

# Data types with



Open **05-Data-Types.Rmd**

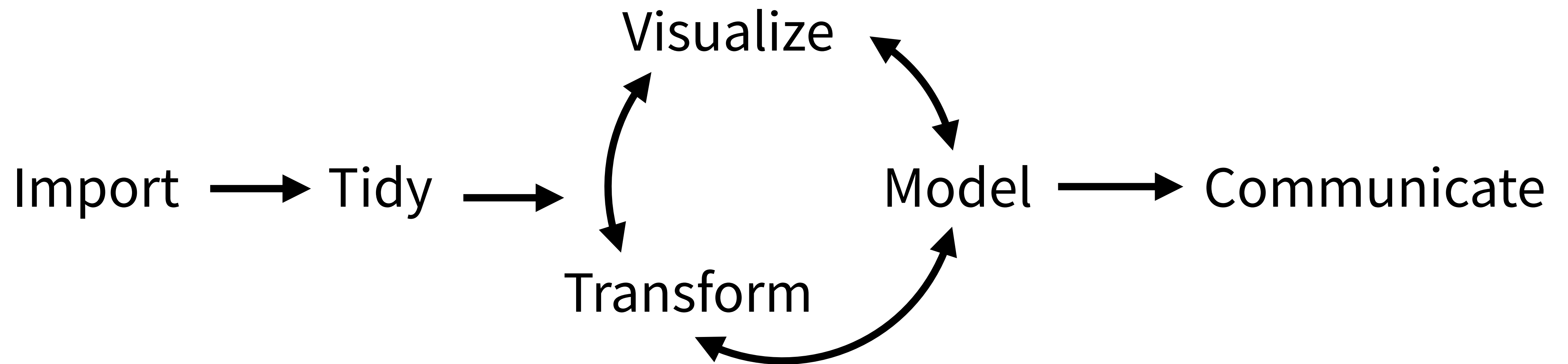


# Quiz

What types of data are in this data set?

|    | time_hour           | name                     | air_time             | distance | day     | delayed |
|----|---------------------|--------------------------|----------------------|----------|---------|---------|
| 1  | 2013-01-01 05:00:00 | United Air Lines Inc.    | 13620s (~3.78 hours) | 1400     | Tuesday | TRUE    |
| 2  | 2013-01-01 05:00:00 | United Air Lines Inc.    | 13620s (~3.78 hours) | 1416     | Tuesday | TRUE    |
| 3  | 2013-01-01 05:00:00 | American Airlines Inc.   | 9600s (~2.67 hours)  | 1089     | Tuesday | TRUE    |
| 4  | 2013-01-01 05:00:00 | JetBlue Airways          | 10980s (~3.05 hours) | 1576     | Tuesday | FALSE   |
| 5  | 2013-01-01 06:00:00 | Delta Air Lines Inc.     | 6960s (~1.93 hours)  | 762      | Tuesday | FALSE   |
| 6  | 2013-01-01 05:00:00 | United Air Lines Inc.    | 9000s (~2.5 hours)   | 719      | Tuesday | TRUE    |
| 7  | 2013-01-01 06:00:00 | JetBlue Airways          | 9480s (~2.63 hours)  | 1065     | Tuesday | TRUE    |
| 8  | 2013-01-01 06:00:00 | ExpressJet Airlines Inc. | 3180s (~53 minutes)  | 229      | Tuesday | FALSE   |
| 9  | 2013-01-01 06:00:00 | JetBlue Airways          | 8400s (~2.33 hours)  | 944      | Tuesday | FALSE   |
| 10 | 2013-01-01 06:00:00 | American Airlines Inc.   | 8280s (~2.3 hours)   | 733      | Tuesday | TRUE    |
| 11 | 2013-01-01 06:00:00 | JetBlue Airways          | 8940s (~2.48 hours)  | 1028     | Tuesday | FALSE   |

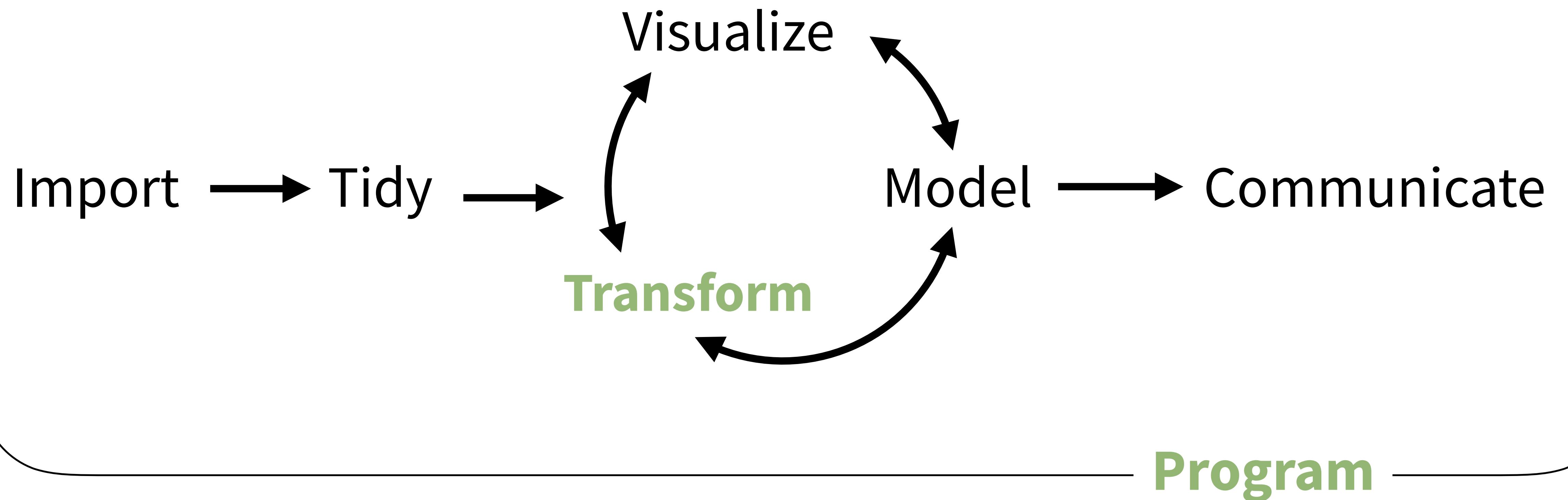
# (Applied) Data Science



Program



# (Applied) Data Science





# Logicals



# Most useful skills

## 1. Math with logicals



# Math

When you do math with logicals, **TRUE becomes 1** and **FALSE becomes 0**.





# Math

When you do math with logicals, **TRUE becomes 1** and **FALSE becomes 0**.

- The **sum** of a logical vector is the **count of TRUEs**

```
sum(c(TRUE, FALSE, TRUE, TRUE))
```

```
## 3
```



# Math

When you do math with logicals, **TRUE becomes 1** and **FALSE becomes 0**.

- The **sum** of a logical vector is the **count of TRUEs**

```
sum(c(TRUE, FALSE, TRUE, TRUE))  
##    3
```

- The **mean** of a logical vector is the **proportion of TRUEs**

```
mean(c(1, 2, 3, 4) < 4)  
## 0.75
```



# Warm Up

Did you fly here?

Did your flight arrive late?



# Your Turn 1

Create a logical variable in flights that displays whether a flight was delayed ( $\text{arr\_delay} > 0$ ). Remove all NAs in the variable.

Then create a summary table that shows:

1. How many flights were delayed
2. What proportion of flights were delayed

04:00



```
flights %>%  
  mutate(delayed = arr_delay > 0) %>%  
  drop_na(delayed) %>%  
  summarise(total = sum(delayed), prop = mean(delayed))  
## # A tibble: 1 × 2  
##   total      prop  
##   <int>    <dbl>  
## 1 133004 0.4063101
```



# Strings



# Warm Up

Decide in your group:

Are boys names or girls names more likely to end in a vowel?

02:00

# (character) **strings**

Anything surrounded by quotes(") or single quotes(').

```
> "one"  
> "1"  
> "one's"  
> '"Hello World"'  
> "foo  
+  
+  
+ oops. I'm stuck in a string."
```





# Most useful skills

1. How to extract/ replace substrings
2. How to find matches for patterns
3. Regular expressions



# stringr



Simple, consistent functions for working with strings.

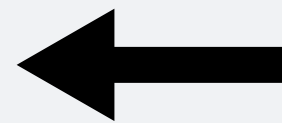
```
# install.packages("tidyverse")  
library(stringr)
```



```
install.packages("tidyverse")
```

does the equivalent of

```
install.packages("ggplot2")
install.packages("dplyr")
install.packages("tidyr")
install.packages("readr")
install.packages("purrr")
install.packages("tibble")
install.packages("hms")
install.packages("stringr")
install.packages("lubridate")
install.packages("forcats")
install.packages("DBI")
install.packages("haven")
install.packages("httr")
install.packages("jsonlite")
install.packages("readxl")
install.packages("rvest")
install.packages("xml2")
install.packages("modelr")
install.packages("broom")
```



```
library("tidyverse")
```

does the equivalent of

```
library("ggplot2")
library("dplyr")
library("tidyr")
library("readr")
library("purrr")
library("tibble")
```

```
install.packages("tidyverse")
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install.packages("stringr")
install.packages("lubridate")
install.packages("forcats")
install.packages("DBI")
install.packages("haven")
install.packages("httr")
install.packages("jsonlite")
install.packages("readxl")
install.packages("rvest")
install.packages("xml2")
install.packages("modelr")
install.packages("broom")
```




```
library("tidyverse")
```

does the equivalent of

```
library("ggplot2")
library("dplyr")
library("tidyr")
library("readr")
library("purrr")
library("tibble")
```



# babynames



| <b>year</b> | <b>sex</b> | <b>name</b> | <b>n</b> | <b>prop</b>  |
|-------------|------------|-------------|----------|--------------|
| <dbl>       | <chr>      | <chr>       | <int>    | <dbl>        |
| 1880        | F          | Mary        | 7065     | 7.238433e-02 |
| 1880        | F          | Anna        | 2604     | 2.667923e-02 |
| 1880        | F          | Emma        | 2003     | 2.052170e-02 |
| 1880        | F          | Elizabeth   | 1939     | 1.986599e-02 |
| 1880        | F          | Minnie      | 1746     | 1.788861e-02 |
| 1880        | F          | Margaret    | 1578     | 1.616737e-02 |
| 1880        | F          | Ida         | 1472     | 1.508135e-02 |
| 1880        | F          | Alice       | 1414     | 1.448711e-02 |
| 1880        | F          | Bertha      | 1320     | 1.352404e-02 |
| 1880        | F          | Sarah       | 1288     | 1.319618e-02 |

1–10 of 1,858,689 rows

Previous

1

2

3

4

5

6

...

100

Next



# str\_sub()

Extract or replace portions of a string with **str\_sub()**

```
str_sub(string, start = 1L, end = -1L)
```

**string(s) to  
manipulate**

**position of first  
character to extract  
within each string**

**position of last  
character to extract  
within each string**

# Quiz

What will this return?

```
str_sub("Garrett", 1, 2)
```



# Quiz

What will this return?

```
str_sub("Garrett", 1, 2)
```

"Ga"



# Quiz

What will this return?

```
str_sub("Garrett", 1, 1)
```

# Quiz

What will this return?

```
str_sub("Garrett", 1, 1)
```

"G"



# Quiz

What will this return?

```
str_sub("Garrett", 2)
```

# Quiz

What will this return?

```
str_sub("Garrett", 2)
```

"arrett"



# Quiz

What will this return?

```
str_sub("Garrett", -3)
```



# Quiz

What will this return?

```
str_sub("Garrett", -3)
```

"ett"

# Quiz

What will this return?

```
g <- "Garrett"
```

```
str_sub(g, -3) <- "eth"
```

```
g
```



# Quiz

What will this return?

```
g <- "Garrett"
```

```
str_sub(g, -3) <- "eth"
```

```
g
```

```
"Garreth"
```

# Your Turn 2

In your group, fill in the blanks to:

1. Isolate the last letter of every name
2. and create a logical variable that displays whether the last letter is one of "a", "e", "i", "o", "u", or "y".
3. Use a weighted mean to calculate the proportion of children whose name ends in a vowel (by year and sex)
4. and then display the results as a line plot.

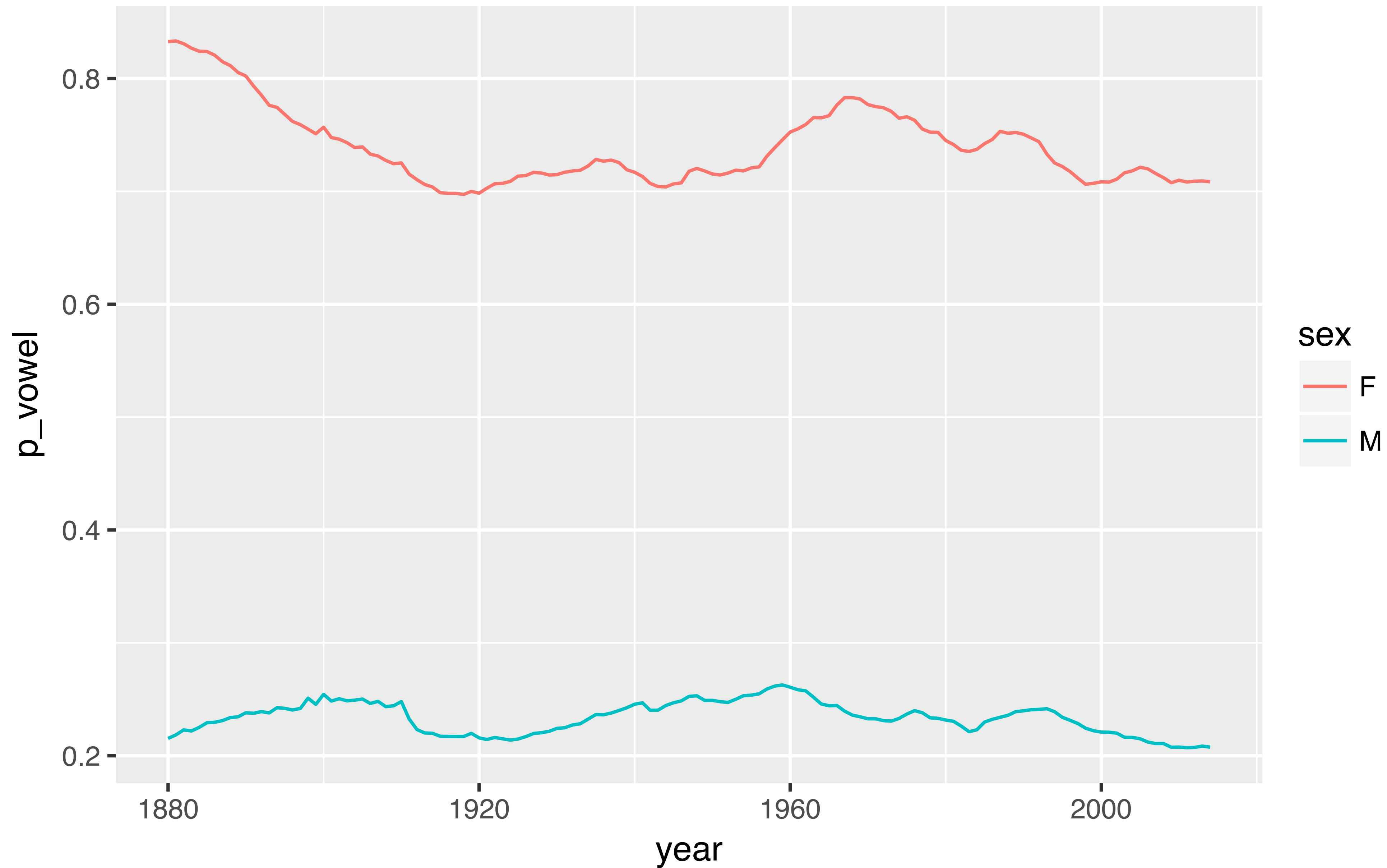
A digital timer display showing the time 05:00. The digits are black with a white outline, set against a white background with a thin black border.



```
babynames %>%  
  mutate(last = str_sub(name, -1),  
         vowel = last %in% c("a", "e", "i", "o", "u", "y")) %>%  
  group_by(year, sex) %>%  
  summarise(p_vowel = weighted.mean(vowel, n)) %>%  
  ggplot(aes(year, p_vowel, color = sex)) +  
    geom_line()
```



# Proportion of names that end in a vowel



```
help(package = stringr)
```

## Simple, Consistent Wrappers for Common String Operations



### Documentation for package 'stringr' version 1.2.0

- [DESCRIPTION file.](#)
- [User guides, package vignettes and other documentation.](#)

### Help Pages

[boundary](#)

Control matching behaviour with modifier functions.

[case](#)

Convert case of a string.

[coll](#)

Control matching behaviour with modifier functions.

[fixed](#)

Control matching behaviour with modifier functions.

[fruit](#)

Sample character vectors for practicing string manipulations.

[invert\\_match](#)

Switch location of matches to location of non-matches.

[modifiers](#)

Control matching behaviour with modifier functions.

[regex](#)

Control matching behaviour with modifier functions.

Sample character vectors for practicing string manipulations.



# Factors





# Warm Up

Decide in your group:

Do married people watch more or less TV than single people?

02:00

# gss\_cat

```
library(forcats)
gss_cat
```

A sample of data from the General Social Survey, a long-running US survey conducted by NORC at the University of Chicago.

| <b>year</b><br><int> | <b>marital</b><br><fctr> | <b>age</b><br><int> | <b>race</b><br><fctr> | <b>rincome</b><br><fctr> | <b>partyid</b><br><fctr> |
|----------------------|--------------------------|---------------------|-----------------------|--------------------------|--------------------------|
| 2000                 | Never married            | 26                  | White                 | \$8000 to 9999           | Ind,near rep             |
| 2000                 | Divorced                 | 48                  | White                 | \$8000 to 9999           | Not str republican       |
| 2000                 | Widowed                  | 67                  | White                 | Not applicable           | Independent              |
| 2000                 | Never married            | 39                  | White                 | Not applicable           | Ind,near rep             |
| 2000                 | Divorced                 | 25                  | White                 | Not applicable           | Not str democrat         |
| 2000                 | Married                  | 25                  | White                 | \$20000 – 24999          | Strong democrat          |
| 2000                 | Never married            | 36                  | White                 | \$25000 or more          | Not str republican       |
| 2000                 | Divorced                 | 44                  | White                 | \$7000 to 7999           | Ind,near dem             |
| 2000                 | Married                  | 44                  | White                 | \$25000 or more          | Not str democrat         |



# factors

R's representation of categorical data. Consists of:

1. A set of **values**
2. An ordered set of **valid levels**

```
eyes <- factor(x = c("blue", "green", "green"),  
              levels = c("blue", "brown", "green"))  
  
eyes  
## [1] blue  green green  
## Levels: blue brown green
```





# factors

Stored as an integer vector with a levels attribute

```
unclass(eyes)
## 1 2 2
## attr(,"levels")
## "blue" "brown" "green"
```





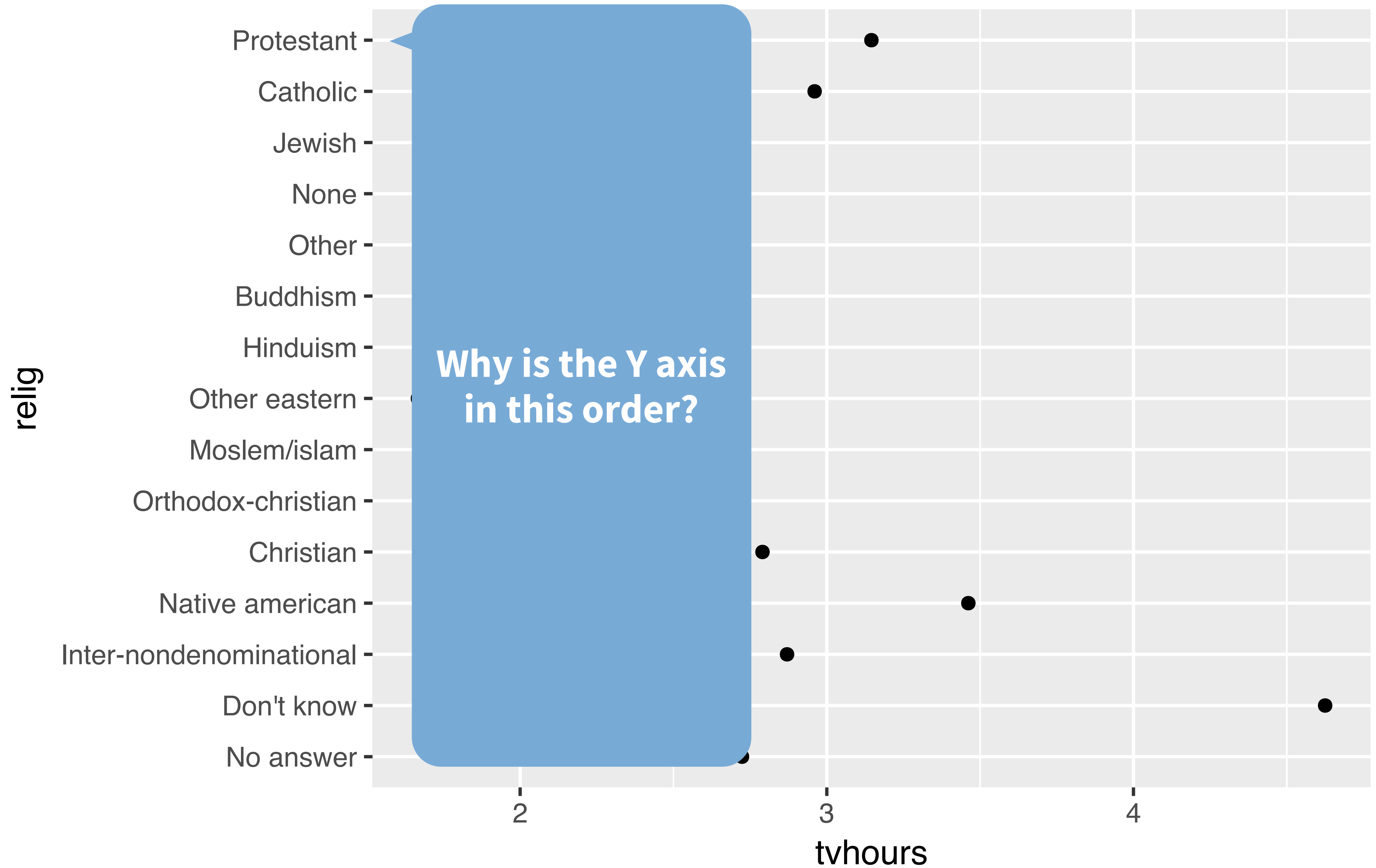
# Most useful skills

1. Reorder the levels
2. Recode the levels
3. Collapse levels

# Which religions watch the most TV?

```
gss_cat %>%  
  drop_na(tvhours) %>%  
  group_by(relig) %>%  
  summarise(tvhours = mean(tvhours)) %>%  
  ggplot(aes(tvhours, relig)) +  
    geom_point()
```



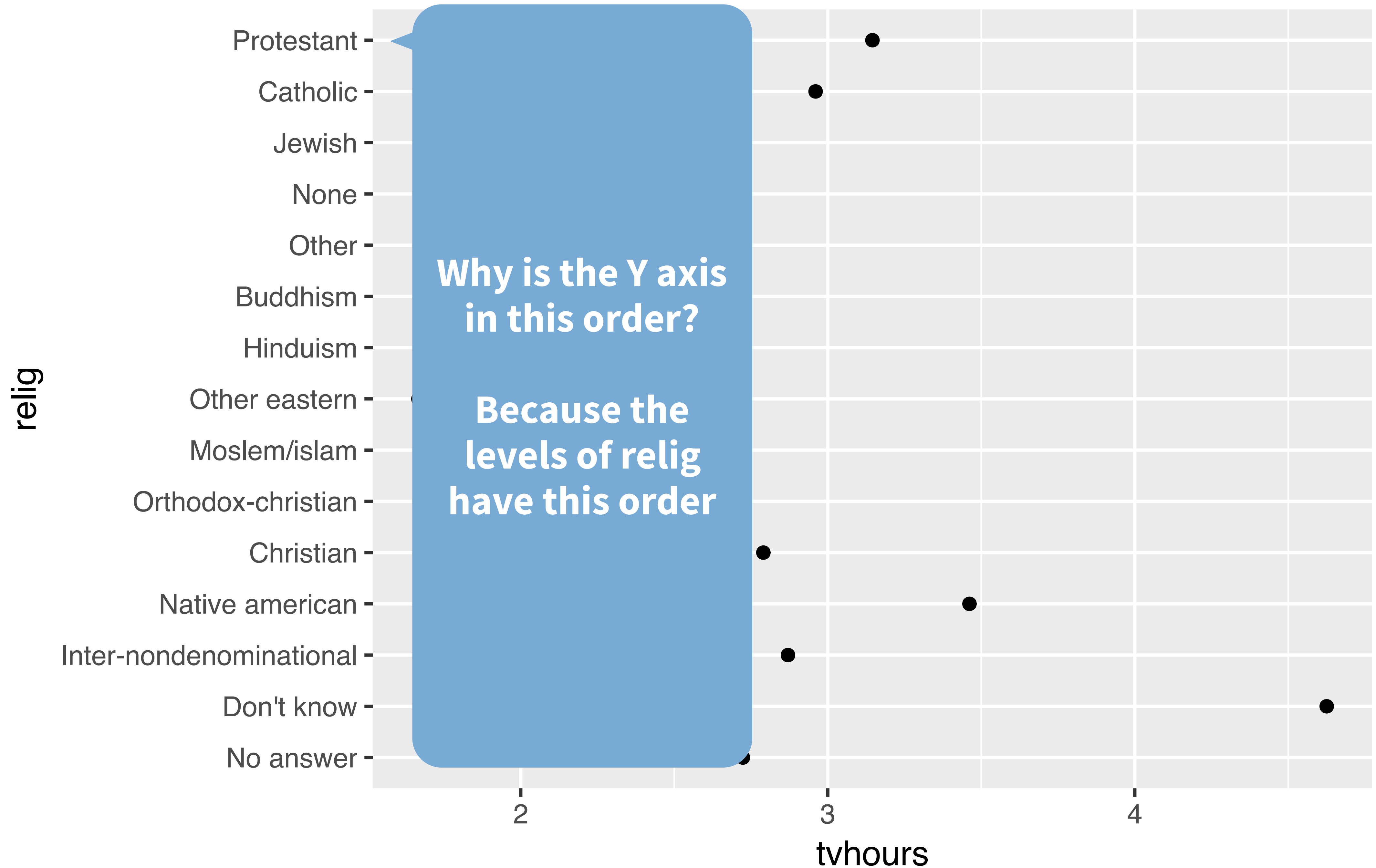


# levels()

Use **levels()** to access a factor's levels

```
levels(gss_cat$relig)
## [1] "No answer" "Don't know"
## [3] "Inter-nondenominational" "Native american"
## [5] "Christian" "Orthodox-christian"
## [7] "Moslem/islam" "Other eastern"
## [9] "Hinduism" "Buddhism"
## [11] "Other" "None"
## [13] "Jewish" "Catholic"
## [15] "Protestant" "Not applicable"
```





# Reordering levels



# forcats



Simple functions for working with factors.

```
# install.packages("tidyverse")  
library(forcats)
```



# fct\_reorder()

Reorders the levels of a factor based on the result of fun(x) applied to each group of cases (grouped by level).

```
fct_reorder(f, x, fun = median, ..., .desc = FALSE)
```

**factor to  
reorder**

**variable to  
reorder by**  
(in conjunction  
with fun)

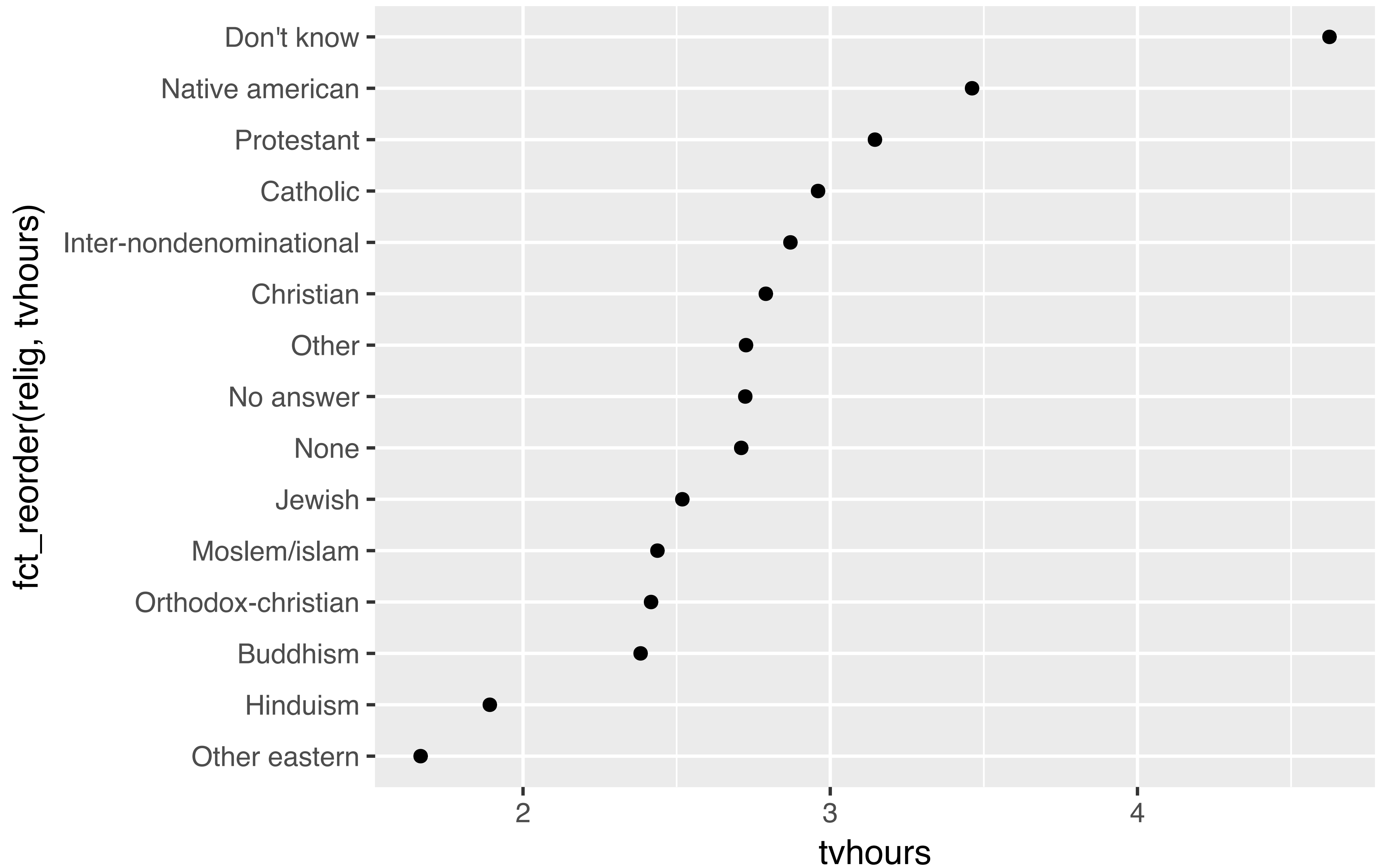
**function to  
reorder by**  
(in conjunction  
with x)

**put in descending  
order?**



```
gss_cat %>%  
  drop_na(tvhours) %>%  
  group_by(relig) %>%  
  summarise(tvhours = mean(tvhours)) %>%  
  ggplot(aes(tvhours, fct_reorder(relig, tvhours))) +  
    geom_point()
```





# Your Turn 3

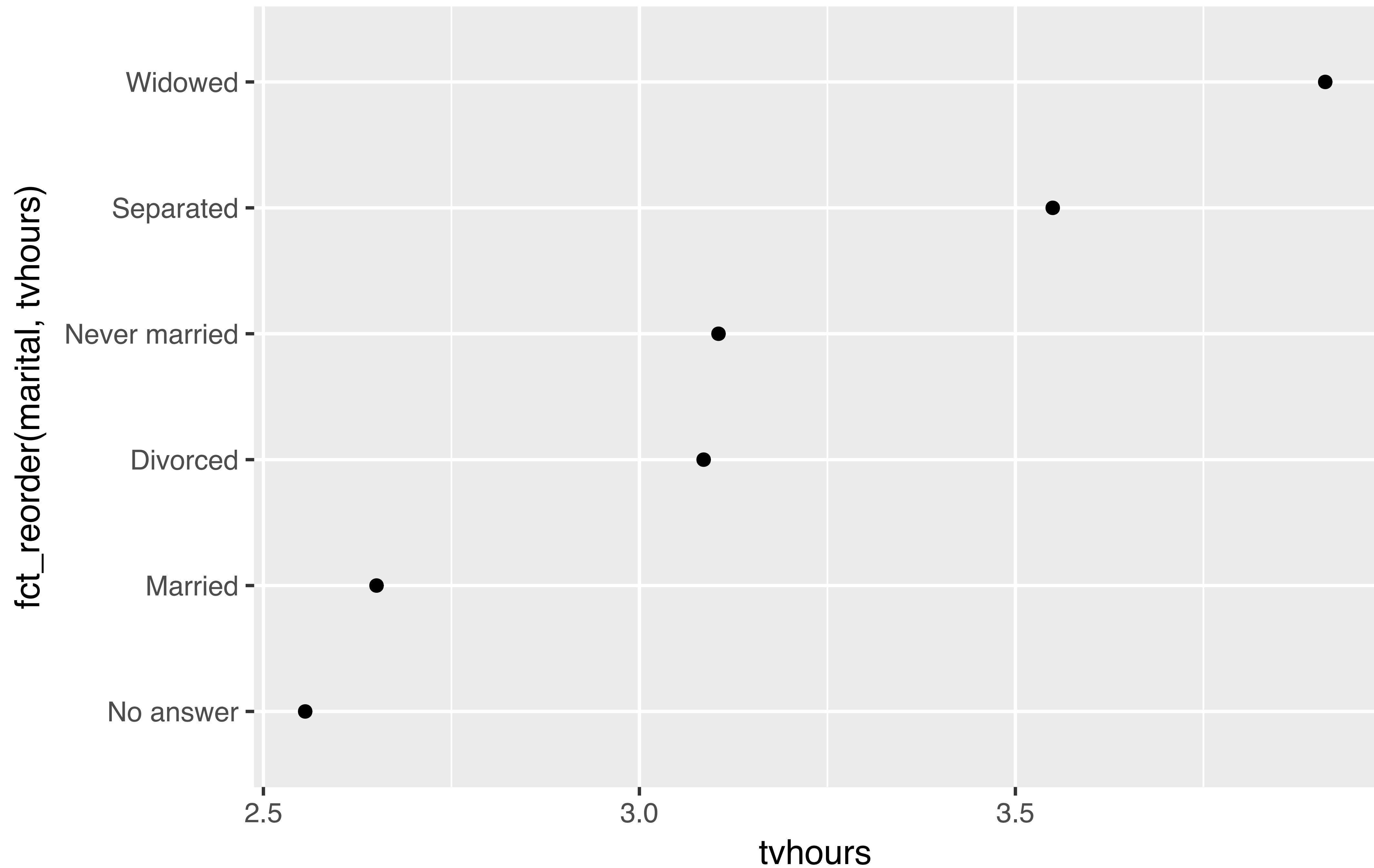
Repeat the previous exercise, some of whose code is in your notebook, to make a sensible graph of average TV consumption by marital status.

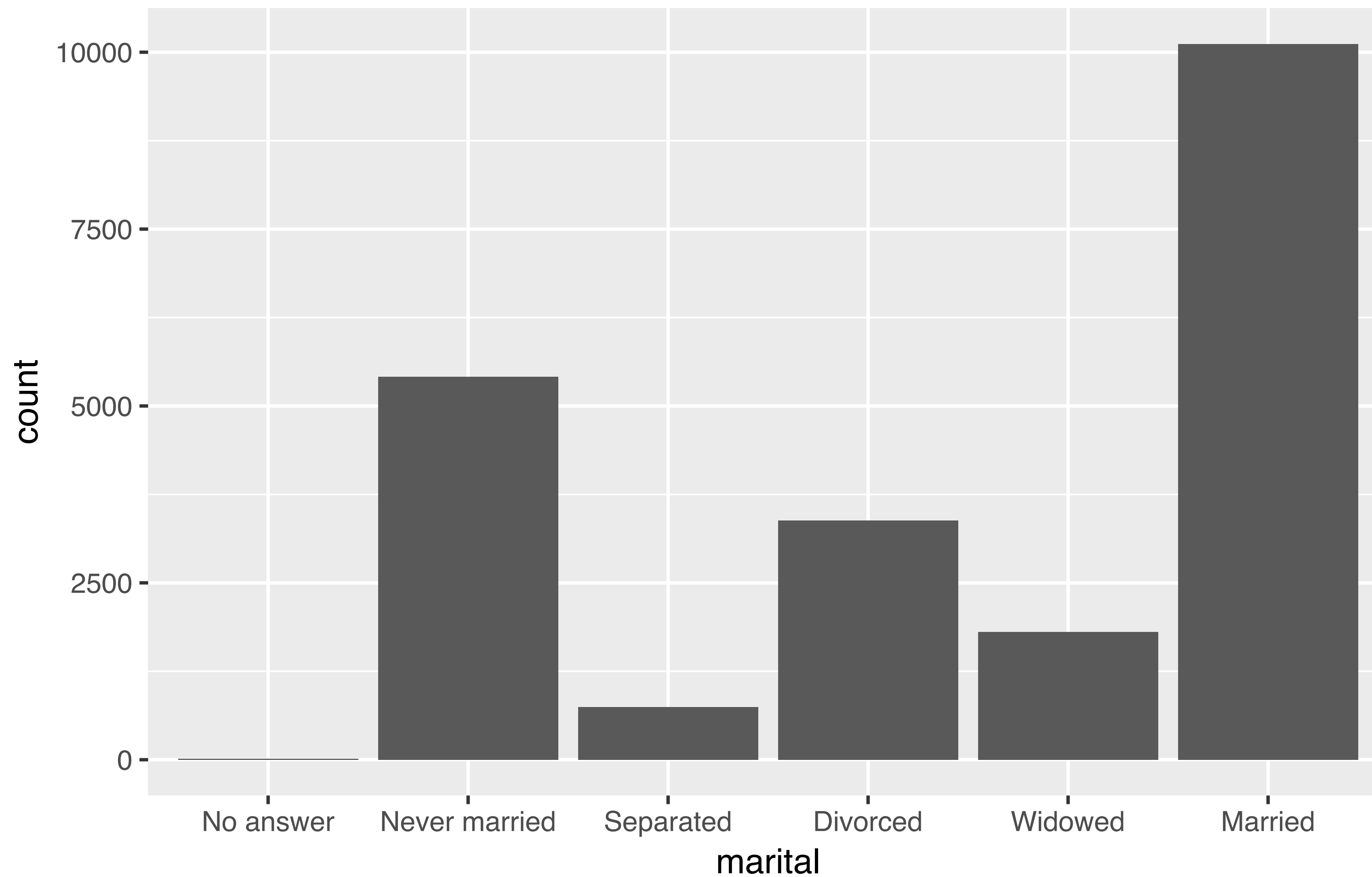
05:00

```
gss_cat %>%  
  drop_na(tvhours) %>%  
  group_by(marital) %>%  
  summarise(tvhours = mean(tvhours)) %>%  
  ggplot(aes(tvhours, fct_reorder(marital, tvhours))) +  
    geom_point()
```





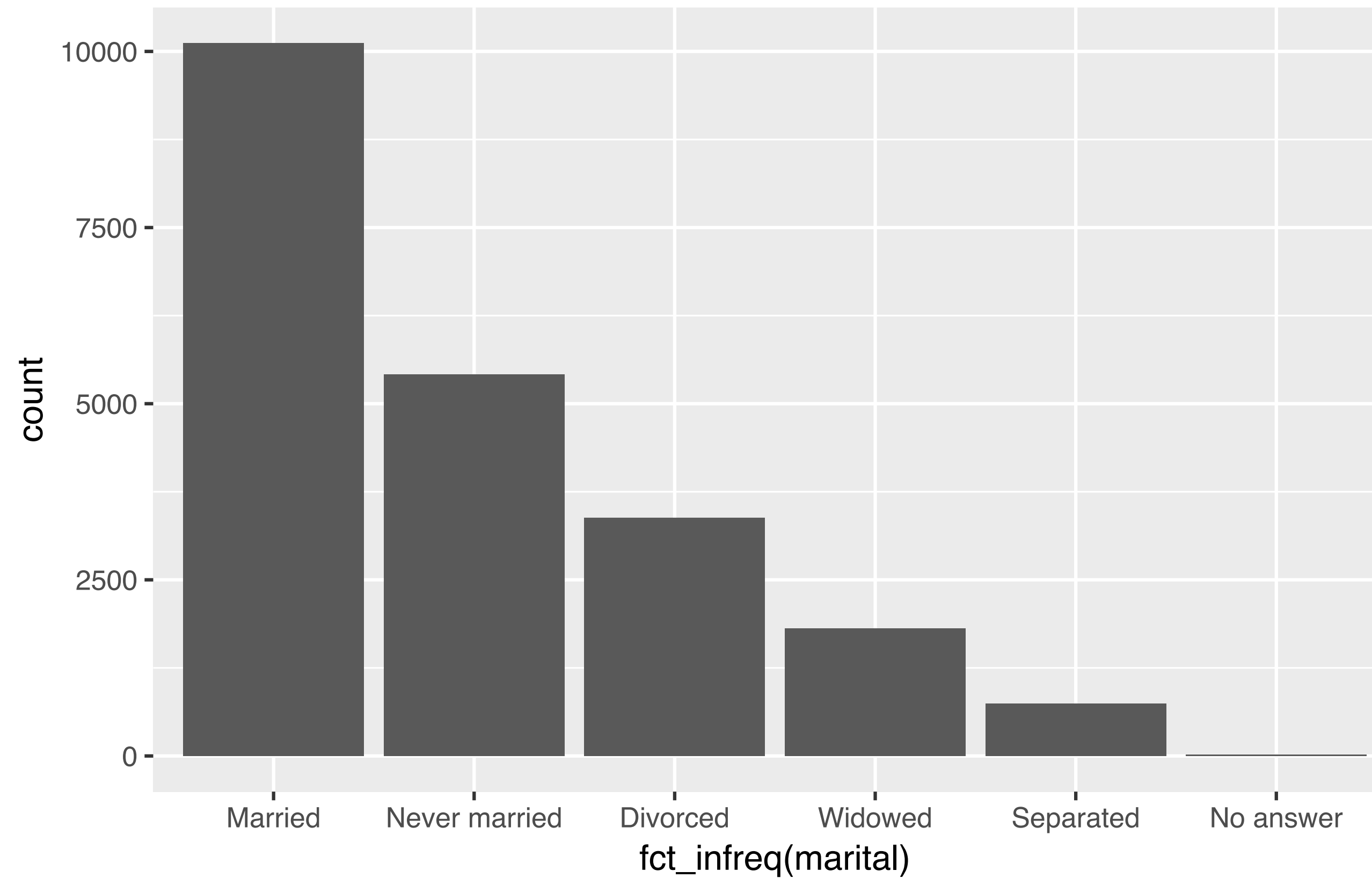




```
gss_cat %>%  
  ggplot(aes(marital)) + geom_bar()
```



# fct\_infreq



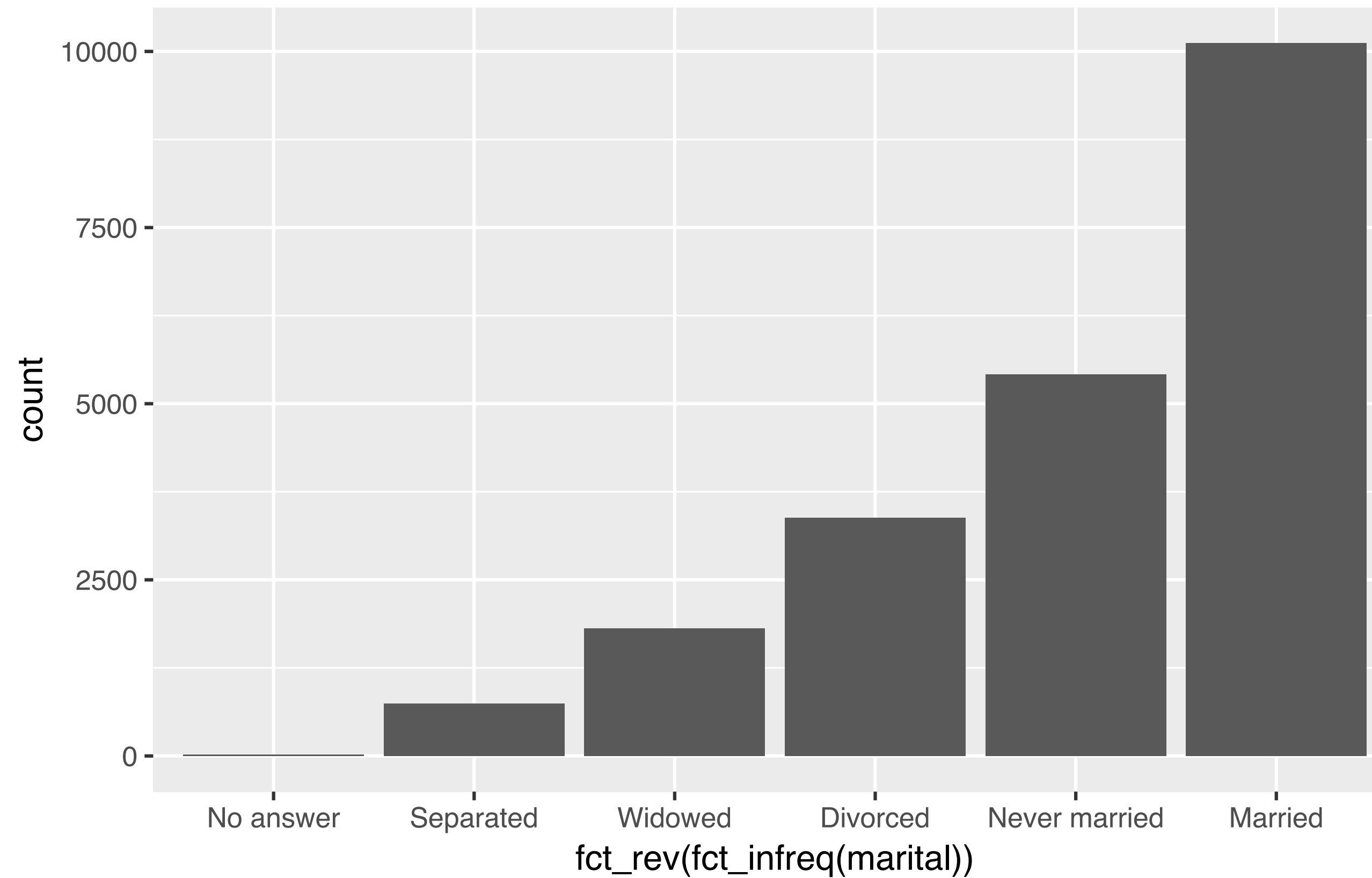
```
gss_cat %>%
```

```
  ggplot(aes(fct_infreq(marital))) + geom_bar()
```





# fct\_rev



```
gss_cat %>%
```

```
  ggplot(aes(fct_rev(fct_infreq(marital)))) + geom_bar()
```

# Changing level values



# Your Turn 4

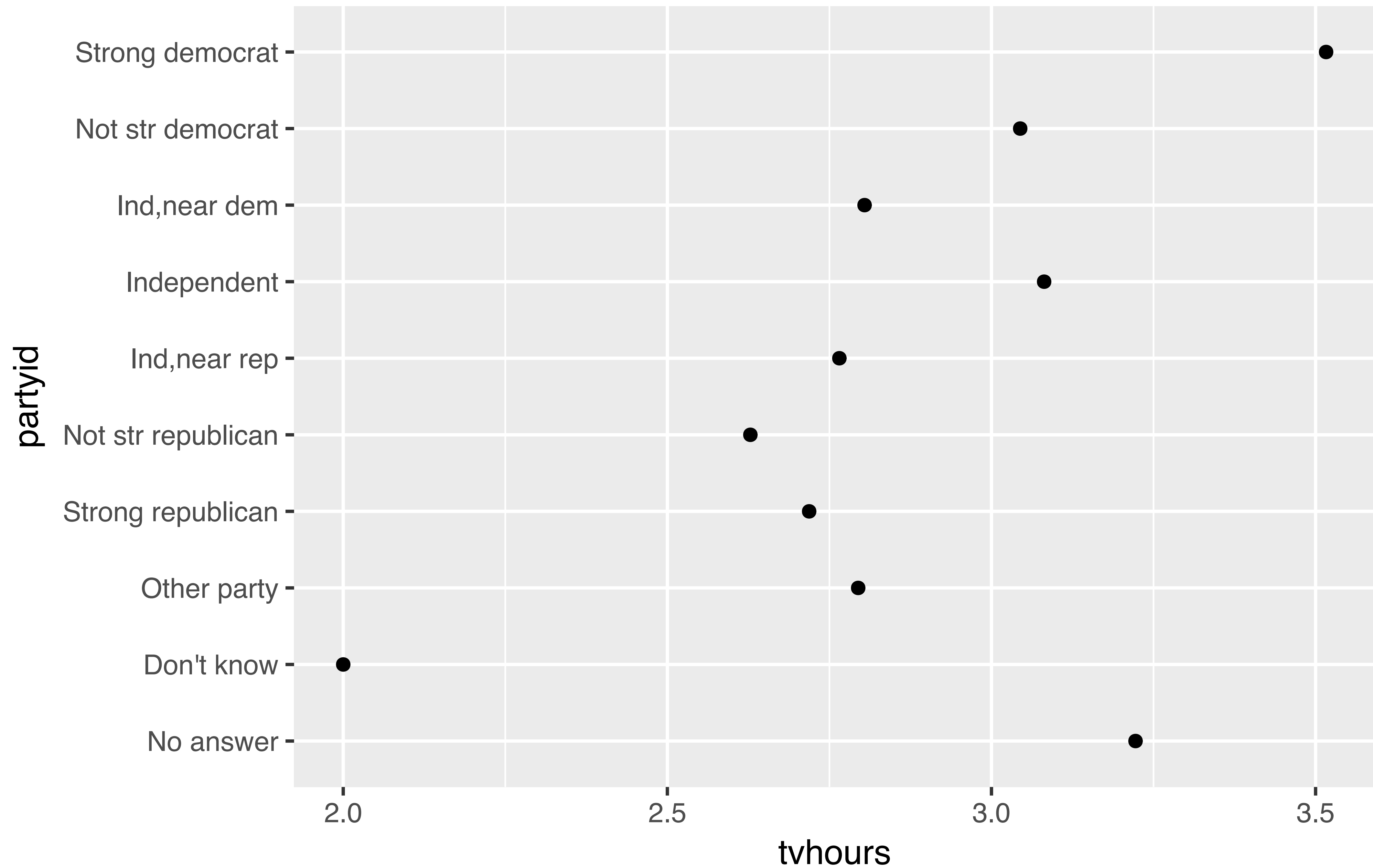
Do you think liberals or conservatives watch more TV?

Compute average tv hours by party ID and then plot the results.

04:00



```
gss_cat %>%  
  drop_na(tvhours) %>%  
  group_by(partyid) %>%  
  summarise(tvhours = mean(tvhours)) %>%  
  ggplot(aes(tvhours, partyid)) +  
    geom_point()
```



# fct\_recode()

Changes values of levels

```
fct_recode(f, ...)
```

**factor with  
levels**

**new level = old level  
pairs** (as a named  
character vector)

```
gss_cat %>%  
  drop_na(tvhours) %>%  
  mutate(partyid = fct_recode(partyid,  
    "Republican, strong"      = "Strong republican",  
    "Republican, weak"        = "Not str republican",  
    "Independent, near rep"   = "Ind,near rep",  
    "Independent, near dem"   = "Ind,near dem",  
    "Democrat, weak"          = "Not str democrat",  
    "Democrat, strong"        = "Strong democrat")) %>%  
  group_by(partyid) %>%  
  summarise(tvhours = mean(tvhours)) %>%  
  ggplot(aes(tvhours, partyid)) +  
    geom_point()
```





partyid

Democrat, strong  
Democrat, weak  
Independent, near dem  
Independent  
Independent, near rep  
Republican, weak  
Republican, strong  
Other party  
Don't know  
No answer

2.0

2.5

3.0

3.5

tvhours



# Collapsing levels



# fct\_collapse()

Changes multiple levels into single levels

```
fct_collapse(f, ...)
```

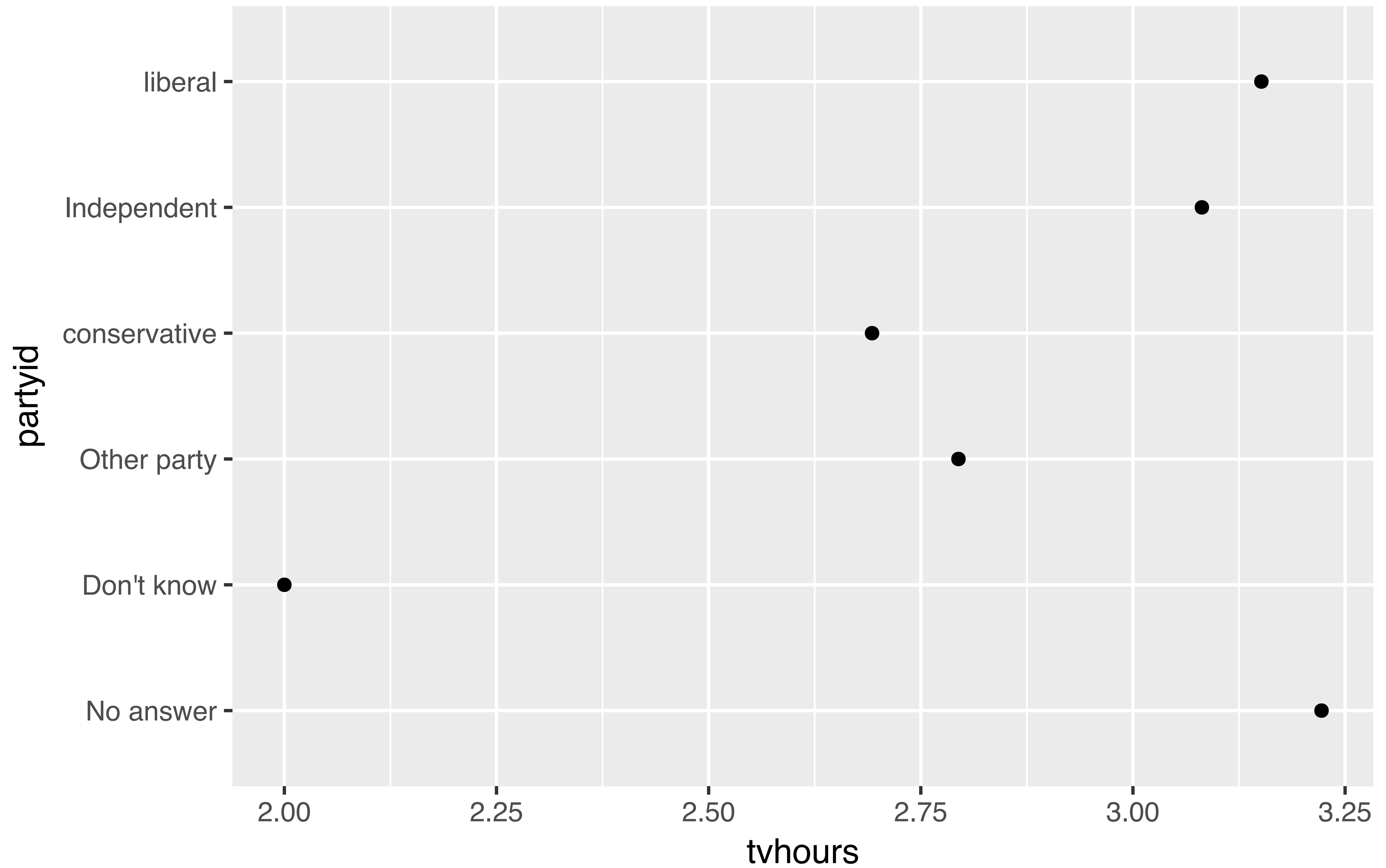
**factor with  
levels**

**named arguments set to a  
character vector** (levels in the  
vector will be collapsed to the name  
of the argument)

```
gss_cat %>%  
  drop_na(tvhours) %>%  
  mutate(partyid = fct_collapse(partyid,  
    conservative = c("Strong republican",  
                      "Not str republican",  
                      "Ind,near rep"),  
    liberal = c("Strong democrat",  
                "Not str democrat",  
                "Ind,near dem")) %>%  
  group_by(partyid) %>%  
  summarise(tvhours = mean(tvhours)) %>%  
  ggplot(aes(tvhours, partyid)) +  
    geom_point()
```







# fct\_lump()

Collapses levels with fewest values into a single level. By default collapses as many levels as possible such that the new level is still the smallest.

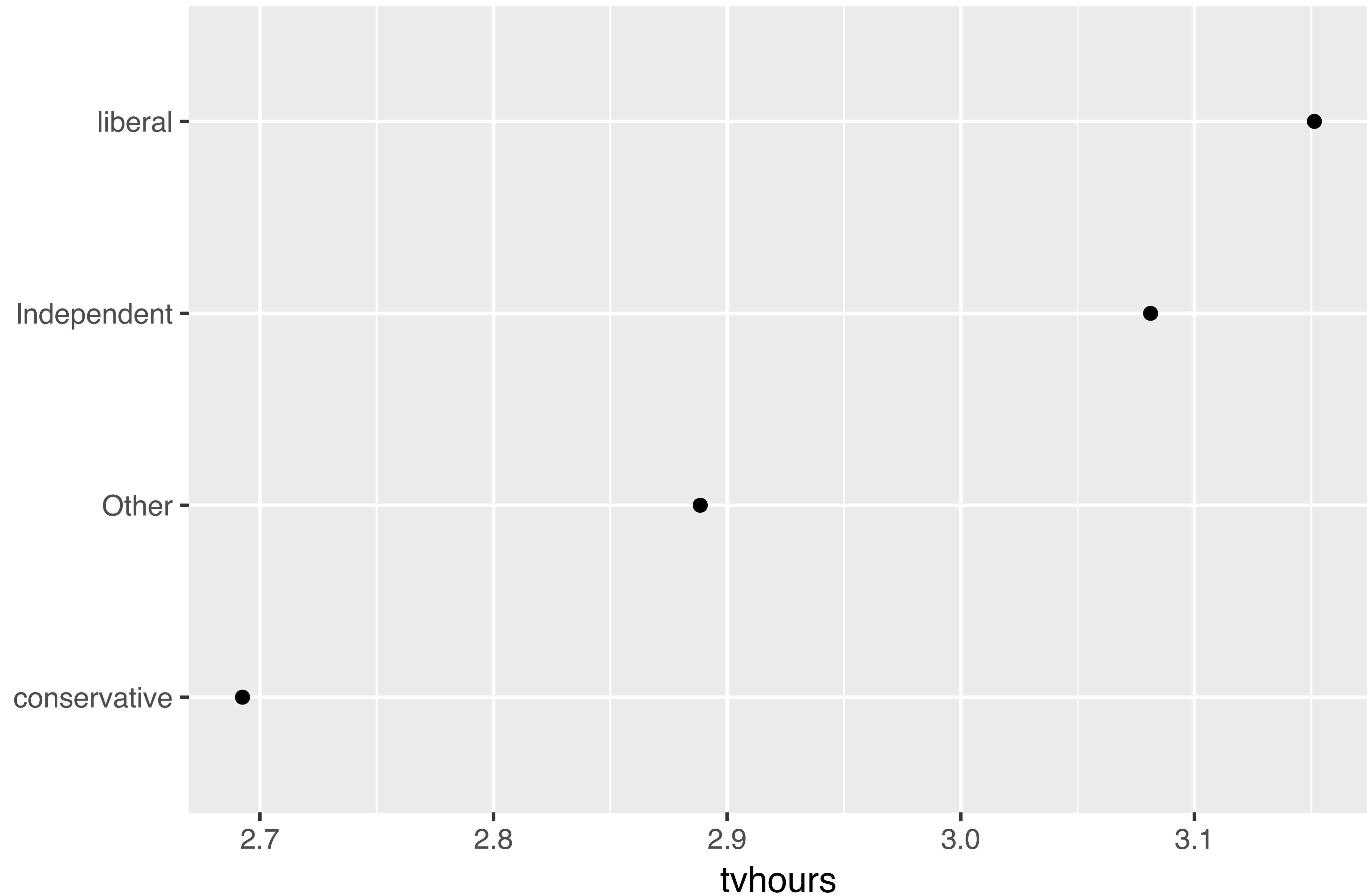
```
fct_lump(f, other_level = "Other", ...)
```

**factor with  
levels**

**name of new level**



partyid



```
gss_cat %>%  
  drop_na(tvhours) %>%  
  mutate(partyid = partyid %>%  
    fct_lump() %>%  
    fct_collapse(  
      conservative = c("Strong republican",  
                        "Not str republican", "Ind,near rep"),  
      liberal = c("Strong democrat", "Not str democrat",  
                  "Ind,near dem")) %>%  
    fct_reorder(tvhours, mean)  
  ) %>%  
  group_by(partyid) %>%  
  summarise(tvhours = mean(tvhours)) %>%  
  ggplot(aes(tvhours, partyid)) +  
    geom_point()
```





# Date times



# Quiz

Does every year have 365 days?

# Quiz

Does every day have 24 hours?



# Quiz

Does every minute have 60 seconds?



# Quiz

What does a month measure?

# Most useful skills

1. Parse a string into a date time class
2. Access and change parts of a date
3. Deal with time zones
4. Do math with instants and time spans

# Warm Up

Decide in your group:

- What is the best time of day to fly?
- What is the best day of the week to fly?

02:00



# Parsing dates and times





# hms



A class for representing just clock times.

```
# install.packages("tidyverse")  
library(hms)
```



# hms

2017-01-01 12:34:56

Stored as the number of seconds since 00:00:00.\*

```
library(hms)
```

```
hms(seconds = 56, min = 34, hour = 12)
```

```
## 12:34:56
```

```
unclass(hms(56, 34, 12))
```

```
## 45296
```



# hms()

2017-01-01 12:34:56

```
library(hms)
```

```
hms(seconds, minutes, hours, days)
```



# Your Turn 5

What is the best time of day to fly?

Use the **hour** and **minute** variables in `flights` to compute the time of day for each flight as an hms. Then use a smooth line to plot the relationship between time of day and **arr\_delay**.

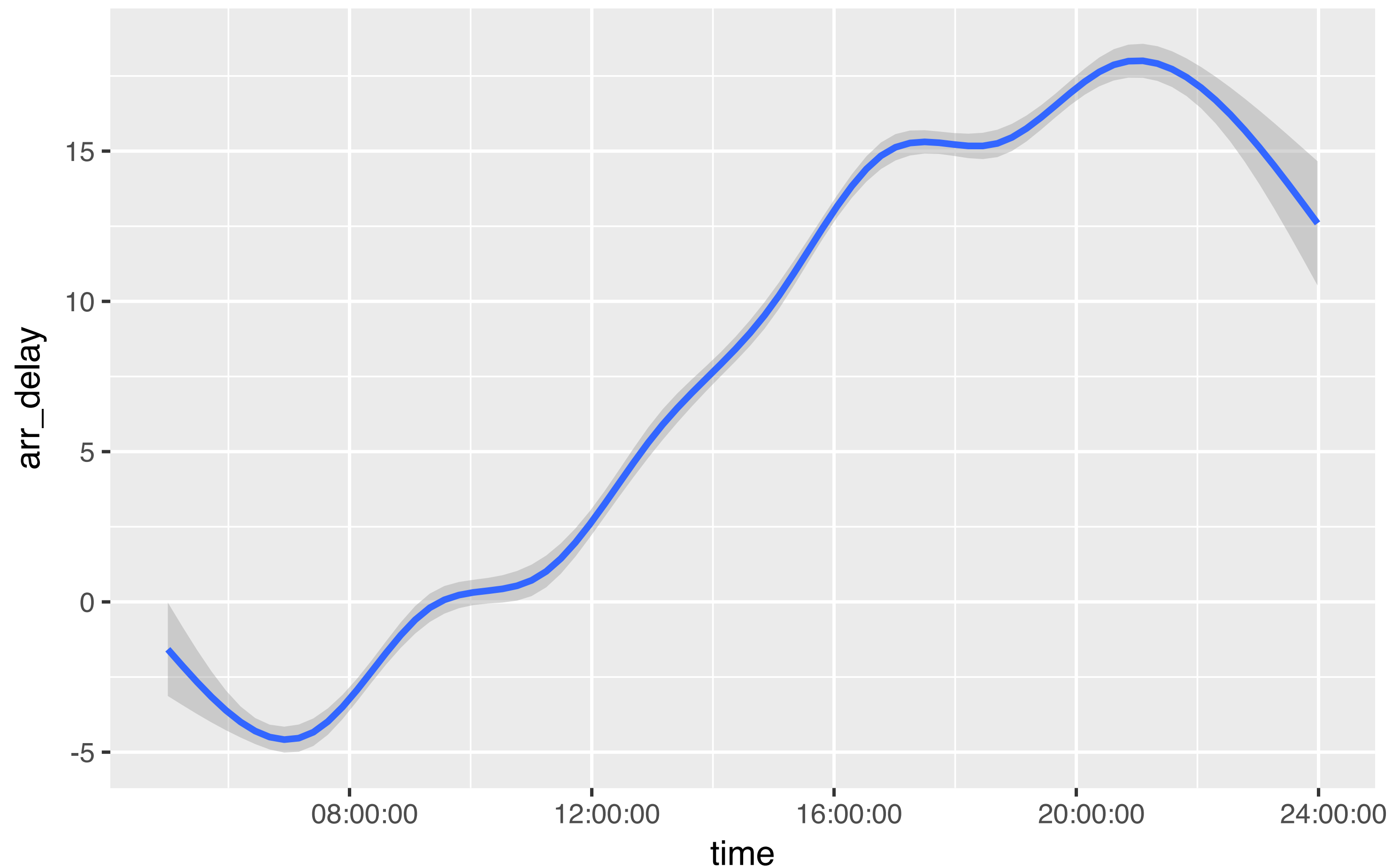
A digital clock display showing the time 05:00 in a black, segmented font on a white background.



```
flights %>%
```

```
  mutate(time = hms(hour = hour, minute = minute)) %>%
```

```
  ggplot(aes(time, arr_delay)) + geom_smooth()
```



# lubridate



Functions for working with dates and time spans

```
# install.packages("tidyverse")  
library(lubridate)
```



# ymd() family

To parse strings as dates, use a y, m, d, h, m, s combo

```
ymd("2017/01/11")
```

```
mdy("January 11, 2017")
```

```
ymd_hms("2017-01-11 01:30:55")
```

# Parsing functions

| function   | parses to                         |
|--|-----------------------------------|
| ymd_hms(), ymd_hm(), ymd_h()<br>ydm_hms(), ydm_hm(), ydm_h()<br>dmy_hms(), dmy_hm(), dmy_h()<br>mdy_hms(), mdy_hm(), mdy_h() | POSIXct                           |
| ymd(), ydm(), mdy()<br>myd(), dmy(), dym(), yq()   | Date<br>(POSIXct if tz specified) |
| hms(), hm(), ms()  | Period                            |





Accessing  
and changing  
components



# Accessing components

Extract components by name with a **singular** name

```
date <- ymd("2017-01-11")  
year(date)  
## 2017
```

# Setting components

Use the same function to set components

```
date  
## "2017-01-11"  
year(date) <- 1999  
date  
## "1999-01-11"
```

# Accessing date time components

| function | extracts       | extra arguments            |
|----------|----------------|----------------------------|
| year()   | year           |                            |
| month()  | month          | label = FALSE, abbr = TRUE |
| week()   | week           |                            |
| day()    | day of month   |                            |
| wday()   | day of week    | label = FALSE, abbr = TRUE |
| qday()   | day of quarter |                            |
| yday()   | day of year    |                            |
| hour()   | hour           |                            |
| minute() | minute         |                            |
| second() | second         |                            |



# Accessing components

```
wday(ymd("2017-01-11"))
```

```
## 2017
```

```
wday(ymd("2017-01-11"), label = TRUE)
```

```
## [1] Wed
```

```
## 7 Levels: Sun < Mon < Tues < Wed < Thurs < ... < Sat
```

```
wday(ymd("2017-01-11"), label = TRUE, abbr = FALSE)
```

```
## [1] Sunday
```

```
## 7 Levels: Sunday < Monday < Tuesday < ... < Saturday
```



# Your Turn 6

Fill in the blanks to:

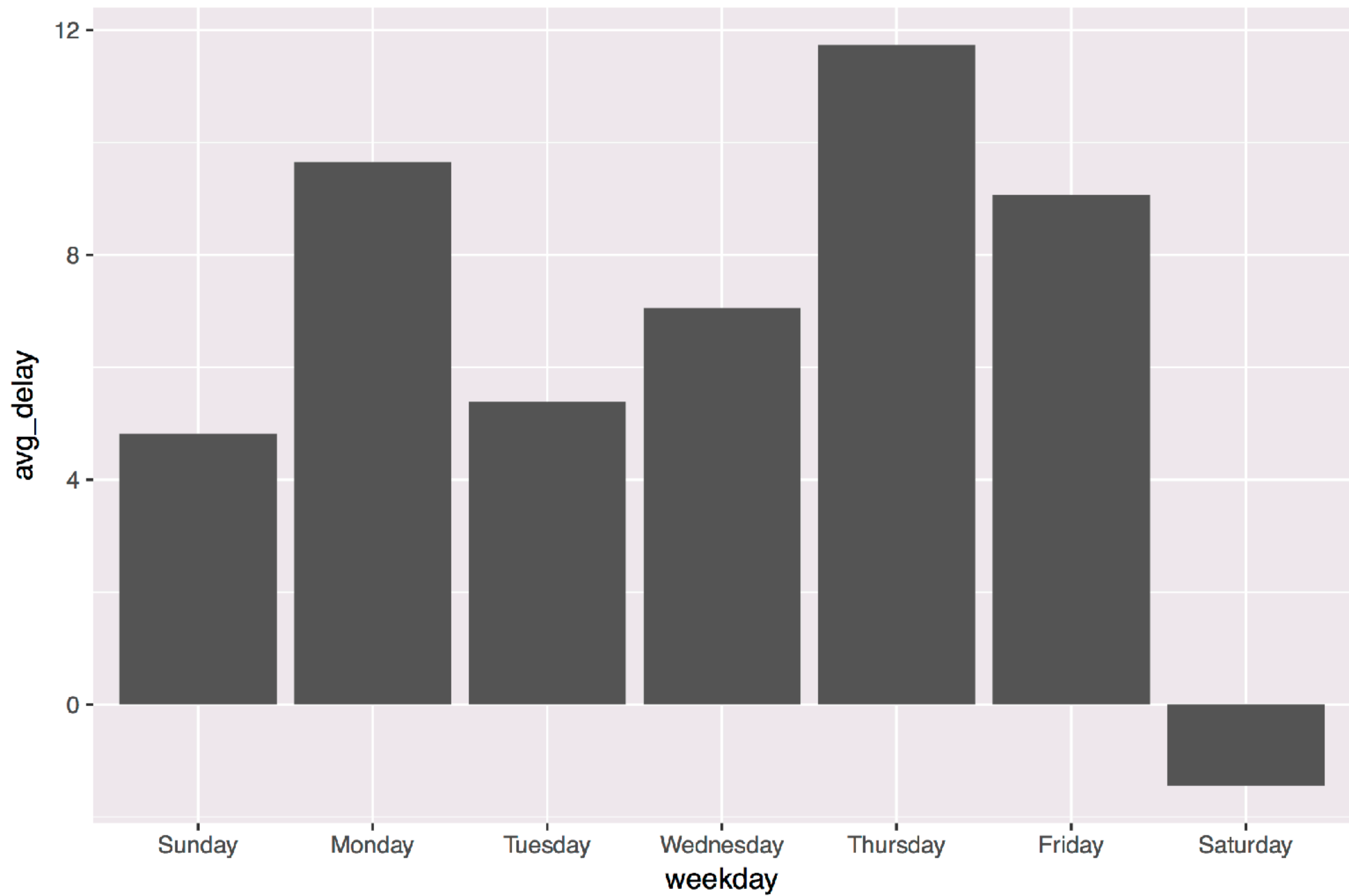
Extract the day of the week of each flight (as a full name) from **time\_hour**.

Calculate the average **dep\_delay** by day of the week.

Plot the results as a column chart (bar chart) with **geom\_col()**.

05:00

```
flights %>%  
  mutate(weekday = wday(time_hour, label = TRUE, abbr = FALSE)) %>%  
  group_by(weekday) %>%  
  drop_na(dep_delay) %>%  
  summarise(avg_delay = mean(dep_delay)) %>%  
  ggplot() +  
    geom_col(mapping = aes(x = weekday, y = avg_delay))
```





# Parsing functions

| function   | parses to                         |
|--|-----------------------------------|
| ymd_hms(), ymd_hm(), ymd_h()<br>ydm_hms(), ydm_hm(), ydm_h()<br>dmy_hms(), dmy_hm(), dmy_h()<br>mdy_hms(), mdy_hm(), mdy_h() | POSIXct                           |
| ymd(), ydm(), mdy()<br>myd(), dmy(), dym(), yq()   | Date<br>(POSIXct if tz specified) |
| hms(), hm(), ms()  | Period                            |



# Parsing functions

| function   | parses to                         |
|--|-----------------------------------|
| ymd_hms(), ymd_hm(), ymd_h()<br>ydm_hms(), ydm_hm(), ydm_h()<br>dmy_hms(), dmy_hm(), dmy_h()<br>mdy_hms(), mdy_hm(), mdy_h() | POSIXct                           |
| ymd(), ydm(), mdy()<br>myd(), dmy(), dym(), yq()   | Date<br>(POSIXct if tz specified) |
| <b>hms()</b> , hm(), ms()  | Period                            |

Same name as  
hms() in hms



# `hms::hms()`

**package  
name**

**function  
name**



`hms::hms()`

`lubridate::hms()`





# hms()

```
hms::hms(seconds = 3, hours = 5)
```

Use the  
hms() function in  
the hms package



# Data types with

