

Module 1 Assignment 2: Getting to Know your Home

Ellen Bledsoe

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Assignment Description

Purpose

The goal of this assignment is to get comfortable using the `tidyverse` with 2-dimensional data sets and compare this process to using base R.

Task

Write R code using the `tidyverse` to successfully answer each question below.

Criteria for Success

- Code is within the provided code chunks
- Code chunks run without errors
- Code produces the correct result
 - Code that produces the correct answer will receive full credit
 - Code attempts with logical direction will receive partial credit
- Written answers address the questions in sufficient detail

Due Date

Feb 1 at 11am MST

Assignment Questions

For this final assignment for Module 1, you'll be working with another real-world data set—a collection of data from climate stations scattered across Antarctica.

1. In your own words, describe what the `tidyverse` is. Your answer should be between 1-3 sentences.
2. Load in the `tidyverse` package.

```
# load packages  
library(tidyverse)
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
## v dplyr      1.1.4      v readr      2.1.5
## v forcats    1.0.0      v stringr    1.5.1
## v ggplot2    3.4.4      v tibble     3.2.1
## v lubridate  1.9.3      v tidyr      1.3.1
## v purrr      1.0.2
## -- Conflicts ----- tidyverse_conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()     masks stats::lag()
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

3. Load in the data file (called aggregated_station_data.csv) using the read_csv() function. Save the data as an object called weather.

```
# read in weather station data
weather <- read_csv("../data/aggregated_station_data.csv")
```

```
## Rows: 139160 Columns: 12
## -- Column specification -----
## Delimiter: ","
## chr  (1): station_id
## dbl (11): year, day, month, running_day, hour, temp, pressure, wind_speed, w...
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

4. Take a look at the data in whichever way you would like (e.g., str(), head(), etc.). How many rows and columns are in the data? Type your answers below:

rows: 139,160

columns: 12

5. Choose rows that only includes temperatures which are above freezing (AKA greater than 0)

```
filter(weather, temp > 0)
```

```
## # A tibble: 769 x 12
##   year  day month running_day hour  temp pressure wind_speed wind_direction
##   <dbl> <dbl> <dbl>      <dbl> <dbl> <dbl>    <dbl>    <dbl>         <dbl>
## 1 2018     5     1         5     300  0.2     985.        2.6           8
## 2 2018     7     1         7    1800  0.2     988.        6.5          49.7
## 3 2018     7     1         7    2100  1.0     988.        8.0           45
## 4 2018     8     1         8       0  1.4     989.       10.2          44.4
## 5 2018     8     1         8     300  0.5     991.        6.0          212.
## 6 2018     8     1         8     600  0.3     992.        5.3          226.
## 7 2018    20     1        20       0  1.3     969.       10.7          204.
## 8 2018    20     1        20     300  2.6     968.       14.6          203.
## 9 2018    20     1        20     600  1.9     968.       11.5          216.
## 10 2018    20     1        20     900  1.6     967.       15.6          200.
## # i 759 more rows
## # i 3 more variables: humidity <dbl>, delta_t <dbl>, station_id <chr>
```

6. Choose *only* the following columns: year, day, month, temp, station_id. Save these columns as a new object called station_temp.

```
station_temp <- select(weather, year:month, temp, station_id)
station_temp
```

```
## # A tibble: 139,160 x 5
##   year   day month   temp station_id
##   <dbl> <dbl> <dbl> <dbl> <chr>
## 1 2018     1     1 -29.5 ag4201801q3h
## 2 2018     1     1 -27.4 ag4201801q3h
## 3 2018     1     1 -25.5 ag4201801q3h
## 4 2018     1     1 -24.9 ag4201801q3h
## 5 2018     1     1 -25   ag4201801q3h
## 6 2018     1     1 -27.5 ag4201801q3h
## 7 2018     1     1 -30.3 ag4201801q3h
## 8 2018     1     1 -30.1 ag4201801q3h
## 9 2018     2     1 -28.8 ag4201801q3h
## 10 2018     2     1 -26.4 ag4201801q3h
## # i 139,150 more rows
```

7. Using the data frame you created in Q6 above (`station_temp`), add a new column to that data frame that converts the temperature column (currently in Celsius) to Fahrenheit. Call the new column `tempF`.
(Hint: we did this in class—use that same equation!)

```
station_temp %>%
  mutate(tempF = temp*(9/5) + 32)
```

```
## # A tibble: 139,160 x 6
##   year   day month   temp station_id   tempF
##   <dbl> <dbl> <dbl> <dbl> <chr>      <dbl>
## 1 2018     1     1 -29.5 ag4201801q3h -21.1
## 2 2018     1     1 -27.4 ag4201801q3h -17.3
## 3 2018     1     1 -25.5 ag4201801q3h -13.9
## 4 2018     1     1 -24.9 ag4201801q3h -12.8
## 5 2018     1     1 -25   ag4201801q3h -13
## 6 2018     1     1 -27.5 ag4201801q3h -17.5
## 7 2018     1     1 -30.3 ag4201801q3h -22.5
## 8 2018     1     1 -30.1 ag4201801q3h -22.2
## 9 2018     2     1 -28.8 ag4201801q3h -19.8
## 10 2018     2     1 -26.4 ag4201801q3h -15.5
## # i 139,150 more rows
```

8. In *your own words* (either bullet points or sentence form is fine), explain two benefits of using the pipe (`%>%`).

Answer: fewer intermediates (meaning reduced risk of error); more readable code for humans

9. Using the *original* data frame (`weather`), find the minimum temperature recorded for each month (in Celsius, the original column called `temp`). (Hint: think about months first (split) and then temperature (apply). You will also want to remove all the NA values.)

```
weather %>%
  group_by(month) %>%
  summarise(min_temp = min(temp, na.rm = TRUE))
```

```
## # A tibble: 12 x 2
##   month min_temp
##   <dbl>   <dbl>
## 1     1    -44.2
## 2     2    -59
## 3     3   -67.9
## 4     4   -72.3
## 5     5   -77.1
## 6     6    -76
## 7     7   -79.5
## 8     8   -80.2
## 9     9   -77.1
## 10    10   -70.8
## 11    11   -59.4
## 12    12   -41.3
```

10. Again, using the *original* data frame, create a data frame with the mean temperature for the month of January for each station.

Some hints:

- take note of how months are represented in the data
- think about using the pipe, how we choose which rows we want, and how we split-apply-combine
- remember to remove the NA values!

```
weather %>%
  filter(month == 1) %>%
  group_by(station_id) %>%
  summarize(mean_temp = mean(temp, na.rm = TRUE))
```

```
## # A tibble: 49 x 2
##   station_id mean_temp
##   <chr>       <dbl>
## 1 ag4201801q3h -31.4
## 2 bal201801q3h -19.1
## 3 brp201801q3h -6.05
## 4 byd201801q3h -15.5
## 5 cbd201801q3h -3.83
## 6 cha201801q3h -3.04
## 7 d10201801q3h -3.32
## 8 d47201801q3h -13.4
## 9 d85201801q3h -24.2
## 10 dc2201801q3h -27.4
## # i 39 more rows
```

Bonus! (up to 2 points)

Write code to determine how many unique stations are in the `weather` data set. (Hint: look up the help file for the `distinct()` and the `count()` functions).

```
# number of unique stations
weather %>%
  distinct(station_id) %>%
  count()
```

```
## # A tibble: 1 x 1
##       n
##   <int>
## 1   571
```

Turning in Your Assignment

Follow these steps to successfully turn in your assignment on D2L.

1. Click the **Knit** button up near the top of this document. This should produce a PDF file that shows up in the **Files** panel on the bottom-right of your screen.
2. Click the empty box to the left of the PDF file.
3. Click on the blue gear near the top of the **Files** panel and choose **Export**.
4. Put your last name at the front of the file name when prompted, then click the **Download** button. The PDF file of your assignment is now in your “Downloads” folder on your device.
5. Head over to D2L and navigate to Module 1 Assignment 2. Submit the PDF file that you just downloaded.