Data Science for Public Policy

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PPOL 670 | Assignment 02 tidyverse Basics

Due Date: Tuesday, February 1st at 11:59 PM

Deliverable: A single, well-documented .R script submitted to Canvas. The .R script should be called assignment02_<NetID>.R. Each exercise should be clearly labeled and the grader should be able to run each segment of R code.

Rubric

- Each exercise is clearly labeled with a comment. Use # to create a comments in your R script.
- Each answer runs successfully and provides a correct answer.
- Attempted exercises will receive more than zero points. Unattempted exercises will receive zero points.

Points: 5 points

Learning and data science are both collaborative practices. We encourage you to discuss class topics and homework topics with each other. However, the work you submit must be your own. A student should never see another student's code or receive explicit coding instructions for a homework problem. Please attend office hours or contact one of the instructors if you need help or clarification.

Plagiarism on homework or projects will be dealt with to the full extent allowed by Georgetown policy (see http://honorcouncil.georgetown.edu).

Setup

- 1. Create a new folder on your computer called assignment02.
- 2. Open R Studio.
- 3. In the top right, click "Project: (None)".
- 4. Click "New Project".
- 5. Create a new project in the existing directory assignment02.
- 6. In R Studio, open a .R script and save it at assignment02/assignment02_<NetID>.R.
- 7. Add your name and NetID as a comment at the top of the .R script.

The next few exercises will work with a data set that contains counts of PhDs awarded by field from 2008 to 2017. The data come from TidyTuesday—a weekly coding project that tackles a new data set each week in R. The documentation and original source are available here.

Exercise 01 (0.5 point)

- 1. Add library(tidyverse) to the top of your script.
- 2. Duplicate the following code in your script and run the code to download a .csv from this GitHub repo. phd_link <-

- 3. Load the .csv into R with read_csv() from library(readr) and assign it to phds. Do not use the point-and-click interface in RStudio.
- 4. Steps 4 through 6 should be connected by pipes (%>%). Create a new tibble called phds2017 that only contains observations from 2017.
- 5. Drop the year column.
- 6. Sort the tibble from greatest to least based on n_phds.

Exercise 02 (0.5 point)

In the next three exercises, we will replace missing values and calculate summary statistics with four different methods using the phds data set.

- 1. Use count() to count the number of observations in each year in phds.
- 2. Use summarize() to sum n_phds for the entire phds data set. Use na.rm to drop the missing values.
- 3. Use group_by() and summarize() to sum n_phds by year in phds, again using na.rm to drop the missing values. Use the pipe (%>%) instead of assignment. Rename the complicated variable name created by summarize() to n_phds. To do this, simply add n_phds = inside of summarize().

Exercise 03 (0.5 point)

- Drop observations from phds with missing n_phd with filter() and assign the results to a new tibble called phds_no_na.
- 2. Using phds_no_na, group_by(), summarize(), and %>% to add up the number of PhDs by broad_field. You don't need to assign the results.
- 3. Using phds_no_na, group_by(), summarize(), and %>% to add up the number of PhDs by major_field. You don't need to assign the results.
- 4. Using phds_no_na, group_by(), summarize(), and %>% to add up the number of PhDs by field. You don't need to assign the results.

Exercise 04 (0.5 point)

Note: don't use assignment for either approach. Use pipes and then let the answer print to the console.

- Use mutate() and if_else() to create a new variable called n_phds_no_na. To create the variable replace NA values in n_phds with zeros. You will need to come up with condition and true in if_else(). false should be false = n_phds. Pipe the result into summarize() and sum the number of PhDs in the entire data set.
- 2. Use replace_na() inside of mutate() to replace missing n_phds with zeros in the phds tibble. Pipe the result into summarize() and sum the number of PhDs in the entire data set.

Note: You've now seen four different ways to summarize() with missing values. 1. Use na.rm. 2. Filter the missing observations. 3. Replace the missing values with zeros with if_else(). 4. Replace missing values with zeros with replace_na(). Different situations call for different approaches. For instance, replacing NA with 0 results in the same answer as dropping NA for summation but would result in different answers for mean.

Exercise 05 (0.5 point)

pivot_longer() reshapes tibbles so they are longer. pivot_wider() reshapes tibbles so they are wider.

- 1. Explore the documentation for both functions using the ? operator.
- 2. Read the pivoting section in R4DS.
- 3. Examine and duplicate the following code into your script.

```
# create an example tibble
ex5 <- tibble::tribble(
    ~student, ~math_score, ~reading_score,
    "Aaron", 98, 97,
    "Alex", 100, 100
)

# transform from wide to long
ex5_longer <- pivot_longer(
    data = ex5,
    cols = c(math_score, reading_score),
    names_to = "subject",
    values_to = "score"
)

# print the result
ex5_longer</pre>
```

```
## # A tibble: 4 x 3
##
     student subject
                            score
##
     <chr>>
             <chr>
                            <dbl>
## 1 Aaron
             math_score
                                98
                                97
## 2 Aaron
             reading_score
## 3 Alex
             math_score
                               100
## 4 Alex
             reading_score
                               100
```

4. Use pivot_wider() to make ex5_longer wider. It should be identical to the original ex5.

Exercise 06 (1 point)

The next set of exercises will use data about trains in France from TidyTuesday.

"The SNCF (National Society of French Railways) is France's national state-owned railway company. Founded in 1938, it operates the country's national rail traffic along with Monaco, including the TGV, France's high-speed rail network. This dataset covers 2015-2018 with a lot of different train stations. The dataset primarily covers aggregate trip times, delay times, cause for delay, etc for each station - lots of different ways to approach the full_trains.csv dataset with it's 27 columns!"

- 1. read_csv() can read data from a URL if the URL contains a raw csv. Accordingly, the step of downloading the data in exercise 1 was unnecessary. Use read_csv() to directly read in this data and assign it to trains.
- 2. The unit of observation for this tibble is year-month-train line. We want to make the data longer so the unit of observation is year-month-train line-delay type. Essentially, we want to turn the column names that begin with "delay" into a character vector called delay_cause and the values in those columns into a column titled delayed_number. Tips: include cols = starts_with("delay") to pick the columns that should pivoted to longer. Assign the result to trains_long. It should have 32,772 observations and 23 variables.
- 3. Pipe the result from subsection 2. to sort trains_long by year, month, and delay_cause.
- 4. Load this data with read_csv() and assign it to trains_small. Sort the data frame by year, month, and delay_cause.
- 5. The trains_long you created should match trains_small. Add mean(pull(trains_long, delay_cause) == pull(trains_small, delay_cause)) to your script it should return the value 1. Perform the same check for delayed_number it should also return the value 1, but you will need to drop missing values with na.rm = TRUE.

Exercise 07 (1.5 points)

We want to recode the values of month in the trains tibble from integers to month names (e.g. "January", "February", "March"). We will do this three different ways and then sum the number total_num_trips by month.

First approach (Optional)

1. Add the following tibble to your script.

```
months <- tibble::tribble(
    ~month, ~month_name,
    1, "January",
    2, "February",
    3, "March",
    4, "April",
    5, "May",
    6, "June",
    7, "July",
    8, "August",
    9, "September",
    10, "October",
    11, "November",
    12, "December"
)</pre>
```

2. left_join() the months months tibble to trains.

3. Pipe (%>%), group_by() month_name, and sum total_num_trips with summarize().

Second approach

case_when() can be used inside mutate() to conditionally code the levels of a variables. This can be used to create new variables or edit existing variables. For example:

```
tribble(
  ~student_id, ~score,
  "101", 83,
  "102", 99,
  "103", 37,
  "104", 78,
  "105", 92
  ) %>%
  mutate(
    grade = case_when(
      score >= 90 ~ "A",
      score \geq 80 \sim "B",
      score >= 70 ~ "C",
      score \geq 60 \sim "D",
      TRUE ~ "F"
    )
  )
```

The logical condition goes to the left of ~ and the new value goes to the right of ~. Conditions are evaluated from top to bottom (i.e. the second condition is really greater than or equal to 80 and less than 90), and TRUE can be used at the end as an "otherwise" at the end.

- 1. Use case_when() and mutate() to replace integer month in trains with month names in a new variable called month_name. You will need to use == for the logical condition.
- 2. Pipe (%>%), group_by() month_name, and sum total_num_trips with summarize().

Third approach

recode() can be used to replace values in a vector with a new vector. Consider the following example from the function documentation:

```
char_vec <- sample(c("a", "b", "c"), 10, replace = TRUE)
recode(char_vec, a = "Apple")
recode(char_vec, a = "Apple", b = "Banana")</pre>
```

Note: recode() operates on a vector and will need to be used inside mutate().

1. Use recode() to replace the integers with month names in a variable called month_names. Unlike the above example, you will need to wrap the integers in ticks (the key under escape) inside of the

recode() function.

2. Pipe (%>%), group_by() month_name, and sum total_num_trips with summarize().

The three different approaches have different advantages. case_when() is useful because the conditions to the left of ~ can be arbitrarily complex. left_join() is useful when the updated variable is stored in a separate tibble. recode() is useful when there are a few one-to-one recodes.