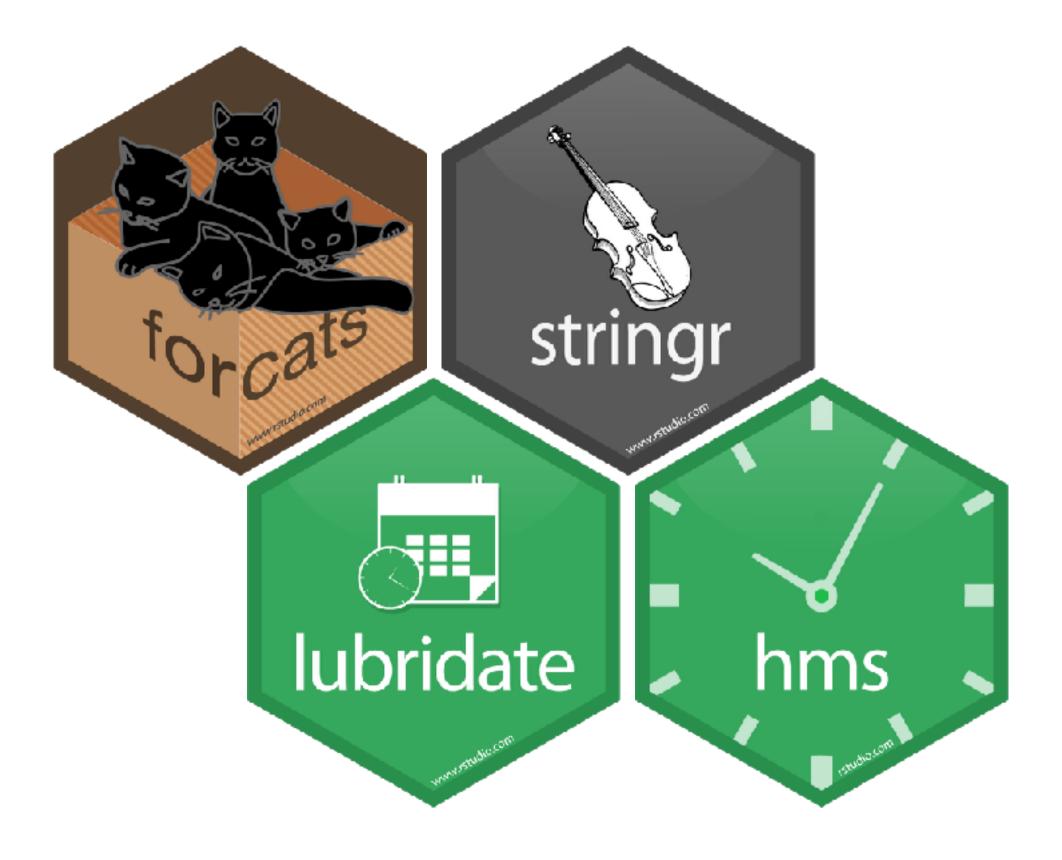
Data Types



gss_cat

gss_cat

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

						~ ~ ~
year <int></int>	marital <fctr></fctr>	age <int></int>	race <fctr></fctr>	rincome <fctr></fctr>	partyid <fctr></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	

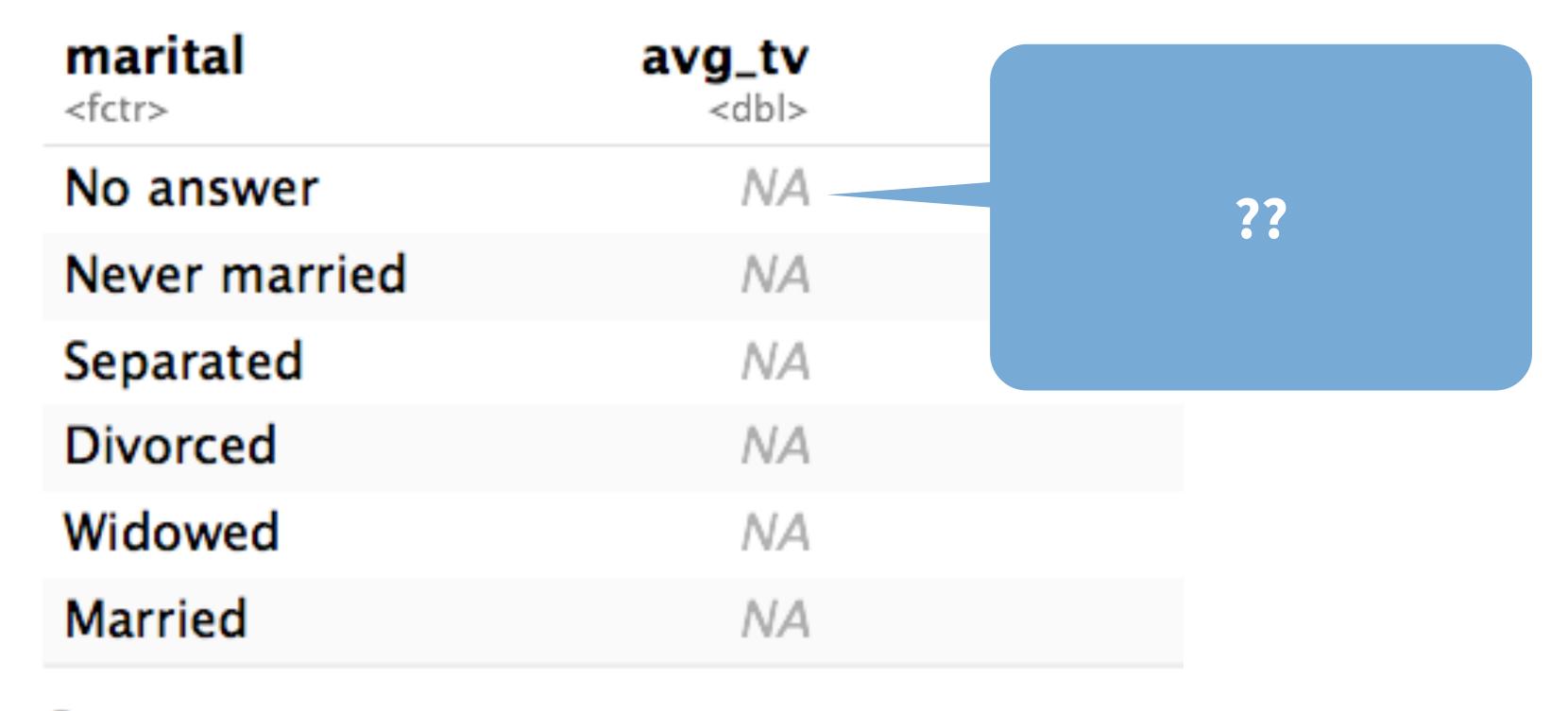


Warm-up/Review

Using the data gss_cat, find the average hours of tw watched (tvhours) for each category of marital status (marital).

(Don't worry if you get something unexpected, we'll fix it soon)

```
gss_cat %>%
  group_by(marital) %>%
  summarise(avg_tv = mean(tvhours))
```



6 rows

Missing values

Missing values propagate

```
NA + 2
## [1] NA
NA == NA
## [1] NA
mean(c(1, NA, 2))
```

is.na()

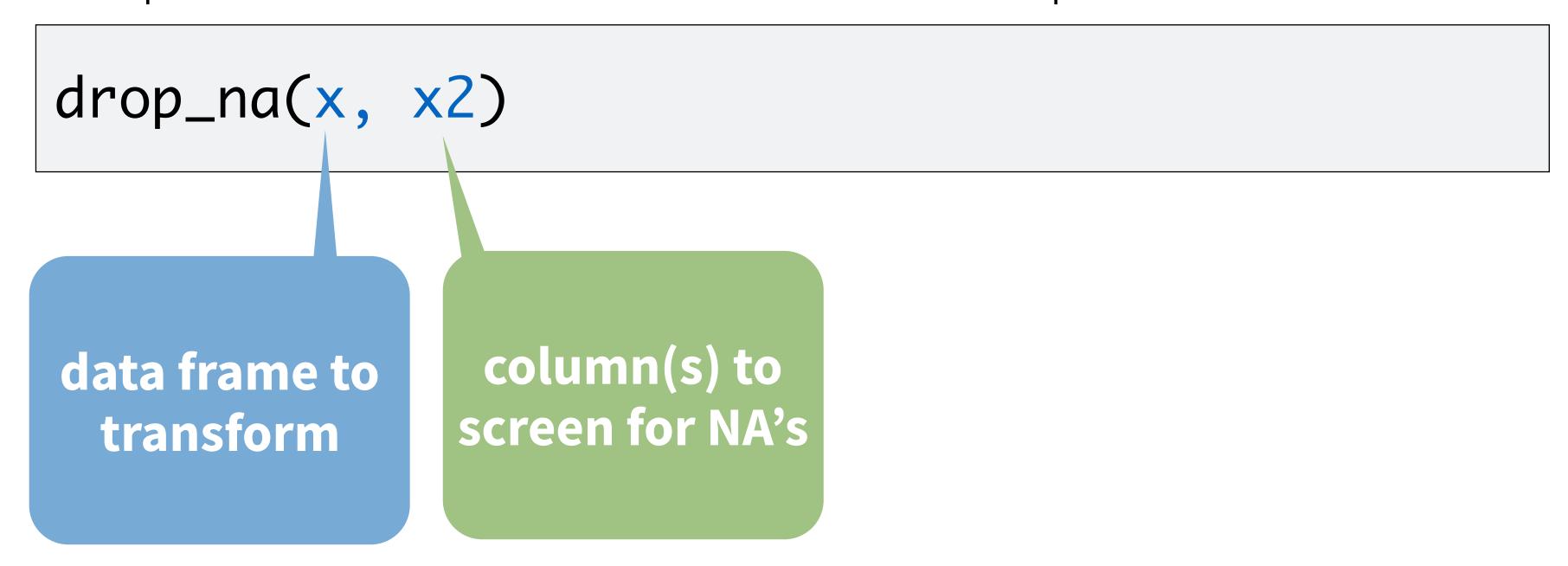
Returns TRUE if the value is NA

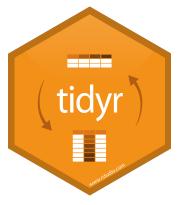
```
filter(gss_cat, tvhours == NA)
                                          # Nope!
## # A tibble: 0 x 9
filter(gss_cat, is.na(tvhours))
                                          # Yes!
# A tibble: 10,146 x 9
  year marital
            age race rincome partyid relig denom
  2000 Divorced 48 White $8000 ... Not str... Prote... Baptis...
  2000 Married
              25 White $20000... Strong ... Prote... Southe...
                                            NA
```



drop_na()

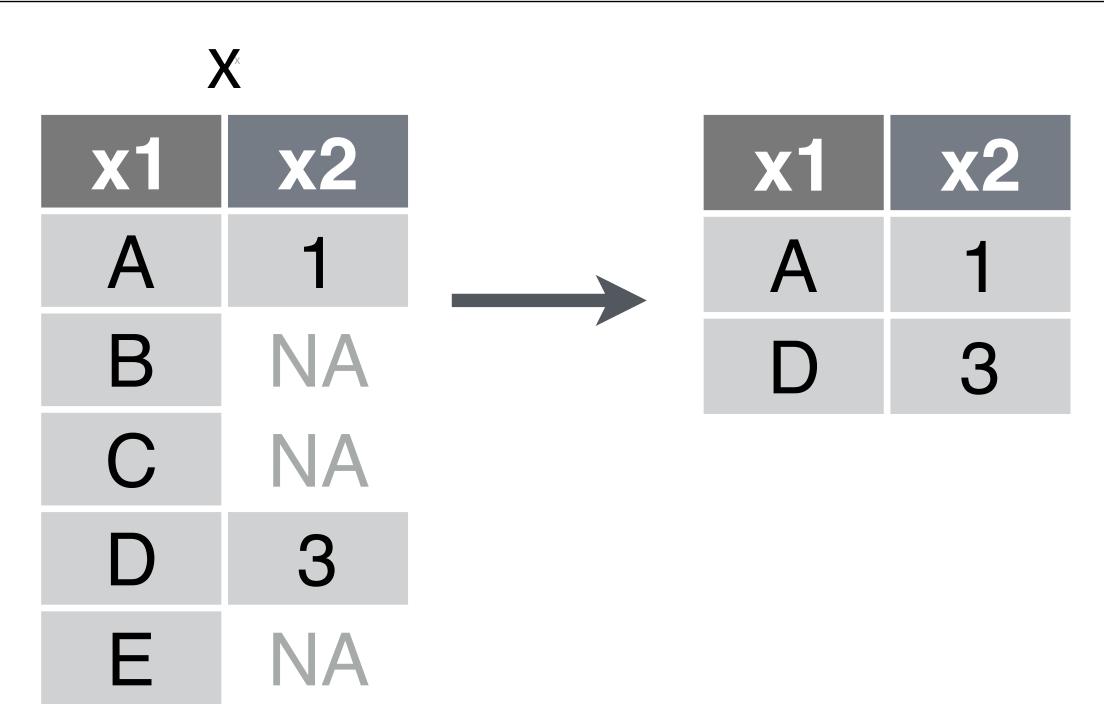
Drops rows that contain NA's in the specified columns.





drop_na()

Drops rows that contain NA's in the specified columns.





```
gss_cat %>%
  group_by(marital) %>%
  drop_na(tvhours) %>%
  summarise(avg_tv = mean(tvhours))
```

marital <fctr></fctr>	avg_tv <dbl></dbl>
No answer	2.55556
Never married	3.105175
Separated	3.549618
Divorced	3.085407
Widowed	3.912000
Married	2.650425

6 rows

Your Turn 1

What kind of object is the marital variable?

Brainstorm with your neighbor, all the things you know about that kind of object.

Factors



factors

R's representation of categorical data. Consists of:

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



factors

Stored as an integer vector with a levels attribute

```
unclass(eyes)
## 1 3 3
## attr(,"levels")
## "blue" "brown" "green"
```



forcats



Simple functions for working with factors.

library(tidyverse)

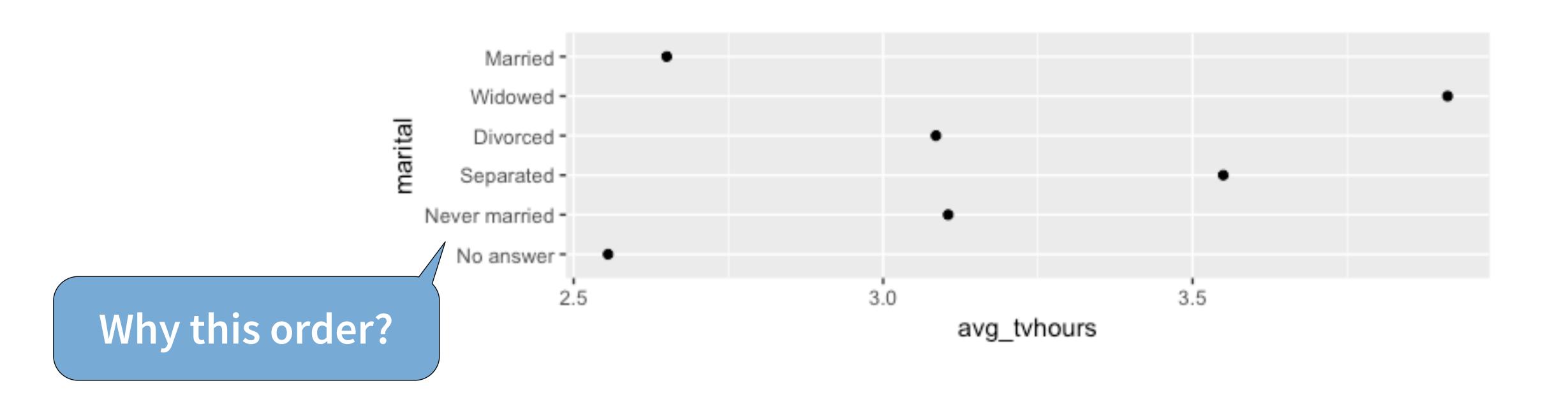
Your Turn 2

Fix your summary of average hours of tv watched (tvhours) by marital status (marital), to ignore missing values,

then create a plot to examine the results.



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



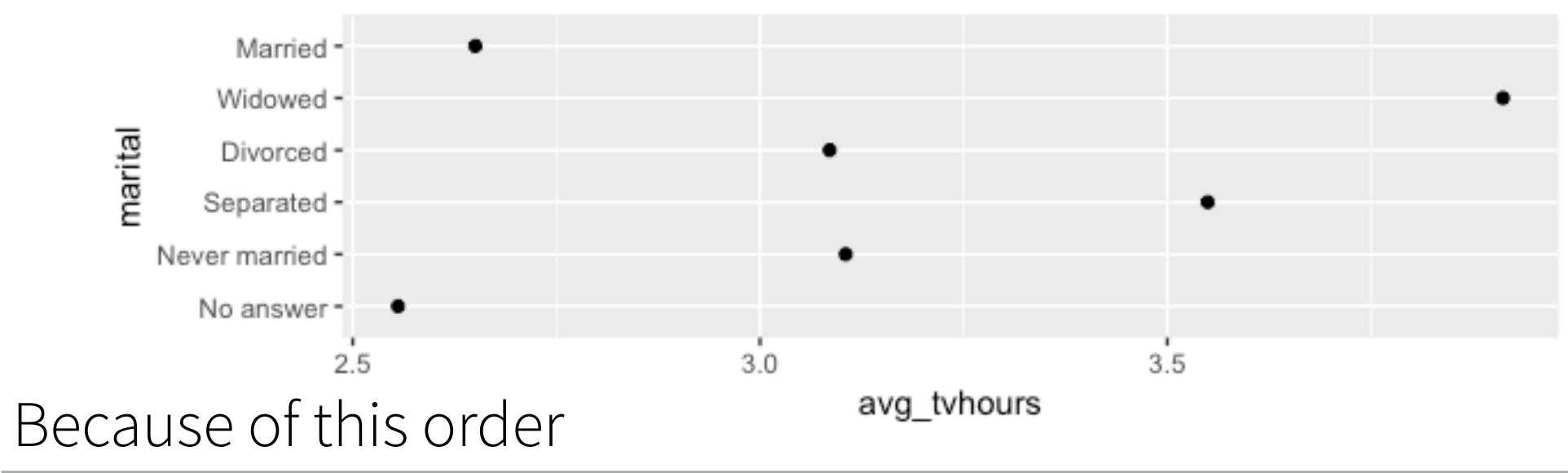
levels()

Use levels() to access a factor's levels

```
x %>% pull(y)
levels(eyes)
                                    same as
# [1] "blue" "brown" "green"
                                     x$y
gss_cat %>% pull(marital) %>% levels()
# [1] "No answer" "Never married" "Separated"
 [4] "Divorced" "Widowed"
                                    "Married"
```



Why this order?



```
gss_cat %>% pull(marital) %>% levels()
# [1] "No answer" "Never married" "Separated"
# [4] "Divorced" "Widowed" "Married"
```

Most useful skills

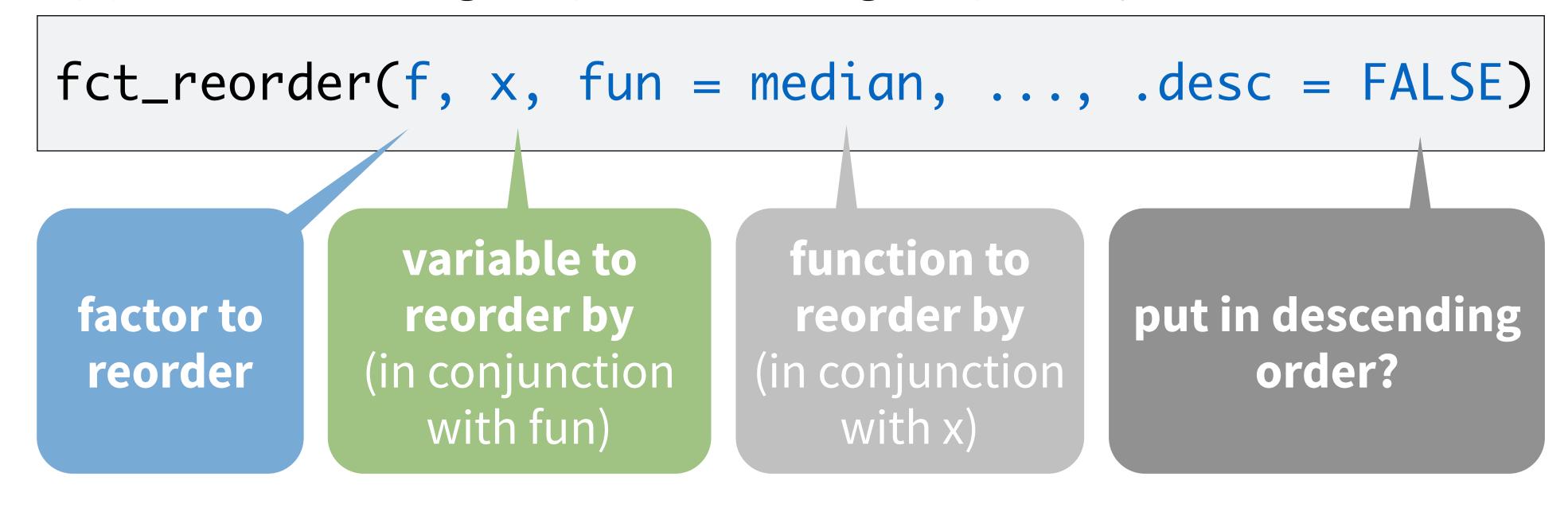
- 1. Reorder the levels
- 2. Manipulate the levels



Reordering levels

fct_reorder()

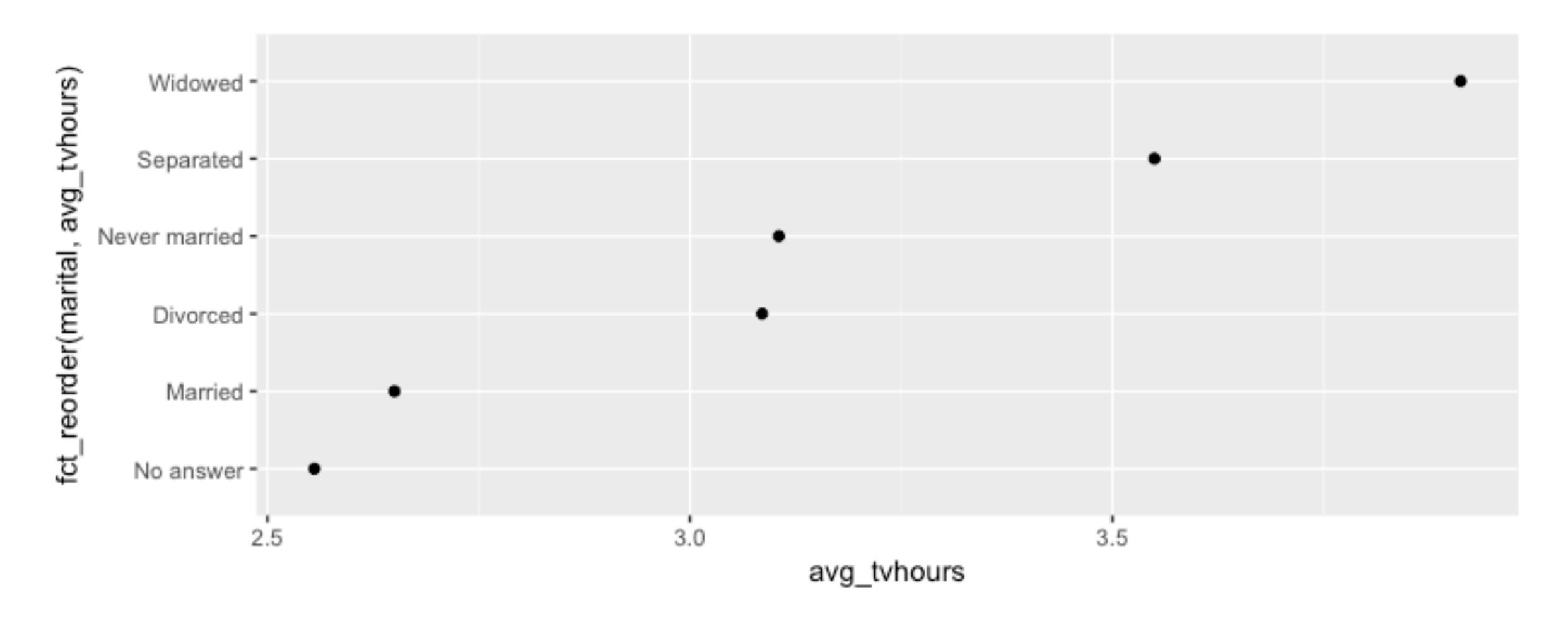
Reorders the levels of a factor based on the result of fun(x) applied to each group of cases (grouped by level).





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







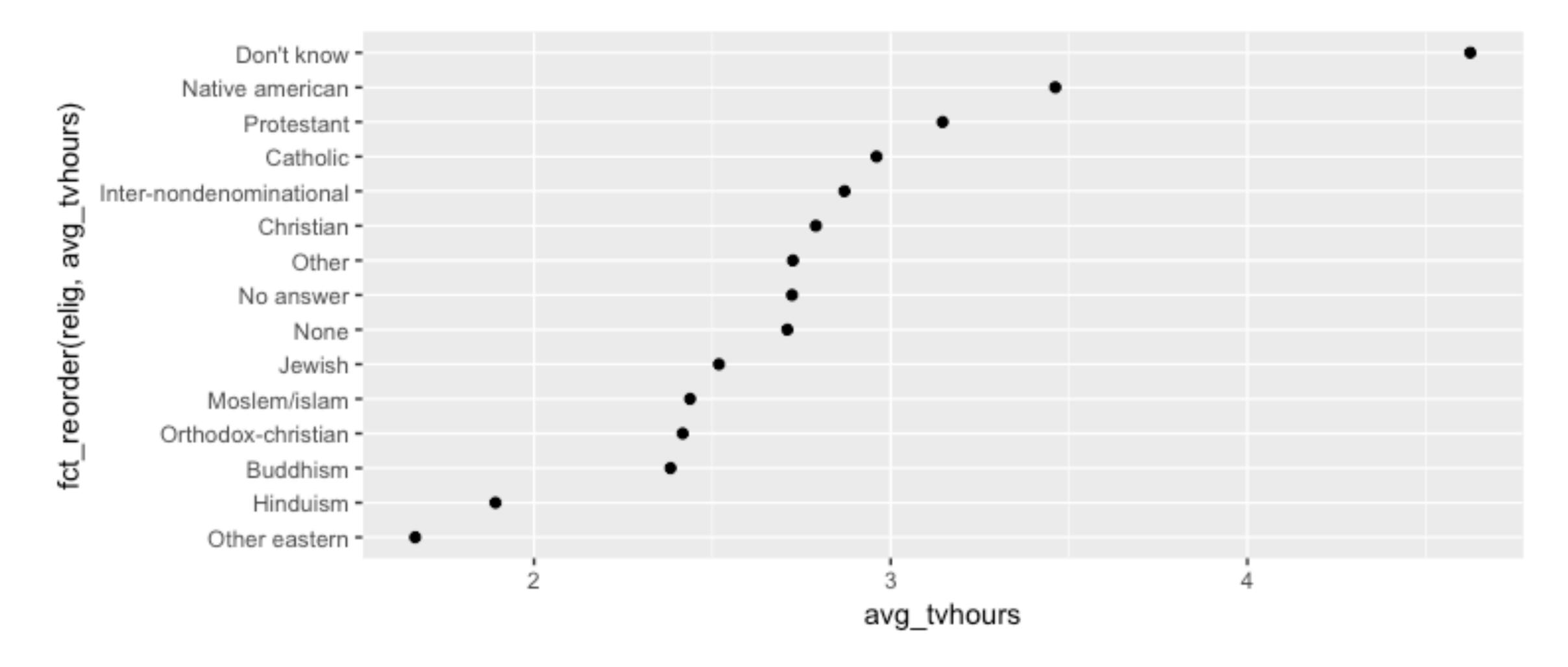
Your Turn 3

Fill in the blanks (____) to explore the average hours of tw watched by religion.

```
gss_cat %>%
  drop_na(___) %>%
  group_by(___) %>%
  summarise(___) %>%
  ggplot() +
    geom_point(mapping = aes(x = ___, y = ___))
```

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







Other reordering functions

Values and level labels are unchanged

fct_shuffle() Randomize order

fct_relevel() Order "by hand"

fct_infreq()

Order from most to least frequent

fct_inorder()

Order from first to last observed

fct_rev()

Reverse the current order

fct_shift()

Shift the order by 1

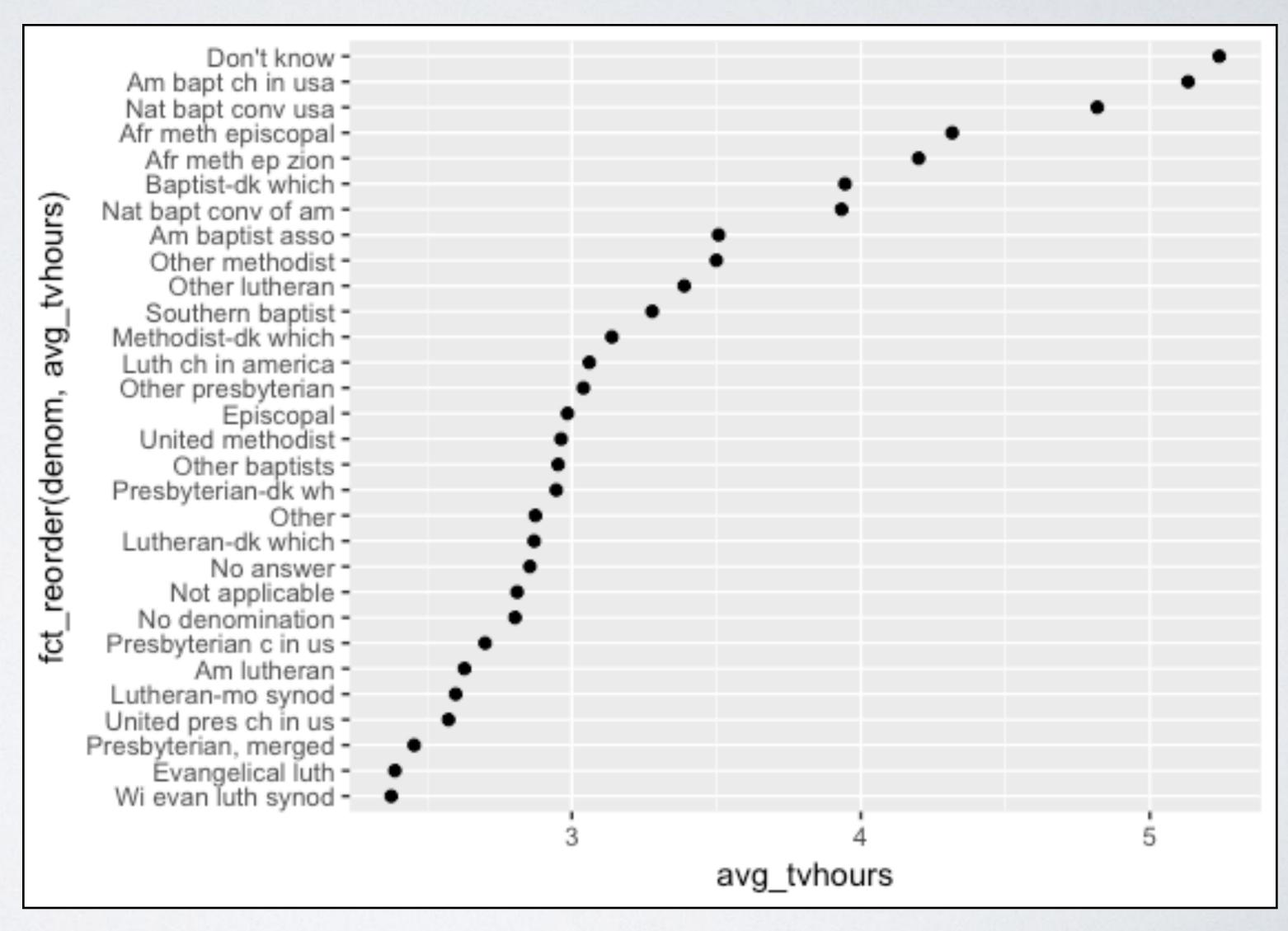
All forcats functions start with fct_



Manipulating levels

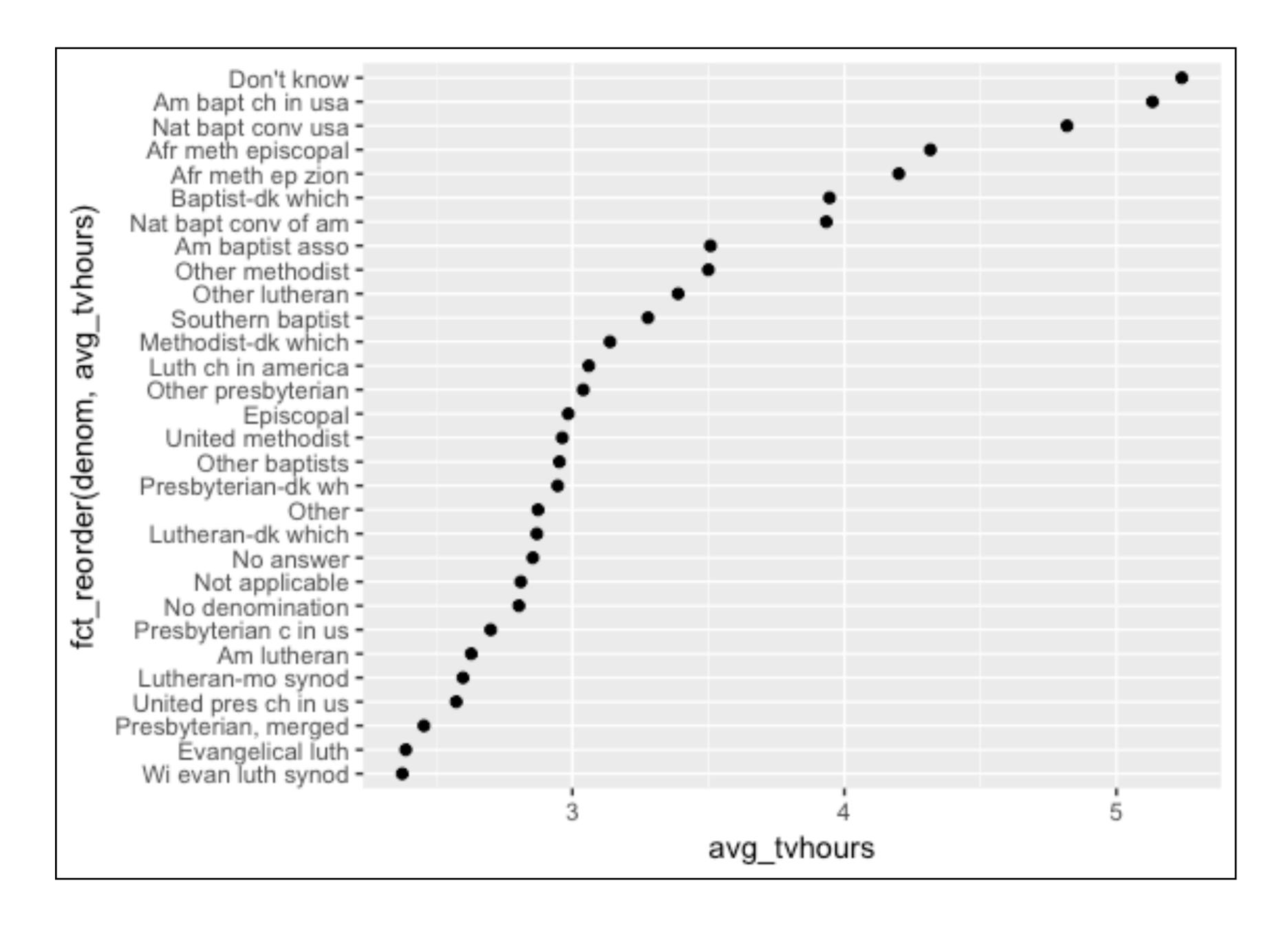
Quiz

Why is this plot not very useful?



Too many categories

Poorly labelled



		Relabel levels	Reduce levels				
First 10 rows							
Obs.	denom	denom	denom				
1	Southern baptist	Baptist - Southern	Baptist				
2	Baptist-dk which	Baptist - Don't know	Baptist				
3	No denomination	No denomination	None				
4	Not applicable	Not applicable	None				
5	Not applicable	Not applicable	None				
6	Southern baptist	Baptist - Southern	Baptist				
7	Not applicable	Not applicable	None				
8	Lutheran-mo synod	Lutheran - Missouri Synod	Lutheran				
9	Other	Other	Other				
10	Southern baptist	Baptist - Southern	Baptist				
• • •	•••	• • •	• • •				

Level manipulation functions

Values change to match levels

Relabel fct_anon()
Relabel levels "by hand"
fct_anon()
Anonymize levels
Relabel using a function
fct_collapse()
Collapse levels "by hand"
fct_lump()
Lump levels with small collapse
fct_other()
Replace levels with "Other

Collapse levels "by hand"

Lump levels with small counts together

Replace levels with "Other"



fct_recode()

Changes values of levels

```
fct_recode(f, ...)

factor with levels

new level = old level pairs
```



```
gss_cat %>%
pull(denom) %>%
levels()
```

```
"Don't know"
    "No answer"
    "No denomination"
                            "Other"
                            "Presbyterian-dk wh"
    "Episcopal"
    "Presbyterian, merged" "Other presbyterian"
    "United pres ch in us" "Presbyterian c in us"
    "Lutheran-dk which"
                            "Evangelical luth"
[13] "Other lutheran"
                            "Wi evan luth synod"
                            "Luth ch in america"
[15] "Lutheran-mo synod"
                            "Methodist-dk which"
    "Am lutheran"
                            "United methodist"
[19] "Other methodist"
                            "Afr meth episcopal"
    "Afr meth ep zion"
    "Baptist-dk which"
                            "Other baptists"
[25] "Southern baptist"
                            "Nat bapt conv usa"
    "Nat bapt conv of am"
                            "Am bapt ch in usa"
    "Am baptist asso"
                            "Not applicable"
```



```
gss_cat %>%
mutate(denom = fct_recode(denom,
    "Baptist - Southern" = "Southern baptist")
) %>%
pull(denom) %>%
levels()
```

[1]	"No answer"	"Don't know"
[3]	"No denomination"	"Other"
[5]	"Episcopal"	"Presbyterian-dk wh"
[7]	"Presbyterian, merged"	"Other presbyterian"
[9]	"United pres ch in us"	"Presbyterian c in us"
[11]	"Lutheran-dk which"	"Evangelical luth"
[13]	"Other lutheran"	"Wi evan luth synod"
[15]	"Lutheran-mo synod"	"Luth ch in america"
[17]	"Am lutheran"	"Methodist-dk which"
[19]	"Other methodist"	"United methodist"
[21]	"Afr meth ep zion"	"Afr meth episcopal"
[23]	"Baptist-dk which"	"Other baptists"
[25]	"Baptist - Southern"	"Nat bapt conv usa"
[27]	"Nat bapt conv of am"	"Am bapt ch in usa"
[29]	"Am baptist asso"	"Not applicable"



factor with levels

```
gss_cat %>%
mutate(denom = fct_recode(denom,
    "Baptist - Southern" = "Southern baptist")
) %>%
pull(denom) %>%
levels()
```

new level = old level pairs

[1]	"No answer"	"Don't know"
[3]	"No denomination"	"Other"
[5]	"Episcopal"	"Presbyterian-dk wh"
[7]	"Presbyterian, merged"	"Other presbyterian"
[9]	"United pres ch in us"	"Presbyterian c in us"
[11]	"Lutheran-dk which"	"Evangelical luth"
[13]	"Other lutheran"	"Wi evan luth synod"
[15]	"Lutheran-mo synod"	"Luth ch in america"
[17]	"Am lutheran"	"Methodist-dk which"
[19]	"Other methodist"	"United methodist"
[21]	"Afr meth ep zion"	"Afr meth episcopal"
[23]	"Baptist-dk which"	"Other baptists"
[25]	"Baptist - Southern"	"Nat bapt conv usa"
[27]	"Nat bapt conv of am"	"Am bapt ch in usa"
[29]	"Am baptist asso"	"Not applicable"



Your Turn 4

Edit the code to also relabel some other Baptist denominations:

- "Baptist-dk which"
- "Other baptists"

```
gss_cat %>%
 mutate(denom = fct_recode(denom,
    "Baptist - Southern" = "Southern baptist",
    "Baptist - Don't know" = "Baptist-dk which",
    "Baptist - Other" = "Other baptists")
 ) %>%
  pull(denom) %>%
  levels()
```

```
gss_cat %>%
  mutate(denom = fct_recode(denom,
    "Baptist - Southern" = "Southern baptist",
    "Baptist-dk which" = "Baptist - Don't know",
    "Baptist - Other" = "Other baptists")
# Unknown levels in `f`: Baptist - Don't know
```

Common mistake

Whoops, around the wrong way!

Message, but no warning or error!

Your Turn 5

What does the function detect_denom() do?

gss_cat %>% pull(denom) %>% levels()

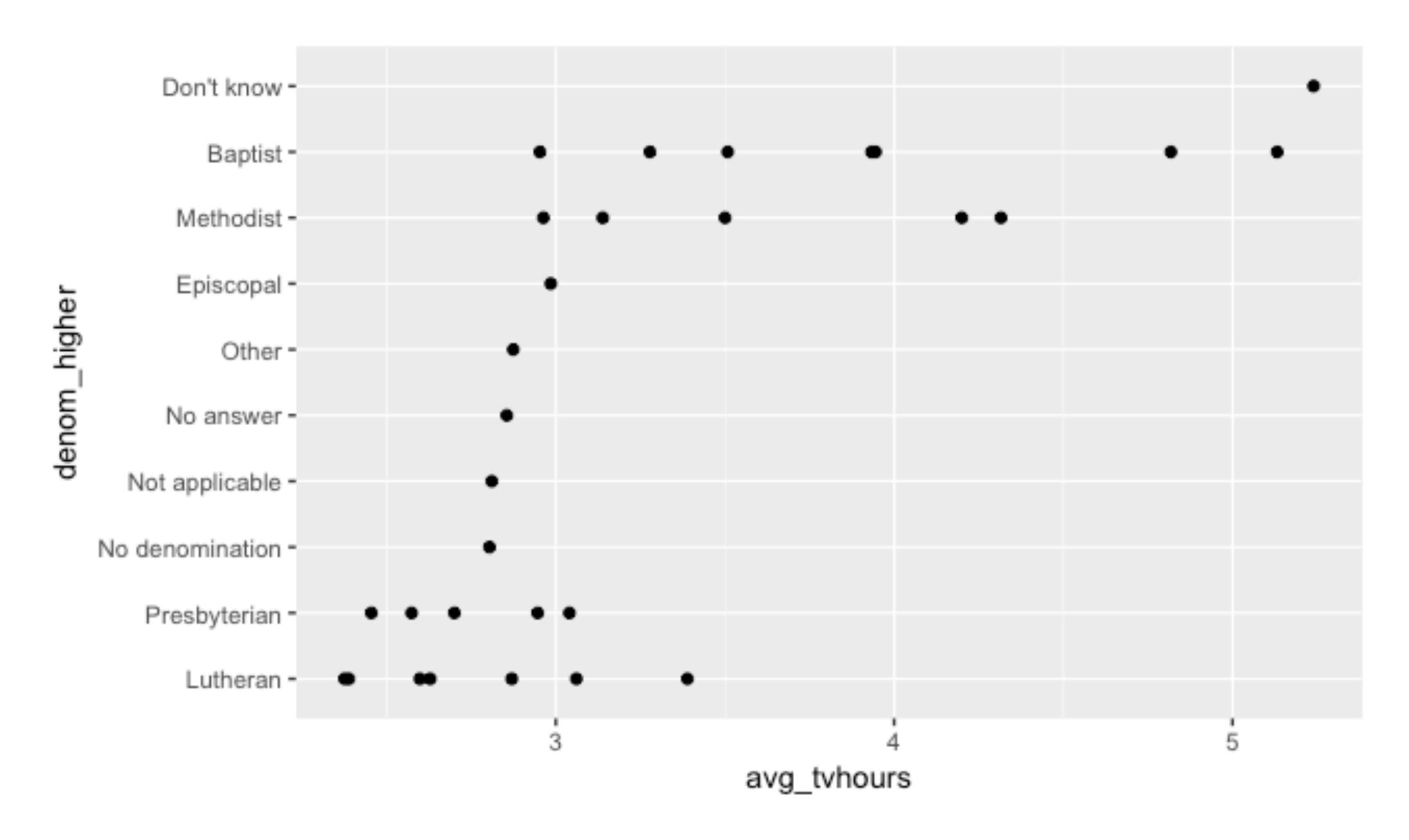
```
[1] "No answer"
                           "Don't know"
 [3] "No denomination"
                           "Other"
 [5] "Episcopal"
                          "Presbyterian-dk wh"
 [7] "Presbyterian, merged" "Other presbyterian"
 [9] "United pres ch in us" "Presbyterian c in us"
[11] "Lutheran-dk which" "Evangelical luth"
[13] "Other lutheran" "Wi evan luth synod"
                           "Luth ch in america"
[15] "Lutheran-mo synod"
                           "Methodist-dk which"
[17] "Am lutheran"
                           "United methodist"
[19] "Other methodist"
[21] "Afr meth ep zion"
                           "Afr meth episcopal"
                           "Other baptists"
[23] "Baptist-dk which"
[25] "Southern baptist"
                           "Nat bapt conv usa"
[27] "Nat bapt conv of am" "Am bapt ch in usa"
[29] "Am baptist asso"
                           "Not applicable"
```

gss_cat %>% pull(denom) %>% levels() %>% detect_denom()

[1]	"No answer"	"Don't know"	
[3]	"No denomination"	"Other"	
[5]	"Episcopal"	"Presbyterian"	
[7]	"Presbyterian"	"Presbyterian"	
[9]	"Presbyterian"	"Presbyterian"	converts to a
[11]	"Lutheran"	"Lutheran"	
[13]	"Lutheran"	"Lutheran"	higher level
[15]	"Lutheran"	"Lutheran"	grouping
[17]	"Lutheran"	"Methodist"	Sioupins
[19]	"Methodist"	"Methodist"	
[21]	"Methodist"	"Methodist"	
[23]	"Baptist"	"Baptist"	
[25]	"Baptist"	"Baptist"	
[27]	"Baptist"	"Baptist"	
[29]	"Baptist"	"Not applicable"	

Use with fct_relabel() to collapse levels

```
gss_cat %>%
  drop_na(tvhours) %>%
  mutate(denom_higher = fct_relabel(denom, detect_denom) %>%
      fct_reorder(tvhours, mean)) %>%
  group_by(denom_higher, denom) %>%
  summarise(avg_tvhours = mean(tvhours)) %>%
  ggplot() +
    geom_point(mapping = aes(x = avg_tvhours,
      y = denom_higher))
```



Strings



(character) strings

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

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

stringr



Simple functions for working with character **strings**.

library(tidyverse)

Most useful skills

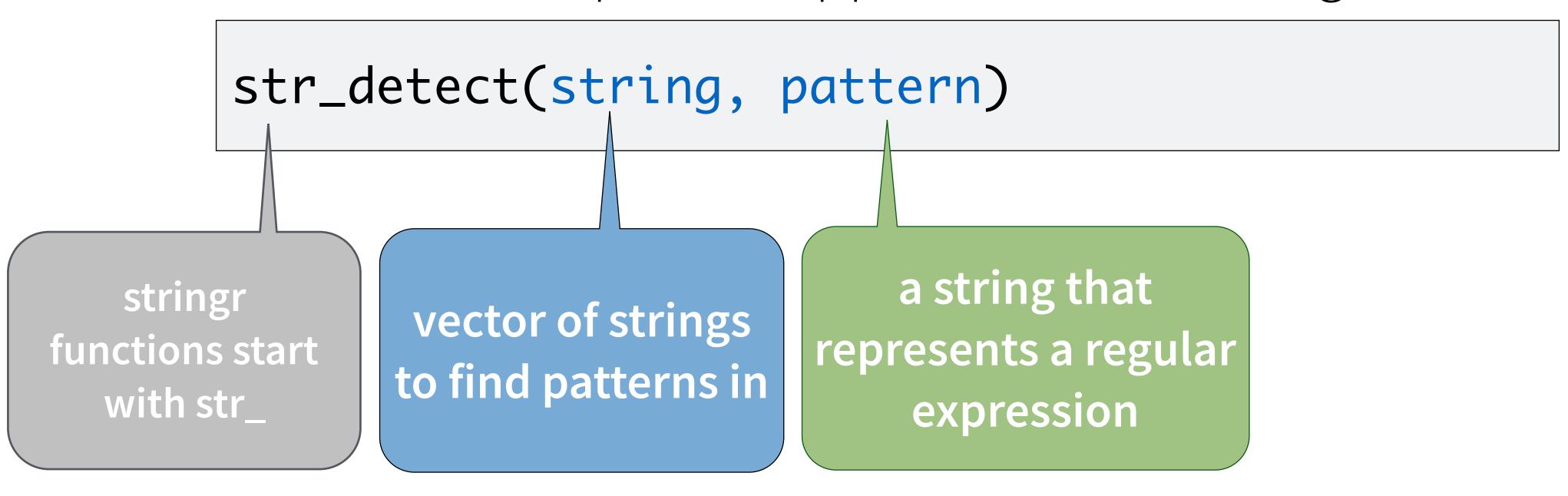
- 1. How to extract/ replace substrings
- 2. How to find matches for patterns
- 3. Regular expressions

```
detect_denom <- function(x){</pre>
  case_when(
    str_detect(x, "[Bb]ap") ~ "Baptist",
    str_detect(x, "[Pp]res") ~ "Presbyterian",
    str_detect(x, "[L1]uth") ~ "Lutheran",
    str_detect(x, "[Mm]eth") ~ "Methodist",
    TRUE ~ X
```

```
detect_denom <- function(x){</pre>
  case_when(
    str_detect(x, "[Bb]ap") ~ "Baptist",
    str_detect(x, "[Pp]res") ~ "Presbyterian",
    str_detect(x, "[L1]uth") ~ "Lutheran",
    str_detect(x, "[Mm]eth") ~ "Methodist",
    TRUE ~ X
```

str_detect()

Test whether a pattern appears within a string.



Returns TRUE or FALSE



Your Turn 6

```
strings <- c("Apple", "Pineapple", "Orange")</pre>
```

With your neighbor, predict what these might return:

```
str_detect(strings, pattern = "pp")
str_detect(strings, pattern = "apple")
str_detect(strings, pattern = "[Aa]pple")
```

Then run them!

```
str_detect(strings, pattern = "pp")
# [1] TRUE TRUE FALSE
```

Apple
Pineapple
Orange

```
str_detect(strings, pattern = "apple")
# [1] FALSE TRUE FALSE
```

Apple
Pineapple
Orange

Apple
Pineapple
Orange

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.

<u>case</u> Convert case of a string.

collControl matching behaviour with modifier functions.fixedControl matching behaviour with modifier functions.

<u>fruit</u>
Sample character vectors for practicing string manipulations.

<u>invert_match</u>
<u>modifiers</u>

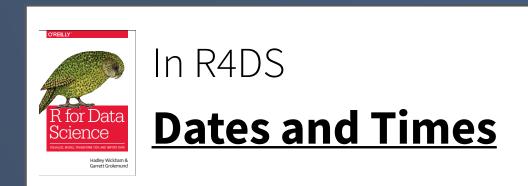
Switch location of matches to location of non-matches.

Control matching behaviour with modifier functions.

Control matching behaviour with modifier functions.



Dates and times



Most useful skills

- 1. Creating dates/times (i.e. parsing)
- 2. Access parts of a date
- 3. Deal with time zones
- 4. Do math with instants and time spans



Creating dates and times

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()

ydm_hms(), ydm_hm(), ydm_h()

dmy_hms(), dmy_hm(), dmy_h()

mdy_hms(), mdy_hm(), mdy_h()

POSIXct

ymd(), ydm(), mdy()

myd(), dmy(), dym(), yq()

Date (POSIXct if tz specified)

hms(), hm(), ms()

Period



Your Turn 7

For each of the following formats (of the same date), pick the right ymd() function to parse them:

- "2018 Feb 02"
- "2-1-18"
- "01/02/2018"

```
ymd("2018 Feb 02")
# [1] "2018-02-02"
mdy("2-1-18")
# [1] "2018-02-01"
dmy("01/02/2018")
# [1] "2018-02-01"
```

Extract components by name with a singular name

```
date <- ymd("2018-02-01")
year(date)
## 2018</pre>
```



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	



```
wday(ymd("2018-02-01"))
## 5
wday(ymd("2018-02-01"), label = TRUE)
# [1] Thu
# Levels: Sun < Mon < Tue < Wed < Thu < Fri < Sat
wday(ymd("2018-02-01"), label = TRUE, abbr = FALSE)
# [1] Thursday
# 7 Levels: Sunday < Monday < Tuesday < Wednesday < ... < Saturday
```



births

Two variables from yesterday's data

	date	births	
	<date></date>	<int></int>	
	1994-01-01	8096	
	1994-01-02	7772	
	1994-01-03	10142	
	1994-01-04	11248	
	1994-01-05	11053	
	1994-01-06	11406	
	1994-01-07	11251	
	1994-01-08	8653	
	1994-01-09	7910	
	1994-01-10	10498	
1-10 of	3,652 rows		Previous 1 2 3 4 5 6 100 Next

Adapted from 'Master the tidyverse' CC by RStudio

Your Turn 8

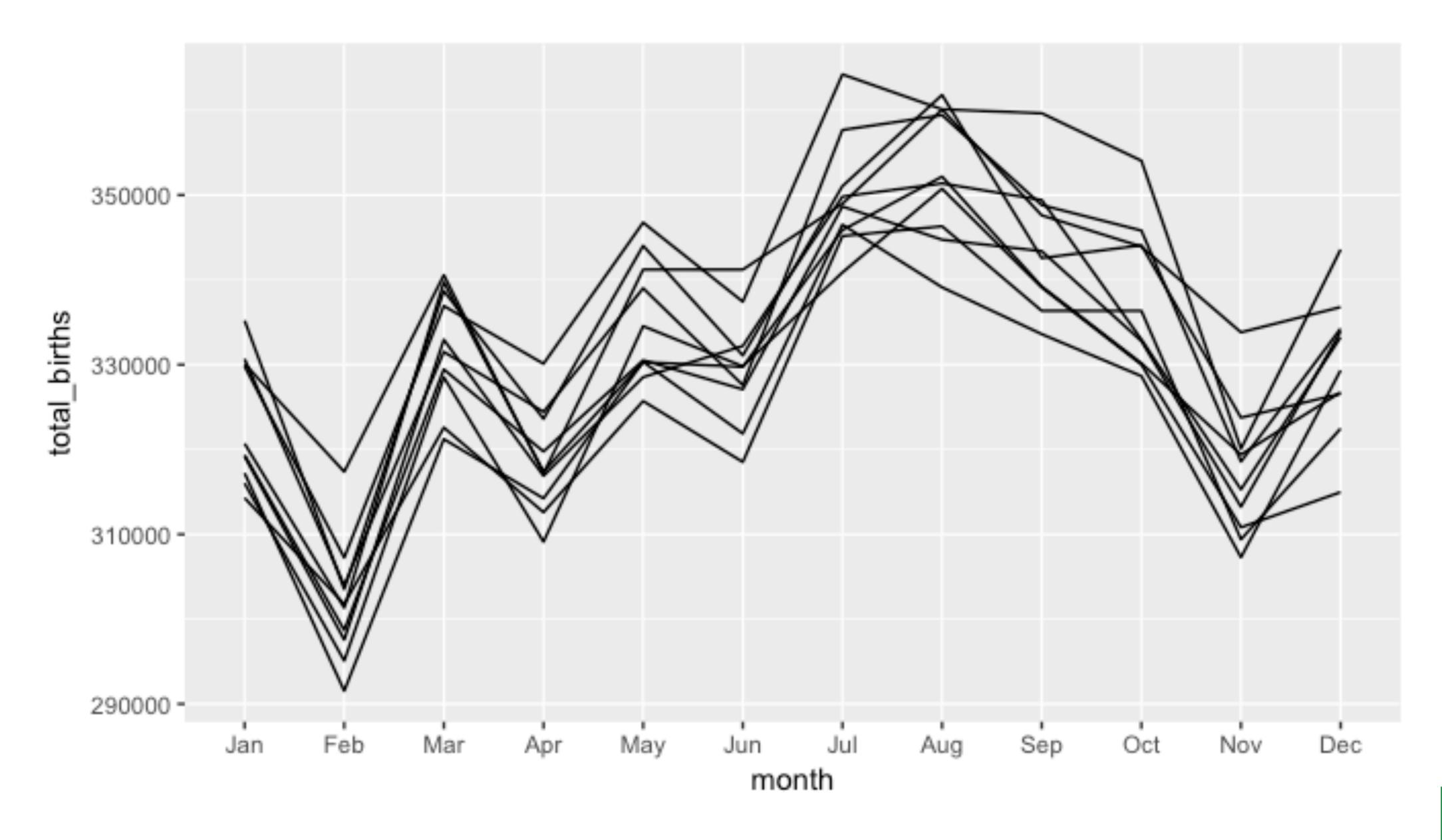
Fill in the blanks to:

- Extract the month from date.
- Extract the year from date.
- Find the total births for each year/month.
- Plot the results as a line chart.



```
births %>%
  mutate(year = year(date),
   month = month(date, label = TRUE)) %>%
  group_by(year, month) %>%
  summarise(total_births = sum(births)) %>%
  ggplot() +
    geom_line(aes(x = month, y = total_births, group = year))
```







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
```



* on a typical day

Data Types

