STA302H1 - Final Project Descriptive Statistics

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Import STA302H1 Study Time and COVID Contemplation Time vs. Quiz Performance Dataset

Data Cleaning

First, I'll clean my data.

```
cleaned_sta302_performance_data <- sta302_performance_data %>%
    # Create a new "country" column, which is just "Country" but whose entries are factors.
   mutate(country = as.factor(Country)) %>%
    # Remove the "X" column: it's simply the row number, which isn't very useful.
    # Remove the "Country" column: column "country" already exists
   select(-X, -Country) %>%
   # Rearrange similar columns side-by-side.
   relocate(country,
             COVID.hours..W1., COVID.hours..W2.,
             COVID.hours..W3., COVID.hours..W4.,
             STA302.hours..W1., STA302.hours..W2.,
             STA302.hours..W3., STA302.hours..W4.,
             Quiz_1_score, Quiz_2_score,
             Quiz_3_score, Quiz_4_score)
    # Identify rows with at least 3 missing quiz marks.
    # These indicate students who have dropped STA302H1, and who
    # should be excluded from the final data.
```

Rows With At Least One NA

Rows with at least one NA deserve closer examination.

Some of the rows might only have 1 - 2 NAs and are therefore salvageable, which is OK.

Other rows may contain 3 or more NAs, and might indicate students who have dropped STA302H1. We'd like to exclude them from our analysis.

```
rows_with_at_least_one_NA = function(data) {
  return (rowSums(is.na(cleaned_sta302_performance_data)) >= 1)
}
rows_with_some_NAs = cleaned_sta302_performance_data[
  rows_with_at_least_one_NA(cleaned_sta302_performance_data),
٦
print(as tibble(data.frame(
  num_row_NAs = nrow(rows_with_some_NAs)
)))
## # A tibble: 1 x 1
##
    num_row_NAs
##
           <int>
## 1
              84
```

Let's identify which rows only have 1 NA.

```
which(rows_with_at_least_one_NA(cleaned_sta302_performance_data))
```

```
[1]
         5
             6
                 9
                    10
                       12 16
                              17
                                   19
                                       23
                                           28
                                               30
                                                   34
                                                       39
                                                           40
                                                               41 43 46 47
## [20]
        51 52 53
                    56
                       58
                            59
                               64
                                   69 78 80
                                               90
                                                  91
                                                       95
                                                           96
                                                               98 100 102 113 117
## [39] 120 121 125 131 132 133 138 143 145 148 168 181 182 188 189 192 194 195 197
## [58] 198 202 203 204 205 206 207 208 209 210 211 212 213 214 215 216 217 218 219
## [77] 220 221 222 223 224 225 226 227
```

Columns with NAs

```
num_NAs_in_column = function(predictor_variable) {
  sum(is.na(predictor_variable))
}
row_nums_of_NA = function(data, predictor_variable) {
  which(is.na(predictor_variable))
}
perform_data = cleaned_sta302_performance_data
print(as tibble(data.frame(
  week1_covid = num_NAs_in_column(perform_data$COVID.hours..W1.),
  week2_covid = num_NAs_in_column(perform_data$COVID.hours..W2.),
  week3_covid = num_NAs_in_column(perform_data$COVID.hours..W3.),
  week4_covid = num_NAs_in_column(perform_data$COVID.hours..W4.)
)))
## # A tibble: 1 x 4
     week1_covid week2_covid week3_covid week4_covid
##
           <int>
                       <int>
                                   <int>
                                               <int>
## 1
                                      21
                                                  40
print(as tibble(data.frame(
  week1_sta302 = num_NAs_in_column(perform_data$STA302.hours..W1.),
  week2_sta302 = num_NAs_in_column(perform_data$STA302.hours..W2.),
  week3_sta302 = num_NAs_in_column(perform_data$STA302.hours..W3.),
  week4 sta302 = num NAs in column(perform data$STA302.hours..W4.)
)))
## # A tibble: 1 x 4
    week1 sta302 week2 sta302 week3 sta302 week4 sta302
##
           <int>
                         <int>
                                 <int>
                                                 <int>
## 1
               26
                                         20
                                                      40
print(as_tibble(data.frame(
  quiz1_score = num_NAs_in_column(perform_data$Quiz_1_score),
  quiz2_score = num_NAs_in_column(perform_data$Quiz_2_score),
  quiz3_score = num_NAs_in_column(perform_data$Quiz_3_score),
  quiz4_score = num_NAs_in_column(perform_data$Quiz_4_score)
)))
## # A tibble: 1 x 4
    quiz1_score quiz2_score quiz3_score quiz4_score
##
           <int>
                       <int>
                                   <int>
                                               <int>
## 1
              13
                          36
                                      31
```

Rows with Mistyped Columns

Rows whose columns are mis-typed may need to be corrected via imputation.

```
rows_with_mistyped_columms = cleaned_sta302_performance_data[c(38, 83, 84, 117),]
# row 83: Country -> "canada" -- DONE
# row 84: Country -> "canada" -- DONE

# row 117: COVID.hours..W4. -> 0.5 hours -- DONE

# row 38: STA302.hours..W3. -> 5.5<U+00A0> -- DONE
# row 117: STA302.hours..W4. -> 7.5 hours -- DONE
```

```
# library(janitor)
# use it to clean up data.
```

Rows Without Country Entry

Taking out the country column can come in handy for functions like cor() where factors aren't allowed.

```
rows_with_no_country = cleaned_sta302_performance_data %>%
select(-country)
```

Find Significance Predictor Variables, Select Predictor Variables Based on Criterion

```
# use week 5b slides -- choose model selection criterion to pick predictor variables.

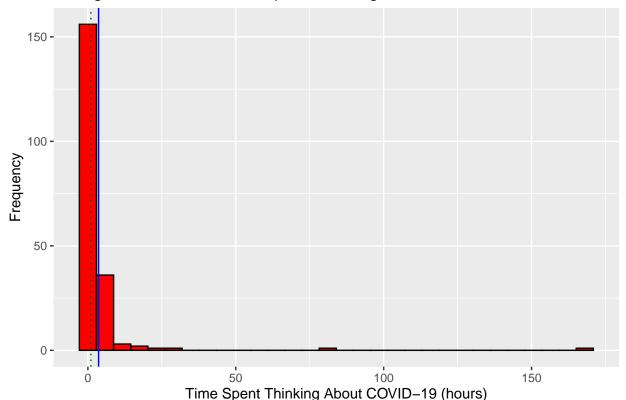
# use lm() on a bunch of predictor variables to determine significant
# predictor variables.
```

Histograms

Histograms of COVID Hours

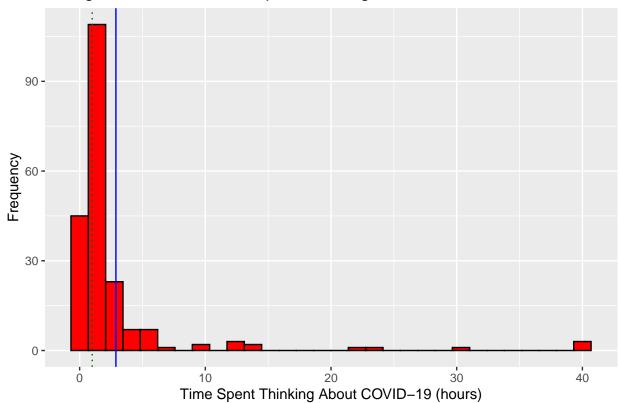
Warning: Removed 26 rows containing non-finite values (stat_bin).

Histogram of Week 1 Time Spent Thinking About COVID-19



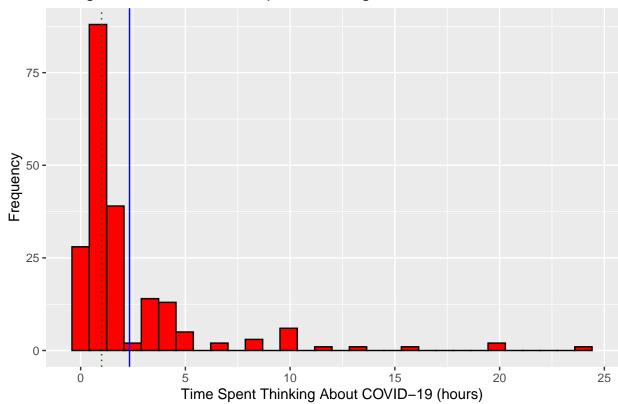
Warning: Removed 22 rows containing non-finite values (stat_bin).

Histogram of Week 2 Time Spent Thinking About COVID-19



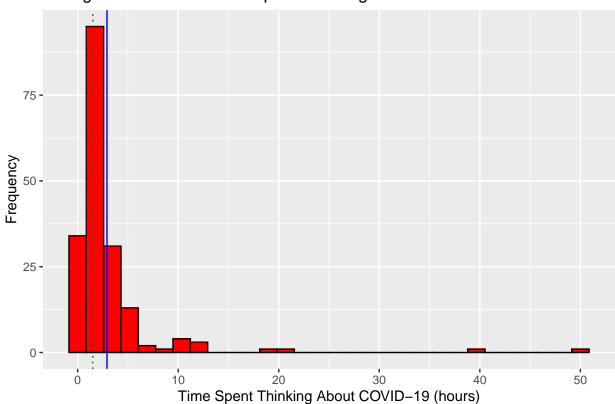
Warning: Removed 21 rows containing non-finite values (stat_bin).

Histogram of Week 3 Time Spent Thinking About COVID-19



Warning: Removed 40 rows containing non-finite values (stat_bin).

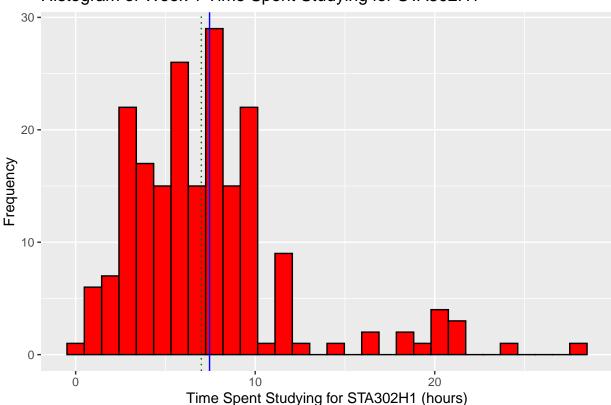
Histogram of Week 4 Time Spent Thinking About COVID-19



Histograms of STA302H1 Hours

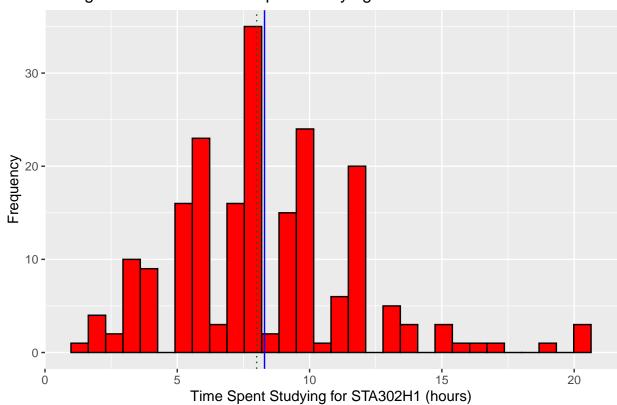
Warning: Removed 26 rows containing non-finite values (stat_bin).

Histogram of Week 1 Time Spent Studying for STA302H1



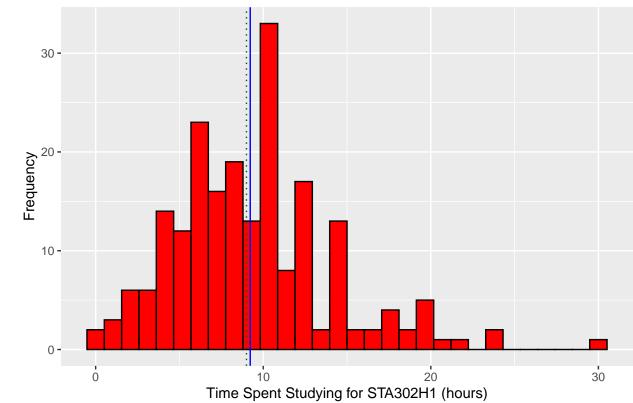
Warning: Removed 22 rows containing non-finite values (stat_bin).

Histogram of Week 2 Time Spent Studying for STA302H1



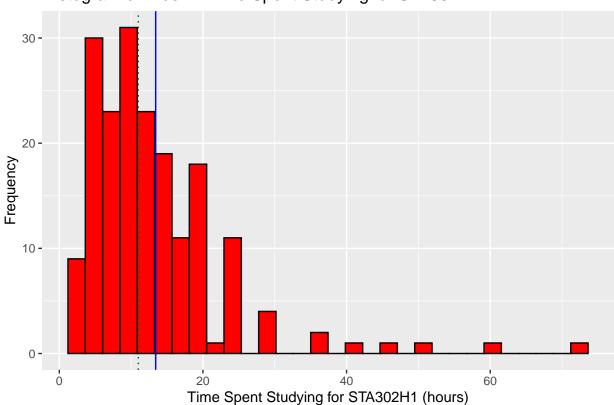
Warning: Removed 20 rows containing non-finite values (stat_bin).

Histogram of Week 3 Time Spent Studying for STA302H1



Warning: Removed 40 rows containing non-finite values (stat_bin).

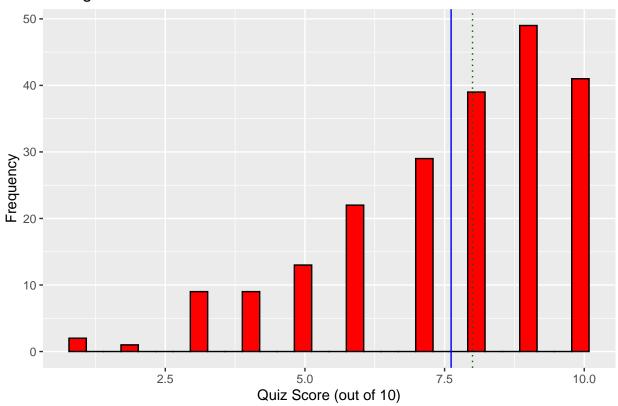
Histogram of Week 4 Time Spent Studying for STA302H1



Histograms of Quiz Scores

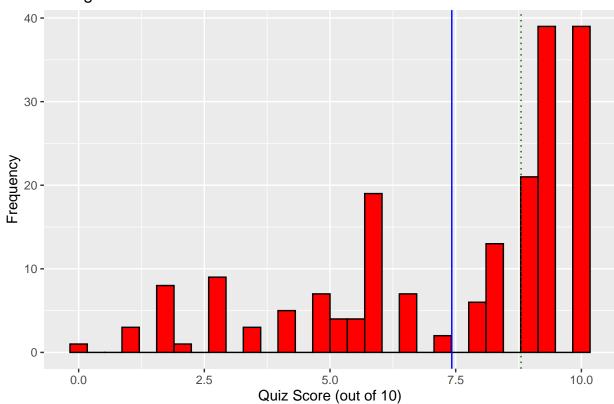
Warning: Removed 13 rows containing non-finite values (stat_bin).

Histogram of Quiz 1 Scores



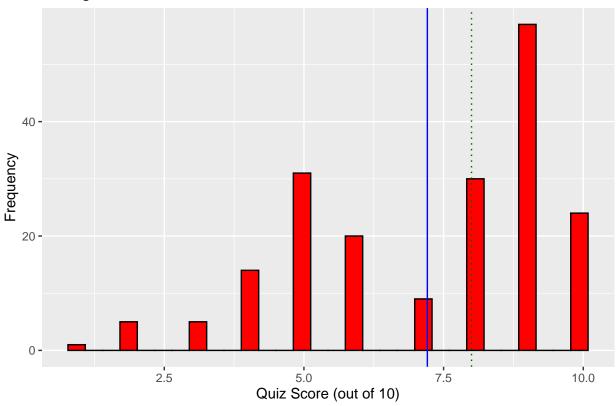
Warning: Removed 36 rows containing non-finite values (stat_bin).

Histogram of Quiz 2 Scores



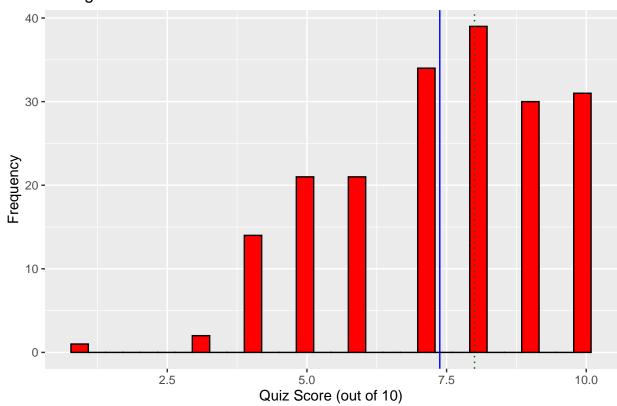
Warning: Removed 31 rows containing non-finite values (stat_bin).

Histogram of Quiz 3 Scores



Warning: Removed 34 rows containing non-finite values (stat_bin).

Histogram of Quiz 4 Scores



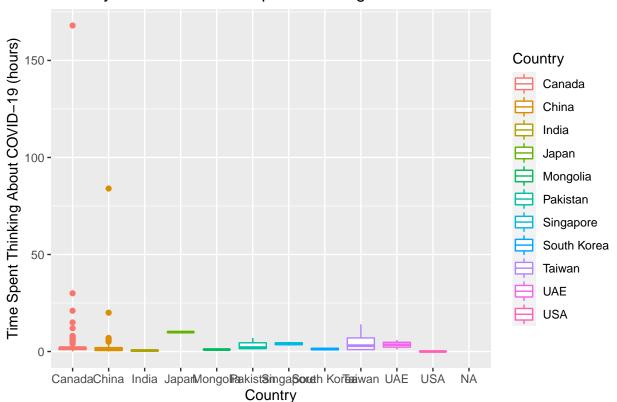
Boxplots

```
display_boxplot <- function(data, predictor_variable, boxplot_title, y_axis_label) {
   ggplot(mapping = aes(x = Country, y = predictor_variable, color = Country)) +
      geom_boxplot(mapping = aes(x = Country, y = predictor_variable)) +
   labs(title = boxplot_title,
      x = "Country",
      y = y_axis_label)
}</pre>
```

Boxplots of COVID Hours

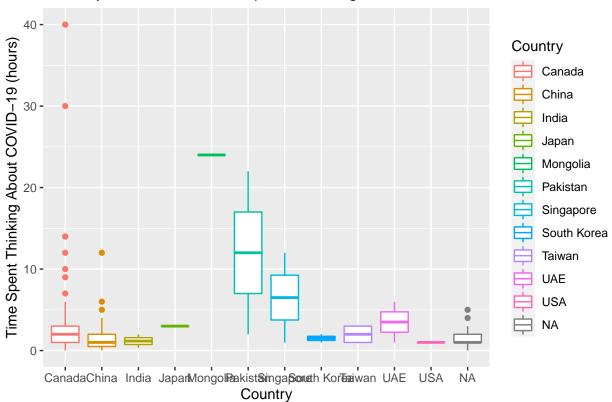
Warning: Removed 26 rows containing non-finite values (stat_boxplot).

Country vs. Week 1 Time Spent Thinking About COVID-19

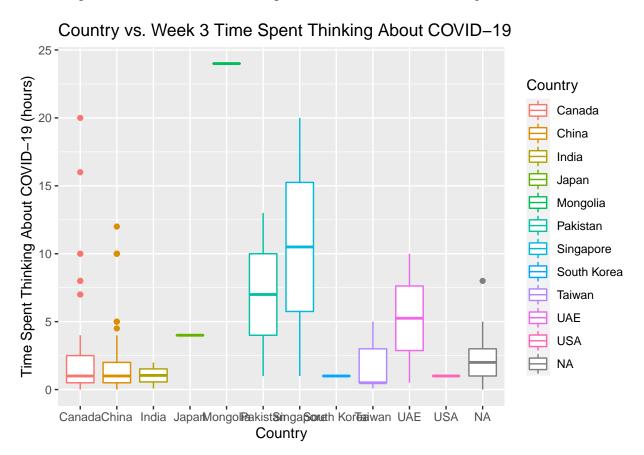


Warning: Removed 22 rows containing non-finite values (stat_boxplot).



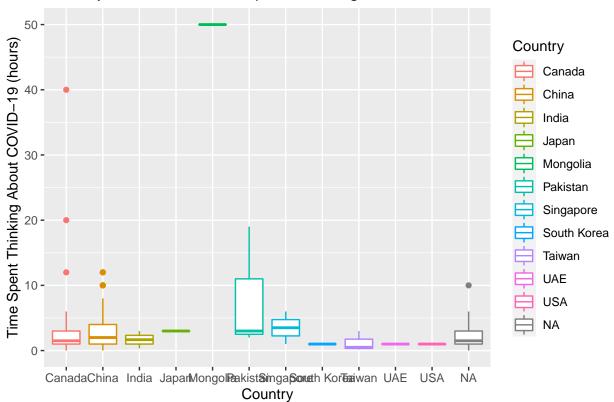


Warning: Removed 21 rows containing non-finite values (stat_boxplot).



Warning: Removed 40 rows containing non-finite values (stat_boxplot).

Country vs. Week 4 Time Spent Thinking About COVID-19



Boxplots of STA302H1 Hours

Warning: Removed 26 rows containing non-finite values (stat_boxplot).

CanadaChina India JapanMongolPeakistSingaSonuth KorFeaiwan UAE USA

Country

Country vs. Week 1 Time Spent Studying For STA302H1 Country Country Country Amongolia Pakistan Singapore South Korea Taiwan UAE USA

Warning: Removed 22 rows containing non-finite values (stat_boxplot).

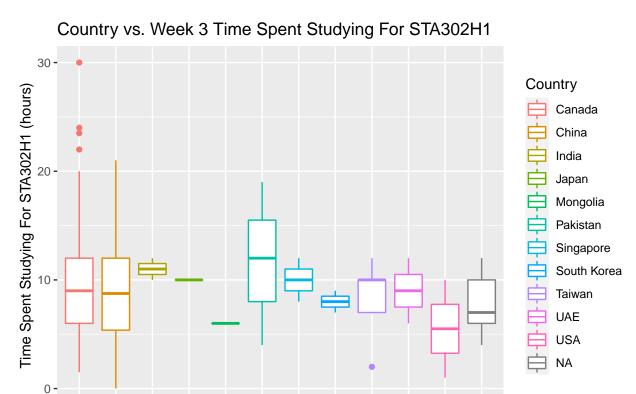
Country vs. Week 2 Time Spent Studying For STA302H1 20 -Country Time Spent Studying For STA302H1 (hours) Canada China India Japan Mongolia Pakistan Singapore South Korea Taiwan UAE USA NA

CanadaChina India JaparMongolPakistSingaSoutth KorEssiwan UAE USA NA Country

Warning: Removed 20 rows containing non-finite values (stat_boxplot).

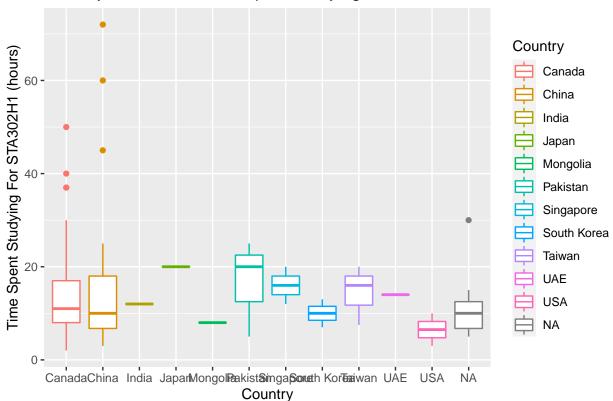
CanadaChina India JapanMongolPakistSingaSonuth KorTeaiwan UAE USA

Country



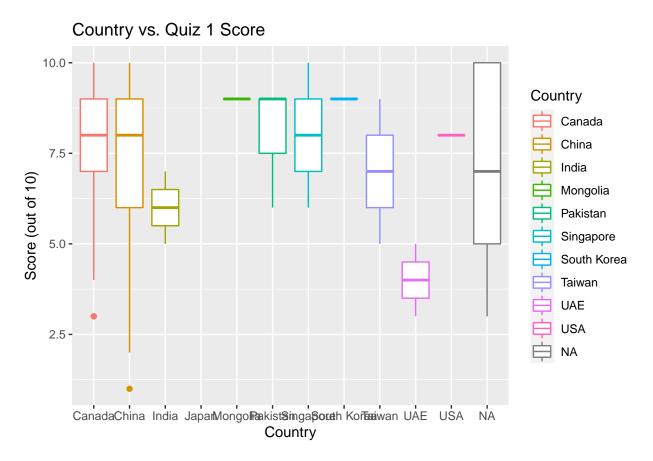
Warning: Removed 40 rows containing non-finite values (stat_boxplot).

Country vs. Week 4 Time Spent Studying For STA302H1

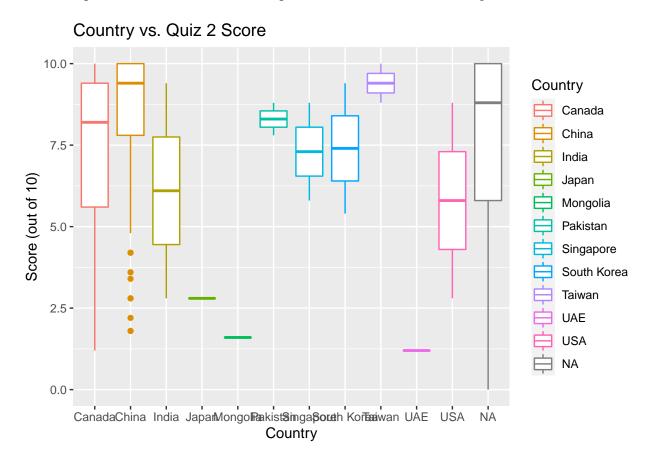


Boxplots of Quiz Scores

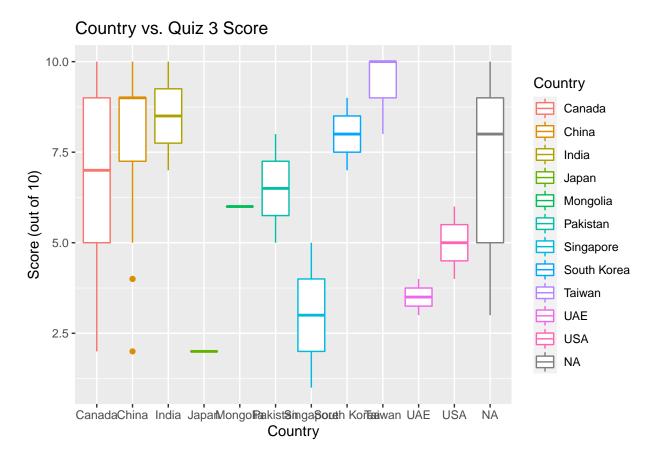
Warning: Removed 13 rows containing non-finite values (stat_boxplot).



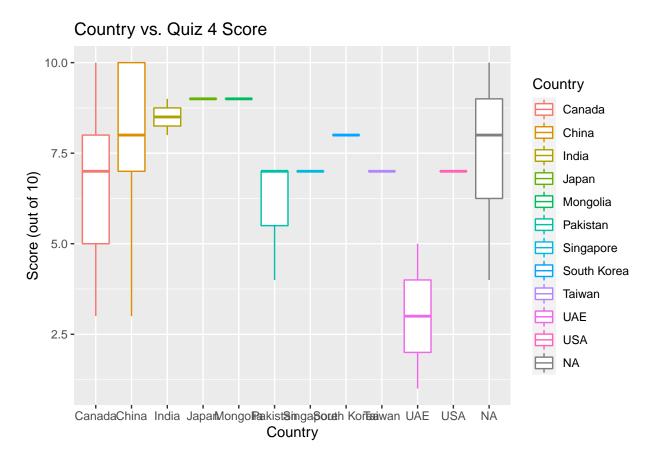
Warning: Removed 36 rows containing non-finite values (stat_boxplot).



Warning: Removed 31 rows containing non-finite values (stat_boxplot).



Warning: Removed 34 rows containing non-finite values (stat_boxplot).



Scatterplots

Comprehensive pairwise scatterplot

```
pairs(~COVID.hours..W1. + COVID.hours..W2. + COVID.hours..W3. + COVID.hours..W4. +
      STA302.hours..W1. + STA302.hours..W2. + STA302.hours..W3. + STA302.hours..W4. +
      Quiz_1_score + Quiz_2_score + Quiz_3_score + Quiz_4_score,
      data = cleaned_sta302_performance_data)
         0 30
                      0 40
                                                0 50
                                                             0 6
          D.hours
                D.hours.
                       D.hours
                             02.hours
                             0 20
                                          0 20
     150
                0 15
                                                       2 8
                                                                    2 8
## GGally
# ggpairs -- removes bottom half of pairs plot
# ggpairs(data = cleaned_sta302_performance_data)
```

Slightly Zoomed In Pairwise Scatterplots

```
# 1. Zoom in a bit by creating 3 - 4 pairs() functions:

# quiz4 ~ quiz 1, 2, 3

# covid4 ~ covid 1, 2, 3

# sta302h14 ~ sta302h1 1, 2, 3
```

Top 4 - 5 Intersting Scatterplots

- # 2. Pick out 4 5 scatterplots that have interesting relationships
- # 3. back up your choices with their correlation (R value).

Correlation Matrix

All Countries

We can find correlation matrix to determine candidate significant predictor values.

```
library(GGally)
## Registered S3 method overwritten by 'GGally':
     method from
     +.gg
          ggplot2
colnames(rows_with_no_country) <- c("W1COV", "W2COV", "W3COV", "W4COV",</pre>
                          "W1302", "W2302", "W3302", "W4302",
                          "Q1", "Q2", "Q3", "Q4")
ggcorr(rows_with_no_country, label = TRUE, label_round = 2)
                                                         Q4
                                                    Q3 0.55
                                               Q2 0.23 0.19
                                          Q1 0.25 0.29 0.29
                                  W43020.010.04-0.05-0.06
                                                                     1.0
                                                                     0.5
                             W330<mark>20.62</mark> 0.03 0.08-0.12-0.08
                                                                     0.0
                        W23020.73 0.48 0.05 0.06 - 0.05 - 0.11
                                                                     -0.5
                   W130<mark>20.64 0.62 0.3 0.09 0.13</mark>-0.04-0.08
                                                                     -1.0
              W4COV.02 0.07 0.09 0.07 0.12 -0.1 0.02 0.06
         W3CO<mark>0,72 0.11 0.11 0.18 0.13 0.07 -0.07-0.11-0.09</mark>
    W2CO 0.68 0.71 0.05 0.09 0.17 0.19 0.11 -0.1-0.12-0.01
V1COV0.55 0.46 0.27 0.04-0.02-0.010.04 0.08 0.06 0.07 0.02
round(cor(rows_with_no_country, use = "pairwise.complete.obs", method = "pearson"), 2)
         W1COV W2COV W3COV W4COV W1302 W2302 W3302 W4302
                                                             Q1
                                                                   Q2
                                                                          Q3
                                                                                Q4
## W1COV 1.00 0.55 0.46 0.27 0.04 -0.02 -0.01 0.04
                                                           0.08 0.06 0.07
## W2COV 0.55 1.00 0.68 0.71 0.05 0.09
                                               0.17
                                                     0.19
                                                           0.11 -0.10 -0.12 -0.01
## W3COV 0.46 0.68 1.00 0.72 0.11 0.11 0.18 0.13 0.07 -0.07 -0.11 -0.09
```

W4COV 0.27 0.71 0.72 1.00 0.02 0.07 0.09 0.07 0.12 -0.10 0.02 0.06 ## W1302 0.04 0.05 0.11 0.02 1.00 0.64 0.62 0.30 0.09 0.13 -0.04 -0.08

```
## W2302 -0.02 0.09 0.11 0.07 0.64 1.00 0.73 0.48 0.05 0.06 -0.05 -0.11 ## W3302 -0.01 0.17 0.18 0.09 0.62 0.73 1.00 0.62 0.03 0.08 -0.12 -0.08 ## W4302 0.04 0.19 0.13 0.07 0.30 0.48 0.62 1.00 -0.01 0.04 -0.05 -0.06 ## Q1 0.08 0.11 0.07 0.12 0.09 0.05 0.03 -0.01 1.00 0.25 0.29 0.29 ## Q2 0.06 -0.10 -0.07 -0.10 0.13 0.06 0.08 0.04 0.25 1.00 0.23 0.19 ## Q3 0.07 -0.12 -0.11 0.02 -0.04 -0.05 -0.12 -0.05 0.29 0.29 0.25 ## Q4 0.02 -0.01 -0.09 0.06 -0.08 -0.11 -0.08 -0.06 0.29 0.19 0.55 1.00
```

By Individual Country

TODO: You could also create separate correlation matrices for each country.

5-Number Summary Statistics

STA302H1 Hours 5-Number Summary

```
summary(sta302_performance_data$STA302.hours..W1.)
##
     Min. 1st Qu. Median
                          Mean 3rd Qu.
                                           Max.
                                                  NA's
    0.000
          4.000 7.000 7.458 9.000 28.000
##
summary(sta302_performance_data$STA302.hours..W2.)
##
     Min. 1st Qu. Median
                                                  NA's
                           Mean 3rd Qu.
                                           Max.
    1.000 6.000 8.000 8.298 10.000 20.000
##
                                                    22
summary(sta302_performance_data$STA302.hours..W3.)
##
     Min. 1st Qu. Median
                           Mean 3rd Qu.
                                           Max.
                                                  NA's
    0.000 6.000 9.000 9.225 11.500 30.000
##
                                                    20
summary(sta302_performance_data$STA302.hours..W4.)
##
     Min. 1st Qu. Median Mean 3rd Qu.
                                          Max.
                                                  NA's
##
     2.00 7.00 11.00 13.42 16.00 72.00
                                                    40
```

COVID Hours 5-Number Summary

```
summary(sta302_performance_data$COVID.hours..W1.)
##
     Min. 1st Qu. Median
                          Mean 3rd Qu.
                                                  NA's
                                           Max.
##
    0.000 1.000 1.000 3.607 2.000 168.000
                                                    26
summary(sta302_performance_data$COVID.hours..W2.)
##
     Min. 1st Qu. Median
                          Mean 3rd Qu.
                                           Max.
                                                  NA's
                  1.000
##
    0.000
          1.000
                           2.884
                                  2.000 40.000
                                                    22
summary(sta302_performance_data$COVID.hours..W3.)
##
     Min. 1st Qu. Median
                           Mean 3rd Qu.
                                                  NA's
##
    0.000 0.500
                  1.000 2.333 2.000 24.000
                                                    21
summary(sta302_performance_data$COVID.hours..W4.)
##
     Min. 1st Qu. Median Mean 3rd Qu.
                                           Max.
                                                  NA's
##
    0.000 1.000 1.500 2.918 3.000 50.000
                                                    40
```

Quiz Scores 5-Number Summary

```
summary(sta302_performance_data$Quiz_1_score)
##
     Min. 1st Qu. Median
                           Mean 3rd Qu.
                                           Max.
                                                  NA's
##
    1.000 6.000 8.000 7.617 9.000 10.000
                                                    13
summary(sta302_performance_data$Quiz_2_score)
##
     Min. 1st Qu. Median
                          Mean 3rd Qu.
                                           Max.
                                                  NA's
                  8.800
                           7.422
                                 9.400 10.000
##
    0.000
          5.800
                                                    36
summary(sta302_performance_data$Quiz_3_score)
##
     Min. 1st Qu. Median
                           Mean 3rd Qu.
                                           Max.
                                                  NA's
##
    1.000 5.000
                  8.000 7.209 9.000 10.000
                                                    31
summary(sta302_performance_data$Quiz_4_score)
##
     Min. 1st Qu. Median
                          Mean 3rd Qu.
                                           Max.
                                                  NA's
##
    1.000 6.000 8.000 7.378 9.000 10.000
                                                    34
```