



# AE 15: Optimizing color spaces

## Suggested answers

APPLICATION EXERCISE

ANSWERS

MODIFIED

March 20, 2025

```
library(tidyverse)
library(colorspace) # for improved color palettes
library(scales) # for improved labels
library(ggthemes) # for scale_color_colorblind()

theme_set(theme_classic(base_size = 12))
```

## Import birth data

The Social Security Administration keeps detailed records on births and deaths in the United States. For our analysis, we will use a dataset of the number of births daily in the United States from 1994-2014.<sup>1</sup>

```
births <- read_rds("data/births.Rds")
births
```

# A tibble: 7,670 × 7

	year	month	date_of_month	day_of_week	births	date_of_month_catego... <sup>1</sup>	weekend
	<dbl>	<ord>	<dbl>	<ord>	<dbl>	<fct>	<lgl>
1	1994	January	1	Saturday	8096	1	TRUE
2	1994	January	2	Sunday	7772	2	TRUE
3	1994	January	3	Monday	10142	3	FALSE
4	1994	January	4	Tuesday	11248	4	FALSE
5	1994	January	5	Wednesday	11053	5	FALSE
6	1994	January	6	Thursday	11406	6	FALSE
7	1994	January	7	Friday	11251	7	FALSE
8	1994	January	8	Saturday	8653	8	TRUE
9	1994	January	9	Sunday	7910	9	TRUE
10	1994	January	10	Monday	10498	10	FALSE

# i 7,660 more rows

# i abbreviated name: <sup>1</sup>date\_of\_month\_categorical

# The Friday the 13th effect

Friday the 13th is considered an unlucky day in Western superstition. Let's see if fewer babies are born on the 13th of each month if it falls on a Friday compared to another week day. Specifically, we will compare the average number of births on the 13th of the month to the average number of births on the 6th and 20th of the month.

**Your turn:** Visualize the results using a bar chart. Emphasize the difference on Fridays compared to other weekdays.<sup>2</sup>

```
friday_13_births <- births |>
  # only look at births on the 6, 13, and 20th
  filter(date_of_month %in% c(6, 13, 20)) |>
  # distinguish 6/20 from 13
  mutate(not_13 = date_of_month == 13) |>
  # calculate average number of births for each week day and whether or not it was the 13th
  summarize(
    avg_births = mean(births),
    .by = c(day_of_week, not_13)
  ) |>
  # calculate the difference in percentage
  pivot_wider(
    names_from = not_13,
    values_from = avg_births
  ) |>
  mutate(pct_diff = (`TRUE` - `FALSE`) / `FALSE`) |>
  arrange(day_of_week)

# highlight one bar in orange
ggplot(
  data = friday_13_births,
  mapping = aes(
    x = day_of_week,
    y = pct_diff,
    fill = day_of_week == "Friday"
  )
) +
  geom_col() +
  scale_y_continuous(labels = label_percent()) +
  scale_fill_manual(values = c("grey50", "orange"), guide = "none") +
  labs(
    x = NULL,
    y = NULL,
    title = "The Friday the 13th effect",
```

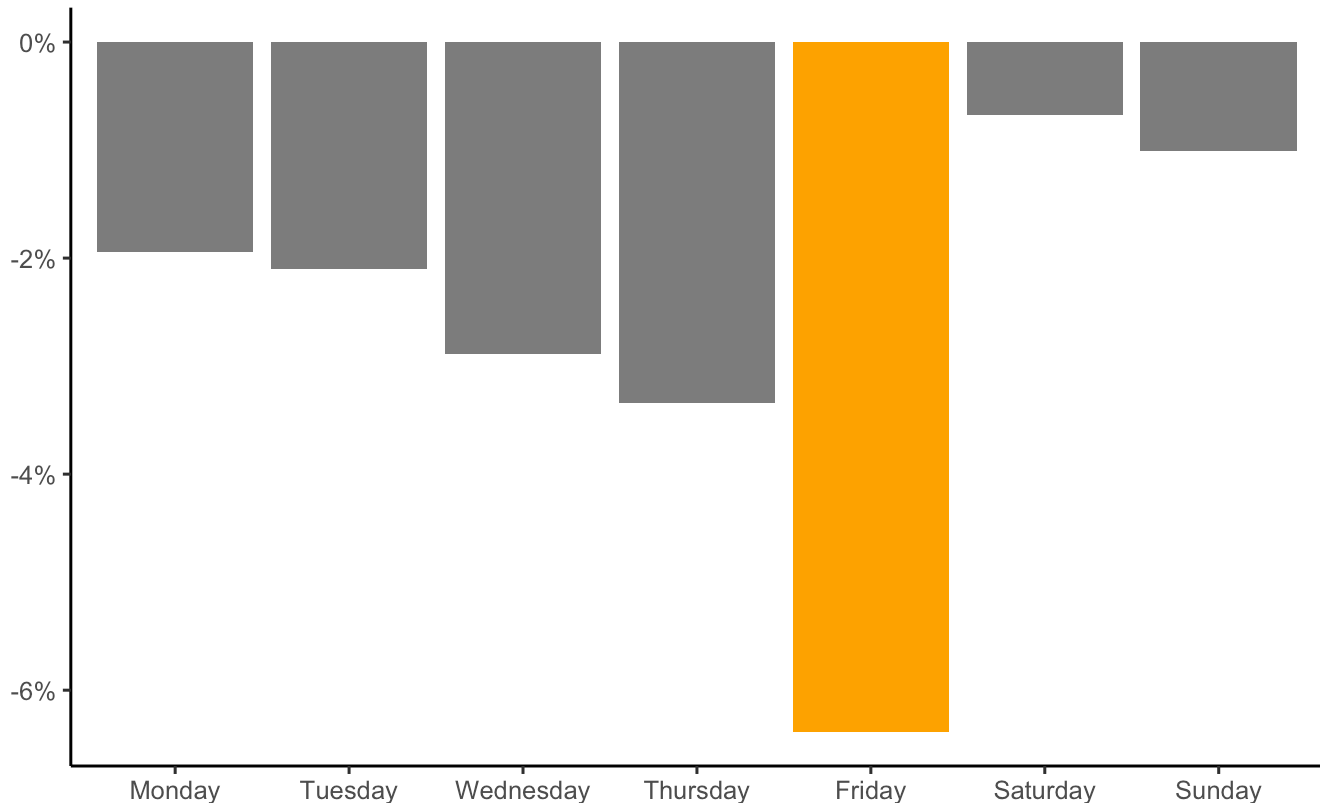
```

subtitle = "Difference in the share of U.S. births on the 13th of each month\nfrom the
          average of births on the 6th and the 20th, 1994-2014"
)

```

## The Friday the 13th effect

Difference in the share of U.S. births on the 13th of each month  
from the average of births on the 6th and the 20th, 1994-2014



```

# use transparency instead
ggplot(
  data = friday_13_births,
  mapping = aes(
    x = day_of_week,
    y = pct_diff,
    alpha = day_of_week == "Friday"
  )
) +
  geom_col(fill = "orange") +
  scale_y_continuous(labels = label_percent()) +
  scale_alpha_manual(values = c(0.4, 1), guide = "none") +
  labs(
    x = NULL,
    y = NULL,
    title = "The Friday the 13th effect",

```

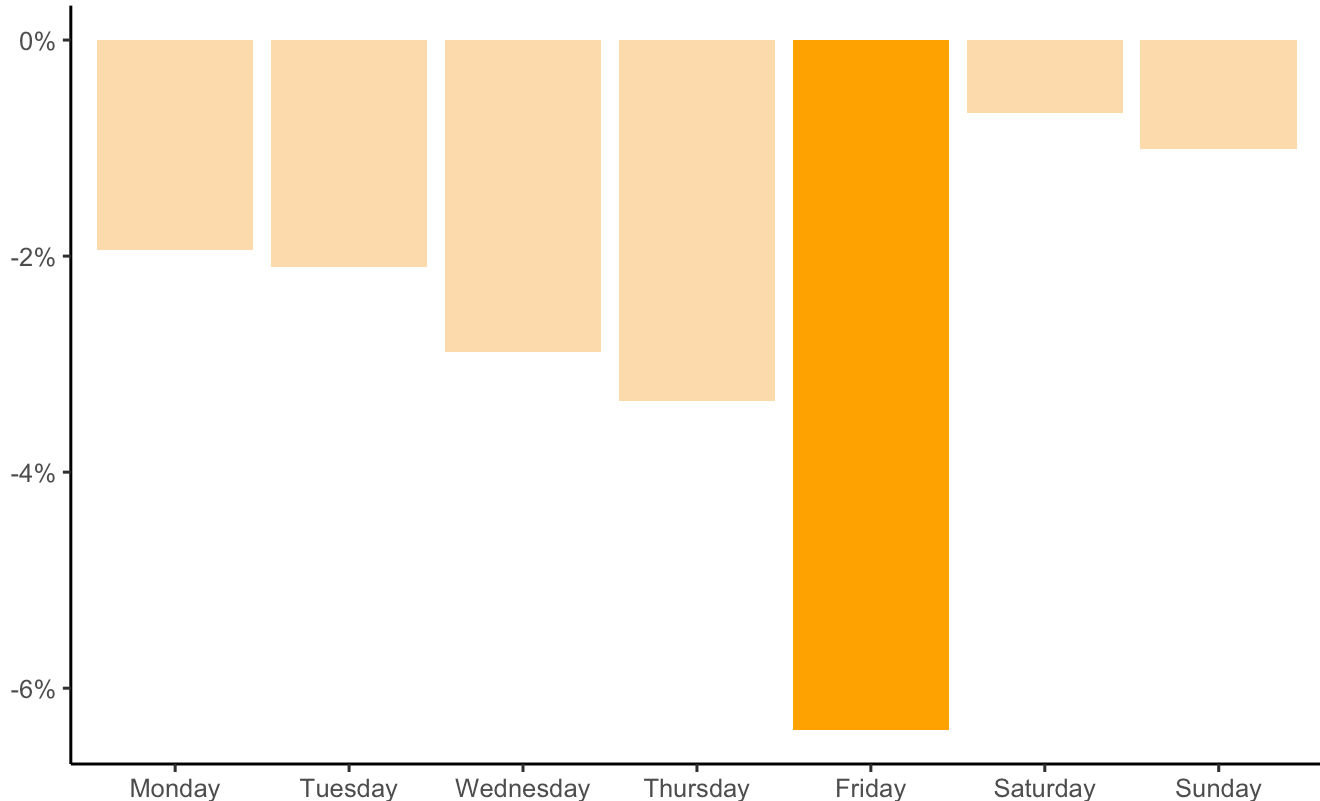
```

subtitle = "Difference in the share of U.S. births on the 13th of each month\nfrom the
          average of births on the 6th and the 20th, 1994-2014"
)

```

## The Friday the 13th effect

Difference in the share of U.S. births on the 13th of each month from the average of births on the 6th and the 20th, 1994-2014



## Create a heatmap showing average number of births by day of year

Let's explore the relative popularity of each calendar day for births. We will create a heatmap showing the relative ratio of births for each day of the year compared to the annual average.

```

avg_births_month_day <- births |>
  group_by(month, date_of_month_categorical) |>
  summarize(avg_births = mean(births), .groups = "drop") |>
  mutate(avg_births_ratio = avg_births / mean(births$births))
avg_births_month_day

```

# A tibble: 366 × 4

month	date_of_month_categorical	avg_births	avg_births_ratio
<ord>	<fct>	<dbl>	<dbl>

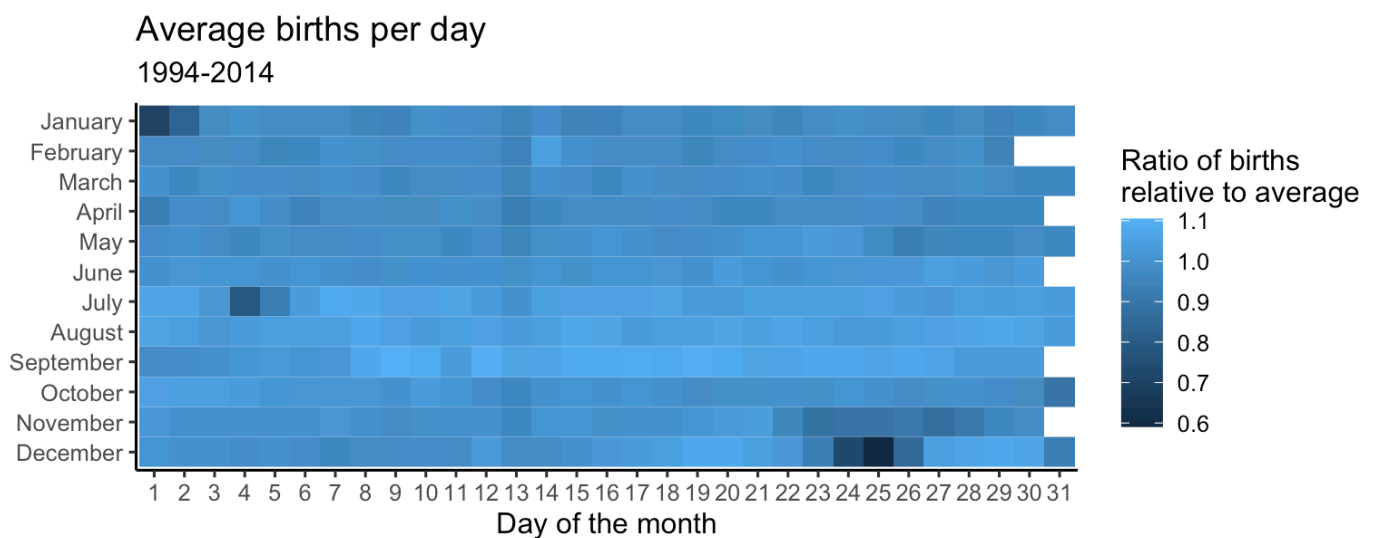
1 January 1	7827.	0.700
2 January 2	9356.	0.837
3 January 3	10869.	0.973
4 January 4	11064.	0.990
5 January 5	10992.	0.984
6 January 6	10942.	0.979
7 January 7	10963.	0.981
8 January 8	10656.	0.954
9 January 9	10672.	0.955
10 January 10	11072.	0.991

# i 356 more rows

```

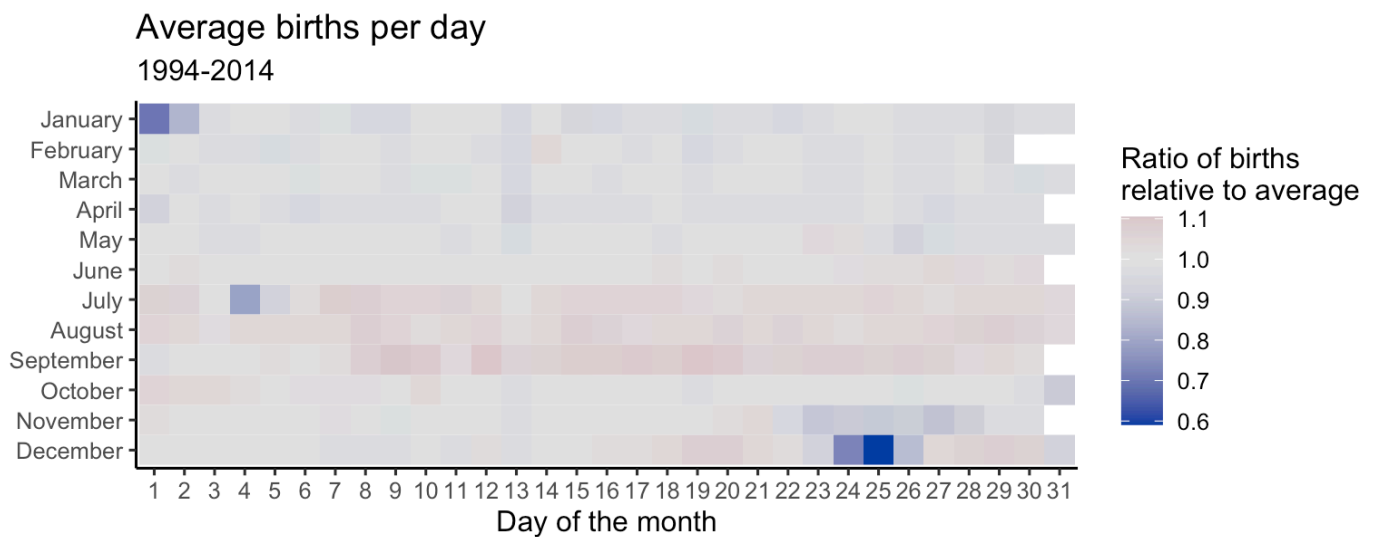
birth_days_plot <- ggplot(
  data = avg_births_month_day,
  # By default, the y-axis will have December at the top, so use fct_rev() to reverse it
  mapping = aes(x = date_of_month_categorical, y = fct_rev(month), fill = avg_births_ratio)
) +
  geom_tile() +
  # Add nice labels
  labs(
    x = "Day of the month", y = NULL,
    title = "Average births per day",
    subtitle = "1994-2014",
    fill = "Ratio of births\nrelative to average"
  ) +
  # Force all the tiles to have equal widths and heights
  coord_equal()
birth_days_plot

```

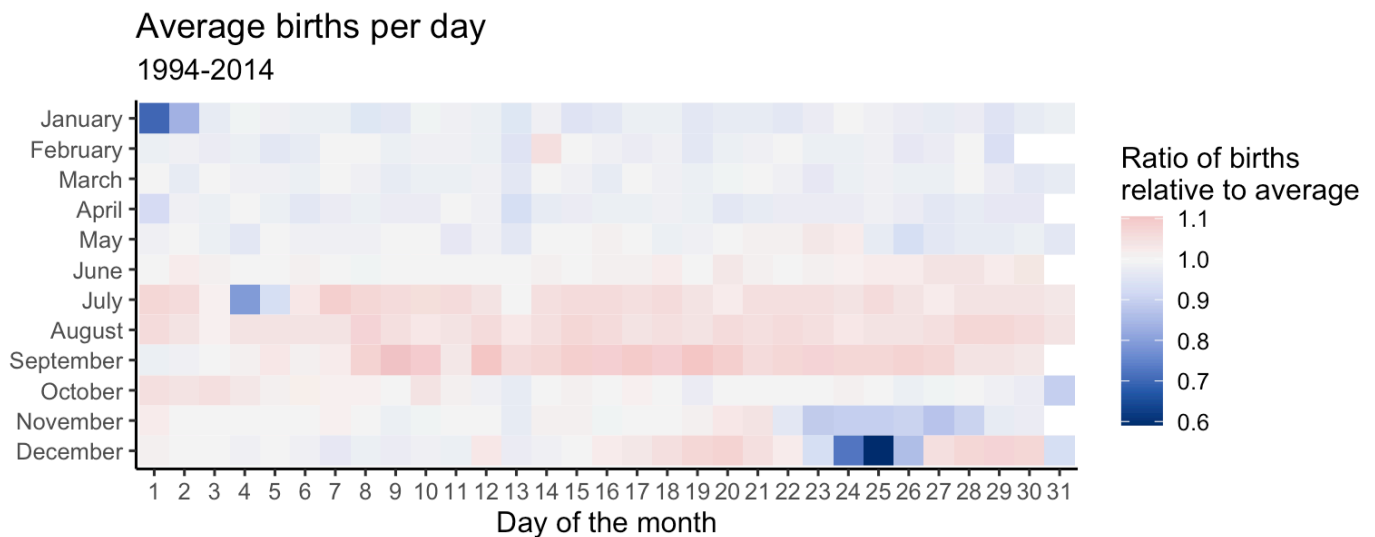


**Your turn:** Modify the plot to use an appropriate color palette. What days have an unusually high or low number of births?

```
birth_days_plot + scale_fill_continuous_diverging(mid = 1)
```



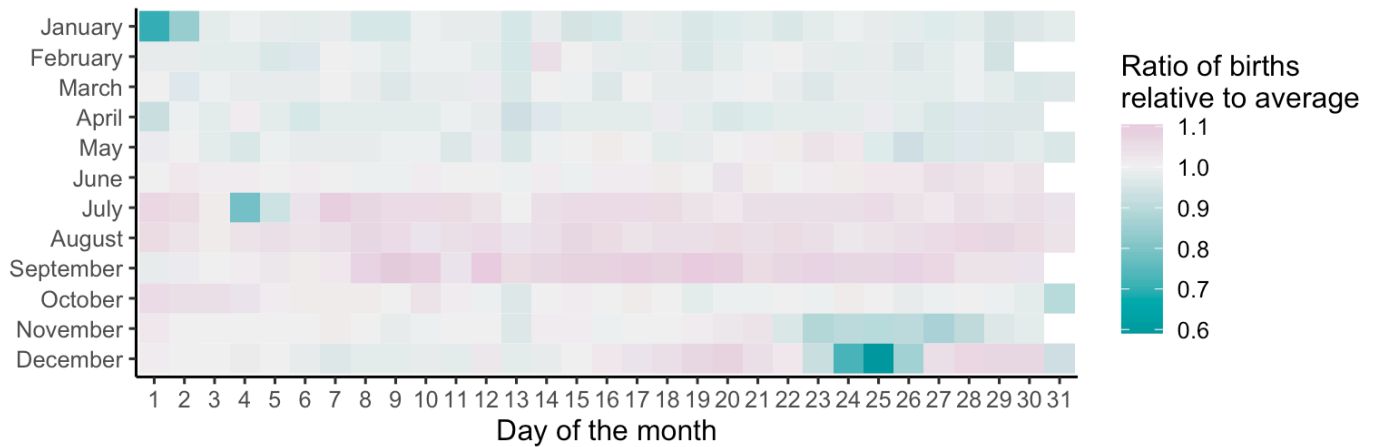
```
birth_days_plot + scale_fill_continuous_diverging(palette = "Blue-Red 3", mid = 1)
```



```
birth_days_plot + scale_fill_continuous_diverging(palette = "Tropic", mid = 1)
```

## Average births per day

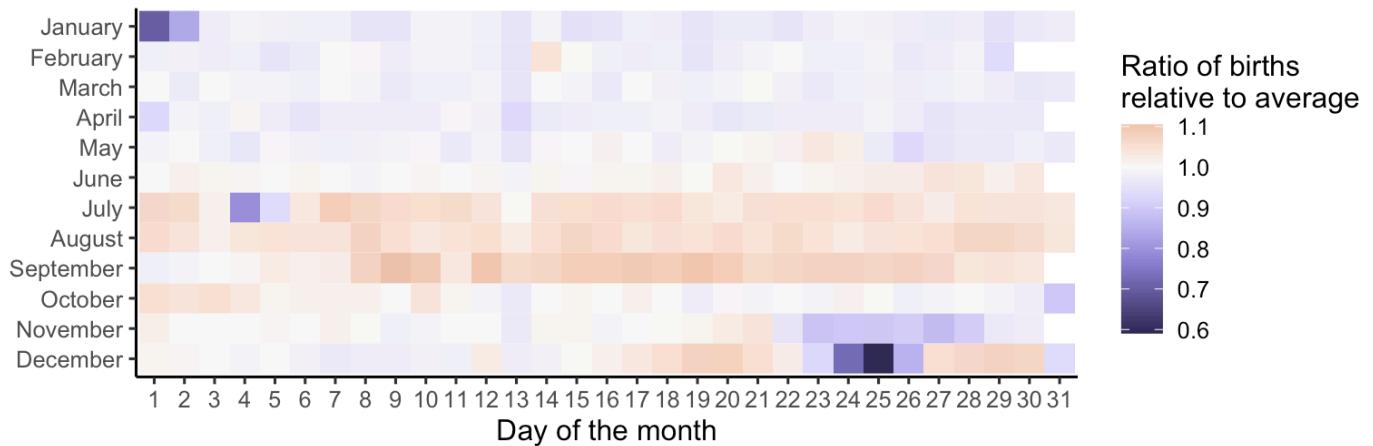
1994-2014



```
birth_days_plot + scale_fill_continuous_diverging(palette = "Purple-Brown", mid = 1)
```

## Average births per day

1994-2014



Add response here.

## Session information

## Footnotes

1. Collected by [FiveThirtyEight](#). ↵
2. Essentially a replication of [Carl Bialik's original chart](#). ↵

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