

# Colors!

*Daniel Anderson  
Week 5, Class 1*



# Agenda

- Color basics
  - 3 basic ways color is used
- Color blindness
- Some common problems with color use
- Quick discussion of palettes

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- Color basics
  - 3 basic ways color is used
- Color blindness
- Some common problems with color use
- Quick discussion of palettes

## *Learning Objectives*

- Understand different types of color palettes
  - ...and when you should use one versus another
- Understand and be able to effectively evaluate concerns related to color blindness
- Be able to fluently change colors/fills within ggplot

# Before we get too deep

*Some very practical advice*

- Keep straight when color is mapped to a variable through `aes` and when it's modifying an element overall
  - Former requires `scale_color_*` or `scale_fill_*` while the latter does not

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- Keep straight colors and fills (see former bullet)

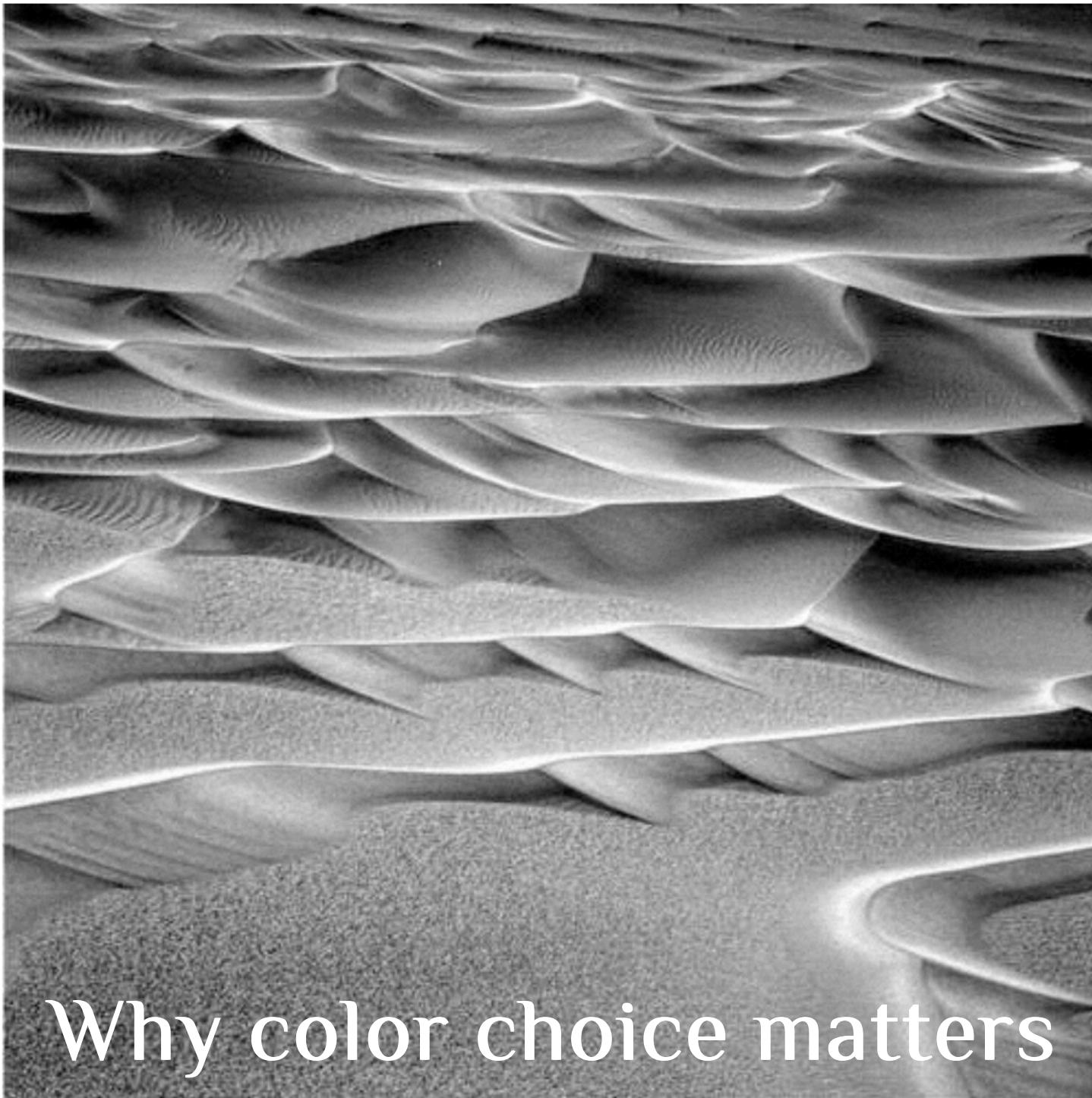
# Before we get too deep

*Some very practical advice*

- Keep straight when color is mapped to a variable through `aes` and when it's modifying an element overall
  - Former requires `scale_color_*` or `scale_fill_*` while the latter does not
- Keep straight colors and fills (see former bullet)
- Use advice of others to your advantage (e.g., <http://colorbrewer2.org/>)



Why color choice matters



Why color choice matters

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# Another quick example

## rayshader

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# Three fundamental uses of color

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1. Distinguish groups from each other

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2. Represent data values

# Three fundamental uses of color

1. Distinguish groups from each other
2. Represent data values
3. Highlight

# Color as a tool to distinguish

# Discrete items

- Often no intrinsic order

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- Often no intrinsic order

## *Qualitative color scale*

- Finite number of colors
  - Chosen to maximize distinctness, while also be equivalent
  - Equivalent
    - No color should stand out
    - No impression of order

# Some examples

Okabe Ito



ColorBrewer Dark2



ggplot2 hue



See more about the Okabe Ito palette origins here: <http://jfly.iam.u-tokyo.ac.jp/color/>

# How do we use them?

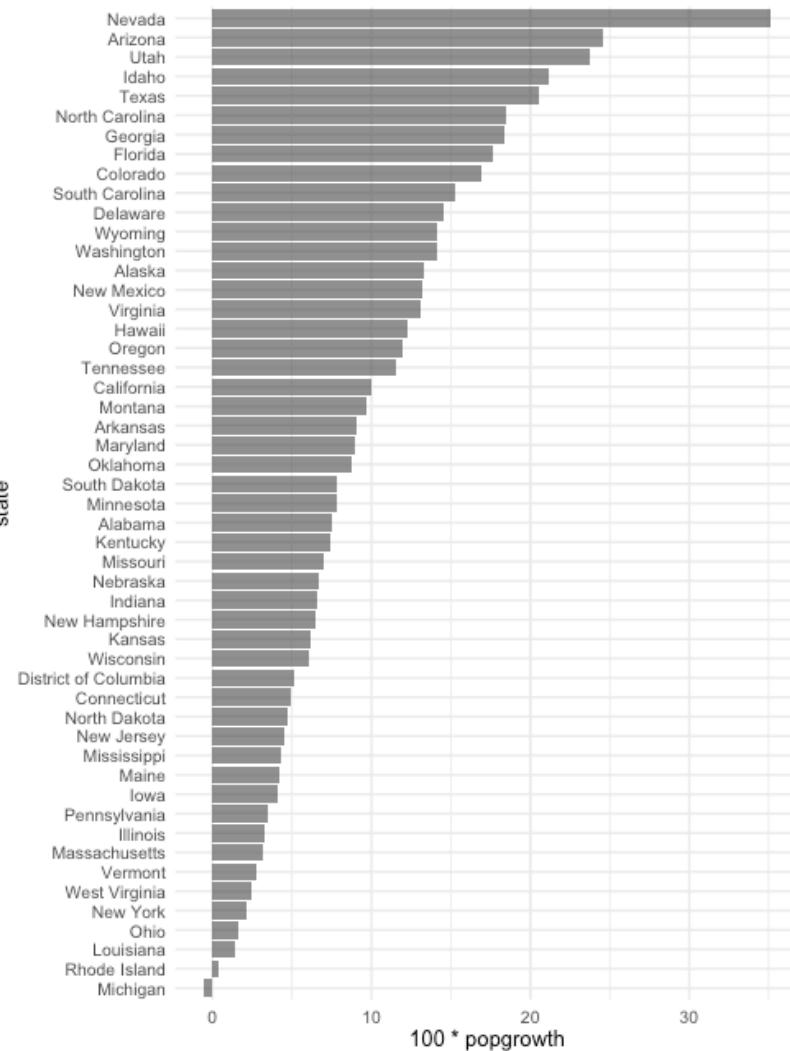
Imagine we have data like this

```
popgrowth_df
```

```
## # A tibble: 51 x 7
##   region    division      state    pop2000    pop2010    popgrowth    area
##   <fct>     <chr>       <fct>     <dbl>      <dbl>      <dbl>      <dbl>
## 1 Midwest   East North Ce... Michigan  9938444  9883640 -0.005514344 56538.92
## 2 Northea... New England   Rhode Is... 1048319   1052567  0.004052202 1033.82
## 3 South     West South Ce... Louisiana 4468976   4533372  0.01440956  43203.93
## 4 Midwest   East North Ce... Ohio     11353140  11536504  0.01615095  40860.73
## 5 Northea... Middle Atlant... New York 18976457  19378102  0.02116544  47126.43
## 6 South     South Atlantic West Vir... 1808344   1852994  0.02469110  24038.21
## # ... with 45 more rows
```

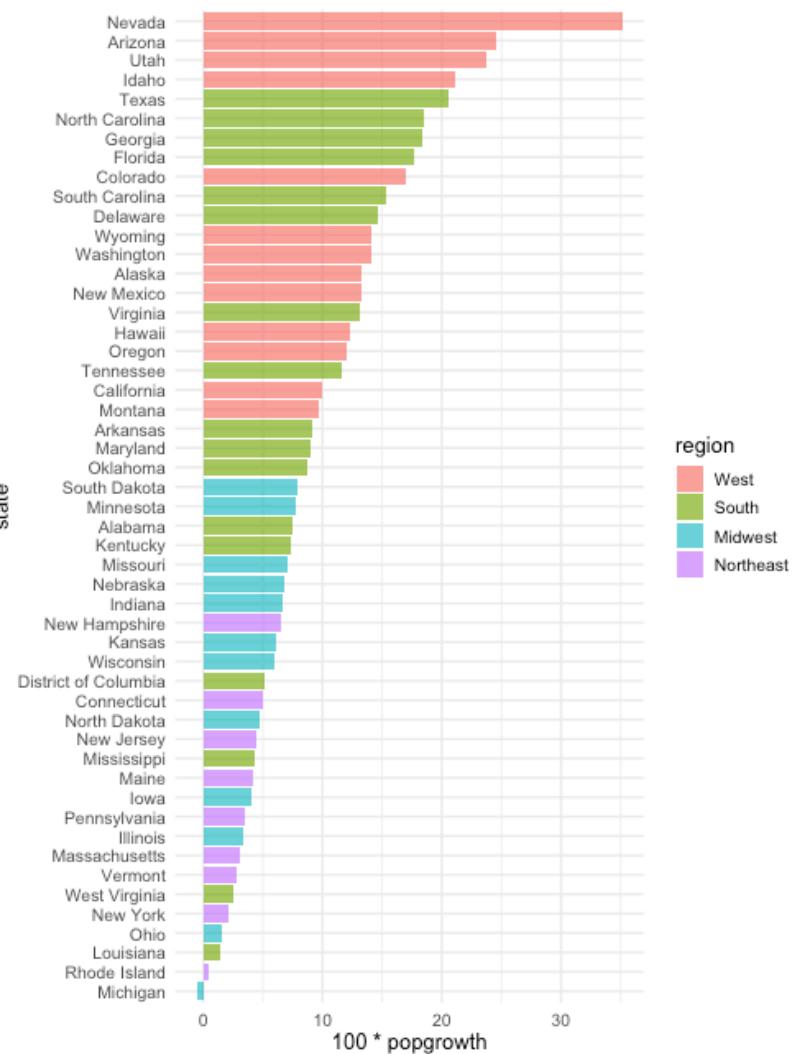
# We might build a plot like this

```
ggplot(popgrowth_df,  
       aes(x = state,  
            y = 100*popgrowth)) +  
  geom_col(alpha = 0.7) +  
  coord_flip()
```

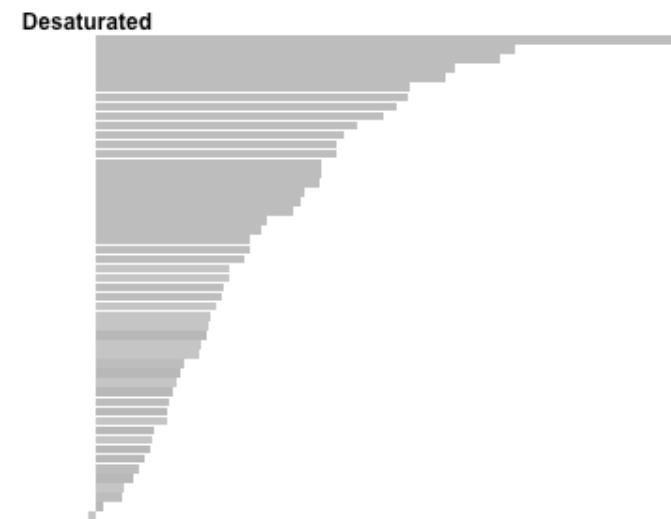
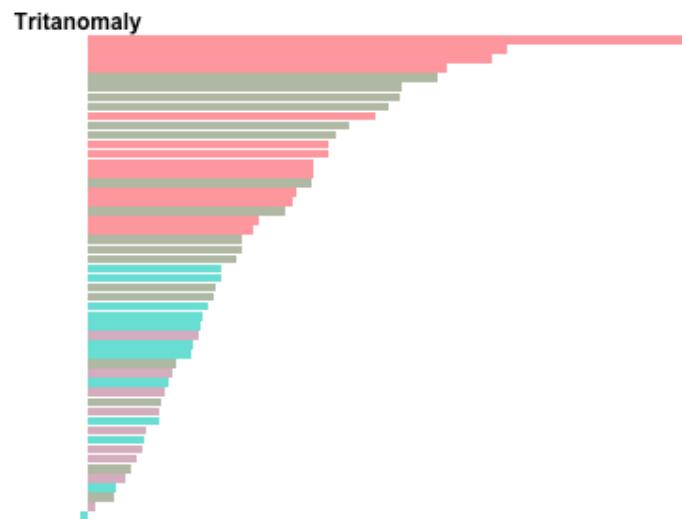
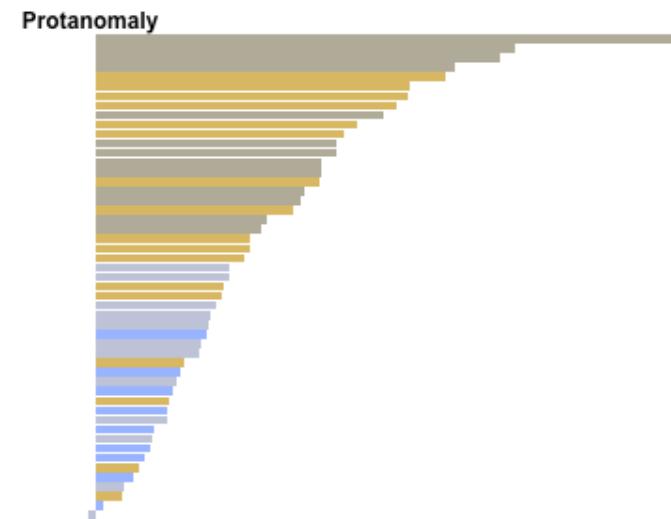
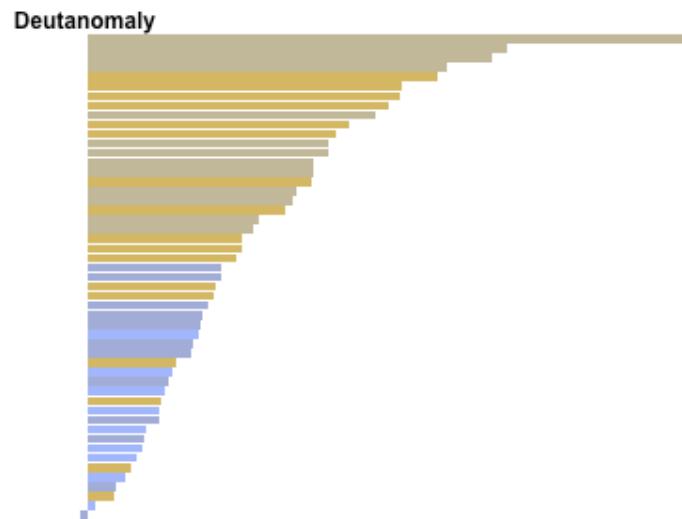


# Alternatively, fill by region

```
ggplot(popgrowth_df,  
       aes(x = state,  
            y = 100*popgrowth)) +  
  geom_col(aes(fill = region),  
           alpha = 0.7) +  
  coord_flip()
```

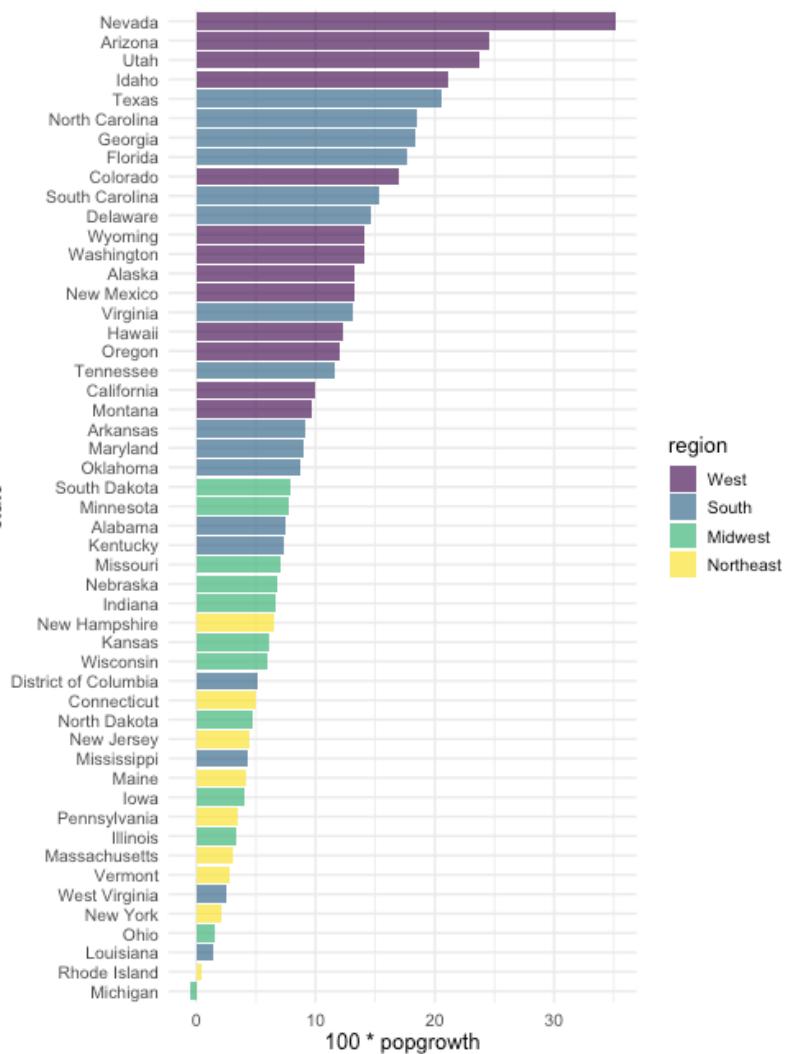


# Problem with default palette

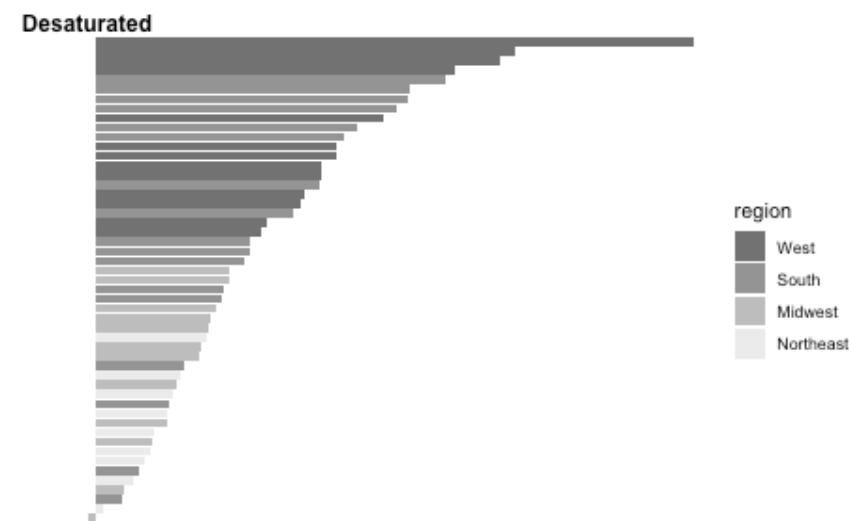
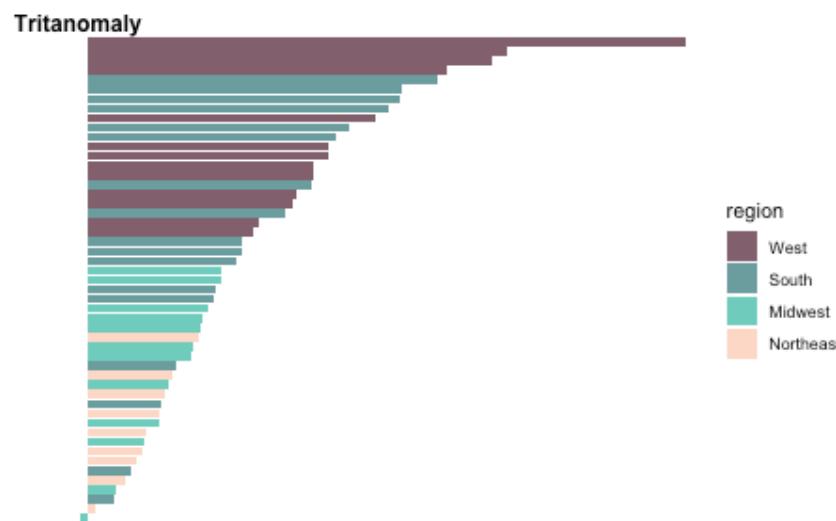
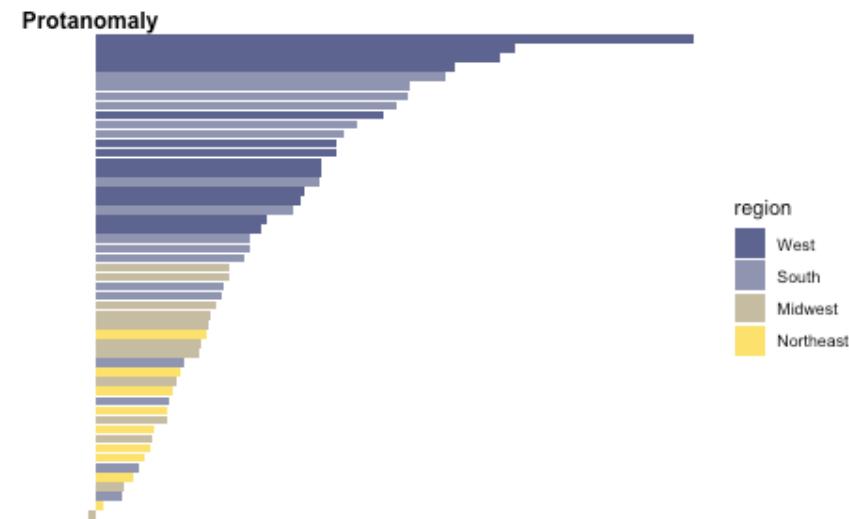
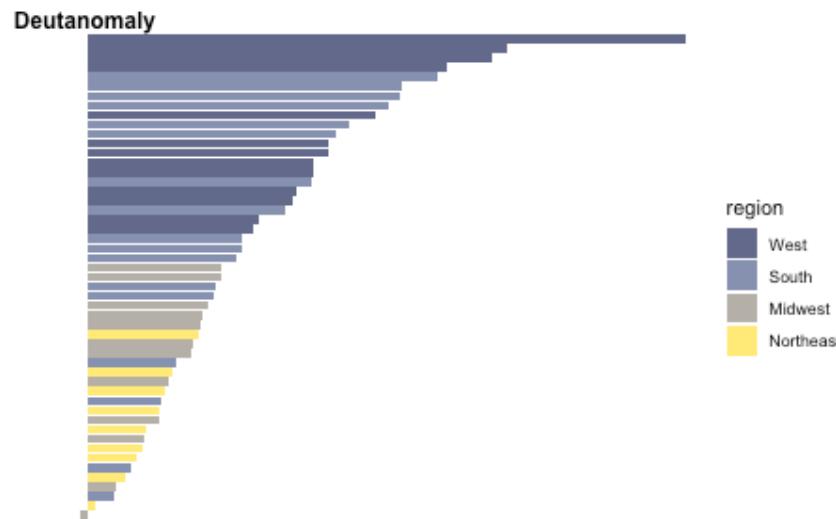


# Alternative: viridis

```
ggplot(popgrowth_df,  
       aes(x = state,  
            y = 100*popgrowth)) +  
  geom_col(aes(fill = region),  
           alpha = 0.7) +  
  scale_fill_viridis_d() +  
  coord_flip()
```



# Revised version

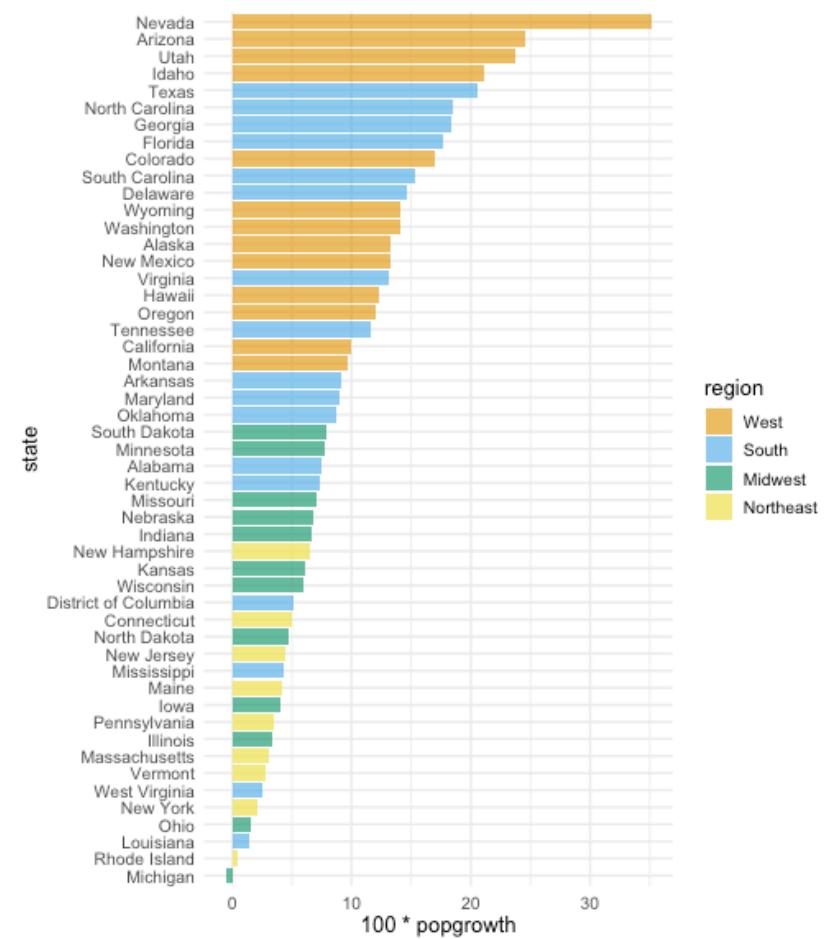


# The Okabe Ito palette

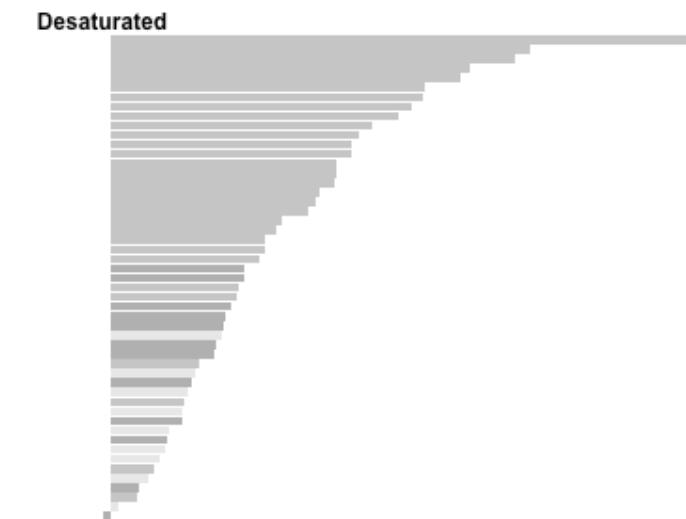
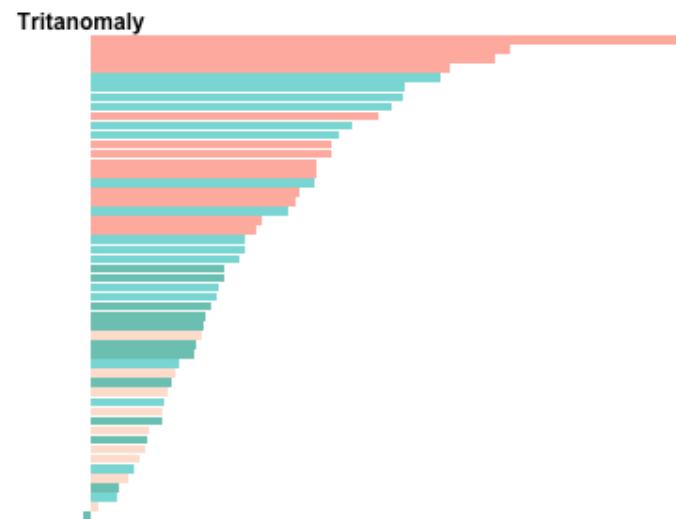
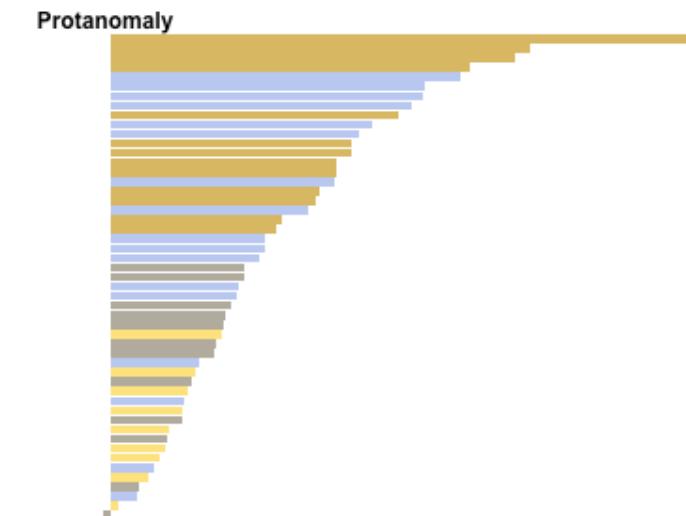
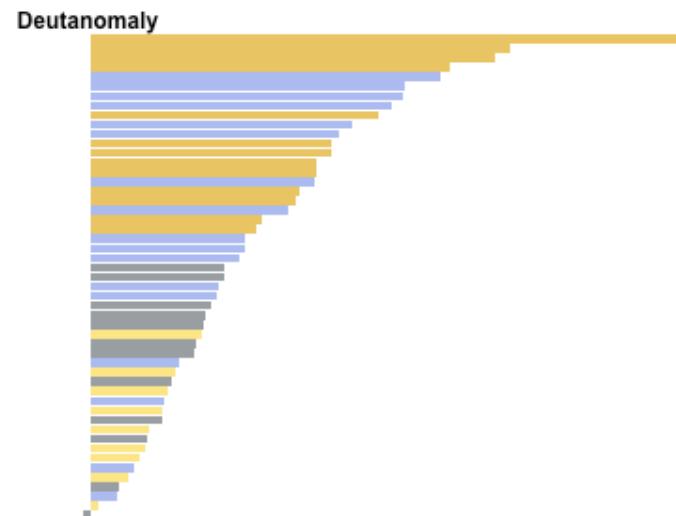
- From [Color Universal Design](#)

```
library(colorblindr)
```

```
ggplot(popgrowth_df,
       aes(x = state,
           y = 100*popgrowth)) +
  geom_col(aes(fill = region),
           alpha = 0.7) +
  scale_fill_OkabeIto() +
  coord_flip()
```



# Okabe Ito for colorblindness



# How am I checking for colorblindness?

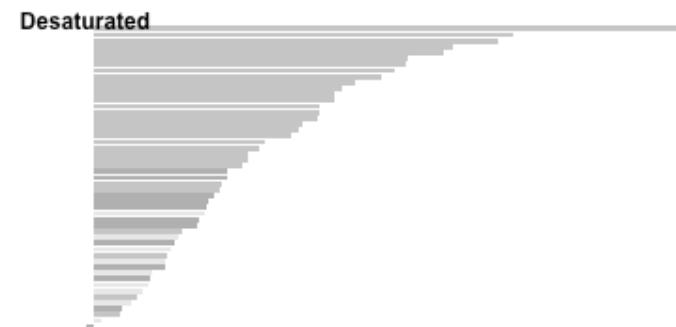
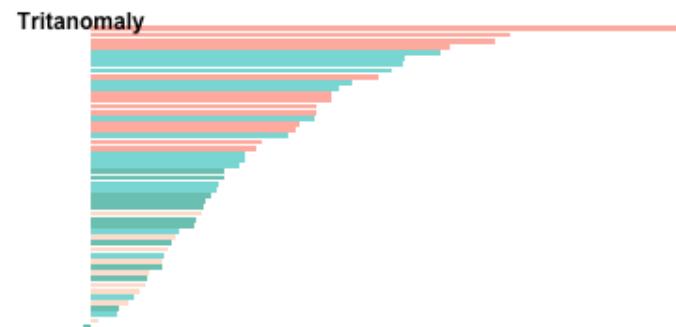
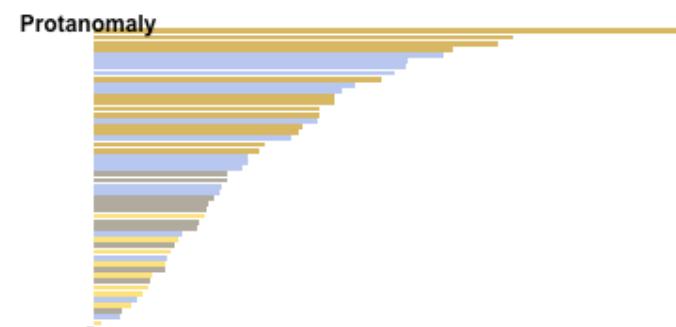
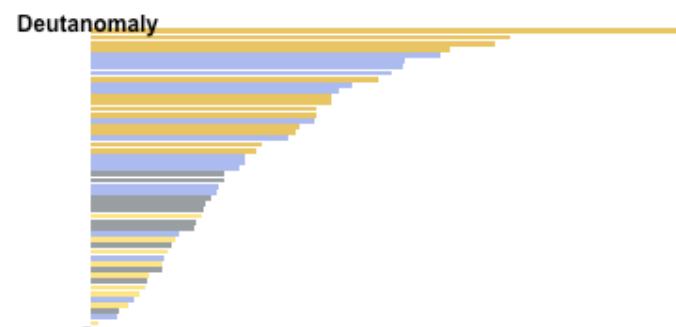
- Also part of the `colorblindr` package ([here](#))
  - depends on the dev versions of `colorspace` and `cowplot`, which are useful packages in their own right

```
devtools::install_github("wilkelab/cowplot")
install.packages("colorspace", repos = "http://R-Forge.R-project.org")

devtools::install_github("clauswilke/colorblindr")
```

```
p <- ggplot(popgrowth_df,
  aes(x = state,
      y = 100*popgrowth)) +
geom_col(aes(fill = region),
         alpha = 0.7) +
scale_fill_OkabeIto() +
coord_flip() +
theme_void() # not necessary but I like it
```

```
colorblindr::cvd_grid(p)
```



Colors to represent continuous  
values

# Sequential scale examples

ColorBrewer Blues



Heat



Viridis



# Sequential scale representation

- Which values are larger/smaller

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- Which values are larger/smaller
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  - Similar to an interval scale, but for color
- Often based on a single **hue**

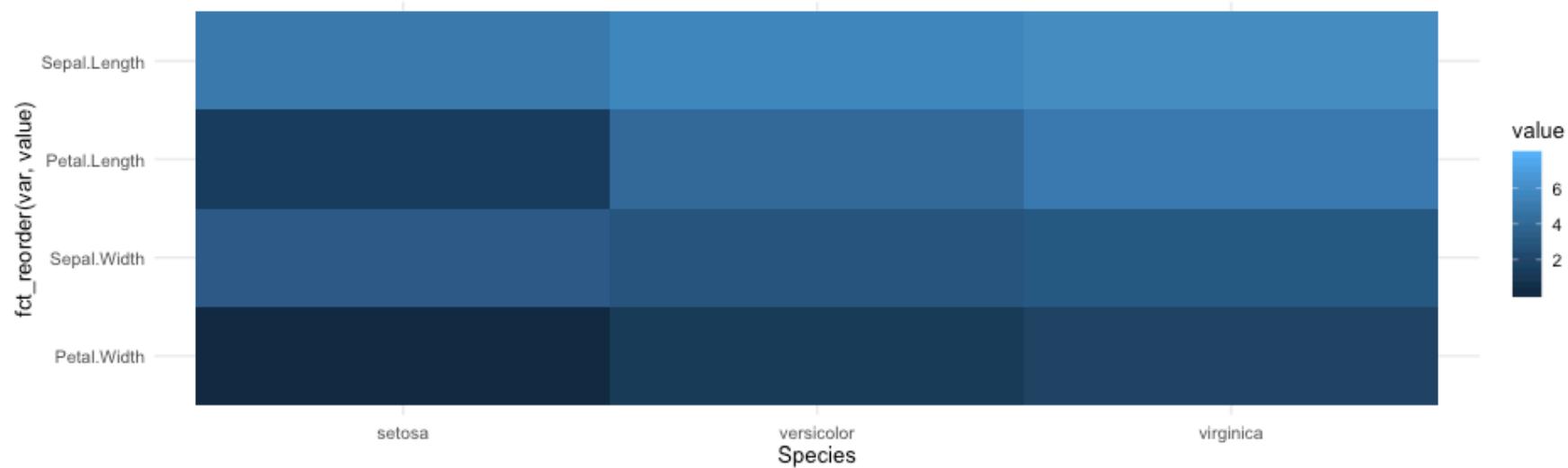
# Sequential scale representation

- Which values are larger/smaller
- How distant two values are from each other
  - Scale must be perceptually uniform across its entire range
  - Similar to an interval scale, but for color
- Often based on a single **hue**
- Multi-hue sequential scales tend to follow gradients in the natural world

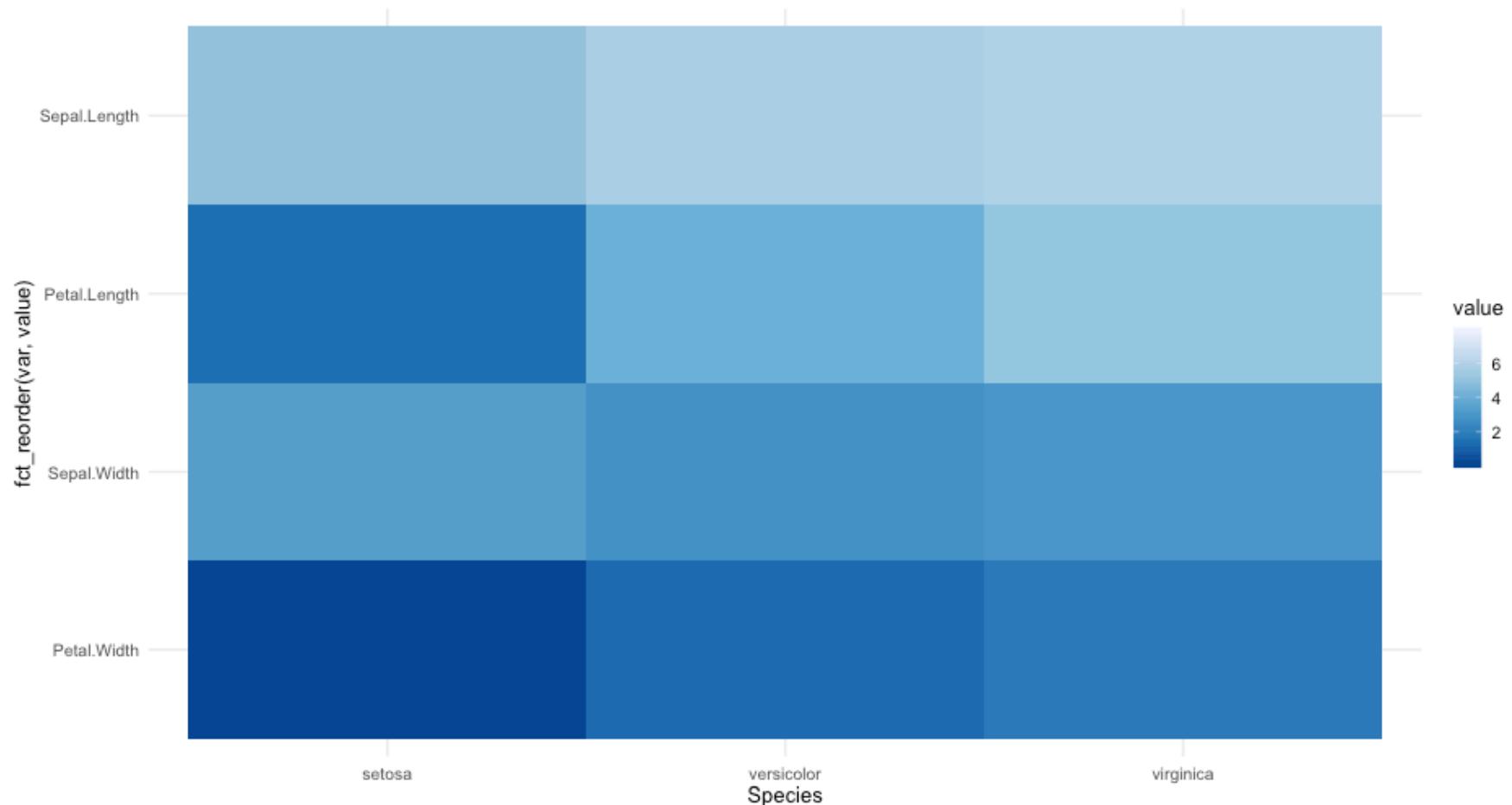
# Common use of sequential palettes

## *Heatmaps*

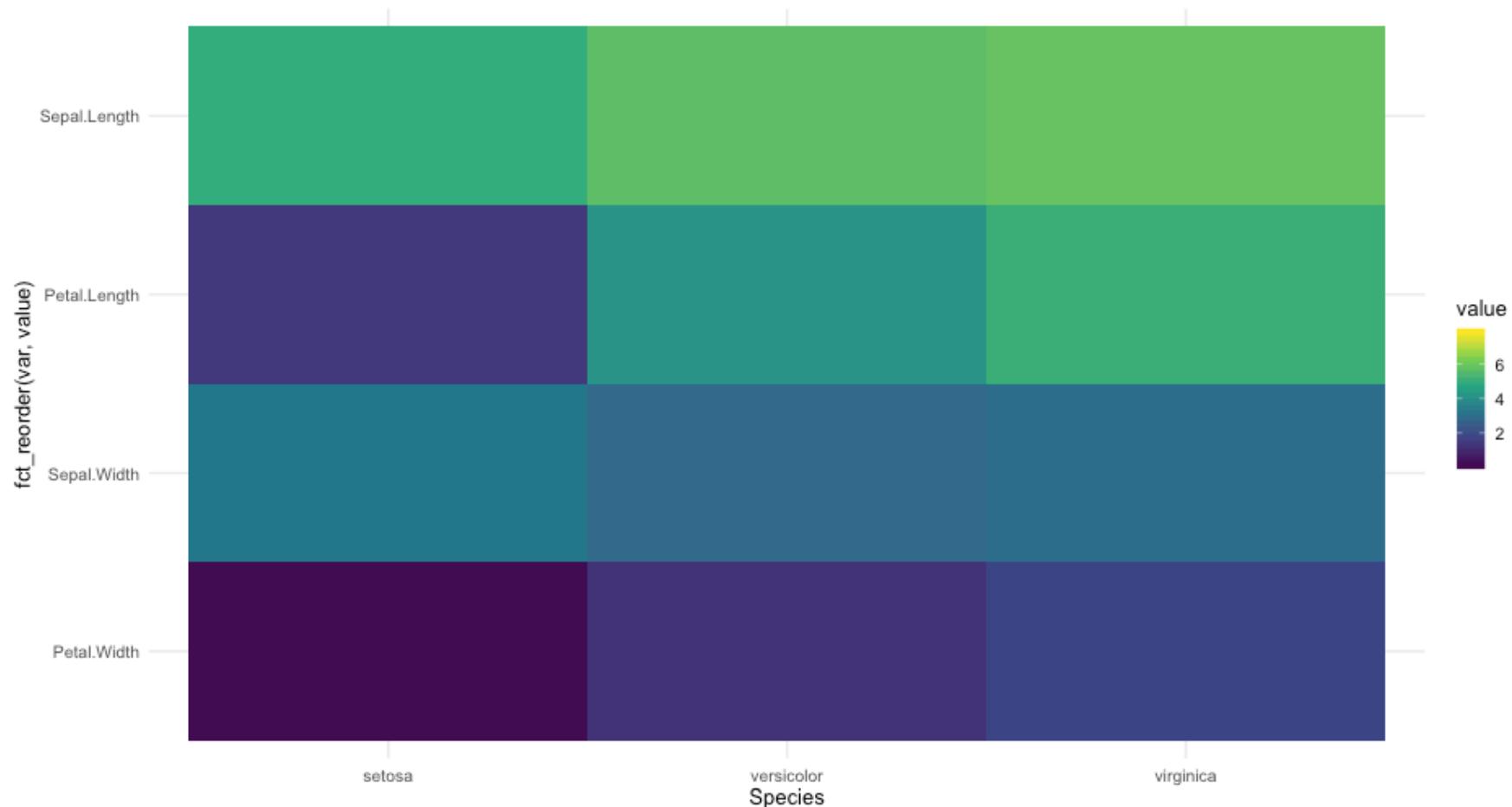
```
iris %>%
  gather(var, value, -Species) %>%
  ggplot(aes(Species, fct_reorder(var, value))) +
  geom_tile(aes(fill = value))
```



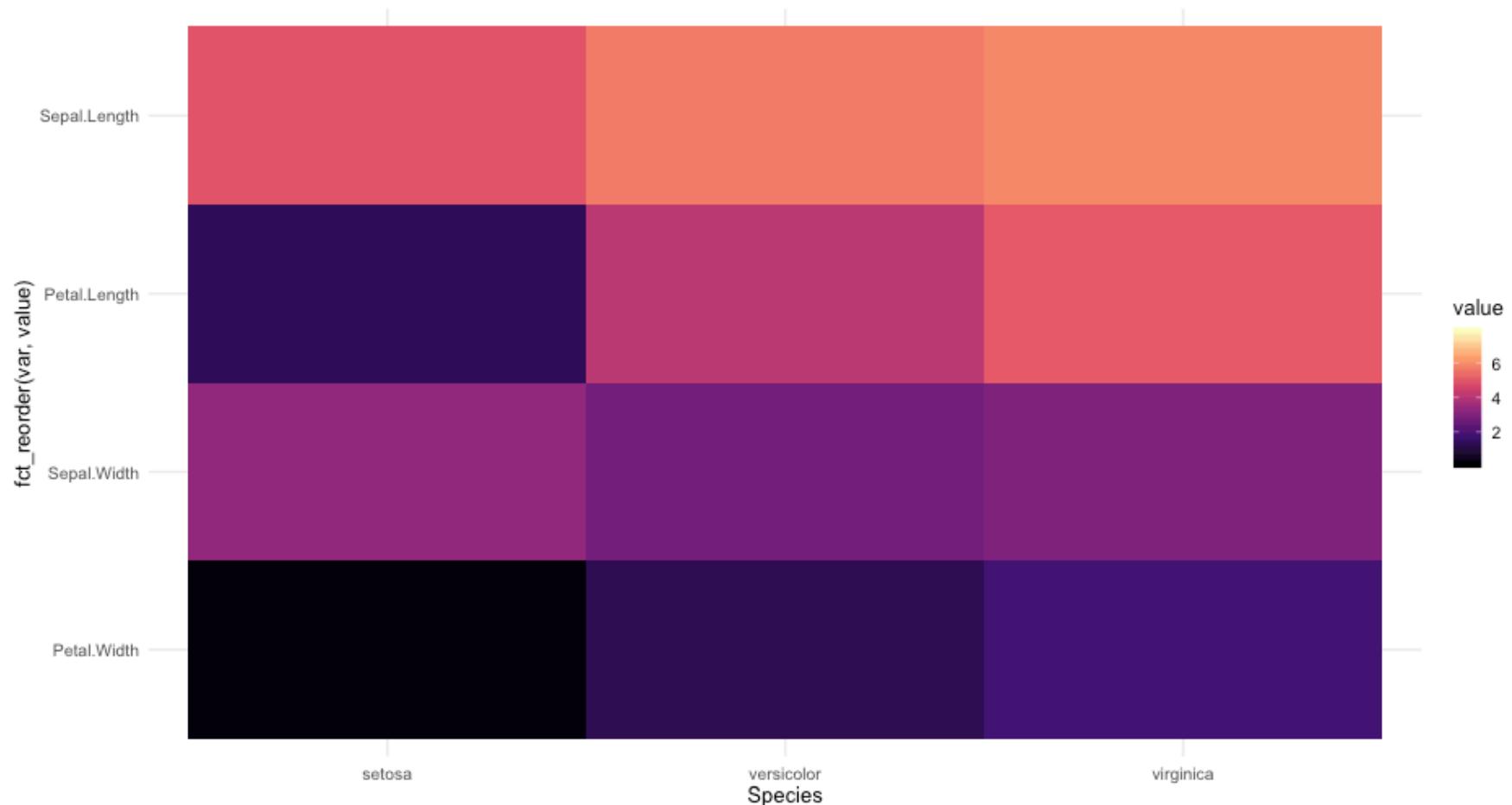
```
iris %>%
  gather(var, value, -Species) %>%
  ggplot(aes(Species, fct_reorder(var, value))) +
  geom_tile(aes(fill = value)) +
  scale_fill_distiller(palette = "Blues")
```



```
iris %>%
  gather(var, value, -Species) %>%
  ggplot(aes(Species, fct_reorder(var, value))) +
  geom_tile(aes(fill = value)) +
  scale_fill_viridis_c()
```



```
iris %>%
  gather(var, value, -Species) %>%
  ggplot(aes(Species, fct_reorder(var, value))) +
  geom_tile(aes(fill = value)) +
  scale_fill_viridis_c(option = "magma")
```

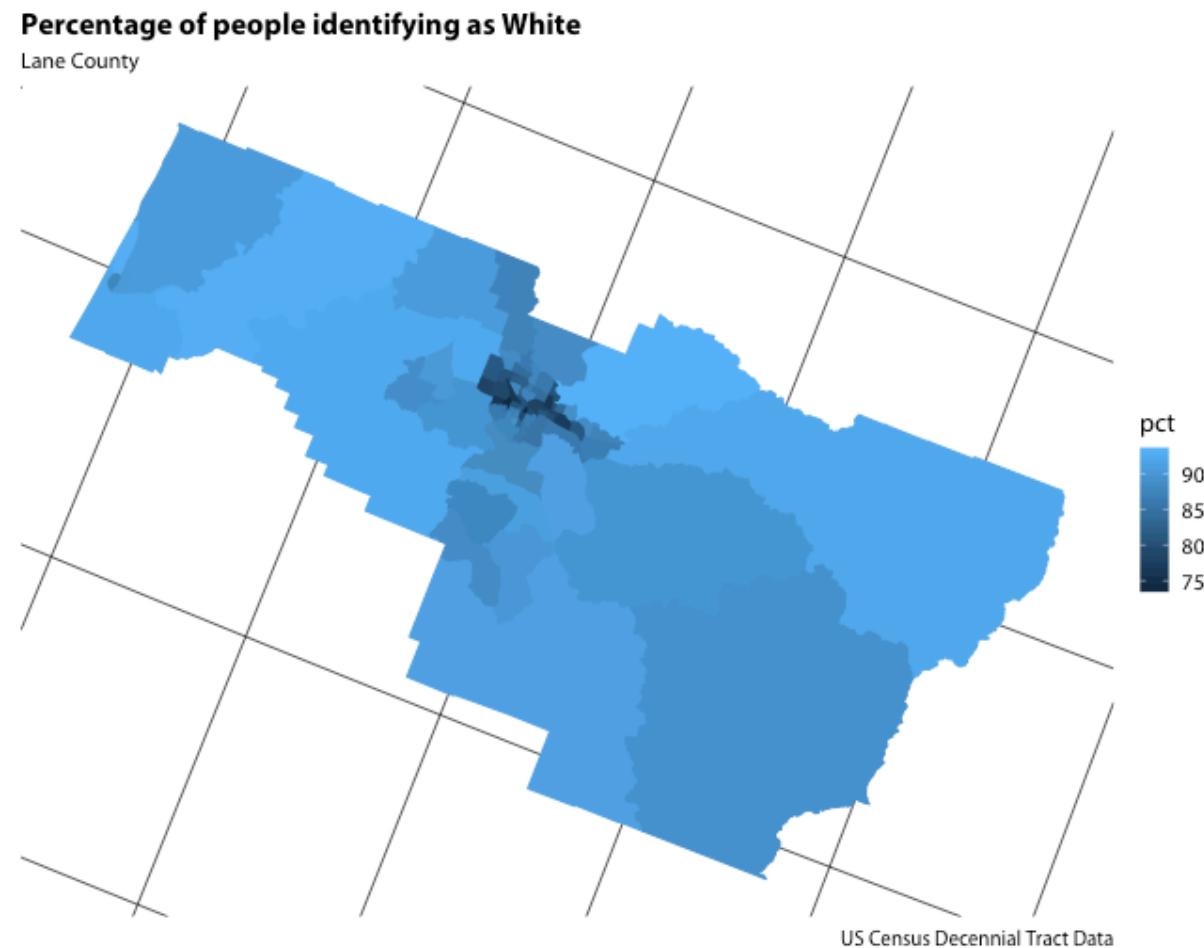


option = c("viridis", "magma", "inferno", "plasma")

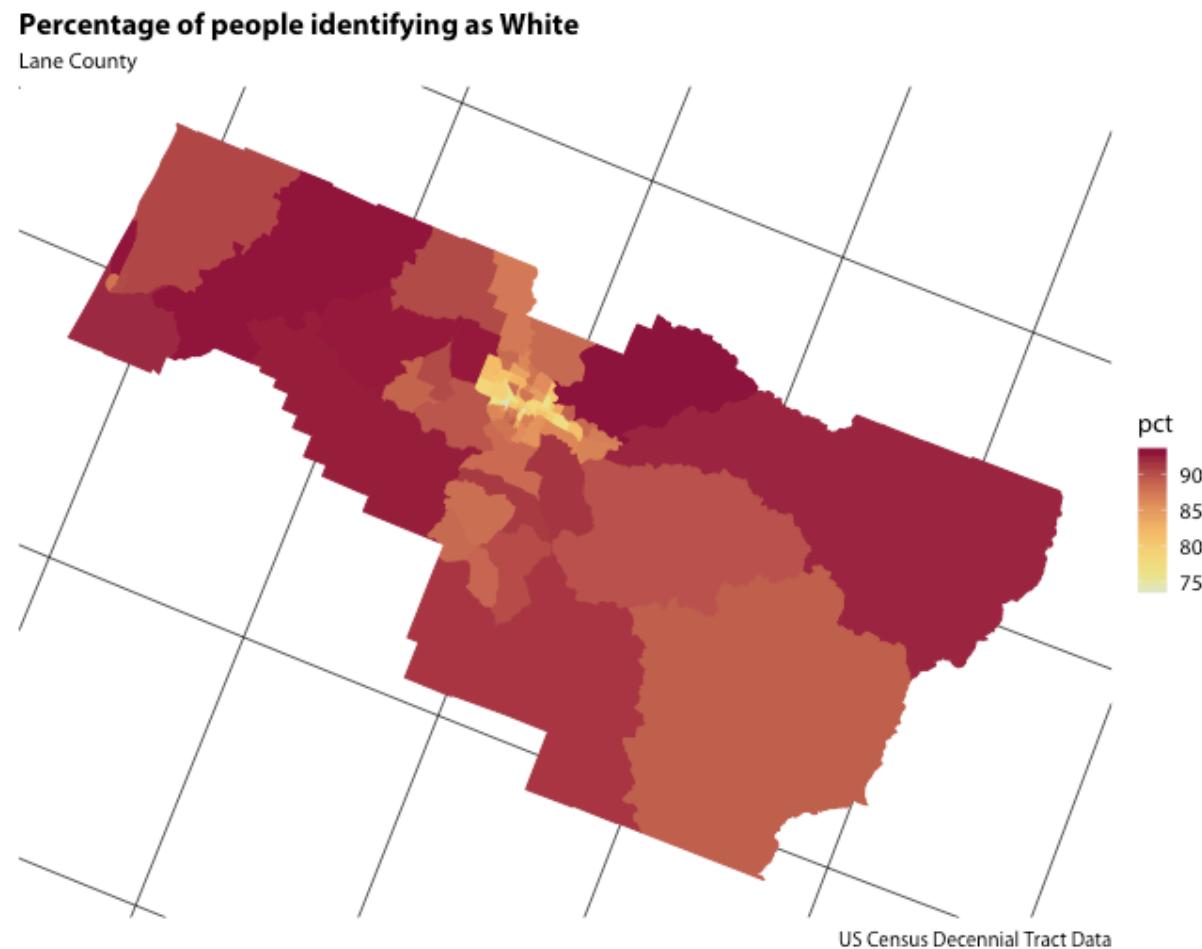
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# Another common use

*Choropleths*



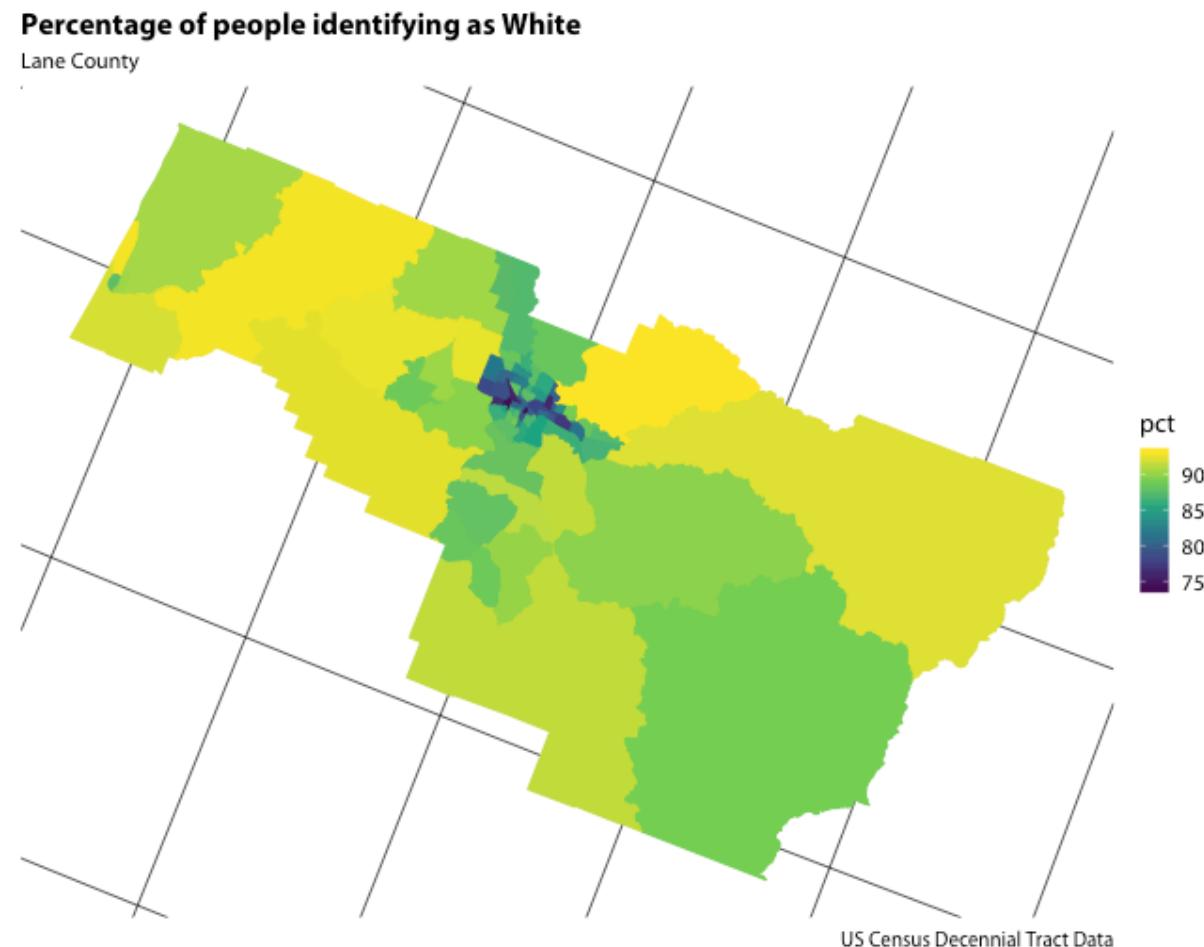
# Heat palette



# Options

- `scale_fill_continuous_sequential("Heat")`
- `scale_color_continuous_sequential("Heat")`
- `scale_fill_discrete_sequential("Heat")`
- `scale_color_discrete_sequential("Heat")`

# viridis palette

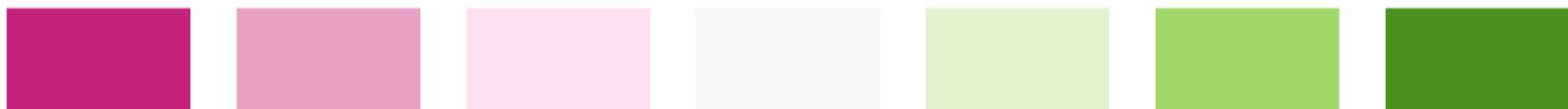


# Diverging palettes

CARTO Earth



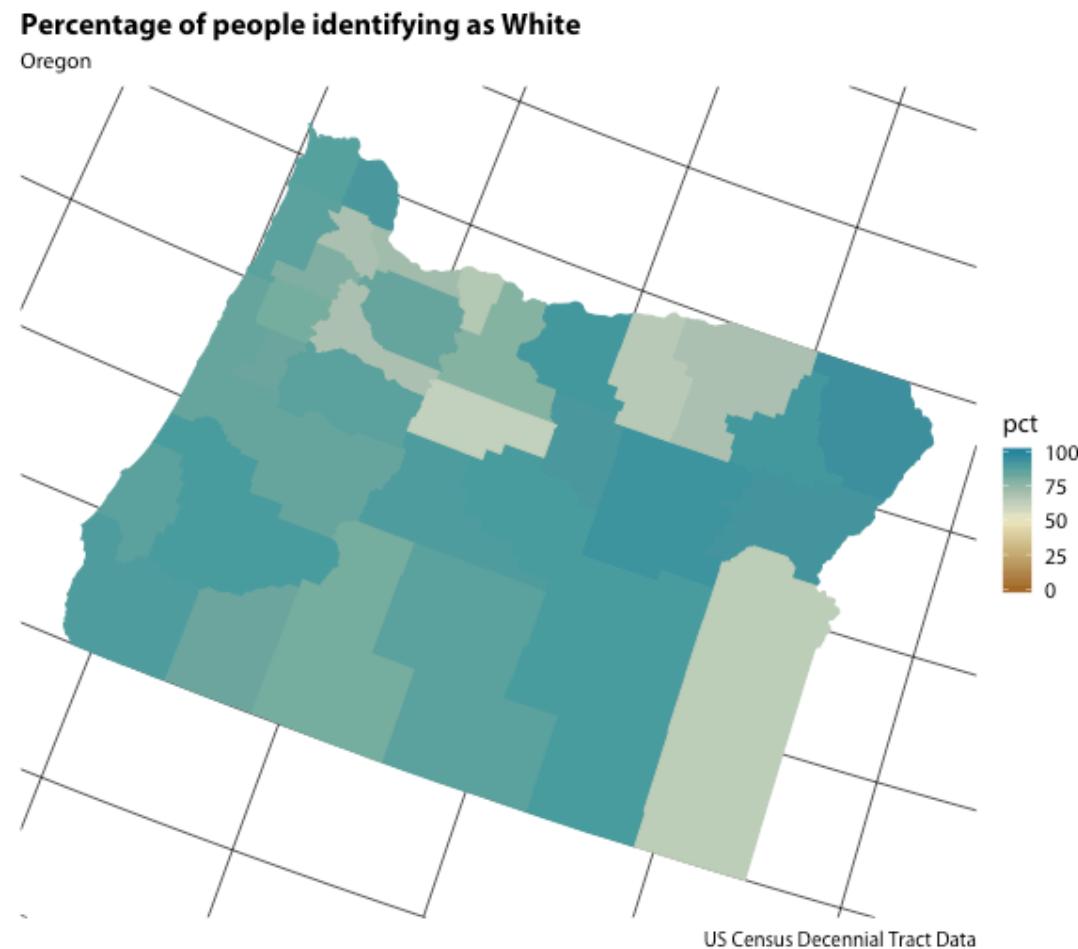
ColorBrewer PiYG



Blue-Red

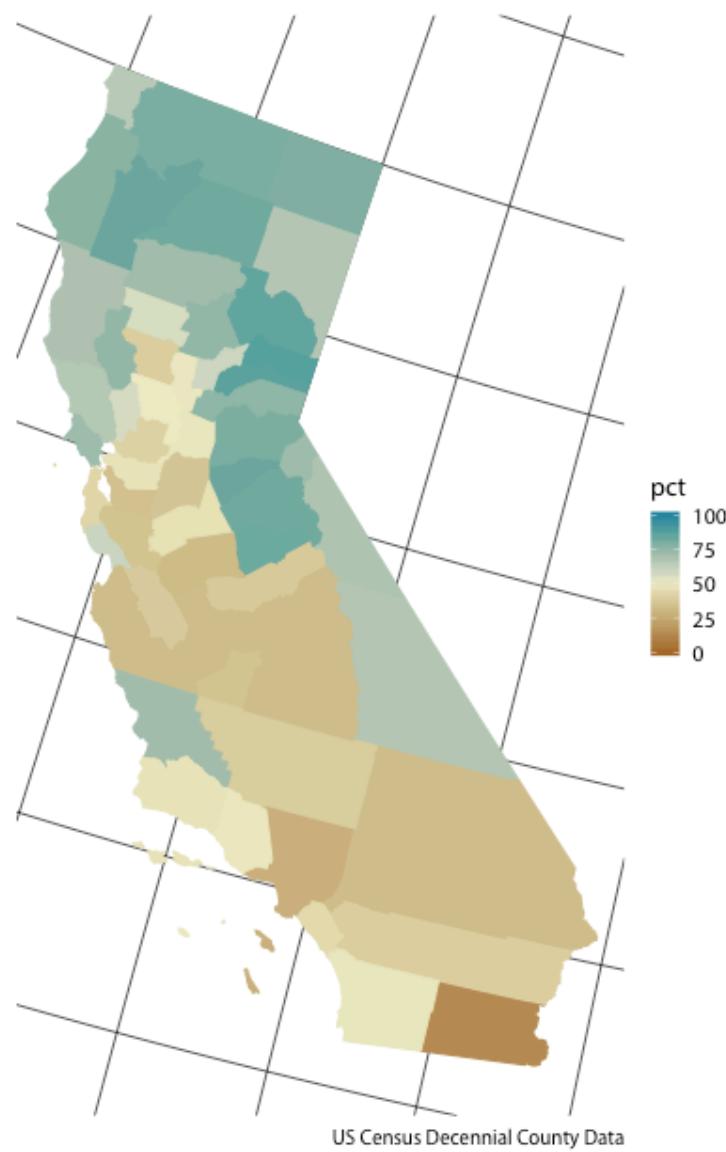


# Earth palette



### Percentage of people identifying as White

California

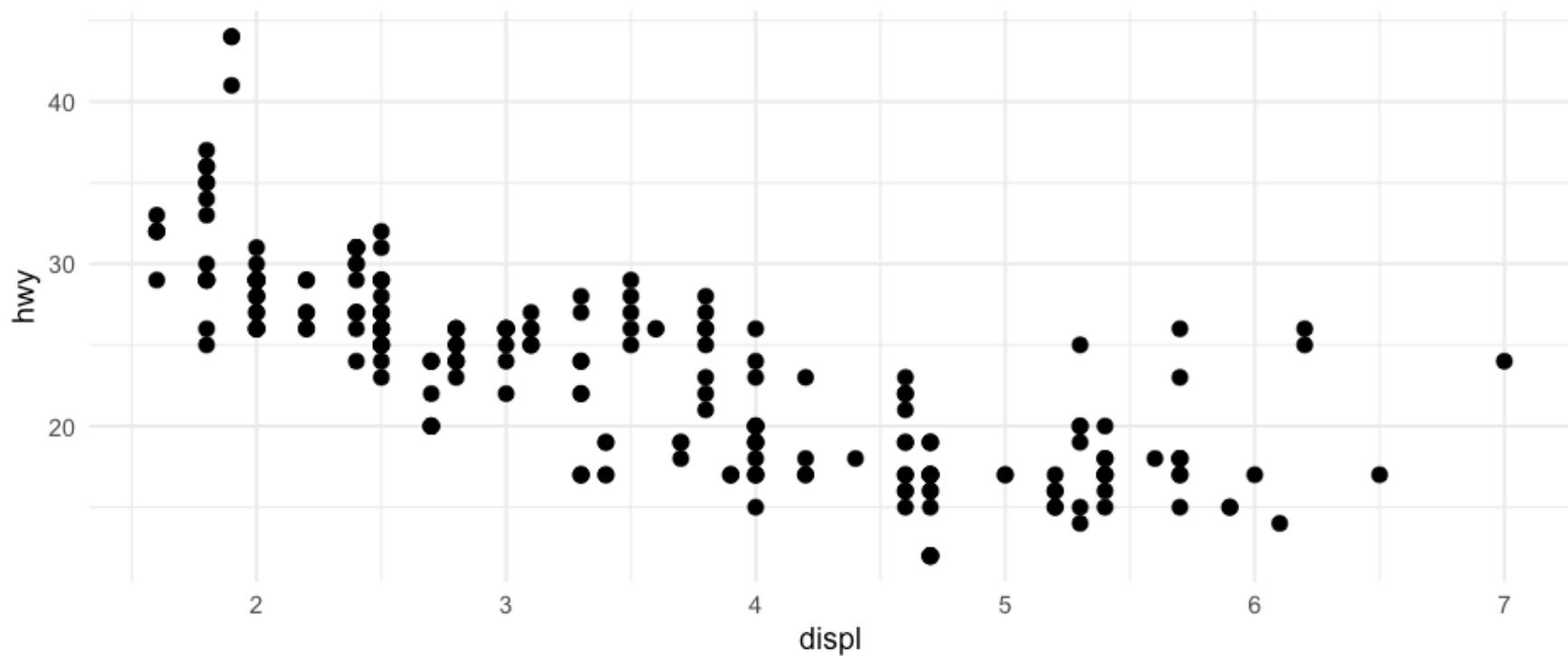


# Color as a tool to highlight

# MPG data

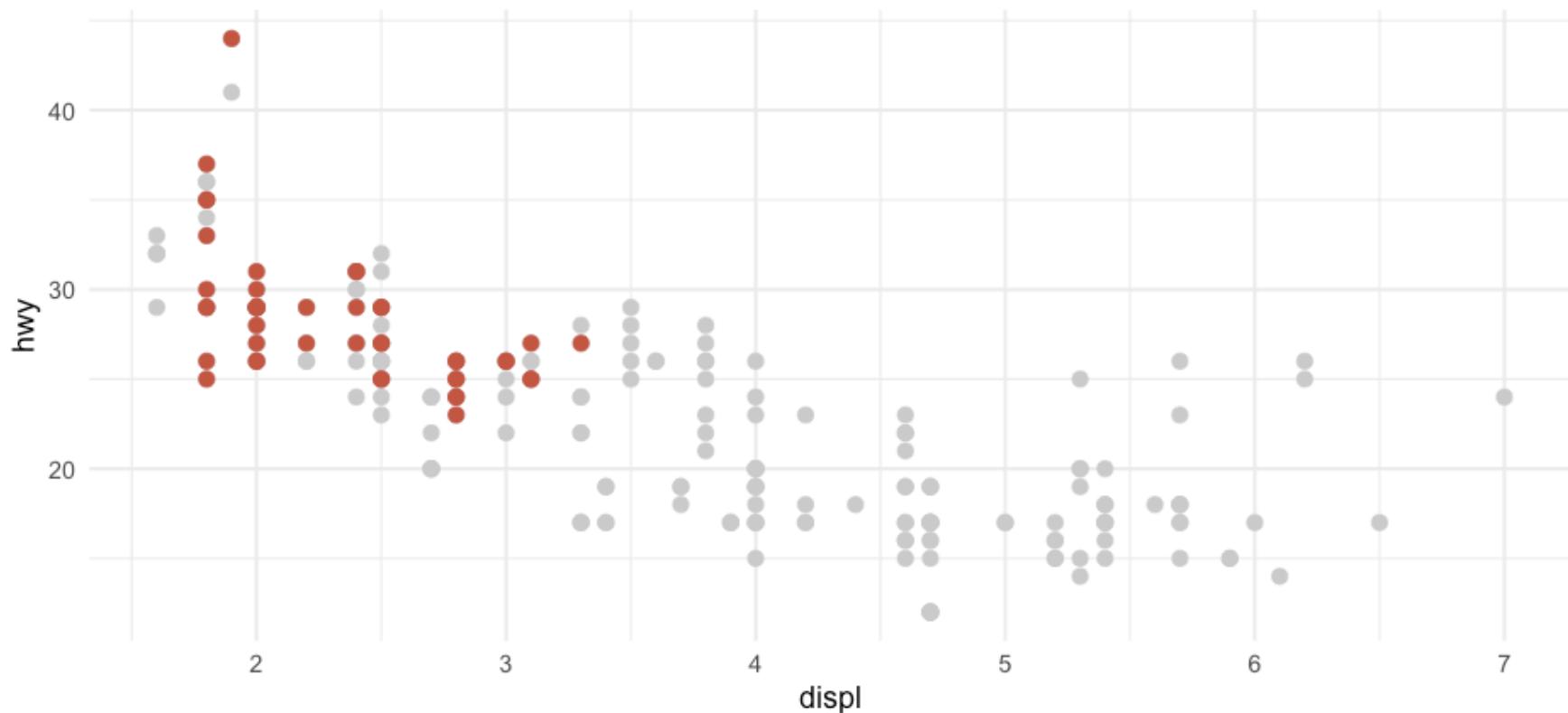
Basic scatterplot of weight to highway mpg

```
ggplot(mpg, aes(displ, hwy)) +  
  geom_point()
```



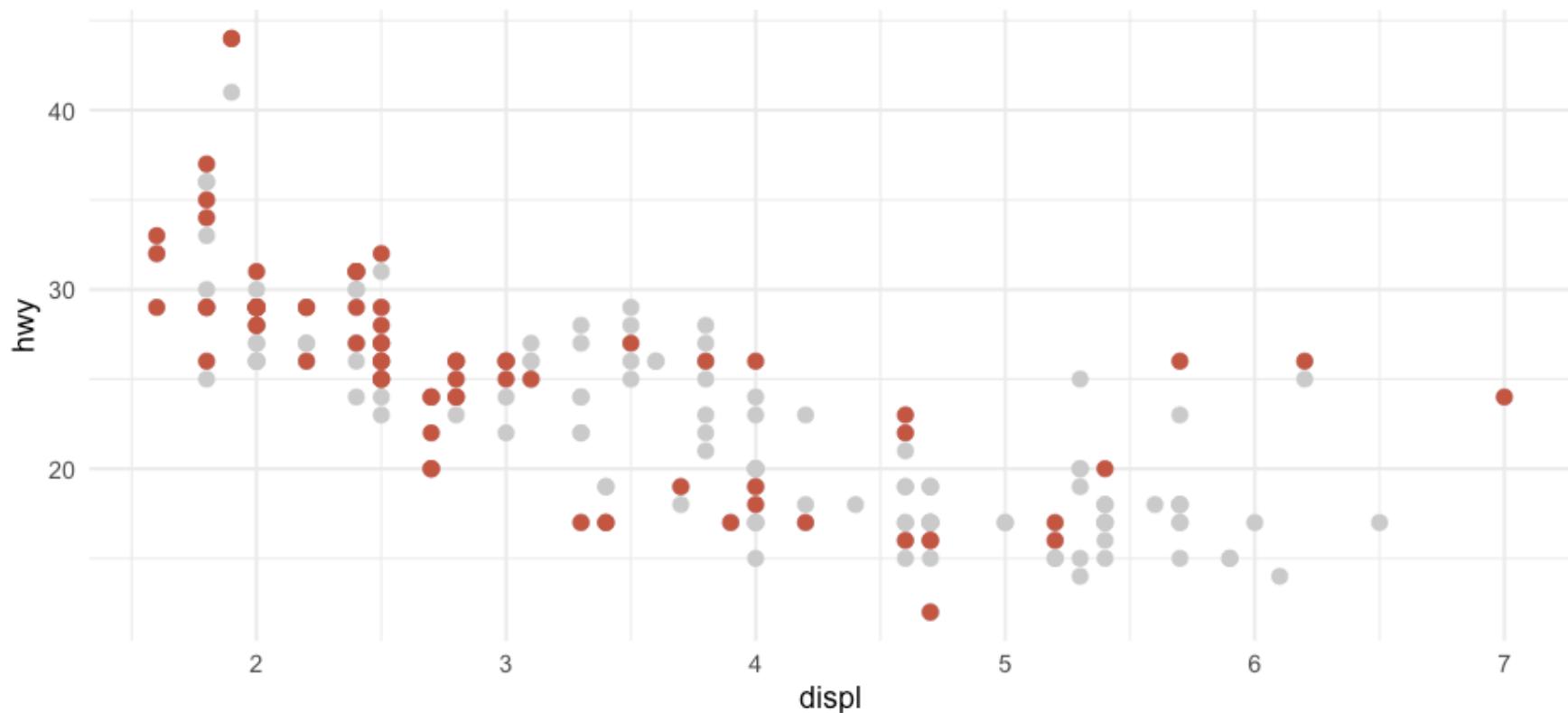
# Highlight compact cars

```
ggplot(mpg, aes(displ, hwy)) +  
  geom_point(color = "gray80") +  
  geom_point(data = filter(mpg, class == "compact"),  
             color = "#C55644")
```



# Highlight manual cars

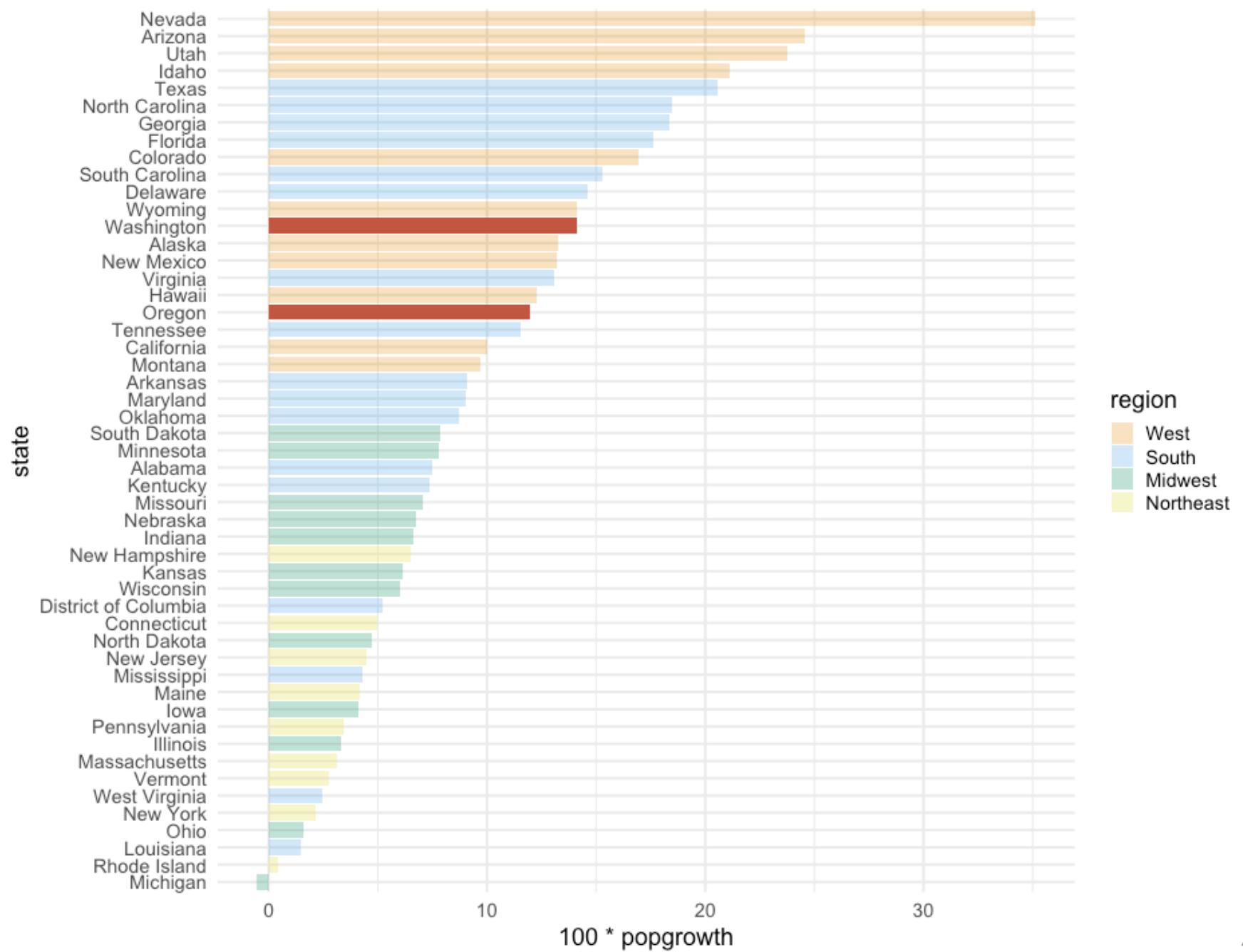
```
ggplot(mpg, aes(displ, hwy)) +  
  geom_point(color = "gray80") +  
  geom_point(data = filter(mpg, str_detect(trans, "manual")),  
             color = "#C55644")
```



# Back to our states plot

*Highlight Oregon and Washington*

```
ggplot(popgrowth_df,
       aes(x = state,
           y = 100*popgrowth)) +
  geom_col(aes(fill = region),
            alpha = 0.3) +
  geom_col(data = filter(popgrowth_df,
                         state == "Oregon" |
                         state == "Washington"),
            fill = "#C55644") +
  scale_fill_OkabeIto() +
  coord_flip()
```



# Color labels

```
states <- unique(popgrowth_df$state)

label_color <- ifelse(states == "Oregon" | states == "Washington",
                      "#C55644",
                      "gray30")

label_color
```

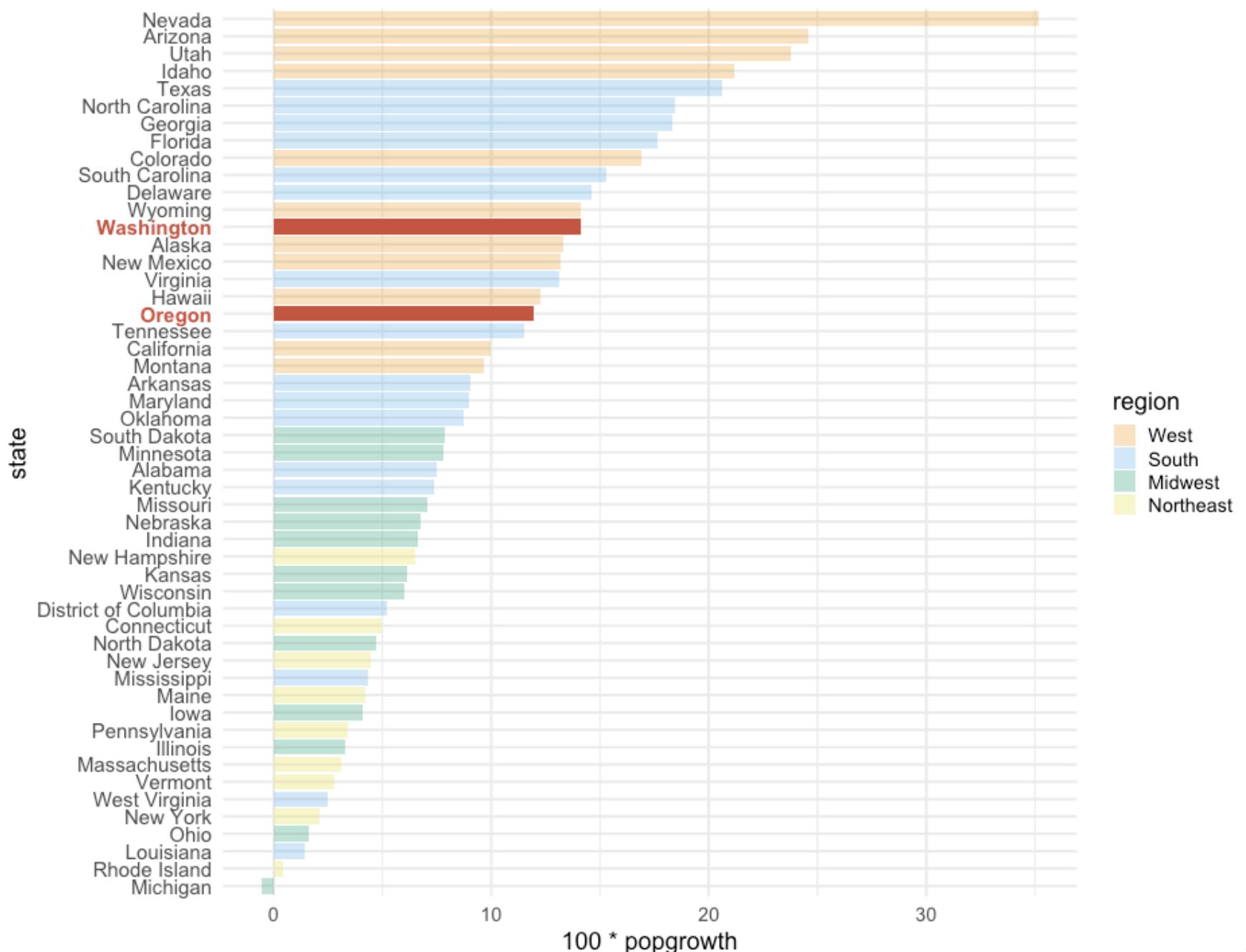
```
## [1] "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"
## [8] "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"
## [15] "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"
## [22] "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"
## [29] "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "#C55644" "gray30"
## [36] "gray30"  "gray30"  "gray30"  "#C55644" "gray30"  "gray30"  "gray30"  "gray30"
## [43] "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"  "gray30"
## [50] "gray30"  "gray30"
```

```
label_face <- ifelse(states == "Oregon" | states == "Washington",
                      "bold",
                      "plain")

label_face
```

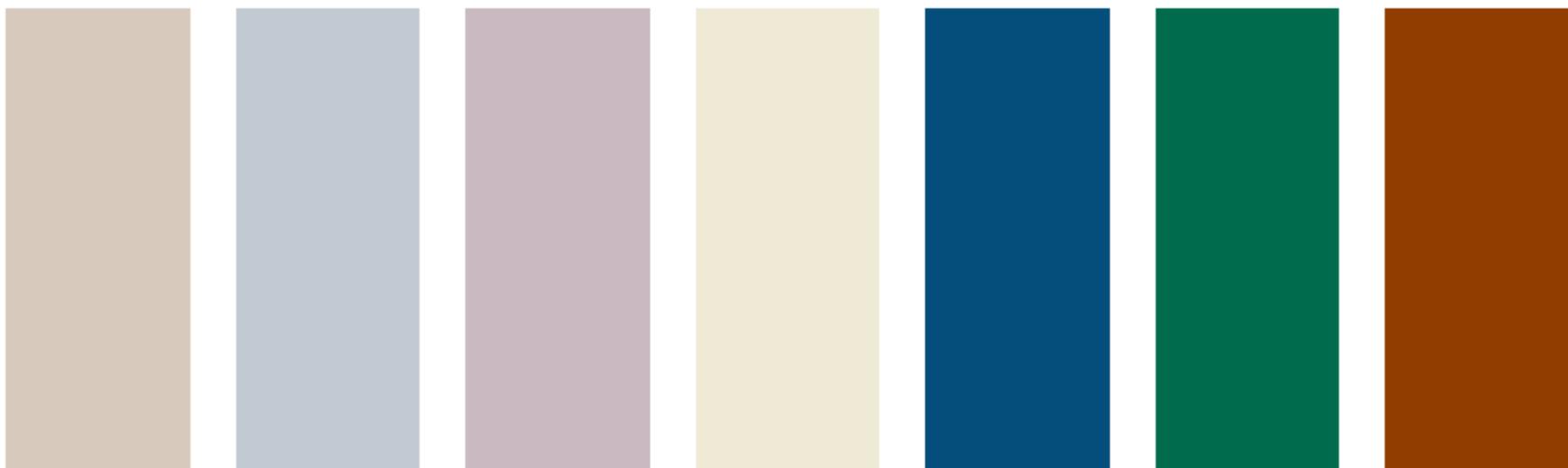
```
## [1] "plain" "plain" "plain" "plain" "plain" "plain" "plain" "plain"
## [9] "plain" "plain" "plain" "plain" "plain" "plain" "plain" "plain"
```

```
ggplot(popgrowth_df,
       aes(x = state,
           y = 100*popgrowth)) +
geom_col(aes(fill = region),
         alpha = 0.3) +
geom_col(data = filter(popgrowth_df,
                       state == "Oregon" |
                       state == "Washington"),
         fill = "#C55644") +
scale_fill_OkabeIto() +
coord_flip() +
theme(axis.text.y = element_text(color = label_color,
                                  face = label_face))
```

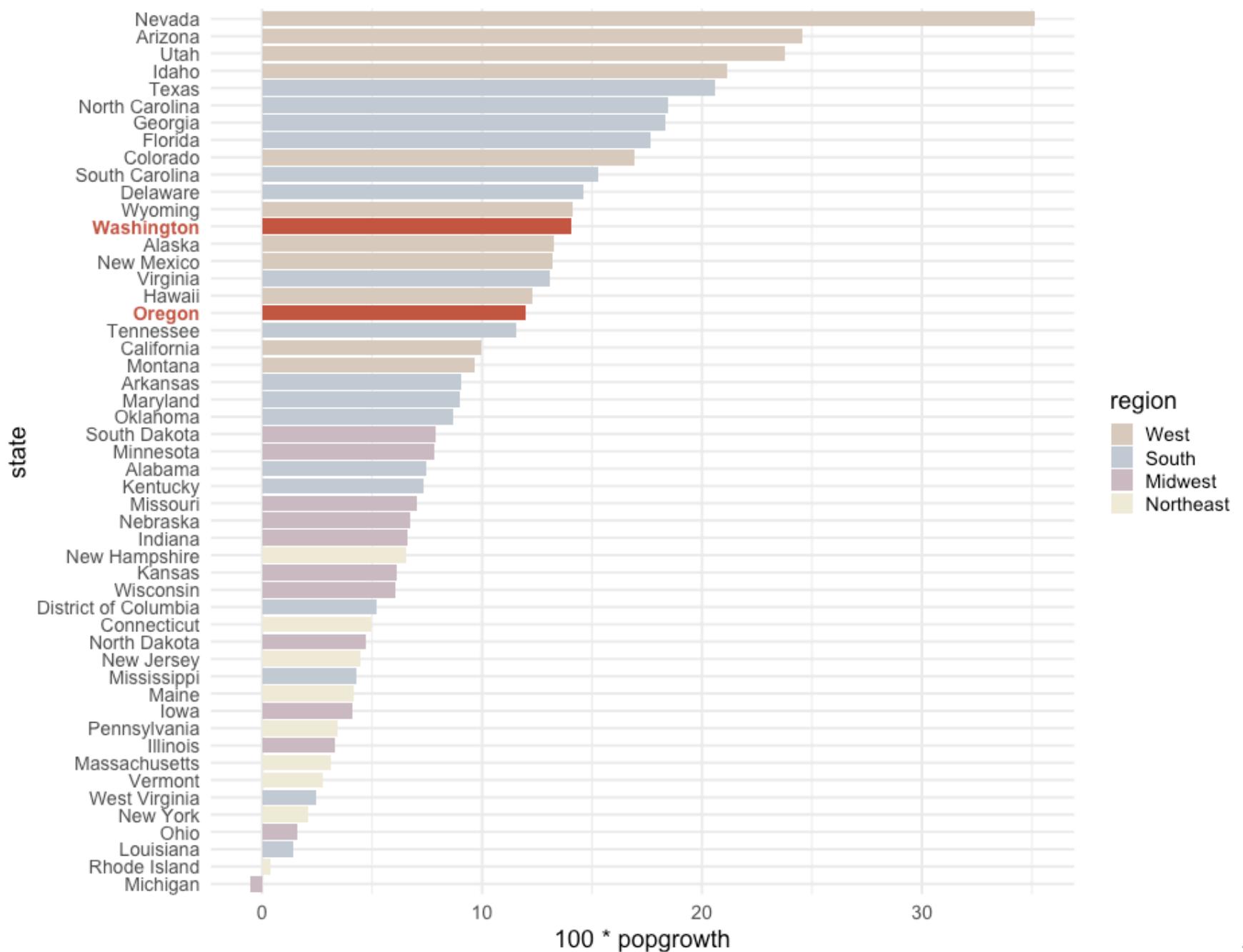


# Even better

```
accent_OkabeIto <- palette_OkabeIto[c(1, 2, 7, 4, 5, 3, 6)]  
accent_OkabeIto[1:4] <- desaturate(lighten(accent_OkabeIto[1:4], .4), .8)  
accent_OkabeIto[5:7] <- darken(accent_OkabeIto[5:7], .3)  
gg_color_swatches(7) +  
  scale_fill_manual(values = accent_OkabeIto)
```



```
ggplot(popgrowth_df,
       aes(x = state,
           y = 100*popgrowth)) +
  geom_col(aes(fill = region)) +
  geom_col(data = filter(popgrowth_df,
                         state == "Oregon" |
                         state == "Washington"),
            fill = "#C55644") +
  scale_fill_manual(values = accent_OkabeIto) +
  coord_flip() +
  theme(axis.text.y = element_text(color = label_color,
                                    face = label_face))
```



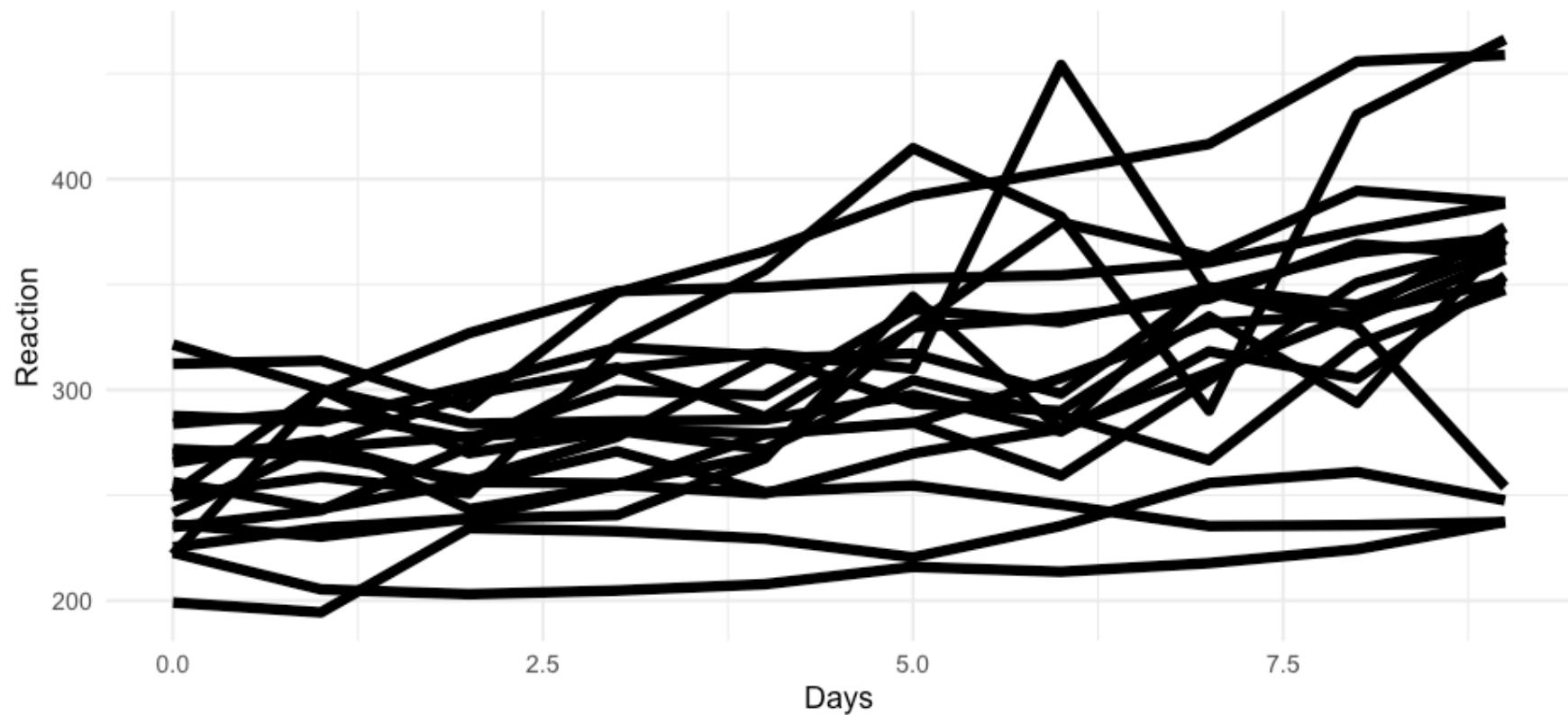
# Last example

```
data(sleepstudy, package = "lme4")
head(sleepstudy)
```

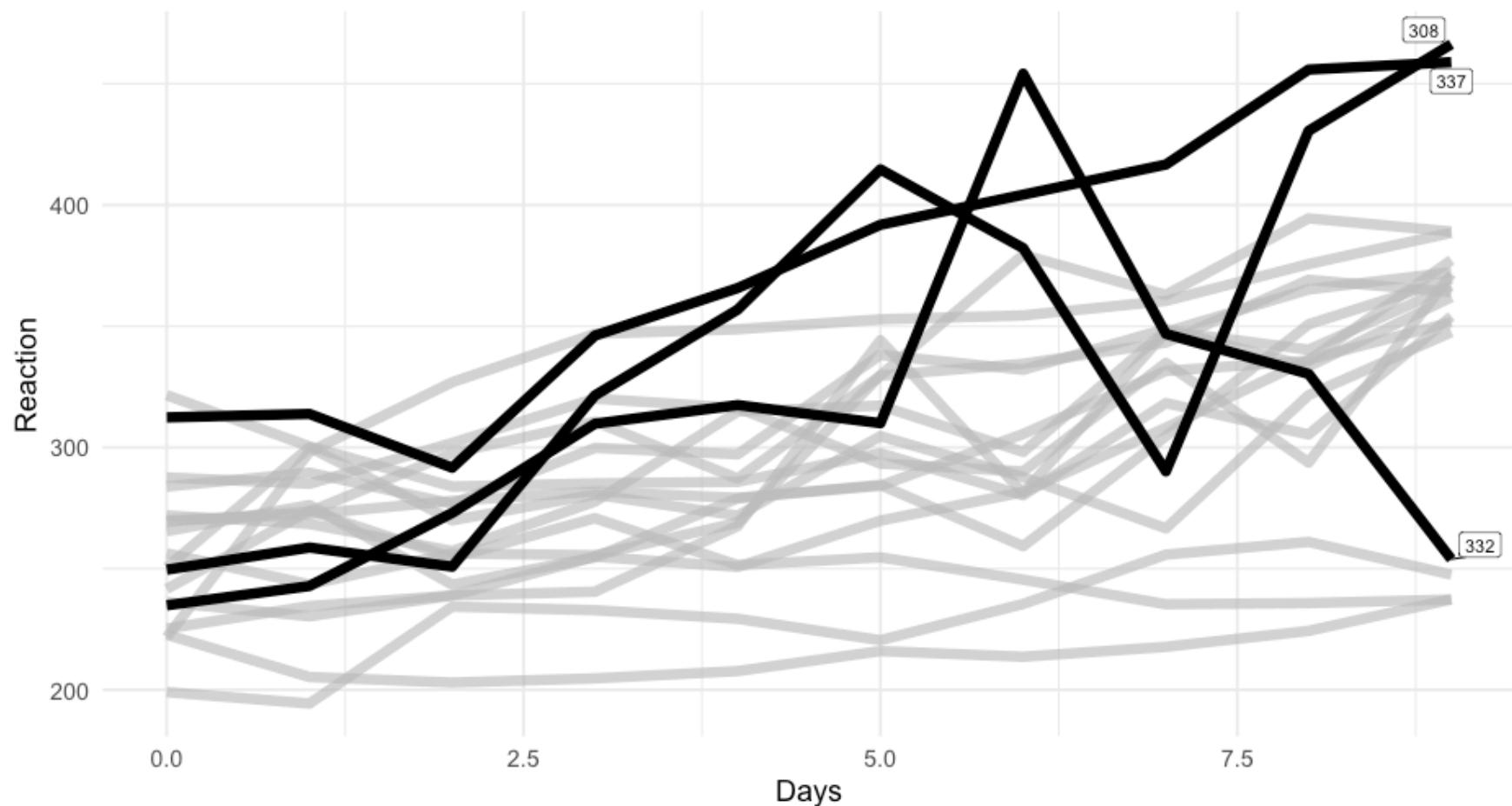
```
##   Reaction Days Subject
## 1      250     0    308
## 2      259     1    308
## 3      251     2    308
## 4      321     3    308
## 5      357     4    308
## 6      415     5    308
```

# Plot by subject

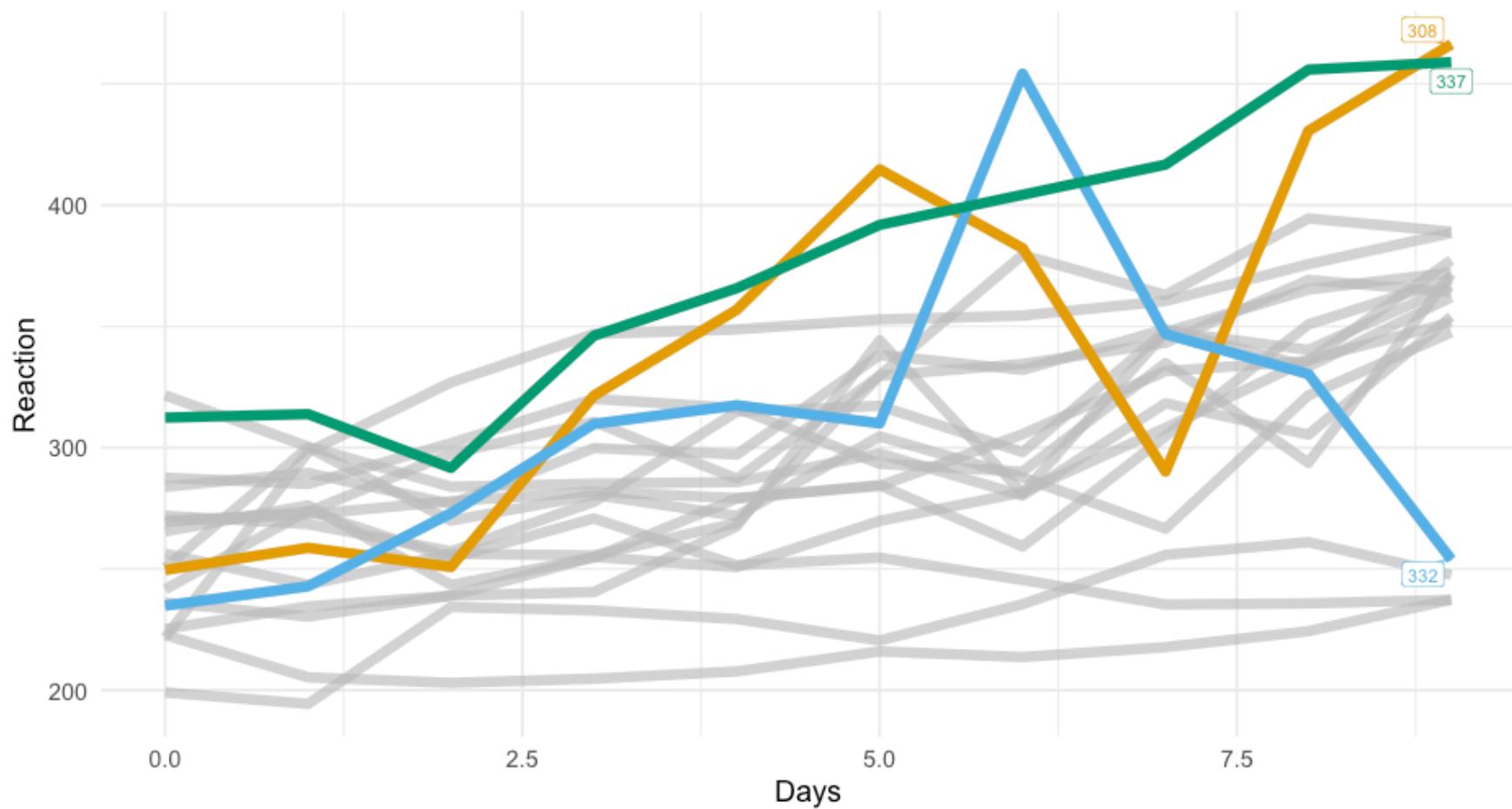
```
ggplot(sleepstudy, aes(Days, Reaction, group = Subject)) +  
  geom_line()
```



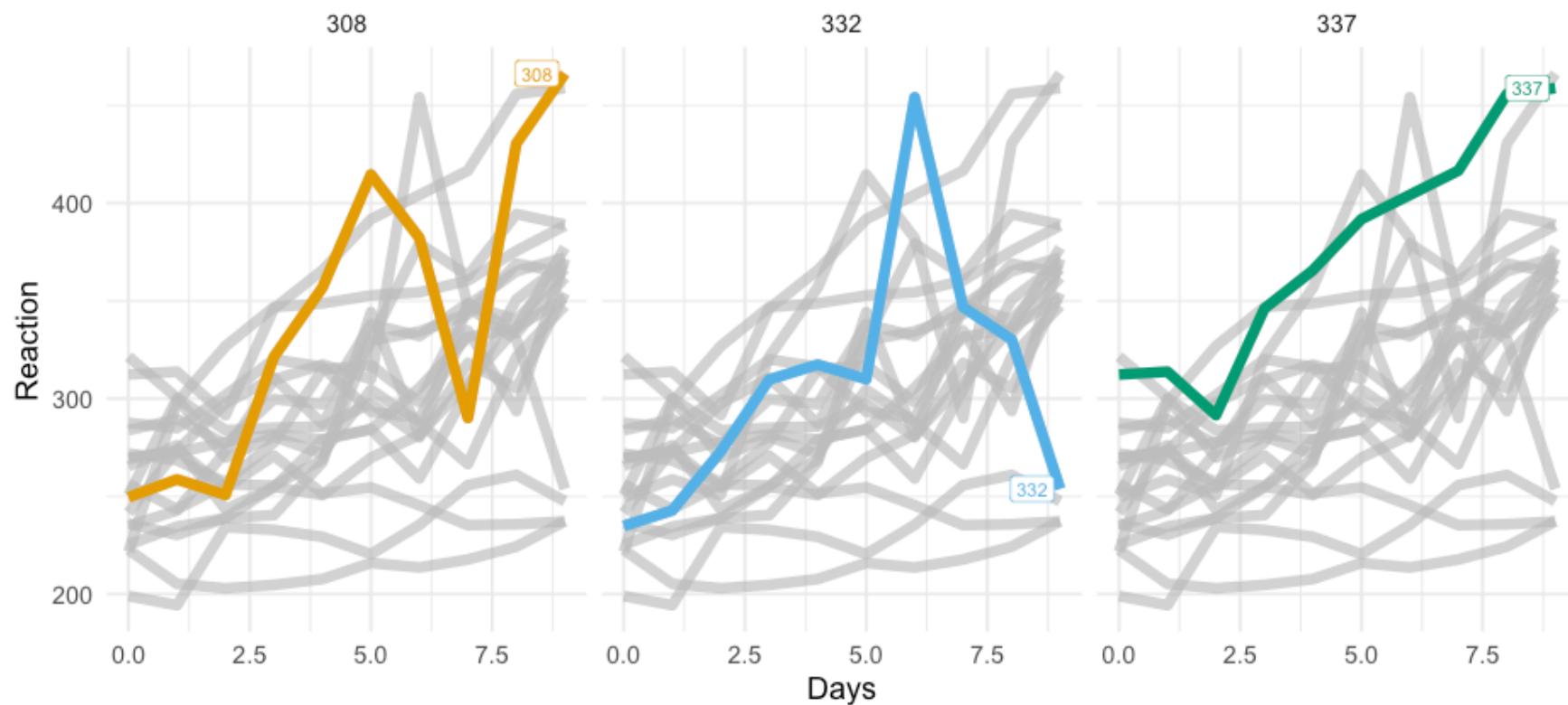
```
library(gghighlight)
ggplot(sleepstudy, aes(Days, Reaction, group = Subject)) +
  geom_line() +
  gghighlight(max(Reaction) > 400)
```



```
library(gghighlight)
ggplot(sleepstudy, aes(Days, Reaction, color = Subject)) +
  geom_line() +
  gghighlight(max(Reaction) > 400) +
  scale_color_OkabeIto()
```

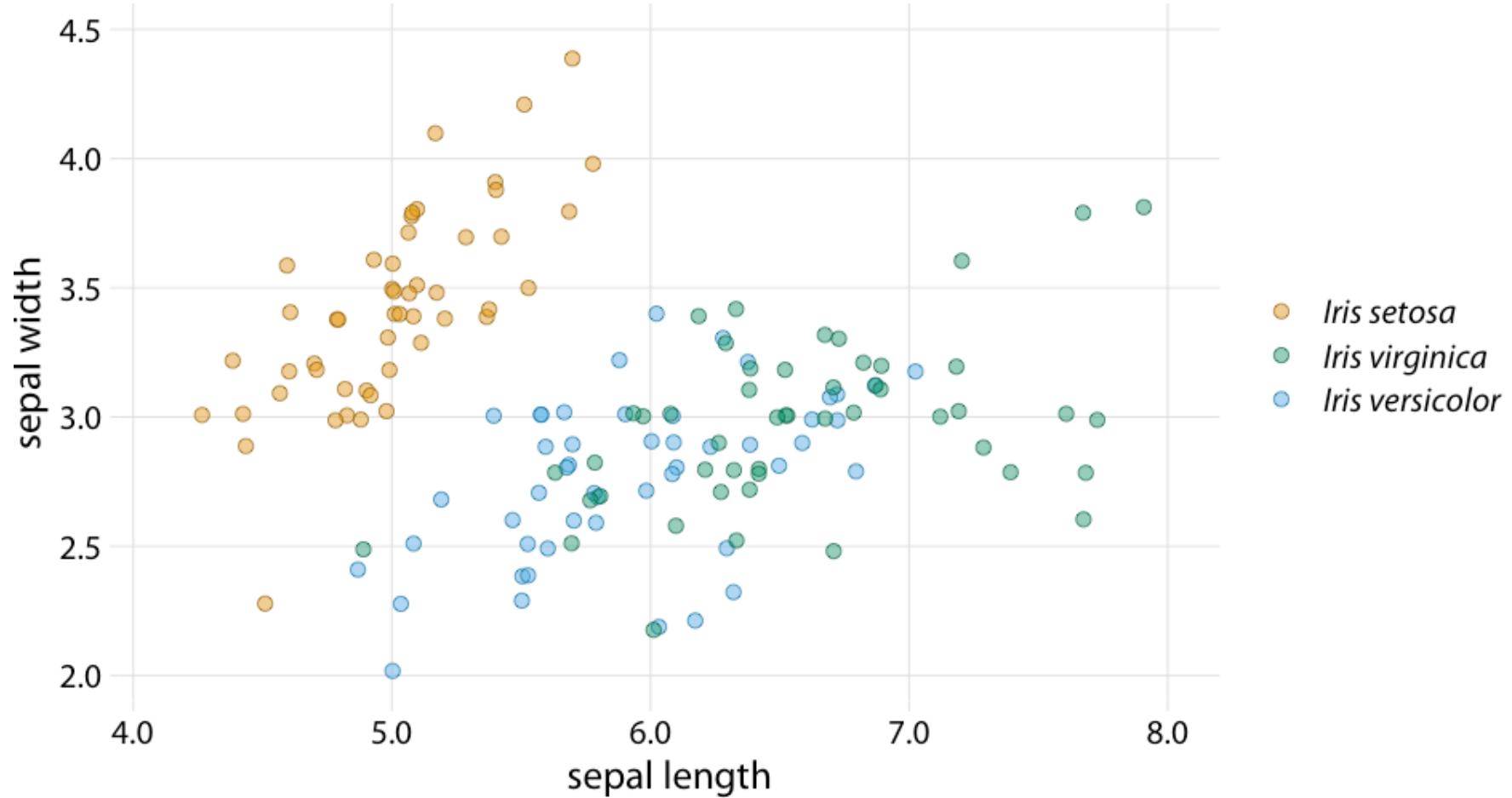


```
library(gghighlight)
ggplot(sleepstudy, aes(Days, Reaction, color = Subject)) +
  geom_line() +
  facet_wrap(~Subject) +
  gghighlight(max(Reaction) > 400) +
  scale_color_OkabeIto()
```

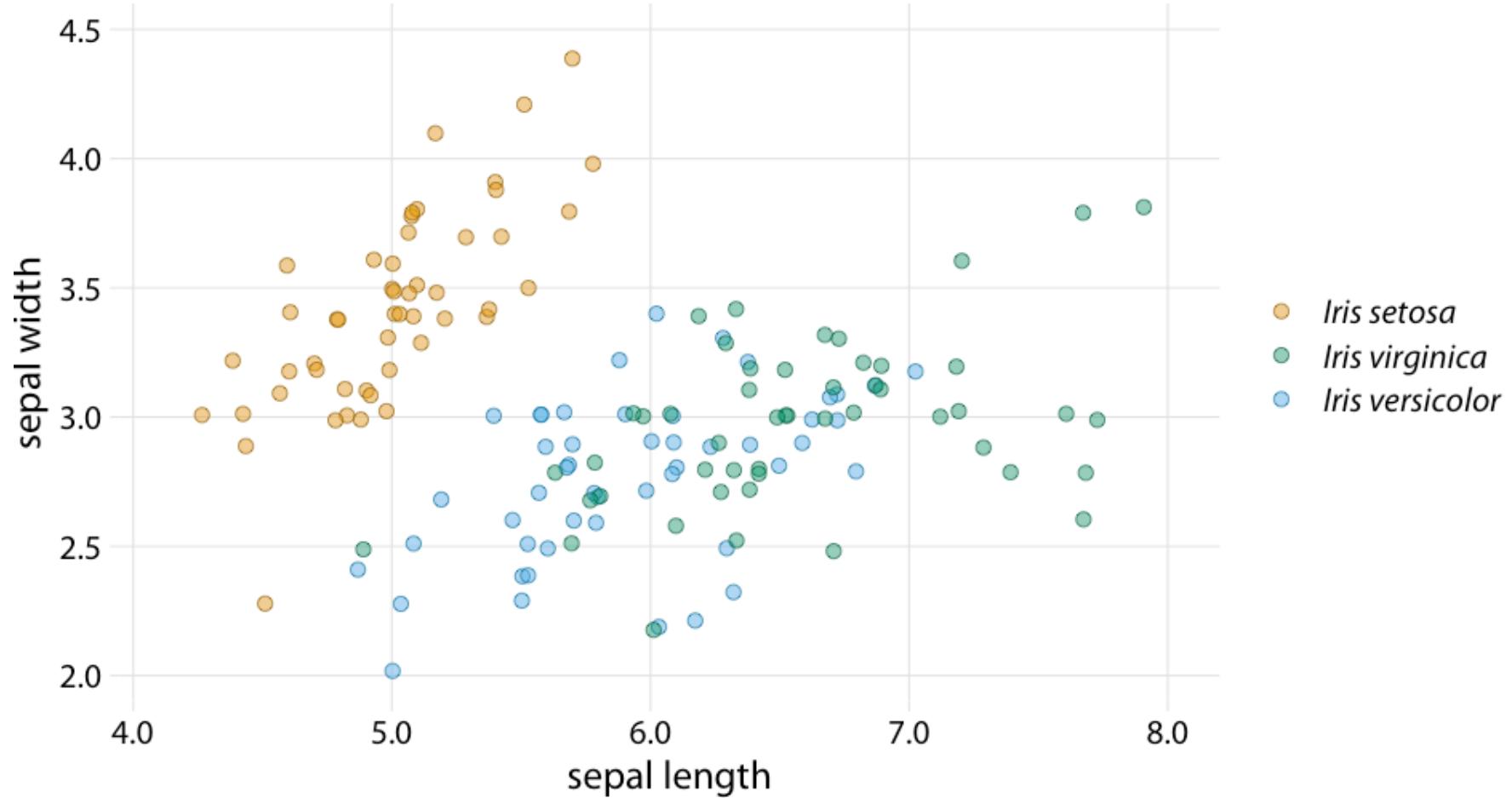


# A few other things to consider

# Double encodings



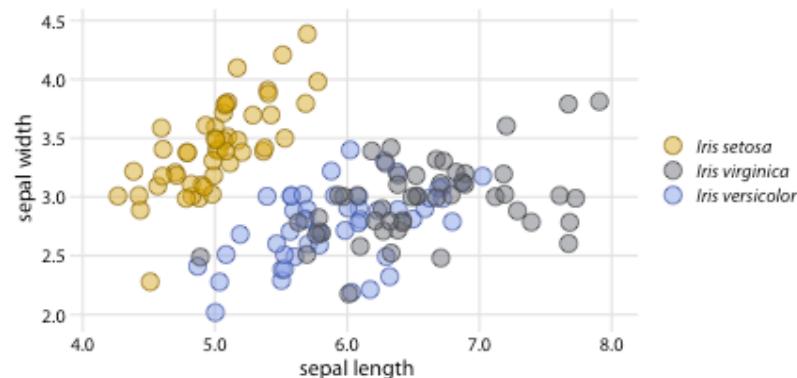
# Double encodings



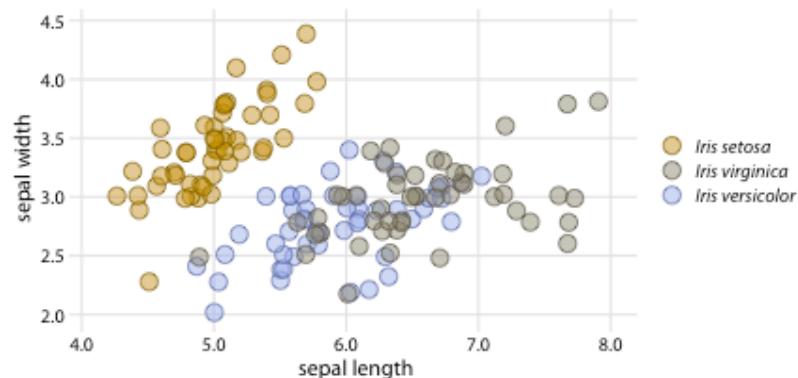
This plot is less than ideal. Why?

# Color blindness

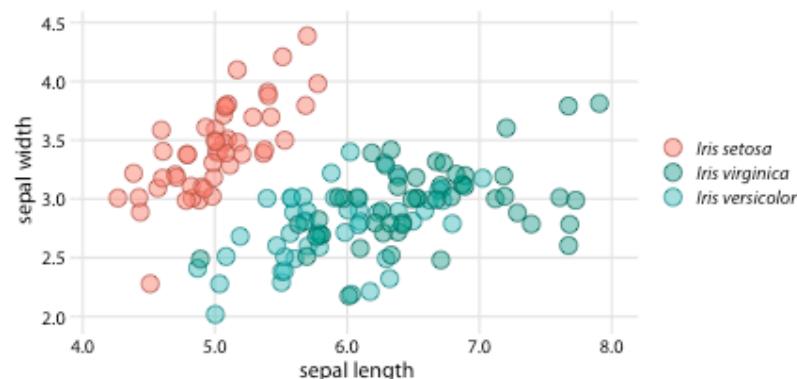
Deutanomaly



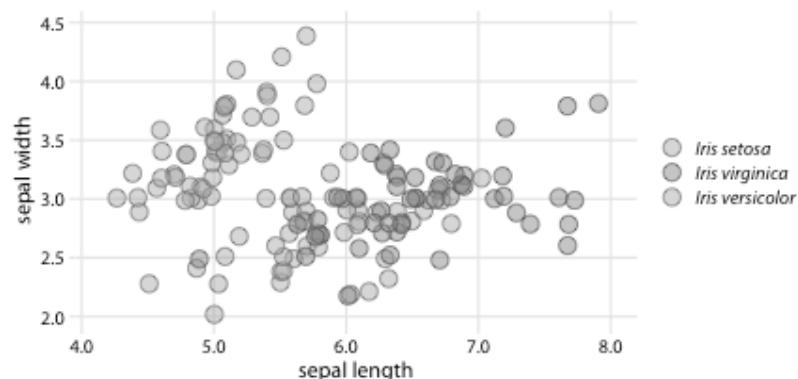
Protanomaly



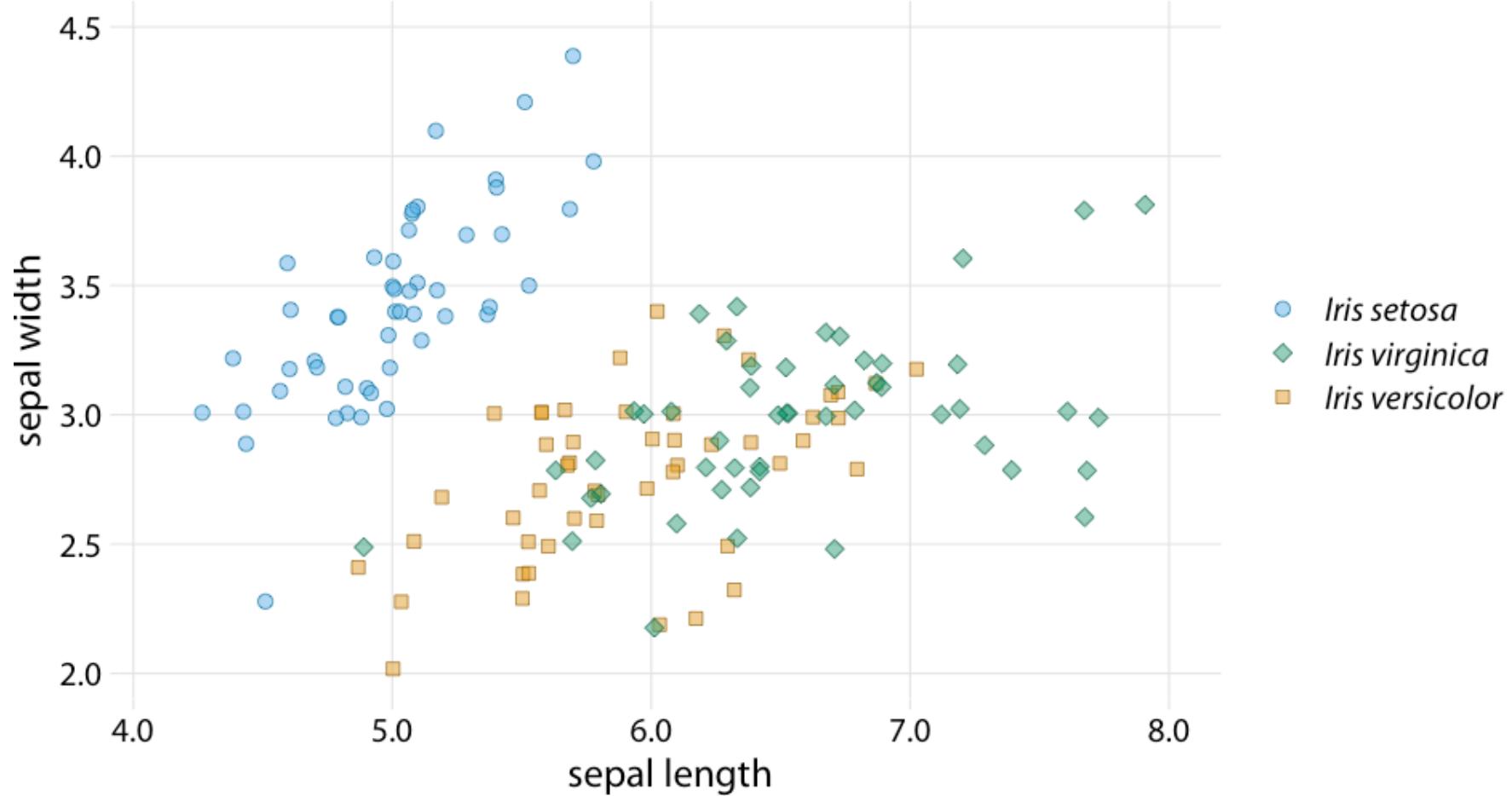
Tritanomaly



Desaturated

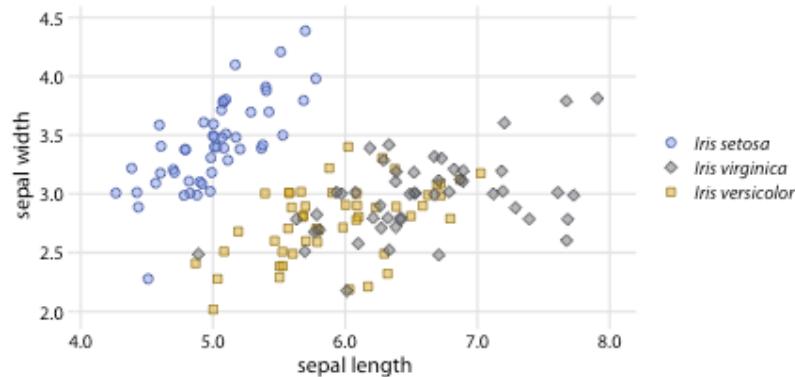


# Better version

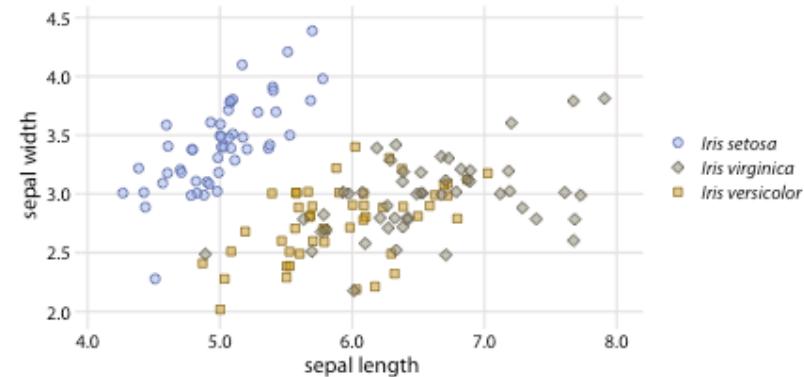


# Color blindness check

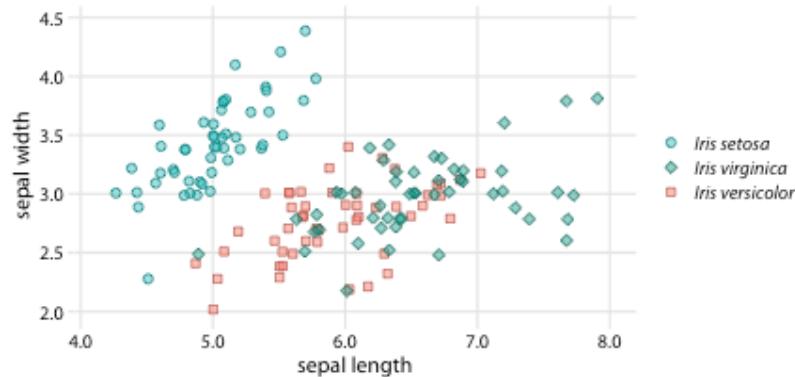
deuteranomaly



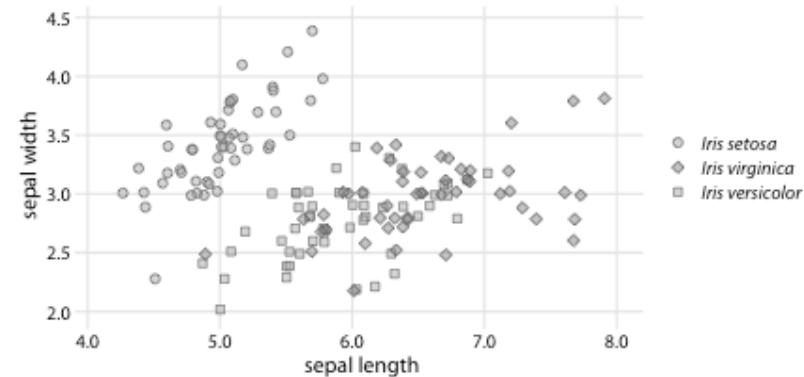
protanomaly



tritanomaly



desaturated

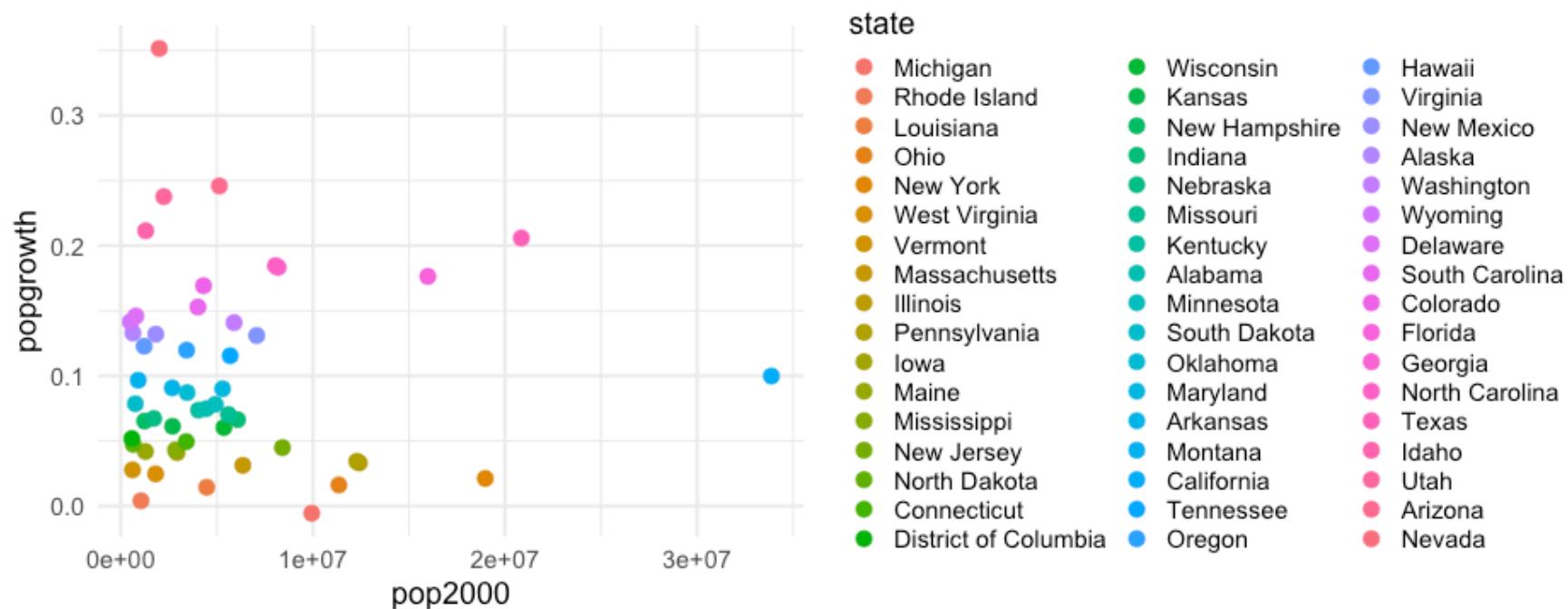


# Common problems with color

# Too many colors

More than 5-ish categories generally becomes too difficult to track

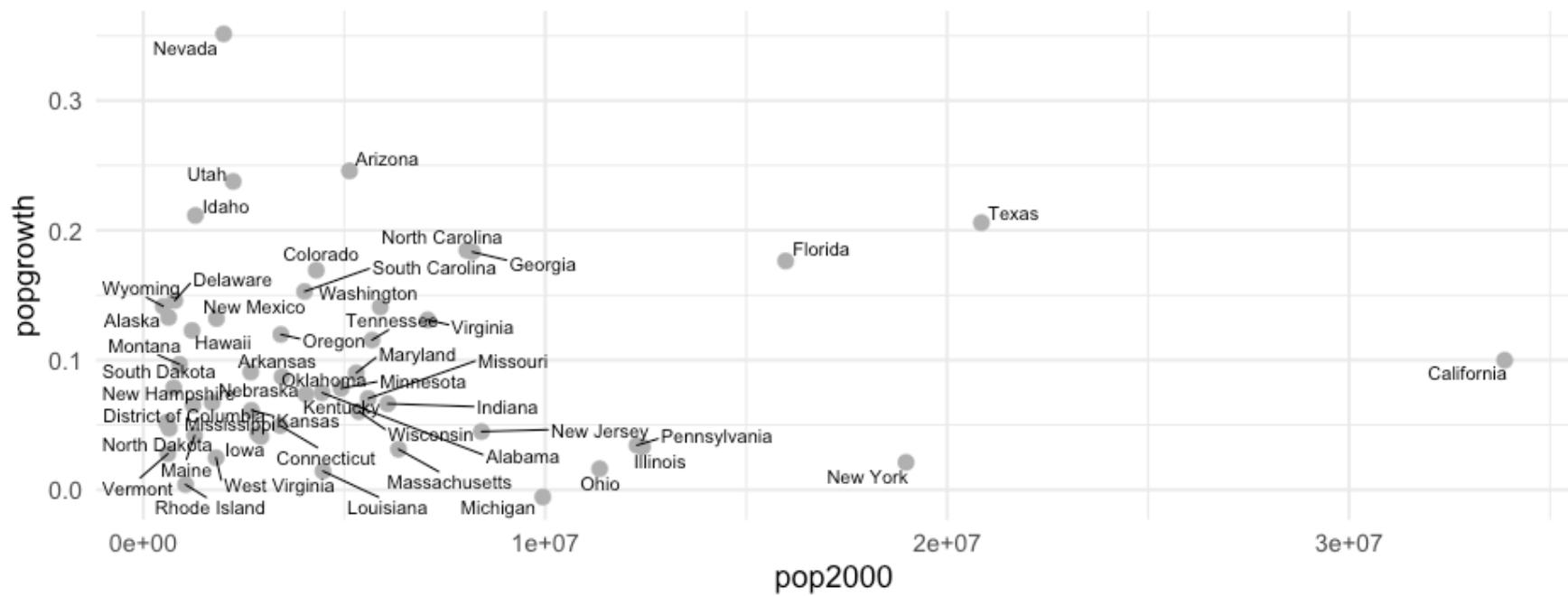
```
ggplot(popgrowth_df, aes(pop2000, popgrowth, color = state)) +  
  geom_point()
```



# Use labels

More than 5-ish categories generally becomes too difficult to track

```
library(ggrepel)  
  
ggplot(popgrowth_df, aes(pop2000, popgrowth)) +  
  geom_point(color = "gray70") +  
  geom_text_repel(aes(label = state))
```



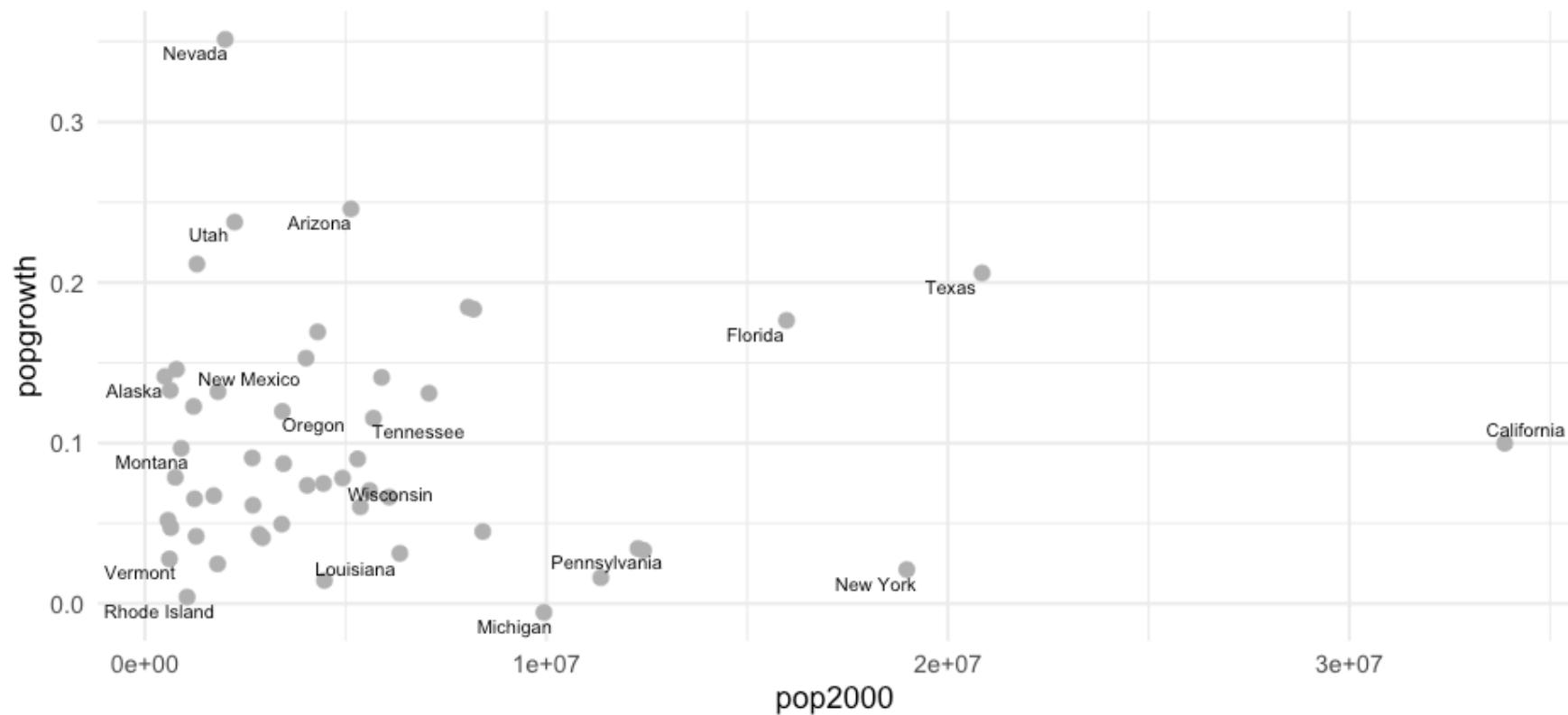
# Better

Get a subset

```
to_label <- c("Alaska", "Arizona", "California", "Florida", "Wisconsin",
             "Louisiana", "Nevada", "Michigan", "Montana", "New Mexico",
             "Pennsylvania", "New York", "Oregon", "Rhode Island",
             "Tennessee", "Texas", "Utah", "Vermont")
subset_states <- popgrowth_df %>%
  filter(state %in% to_label)
```

```
library(ggrepel)

ggplot(popgrowth_df, aes(pop2000, popgrowth)) +
  geom_point(color = "gray70") +
  geom_text_repel(aes(label = state), data = subset_states)
```



(still lots more cleaning up we could do here...)

# Rainbow palette

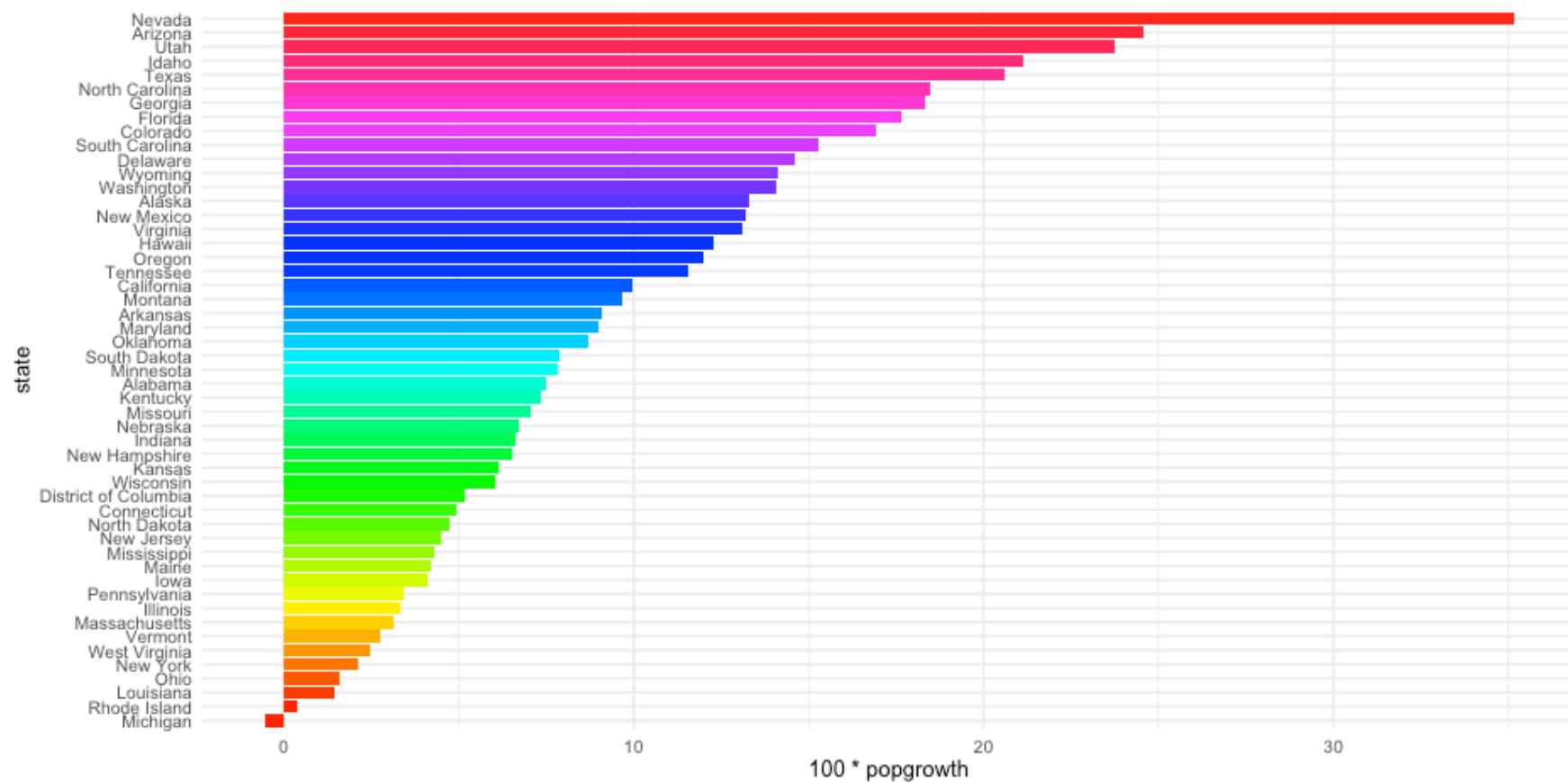
```
rainbow(3)
```

```
## [1] "#FF0000FF" "#00FF00FF" "#0000FFFF"
```

```
rainbow(7)
```

```
## [1] "#FF0000FF" "#FFDB00FF" "#49FF00FF" "#00FF92FF" "#0092FFFF" "#4900FFFF"  
## [7] "#FF00DBFF"
```

```
ggplot(popgrowth_df,  
       aes(x = state,  
            y = 100*popgrowth)) +  
  geom_col(aes(fill = state)) +  
  scale_fill_manual(values = rainbow(51)) +  
  coord_flip() +  
  guides(fill = "none")
```

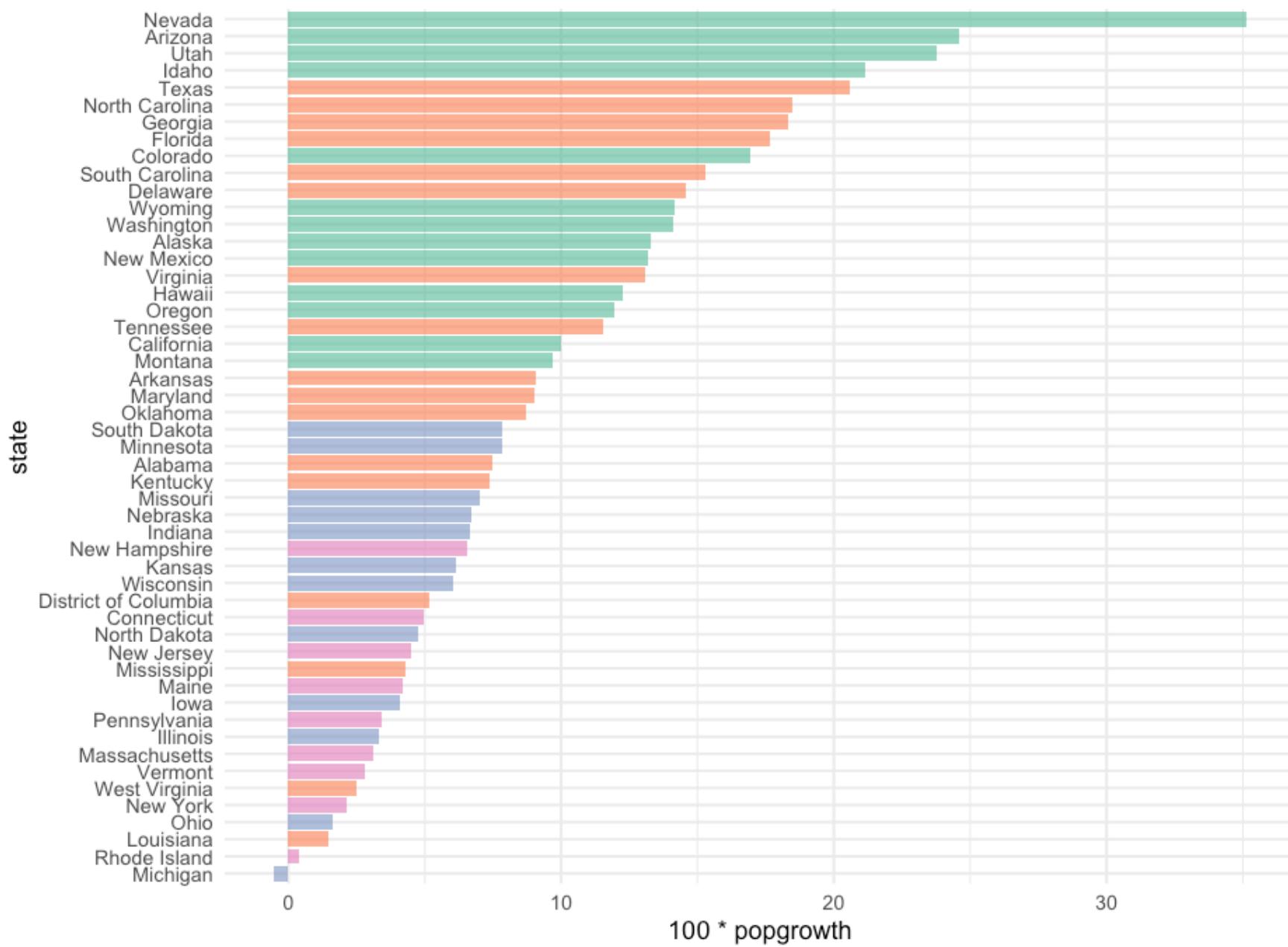


# Last few note on palettes

- Do some research, find what you like and what tends to work well
- Check for colorblindness
- Look into <http://colorbrewer2.org/>
  - `scale_color_brewer()` and `scale_fill_brewer()` ship with `ggplot2`

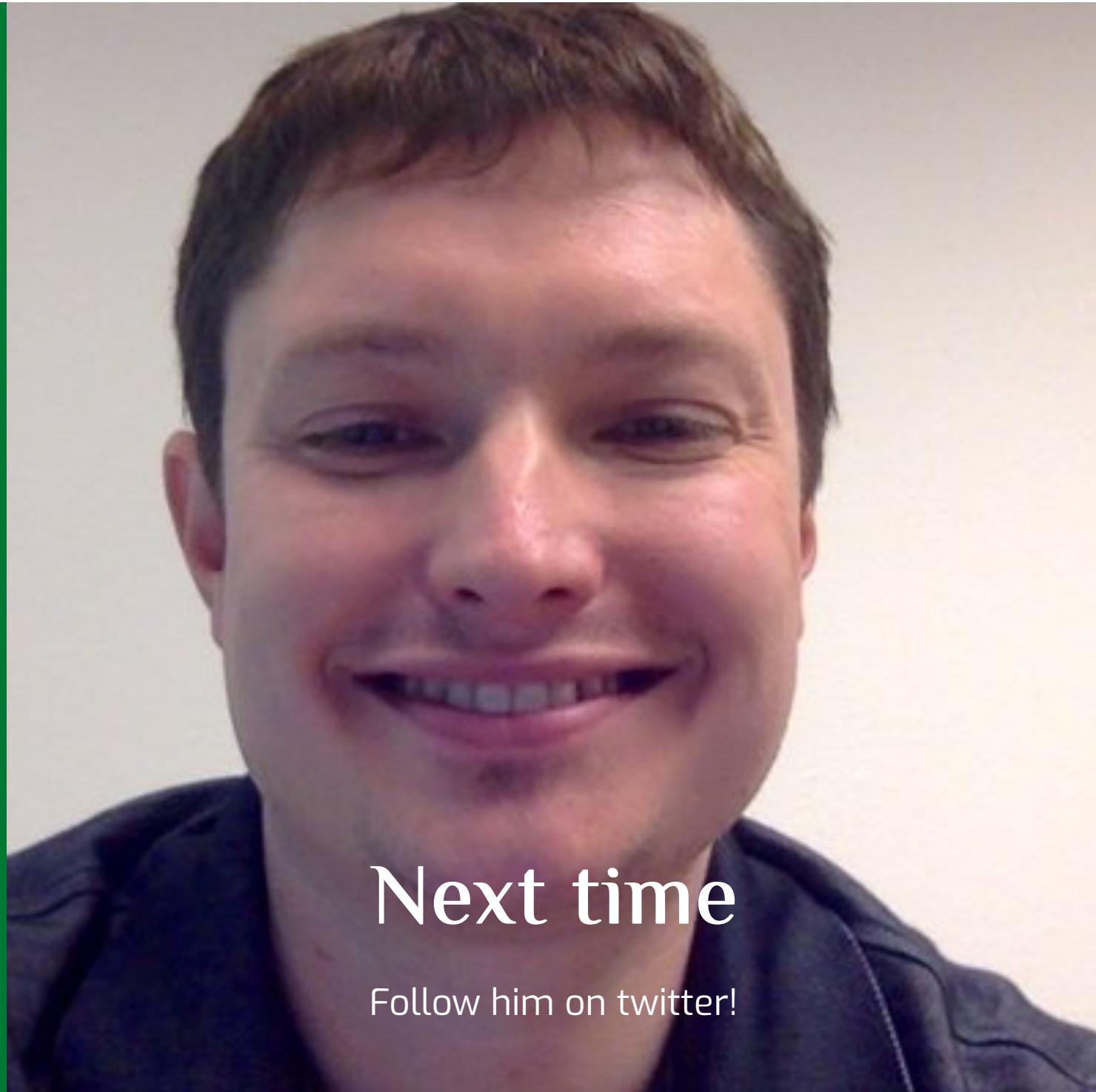
e.g.,

```
ggplot(popgrowth_df,  
       aes(x = state,  
            y = 100*popgrowth)) +  
  geom_col(aes(fill = region),  
           alpha = 0.7) +  
  scale_fill_brewer(palette = "Set2") +  
  coord_flip() +  
  guides(fill = "none")
```



# Paleteer package





Next time

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