Lecture 19: Factors

STAT 450, Fall 2021

Reading: Chapter 15 from R for Data Science

Useful reference: https://forcats.tidyverse.org/

In R, factors are used to work with categorical variables. Recall that a categorical variable takes on values that fall into distinct categories. Some examples are gender, education level, and political affiliation.

We will use the forcats package, which is a core tidyverse package that provides tools for working with factors.

library(tidyverse)

```
## -- Attaching packages -----
                                              ----- tidyverse 1.3.1 --
## v ggplot2 3.3.5
                    v purrr
                             0.3.4
## v tibble 3.1.3
                    v dplyr
                             1.0.7
## v tidyr
           1.1.3
                    v stringr 1.4.0
## v readr
                    v forcats 0.5.1
           2.0.1
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                  masks stats::lag()
```

Creating Factors

```
x <- c("low", "medium", "high", "low", "high")
class(x)</pre>
```

[1] "character"

By default, R sorts alphabetically, which is not that useful here.

sort(x)

```
## [1] "high" "high" "low" "low" "medium" "medium"
```

We can use the factor() function to convert the character vector \mathbf{x} into a factor. The ordering of the categories is specified by the levels argument.

```
y <- factor(x, levels = c("low", "medium", "high"))
class(y)</pre>
```

```
## [1] "factor"
sort(y)
```

```
## [1] low low medium medium high high
## Levels: low medium high
```

General Social Survey

The gss_cat data frame contains data from the General Social Survey, which is a long-running sociological survey conducted by the National Opinion Research Center at the University of Chicago.

gss_cat

```
## # A tibble: 21,483 x 9
##
       year marital
                            age race rincome
                                                      partyid relig denom
                                                                             tvhours
##
      <int> <fct>
                          <int> <fct> <fct>
                                                      <fct>
                                                               <fct>
                                                                      <fct>
                                                                                <int>
       2000 Never married
##
                             26 White $8000 to 9999
                                                      Ind, near Proter South
                                                                                   12
   1
                                                     Not str~ Prote~ Bapti~
##
       2000 Divorced
                             48 White $8000 to 9999
                                                                                   NA
##
   3
       2000 Widowed
                             67 White Not applicable Indepen~ Prote~ No de~
                                                                                    2
##
       2000 Never married
                             39 White Not applicable Ind, nea~ Ortho~ Not a~
                                                                                    4
       2000 Divorced
##
   5
                             25 White Not applicable Not str~ None
                                                                                    1
                                                                      Not a~
                             25 White $20000 - 24999 Strong ~ Prote~ South~
##
   6
       2000 Married
                                                                                   NA
##
   7
       2000 Never married
                             36 White $25000 or more Not str~ Chris~ Not a~
                                                                                    3
##
       2000 Divorced
                             44 White $7000 to 7999 Ind, near Proter Luther
                                                                                   NA
##
   9
       2000 Married
                             44 White $25000 or more Not str~ Prote~ Other
                                                                                    0
## 10
       2000 Married
                             47 White $25000 or more Strong ~ Prote~ South~
                                                                                    3
## # ... with 21,473 more rows
```

Consider the categorical variable race, which is represented as a factor. The levels() function can be used to extract the different categories:

```
levels(gss_cat$race)
```

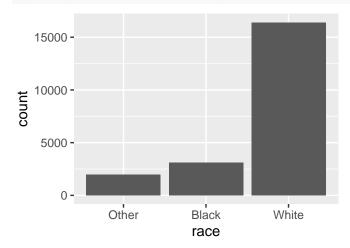
```
## [1] "Other" "Black" "White" "Not applicable"
```

We can also make a table and bar plot displaying the number of respondents in each race category:

```
gss_cat %>%
  count(race)
```

```
## # A tibble: 3 x 2
## race n
## <fct> <int>
## 1 Other 1959
## 2 Black 3129
## 3 White 16395
```

```
ggplot(gss_cat, aes(race)) + geom_bar()
```



By default ggplot2 will drop any levels that don't have values. That's why Not applicable does not show up in the bar plot.

Exercise 1: Explore the distribution of rincome (reported income). What makes the default bar plot difficult to understand? How could you improve the plot?

Modifying factor order

It is often useful to change the order of the levels of a factor for data visualization.

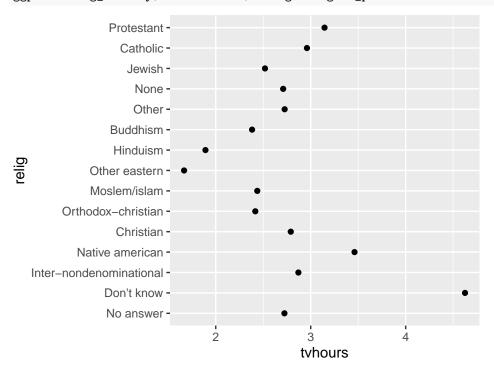
fct_reorder()

Suppose we want to explore the average number of hours per day spent watching TV across religions:

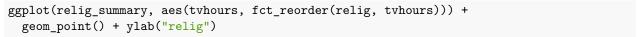
```
relig_summary <- gss_cat %>%
  group_by(relig) %>%
  summarise(
    n = n(),
    tvhours = mean(tvhours, na.rm = TRUE)
)
relig_summary
```

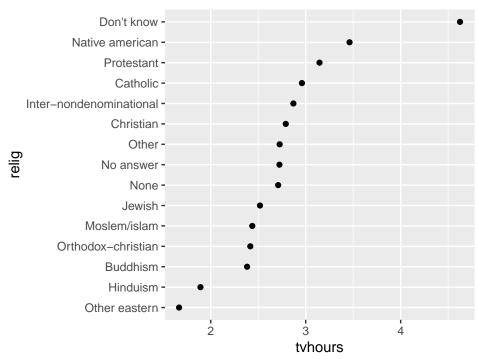
```
## # A tibble: 15 x 3
                                  n tvhours
##
      relig
##
      <fct>
                              <int>
                                      <dbl>
   1 No answer
                                 93
                                       2.72
##
  2 Don't know
                                 15
                                       4.62
                                109
                                       2.87
##
   3 Inter-nondenominational
## 4 Native american
                                 23
                                       3.46
## 5 Christian
                                689
                                       2.79
## 6 Orthodox-christian
                                 95
                                       2.42
##
  7 Moslem/islam
                                104
                                       2.44
                                       1.67
## 8 Other eastern
                                 32
## 9 Hinduism
                                 71
                                       1.89
## 10 Buddhism
                                147
                                       2.38
## 11 Other
                                224
                                       2.73
## 12 None
                               3523
                                       2.71
## 13 Jewish
                                388
                                       2.52
## 14 Catholic
                               5124
                                       2.96
## 15 Protestant
                              10846
                                       3.15
```

ggplot(relig_summary, aes(tvhours, relig)) + geom_point()



We can improve this plot by using the function fct_reorder() to reorder the levels of relig according to the average number of hours per day spent watching TV:



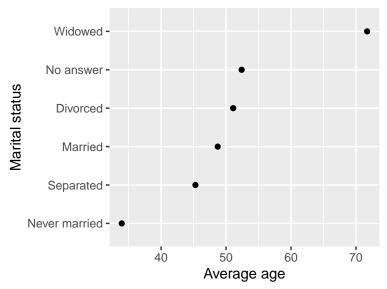


The following code which uses mutate() will create the same graph:

```
relig_summary %>%
  mutate(relig = fct_reorder(relig, tvhours)) %>%
ggplot(aes(tvhours, relig)) +
  geom_point()
```

This code is somewhat easier to read in my view, since it clearly shows the two steps here: (1) reorder the levels of relig according to tvhours, the average number of hours spent watching tv; and (2) make a plot with tvhours on the x-axis and relig on the y-axis.

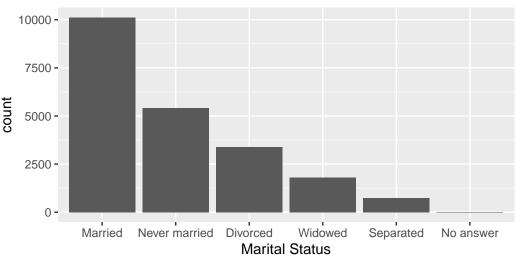
Exercise 2: Use group_by() and summarise() to compute the average age for each category of marital. Call the data frame with the grouped summaries marital_summary. Then recreate the R code that makes the graph below.



fct_infreq()

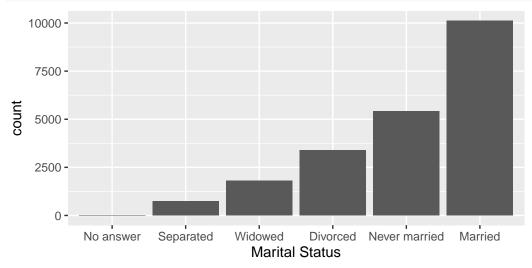
fct_infreq() orders the levels of a factor by frequency (number of observations in each level).

```
ggplot(gss_cat, aes(fct_infreq(marital))) +
geom_bar() + xlab("Marital Status")
```



You can combine with fct_rev() to reverse the order.

```
ggplot(gss_cat, aes(fct_rev(fct_infreq(marital)))) +
  geom_bar() + xlab("Marital Status")
```



The following code which uses mutate() and the %>% pipe operator will create the same graph:

```
gss_cat %>%
  mutate(marital = marital %>% fct_infreq() %>% fct_rev()) %>%
  ggplot(aes(marital)) +
    geom_bar()
```

Modifying factor levels

fct_recode()

Use fct_recode() to change the names of the levels. For example:

```
gss_cat %>% count(partyid)
```

```
## # A tibble: 10 x 2
##
     partyid
                            n
##
      <fct>
                         <int>
## 1 No answer
                           154
## 2 Don't know
                            1
## 3 Other party
                          393
## 4 Strong republican
                          2314
## 5 Not str republican 3032
                          1791
## 6 Ind, near rep
## 7 Independent
                         4119
## 8 Ind, near dem
                          2499
## 9 Not str democrat
                          3690
## 10 Strong democrat
                          3490
```

The levels for partyid are inconsistent, and can be improved.

```
gss_cat %>%
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"
)) %>%
count(partyid)
```

```
## # A tibble: 10 x 2
##
     partyid
                               n
##
      <fct>
                           <int>
## 1 No answer
                             154
## 2 Don't know
                               1
## 3 Other party
                             393
## 4 Republican, strong
                            2314
## 5 Republican, weak
                            3032
## 6 Independent, near rep 1791
## 7 Independent
                            4119
## 8 Independent, near dem 2499
## 9 Democrat, weak
                            3690
## 10 Democrat, strong
                            3490
```

 ${\tt fct_recode}()$ will not alter any levels that aren't explicitly mentioned.

fct_collapse()

Use fct_collapse() to combine levels. For example:

```
gss_cat %>%
  mutate(partyid = fct_collapse(partyid,
      other = c("No answer", "Don't know", "Other party"),
  rep = c("Strong republican", "Not str republican"),
  ind = c("Ind,near rep", "Independent", "Ind,near dem"),
  dem = c("Not str democrat", "Strong democrat")
)) %>%
  count(partyid)
```

```
## # A tibble: 4 x 2
## partyid n
## <fct> <int>
## 1 other 548
## 2 rep 5346
## 3 ind 8409
## 4 dem 7180
```