

TEND_Subsetting_Data

Me

2023-05-27

```
knitr::opts_chunk$set(root.dir = "C:\\Users\\adcre\\OneDrive\\  
Documents\\Desktop_RStudio",  
echo = TRUE)  
getwd()
```

```
## [1] "C:/Users/adcre/OneDrive/Documents/R_Programming/TEND"
```

Adding libraries/packages

```
## Warning: package 'tidyverse' was built under R version 4.2.3
```

```
## Warning: package 'ggplot2' was built under R version 4.2.3
```

```
## Warning: package 'tibble' was built under R version 4.2.3
```

```
## Warning: package 'dplyr' was built under R version 4.2.3
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --
```

```
## v dplyr      1.1.2      v readr      2.1.4
```

```
## v forcats   1.0.0      v stringr   1.5.0
```

```
## v ggplot2    3.4.2      v tibble    3.2.1
```

```
## v lubridate  1.9.2      v tidyr     1.3.0
```

```
## v purrr      1.0.1
```

```
## -- 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
```

```
## Warning: package 'psych' was built under R version 4.2.3
```

```
##
```

```
## Attaching package: 'psych'
```

```
##
```

```
## The following objects are masked from 'package:ggplot2':
```

```
##
```

```
## %+%, alpha
```

```
##
```

```
##
```

```
## Please cite as:
```

```
##
```

```
## Hlavac, Marek (2022). stargazer: Well-Formatted Regression and Summary Statistics Tables.
```

```
## R package version 5.2.3. https://CRAN.R-project.org/package=stargazer
```

R Markdown

This markdown will serve as a reference sheet for different subsetting methods, packages, applications, and more. More specifically, how to subset data casewise, listwise, by ID, or by variable. Having a large and complex df could prevent you from gaining insight about specific data. Having the ability to call for certain ID numbers or variables, like gender or age, allows you to use a magnifying glass on your large df. I also plan on learning and practicing substringing values, and being able to remove the “sub” from “sub-xxxx” in the subject ID column. The key differences between substringing and subsetting is that substringing is applied to character strings like text files, and can output specific ranges or positions in a long character string. A subset is more widely used because it can be applied to data frames and other data structures. Based off of certain arguments or criteria, a subset can give you a closer look on specific points in your data structure. Subsets are represented by brackets[], and more of the syntax will be learned with time and practice.

Subsetting

```
# We will first create a df that allows us to visualize subsetting, and learn the syntax used to call f

# Create a data frame
df <- data.frame(
  Name = c("John", "Emily", "David", "Sarah", "Michael"),
  Age = c(25, 32, 28, 35, 30),
  City = c("New York", "London", "Paris", "Tokyo", "Sydney")
)

# Print the data frame
print(df)
```

The first code chunk will be examples of subsetting on a simple data structure that is created within the chunk. In the interest of truly being able to apply these skills, I will use a df imported from Kaggle after this code chunk in order to test it in a real-life situation.

```
##      Name Age   City
## 1   John  25 New York
## 2  Emily  32  London
## 3  David  28   Paris
## 4  Sarah  35   Tokyo
## 5 Michael 30  Sydney
```

```
# Notice the use of brackets[] and a dollar sign$ to indicate that you want to subset, and that you wou

# Subset the data frame by age, specifically ppl over the age of 30
subset_df <- df[df$Age > 30, ]

# Print the subsetted data frame
print(subset_df)
```

```
##      Name Age   City
## 2 Emily  32  London
## 4 Sarah  35   Tokyo
```

Quick note: is.na()

a logical function that checks the given the data structure and returns TRUE if there are missing values

Operators

```
# first, let's look at is.na(), '&', and '!' in action on the TEND example data
Brain_Data <- read.csv("Brain_Data.csv", header=T, sep=";", na.strings=c("NA", "888", "999"))
Behavioral_Data<-read.csv("Bx_Data.csv", header=T, sep=";", na.strings=c("NA", "888", "999"))
DATA <- merge(Brain_Data, Behavioral_Data, "ID", all=T)

# With this line of code, we are extracting all of the cases that did not have missing values in the sp
SUBdata<- subset(DATA, !is.na(DATA$CDRSR_total) & !is.na(DATA$RSFC1))
View(SUBdata)
```

operators like '!' and '&' can be useful when used properly, like in conjunction with functions within a code chunk. While they don't necessarily do anything powerful, they are very useful when used !

Let's try subsetting by variables and conditions on a more realistic data set found on Kaggle. We will explore the same data set found in my "Importing_Data_Sets" repo, the Depression Dataset found at this link on Kaggle: <https://www.kaggle.com/datasets/arashnic/the-depression-dataset>

```
# First, I will load in the scores from the selected dataset. It can be done many ways, and should be w
dep_scores <- read_csv("scores.csv")
```

This data set is represented as 'scores.csv' in my Files

```
## Rows: 55 Columns: 12
## -- Column specification -----
## Delimiter: ","
## chr (3): number, age, edu
## dbl (9): days, gender, afftype, melanch, inpatient, marriage, work, madsr1, ...
##
## 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.
```

```
str(dep_scores)
```

```
## spc_tbl_ [55 x 12] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ number   : chr [1:55] "condition_1" "condition_2" "condition_3" "condition_4" ...
## $ days     : num [1:55] 11 18 13 13 13 7 11 5 13 9 ...
## $ gender   : num [1:55] 2 2 1 2 2 1 1 2 2 2 ...
## $ age      : chr [1:55] "35-39" "40-44" "45-49" "25-29" ...
```

```
## $ afftype : num [1:55] 2 1 2 2 2 2 1 2 1 2 ...
## $ melanch : num [1:55] 2 2 2 2 2 2 NA NA NA 2 ...
## $ inpatient: num [1:55] 2 2 2 2 2 2 2 2 2 2 ...
## $ edu      : chr [1:55] "6-10" "6-10" "6-10" "11-15" ...
## $ marriage : num [1:55] 1 2 2 1 2 1 2 1 1 1 ...
## $ work     : num [1:55] 2 2 2 1 2 2 1 2 2 2 ...
## $ madsr1   : num [1:55] 19 24 24 20 26 18 24 20 26 28 ...
## $ madsr2   : num [1:55] 19 11 25 16 26 15 25 16 26 21 ...
## - attr(*, "spec")=
## .. cols(
## ..   number = col_character(),
## ..   days = col_double(),
## ..   gender = col_double(),
## ..   age = col_character(),
## ..   afftype = col_double(),
## ..   melanch = col_double(),
## ..   inpatient = col_double(),
## ..   edu = col_character(),
## ..   marriage = col_double(),
## ..   work = col_double(),
## ..   madsr1 = col_double(),
## ..   madsr2 = col_double()
## .. )
## - attr(*, "problems")=<externalptr>
```

By using str(), I can view the different columns by name. This is essentially a list of the variables

```
dep_subset <- dep_scores[dep_scores$age == "40-44", ]
print(dep_subset)
```

```
## # A tibble: 5 x 12
##   number      days gender age  afftype melanch inpatient edu  marriage  work
##   <chr>      <dbl> <dbl> <chr>   <dbl>   <dbl>      <dbl> <chr>    <dbl> <dbl>
## 1 condition_2    18     2 40-44     1     2         2 6-10      2     2
## 2 condition_12   12     2 40-44     1     2         2 6-10      2     2
## 3 condition_18   13     2 40-44     3     2         2 11-15     2     2
## 4 control_8      13     2 40-44    NA     NA        NA <NA>     NA     NA
## 5 control_16     13     2 40-44    NA     NA        NA <NA>     NA     NA
## # i 2 more variables: madsr1 <dbl>, madsr2 <dbl>
```

Notice that we used '==' instead of the logical condition '<' or '>'. So, rather than calling for sub.
What if I wanted to call for a specific ID? There is a 'number' column, and each subject is either re

```
test_subject1 <- dep_scores[dep_scores$number == "condition_1", ]
control_subject1 <- dep_scores[dep_scores$number == "control_1", ]
```

```
View(control_subject1)
View(test_subject1)
```

After looking further into the data, I noticed that the 'gender' column has 1's and 2's, 1 presumably
male_subject_dep <- dep_scores[dep_scores\$gender == 1,]

Let's go one step further and subset by age AND gender by using the '&' operator in the brackets. Make

```
new_subset <- dep_scores[dep_scores$gender == 2 &
                        dep_scores$age == "45-49", ]
View(new_subset)
```

This next code chunk will look into subsetting list-wise and case-wise!

What are the key differences between these methods of subsetting, and when would you use them? To quote GPT-4, “In summary, case-wise subsetting involves selecting specific rows based on conditions, while list-wise subsetting involves selecting specific columns.”

Case-wise refers to subsetting for a specific cases, or ROWS, of the data frame. You could set criteria for case-wise subsetting, such as selecting rows where the 'days' column is greater than 13. Case-wise subsetting is used when you want to focus on specific rows of data.
List-wise, on the other hand, is for when we would like to include all of the data from specific variables. List-wise subsetting is used when you want to focus on specific columns of data.

```
# Let's use the Kaggle depression data set from the previous chunk to observe these subset methods.
dep_scores <- read_csv("scores.csv", show_col_types = FALSE)
str(dep_scores)
```

```
## spc_tbl_ [55 x 12] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ number   : chr [1:55] "condition_1" "condition_2" "condition_3" "condition_4" ...
## $ days     : num [1:55] 11 18 13 13 13 7 11 5 13 9 ...
## $ gender   : num [1:55] 2 2 1 2 2 1 1 2 2 2 ...
## $ age      : chr [1:55] "35-39" "40-44" "45-49" "25-29" ...
## $ afftype  : num [1:55] 2 1 2 2 2 2 1 2 1 2 ...
## $ melanch  : num [1:55] 2 2 2 2 2 2 NA NA NA 2 ...
## $ inpatient: num [1:55] 2 2 2 2 2 2 2 2 2 2 ...
## $ edu      : chr [1:55] "6-10" "6-10" "6-10" "11-15" ...
## $ marriage : num [1:55] 1 2 2 1 2 1 2 1 1 1 ...
## $ work     : num [1:55] 2 2 2 1 2 2 1 2 2 2 ...
## $ madsr1   : num [1:55] 19 24 24 20 26 18 24 20 26 28 ...
## $ madsr2   : num [1:55] 19 11 25 16 26 15 25 16 26 21 ...
## - attr(*, "spec")=
## .. cols(
## ..   number = col_character(),
## ..   days = col_double(),
## ..   gender = col_double(),
## ..   age = col_character(),
## ..   afftype = col_double(),
## ..   melanch = col_double(),
## ..   inpatient = col_double(),
## ..   edu = col_character(),
## ..   marriage = col_double(),
## ..   work = col_double(),
## ..   madsr1 = col_double(),
## ..   madsr2 = col_double()
## .. )
## - attr(*, "problems")=<externalptr>
```

First, we will look at CASE-WISE subsetting. There is a 'days' column, which we will assume refers to the number of days since the onset of symptoms. We will subset the data to only include cases where the number of days is greater than 13.

```
subset_twowksubs <- dep_scores[dep_scores$days > 13, ]
View(subset_twowksubs)
```

Now, it will return cases that are 2 weeks old or older! Thus, case-wise subsetting

It's time to look at LIST-WISE subsetting, which is more focused on certain columns, or variables, within the data frame.

You will notice that we start the [] section of the following code with a ','. This simply means that we are selecting all rows, but only specific columns.

```
listwise_subs <- dep_scores[, c("age", "afftype", "edu")]
View(listwise_subs)
```

```
# For our Kaggle data set, there appears to be two prefixes in the 'number' column. This chunk will show
dep_scores <- read_csv("scores.csv", show_col_types = FALSE)
print(dep_scores)
```

grep! this function is from base R, and is used when pattern matching is needed, in this case, for a subset of prefixed rows.

```
## # A tibble: 55 x 12
##   number      days gender age   afftype melanch inpatient edu  marriage  work
##   <chr>      <dbl> <dbl> <chr>   <dbl>   <dbl>      <dbl> <chr>    <dbl> <dbl>
## 1 condition_1    11     2 35-39     2     2        2 6-10      1     2
## 2 condition_2    18     2 40-44     1     2        2 6-10      2     2
## 3 condition_3    13     1 45-49     2     2        2 6-10      2     2
## 4 condition_4    13     2 25-29     2     2        2 11-15     1     1
## 5 condition_5    13     2 50-54     2     2        2 11-15     2     2
## 6 condition_6     7     1 35-39     2     2        2 6-10      1     2
## 7 condition_7    11     1 20-24     1    NA        2 11-15     2     1
## 8 condition_8     5     2 25-29     2    NA        2 11-15     1     2
## 9 condition_9    13     2 45-49     1    NA        2 6-10      1     2
## 10 condition_~    9     2 45-49     2     2        2 6-10      1     2
## # i 45 more rows
## # i 2 more variables: madsr1 <dbl>, madsr2 <dbl>
```

```
# Subset only the "condition" rows
condition_df <- dep_scores[grepl("^condition_", dep_scores$number), ]

# Subset only the "control" rows
control_df <- dep_scores[grepl("^control_", dep_scores$number), ]

# Check the subsetted data frames
str(condition_df)
```

```
## tibble [23 x 12] (S3: tbl_df/tbl/data.frame)
## $ number   : chr [1:23] "condition_1" "condition_2" "condition_3" "condition_4" ...
## $ days     : num [1:23] 11 18 13 13 13 7 11 5 13 9 ...
## $ gender   : num [1:23] 2 2 1 2 2 1 1 2 2 2 ...
## $ age      : chr [1:23] "35-39" "40-44" "45-49" "25-29" ...
## $ afftype  : num [1:23] 2 1 2 2 2 2 1 2 1 2 ...
## $ melanch  : num [1:23] 2 2 2 2 2 2 NA NA NA 2 ...
## $ inpatient: num [1:23] 2 2 2 2 2 2 2 2 2 2 ...
## $ edu      : chr [1:23] "6-10" "6-10" "6-10" "11-15" ...
## $ marriage : num [1:23] 1 2 2 1 2 1 2 1 1 1 ...
## $ work     : num [1:23] 2 2 2 1 2 2 1 2 2 2 ...
## $ madsr1   : num [1:23] 19 24 24 20 26 18 24 20 26 28 ...
## $ madsr2   : num [1:23] 19 11 25 16 26 15 25 16 26 21 ...
```

```
str(control_df)
```

```
## tibble [32 x 12] (S3: tbl_df/tbl/data.frame)
## $ number   : chr [1:32] "control_1" "control_2" "control_3" "control_4" ...
```

```
## $ days      : num [1:32] 8 20 12 13 13 13 13 13 13 8 ...
## $ gender    : num [1:32] 2 1 2 1 1 1 2 2 1 ...
## $ age       : chr [1:32] "25-29" "30-34" "30-34" "25-29" ...
## $ afftype   : num [1:32] NA NA NA NA NA NA NA NA NA NA NA ...
## $ melanch   : num [1:32] NA NA NA NA NA NA NA NA NA NA NA ...
## $ inpatient: num [1:32] NA NA NA NA NA NA NA NA NA NA NA ...
## $ edu       : chr [1:32] NA NA NA NA ...
## $ marriage  : num [1:32] NA NA NA NA NA NA NA NA NA NA NA ...
## $ work      : num [1:32] NA NA NA NA NA NA NA NA NA NA NA ...
## $ madsr1    : num [1:32] NA NA NA NA NA NA NA NA NA NA NA ...
## $ madsr2    : num [1:32] NA NA NA NA NA NA NA NA NA NA NA ...
```

```
ifelse()
```

```
#Here we are creating a new variable "Current_Med_String" to have strings of what was in numeric 1s (yes)
DATA$Current_Med_String <- ifelse(DATA$Current_Med == 1, "Meds", "noMeds")
```

Subsetting Alternatives

```
# This code chunk contains many different ways to extract rows and columns with more simple syntax.
dep_scores[,] # all rows and all columns
```

Below are some very useful examples that Johanna offered as alternatives to the more complex syntax or functions previously mentioned

```
## # A tibble: 55 x 12
##   number      days gender age   afftype melanch inpatient edu   marriage work
##   <chr>      <dbl> <dbl> <chr>   <dbl>   <dbl>      <dbl> <chr>   <dbl> <dbl>
## 1 condition_1    11      2 35-39      2      2      2 6-10      1      2
## 2 condition_2    18      2 40-44      1      2      2 6-10      2      2
## 3 condition_3    13      1 45-49      2      2      2 6-10      2      2
## 4 condition_4    13      2 25-29      2      2      2 11-15     1      1
## 5 condition_5    13      2 50-54      2      2      2 11-15     2      2
## 6 condition_6     7      1 35-39      2      2      2 6-10      1      2
## 7 condition_7    11      1 20-24      1     NA      2 11-15     2      1
## 8 condition_8     5      2 25-29      2     NA      2 11-15     1      2
## 9 condition_9    13      2 45-49      1     NA      2 6-10      1      2
## 10 condition_~    9      2 45-49      2      2      2 6-10      1      2
## # i 45 more rows
## # i 2 more variables: madsr1 <dbl>, madsr2 <dbl>
```

```
dep_scores[1,] #first row and all columns
```

```
## # A tibble: 1 x 12
##   number      days gender age   afftype melanch inpatient edu   marriage work
##   <chr>      <dbl> <dbl> <chr>   <dbl>   <dbl>      <dbl> <chr>   <dbl> <dbl>
## 1 condition_1    11      2 35-39      2      2      2 6-10      1      2
## # i 2 more variables: madsr1 <dbl>, madsr2 <dbl>
```

```
dep_scores[1:10,] #first 10 rows and all columns
```

```
## # A tibble: 10 x 12
##   number      days gender age   afftype melanch inpatient edu   marriage work
##   <chr>      <dbl> <dbl> <chr>   <dbl>   <dbl>      <dbl> <chr>   <dbl> <dbl>
## 1 condition_1    11      2 35-39      2      2      2 6-10      1      2
## 2 condition_2    18      2 40-44      1      2      2 6-10      2      2
## 3 condition_3    13      1 45-49      2      2      2 6-10      2      2
## 4 condition_4    13      2 25-29      2      2      2 11-15     1      1
## 5 condition_5    13      2 50-54      2      2      2 11-15     2      2
## 6 condition_6     7      1 35-39      2      2      2 6-10      1      2
## 7 condition_7    11      1 20-24      1     NA      2 11-15     2      1
## 8 condition_8     5      2 25-29      2     NA      2 11-15     1      2
## 9 condition_9    13      2 45-49      1     NA      2 6-10      1      2
## 10 condition_~   9      2 45-49      2      2      2 6-10      1      2
## # i 2 more variables: madsr1 <dbl>, madsr2 <dbl>
```

```
dep_scores[c(1,3),1:10] #first and third row and the first 10 columns
```

```
## # A tibble: 2 x 10
##   number      days gender age   afftype melanch inpatient edu   marriage work
##   <chr>      <dbl> <dbl> <chr>   <dbl>   <dbl>      <dbl> <chr>   <dbl> <dbl>
## 1 condition_1    11      2 35-39      2      2      2 6-10      1      2
## 2 condition_3    13      1 45-49      2      2      2 6-10      2      2
```

Substringing

To stay consistent, we will first create a simple data set within our code to visualize and practice
Substr() takes three arguments: the character vector being modified, the starting position, and the end position

```
#First, create the df
substr_df <- data.frame(
  ParticipantID = c("sub_1234", "sub_5678", "sub_9101", "sub_2345"),
  Score = c(15, 20, 18, 22),
  stringsAsFactors = FALSE
)

# Remove the prefix with substr(), no need to assign a variable!
substr_df$ParticipantID <- substr(substr_df$ParticipantID, start = 5, stop = 8)
print(substr_df)
```

Let's say we had a data frame, but in our subject's row3, there is a prefix called "sub_". This means that each case is represented as 'sub_XXXX'. If I wanted to remove that prefix, or any prefixes for that matter, I would use substringing! Substringing is useful when you know the specific ranges or positions that you want to call for. In this case, we know that the last 4 characters are the ID, and we don't need the prefix. This code will show you how to solve this problem, so that subjects are only represented by integers, not a prefix as well.

```
##   ParticipantID Score
```



```
## 1      1234    15
## 2      5678    20
## 3      9101    18
## 4      2345    22
```

```
dep_scores <- read_csv("scores.csv", show_col_types = FALSE)
str(dep_scores)
```

Now that we have looked at a basic example of substringing to remove unwanted prefixes, let's try it on our Kaggle depression data set

```
## spc_tbl_ [55 x 12] (S3: spec_tbl_df/tbl_df/tbl/data.frame)
## $ number   : chr [1:55] "condition_1" "condition_2" "condition_3" "condition_4" ...
## $ days     : num [1:55] 11 18 13 13 13 7 11 5 13 9 ...
## $ gender   : num [1:55] 2 2 1 2 2 1 1 2 2 2 ...
## $ age      : chr [1:55] "35-39" "40-44" "45-49" "25-29" ...
## $ afftype  : num [1:55] 2 1 2 2 2 2 1 2 1 2 ...
## $ melanch  : num [1:55] 2 2 2 2 2 2 NA NA NA 2 ...
## $ inpatient: num [1:55] 2 2 2 2 2 2 2 2 2 2 ...
## $ edu      : chr [1:55] "6-10" "6-10" "6-10" "11-15" ...
## $ marriage : num [1:55] 1 2 2 1 2 1 2 1 1 1 ...
## $ work     : num [1:55] 2 2 2 1 2 2 1 2 2 2 ...
## $ madsr1   : num [1:55] 19 24 24 20 26 18 24 20 26 28 ...
## $ madsr2   : num [1:55] 19 11 25 16 26 15 25 16 26 21 ...
## - attr(*, "spec")=
## .. cols(
## ..   number = col_character(),
## ..   days = col_double(),
## ..   gender = col_double(),
## ..   age = col_character(),
## ..   afftype = col_double(),
## ..   melanch = col_double(),
## ..   inpatient = col_double(),
## ..   edu = col_character(),
## ..   marriage = col_double(),
## ..   work = col_double(),
## ..   madsr1 = col_double(),
## ..   madsr2 = col_double()
## .. )
## - attr(*, "problems")=<externalptr>
```

Using str() allows me to peek at the names of the rows that pertain to certain subjects. What I notice

First, the method using sub()

```
dep_scores$number <- sub("^condition_|^control_", "", dep_scores$number)
print(dep_scores)
```

```
## # A tibble: 55 x 12
##   number days gender age   afftype melanch inpatient edu   marriage work
##   <chr>  <dbl>  <dbl> <chr>   <dbl>   <dbl>   <dbl> <chr>   <dbl> <dbl>
```

```
## 1 1      11      2 35-39      2      2      2 6-10      1      2
## 2 2      18      2 40-44      1      2      2 6-10      2      2
## 3 3      13      1 45-49      2      2      2 6-10      2      2
## 4 4      13      2 25-29      2      2      2 11-15     1      1
## 5 5      13      2 50-54      2      2      2 11-15     2      2
## 6 6       7      1 35-39      2      2      2 6-10      1      2
## 7 7      11      1 20-24      1     NA      2 11-15     2      1
## 8 8       5      2 25-29      2     NA      2 11-15     1      2
## 9 9      13      2 45-49      1     NA      2 6-10      1      2
## 10 10     9      2 45-49      2      2      2 6-10      1      2
## # i 45 more rows
## # i 2 more variables: madsr1 <dbl>, madsr2 <dbl>
```

I want to emphasize that we only practiced that method for the sake of practice. It may be counterproductive.

DURING-Session REFLECTION

I want to thank Krupali, who helped me find my way when navigating this software. I was in need of some short-term goals, and she helped contextualize RStudio in a way that inspired me to get back to the computer and keep going! Thank you, Krupali!

I need to look more into `ifelse()` statements and levelling

I noticed myself deviating from Johanna's videos and using the internet for some troubleshooting which I think is totally fine, however it ended up costing me time and cognitive load, because had I finished watching the video from Week 2, I would've learned many functions or concepts that I had trouble understanding. Instead of following her videos or even the Rmd files, I was just asking the internet for answers to questions I didn't end up needing an answer to. While I don't think it is wise to follow only one source of content, I must realize that these skills are only useful if they make sense in your specific environment, TEND lab. If I receive advice from the internet, they won't understand the full context of the problems I am trying to solve. THAT, is why I believe following the recordings/markdown/schedule will allow me to set realistic, useful, and short-term goals for learning RStudio and how to be of use to my peers!

POST-Session REFLECTION

TBD....