## Exercise 5: Functions

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9/24/2020

## **Prerequisites**

- 1. Open you ps811-exercises folder.
- 2. Go to File > New File > R Script.
- 3. Save file as exercise-6-functions.R.
- 4. Load R packages installed in Lectures 5 and 6.

```
library(tidyverse)
library(here)
```

- 5. Download the national.csv file from Kaggle.
- 6. Load the CSV file into your R environment.

```
national <- read.csv(here("data/national.csv"))</pre>
```

7. View the CSV.

## Loops

1. Write a loop to find how many variables there are per observations.

```
rows_loop <- numeric()

for (i in 1:ncol(national)) {
  rows_loop[i] <- length(national[i, ])
}

rows_loop</pre>
```

2. Find the mean number of Protestant Christians in each country (i.e., the state column) using tapply() in both base R and Tidyverse.

```
# base R
tapply(
   X = national$christianity_protestant,
   INDEX = list(national$state),
```

```
FUN = mean,
  na.rm = TRUE
)

# tidyverse
national %>%
  group_by(state) %>%
  summarize(
   mean_christ= mean(christianity_protestant, na.rm = TRUE)
)
```

3. Check the column type for each variable that is just characters.

Hint: You can use sapply(dataset-name-here, class).

```
sapply(national, is.character)
```

4. Log the buddhism variables.

```
national %>%
  mutate_at(
    .vars = vars(starts_with("buddhism")),
    .funs = function(x) log(x)
) %>%
  select(starts_with("buddhism"))

# answers referring to only the "buddhism_all" variable are also acceptable
# I realize that this question is unclear (sorry!)
```

5. Write a function that lists all the unique years with more than 300,000 Christians in total.

```
unique_chr300k_yr <- function(x) {
    if(national$christianity_all > 300000) {
    return(unique(national$year))
    }
}
unique_chr300k_yr()
```

- ## Warning in if (national\$christianity\_all > 3e+05) {: the condition has length >
  ## 1 and only the first element will be used
- ## [1] 1945 1950 1955 1960 1965 1970 1975 1980 1985 1990 1995 2000 2005 2010
  - 6. Group by the code variable.

Hint: Use nesting.

```
nested_code <- national %>%
group_by(code) %>%
nest()
```

7. Create a model column. You need to include a linear regression model that examines whether the proportion of Jews (judaism\_percent) affect whether a country has dual religions.

```
nested_models <- nested_code %>%
  mutate(
  model = map(
    .x = data,
```

```
.f = ~ lm(dual_religion ~ judaism_percent, data = .x)
    )
  ) %>%
 print()
## # A tibble: 200 x 3
## # Groups:
               code [200]
##
      code data
                                model
##
      <chr> <list>
                                st>
##
   1 USA
            <tibble [14 x 78]> <lm>
##
   2 CAN
            <tibble [14 x 78] > <lm>
##
  3 BHM
            <tibble [8 x 78]> <lm>
## 4 CUB
            <tibble [14 x 78]> <lm>
            <tibble [14 x 78]> <lm>
## 5 HAI
## 6 DOM
            <tibble [14 x 78]> <lm>
## 7 JAM
            <tibble [11 x 78]> <lm>
## 8 TRI
            <tibble [11 x 78]> <lm>
## 9 BAR
            <tibble [9 \times 78]> <lm>
## 10 DMA
            <tibble [7 x 78]> <lm>
## # ... with 190 more rows
  8. Extract the coefficients into a new column
nested_coefs <- nested_models %>%
  mutate(coefs = map(model, coefficients)) %>%
  print()
## # A tibble: 200 x 4
## # Groups:
               code [200]
##
      code data
                                model
                                       coefs
##
      <chr> <list>
                                t> <list>
##
   1 USA
            <tibble [14 x 78]> <lm>
                                       <dbl [2]>
##
   2 CAN
            <tibble [14 x 78]> <lm>
                                       <dbl [2]>
##
  3 BHM
            <tibble [8 x 78]> <lm>
                                       <dbl [2]>
## 4 CUB
            <tibble [14 x 78]> <lm>
                                       <dbl [2]>
## 5 HAI
            <tibble [14 x 78]> <lm>
                                       <dbl [2]>
            <tibble [14 x 78]> <lm>
## 6 DOM
                                       <dbl [2]>
  7 JAM
            <tibble [11 x 78]> <lm>
                                       <dbl [2]>
## 8 TRI
            <tibble [11 x 78]> <lm>
                                       <dbl [2]>
## 9 BAR
            <tibble [9 x 78]> <lm>
                                       <dbl [2]>
## 10 DMA
            <tibble [7 \times 78]> <lm>
                                       <dbl [2]>
## # ... with 190 more rows
  9. Look at the coefficients.
nested_coefs$coefs
 10. Pull out the model column.
nested_coefs %>% pull(model)
 11. Unnest the coefficients column.
coefs <- nested_coefs %>%
  unnest(coefs) %>%
  print()
```

```
## # Groups: code [200]
##
     code data
                            model coefs
##
     <chr> <list>
                             t> <dbl>
## 1 USA
           <tibble [14 x 78]> <lm>
## 2 USA
           <tibble [14 x 78]> <lm>
           <tibble [14 x 78]> <1m>
## 3 CAN
                                     0
           <tibble [14 x 78]> <lm>
## 4 CAN
           <tibble [8 x 78]> <1m>
## 5 BHM
                                     0
           <tibble [8 x 78]> <lm>
## 6 BHM
                                     0
## 7 CUB
           <tibble [14 x 78]> <lm>
                                     1.00
## 8 CUB
           <tibble [14 x 78]> <lm>
## 9 HAI
           <tibble [14 x 78]> <lm>
                                     1.00
## 10 HAI
          <tibble [14 x 78]> <lm>
## # ... with 390 more rows
```