Session III Recap

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This week we focused on chapters 4 and 5 of R for Data Science.

Chapter 4

To start, we had a brief conversation about naming conventions. For syntax, we want to choose either CamelCase or snake_case. I would recommend snake_case but there is nothing wrong with CamelCase. Some people use (unoffical name) dot.case. I recommend against this strongly as there are some functions written with that notation and we don't want to mix up variable names with function names.

When choosing a name for a data frame or variable, we want to be specific enough to know what the variable is without being too lengthy. For example if we had a dataset on Kobe Bryant's career stats, we wouldn't name the variable "kobe_bryant_points_per_game". Instead we would call the data frame "kobe_career_stats" and the variable "ppg" or "pts_game".

The next part of chapter 4 involves looking at functions. The syntax for a function is function(input..) where input could be a single value or multiple. For examples well use the seq function from base R.

```
seq(1,10)
```

```
## [1] 1 2 3 4 5 6 7 8 9 10
```

As we can see, this function prints all numbers from the first input to the second. By default to and from are set to 1. Then we can accomplish the same as above using:

```
seq(from = 10)
```

```
## [1] 1 2 3 4 5 6 7 8 9 10
```

Another input or argument for this function is by. This will tell R how many numbers to count by. Default is to go by 1 (as above).

```
seq(1, 10, by = 2)
## [1] 1 3 5 7 9
seq(2, 10, by = 3)
```

```
## [1] 2 5 8
```

Alternatively, we can specify length.out which tells R the length of the output vector or how many numbers to print. Then length.out = 15 tells R to print 15 even spaced numbers between and including to and from

```
seq(1, 1000, length.out = 15)
           1.00000
                                            215.07143
    [1]
                      72.35714
                                143.71429
                                                        286.42857
                                                                    357.78571
                                            643.21429
    [7]
         429.14286
                     500.50000
                                571.85714
                                                        714.57143
                                                                    785.92857
## [13]
         857.28571
                     928.64286 1000.00000
```

If we are saving something like this in a variable and want to check that it is evaluating correctly, we can wrap the definition inside parenthesis.

```
# Store the sequence in the variable a
a \leftarrow seg(1, 1000, length.out = 15)
# Store the sequence in the variable b AND print to console
(b \leftarrow seq(1, 1000, length.out = 15))
##
    [1]
           1.00000
                      72.35714 143.71429
                                            215.07143
                                                        286.42857
                                                                    357.78571
   [7]
                                            643.21429
         429.14286
                     500.50000 571.85714
                                                        714.57143
                                                                    785.92857
         857.28571
                     928.64286 1000.00000
```

Chapter 5

The focus of this chapter is data transformations or data wrangling; this is the process of manipulating our data into a certain format that makes summaries and/or plotting a specific characteristic easier. For these examples, we used the scoobydoo dataset from the most recent TidyTuesday.

Data wrangling is done largly with {dplyr}. The main functions are filter, arrange, group_by, mutate, summarize and select.

```
num_obs <- scoobydoo %>%
  group_by(format) %>%
  summarize(count = n()) %>%
  arrange(-count)

knitr::kable(num_obs)
```

format	count
TV Series	374
TV Series (segmented)	175
Movie	43
Crossover	8
Movie (Theatrical)	3

In the above:

- group_by(format) tells R that the manipulation in the following lines should be done for each category as a whole
- summarize(count = n()) creates a new variable called count and condenses the dataframe so that it only has one observation for the group_by variable.
- arrange(-count) arranges the data by the count variable from greatest to least

The mutate function behaves very similar to summarize. The difference is that it will tack the new variable on to the end of the dataframe, instead of summarizing the data. Note: the example below code below is not displayed because the resulting data frame has 603 rows and 76 columns.

```
num_obs_2 <- scoobydoo %>%
  group_by(format) %>%
  mutate(count = n()) %>%
  arrange(-count)
```

Let's say we wanted to calculate the average imdb score for each season of each show.

```
avg_imdb_rating <- scoobydoo %>%
  filter(format == "TV Series") %>%
  select(c(series_name, season, imdb)) %>%
  group_by(series_name, season) %>%
  mutate(imdb = as.numeric(imdb)) %>%
  summarize(avg_imdb = mean(imdb))
```

Warning in mask\$eval_all_mutate(quo): NAs introduced by coercion

`summarise()` has grouped output by 'series_name'. You can override using the `.groups` argument.
knitr::kable(avg_imdb_rating)

series_name	season	avg_imdb
A Pup Named Scooby-Doo	1	7.453846
A Pup Named Scooby-Doo	2	7.625000
A Pup Named Scooby-Doo	3	7.050000
A Pup Named Scooby-Doo	4	7.200000
Be Cool, Scooby-Doo!	1	7.538462
Be Cool, Scooby-Doo!	2	7.329630
Lego	Special	6.600000
Night of the Living Doo	Special	7.200000
Scooby-Doo and Guess Who?	1	7.865385
Scooby-Doo and Guess Who?	2	NA
Scooby-Doo and Scrappy-Doo (first series)	1	7.450000
Scooby-Doo Mystery Incorporated	1	8.180769
Scooby-Doo Mystery Incorporated	2	8.407692
Scooby Doo, Where Are You!	1	8.123529
Scooby Doo, Where Are You!	2	8.087500
Shaggy & Scooby-Doo Get a Clue!	1	5.800000
Shaggy & Scooby-Doo Get a Clue!	2	5.407692
The 13 Ghosts of Scooby-Doo	1	7.476923
The New Scooby-Doo Movies	1	7.743750
The New Scooby-Doo Movies	2	7.562500
The New Scooby-Doo Mysteries	1	7.516667
The New Scooby and Scrappy Doo Show	1	6.500000
The Scooby-Doo Project	Special	7.400000
The Scooby-Doo Show	1	7.481250
The Scooby-Doo Show	2	7.687500
The Scooby-Doo Show	3	7.450000
Warner Home Video	Special	6.900000
What's New Scooby-Doo?	1	7.450000
What's New Scooby-Doo?	2	7.407143
What's New Scooby-Doo?	3	7.428571

In the above code:

- filter(format == "TV Series") goes through the data and removes rows that have a format other than TV Series. In general, the filter function takes in a logical argument and then uses that to parse the dataset for rows where this is true.
- select(c(series_name, season, imdb)) removes all variables except for those listed. Alternatively, a in front of the c(would *remove* the variables listed and keep all others.
- mutate(imdb = as.numeric(imdb)) This is the code that I couldn't see during our session but you

were able to figure out!

Next week, we'll focus on creating visualizations based on these manipulations (and others) as well as putting them into an RMarkdown document.