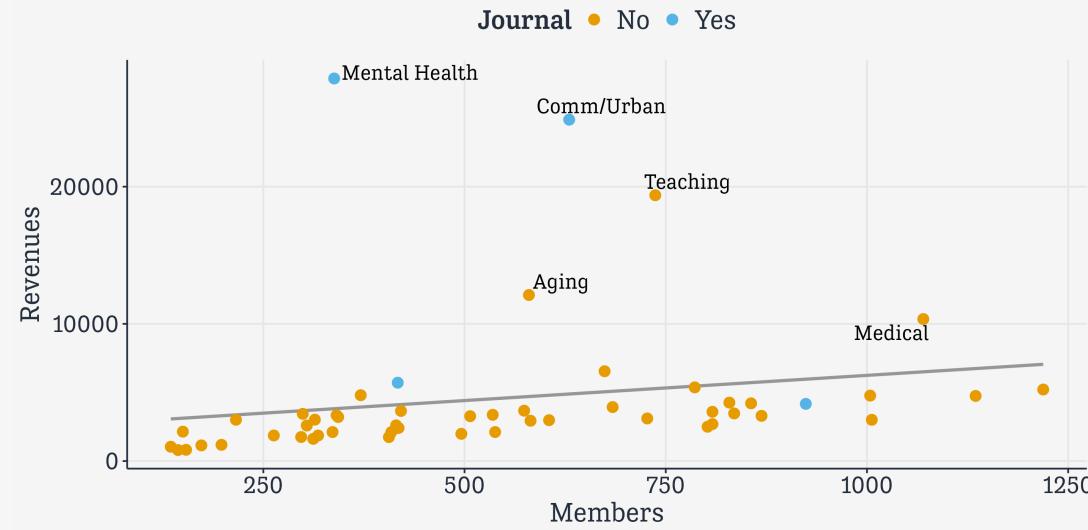
Build your plots a piece at a time

```
df_rich ← asasec ▷ filter(Year = 2014 &  
                           Revenues > 7000)  
  
asasec ▷  
  filter(Year = 2014) ▷  
  ggplot(mapping = aes(x = Members,  
                        y = Revenues,  
                        label = Sname)) +  
  geom_smooth(method = "lm",  
              se = FALSE,  
              color = "gray60") +  
  geom_point(mapping = aes(color = Journal),  
             size = rel(3))
```



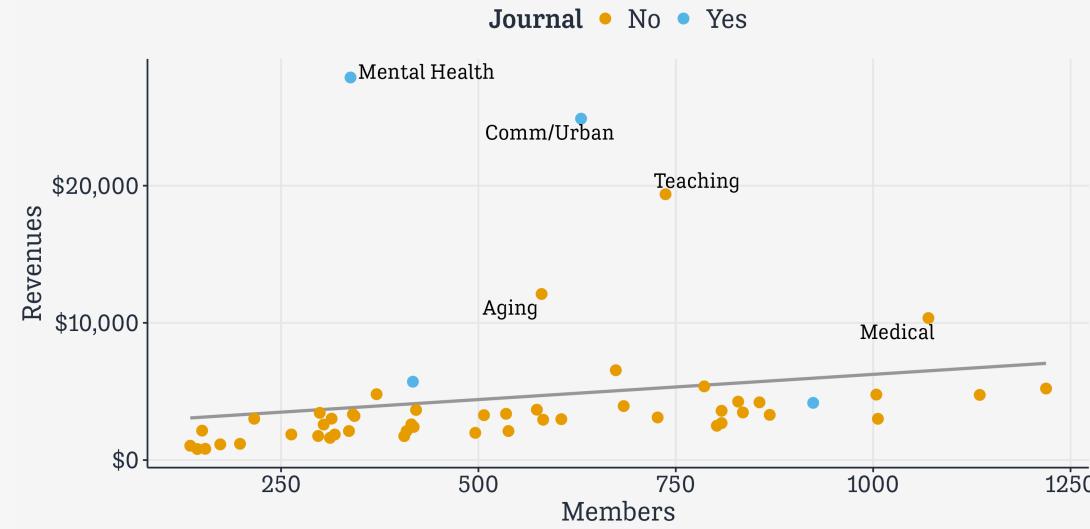
Build your plots a piece at a time

```
df_rich ← asasec %>% filter(Year = 2014 &  
  Revenues > 7000)  
  
asasec %>  
  filter(Year = 2014) %>  
  ggplot(mapping = aes(x = Members,  
    y = Revenues,  
    label = Sname)) +  
  geom_smooth(method = "lm",  
    se = FALSE,  
    color = "gray60") +  
  geom_point(mapping = aes(color = Journal),  
    size = rel(3)) +  
  geom_text_repel(data = df_rich,  
    size = rel(5),  
    mapping =  
    aes(family = "Tenso Slide"))
```



Build your plots a piece at a time

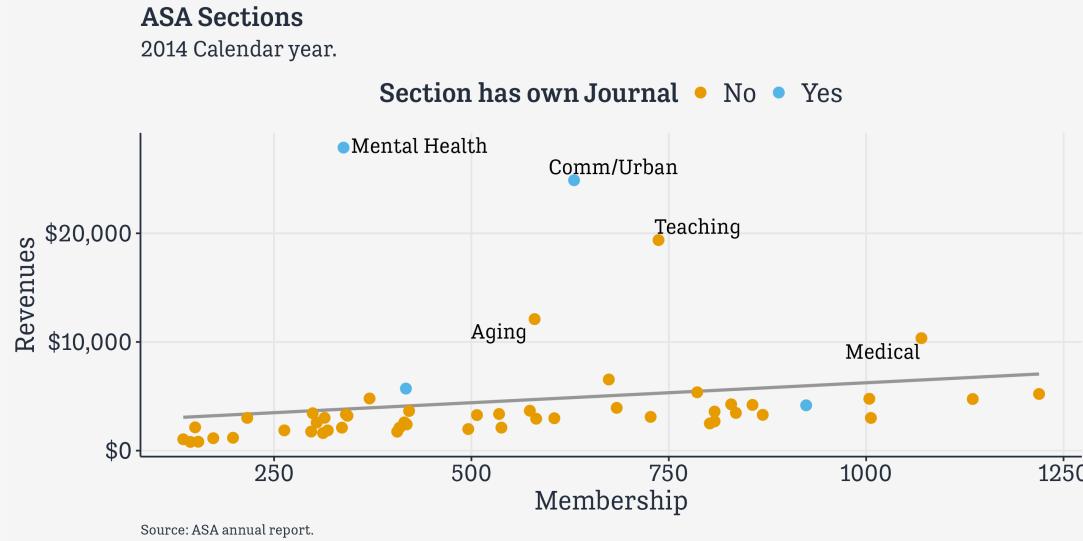
```
df_rich ← asasec %>% filter(Year = 2014 &  
  Revenues > 7000)  
  
asasec %>  
  filter(Year = 2014) %>  
  ggplot(mapping = aes(x = Members,  
    y = Revenues,  
    label = Sname)) +  
  geom_smooth(method = "lm",  
    se = FALSE,  
    color = "gray60") +  
  geom_point(mapping = aes(color = Journal),  
    size = rel(3)) +  
  geom_text_repel(data = df_rich,  
    size = rel(5),  
    mapping =  
      aes(family = "Tenso Slide")) +  
  scale_y_continuous(labels =  
    scales::label_dollar())
```



Build your plots a piece at a time

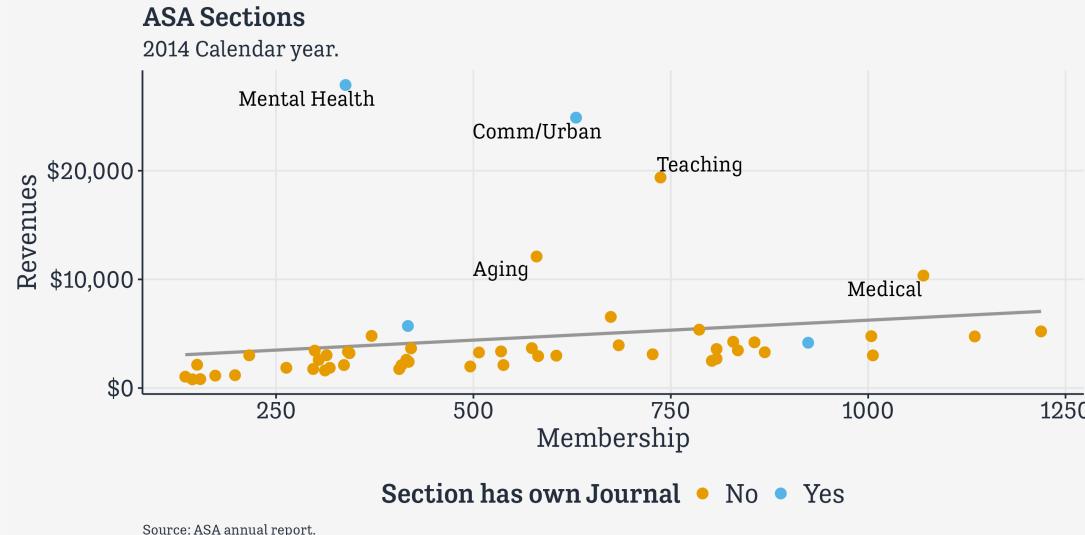
```
df_rich ← asasec %> filter(Year = 2014 &
  Revenues > 7000)

asasec %>
  filter(Year = 2014) %>
  ggplot(mapping = aes(x = Members,
    y = Revenues,
    label = Sname)) +
  geom_smooth(method = "lm",
    se = FALSE,
    color = "gray60") +
  geom_point(mapping = aes(color = Journal),
    size = rel(3)) +
  geom_text_repel(data = df_rich,
    size = rel(5),
    mapping =
      aes(family = "Tenso Slab")) +
  scale_y_continuous(labels =
    scales::label_dollar()) +
  labs(x="Membership", y="Revenues",
    color = "Section has own Journal",
    title = "ASA Sections",
    subtitle = "2014 Calendar year.",
    caption = "Source: ASA annual report.")
```



Build your plots a piece at a time

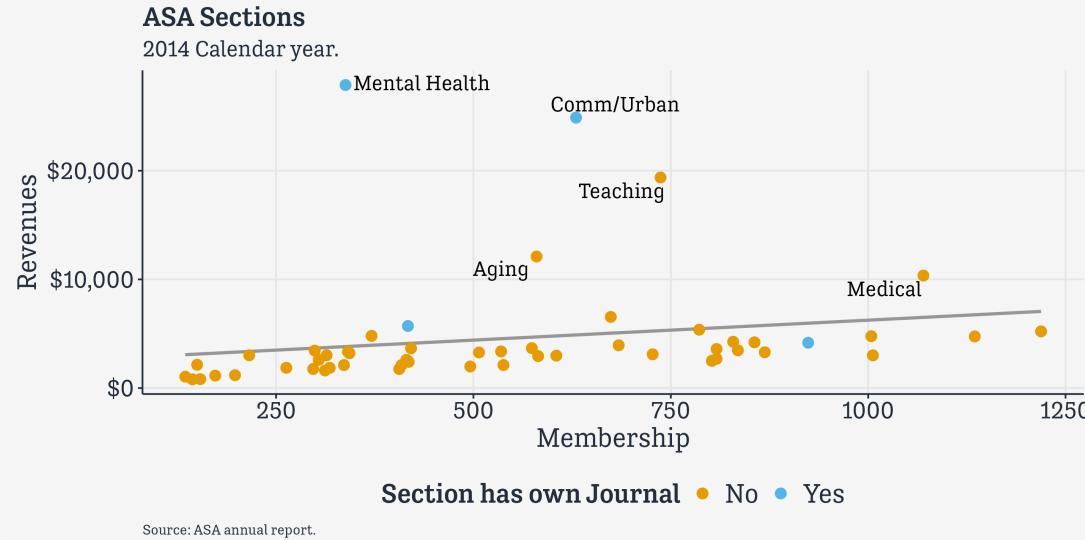
```
df_rich ← asasec %> filter(Year == 2014 &  
                           Revenues > 7000)  
  
asasec %>  
  filter(Year == 2014) %>  
  ggplot(mapping = aes(x = Members,  
                        y = Revenues,  
                        label = Sname)) +  
  geom_smooth(method = "lm",  
              se = FALSE,  
              color = "gray60") +  
  geom_point(mapping = aes(color = Journal),  
             size = rel(3)) +  
  geom_text_repel(data = df_rich,  
                 size = rel(5),  
                 mapping =  
                   aes(family = "Tenso Slide")) +  
  scale_y_continuous(labels =  
    scales::label_dollar()) +  
  labs(x = "Membership", y = "Revenues",  
       color = "Section has own Journal",  
       title = "ASA Sections",  
       subtitle = "2014 Calendar year.",  
       caption = "Source: ASA annual report.") +  
  theme(legend.position = "bottom")
```



Build your plots a piece at a time

```
df_rich ← asasec %> filter(Year = 2014 &
  Revenues > 7000)

asasec %>
  filter(Year = 2014) %>
  ggplot(mapping = aes(x = Members,
    y = Revenues,
    label = Sname)) +
  geom_smooth(method = "lm",
    se = FALSE,
    color = "gray60") +
  geom_point(mapping = aes(color = Journal),
    size = rel(3)) +
  geom_text_repel(data = df_rich,
    size = rel(5),
    mapping =
      aes(family = "Tenso Slide")) +
  scale_y_continuous(labels =
    scales::label_dollar()) +
  labs(x="Membership", y="Revenues",
    color = "Section has own Journal",
    title = "ASA Sections",
    subtitle = "2014 Calendar year.",
    caption = "Source: ASA annual report.") +
  theme(legend.position = "bottom")
```



**Layer color and text
to your advantage**

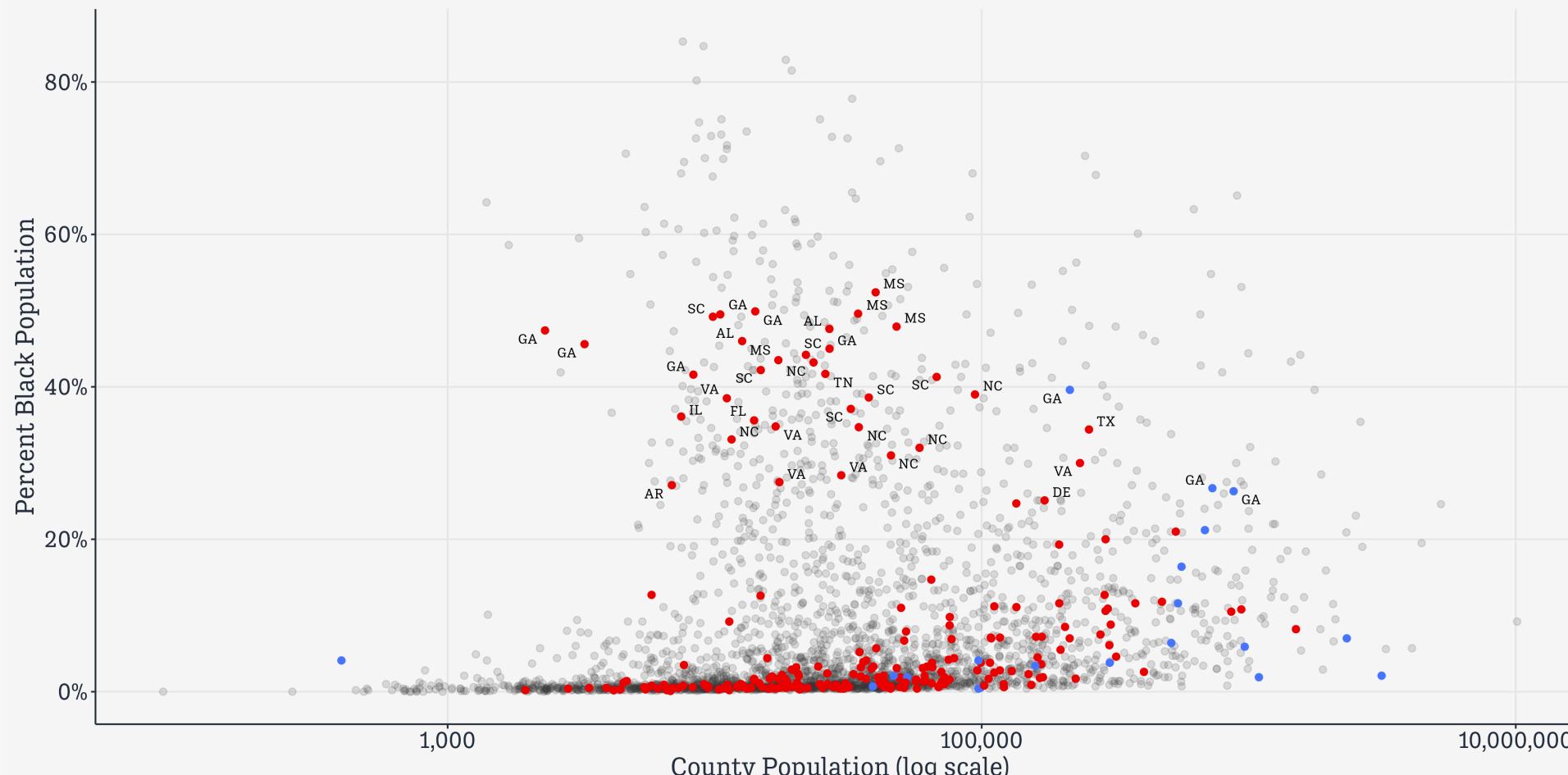
Some 2016 Election Data

```
county_data %>  
  select(name, state, pop, black, partywinner16, flipped) %>  
  sample_n(10)
```

		name	state	pop	black	partywinner16	flipped
1		Racine County	WI	195163	11.6	Republican	Yes
2		Monroe County	AL	21947	41.3	Republican	No
3		Owyhee County	ID	11353	0.9	Republican	No
4		Norton city	VA	4031	6.7	Republican	No
5		Worcester County	MD	51675	13.9	Republican	No
6		New London County	CT	273676	6.8	Democrat	No
7		Clermont County	OH	201560	1.4	Republican	No
8		Chattooga County	GA	24939	10.4	Republican	No
9		Madison County	KY	87340	4.5	Republican	No
10		Westmoreland County	PA	359320	2.5	Republican	No

Flipped counties, 2016

County flipped to ... • Democrat • Republican



We build it a piece at a time

```
## Brighter Blue and Red  
party_colors ← c("royalblue1", "red2")
```

We build it a piece at a time

```
## Brighter Blue and Red
party_colors ← c("royalblue1", "red2")

## Data subsets for our geoms
df_noflip ← county_data %>% filter(flipped == "No")
```

We build it a piece at a time

```
## Brighter Blue and Red
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df_noflip ← county_data %>% filter(flipped == "No")
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```

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```
## Brighter Blue and Red
party_colors ← c("royalblue1", "red2")

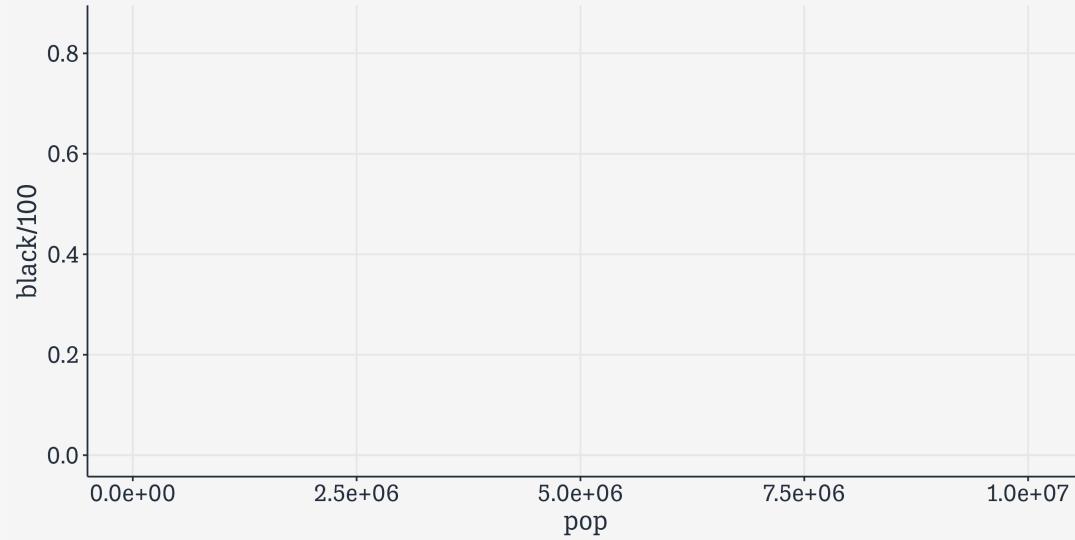
## Data subsets for our geoms
df_noflip ← county_data %>% filter(flipped == "No")
df_flip ← county_data %>% filter(flipped == "Yes")
df_text ← county_data %>% filter(flipped == "Yes" & black
```

We build it a piece at a time

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df_noflip ← county_data %>% filter(flipped = "No")
df_flip ← county_data %>% filter(flipped = "Yes")
df_text ← county_data %>% filter(flipped = "Yes" & black > 0)

ggplot(data = df_noflip,
       mapping = aes(x = pop,
                      y = black/100))
```

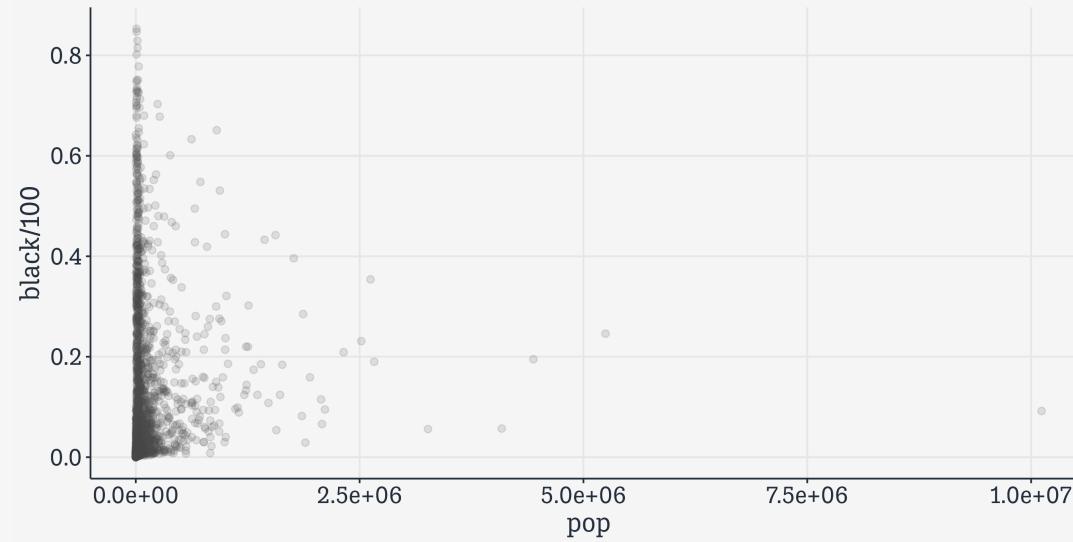


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df_noflip <- county_data %>% filter(flipped = "No")
df_flip <- county_data %>% filter(flipped = "Yes")
df_text <- county_data %>% filter(flipped = "Yes" & black

ggplot(data = df_noflip,
       mapping = aes(x = pop,
                      y = black/100)) +
  geom_point(alpha = 0.15,
             color = "gray30", size = rel(2))
```

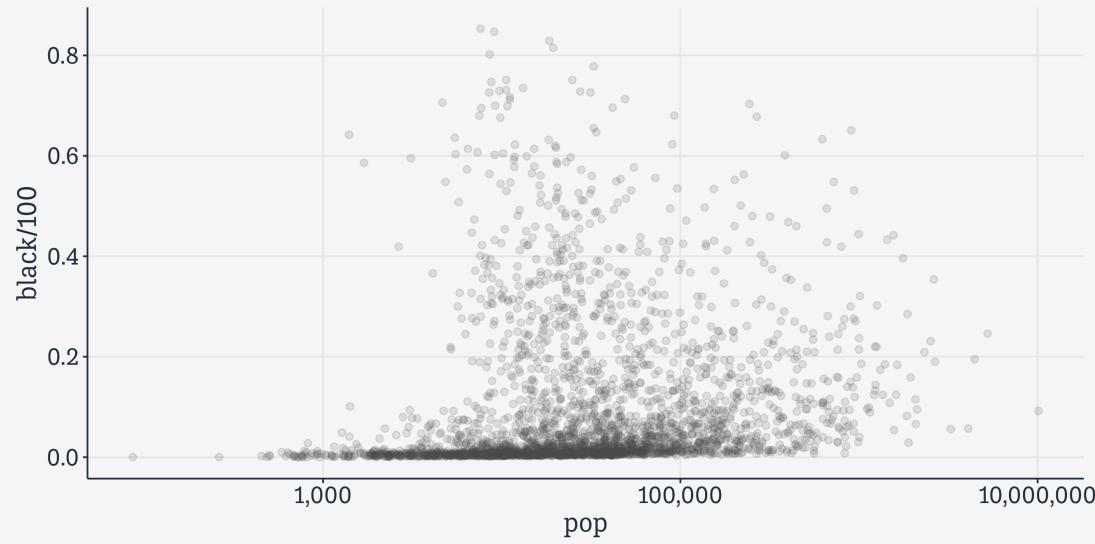


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df_noflip <- county_data %>% filter(flipped = "No")
df_flip <- county_data %>% filter(flipped = "Yes")
df_text <- county_data %>% filter(flipped = "Yes" & black > 0)

ggplot(data = df_noflip,
       mapping = aes(x = pop,
                     y = black/100)) +
  geom_point(alpha = 0.15,
             color = "gray30", size = rel(2)) +
  scale_x_log10(labels = label_comma())
```

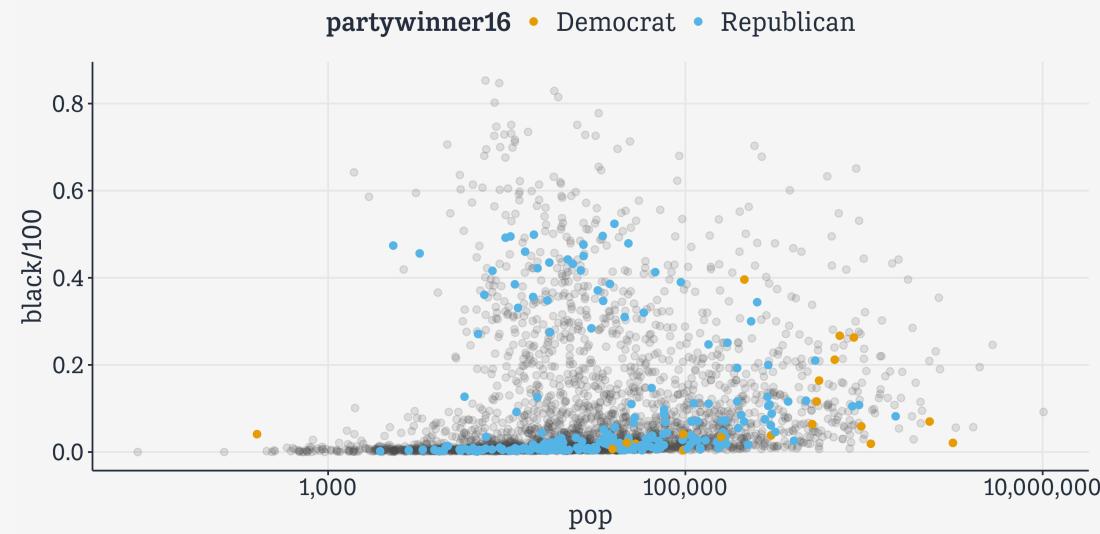


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df_text <- county_data %>% filter(flipped = "Yes" & black > 0)

ggplot(data = df_noflip,
       mapping = aes(x = pop,
                      y = black/100)) +
  geom_point(alpha = 0.15,
             color = "gray30", size = rel(2)) +
  scale_x_log10(labels = label_comma()) +
  geom_point(data = df_flip,
             mapping = aes(x = pop, y = black/100,
                           color = partywinner16),
             size = rel(2))
```

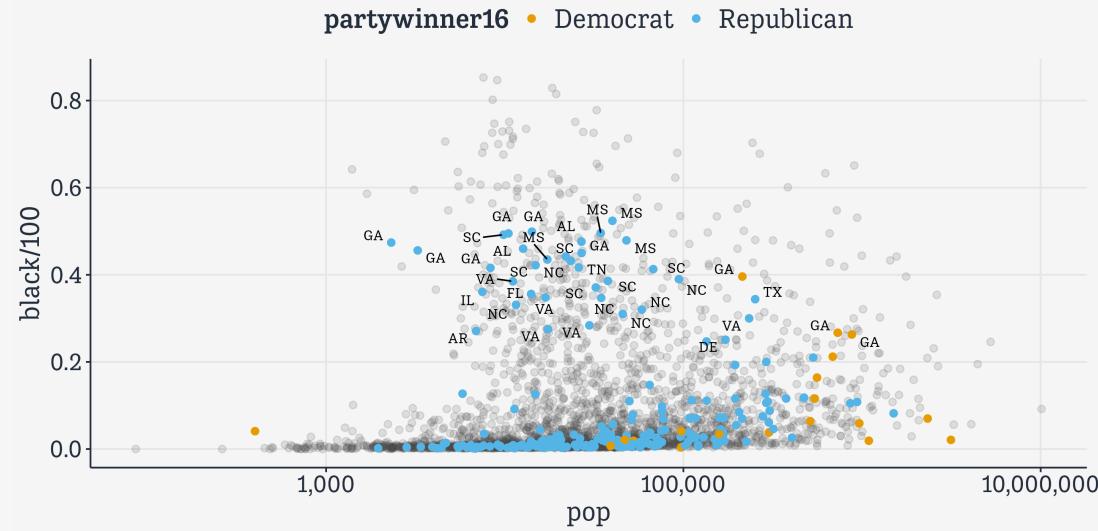


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ggplot(data = df_noflip,
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  geom_point(alpha = 0.15,
             color = "gray30", size = rel(2)) +
  scale_x_log10(labels = label_comma()) +
  geom_point(data = df_flip,
             mapping = aes(x = pop, y = black/100,
                           color = partywinner16),
             size = rel(2)) +
  geom_text_repel(data = df_text,
                  mapping = aes(label = state,
                                family = "Tenso Slab"),
                  size = rel(3.5))
```

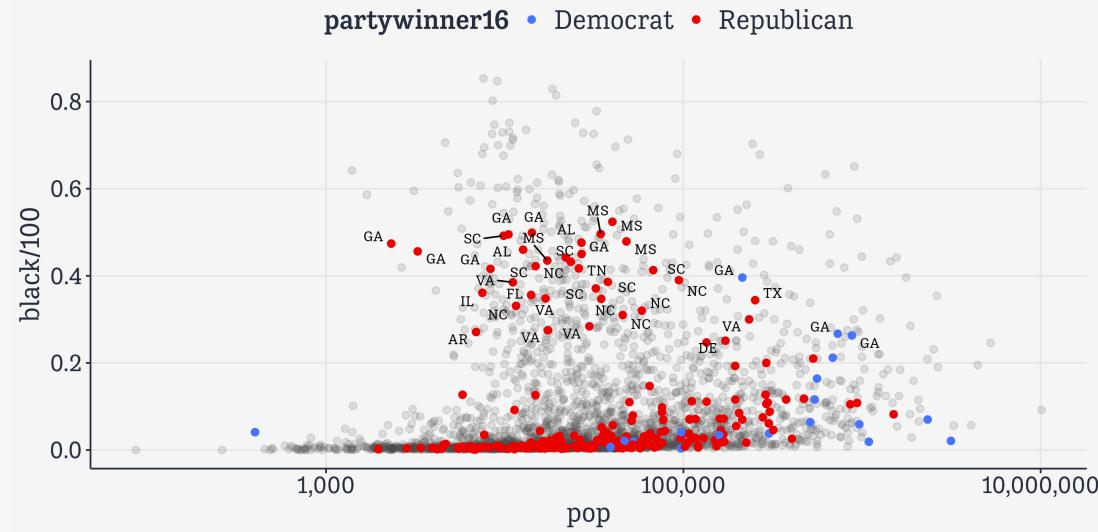


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ggplot(data = df_noflip,
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  scale_x_log10(labels = label_comma()) +
  geom_point(data = df_flip,
             mapping = aes(x = pop, y = black/100,
                           color = partywinner16),
             size = rel(2)) +
  geom_text_repel(data = df_text,
                  mapping = aes(label = state,
                                family = "Tenso Slab"),
                  size = rel(3.5)) +
  scale_color_manual(values = party_colors)
```

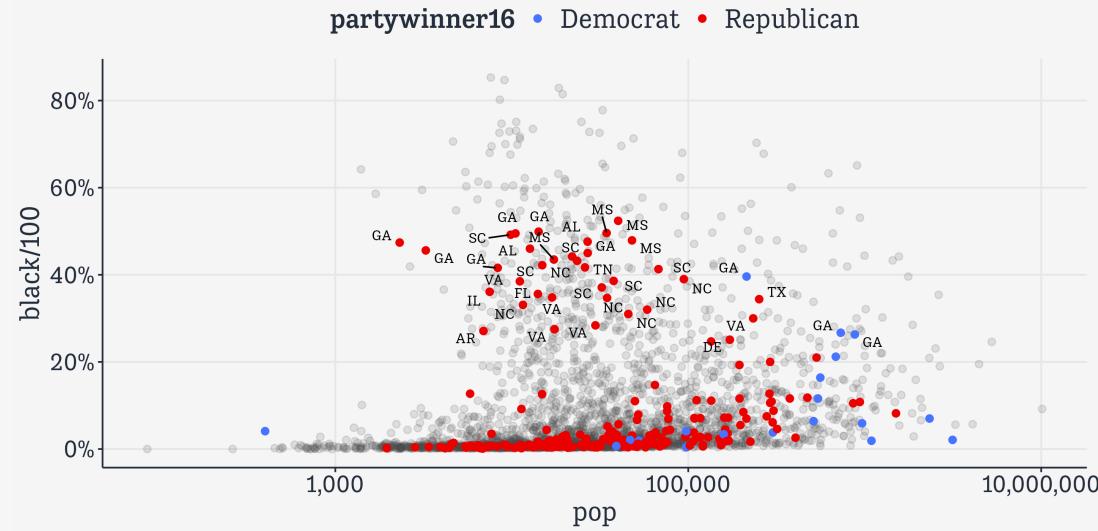


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party_colors ← c("royalblue1", "red2")

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ggplot(data = df_noflip,
        mapping = aes(x = pop,
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  geom_point(alpha = 0.15,
             color = "gray30", size = rel(2)) +
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  geom_point(data = df_flip,
             mapping = aes(x = pop, y = black/100,
                           color = partywinner16),
             size = rel(2)) +
  geom_text_repel(data = df_text,
                  mapping = aes(label = state,
                                family = "Tenso Slide"),
                  size = rel(3.5)) +
  scale_color_manual(values = party_colors) +
  scale_y_continuous(labels = label_percent())
```

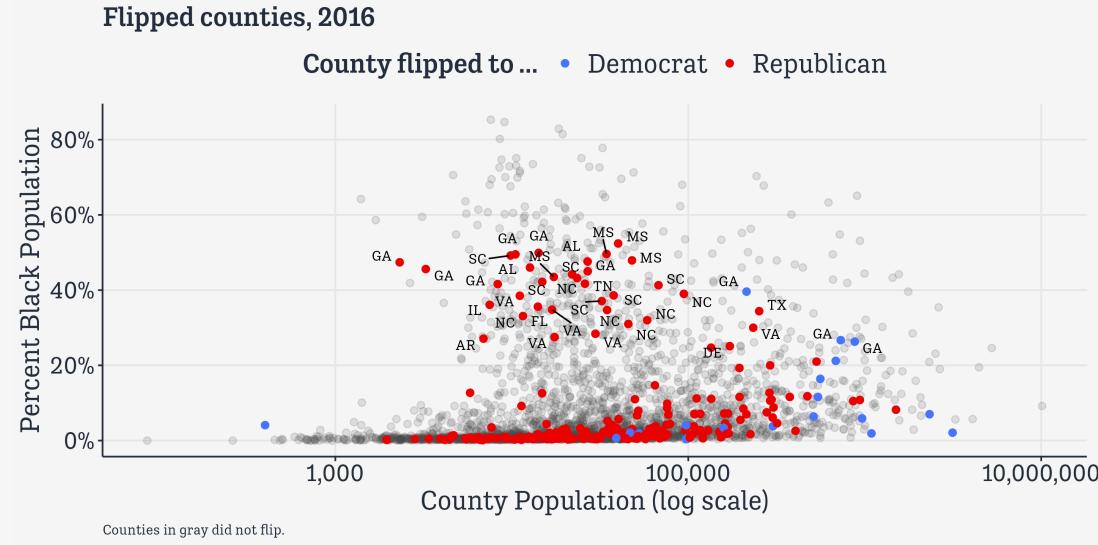


We build it a piece at a time

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ggplot(data = df_noflip,
       mapping = aes(x = pop,
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  geom_point(alpha = 0.15,
             color = "gray30", size = rel(2)) +
  scale_x_log10(labels = label_comma()) +
  geom_point(data = df_flip,
             mapping = aes(x = pop, y = black/100,
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             size = rel(2)) +
  geom_text_repel(data = df_text,
                  mapping = aes(label = state,
                                family = "Tenso Slide"),
                  size = rel(3.5)) +
  scale_color_manual(values = party_colors) +
  scale_y_continuous(labels = label_percent()) +
  labs(color = "County flipped to ...",
       x = "County Population (log scale)",
       y = "Percent Black Population",
       title = "Flipped counties, 2016",
       caption = "Counties in gray did not flip.")
```

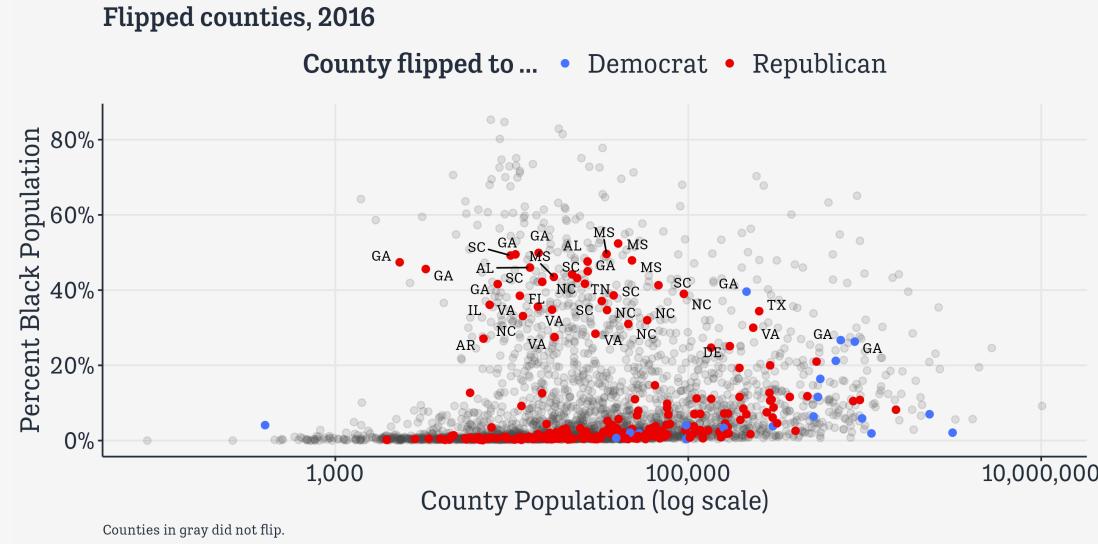


We build it a piece at a time

```
## Brighter Blue and Red
party_colors <- c("royalblue1", "red2")

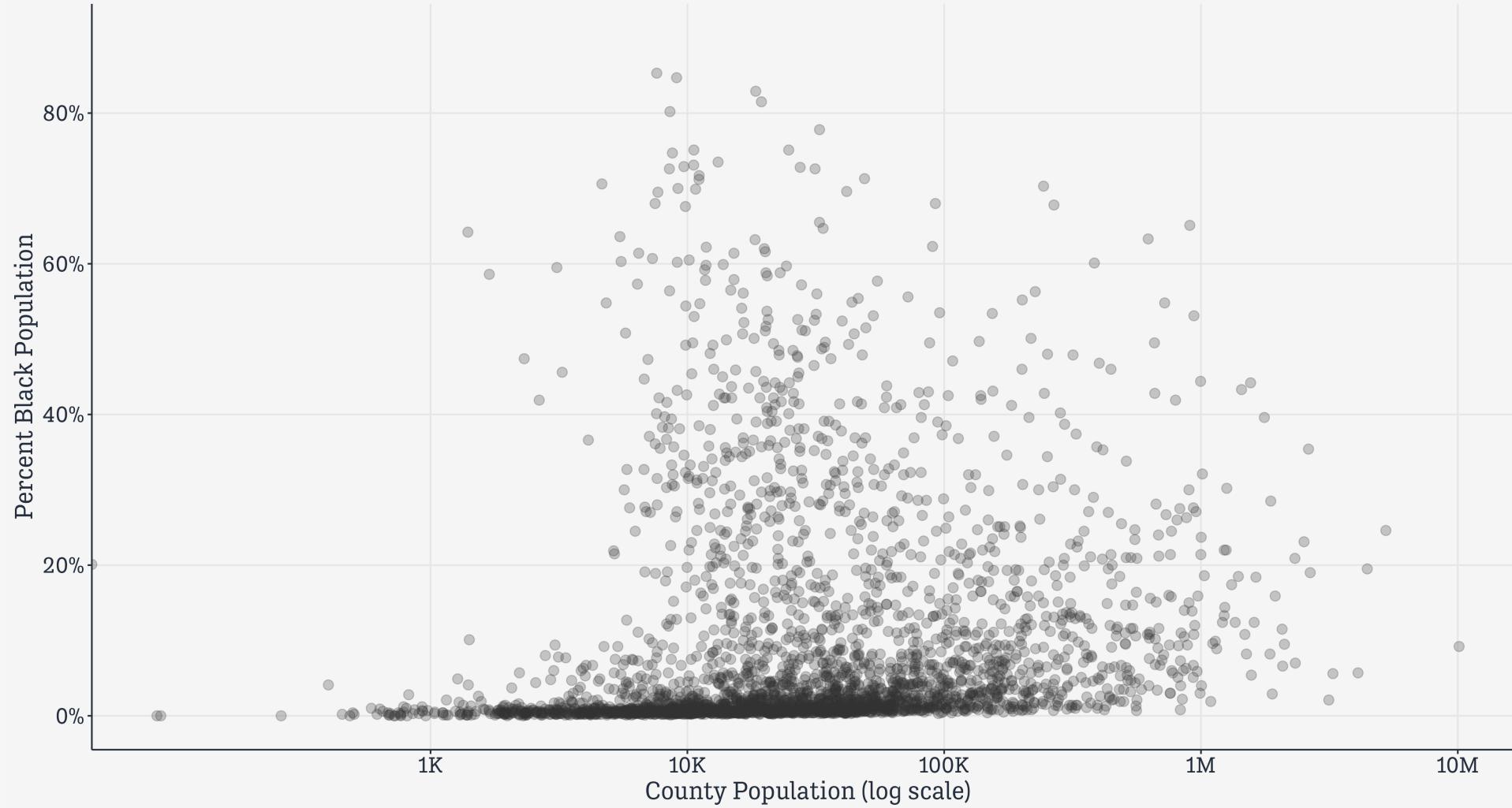
## Data subsets for our geoms
df_noflip <- county_data %>% filter(flipped == "No")
df_flip <- county_data %>% filter(flipped == "Yes")
df_text <- county_data %>% filter(flipped == "Yes" & black

ggplot(data = df_noflip,
       mapping = aes(x = pop,
                      y = black/100)) +
  geom_point(alpha = 0.15,
             color = "gray30", size = rel(2)) +
  scale_x_log10(labels = label_comma()) +
  geom_point(data = df_flip,
             mapping = aes(x = pop, y = black/100,
                           color = partywinner16),
             size = rel(2)) +
  geom_text_repel(data = df_text,
                  mapping = aes(label = state,
                                family = "Tenso Slide"),
                  size = rel(3.5)) +
  scale_color_manual(values = party_colors) +
  scale_y_continuous(labels = label_percent()) +
  labs(color = "County flipped to ...",
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       caption = "Counties in gray did not flip.")
```

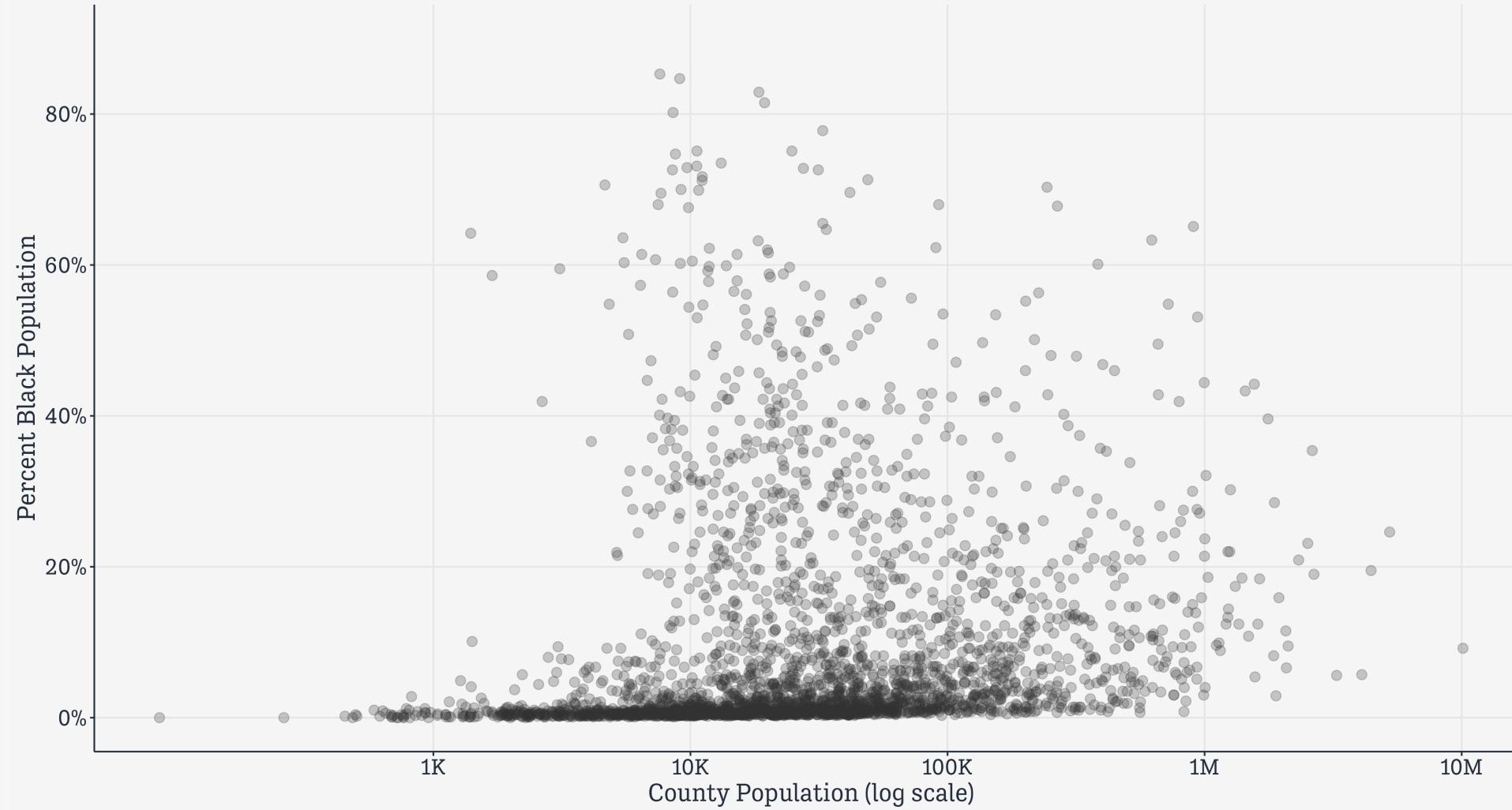


Leverage ggplot's
layered approach

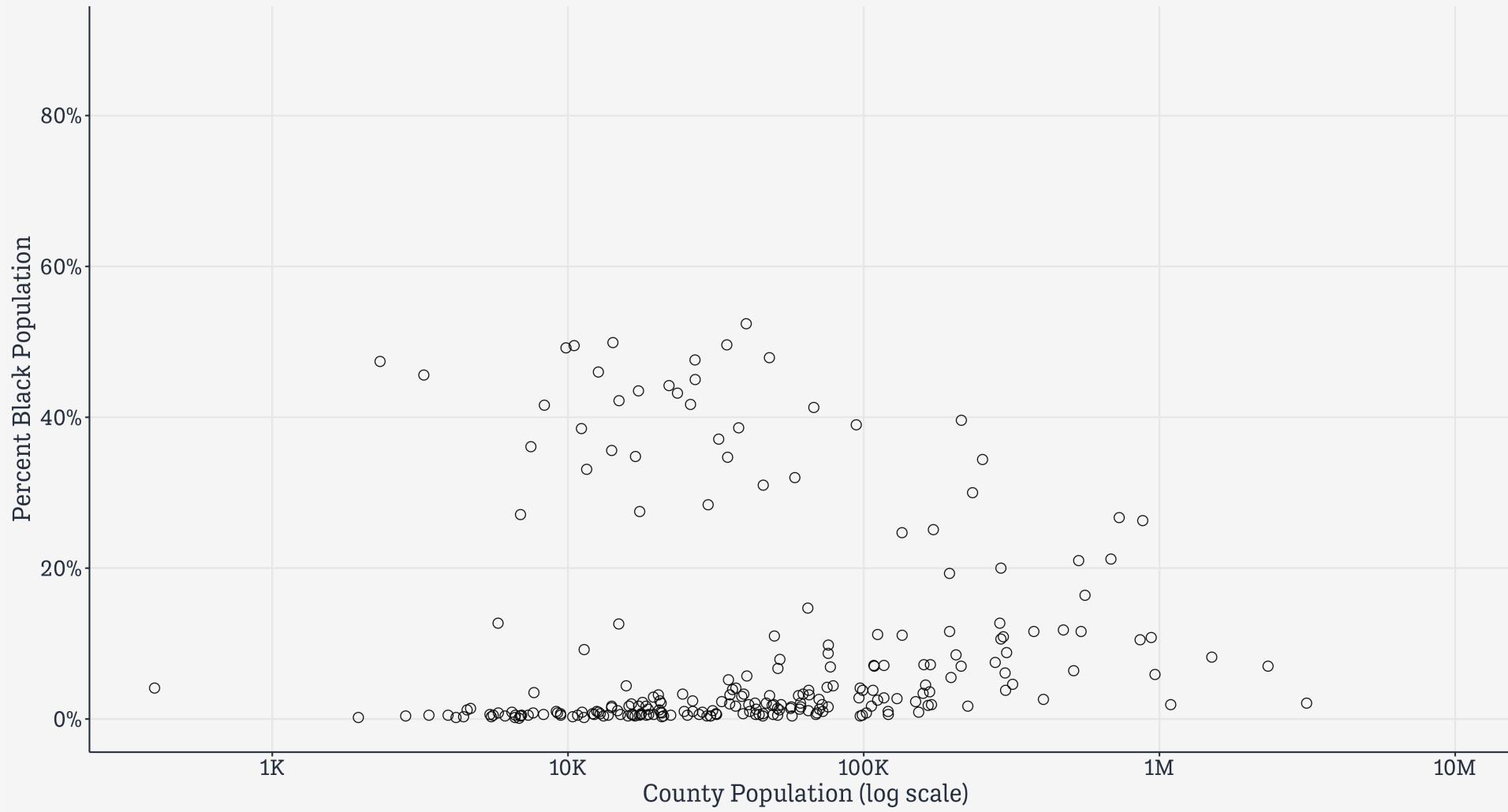
U.S. Counties by Population and Percent Black



These counties did not flip in 2016

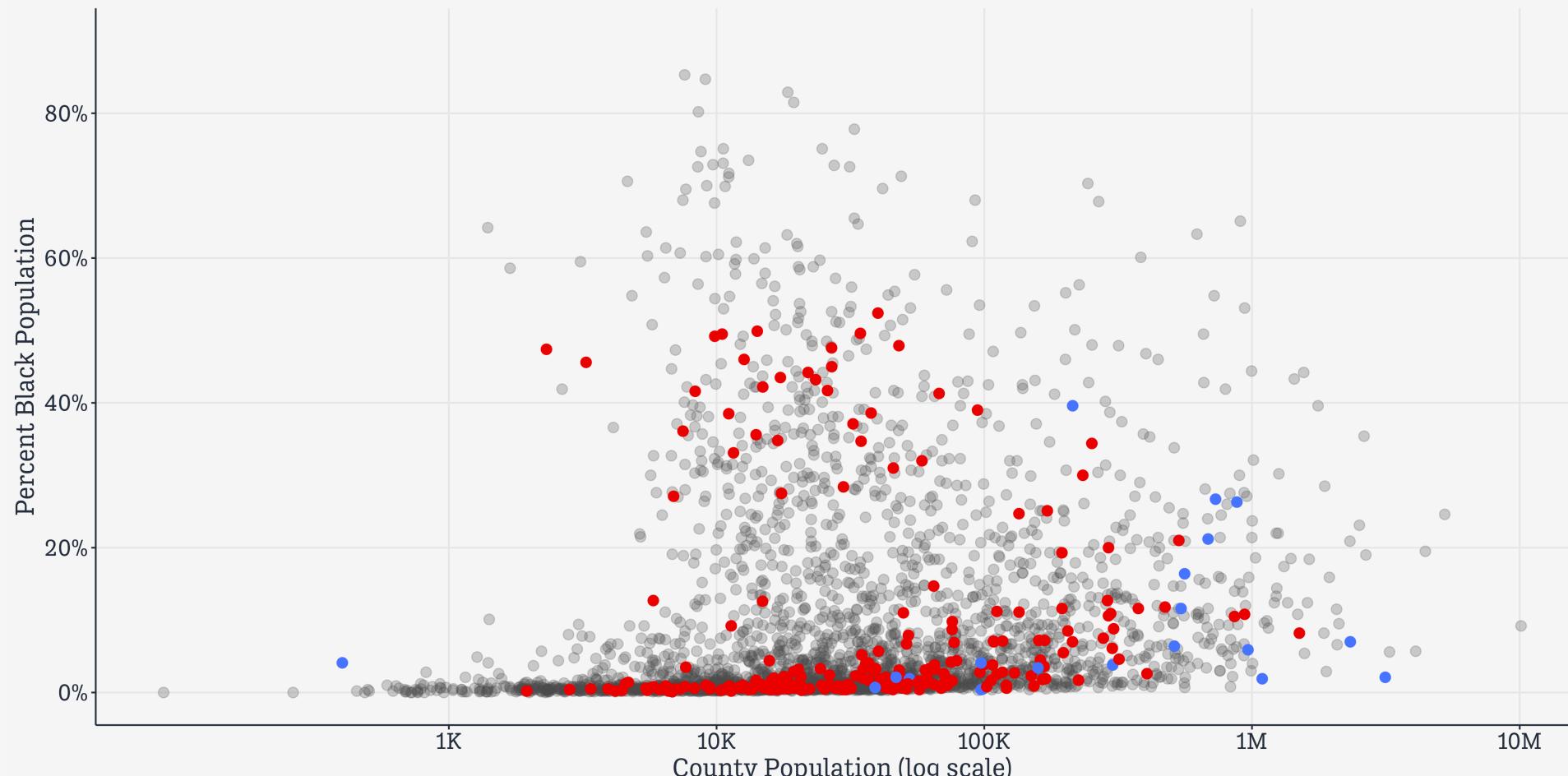


These counties did



Counties that flipped shown by party color

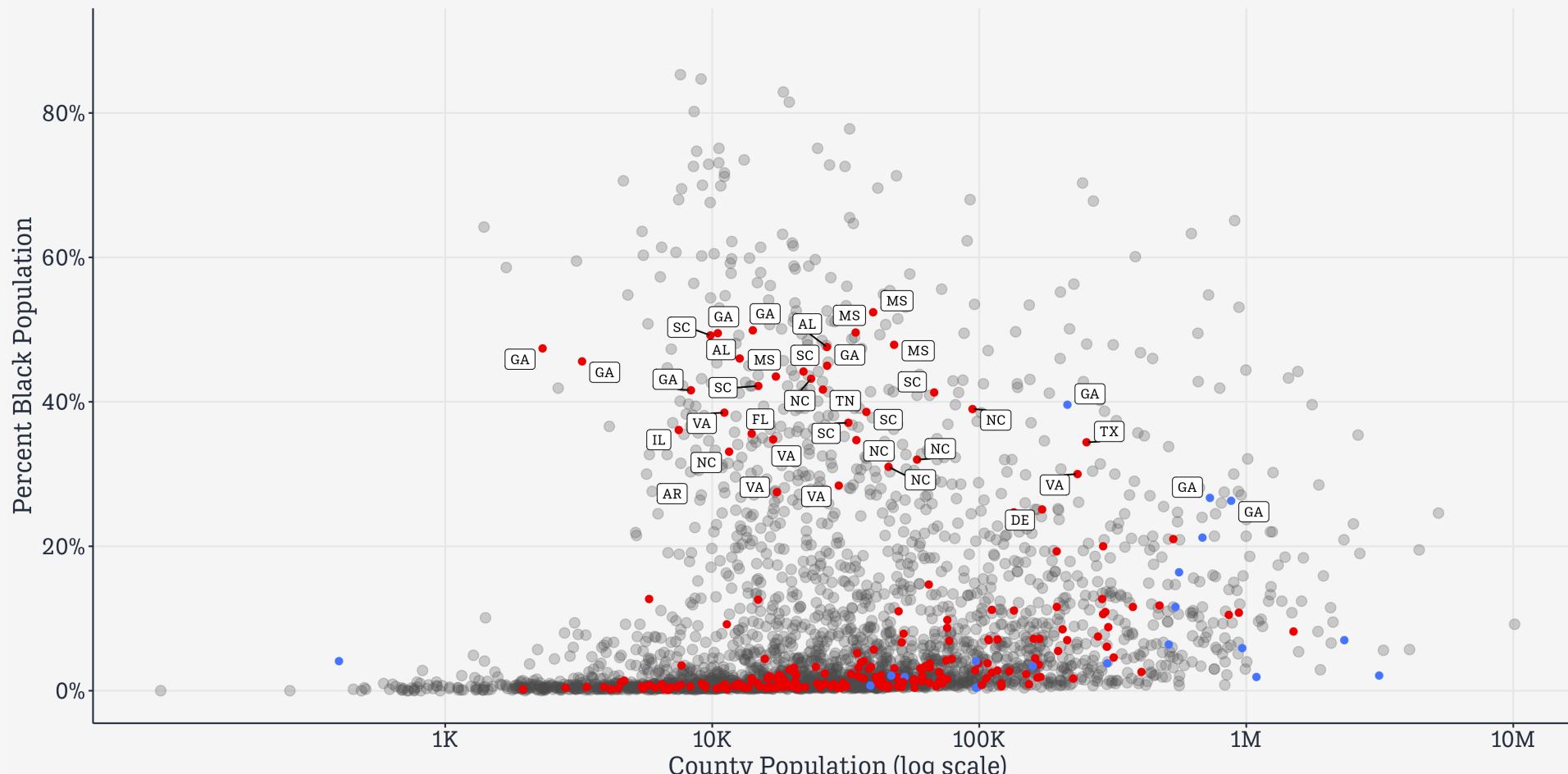
Flipped to • Democrat • Republican



Counties in gray did not flip.

Counties that flipped shown by party color, and labeled by state

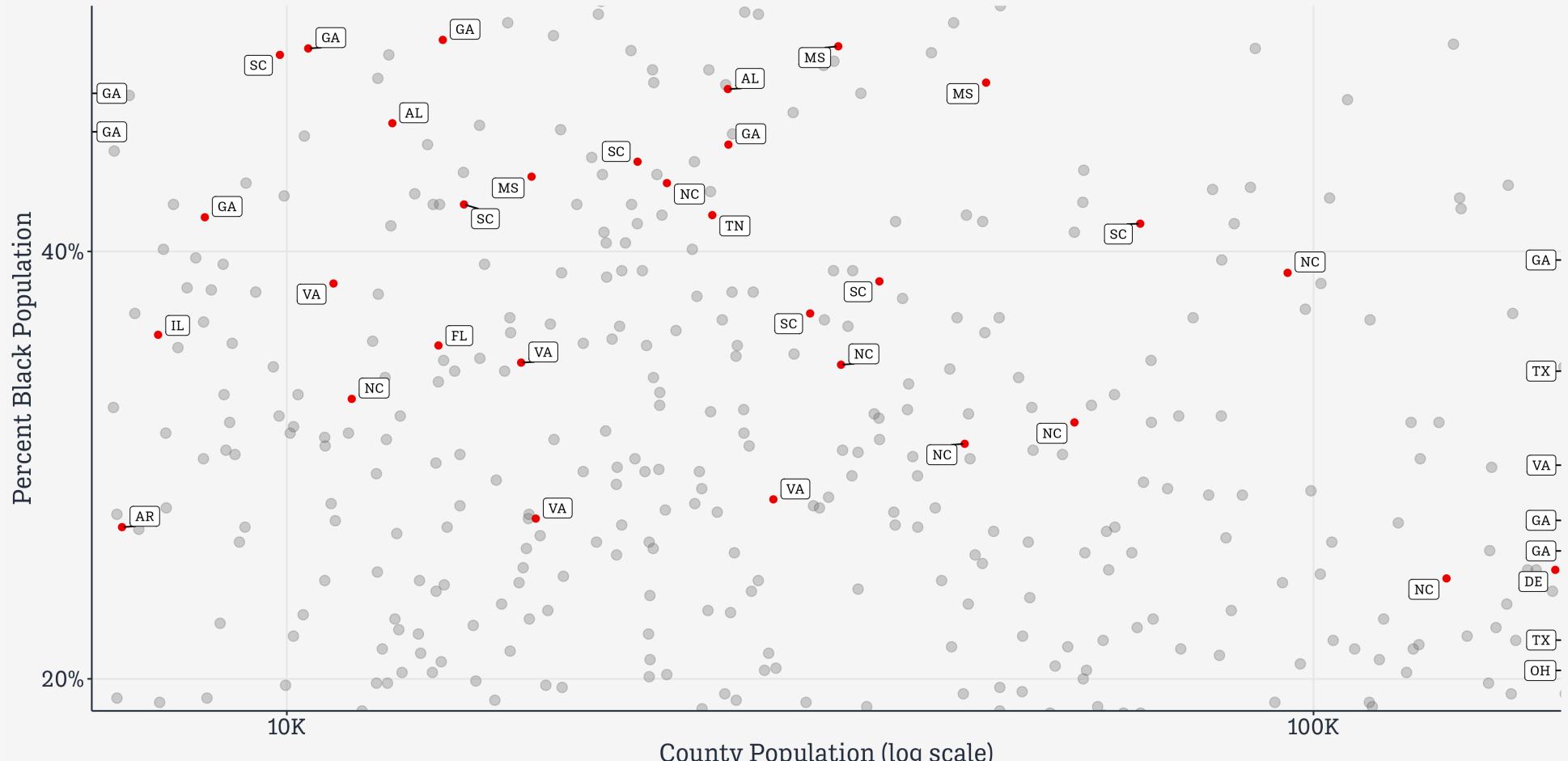
Flipped to • Democrat • Republican



Counties in gray did not flip.

Counties that flipped shown by party color, and labeled by state; zoomed-in

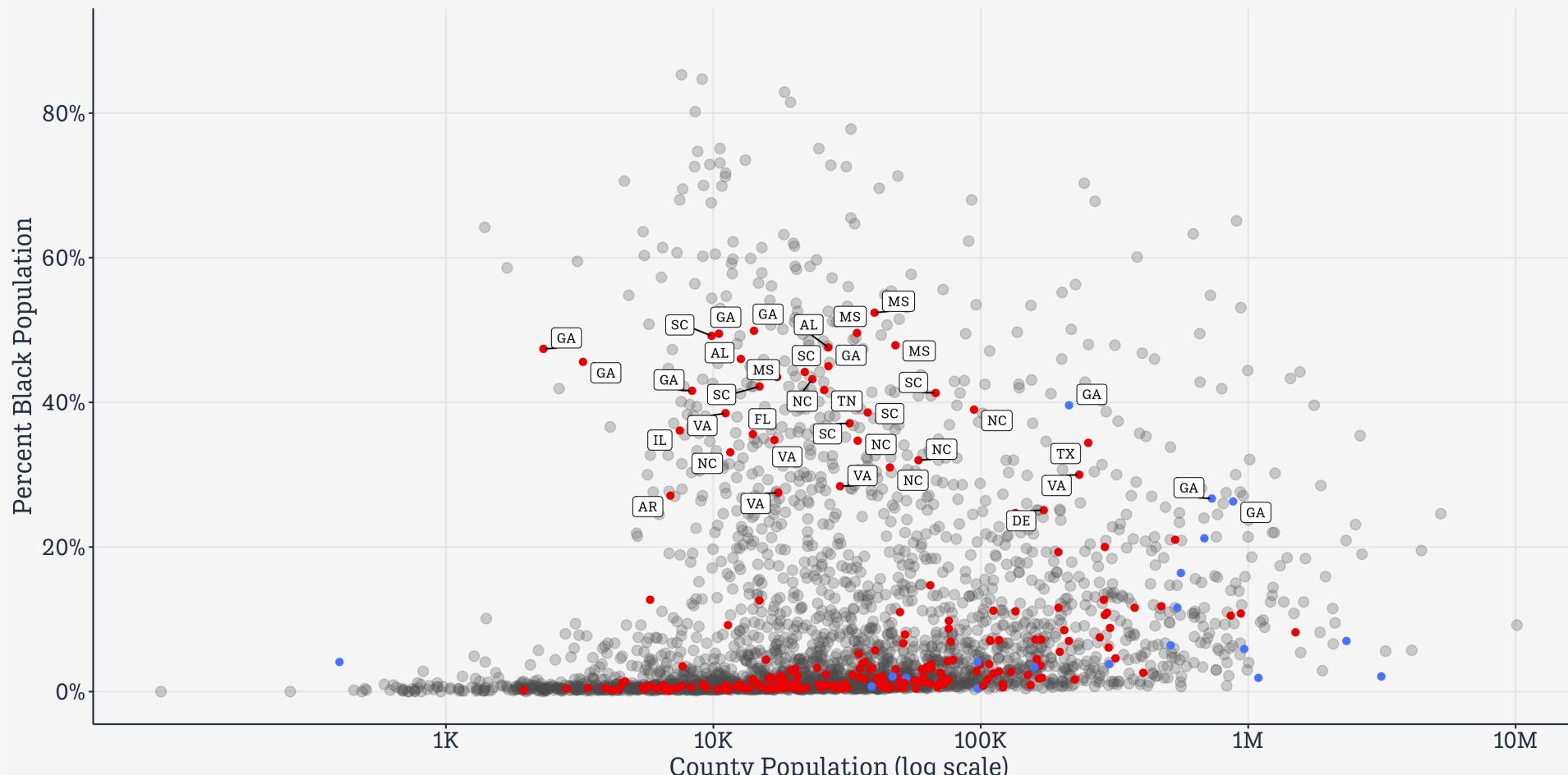
Flipped to • Democrat • Republican



Counties in gray did not flip.

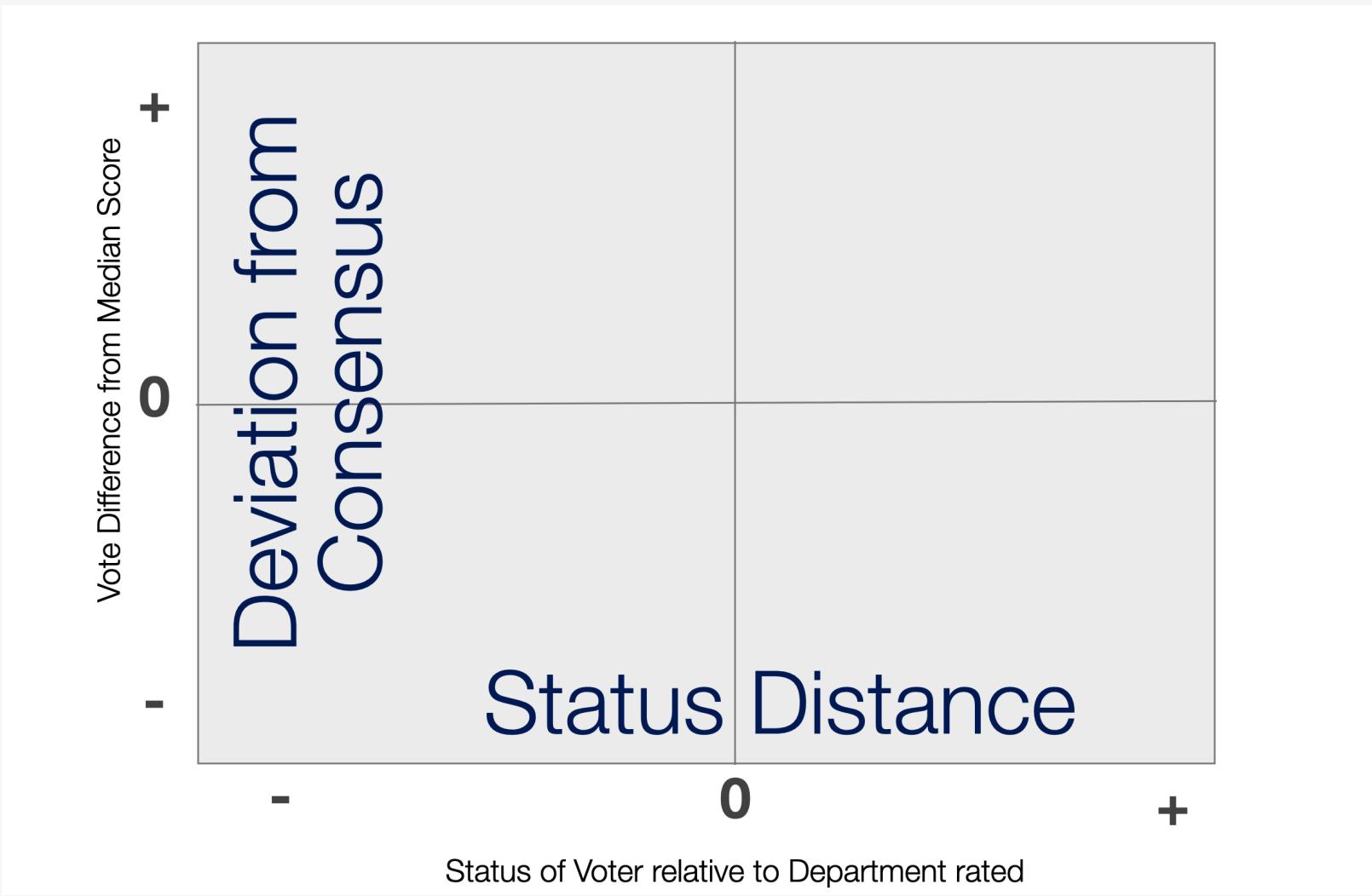
Counties that flipped shown by party color, and labeled by state

Flipped to • Democrat • Republican



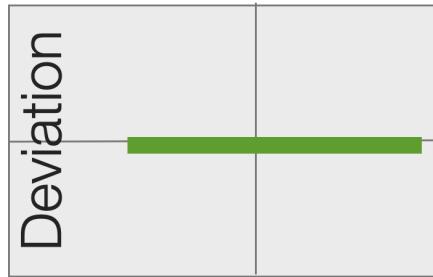
Layer,
Highlight,
Repeat

Build from ideas to data



The relationship of interest

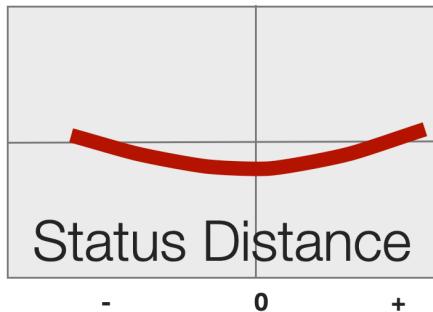
Build from ideas to data



1. Pure Objectivity



2. Distant Envy

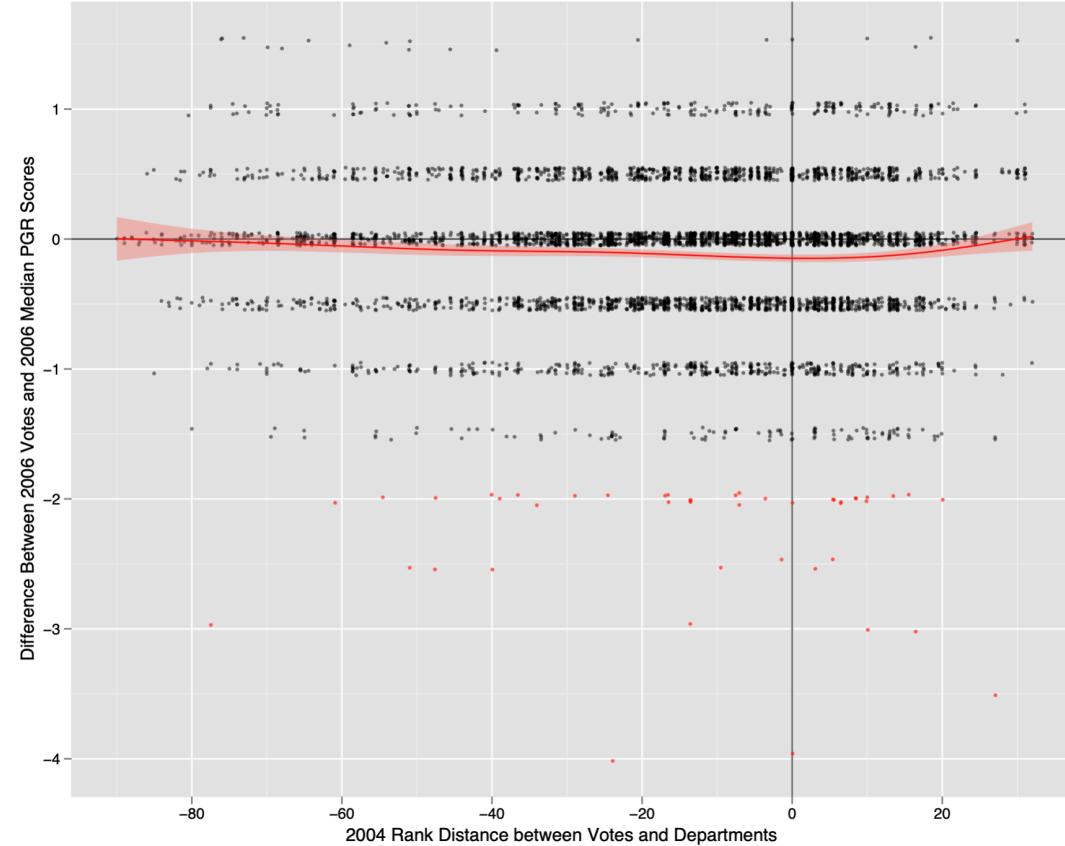


3. Local Competition

Theory says ...

Build from ideas to data

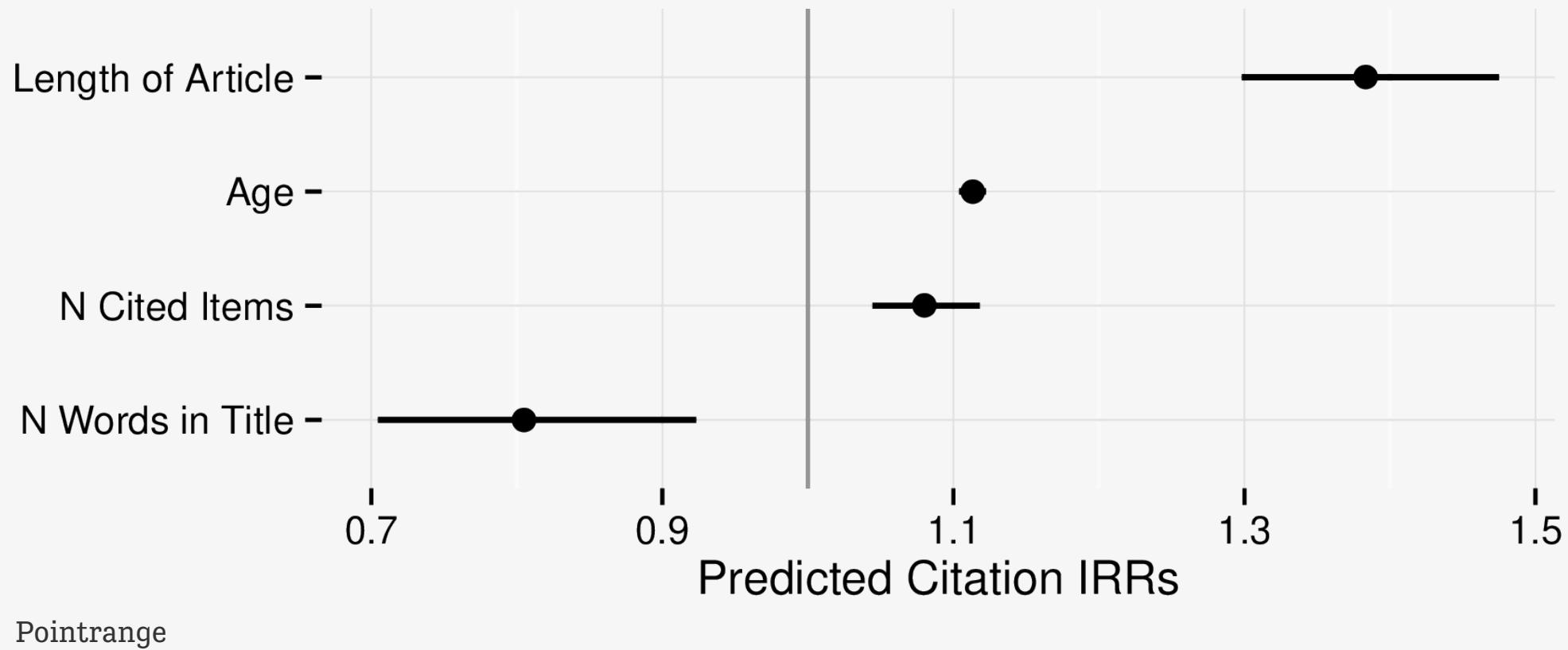
Deviation



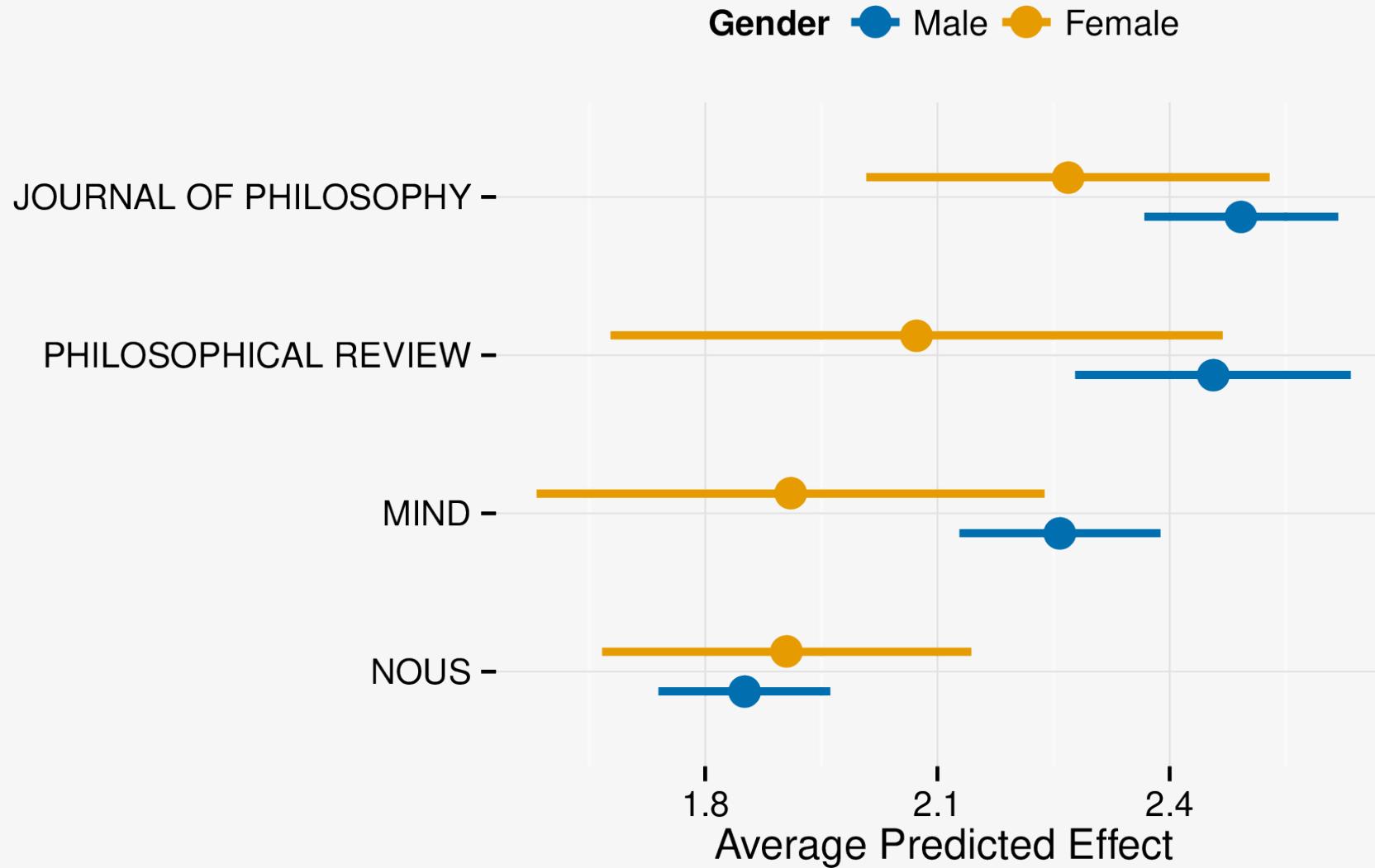
Status Distance

Data suggests ...

Repeat to differentiate



Repeat to differentiate

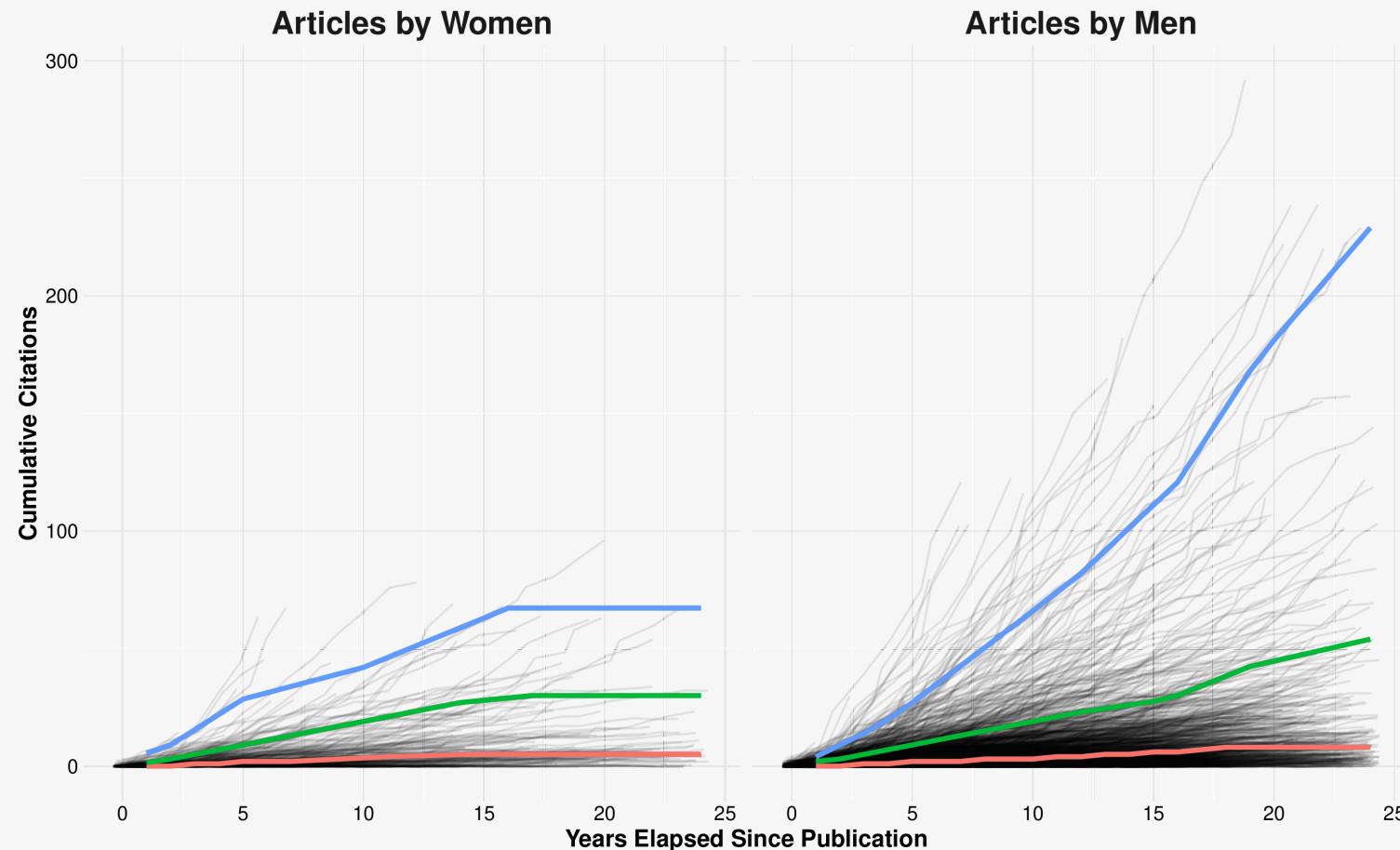


Add a comparison group with color and `position_dodge()`

Layer and repeat with facets

Citation Growth Curves for all Articles

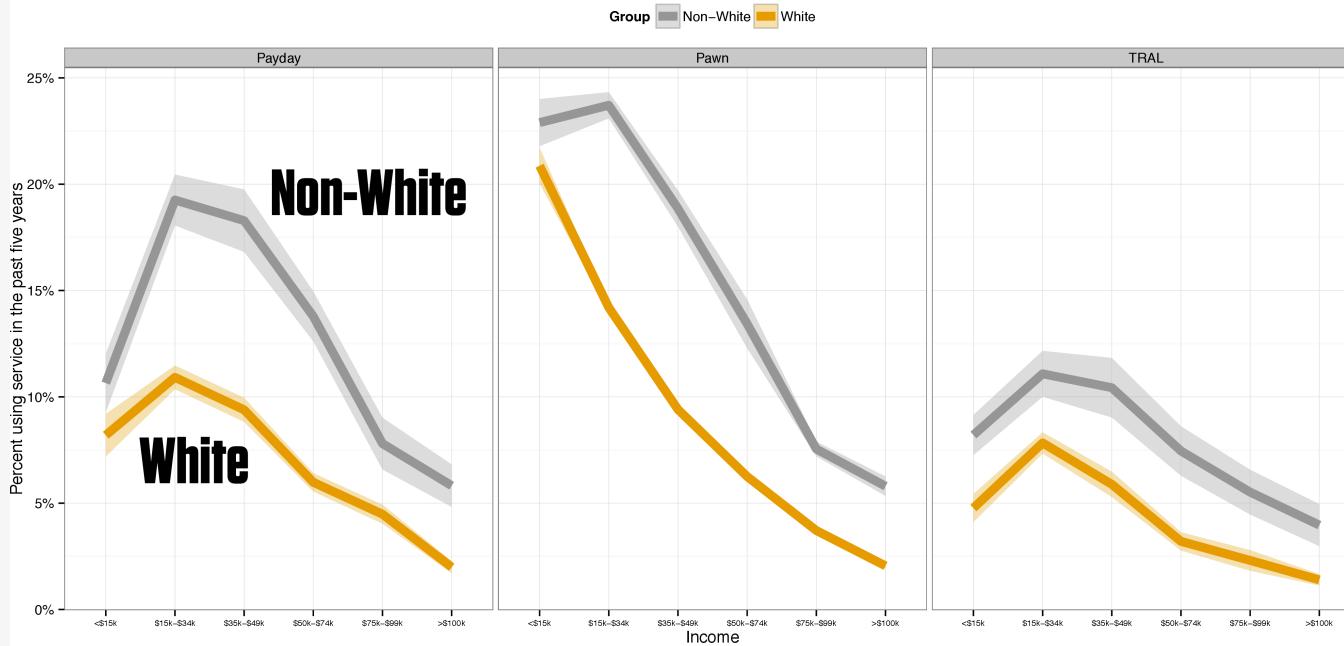
Percentile — at the Median — at the 90th Percentile — in the top One Percent



Compare across facets

Layer and repeat across facets

Categorical Gaps: Alternative Financial Services



Payday Loan

Pawn Shop

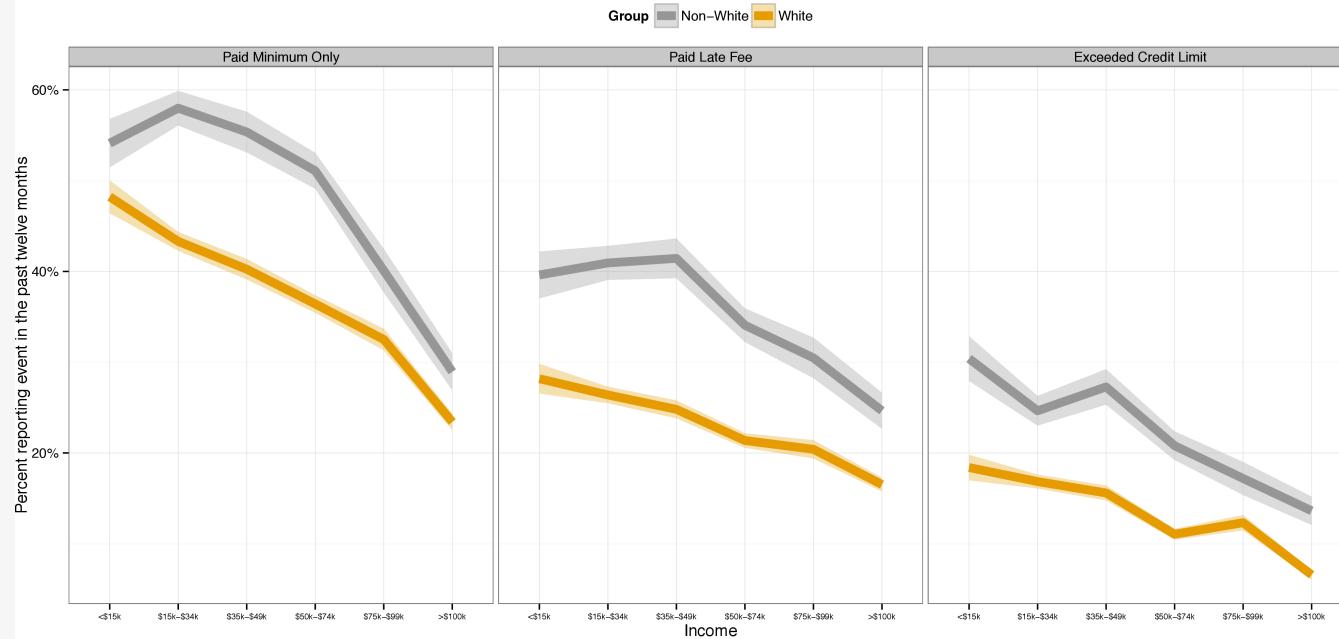
TRAL

8

Layer to compare

Layer and repeat across facets

Categorical Gaps: Adverse Credit Events



Minimum Only

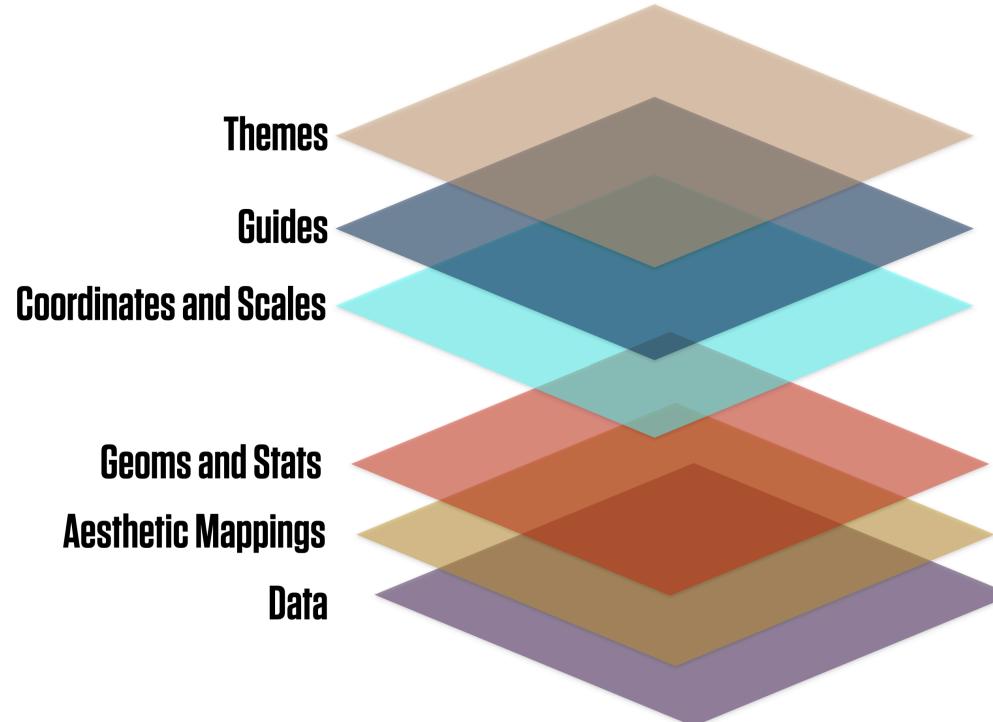
Late Fee

Over Limit

9

Layer to compare

X-Ray Vision



Seeing through it

Themes

Themes ...

are controlled by the `theme()` function

can be bundled into functions of their own, like `theme_bw()` or
`theme_minimal()`

can be set for the duration of a file or project with `theme_set()`

make changes that are applied *additively*

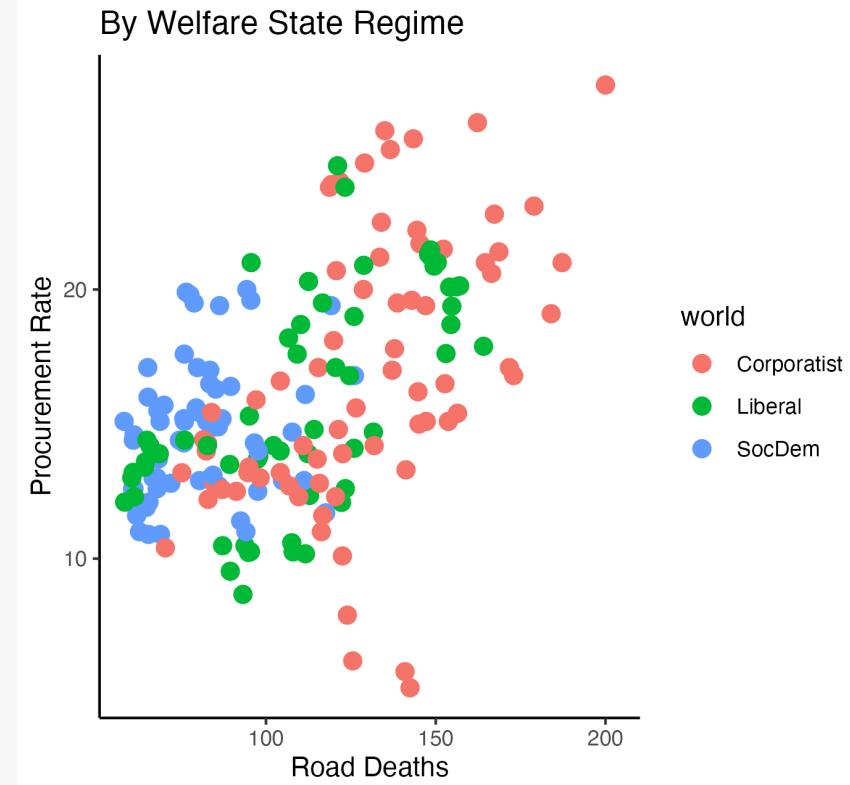
and most importantly ...

**Thematic elements
do not represent
data directly**

Make a plot

```
p <- organdata %>
  drop_na(world) %>
  ggplot(mapping = aes(x = roads, y = donors,
                        color = world)) +
  geom_point(size = 3) +
  labs(x = "Road Deaths",
       y = "Procurement Rate",
       title = "By Welfare State Regime")
```

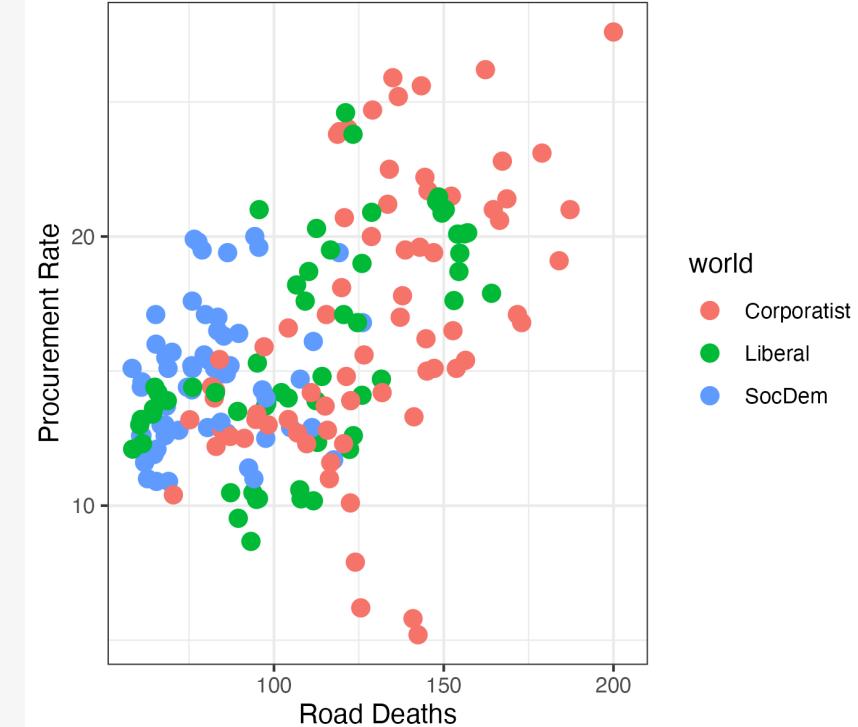
```
p
```



Add a theme ... `theme_bw()`

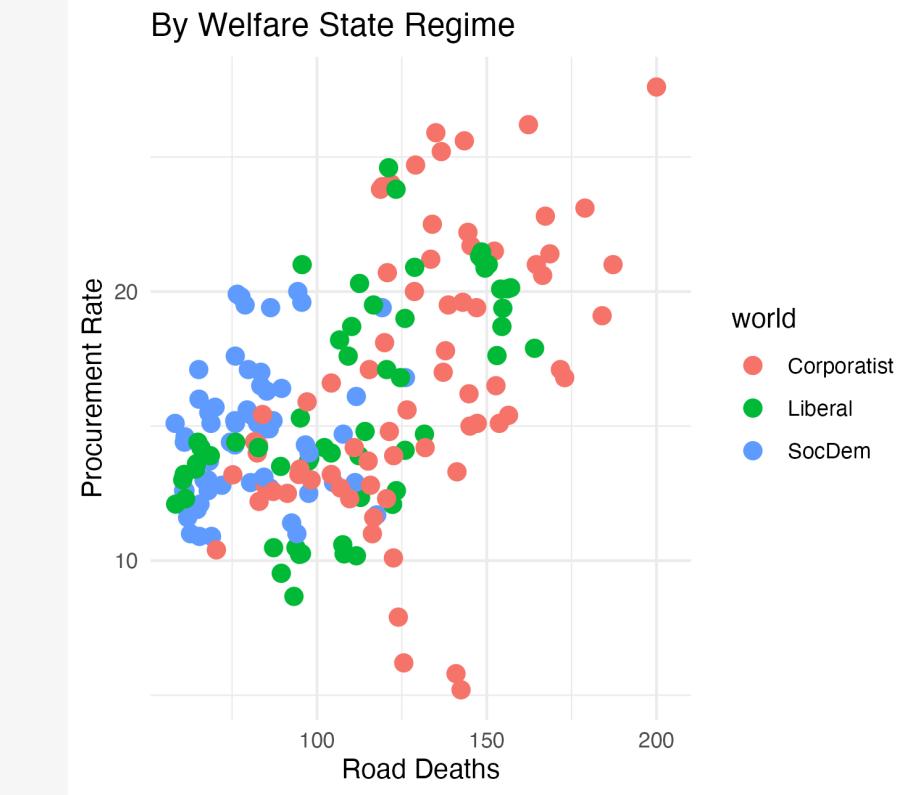
```
p + theme_bw()
```

By Welfare State Regime



Add a theme ... `theme_minimal()`

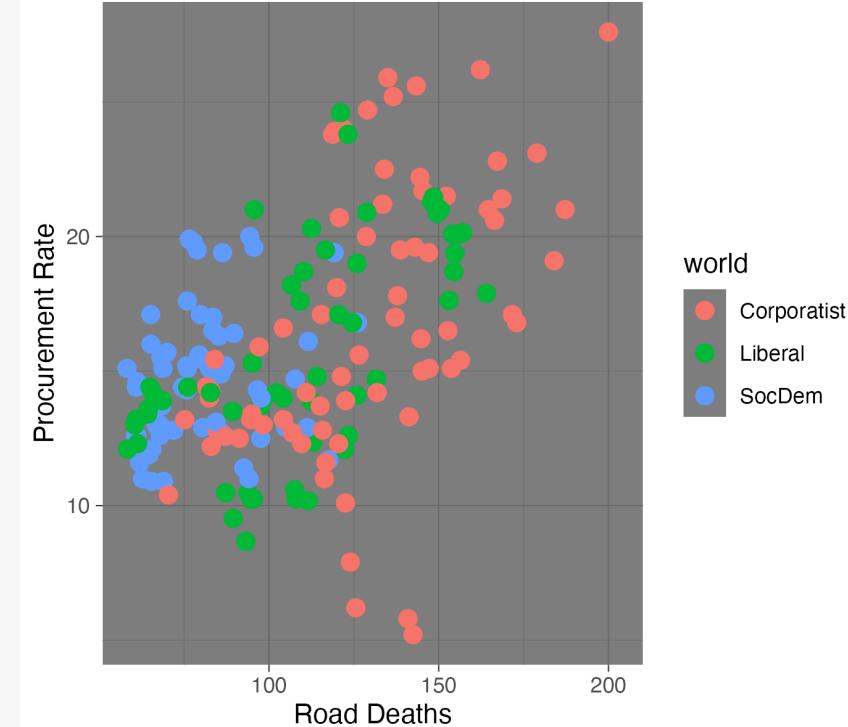
```
p + theme_minimal()
```



Add a theme ... `theme_dark()`

```
p + theme_dark()
```

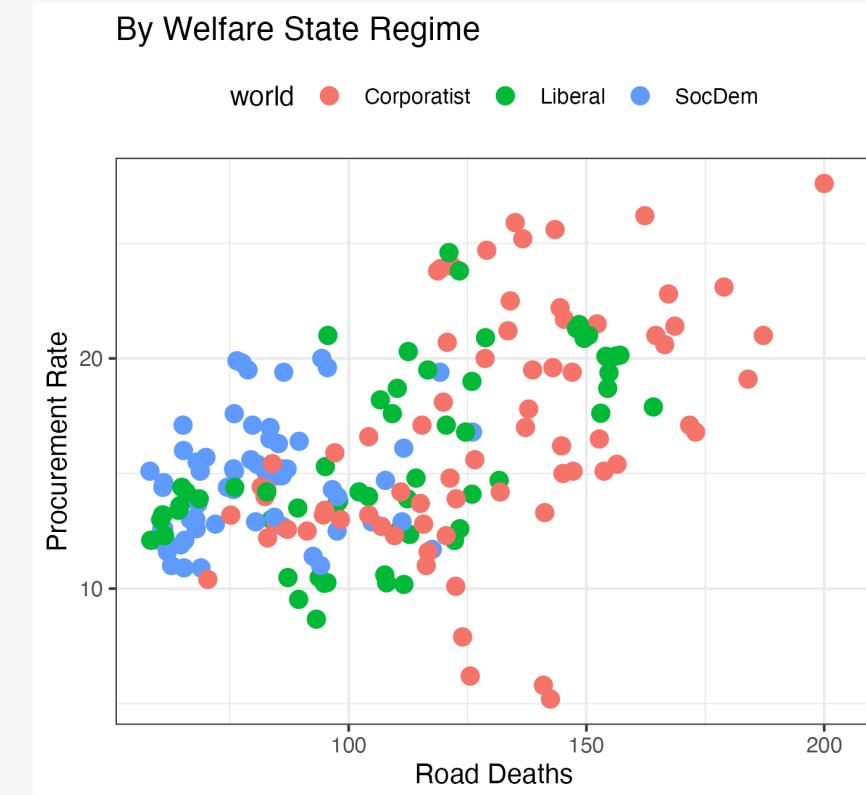
By Welfare State Regime



Adjust with the `theme()` function

None of this directly touches the parts of the plot that are representing your data—i.e. the visual parts that are mapped to a variable, and thus have a scale. Adjusting those is the job of the `scale_` and `guide()` functions.

```
p + theme_bw() +  
  theme(legend.position = "top")
```



There are *many* theme elements

line rect text title aspect.ratio
axis.title axis.title.x axis.title.x.top axis.title.x.bottom
axis.title.y axis.title.y.left axis.title.y.right axis.text
axis.text.x axis.text.x.top axis.text.x.bottom axis.text.y
axis.text.y.left axis.text.y.right axis.ticks axis.ticks.x
axis.ticks.x.top axis.ticks.x.bottom axis.ticks.y
axis.ticks.y.left axis.ticks.y.right axis.ticks.length
axis.ticks.length.x axis.ticks.length.x.top
axis.ticks.length.x.bottom axis.ticks.length.y
axis.ticks.length.y.left axis.ticks.length.y.right axis.line
axis.line.x axis.line.x.top axis.line.x.bottom axis.line.y
axis.line.y.left axis.line.y.right

legend.background legend.margin legend.spacing
legend.spacing.x legend.spacing.y legend.key
legend.key.size legend.key.height legend.key.width
legend.text legend.text.align legend.title
legend.title.align legend.position legend.direction
legend.justification legend.box legend.box.just
legend.box.margin legend.box.background
legend.box.spacing

panel.background panel.border panel.spacing
panel.spacing.x panel.spacing.y panel.grid
panel.grid.major panel.grid.minor panel.grid.major.x
panel.grid.major.y panel.grid.minor.x panel.grid.minor.y
panel.on top plot.background

plot.title plot.title.position plot.subtitle plot.caption
plot.caption.position plot.tag plot.tag.position
plot.margin

strip.background strip.background.x strip.background.y
strip.placement strip.text strip.text.x strip.text.y
strip.switch.pad.grid strip.switch.pad.wrap

But they are structured

line rect text title aspect.ratio

axis.title axis.title.x axis.title.x.top axis.title.x.bottom
axis.title.y axis.title.y.left axis.title.y.right axis.text
axis.text.x axis.text.x.top axis.text.x.bottom axis.text.y
axis.text.y.left axis.text.y.right axis.ticks axis.ticks.x
axis.ticks.x.top axis.ticks.x.bottom axis.ticks.y
axis.ticks.y.left axis.ticks.y.right axis.ticks.length
axis.ticks.length.x axis.ticks.length.x.top
axis.ticks.length.x.bottom axis.ticks.length.y
axis.ticks.length.y.left axis.ticks.length.y.right axis.line
axis.line.x axis.line.x.top axis.line.x.bottom axis.line.y
axis.line.y.left axis.line.y.right

legend.background legend.margin legend.spacing
legend.spacing.x legend.spacing.y legend.key
legend.key.size legend.key.height legend.key.width
legend.text legend.text.align legend.title
legend.title.align legend.position legend.direction
legend.justification legend.box legend.box.just
legend.box.margin legend.box.background
legend.box.spacing

panel.background panel.border panel.spacing
panel.spacing.x panel.spacing.y panel.grid
panel.grid.major panel.grid.minor panel.grid.major.x
panel.grid.major.y panel.grid.minor.x panel.grid.minor.y
panel.on top

plot.background plot.title plot.title.position plot.subtitle
plot.caption plot.caption.position plot.tag
plot.tag.position plot.margin

strip.background strip.background.x strip.background.y
strip.placement strip.text strip.text.x strip.text.y
strip.switch.pad.grid strip.switch.pad.wrap

And *inherit*

line rect text title aspect.ratio

axis.title axis.title.x axis.title.x.top axis.title.x.bottom
axis.title.y axis.title.y.left axis.title.y.right axis.text
axis.text.x axis.text.x.top axis.text.x.bottom axis.text.y
axis.text.y.left axis.text.y.right axis.ticks axis.ticks.x
axis.ticks.x.top axis.ticks.x.bottom axis.ticks.y
axis.ticks.y.left axis.ticks.y.right axis.ticks.length
axis.ticks.length.x axis.ticks.length.x.top
axis.ticks.length.x.bottom axis.ticks.length.y
axis.ticks.length.y.left axis.ticks.length.y.right axis.line
axis.line.x axis.line.x.top axis.line.x.bottom axis.line.y
axis.line.y.left axis.line.y.right

legend.background legend.margin legend.spacing
legend.spacing.x legend.spacing.y legend.key
legend.key.size legend.key.height legend.key.width
legend.text legend.text.align legend.title
legend.title.align legend.position legend.direction
legend.justification legend.box legend.box.just
legend.box.margin legend.box.background
legend.box.spacing

panel.background panel.border panel.spacing
panel.spacing.x panel.spacing.y panel.grid
panel.grid.major panel.grid.minor panel.grid.major.x
panel.grid.major.y panel.grid.minor.x panel.grid.minor.y
panel.on top

plot.background plot.title plot.title.position
plot.subtitle plot.caption plot.caption.position plot.tag
plot.tag.position plot.margin

strip.background strip.background.x
strip.background.y strip.placement strip.text
strip.text.x strip.text.y strip.switch.pad.grid
strip.switch.pad.wrap

Two kinds of adjustment

It's a single setting.

E.g., `legend.position` can be "none", "left", "right", "bottom", or "top"

Hence, e.g., `theme(legend.position = "top")`, which we have seen several times. Similarly for e.g. `legend.direction` (can be "horizontal" or "vertical").

It's a component of the plot that might be styled in several ways. E.g., The text on the axes, or the lines in the plot panel.

If the latter ...

If adjusting a thematic element ask...

Where on the plot is it?

If adjusting a thematic element ask...

Where on the plot is it?

Is it part of an *axis*, part of the *panel*, the *strip* (facet title) box, or the *legend*?
This will help you find the name of the thing you want to adjust.

If adjusting a thematic element ask...

Where on the plot is it?

Is it part of an *axis*, part of the *panel*, the *strip* (facet title) box, or the *legend*?

This will help you find the name of the thing you want to adjust.

E.g. “I want to adjust the text for the markings on the x-axis”

You want `axis.ticks.x`

E.g. “I want to adjust the styling of the main y-axis grid lines inside the plot”

You want `panel.grid.major.y`

If adjusting a thematic element, ask...

What *kind* of element is it?

If adjusting a thematic element, ask...

What *kind* of element is it?

Is it *text*, or a *line*, or a *rectangle*?

This will tell you what function to use to make the adjustment to the named element.

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If it's text, adjust the element with `element_text()`

If adjusting a thematic element, ask...

What *kind* of element is it?

Is it *text*, or a *line*, or a *rectangle*?

This will tell you what function to use to make the adjustment to the named element.

If it's text, adjust the element with `element_text()`

If it's a line, adjust it with `element_line()`

If adjusting a thematic element, ask...

What *kind* of element is it?

Is it *text*, or a *line*, or a *rectangle*?

This will tell you what function to use to make the adjustment to the named element.

If it's text, adjust the element with `element_text()`

If it's a line, adjust it with `element_line()`

If it's a rectangle, with `element_rect()`

If adjusting a thematic element, ask...

What *kind* of element is it?

Is it *text*, or a *line*, or a *rectangle*?

This will tell you what function to use to make the adjustment to the named element.

If it's text, adjust the element with `element_text()`

If it's a line, adjust it with `element_line()`

If it's a rectangle, with `element_rect()`

If you want to *fully turn off* an element, use `element_blank()`

For example ...

“I want to adjust the styling of the plot title”

For example ...

“I want to adjust the styling of the plot title”

The relevant element is `plot.title`.

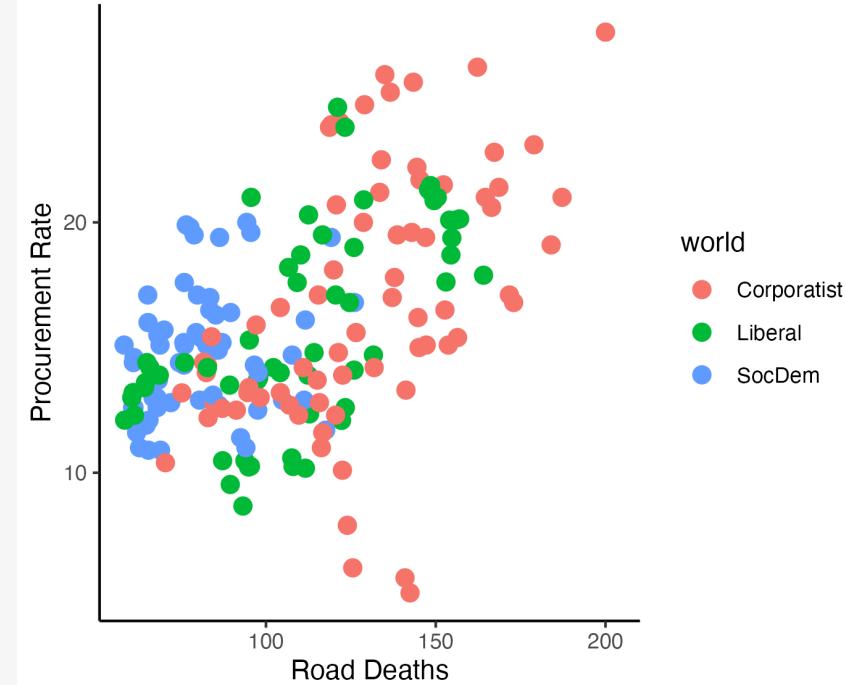
It's *text*.

Inside the theme function, adjust it with `element_text()`.

For example ...

```
p + theme(plot.title =  
          element_text(size = rel(3),  
                      face = "bold",  
                      color = "orange"))
```

By Welfare State Re



For example ...

“I want to adjust y axis grid lines on the plot”

For example ...

“I want to adjust y axis grid lines on the plot”

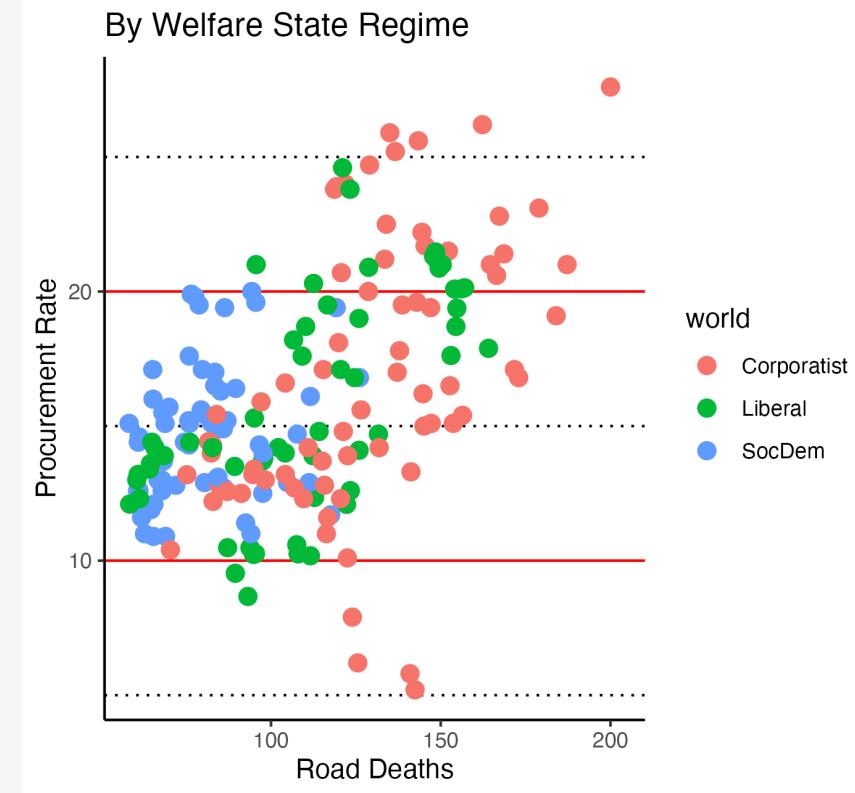
The relevant elements are `panel.grid.major.y` and `panel.grid.minor.y`.

These are *lines*.

Inside the theme function, adjust it with `element_line()`.

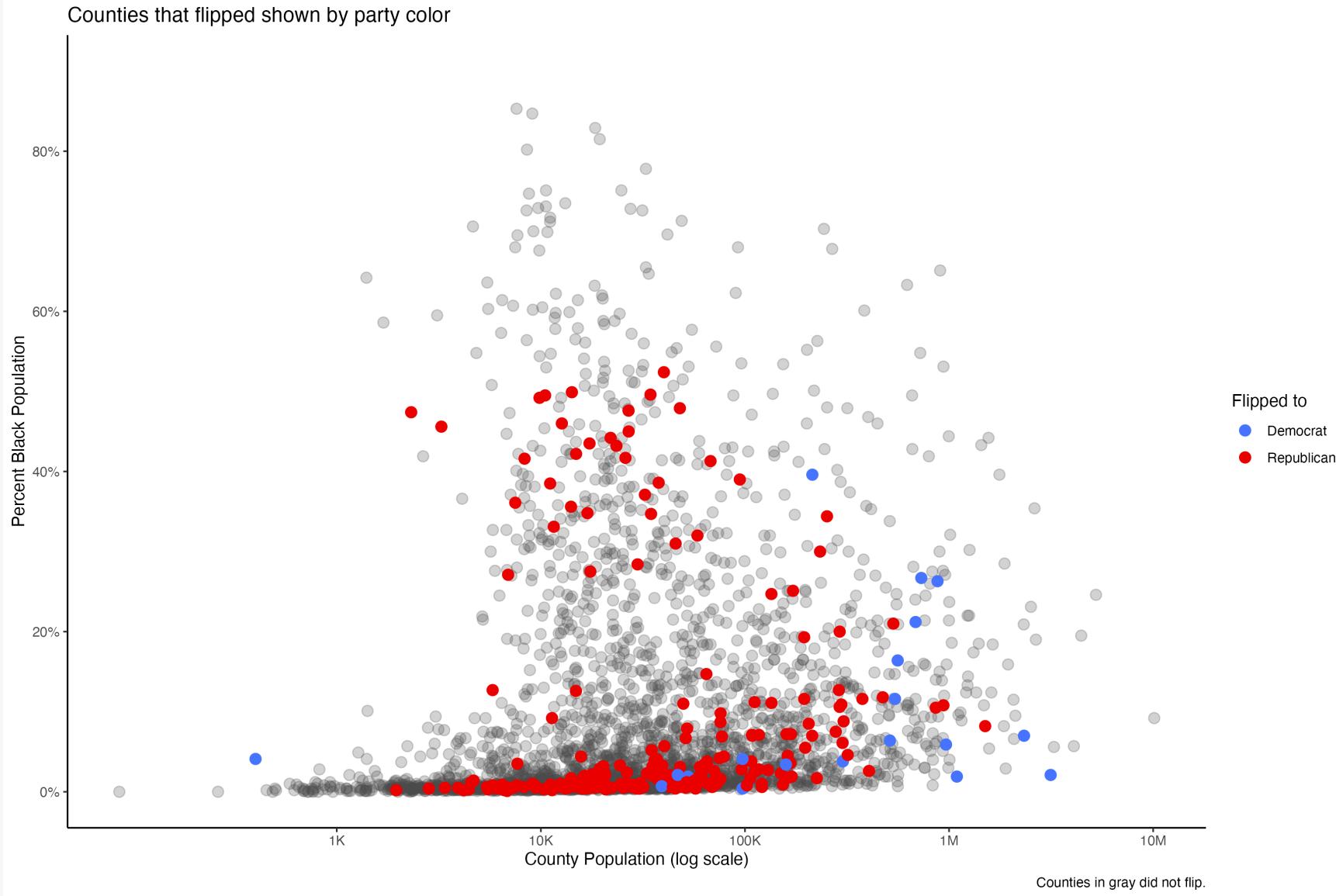
For example ...

```
p + theme(panel.grid.major.y =  
          element_line(color = "red"),  
          panel.grid.minor.y =  
          element_line(color = "black",  
                      linetype = "dotted"))
```



The ggthemes package

Counties that flipped shown by party color



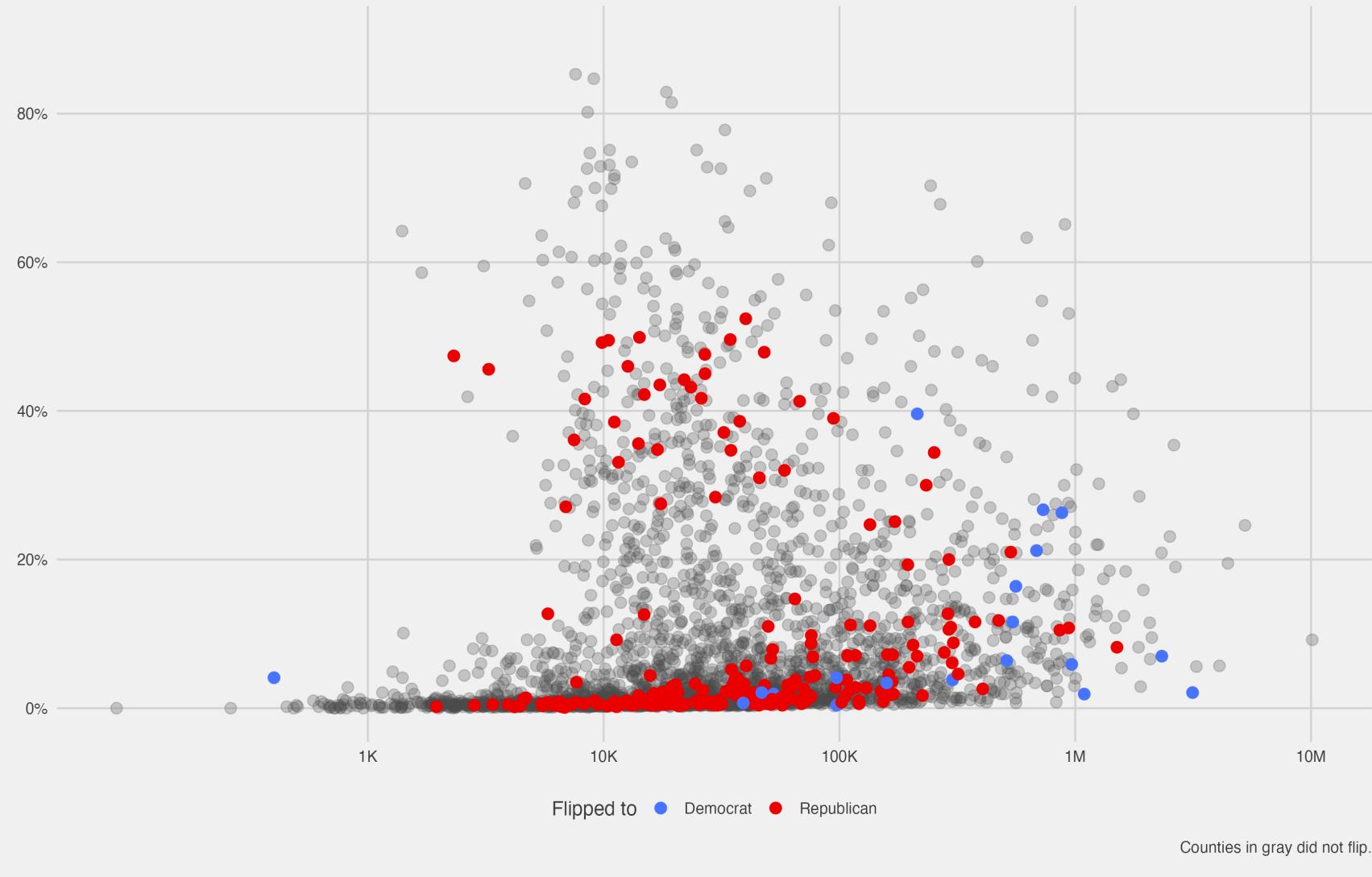
We made this earlier. Here it is in a default theme.

Theming a plot

```
library(ggthemes)
theme_set(theme_fivethirtyeight())
```

See how the full function call goes inside `theme_set()`, including the parentheses, because we are actually running that function to set all the elements.

Counties that flipped shown by party color



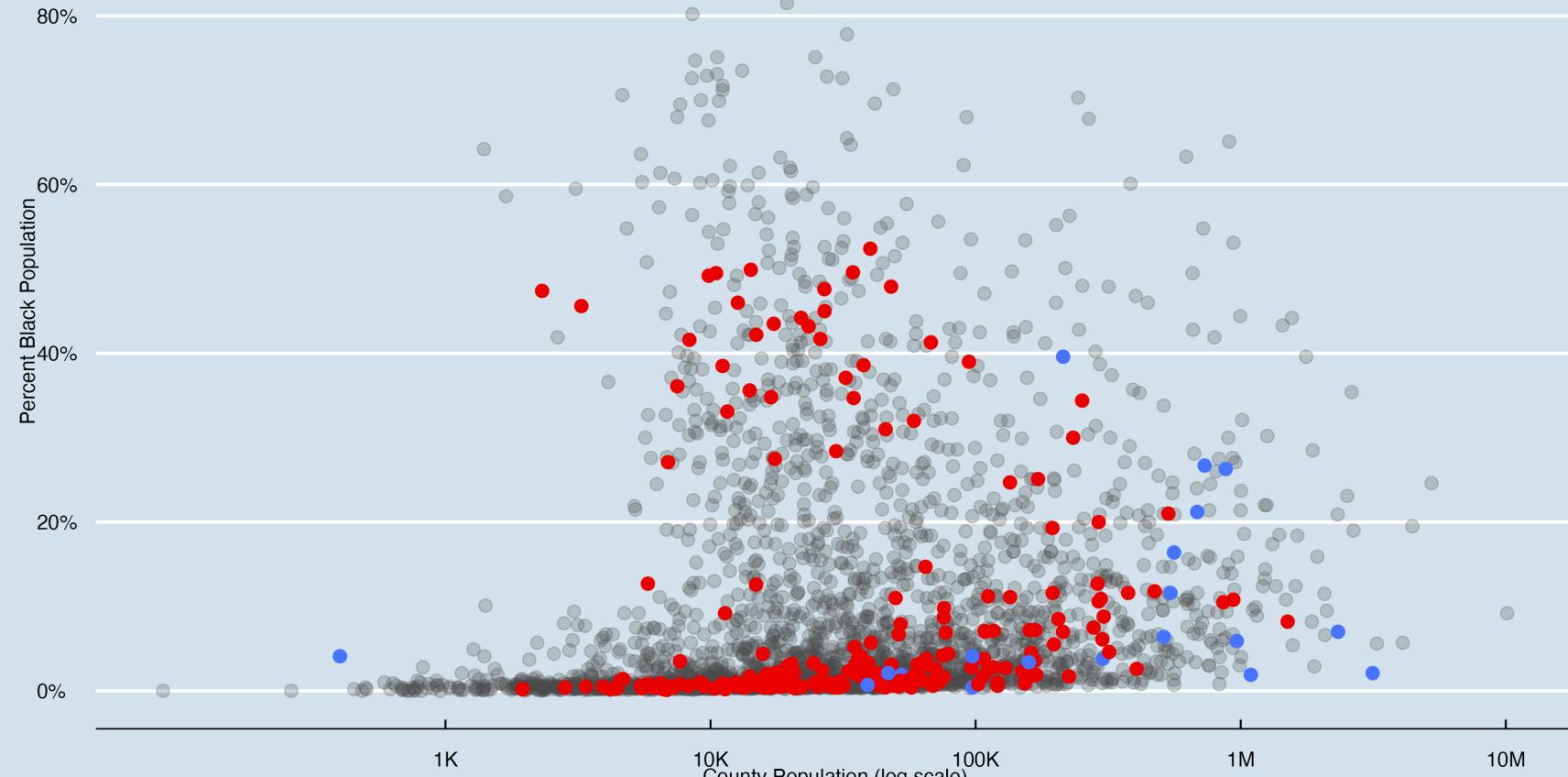
Calling the object now draws the plot with the thematic elements added.

Theming a plot

```
theme_set(theme_economist())
```

Counties that flipped shown by party color

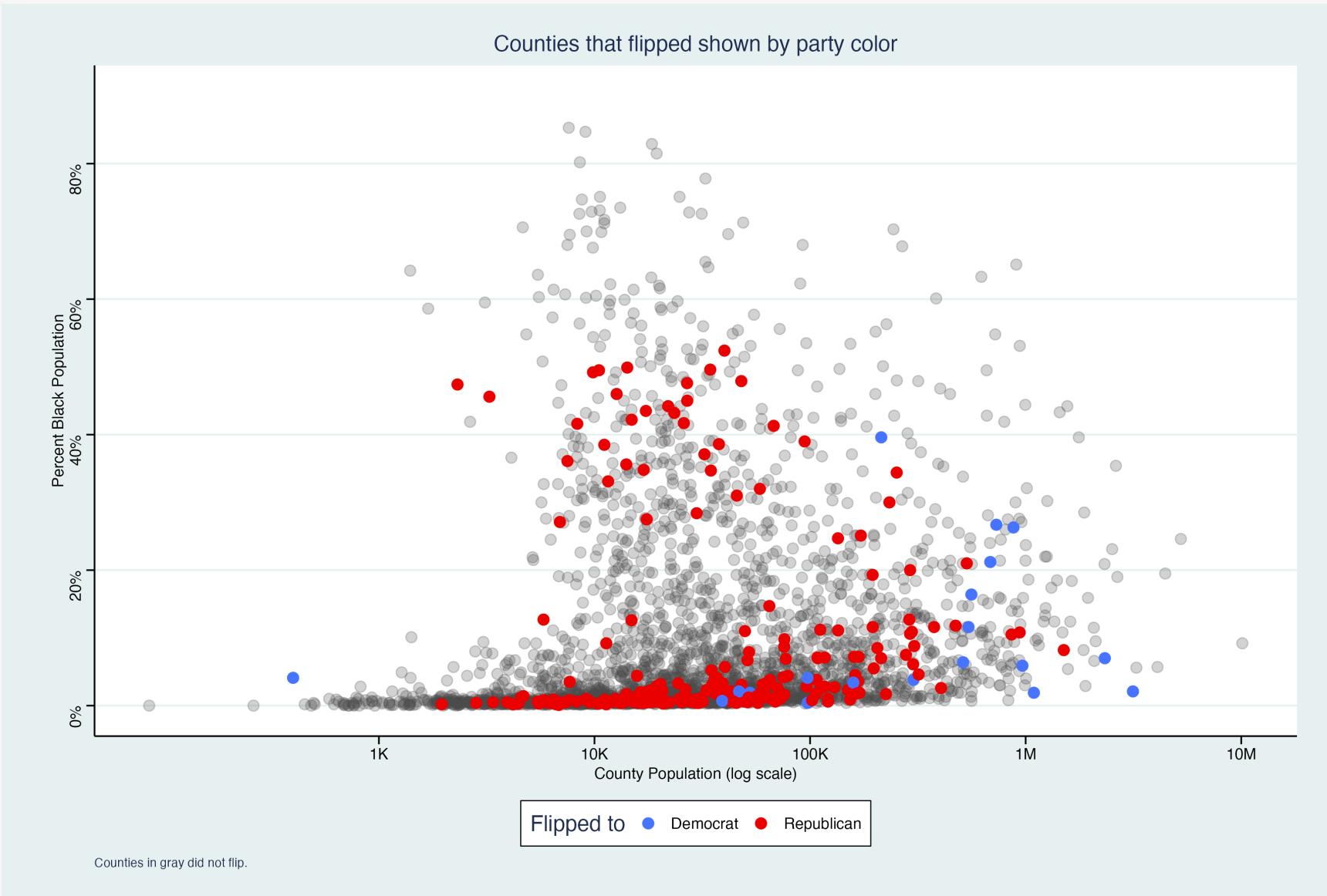
Flipped to ● Democrat ● Republican



Calling the object now draws the plot with the thematic elements added.

This seems morally wrong

```
theme_set(theme_stata())
```



Why would you do this to yourself?

Themes are just functions

E.g. the built-in `theme_bw()` modifies `theme_grey()` with a bunch of additional options, expressed in a `theme()` call:

```
theme_bw

function (base_size = 11, base_family = "", base_line_size = base_size/22,
          base_rect_size = base_size/22)
{
  theme_grey(base_size = base_size, base_family = base_family,
             base_line_size = base_line_size, base_rect_size = base_rect_size) %+replace%
  theme(panel.background = element_rect(fill = "white",
                                         colour = NA), panel.border = element_rect(fill = NA,
                                         colour = "grey20"), panel.grid = element_line(colour = "grey92"),
                                         panel.grid.minor = element_line(linewidth = rel(0.5)),
                                         strip.background = element_rect(fill = "grey85",
                                         colour = "grey20"), complete = TRUE)
}
<bytecode: 0x16f368070>
<environment: namespace:ggplot2>
```

but with the special `%+replace%` operator. You can use `+` if you like. The difference (read the help for `theme_get()`) is that `%+replace%` swaps out the *entire* element being updated, whereas `+` does not.

You can make themes

To make a theme yourself, you can begin entirely from scratch or you can adapt an existing theme (like `theme_bw()` does. For example:

```
theme_rice ← function(base_size = 14,
                      base_family = "",
                      base_line_size = base_size/22,
                      base_rect_size = base_size/22) {

  ## See https://brand.rice.edu/colors/
  riceblue1 ← "#00205B"
  riceblue2 ← "#6D7D9C"
  ricegray1 ← "#686B6C"
  ricegray2 ← "#B0B2B2"
  textcolor ← "#FFFFFF"
  half_line ← base_size / 2

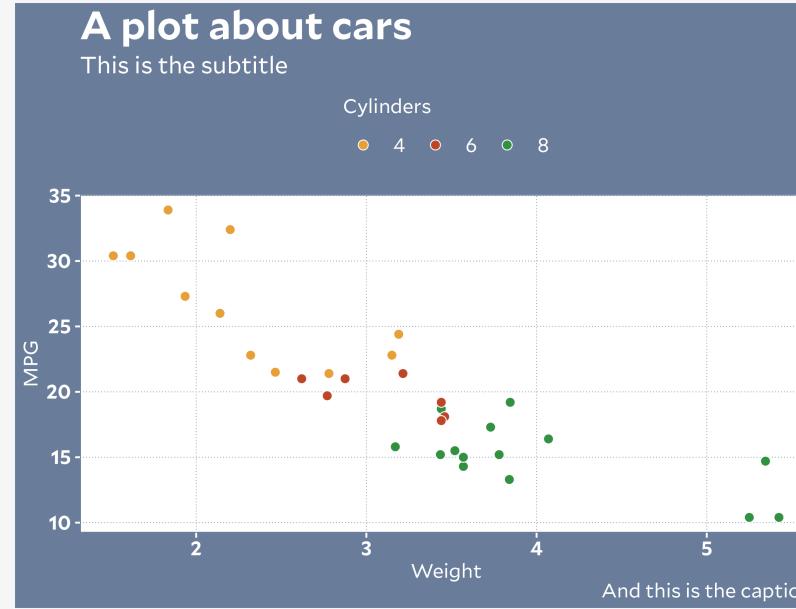
  ggplot2::theme_minimal(base_size = base_size, base_family = base_family,
                        base_line_size = base_line_size, base_rect_size = base_rect_size) %>%
    theme(
      axis.text = element_text(color = textcolor,
                               face = "bold"),
      axis.ticks = element_line(color = textcolor),
      axis.title = element_text(color = textcolor),
      plot.title = element_text(color = textcolor,
                                face = "bold",
                                size = rel(2),
                                # left-justified text
```

Base Plot

```
p ← ggplot(data = mtcars,
             mapping = aes(x = wt,
                            y = mpg,
                            fill = factor(cyl))) +
  geom_point(size = rel(3),
             shape = 21, color = "white") +
  labs(x = "Weight", y = "MPG", fill = "Cylinders",
       title = "A plot about cars",
       subtitle = "This is the subtitle",
       caption = "And this is the caption")
```

Now we can do, e.g.,

```
rice_accents ← c("#E9A139", "#C04829", "#359245", "#362E52", "#005B50")  
  
p + scale_fill_manual(values = rice_accents) +  
  theme_rice(base_family = "Mallory") # See https://brand.rice.edu
```



This is very rough, and I daresay ill-advised. But you see how much control you have over how things look. Also it serves as a reminder that you will likely have to do quite a bit of detail-oriented work if you want a properly good custom theme. Choose one and stick with it.