

# FOOKAIYAN33085625

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## Statement: Generative AI was used in this assignment

- ChatGPT was used to help to rectify an error in generating barchart by providing potential errors and solution. Prompt used was the error message shown after code has been runned. Output was how the code was modified to change and add dataframe before the barchart was generated.
- BingAI was used to bring an idea to code. Prompt used was to combine 2 columns described to the AI as an example and to combine the data in the cells under the columns to be within only 1 cell of a new column. Output was an example code.
- ChatGPT was used to find on how t-value will affect the confidence of the coefficient as a predictor. Prompt used was 'how t-value is used to get the best predictor'. Output: Generally, any t-value greater than +2 or less than - 2 is acceptable. The higher the t-value, the greater the confidence we have in the coefficient as a predictor. Low t-values are indications of low reliability of the predictive power of that coefficient.

## Set working directory

```
setwd("C:/Monash/FIT3152")
```

## Install and load the libraries used

```
library(ggplot2)
library(dplyr)

##
## Attaching package: 'dplyr'

## The following objects are masked from 'package:stats':
##
##   filter, lag

## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union

library(tidyr)
library(factoextra)

## Welcome! Want to learn more? See two factoextra-related books at https://goo.gl/ve3WBa

library(ggpubr)
library(corrplot)

## corrplot 0.92 loaded

library(reshape)
```

```
##
## Attaching package: 'reshape'

## The following objects are masked from 'package:tidyr':
##
##     expand, smiths

## The following object is masked from 'package:dplyr':
##
##     rename
```

#### Load data in PsyCoronaBaselineExtract.csv

```
rm(list = ls())
set.seed(33085625)
cvbase = read.csv("PsyCoronaBaselineExtract.csv", header = TRUE)
cvbase <- cvbase[sample(nrow(cvbase), 40000), ] # 40000 rows
```

#### Question 1a: Basic pre-processing and descriptive analysis for the corona dataset

```
dim(cvbase)

## [1] 40000    52
```

There is a total of 40,000 rows and 52 columns present in the dataset.

```
names(cvbase)

## [1] "employstatus_1"      "employstatus_2"      "employstatus_3"
## [4] "employstatus_4"      "employstatus_5"      "employstatus_6"
## [7] "employstatus_7"      "employstatus_8"      "employstatus_9"
## [10] "employstatus_10"     "isoFriends_inPerson" "isoOthPpl_inPerson"
## [13] "isoFriends_online"   "isoOthPpl_online"    "lone01"
## [16] "lone02"              "lone03"              "happy"
## [19] "lifeSat"             "MLQ"                 "bor01"
## [22] "bor02"              "bor03"              "consp01"
## [25] "consp02"            "consp03"            "rankOrdLife_1"
## [28] "rankOrdLife_2"       "rankOrdLife_3"       "rankOrdLife_4"
## [31] "rankOrdLife_5"       "rankOrdLife_6"       "c19perBeh01"
## [34] "c19perBeh02"         "c19perBeh03"         "c19RCA01"
## [37] "c19RCA02"           "c19RCA03"           "coronaClose_1"
## [40] "coronaClose_2"       "coronaClose_3"       "coronaClose_4"
## [43] "coronaClose_5"       "coronaClose_6"       "gender"
## [46] "age"                 "edu"                 "coded_country"
## [49] "c19ProSo01"          "c19ProSo02"          "c19ProSo03"
## [52] "c19ProSo04"
```

The 52 variables included in the dataset are employstatus\_1, employstatus\_2, employstatus\_3, employstatus\_4, employstatus\_5, employstatus\_6, employstatus\_7, employstatus\_8, employstatus\_9, employstatus\_10, isoFriends\_inPerson, isoOthPpl\_inPerson, isoFriends\_online, isoOthPpl\_online, lone01, lone02, lone03, happy, lifeSat, MLQ, bor01, bor02, bor03, consp01, consp02, consp03, rankOrdLife\_1, rankOrdLife\_2, rankOrdLife\_3, rankOrdLife\_4, rankOrdLife\_5, rankOrdLife\_6, c19perBeh01, c19perBeh02, c19perBeh03, c19RCA01, c19RCA02, c19RCA03, coronaClose\_1, coronaClose\_2, coronaClose\_3, coronaClose\_4, coronaClose\_5, coronaClose\_6, gender, age, edu, coded\_country, c19ProSo01, c19ProSo02, c19ProSo03 and c19ProSo04.

```
# Summary is used to obtain basic summary of each column present in the dataset 'cvbase'
# summary(cvbase)
# str(cvbase)
# The commented code above has been run in appendix section
```

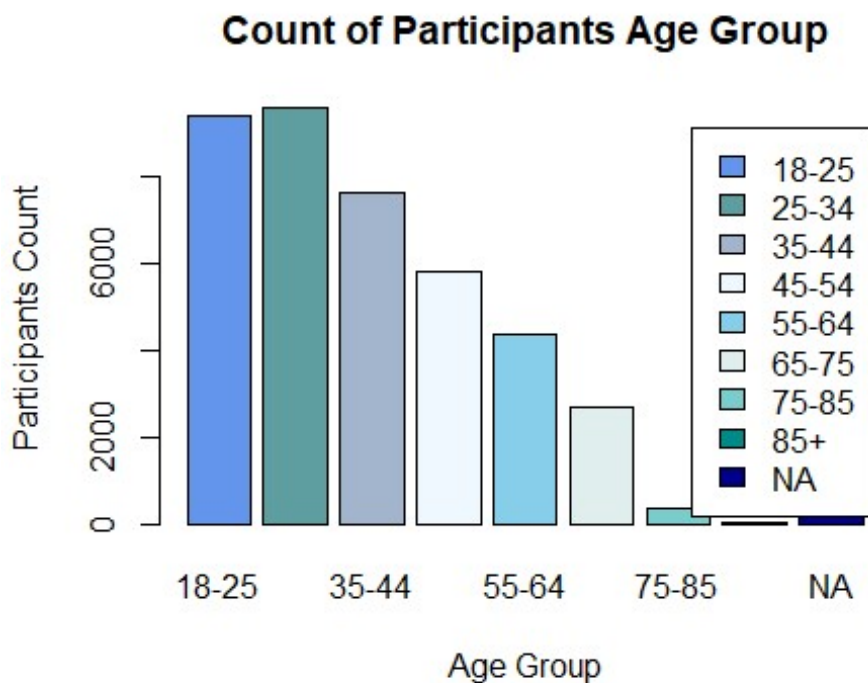
From the code above, we can see that there is some missing values ("NA") present in the dataset.

Most missing values are found in the columns `employstatus_1`, `employstatus_2`, `employstatus_3`, `employstatus_4`, `employstatus_5`, `employstatus_6`, `employstatus_7`, `employstatus_8`, `employstatus_9` and `employstatus_10` as these few columns ask on the concept of the employment status of the participants. Each options represents different employment status the participants might be in. Participants might only choose 1 of the many options provided even though they are allowed to choose multiple.

The Corona Proximity concept columns consisted of `coronaClose_1`, `coronaClose_2`, `coronaClose_3`, `coronaClose_4`, `coronaClose_5` and `coronaClose_6` which is also where the majority of the missing values are found in the dataset. These few columns ask the participants on whether they know people who have corona virus and each option represents different groups of people that have the corona virus and also a choice on not knowing anyone who have the corona virus.

Most of the responses is numerical except for `rankOrdLife_1`, `rankOrdLife_2`, `rankOrdLife_3`, `rankOrdLife_4`, `rankOrdLife_5` and `rankOrdLife_6` which responses came in alphabets and also `coded_country` which is the participants response on the country they currently live in or the country they spent most of their time in.

```
participants_age = cvbase %>% group_by(age) %>% summarise(COUNT = n())
participants_age <- as.data.frame(participants_age, row.names = NULL, optional = FALSE)
participants_age_barchart = participants_age$COUNT
names(participants_age_barchart) <- c("18-25", "25-34", "35-44", "45-54", "55-64", "65-75",
"75-85", "85+", "NA")
barplot(participants_age_barchart, main = "Count of Participants Age Group", xlab = "Age Group",
ylab = "Participants Count", col = c("cornflowerblue", "cadetblue", "lightsteelblue3",
"aliceblue", "skyblue", "azure2", "darkslategray3", "darkcyan", "darkblue"), border = "black",
legend=TRUE)
```



The code above produce a bar chart that displayed the age range of the participants. From the bar chart, we can see that most participants are at the age range of 25-34 but the age range 18-25 is also quite high. This shows that most participants are teenagers or young adults and these people are the people that mostly spent their time on their devices.

### Question 1b:

```
# Add one new column to the dataset named employment_status
# If employstatus_1 then under employment_status column put a 1
# If employstatus_1 and employstatus_2 then under employment_status column put 1, 2
# Once done, remove the columns for employstatus_1, employstatus_2, employstatus_3, employstatus_4, employstatus_5, employstatus_6, employstatus_7, employstatus_8, employstatus_9 and employstatus_10

# Adding the new column and populate it
employment_columns <- c("employstatus_1", "employstatus_2", "employstatus_3", "employstatus_4", "employstatus_5", "employstatus_6", "employstatus_7", "employstatus_8", "employstatus_9", "employstatus_10")
cvbase$employment_status <- NA
cvbase$employment_status <- apply(cvbase[, employment_columns], 1, function(row) {
  not_na <- which(!is.na(row))
  if (length(not_na) > 0) {
    paste(not_na, collapse = ", ")
  } else {
    NA
  }
})
head(cvbase$employment_status)

## [1] "4" "9" "9" "3" "10" "2"

# Removing the columns
cvbase = subset(cvbase, select = -c(employstatus_1, employstatus_2, employstatus_3, employstatus_4, employstatus_5, employstatus_6, employstatus_7, employstatus_8, employstatus_9, employstatus_10))

# Add one new column to the dataset named corona_close
# If coronaClose_1 then under corona_close column put a 1
# If coronaClose_1 and coronaClose_2 then under corona_close column put 1, 2
# Once done, remove the columns for coronaClose_1, coronaClose_2, coronaClose_3, coronaClose_4, coronaClose_5 and coronaClose_6

# Adding the new column and populate it
close_contact_column <- c("coronaClose_1", "coronaClose_2", "coronaClose_3", "coronaClose_4", "coronaClose_5", "coronaClose_6")
cvbase$corona_close <- NA
cvbase$corona_close <- apply(cvbase[, close_contact_column], 1, function(row) {
  not_na <- which(!is.na(row))
  if (length(not_na) > 0) {
    paste(not_na, collapse = ", ")
  } else {
    NA
  }
})
head(cvbase$corona_close)

## [1] "6" "6" "6" "6" "5" "6"

# Removing the columns
cvbase = subset(cvbase, select = -c(coronaClose_1, coronaClose_2, coronaClose_3, coronaClose_4, coronaClose_5, coronaClose_6))
```

## Question 2a: Focus country by Student ID 33085625 is Malaysia

*# Group 1 --> Participants from Malaysia*

```
msia = cvbase %>% filter(coded_country == "Malaysia")
dim(msia)
```

```
## [1] 548 38
```

*# Group 2 --> Participants not from Malaysia*

```
not_msia = cvbase %>% filter(coded_country != "Malaysia")
dim(not_msia)
```

```
## [1] 39452 38
```

After filtering the newly processed dataset, another filter is applied to pinpoint the responses has its coded\_country listed as Malaysia or not Malaysia. Not Malaysia includes all the other countries that is not named Malaysia like Japan, Russia and more.

After future filtering, it is known that there is 548 rows and 38 columns for responses with focus country listed as Malaysia. Whereas on the other hand, there is 39452 rows and 38 columns for responses with focus country not listed as Malaysia.

*# head() is used to get a glimpse of the data*

```
covid19ProSo <- cvbase[, c("coded_country", "c19ProSo01", "c19ProSo02", "c19ProSo03", "c19ProSo04")]
head(covid19ProSo)
```

##	coded_country	c19ProSo01	c19ProSo02	c19ProSo03	c19ProSo04
## 24995	Greece	2	0	2	-2
## 47631	Egypt	1	1	1	1
## 33923	Romania	3	0	0	3
## 4530	Italy	0	0	-1	0
## 3978	China	3	3	3	2
## 36761	Netherlands	2	-2	2	3

```
prosocal_msia <- covid19ProSo[covid19ProSo$coded_country == "Malaysia", ]
head(prosocal_msia)
```

##	coded_country	c19ProSo01	c19ProSo02	c19ProSo03	c19ProSo04
## 24272	Malaysia	0	0	0	1
## 52597	Malaysia	0	0	0	3
## 28793	Malaysia	0	2	1	1
## 43389	Malaysia	2	3	0	0
## 54798	Malaysia	1	1	0	2
## 301	Malaysia	-3	2	2	2

First, before filtering the pro-social attitudes of the participants according to country, the columns that store participants' country and their pro-social attitudes is moved to a new dataframe 'covid19ProSo' for easy view of the participants' responses. This is also to not touch on the newly processed dataset to prevent any mistakes done that would change the original dataset.

prosocal\_msia stores the Malaysian participants' pros-social attitudes responses.

```
covid19ProSo <- cvbase[, c("coded_country", "c19ProSo01", "c19ProSo02", "c19ProSo03", "c19ProSo04")]
```

```
prosocal_not_msia <- covid19ProSo[covid19ProSo$coded_country != "Malaysia", ]
head(prosocal_not_msia)
```

##	coded_country	c19ProSo01	c19ProSo02	c19ProSo03	c19ProSo04
## 24995	Greece	2	0	2	-2
## 47631	Egypt	1	1	1	1
## 33923	Romania	3	0	0	3
## 4530	Italy	0	0	-1	0
## 3978	China	3	3	3	2
## 36761	Netherlands	2	-2	2	3

prosocial\_not\_msia stores the non-Malaysian participants' pros-social attitudes responses.

```
# Malaysia Pro-Social Attitude
paste("Malaysia Pro-Social Attitude")

## [1] "Malaysia Pro-Social Attitude"

# msia c19ProSo01
paste("c19ProSo01")

## [1] "c19ProSo01"

c19ProSo01_count_values <- table(prosocial_msia$c19ProSo01)
c19ProSo01_count_values

##
## -3 -2 -1 0 1 2 3
## 10 8 18 88 130 213 80

# msia c19ProSo02
paste("c19ProSo02")

## [1] "c19ProSo02"

c19ProSo02_count_values <- table(prosocial_msia$c19ProSo02)
c19ProSo02_count_values

##
## -3 -2 -1 0 1 2 3
## 10 8 11 68 116 221 113

# msia c19ProSo03
paste("c19ProSo03")

## [1] "c19ProSo03"

c19ProSo03_count_values <- table(prosocial_msia$c19ProSo03)
c19ProSo03_count_values

##
## -3 -2 -1 0 1 2 3
## 7 15 36 127 152 142 68

# msia c19ProSo04
paste("c19ProSo04")

## [1] "c19ProSo04"

c19ProSo04_count_values <- table(prosocial_msia$c19ProSo04)
c19ProSo04_count_values
```

```
##
##   -3   -2   -1    0    1    2    3
##    7   28   23 100 114 166 109

# Non-Malaysia Pro-Social Attitude
paste("Non-Malaysia Pro-Social Attitude")

## [1] "Non-Malaysia Pro-Social Attitude"

# not msia c19ProSo01
paste("c19ProSo01")

## [1] "c19ProSo01"

c19ProSo01_count_values <- table(prosocial_not_msia$c19ProSo01)
c19ProSo01_count_values

##
##   -3   -2   -1    0    1    2    3
##  1173  1996  2397  7482  9742 11552  4982

# not msia c19ProSo02
paste("c19ProSo02")

## [1] "c19ProSo02"

c19ProSo02_count_values <- table(prosocial_not_msia$c19ProSo02)
c19ProSo02_count_values

##
##   -3   -2   -1    0    1    2    3
##  2111  3300  2942  7982  8517 10065  4402

# not msia c19ProSo03
paste("c19ProSo03")

## [1] "c19ProSo03"

c19ProSo03_count_values <- table(prosocial_not_msia$c19ProSo03)
c19ProSo03_count_values

##
##   -3   -2   -1    0    1    2    3
##  2162  3766  3625  8445  8251  8531  4524

# not msia c19ProSo04
paste("c19ProSo04")

## [1] "c19ProSo04"

c19ProSo04_count_values <- table(prosocial_not_msia$c19ProSo04)
c19ProSo04_count_values

##
##   -3   -2   -1    0    1    2    3
##  1228  1891  2000  5375  7395 11755  9659
```

From the code above, we can see that there is a total of 7 kinds of responses from the range of -3 to 3. The 7 types of responses are Strongly disagree, Disagree, Somewhat disagree, Neither agree nor disagree, Somewhat agree, Agree, Strongly agree.

If the response is -3, then the response is “Strongly disagree”.

If the response is -2, then the response is “Disagree”.

If the response is -1, then the response is “Somewhat disagree”.

If the response is 0, then the response is “Neither agree nor disagree”.

If the response is 1, then the response is “Somewhat agree”.

If the response is 2, then the response is “Agree”.

If the response is 3, then the response is “Strongly agree”.

We can see that from c19ProSo01 column, responses from Malaysia is mostly at ‘2’ where 213 “Agree” to the statement that “I am willing to help others who suffer from coronavirus.” We can also see that responses by participants not from Malaysia also mostly chose ‘2’ whereby the 11552 participants also “Agree” to the statement that “I am willing to help others who suffer from coronavirus.”

Next, it would be the c19ProSo02 column. 221 participants from Malaysia responded ‘2’ which is that they “Agree” to the statement that “I am willing to make donations to help others that suffer from coronavirus.” We can also see that responses by participants not from Malaysia also mostly chose ‘2’ whereby the 10065 participants also “Agree” to the statement that “I am willing to make donations to help others that suffer from coronavirus.”

From the c19ProSo03 column, 152 participants from Malaysia responded ‘1’ where they “Somewhat agree” to the statement that “I am willing to protect vulnerable groups from coronavirus even at my own expense.” There is a slight difference here whereby 8531 participants not from Malaysia responded that they “Agree” to the statement “I am willing to protect vulnerable groups from coronavirus even at my own expense.” But there are also a high number of participants not from Malaysia responded that they “Neither agree nor disagree” to the same statement.

Last but not least, the c19ProSo04 column. 166 participants from Malaysia responded ‘2’ which is that they “Agree” to the statement that “I am willing to make personal sacrifices to prevent the spread of coronavirus.” Most of the non-Malaysian participants which is 11755 participants also have the same response.

#### Question 2b:

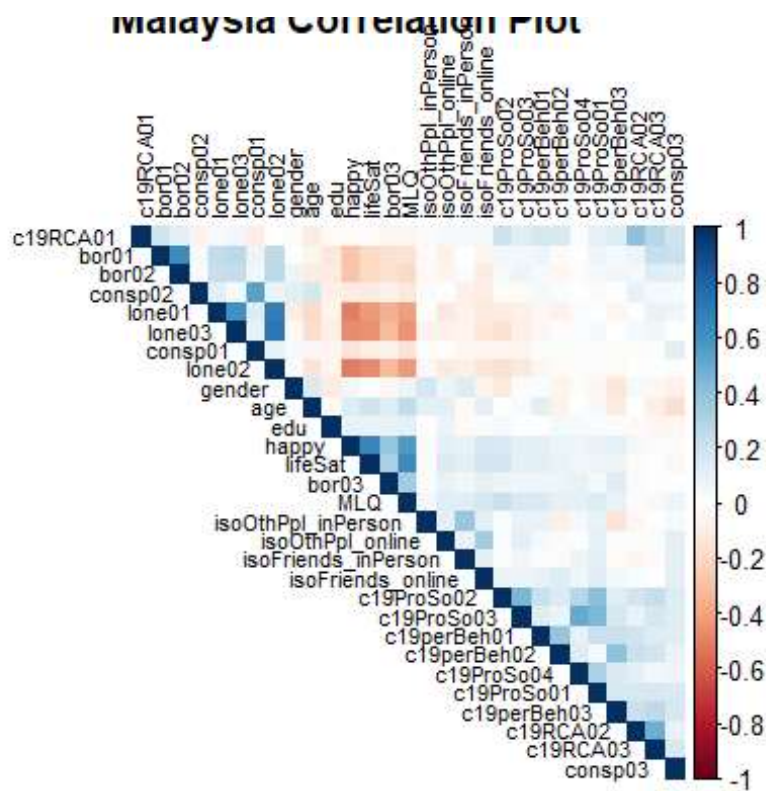
```
msia_corr <- msia
msia_corr = subset(msia_corr, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3, rankOrdLife_4, rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close))

# Compute the correlation matrix
cor_matrix <- cor(msia_corr, use = "complete.obs") # Omit NA values

# Create a correlation plot
corrplot(cor_matrix, method = "color", order = "AOE", tl.cex = 0.7, tl.col = "black",
          title = "Malaysia Correlation Plot", type = "upper")

# Add a Legend
legend("bottomleft", legend = c("Low correlation", "High correlation"),
      fill = c("white", "blue"), bty = "n", cex = 0.8, title = "Correlation Level")
```





Above presents the correlation plot for each of the predictors for Malaysia. The darker the color, the higher correlation the predictors have.

```
str(msia)

## 'data.frame':    548 obs. of  38 variables:
## $ isoFriends_inPerson: int  0 0 0 0 0 0 0 0 0 1 ...
## $ isoOthPpl_inPerson : int  0 0 1 1 1 0 2 0 0 1 ...
## $ isoFriends_online  : int  4 0 7 7 0 3 1 7 5 5 ...
## $ isoOthPpl_online   : int  1 1 0 0 0 0 3 4 3 6 ...
## $ lone01              : int  2 1 1 1 4 3 2 2 3 3 ...
## $ lone02              : int  1 1 1 1 5 1 2 2 3 2 ...
## $ lone03              : int  1 1 1 1 4 1 2 2 3 1 ...
## $ happy               : int  7 7 9 6 3 3 7 5 6 7 ...
## $ lifeSat             : int  5 4 5 5 2 2 4 4 4 5 ...
## $ MLQ                 : int  2 3 2 3 -2 -2 1 0 2 2 ...
## $ bor01               : int  1 -2 3 -2 3 3 1 2 1 3 ...
## $ bor02               : int -1 -2 2 -2 2 3 0 -2 1 2 ...
## $ bor03               : int  1 2 1 0 1 1 1 0 -2 2 ...
## $ consp01             : int  5 10 4 5 7 10 6 NA 9 3 ...
## $ consp02             : int  6 10 4 7 8 8 8 NA 8 8 ...
## $ consp03             : int  5 6 5 3 8 5 8 NA 6 6 ...
## $ rankOrdLife_1       : chr  "F" NA "F" "D" ...
## $ rankOrdLife_2       : chr  "D" NA "D" "E" ...
## $ rankOrdLife_3       : chr  "E" NA "E" "F" ...
## $ rankOrdLife_4       : chr  "B" NA "C" "C" ...
## $ rankOrdLife_5       : chr  "C" NA "A" "A" ...
## $ rankOrdLife_6       : chr  "A" NA "B" "B" ...
## $ c19perBeh01         : int  2 2 3 3 3 2 3 3 1 2 ...
## $ c19perBeh02         : int  2 2 3 3 2 2 3 3 3 3 ...
## $ c19perBeh03         : int  2 2 3 -2 -1 3 2 3 3 2 ...
## $ c19RCA01            : int  2 0 3 3 3 3 3 3 -3 3 ...
## $ c19RCA02            : int  2 2 3 3 3 3 3 3 3 2 ...
## $ c19RCA03            : int  2 2 3 2 3 3 1 2 2 3 ...
```

```
## $ gender      : int  1 2 1 1 2 1 2 1 1 1 ...
## $ age         : int  5 4 3 5 3 2 3 3 4 4 ...
## $ edu         : int  5 2 5 6 5 5 5 5 7 6 ...
## $ coded_country : chr  "Malaysia" "Malaysia" "Malaysia" "Malaysia" ...
## $ c19ProSo01   : int  0 0 0 2 1 -3 1 1 1 0 ...
## $ c19ProSo02   : int  0 0 2 3 1 2 2 -1 0 3 ...
## $ c19ProSo03   : int  0 0 1 0 0 2 1 2 2 2 ...
## $ c19ProSo04   : int  1 3 1 0 2 2 1 -1 3 3 ...
## $ employment_status : chr  "5" "5" "1" "1, 10" ...
## $ corona_close : chr  "6" "6" "5" "6" ...

msia_q2b <- msia
msia_q2b = subset(msia_q2b, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3, rankOrdLife_4, rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close))

# Best predictor for c19ProSo01
# Fit the linear model for c19ProSo01
c19ProSo01_best <- lm(c19ProSo01 ~., data = msia_q2b)
# Extract coefficients and p-values
c19ProSo01_best_summary <- summary(c19ProSo01_best)$coefficients
# Find significant predictors (p-value < 0.05)
c19ProSo01_significant_predictors <- c19ProSo01_best_summary[c19ProSo01_best_summary[, "Pr(>|t|)"] < 0.05, ]
# Sort predictors by absolute coefficient magnitude (significant)
c19ProSo01_strongest_predictors <- c19ProSo01_significant_predictors[order(abs(c19ProSo01_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
# Display the 2 highest/strongest predictors (those with ***)
c19ProSo01_strongest_predictors[1:2, ]

##           Estimate Std. Error  t value      Pr(>|t|)
## c19ProSo02 0.2559605 0.04819536  5.310895 1.723252e-07
## c19ProSo03 0.2469945 0.04894136  5.046745 6.551011e-07

# Get the R-Squared value
summary(c19ProSo01_best)

##
## Call:
## lm(formula = c19ProSo01 ~ ., data = msia_q2b)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.5360 -0.4580  0.1606  0.6137  3.3645
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.578096   0.569329  -1.015   0.31047
## isoFriends_inPerson  0.052118   0.023262   2.240   0.02555 *
## isoOthPpl_inPerson   0.011132   0.028188   0.395   0.69310
## isoFriends_online    0.030434   0.023507   1.295   0.19611
## isoOthPpl_online     0.014629   0.021591   0.678   0.49840
## lone01              0.008974   0.068040   0.132   0.89513
## lone02             -0.019923   0.072058  -0.276   0.78231
## lone03              0.173299   0.068569   2.527   0.01184 *
## happy              0.033681   0.039496   0.853   0.39425
## lifeSat           -0.046903   0.063441  -0.739   0.46010
```

```
## MLQ                0.050256    0.049029    1.025    0.30590
## bor01              0.005426    0.037210    0.146    0.88413
## bor02             -0.006778    0.037889   -0.179    0.85811
## bor03             -0.006643    0.037946   -0.175    0.86112
## consp01           0.016603    0.024270    0.684    0.49426
## consp02           0.007456    0.027997    0.266    0.79011
## consp03           0.018221    0.023081    0.789    0.43028
## c19perBeh01        0.178897    0.060762    2.944    0.00341 **
## c19perBeh02       -0.310739    0.104257   -2.981    0.00303 **
## c19perBeh03        0.112181    0.053571    2.094    0.03682 *
## c19RCA01          -0.036373    0.041648   -0.873    0.38295
## c19RCA02           0.136828    0.083725    1.634    0.10291
## c19RCA03           0.002407    0.049751    0.048    0.96143
## gender            -0.111604    0.119466   -0.934    0.35071
## age               0.035402    0.044815    0.790    0.42996
## edu               0.027532    0.046272    0.595    0.55214
## c19ProSo02         0.255961    0.048195    5.311 1.72e-07 ***
## c19ProSo03         0.246995    0.048941    5.047 6.55e-07 ***
## c19ProSo04         0.072733    0.044153    1.647    0.10020
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.062 on 447 degrees of freedom
## (72 observations deleted due to missingness)
## Multiple R-squared:  0.3358, Adjusted R-squared:  0.2941
## F-statistic: 8.069 on 28 and 447 DF,  p-value: < 2.2e-16
```

#### # Best predictor for c19ProSo02

```
c19ProSo02_best <- lm(c19ProSo02 ~., data = msia_q2b)
c19ProSo02_best_summary <- summary(c19ProSo02_best)$coefficients
c19ProSo02_significant_predictors <- c19ProSo02_best_summary[, "Pr(
>|t|)" < 0.05, ]
c19ProSo02_strongest_predictors <- c19ProSo02_significant_predictors[order(abs(c19ProSo02_s
ignificant_predictors[, "Pr(>|t|)")), decreasing = FALSE), ]
c19ProSo02_strongest_predictors[1:2, ]
```

```
##              Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo03 0.2532347 0.04637006 5.461167 7.863349e-08
## c19ProSo01 0.2318894 0.04366296 5.310895 1.723252e-07
```

```
summary(c19ProSo02_best)
```

```
##
```

```
## Call:
```

```
## lm(formula = c19ProSo02 ~ ., data = msia_q2b)
```

```
##
```

```
## Residuals:
```

```
##      Min       1Q   Median       3Q      Max
## -4.3334 -0.4782  0.1387  0.5467  3.0780
```

```
##
```

```
## Coefficients:
```

```
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    5.843e-01  5.418e-01   1.078  0.28143
## isoFriends_inPerson -2.080e-02  2.224e-02  -0.935  0.35014
## isoOthPpl_inPerson  -3.295e-03  2.683e-02  -0.123  0.90232
## isoFriends_online  -6.813e-05  2.242e-02  -0.003  0.99758
## isoOthPpl_online  -1.893e-02  2.054e-02  -0.922  0.35716
```

```
## lone01      7.366e-03  6.476e-02  0.114  0.90950
## lone02     -4.826e-03  6.859e-02 -0.070  0.94394
## lone03     -1.347e-01  6.542e-02 -2.059  0.04003 *
## happy      -6.661e-03  3.762e-02 -0.177  0.85955
## lifeSat     6.880e-02  6.033e-02  1.140  0.25474
## MLQ         6.328e-02  4.663e-02  1.357  0.17537
## bor01       9.240e-04  3.542e-02  0.026  0.97920
## bor02       2.553e-02  3.604e-02  0.708  0.47915
## bor03      -1.594e-02  3.611e-02 -0.442  0.65903
## consp01     -8.978e-03  2.311e-02 -0.389  0.69781
## consp02      1.299e-02  2.664e-02  0.488  0.62611
## consp03      4.767e-03  2.198e-02  0.217  0.82841
## c19perBeh01  8.469e-02  5.826e-02  1.454  0.14669
## c19perBeh02  1.212e-01  1.001e-01  1.211  0.22639
## c19perBeh03 -9.485e-02  5.104e-02 -1.858  0.06379 .
## c19RCA01     9.213e-02  3.944e-02  2.336  0.01992 *
## c19RCA02     4.668e-02  7.990e-02  0.584  0.55938
## c19RCA03     9.665e-02  4.713e-02  2.051  0.04089 *
## gender      -2.966e-01  1.130e-01 -2.625  0.00895 **
## age         -2.868e-02  4.266e-02 -0.672  0.50174
## edu         6.858e-03  4.406e-02  0.156  0.87637
## c19ProSo01  2.319e-01  4.366e-02  5.311  1.72e-07 ***
## c19ProSo03  2.532e-01  4.637e-02  5.461  7.86e-08 ***
## c19ProSo04  4.560e-02  4.210e-02  1.083  0.27927
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.011 on 447 degrees of freedom
## (72 observations deleted due to missingness)
## Multiple R-squared:  0.3549, Adjusted R-squared:  0.3145
## F-statistic: 8.782 on 28 and 447 DF,  p-value: < 2.2e-16

# Best predictor for c19ProSo03
c19ProSo03_best <- lm(c19ProSo03 ~., data = msia_q2b)
c19ProSo03_best_summary <- summary(c19ProSo03_best)$coefficients
c19ProSo03_significant_predictors <- c19ProSo03_best_summary[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo03_strongest_predictors <- c19ProSo03_significant_predictors[order(abs(c19ProSo03_s
ignificant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo03_strongest_predictors[1:2, ]

##              Estimate Std. Error  t value      Pr(>|t|)
## c19ProSo04  0.3753915  0.03765420  9.969445 2.846259e-21
## c19ProSo02  0.2469957  0.04522763  5.461167 7.863349e-08

summary(c19ProSo03_best)

##
## Call:
## lm(formula = c19ProSo03 ~ ., data = msia_q2b)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.1859 -0.4924  0.1367  0.5983  3.6586
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept)      0.497121    0.535282    0.929    0.3535
## isoFriends_inPerson -0.025947    0.021955   -1.182    0.2379
## isoOthPpl_inPerson  0.034794    0.026450    1.315    0.1890
## isoFriends_online  -0.012809    0.022131   -0.579    0.5630
## isoOthPpl_online   0.037022    0.020231    1.830    0.0679 .
## lone01            -0.033731    0.063940   -0.528    0.5981
## lone02            -0.067802    0.067665   -1.002    0.3169
## lone03            -0.037022    0.064892   -0.571    0.5686
## happy             -0.007895    0.037155   -0.212    0.8318
## lifeSat           -0.011119    0.059670   -0.186    0.8523
## MLQ               0.007668    0.046141    0.166    0.8681
## bor01            -0.010007    0.034976   -0.286    0.7749
## bor02             0.045973    0.035552    1.293    0.1966
## bor03            -0.030996    0.035641   -0.870    0.3849
## consp01          -0.032628    0.022774   -1.433    0.1526
## consp02           0.006595    0.026318    0.251    0.8022
## consp03           0.016865    0.021697    0.777    0.4374
## c19perBeh01       -0.030532    0.057651   -0.530    0.5967
## c19perBeh02       -0.066186    0.098923   -0.669    0.5038
## c19perBeh03        0.086437    0.050439    1.714    0.0873 .
## c19RCA01           0.044679    0.039126    1.142    0.2541
## c19RCA02          -0.184843    0.078453   -2.356    0.0189 *
## c19RCA03           0.092406    0.046563    1.985    0.0478 *
## gender            0.106968    0.112296    0.953    0.3413
## age               0.000315    0.042156    0.007    0.9940
## edu              -0.031452    0.043488   -0.723    0.4699
## c19ProSo01         0.218254    0.043246    5.047 6.55e-07 ***
## c19ProSo02         0.246996    0.045228    5.461 7.86e-08 ***
## c19ProSo04         0.375391    0.037654    9.969 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9981 on 447 degrees of freedom
## (72 observations deleted due to missingness)
## Multiple R-squared:  0.4445, Adjusted R-squared:  0.4097
## F-statistic: 12.78 on 28 and 447 DF,  p-value: < 2.2e-16
```

#### # Best predictor for c19ProSo04

```
c19ProSo04_best <- lm(c19ProSo04 ~., data = msia_q2b)
c19ProSo04_best_summary <- summary(c19ProSo04_best)$coefficients
c19ProSo04_significant_predictors <- c19ProSo04_best_summary[c19ProSo04_best_summary[, "Pr(>|t|)"] < 0.05, ]
c19ProSo04_strongest_predictors <- c19ProSo04_significant_predictors[order(abs(c19ProSo04_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo04_strongest_predictors[1:2, ]
```

```
##              Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo03  0.4845682 0.04860533  9.969445 2.846259e-21
## c19RCA03    -0.1187091 0.05283717 -2.246697 2.514670e-02
```

```
summary(c19ProSo04_best)
```

```
##
## Call:
## lm(formula = c19ProSo04 ~ ., data = msia_q2b)
##
## Residuals:
```

```
##      Min      1Q  Median      3Q      Max
## -3.7859 -0.6533  0.1128  0.6938  2.9062
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    -0.9382774   0.6071262   -1.545   0.1229
## isoFriends_inPerson  0.0281572   0.0249473    1.129   0.2596
## isoOthPpl_inPerson -0.0057822   0.0301085   -0.192   0.8478
## isoFriends_online   0.0139947   0.0251444    0.557   0.5781
## isoOthPpl_online   -0.0357561   0.0230089   -1.554   0.1209
## lone01             0.0663410   0.0726000    0.914   0.3613
## lone02             0.0448717   0.0769350    0.583   0.5600
## lone03             0.0791968   0.0736582    1.075   0.2829
## happy              0.0043484   0.0422156    0.103   0.9180
## lifeSat            -0.0056451   0.0677962   -0.083   0.9337
## MLQ                0.0345365   0.0523988    0.659   0.5102
## bor01              -0.0141496   0.0397356   -0.356   0.7219
## bor02              -0.0092274   0.0404650   -0.228   0.8197
## bor03              0.0727636   0.0403811    1.802   0.0722 .
## consp01            0.0309582   0.0258924    1.196   0.2325
## consp02            -0.0144204   0.0298952   -0.482   0.6298
## consp03            -0.0008683   0.0246677   -0.035   0.9719
## c19perBeh01        -0.0385479   0.0654951   -0.589   0.5565
## c19perBeh02         0.2041366   0.1120322    1.822   0.0691 .
## c19perBeh03         0.0321575   0.0574736    0.560   0.5761
## c19RCA01           -0.0665195   0.0444068   -1.498   0.1348
## c19RCA02            0.1986267   0.0891923    2.227   0.0264 *
## c19RCA03           -0.1187091   0.0528372   -2.247   0.0251 *
## gender             0.1042333   0.1276198    0.817   0.4145
## age                0.0009672   0.0478955    0.020   0.9839
## edu                0.0413589   0.0493995    0.837   0.4029
## c19ProSo01         0.0829613   0.0503622    1.647   0.1002
## c19ProSo02         0.0574158   0.0530024    1.083   0.2793
## c19ProSo03         0.4845682   0.0486053    9.969   <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.134 on 447 degrees of freedom
## (72 observations deleted due to missingness)
## Multiple R-squared:  0.319, Adjusted R-squared:  0.2764
## F-statistic: 7.478 on 28 and 447 DF,  p-value: < 2.2e-16
```

Due to some of the variables having data of not being non-numerical type, the non-numerical columns are removed so a linear regression model (lm) could be used to determine which of the variables are the best predictors by the number of \* present at the most right-side as the more \* there is on the rightmost side the more significant the variables are.

The linear model generated for c19ProSo01 have an R-Square value is 0.3225 which means that 32.25% of the participants from Malaysia can be explained by the independent variable when these participants from Malaysia are willing to help others who suffered from Covid19. This R-squared value of 0.3225 strongly suggests that this linear model's ability to predict is somewhat limited as while it does provide some insight, it's not highly reliable but it still have some significant predictive ability.

The linear model generated for c19ProSo02 have an R-Square value is 0.3834 which means that 38.34% of the participants from Malaysia can be explained by the independent variable when these participants from Malaysia are willing to help others who suffered from Covid19 in a form of donations. This R-squared value



of 0.3834 strongly suggests that this linear model's ability to predict is somewhat limited as while it does provide some insight, it's not highly reliable but it have a more significant predictive ability than the linear model for c19ProSo01.

The linear model generated for c19ProSo03 have an R-Square value is 0.4369 which means that 43.69% of the participants from Malaysia can be explained by the independent variable when these participants from Malaysia are willing to protect others who suffered from Covid19 with their own expenses. This R-squared value of 0.4369 strongly suggests that this linear model's ability to predict is somewhat moderate where it can explain almost half of the variability observed.

The linear model generated for c19ProSo04 have an R-Square value is 0.3023 which means that 30.23% of the participants from Malaysia can be explained by the independent variable when these participants from Malaysia are willing to make personal sacrifice to prevent spread of Covid19 virus. This R-squared value of 0.3023 strongly suggests that this linear model's ability to predict is very limited and the lowest between all 4 pro-social attribute.

When the linear model was generated for c19ProSo01 there were more than 1 predictors and/or variables that have \*\*\* present so indices was used to get the top 2 best predictors and/or variables for each c19ProSo01, c19ProSo02, c19ProSo03, c19ProSo04. 2 best predictors was displayed as if only 1 was displayed, there will be no variable name shown at output.

For the focus country Malaysia, for c19ProSo01, the best predictor is c19ProSo02 as it has the lowest p-value when compared to the other variables/predictors present. The p\_value is 2.637061e-13 which is extremely small as it is smaller than 0.05 which strongly suggest that c19ProSo02 is significantly related to c19ProSo01.

For the focus country Malaysia, for c19ProSo02, the best predictor is c19ProSo03 as it has the lowest p-value when compared to the other variables/predictors present. The p\_value is 1.508749e-14 which is extremely small as it is smaller than 0.05 which strongly suggest that c19ProSo03 is significantly related to c19ProSo02.

For the focus country Malaysia, for c19ProSo03, the best predictor is c19ProSo04. c19ProSo04 is a better predictor than c19ProSo02 because its coefficient estimate is higher so it indicated that c19ProSo04 has a stronger impact on c19ProSo03 than c19ProSo02 does impact on c19ProSo03. The t-value for c19ProSo04 is higher than that for c19ProSo02 and a higher t value indicates that the greater the confidence in c19ProSo04 as a predictor than c19ProSo02.

For the focus country Malaysia, for c19ProSo04, the best predictor is c19ProSo03. c19ProSo03 is a better predictor than c19RCA03 because its coefficient estimate is higher so it indicated that c19ProSo03 has a stronger impact on c19ProSo04 than c19RCA03 does impact on c19ProSo04. The t-value for c19ProSo03 is higher than that for c19RCA03 and a higher t value indicates that the greater the confidence in c19ProSo04 as a predictor than c19RCA03.

#### Question 2c:

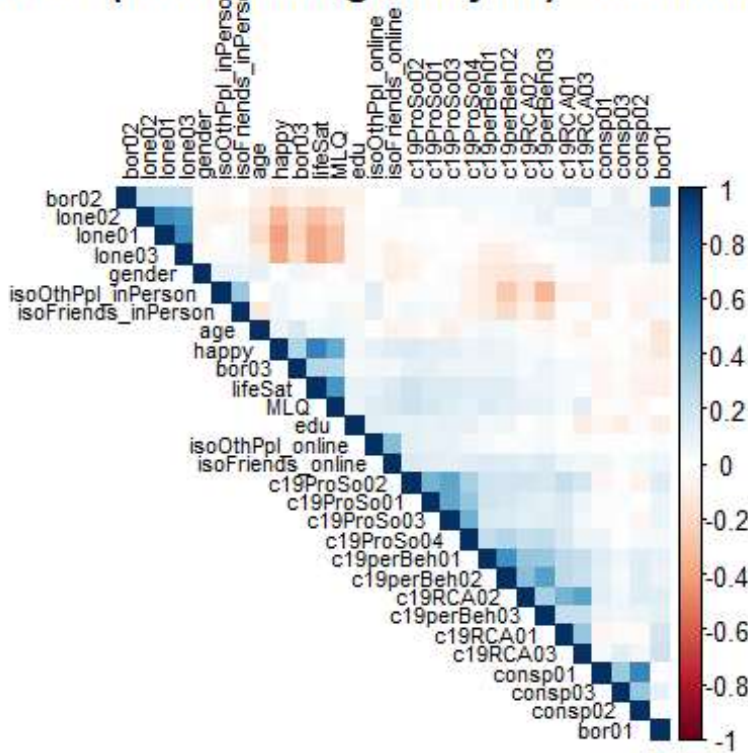
```
not_msia_corr <- not_msia
not_msia_corr = subset(not_msia_corr, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3, rankOrdLife_4, rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close))

# Compute the correlation matrix
cor_matrix <- cor(not_msia_corr, use = "complete.obs") # Omit NA values

# Create a correlation plot
corrplot(cor_matrix, method = "color", order = "AOE", tl.cex = 0.7, tl.col = "black",
          title = "Other Countries (Not including Malaysia) Correlation Plot", type = "upper")
```

```
# Add a Legend
legend("bottomleft", legend = c("Low correlation", "High correlation"),
      fill = c("white", "blue"), bty = "n", cex = 0.8, title = "Correlation Level")
```

Other Countries (NOT including Malaysia) Correlation Plot



Above presents the correlation plot for each of the predictors for other countries not including Malaysia. The darker the color, the higher correlation the predictors have.

```
str(not_msia)

## 'data.frame': 39452 obs. of 38 variables:
## $ isoFriends_inPerson: int 2 3 4 2 4 7 2 7 3 1 ...
## $ isoOthPpl_inPerson : int 0 0 3 0 2 4 3 7 3 0 ...
## $ isoFriends_online : int 7 0 5 4 3 4 5 7 4 7 ...
## $ isoOthPpl_online : int 7 0 0 4 6 0 0 7 0 3 ...
## $ lone01 : int 3 2 1 3 1 2 2 1 3 2 ...
## $ lone02 : int 2 2 1 4 1 4 4 1 3 1 ...
## $ lone03 : int 2 2 1 4 1 3 1 1 2 1 ...
## $ happy : int 1 6 10 7 8 2 7 7 6 8 ...
## $ lifeSat : int 1 4 6 4 6 2 5 4 4 4 ...
## $ MLQ : int 0 2 3 0 3 -2 1 1 -1 -1 ...
## $ bor01 : int 0 2 -3 0 -2 -1 3 2 0 1 ...
## $ bor02 : int -1 1 -3 1 -2 -1 1 2 1 0 ...
## $ bor03 : int -1 -1 3 1 3 -1 2 -1 1 -1 ...
## $ consp01 : int 10 5 8 7 NA 2 3 NA 10 4 ...
## $ consp02 : int 10 10 8 7 NA 2 3 NA 10 6 ...
## $ consp03 : int 0 5 8 7 NA 7 1 NA 9 5 ...
## $ rankOrdLife_1 : chr "D" "C" "B" "A" ...
## $ rankOrdLife_2 : chr "E" "D" "F" "C" ...
## $ rankOrdLife_3 : chr "C" "E" "C" "D" ...
## $ rankOrdLife_4 : chr "A" "B" "D" "E" ...
## $ rankOrdLife_5 : chr "B" "A" "A" "B" ...
## $ rankOrdLife_6 : chr "F" "F" "E" "F" ...
## $ c19perBeh01 : int 3 2 2 0 3 2 3 3 2 2 ...
## $ c19perBeh02 : int -2 2 3 0 3 3 3 3 2 3 ...
```



```
## $ c19perBeh03      : int  -2 1 3 1 3 3 3 2 2 3 ...
## $ c19RCA01         : int  -3 -2 -3 0 3 1 3 2 -3 1 ...
## $ c19RCA02         : int  -1 2 -1 1 3 1 3 3 -2 3 ...
## $ c19RCA03         : int  -3 2 -2 0 3 -1 1 2 1 3 ...
## $ gender           : int  2 1 2 2 2 1 1 2 2 1 ...
## $ age              : int  3 1 2 3 2 2 2 3 2 1 ...
## $ edu              : int  3 4 4 5 6 7 3 5 4 4 ...
## $ coded_country    : chr  "Greece" "Egypt" "Romania" "Italy" ...
## $ c19ProSo01       : int  2 1 3 0 3 2 0 0 1 -2 ...
## $ c19ProSo02       : int  0 1 0 0 3 -2 1 2 1 1 ...
## $ c19ProSo03       : int  2 1 0 -1 3 2 0 2 0 1 ...
## $ c19ProSo04       : int  -2 1 3 0 2 3 3 1 2 3 ...
## $ employment_status : chr  "4" "9" "9" "3" ...
## $ corona_close     : chr  "6" "6" "6" "6" ...

msia_q2c <- not_msia
msia_q2c = subset(msia_q2c, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3, rankOrdLife_4, rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close))

# Best predictor for c19ProSo01
c19ProSo01_best <- lm(c19ProSo01 ~., data = msia_q2c)
c19ProSo01_best_summary <- summary(c19ProSo01_best)$coefficients
c19ProSo01_significant_predictors <- c19ProSo01_best_summary[c19ProSo01_best_summary[, "Pr(>|t|)"] < 0.05, ]
c19ProSo01_strongest_predictors <- c19ProSo01_significant_predictors[order(abs(c19ProSo01_significant_predictors[, "Pr(>|t|)"])), decreasing = FALSE), ]
c19ProSo02_strongest_predictors[1:2, ]

##              Estimate Std. Error  t value      Pr(>|t|)
## c19ProSo03 0.2532347 0.04637006  5.461167 7.863349e-08
## c19ProSo01 0.2318894 0.04366296  5.310895 1.723252e-07

summary(c19ProSo01_best)

##
## Call:
## lm(formula = c19ProSo01 ~ ., data = msia_q2c)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.7121 -0.6055  0.1400  0.7220  4.3658
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.251897   0.055826  -4.512 6.44e-06 ***
## isoFriends_inPerson  0.003886   0.002907   1.337  0.18126
## isoOthPpl_inPerson  0.019609   0.003272   5.994 2.07e-09 ***
## isoFriends_online   0.011816   0.002918   4.049 5.15e-05 ***
## isoOthPpl_online    0.002497   0.002673   0.934  0.35011
## lone01             0.055512   0.007928   7.002 2.57e-12 ***
## lone02            -0.014326   0.006970  -2.055  0.03986 *
## lone03            -0.014190   0.007534  -1.884  0.05963 .
## happy              0.014078   0.004457   3.158  0.00159 **
## lifeSat           -0.019449   0.007690  -2.529  0.01144 *
## MLQ                0.048934   0.005166   9.472 < 2e-16 ***
## bor01              0.010262   0.004471   2.295  0.02172 *
## bor02            -0.001008   0.004471  -0.226  0.82157
```

```
## bor03          0.024173    0.004162    5.808 6.39e-09 ***
## consp01        0.013665    0.003275    4.172 3.02e-05 ***
## consp02        0.004232    0.003439    1.231 0.21851
## consp03        0.005829    0.002597    2.245 0.02479 *
## c19perBeh01     0.075607    0.007499   10.083 < 2e-16 ***
## c19perBeh02     0.040660    0.008992    4.522 6.15e-06 ***
## c19perBeh03    -0.036014    0.005306   -6.788 1.16e-11 ***
## c19RCA01        0.020596    0.004023    5.120 3.08e-07 ***
## c19RCA02        0.005333    0.006414    0.831 0.40574
## c19RCA03       -0.022928    0.004277   -5.360 8.35e-08 ***
## gender          0.069741    0.012957    5.383 7.39e-08 ***
## age             0.004622    0.004137    1.117 0.26397
## edu             0.012099    0.004521    2.676 0.00745 **
## c19ProSo02      0.210313    0.004698   44.765 < 2e-16 ***
## c19ProSo03      0.271932    0.004794   56.727 < 2e-16 ***
## c19ProSo04      0.121185    0.004820   25.141 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.184 on 35932 degrees of freedom
## (3491 observations deleted due to missingness)
## Multiple R-squared:  0.3529, Adjusted R-squared:  0.3524
## F-statistic: 700 on 28 and 35932 DF, p-value: < 2.2e-16

# Best predictor for c19ProSo02
c19ProSo02_best <- lm(c19ProSo02 ~., data = msia_q2c)
c19ProSo02_best_summary <- summary(c19ProSo02_best)$coefficients
c19ProSo02_significant_predictors <- c19ProSo02_best_summary[c19ProSo02_best_summary[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo02_strongest_predictors <- c19ProSo02_significant_predictors[order(abs(c19ProSo02_s
ignificant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo02_strongest_predictors[1:2, ]

##              Estimate Std. Error t value Pr(>|t|)
## c19ProSo01 0.2511643 0.005610749 44.76485      0
## c19ProSo03 0.3333816 0.005177592 64.38931      0

summary(c19ProSo02_best)

##
## Call:
## lm(formula = c19ProSo02 ~ ., data = msia_q2c)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.7407 -0.6920  0.1544  0.8176  5.4181
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.598424    0.060943  -9.819 < 2e-16 ***
## isoFriends_inPerson  0.025068    0.003174   7.899 2.90e-15 ***
## isoOthPpl_inPerson -0.012940    0.003577  -3.618 0.000297 ***
## isoFriends_online   0.011802    0.003189   3.701 0.000215 ***
## isoOthPpl_online    0.017238    0.002919   5.905 3.56e-09 ***
## lone01           0.039201    0.008668   4.523 6.13e-06 ***
## lone02          -0.022160    0.007617  -2.909 0.003623 **
## lone03           0.005482    0.008233   0.666 0.505511
```

```
## happy          0.015924  0.004871  3.269 0.001080 **
## lifeSat        0.071850  0.008396  8.558 < 2e-16 ***
## MLQ            0.056314  0.005645  9.976 < 2e-16 ***
## bor01          0.044335  0.004881  9.084 < 2e-16 ***
## bor02         -0.008400  0.004886 -1.719 0.085553 .
## bor03          0.005069  0.004551  1.114 0.265338
## consp01        -0.021403  0.003578 -5.982 2.23e-09 ***
## consp02        -0.015563  0.003757 -4.142 3.45e-05 ***
## consp03         0.006589  0.002838  2.322 0.020239 *
## c19perBeh01     0.037122  0.008204  4.525 6.06e-06 ***
## c19perBeh02     0.014409  0.009829  1.466 0.142663
## c19perBeh03     0.028585  0.005800  4.929 8.32e-07 ***
## c19RCA01        0.067103  0.004384 15.308 < 2e-16 ***
## c19RCA02       -0.005278  0.007009 -0.753 0.451456
## c19RCA03        0.072639  0.004660 15.587 < 2e-16 ***
## gender         -0.084679  0.014158 -5.981 2.24e-09 ***
## age            -0.016314  0.004521 -3.609 0.000308 ***
## edu            0.045601  0.004935  9.240 < 2e-16 ***
## c19ProSo01      0.251164  0.005611 44.765 < 2e-16 ***
## c19ProSo03      0.333382  0.005178 64.389 < 2e-16 ***
## c19ProSo04      0.032707  0.005311  6.158 7.43e-10 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.294 on 35932 degrees of freedom
## (3491 observations deleted due to missingness)
## Multiple R-squared:  0.3834, Adjusted R-squared:  0.3829
## F-statistic: 797.9 on 28 and 35932 DF, p-value: < 2.2e-16
```

#### # Best predictor for c19ProSo03

```
c19ProSo03_best <- lm(c19ProSo03 ~., data = msia_q2c)
c19ProSo03_best_summary <- summary(c19ProSo03_best)$coefficients
c19ProSo03_significant_predictors <- c19ProSo03_best_summary[, "Pr(>|t|)" < 0.05, ]
c19ProSo03_strongest_predictors <- c19ProSo03_significant_predictors[order(abs(c19ProSo03_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo03_strongest_predictors[1:2, ]
```

```
##           Estimate Std. Error t value Pr(>|t|)
## c19ProSo01 0.3022665 0.005328433 56.72708      0
## c19ProSo02 0.3102988 0.004819104 64.38931      0
```

```
summary(c19ProSo03_best)
```

```
##
## Call:
## lm(formula = c19ProSo03 ~ ., data = msia_q2c)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.4704 -0.7121  0.1728  0.7290  6.0624
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.214205   0.058863  -3.639 0.000274 ***
## isoFriends_inPerson  0.008083   0.003064   2.638 0.008349 **
## isoOthPpl_inPerson  0.010258   0.003451   2.973 0.002954 **
```

```
## isoFriends_online -0.004786 0.003077 -1.555 0.119897
## isoOthPpl_online 0.013027 0.002817 4.625 3.77e-06 ***
## lone01 -0.016180 0.008364 -1.934 0.053071 .
## lone02 -0.006938 0.007349 -0.944 0.345120
## lone03 0.043352 0.007940 5.460 4.79e-08 ***
## happy -0.002980 0.004700 -0.634 0.526068
## lifeSat 0.035160 0.008106 4.337 1.45e-05 ***
## MLQ -0.008942 0.005453 -1.640 0.101051
## bor01 -0.002725 0.004714 -0.578 0.563154
## bor02 0.014114 0.004713 2.995 0.002749 **
## bor03 0.006763 0.004390 1.541 0.123435
## consp01 -0.003544 0.003454 -1.026 0.304770
## consp02 -0.020259 0.003624 -5.590 2.29e-08 ***
## consp03 0.004290 0.002738 1.567 0.117180
## c19perBeh01 -0.006198 0.007917 -0.783 0.433734
## c19perBeh02 -0.019882 0.009483 -2.097 0.036024 *
## c19perBeh03 0.002620 0.005597 0.468 0.639749
## c19RCA01 0.004665 0.004243 1.100 0.271539
## c19RCA02 -0.021963 0.006761 -3.248 0.001162 **
## c19RCA03 -0.028774 0.004509 -6.382 1.77e-10 ***
## gender 0.020823 0.013666 1.524 0.127571
## age -0.067999 0.004347 -15.641 < 2e-16 ***
## edu 0.019132 0.004766 4.015 5.97e-05 ***
## c19ProSo01 0.302266 0.005328 56.727 < 2e-16 ***
## c19ProSo02 0.310299 0.004819 64.389 < 2e-16 ***
## c19ProSo04 0.309957 0.004859 63.794 < 2e-16 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.248 on 35932 degrees of freedom
## (3491 observations deleted due to missingness)
## Multiple R-squared: 0.4451, Adjusted R-squared: 0.4447
## F-statistic: 1029 on 28 and 35932 DF, p-value: < 2.2e-16

# Best predictor for c19ProSo04
c19ProSo04_best <- lm(c19ProSo04 ~., data = msia_q2c)
c19ProSo04_best_summary <- summary(c19ProSo04_best)$coefficients
c19ProSo04_significant_predictors <- c19ProSo04_best_summary[, "Pr(>|t|)" < 0.05, ]
c19ProSo04_strongest_predictors <- c19ProSo04_significant_predictors[order(abs(c19ProSo04_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo04_strongest_predictors[1:2, ]

## Estimate Std. Error t value Pr(>|t|)
## c19ProSo03 0.3282302 0.005145169 63.79386 0.000000e+00
## c19ProSo01 0.1426455 0.005673833 25.14094 2.794645e-138

summary(c19ProSo04_best)

##
## Call:
## lm(formula = c19ProSo04 ~ ., data = msia_q2c)
##
## Residuals:
## Min 1Q Median 3Q Max
## -5.6813 -0.6761 0.1383 0.8036 5.1885
##
```

```
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.378255   0.060552  -6.247 4.24e-10 ***
## isoFriends_inPerson -0.018458   0.003152  -5.856 4.79e-09 ***
## isoOthPpl_inPerson   0.004468   0.003551   1.258 0.208333
## isoFriends_online    0.019428   0.003165   6.138 8.44e-10 ***
## isoOthPpl_online    -0.010055   0.002899  -3.468 0.000525 ***
## lone01           -0.035174   0.008606  -4.087 4.37e-05 ***
## lone02             0.046010   0.007559   6.087 1.16e-09 ***
## lone03             0.020703   0.008173   2.533 0.011313 *
## happy             -0.009097   0.004836  -1.881 0.059999 .
## lifeSat            0.041340   0.008341   4.956 7.22e-07 ***
## MLQ               -0.016540   0.005611  -2.948 0.003204 **
## bor01             -0.019673   0.004850  -4.056 5.00e-05 ***
## bor02              0.019305   0.004850   3.981 6.89e-05 ***
## bor03              0.009225   0.004518   2.042 0.041169 *
## consp01            0.026531   0.003551   7.471 8.15e-14 ***
## consp02           -0.004384   0.003731  -1.175 0.239979
## consp03           -0.009806   0.002817  -3.481 0.000500 ***
## c19perBeh01         0.041610   0.008144   5.109 3.25e-07 ***
## c19perBeh02         0.143853   0.009729  14.786 < 2e-16 ***
## c19perBeh03         0.085087   0.005743  14.817 < 2e-16 ***
## c19RCA01            0.015992   0.004365   3.663 0.000249 ***
## c19RCA02            0.124097   0.006928  17.913 < 2e-16 ***
## c19RCA03           -0.060113   0.004631 -12.979 < 2e-16 ***
## gender            -0.034695   0.014062  -2.467 0.013616 *
## age                0.057287   0.004479  12.791 < 2e-16 ***
## edu                0.028752   0.004903   5.864 4.55e-09 ***
## c19ProSo01         0.142645   0.005674  25.141 < 2e-16 ***
## c19ProSo02         0.032237   0.005235   6.158 7.43e-10 ***
## c19ProSo03         0.328230   0.005145  63.794 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.285 on 35932 degrees of freedom
## (3491 observations deleted due to missingness)
## Multiple R-squared:  0.3233, Adjusted R-squared:  0.3228
## F-statistic: 613.2 on 28 and 35932 DF,  p-value: < 2.2e-16
```

The linear model generated for c19ProSo01 have an R-Square value is 0.3527 which means that 35.27% of the participants not from Malaysia can be explained by the independent variable when these participants not from Malaysia are willing to help others who suffered from Covid19. This R-squared value of 0.3527 strongly suggests that this linear model's ability to predict is somewhat limited as while it does provide some insight, it's not highly reliable but it still have some significant predictive ability.

The linear model generated for c19ProSo01 have an R-Square value is 0.383 which means that 38.3% of the participants not from Malaysia can be explained by the independent variable when these participants not from Malaysia are willing to help others who suffered from Covid19 in a form of donations. This R-squared value of 0.383 strongly suggests that this linear model's ability to predict is somewhat limited as while it does provide some insight, it's not highly reliable but it have a more significant predictive ability than the linear model for c19ProSo01.

The linear model generated for c19ProSo03 have an R-Square value is 0.4475 which means that 44.75% of the participants not from Malaysia can be explained by the independent variable when these participants not from Malaysia are willing to protect others who suffered from Covid19 with their own expenses. This R-

squared value of 0.4475 strongly suggests that this linear model's ability to predict is somewhat moderate where it can explain almost half of the variability observed.

The linear model generated for c19ProSo04 have an R-Square value is 0.3213 which means that 32.13% of the participants not from Malaysia can be explained by the independent variable when these participants not from Malaysia are willing to make personal sacrifice to prevent spread of Covid19 virus. This R-squared value of 0.3213 strongly suggests that this linear model's ability to predict is very limited and the lowest between all 4 pro-social attribute.

For the other countries, for c19ProSo01, the best predictor is c19ProSo03 as it has the lowest p-value when compared to the other variables/predictors present. The p\_value is 1.508749e-14 which is extremely small as it is smaller than 0.05 which strongly suggest that c19ProSo03 is significantly related to c19ProSo01.

For the other countries, for c19ProSo02, the best predictor is c19ProSo01. The reason on why c19ProSo03 is not the better predictor than c19ProSo01 other than the ordering is that c19ProSo03 have a larger t value than c19ProSo01 which shows that c19ProSo03 have a less stable relationship with c19ProSo02 compared to c19ProSo01 which have a lower t value which indicated that c19ProSo01 have a more stable variable relationship with c19ProSo02. The t value of c19ProSo01 is 53.75890 which is smaller than the t value of c19ProSo03 which is 77.95488.

For the other countries, for c19ProSo03, the best predictor is c19ProSo01. The reason on why c19ProSo02 is not the better predictor than c19ProSo01 other than the ordering is that c19ProSo02 have a larger t value than c19ProSo01 which shows that c19ProSo02 have a less stable relationship with c19ProSo03 compared to c19ProSo01 which have a lower t value which indicated that c19ProSo01 have a more stable variable relationship with c19ProSo03. The t value of c19ProSo01 is 68.26182 which is smaller than the t value of c19ProSo02 which is 77.95488.

For the other countries, for c19ProSo04, the best predictor is c19ProSo03 as it has the lowest p-value when compared to the other variables/predictors present. The p\_value is 0.000000e+00 which is extremely small as it is smaller than 0.05 which strongly suggests that c19ProSo03 is significantly related to c19ProSo04.

### Question 3a:

```
unique(cvbase$coded_country)
```

```
##      [1] "Greece"           "Egypt"
##      [3] "Romania"          "Italy"
##      [5] "China"            "Netherlands"
##      [7] "Spain"            "South Africa"
##      [9] "Argentina"        "Peru"
##     [11] "United States of America" "Canada"
##     [13] "United Kingdom"    "Germany"
##     [15] "Russia"            "France"
##     [17] "South Korea"       "Algeria"
##     [19] "Ukraine"           "Brazil"
##     [21] "Turkey"            "Malaysia"
##     [23] "Poland"             "Montenegro"
##     [25] "Philippines"       "Saudi Arabia"
##     [27] "Singapore"         "Chile"
##     [29] "Australia"         "Republic of Serbia"
##     [31] "Iran"              "Indonesia"
##     [33] "Japan"             "Croatia"
##     [35] "Pakistan"          "New Zealand"
##     [37] "Kosovo"            "Venezuela"
##     [39] "Kazakhstan"        "Cyprus"
##     [41] "Taiwan"            "Hong Kong S.A.R."
##     [43] "Hungary"           "Morocco"
```

```
## [45] "Trinidad and Tobago" "Moldova"
## [47] "Bangladesh" "Iraq"
## [49] "Austria" ""
## [51] "Colombia" "Vietnam"
## [53] "India" "Portugal"
## [55] "Tunisia" "El Salvador"
## [57] "Czech Republic" "Norway"
## [59] "Belgium" "Israel"
## [61] "Thailand" "Sweden"
## [63] "Palestine" "Myanmar"
## [65] "Mexico" "Jamaica"
## [67] "United Arab Emirates" "Lebanon"
## [69] "Lithuania" "Mali"
## [71] "Slovakia" "Bulgaria"
## [73] "Dominican Republic" "Laos"
## [75] "Finland" "Guatemala"
## [77] "Switzerland" "Georgia"
## [79] "Libya" "Uruguay"
## [81] "Kuwait" "Bosnia and Herzegovina"
## [83] "Luxembourg" "Oman"
## [85] "Armenia" "Ireland"
## [87] "Ecuador" "Denmark"
## [89] "Bahrain" "Slovenia"
## [91] "Albania" "Ethiopia"
## [93] "Panama" "Nigeria"
## [95] "Malta" "Jordan"
## [97] "Belarus" "Estonia"
## [99] "Cameroon" "Benin"
## [101] "Nepal" "Azerbaijan"
## [103] "Iceland" "Uzbekistan"
## [105] "Mauritius" "Cambodia"
## [107] "Costa Rica" "Kenya"
## [109] "Brunei" "Kyrgyzstan"
## [111] "Botswana" "Mongolia"
## [113] "Andorra"
```

From the output above we can see that there are 113 unique countries present in the dataset provided to complete this assignment.

```
# Load in external dataset
corona = read.csv("2021-GHS-Index-April-2022.csv", header = TRUE)

# Remove NA from corona dataset
corona <- na.omit(corona)

# Identify unique countries
# Refer to appendix

# Identify potential indicators
# Refer to appendix
```

The social, economic, health and political indicators used to identify similar countries from Malaysia is listed down as below with the column number and what are the data recorded in each of the columns.

#### Indicators used:

- [59] "X1.6.1..Vaccination.rates"



- [98] "X2.5.1a..National.support.to.conduct.contact.tracing.in.the.event.of.a.public.health.emergency"
- [270] "X6.2..Socio.economic.resilience"
- [309] "X6.5.3..Public.healthcare.spending.levels.per.capita"
- [311] "X6.5.4..Trust.in.medical.and.health.advice"

*# Columns 59, 98, 270, 309 and 311 from the corona dataset is to be used in the clustering process*

```
corona_kmeans <- kmeans(corona[, c(59, 98, 270, 309, 311)], 6, nstart = 75)
corona_kmeans_cluster <- corona_kmeans$cluster
```

*# Dataframe is used to store the countries and the cluster numbers*

```
cluster_df <- data.frame(Value = as.vector(t(corona$Country)), Cluster = corona_kmeans_cluster)
head(cluster_df)
```

```
##           Value Cluster
## 1      Afghanistan      4
## 2          Albania      3
## 3          Algeria      4
## 4          Andorra      3
## 5           Angola      4
## 6 Antigua & Barbuda      4
```

*# Find which cluster is the country Malaysia is in*

```
msia_cluster = unique(subset(cluster_df, grepl("Malaysia", Value))$Cluster)
msia_cluster
```

```
## [1] 2
```

*# Identify all countries in the same cluster as Malaysia*

```
unique(subset(cluster_df, grepl(msia_cluster, Cluster))$Value)
```

```
## [1] "Azerbaijan"      "Bangladesh"      "Burundi"
## [4] "Costa Rica"      "Egypt"           "Georgia"
## [7] "Hungary"         "India"           "Iran"
## [10] "Israel"          "Jordan"          "Kyrgyz Republic"
## [13] "Malawi"          "Malaysia"        "Mongolia"
## [16] "Philippines"     "Poland"          "Rwanda"
## [19] "Saudi Arabia"    "Serbia"          "Slovakia"
## [22] "Sri Lanka"       "Tajikistan"      "Tanzania"
## [25] "Turkmenistan"    "United Arab Emirates" "Uruguay"
## [28] "Uzbekistan"      "Zimbabwe"        "Argentina"
## [31] "Ethiopia"        "Gambia"          "Ghana"
## [34] "Kuwait"          "Myanmar"         "Portugal"
## [37] "Singapore"       "South Korea"     "Thailand"
```

There are 39 countries in the same cluster as Malaysia which is cluster 1. The 5 similar countries selected are Hungary, Iran, Philippines, Poland, and Saudi Arabia as these 5 countries are the countries positioned near Malaysia within the cluster and these countries also exist in the 'cvbase' dataset.

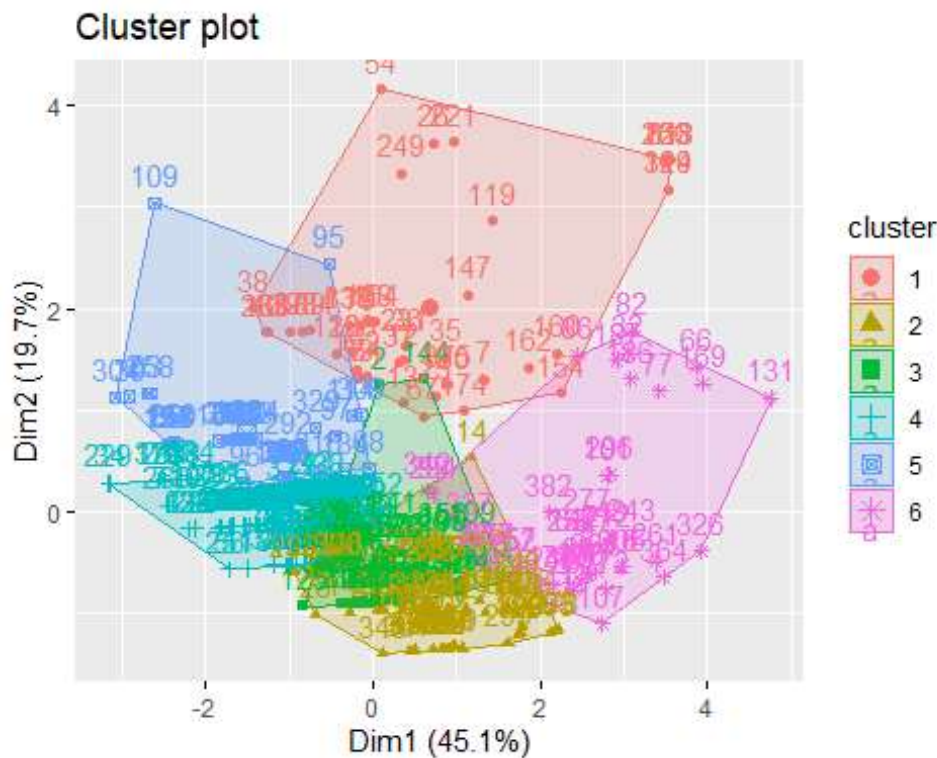
K-means clustering is used on the external dataset 'corona' with the columns 59, 98, 270, 309 and 311 is used in the clustering process. Columns 59, 98, 270, 309 and 311 from the corona dataset are the indicators selected from before. Then a dataframe is used to store the countries and the cluster numbers for easy manipulation of data on the later step. From the dataframe, I then find which cluster is the country Malaysia is in and Malaysia is in the first cluster. After that, I identify all countries in the same cluster as Malaysia as these



countries are like Malaysia for them to end up in the same cluster as Malaysia. Below is a plot on the cluster for visualisation purposes.

```
# Plot out the kmeans cluster
```

```
fviz_cluster(list(data = corona[, c(59, 98, 270, 309, 311)], cluster = corona_kmeans_cluster))
```



### Question 3b:

```
hungary = cvbase %>% filter(coded_country == "Hungary")
hungary <- subset(hungary, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3, rankOrdLife_4, rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close))
```

```
# Linear Model
```

```
hungary_lm_1 <- lm(c19ProSo01 ~., data = hungary)
```

```
# Best Predictors
```

```
# c19ProSo01
```

```
c19ProSo01_coefficients <- summary(hungary_lm_1)$coefficients
```

```
c19ProSo01_significant_predictors <- c19ProSo01_coefficients[c19ProSo01_coefficients[, "Pr(>|t|)"] < 0.05, ]
```

```
c19ProSo01_best_predictors <- c19ProSo01_significant_predictors[order(abs(c19ProSo01_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
```

```
c19ProSo01_best_predictors[1:2, ]
```

```
##              Estimate Std. Error  t value      Pr(>|t|)
## c19ProSo02  0.3686228  0.04926850  7.481916 1.110286e-12
## c19ProSo03  0.2511636  0.05065207  4.958606 1.277359e-06
```

```
# Plot responses from Hungary
```

```
hungary_plot_1 <- qplot(seq_along(hungary$c19ProSo01), hungary$c19ProSo01, main = "Participants Responses for c19ProSo01 for Hungary", xlab = "c19ProSo01", ylab = "c19ProSo01") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)
```

```
## Warning: `qplot()` was deprecated in ggplot2 3.4.0.
```

```
## This warning is displayed once every 8 hours.
```

```

## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.

# c19ProSo02
hungary_lm_2 <- lm(c19ProSo02 ~., data = hungary)
c19ProSo02_coefficients <- summary(hungary_lm_2)$coefficients
c19ProSo02_significant_predictors <- c19ProSo02_coefficients[c19ProSo02_coefficients[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo02_best_predictors <- c19ProSo02_significant_predictors[order(abs(c19ProSo02_signif
icant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo01_best_predictors[1:2, ]

##              Estimate Std. Error  t value      Pr(>|t|)
## c19ProSo02  0.3686228  0.04926850  7.481916  1.110286e-12
## c19ProSo03  0.2511636  0.05065207  4.958606  1.277359e-06

hungary_plot_2 <- qplot(seq_along(hungary$c19ProSo02), hungary$c19ProSo02, main = "Particip
ants Responses for c19ProSo02 for Hungary", xlab = "c19ProSo02", ylab = "c19ProSo02") + geo
m_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

# c19ProSo03
hungary_lm_3 <- lm(c19ProSo03 ~., data = hungary)
c19ProSo03_coefficients <- summary(hungary_lm_3)$coefficients
c19ProSo03_significant_predictors <- c19ProSo03_coefficients[c19ProSo03_coefficients[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo03_best_predictors <- c19ProSo03_significant_predictors[order(abs(c19ProSo03_signif
icant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo03_best_predictors[1:2, ]

##              Estimate Std. Error  t value      Pr(>|t|)
## c19ProSo01  0.3415897  0.06888826  4.958606  1.277359e-06
## c19ProSo04  0.2873640  0.06051383  4.748733  3.373682e-06

hungary_plot_3 <- qplot(seq_along(hungary$c19ProSo03), hungary$c19ProSo03, main = "Particip
ants Responses for c19ProSo03 for Hungary", xlab = "c19ProSo03", ylab = "c19ProSo03") + geo
m_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

# c19ProSo04
hungary_lm_4 <- lm(c19ProSo04 ~., data = hungary)
c19ProSo04_coefficients <- summary(hungary_lm_4)$coefficients
c19ProSo04_significant_predictors <- c19ProSo04_coefficients[c19ProSo04_coefficients[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo04_best_predictors <- c19ProSo04_significant_predictors[order(abs(c19ProSo04_signif
icant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo04_best_predictors[1:2, ]

##              Estimate Std. Error  t value      Pr(>|t|)
## c19ProSo03  0.2757806  0.05807458  4.748733  3.373682e-06
## c19ProSo01  0.1829573  0.06967026  2.626046  9.146617e-03

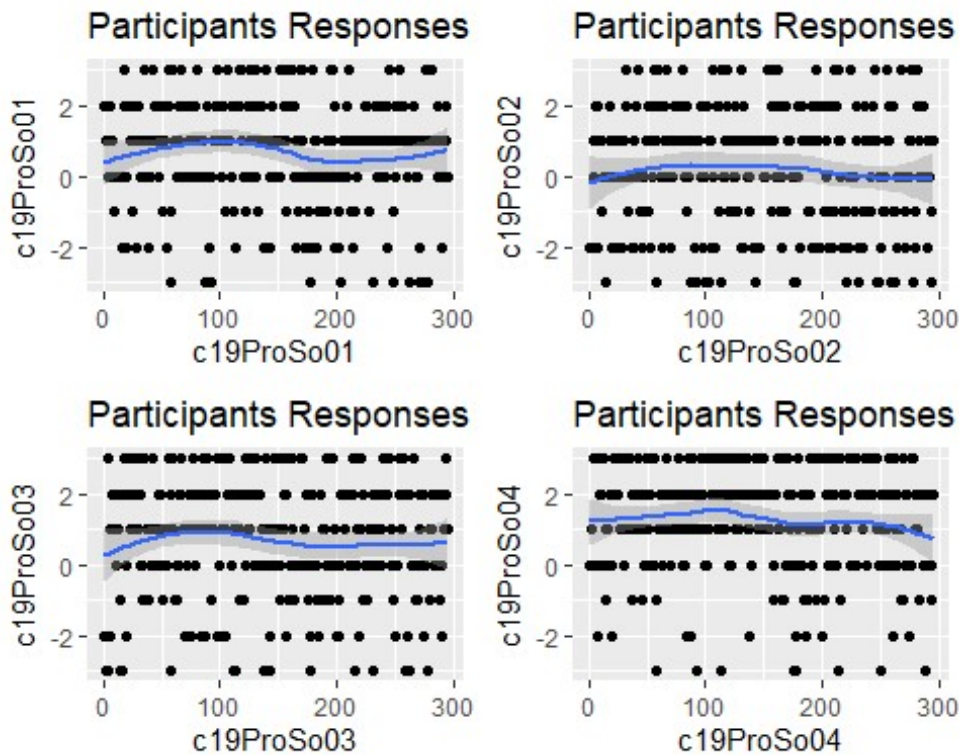
hungary_plot_4 <- qplot(seq_along(hungary$c19ProSo04), hungary$c19ProSo04, main = "Particip
ants Responses for c19ProSo04 for Hungary", xlab = "c19ProSo04", ylab = "c19ProSo04") + geo
m_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

ggarrange(hungary_plot_1, hungary_plot_2, hungary_plot_3, hungary_plot_4, ncol = 2, nrow =
2)

```

```
## Warning: Removed 1 row containing missing values or values outside the scale range
## (`geom_point()`).

## Warning: Removed 1 row containing missing values or values outside the scale range
## (`geom_point()`).
## Removed 1 row containing missing values or values outside the scale range
## (`geom_point()`).
## Removed 1 row containing missing values or values outside the scale range
## (`geom_point()`).
```



For c19ProSo01, participants from Hungary responses fluctuated between 0 and 1, which means that most participants from Hungary either “Neither agree nor disagree” or “Somewhat agree” to the statement that “I am willing to help others who suffer from coronavirus.”

For c19ProSo02, participants from Hungary responded 0, which means that most participants from Hungary “Neither agree nor disagree” to the statement that “I am willing to make donations to help others that suffer from coronavirus.”

For c19ProSo03, participants from Hungary responses fluctuated between 0 and 1, which means that most participants from Hungary either “Neither agree nor disagree” or “Somewhat agree” to the statement that “I am willing to protect vulnerable groups from coronavirus even at my own expense.”

For c19ProSo04, participants from Hungary responses fluctuated between 1 and 2, which means that most participants from Hungary either “Somewhat agree” or “Agree” to the statement that “I am willing to make personal sacrifices to prevent the spread of coronavirus.”

### Best Predictors for Hungary:

- The best predictor for c19ProSo01 is c19ProSo02 with a p-value of 1.110286e-12 which is less than 0.05.
- The best predictor for c19ProSo02 is c19ProSo02 with a p-value of 1.110286e-12 which is less than 0.05. But c19ProSo02 can't be its own best predictor so c19ProSo03 would be the best predictor for c19ProSo02 instead with a p-value of 1.277359e-06 which is still less than 0.05 so its still a strong predictor for c19ProSo02.

- The best predictor for c19ProSo03 is c19ProSo01 with a p-value of 1.277359e-06 which is less than 0.05.
- The best predictor for c19ProSo04 is c19ProSo02 with a p-value of 3.373682e-06 which is less than 0.05.

```
iran = cvbase %>% filter(coded_country == "Iran")
iran <- subset(iran, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3, rankOrdLife_4,
rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close))

iran_lm_1 <- lm(c19ProSo01 ~., data = iran)
c19ProSo01_coefficients <- summary(iran_lm_1)$coefficients
c19ProSo01_significant_predictors <- c19ProSo01_coefficients[c19ProSo01_coefficients[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo01_best_predictors <- c19ProSo01_significant_predictors[order(abs(c19ProSo01_signif
icant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo01_best_predictors[1:2, ]

##           Estimate Std. Error   t value    Pr(>|t|)
## consp02 -0.3550268  0.1057367 -3.357651 0.001033475
## age      0.2486805  0.1012302  2.456585 0.015355829

iran_plot_1 <- qplot(seq_along(iran$c19ProSo01), iran$c19ProSo01, main = "Participants Resp
onses for c19ProSo01 for Iran", xlab = "c19ProSo01", ylab = "c19ProSo01") + geom_smooth(met
hod = "loess", formula = y ~ x, na.rm = TRUE)

iran_lm_2 <- lm(c19ProSo02 ~., data = iran)
c19ProSo02_coefficients <- summary(iran_lm_2)$coefficients
c19ProSo02_significant_predictors <- c19ProSo02_coefficients[c19ProSo02_coefficients[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo02_best_predictors <- c19ProSo02_significant_predictors[order(abs(c19ProSo02_signif
icant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo02_best_predictors[1:2, ]

##           Estimate Std. Error   t value    Pr(>|t|)
## c19ProSo03 0.7540213 0.06181530 12.197973 3.103176e-23
## consp02     0.2008956 0.07588172  2.647484 9.120422e-03

iran_plot_2 <- qplot(seq_along(iran$c19ProSo02), iran$c19ProSo02, main = "Participants Resp
onses for c19ProSo02 for Iran", xlab = "c19ProSo02", ylab = "c19ProSo02") + geom_smooth(met
hod = "loess", formula = y ~ x, na.rm = TRUE)

iran_lm_3 <- lm(c19ProSo03 ~., data = iran)
c19ProSo03_coefficients <- summary(iran_lm_3)$coefficients
c19ProSo03_significant_predictors <- c19ProSo03_coefficients[c19ProSo03_coefficients[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo03_best_predictors <- c19ProSo03_significant_predictors[order(abs(c19ProSo03_signif
icant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo03_best_predictors[1:2, ]

##           Estimate Std. Error   t value    Pr(>|t|)
## c19ProSo02     0.7103530 0.05823534 12.197973 3.103176e-23
## iso0thPpl_online 0.1010448 0.03942672  2.562851 1.152980e-02

iran_plot_3 <- qplot(seq_along(iran$c19ProSo03), iran$c19ProSo03, main = "Participants Resp
onses for c19ProSo03 for Iran", xlab = "c19ProSo03", ylab = "c19ProSo03") + geom_smooth(met
hod = "loess", formula = y ~ x, na.rm = TRUE)
```



```
iran_lm_4 <- lm(c19ProSo04 ~., data = iran)
c19ProSo04_coefficients <- summary(iran_lm_4)$coefficients
c19ProSo04_significant_predictors <- c19ProSo04_coefficients[c19ProSo04_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo04_best_predictors <- c19ProSo04_significant_predictors[order(abs(c19ProSo04_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo04_best_predictors[1:2, ]

##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo01 0.2071479 0.08471397 2.445262 0.01582383
## c19ProSo02 0.2890333 0.11995751 2.409464 0.01738797

iran_plot_4 <- qplot(seq_along(iran$c19ProSo04), iran$c19ProSo04, main = "Participants Responses for c19ProSo04 for Iran", xlab = "c19ProSo04", ylab = "c19ProSo04") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

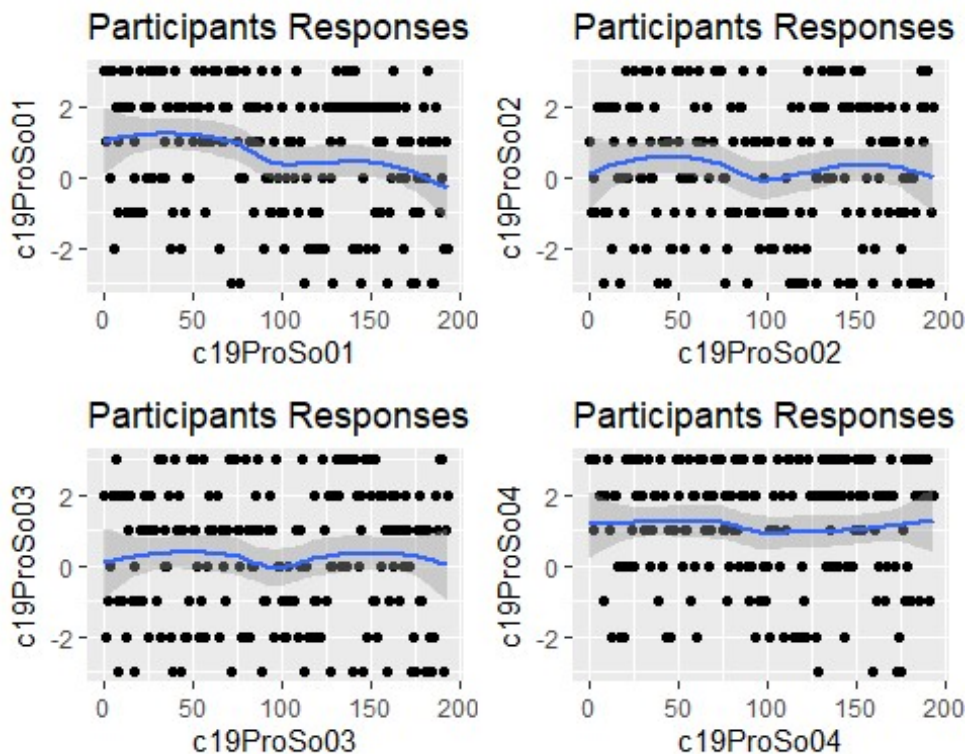
ggarrange(iran_plot_1, iran_plot_2, iran_plot_3, iran_plot_4, ncol = 2, nrow = 2)

## Warning: Removed 3 rows containing missing values or values outside the scale range
## (`geom_point()`).

## Warning: Removed 2 rows containing missing values or values outside the scale range
## (`geom_point()`).

## Warning: Removed 4 rows containing missing values or values outside the scale range
## (`geom_point()`).

## Warning: Removed 3 rows containing missing values or values outside the scale range
## (`geom_point()`).
```



For c19ProSo01, participants from Iran responses fluctuated between 0 and 1, which means that most participants from Iran either “Neither agree nor disagree” or “Somewhat agree” to the statement that “I am willing to help others who suffer from coronavirus.”

For c19ProSo02, participants from Iran responded 0, which means that most participants from Iran “Neither agree nor disagree” to the statement that “I am willing to make donations to help others that suffer from coronavirus.”

For c19ProSo03, participants from Iran responded 0 which means that most participants from Iran “Neither agree nor disagree” to the statement that “I am willing to protect vulnerable groups from coronavirus even at my own expense.”

For c19ProSo04, participants from Iran responded 1, which means that most participants from Iran “Somewhat agree” to the statement that “I am willing to make personal sacrifices to prevent the spread of coronavirus.”

#### Best Predictors for Iran:

- The best predictor for c19ProSo01 is consp02 with a p-value of 0.001033475 which is less than 0.05.
- The best predictor for c19ProSo02 is c19ProSo03 with a p-value of 3.103176e-23 which is less than 0.05.
- The best predictor for c19ProSo03 is c19ProSo02 with a p-value of 3.103176e-23 which is less than 0.05.
- The best predictor for c19ProSo04 is c19ProSo01 with a p-value of 0.01582383 which is less than 0.05.

```
philippines = cvbase %>% filter(coded_country == "Philippines")
philippines <- subset(philippines, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3,
rankOrdLife_4, rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close
))
```

```
philippines_lm_1 <- lm(c19ProSo01 ~., data = philippines)
c19ProSo01_coefficients <- summary(philippines_lm_1)$coefficients
c19ProSo01_significant_predictors <- c19ProSo01_coefficients[c19ProSo01_coefficients[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo01_best_predictors <- c19ProSo01_significant_predictors[order(abs(c19ProSo01_signif
icant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo01_best_predictors[1:2, ]
```

```
##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo02 0.3403081 0.03307542 10.288852 1.568513e-23
## c19ProSo03 0.2028857 0.02874132  7.059025 3.385106e-12
```

```
philippines_plot_1 <- qplot(seq_along(philippines$c19ProSo01), philippines$c19ProSo01, main
= "Participants Responses for c19ProSo01 for Philippines", xlab = "c19ProSo01", ylab = "c19
ProSo01") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)
```

```
philippines_lm_2 <- lm(c19ProSo02 ~., data = philippines)
c19ProSo02_coefficients <- summary(philippines_lm_2)$coefficients
c19ProSo02_significant_predictors <- c19ProSo02_coefficients[c19ProSo02_coefficients[, "Pr(
>|t|)"] < 0.05, ]
c19ProSo02_best_predictors <- c19ProSo02_significant_predictors[order(abs(c19ProSo02_signif
icant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo02_best_predictors[1:2, ]
```

```
##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo03 0.2857587 0.02668230 10.70967 2.977494e-25
## c19ProSo01 0.3136252 0.03048204 10.28885 1.568513e-23
```

```
philippines_plot_2 <- qplot(seq_along(philippines$c19ProSo02), philippines$c19ProSo02, main
= "Participants Responses for c19ProSo02 for Philippines", xlab = "c19ProSo02", ylab = "c19
ProSo02") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)
```

```

philippines_lm_3 <- lm(c19ProSo03 ~., data = philippines)
c19ProSo03_coefficients <- summary(philippines_lm_3)$coefficients
c19ProSo03_significant_predictors <- c19ProSo03_coefficients[, "Pr(>|t|)" < 0.05, ]
c19ProSo03_best_predictors <- c19ProSo03_significant_predictors[order(abs(c19ProSo03_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo03_best_predictors[1:2, ]

##           Estimate Std. Error   t value    Pr(>|t|)
## c19ProSo02 0.4010978 0.03745192 10.709672 2.977494e-25
## c19ProSo04 0.2349850 0.03040284  7.729046 2.934349e-14

philippines_plot_3 <- qplot(seq_along(philippines$c19ProSo03), philippines$c19ProSo03, main = "Participants Responses for c19ProSo03 for Philippines", xlab = "c19ProSo03", ylab = "c19ProSo03") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

philippines_lm_4 <- lm(c19ProSo04 ~., data = philippines)
c19ProSo04_coefficients <- summary(philippines_lm_4)$coefficients
c19ProSo04_significant_predictors <- c19ProSo04_coefficients[, "Pr(>|t|)" < 0.05, ]
c19ProSo04_best_predictors <- c19ProSo04_significant_predictors[order(abs(c19ProSo04_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo04_best_predictors[1:2, ]

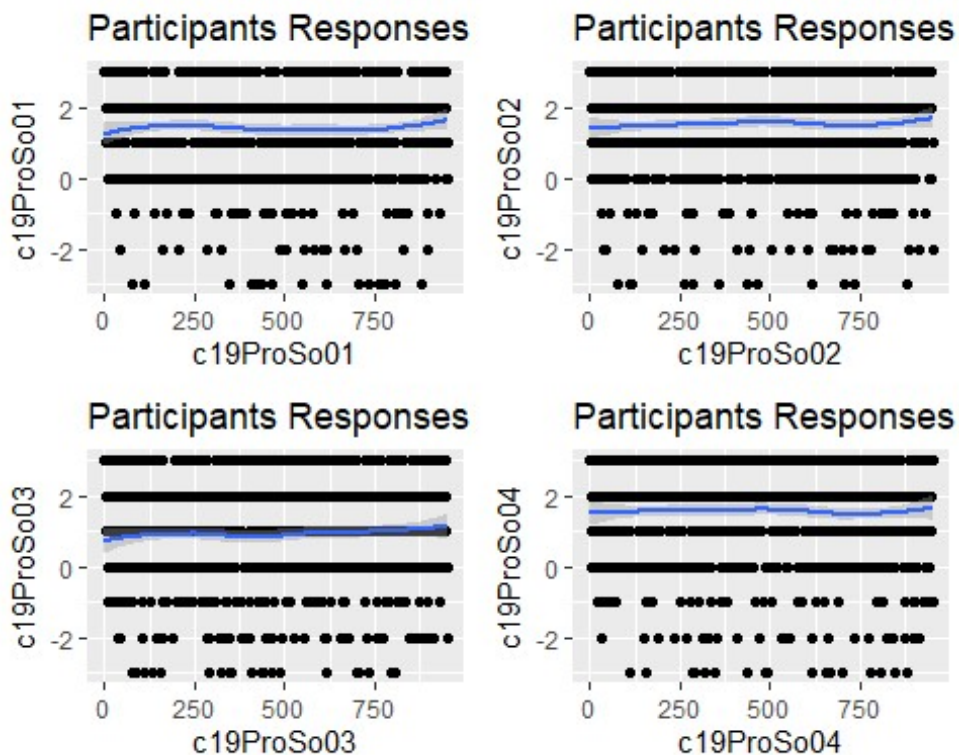
##           Estimate Std. Error   t value    Pr(>|t|)
## c19ProSo03 0.2688072 0.03477883  7.729046 2.934349e-14
## c19ProSo01 0.2129773 0.04023628  5.293165 1.516511e-07

philippines_plot_4 <- qplot(seq_along(philippines$c19ProSo04), philippines$c19ProSo04, main = "Participants Responses for c19ProSo04 for Philippines", xlab = "c19ProSo04", ylab = "c19ProSo04") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

ggarrange(philippines_plot_1, philippines_plot_2, philippines_plot_3, philippines_plot_4, ncol = 2, nrow = 2)

## Warning: Removed 1 row containing missing values or values outside the scale range
## (`geom_point()`).
## Removed 1 row containing missing values or values outside the scale range
## (`geom_point()`).

```



For c19ProSo01, participants from Philippines responses fluctuated between 1 and 2, which means that most participants from Philippines either "Somewhat agree" or "Agree" to the statement that "I am willing to help others who suffer from coronavirus."

For c19ProSo02, participants from Philippines responses fluctuated between 1 and 2, which means that most participants from Philippines either "Somewhat agree" or "Agree" to the statement that "I am willing to make donations to help others that suffer from coronavirus."

For c19ProSo03, participants from Philippines responded 1 which means that most participants from Philippines "Somewhat agree" to the statement that "I am willing to protect vulnerable groups from coronavirus even at my own expense."

For c19ProSo04, participants from Philippines responses fluctuated between 1 and 2, which means that most participants from Philippines either "Somewhat agree" or "Agree" to the statement that "I am willing to make personal sacrifices to prevent the spread of coronavirus."

#### Best Predictors for Philippines:

- The best predictor for c19ProSo01 is c19ProSo02 with a p-value of 1.568513e-23 which is less than 0.05.
- The best predictor for c19ProSo02 is c19ProSo03 with a p-value of 2.977494e-25 which is less than 0.05.
- The best predictor for c19ProSo03 is c19ProSo02 with a p-value of 2.977494e-25 which is less than 0.05.
- The best predictor for c19ProSo04 is c19ProSo03 with a p-value of 2.934349e-14 which is less than 0.05.

```
poland = cvbase %>% filter(coded_country == "Poland")
poland <- subset(poland, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3, rankOrdLife_4, rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close))

poland_lm_1 <- lm(c19ProSo01 ~., data = poland)
```



```

c19ProSo01_coefficients <- summary(poland_lm_1)$coefficients
c19ProSo01_significant_predictors <- c19ProSo01_coefficients[c19ProSo01_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo01_best_predictors <- c19ProSo01_significant_predictors[order(abs(c19ProSo01_significant_predictors[, "Pr(>|t|)"])), decreasing = FALSE), ]
c19ProSo01_best_predictors[1:2, ]

##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo03 0.2609982 0.04048517 6.446761 3.262814e-10
## c19ProSo02 0.2136891 0.03589277 5.953540 5.703087e-09

poland_plot_1 <- qplot(seq_along(poland$c19ProSo01), poland$c19ProSo01, main = "Participant
s Responses for c19ProSo01 for Poland", xlab = "c19ProSo01", ylab = "c19ProSo01") + geom_smo
oth(method = "loess", formula = y ~ x, na.rm = TRUE)

poland_lm_2 <- lm(c19ProSo02 ~., data = poland)
c19ProSo02_coefficients <- summary(poland_lm_2)$coefficients
c19ProSo02_significant_predictors <- c19ProSo02_coefficients[c19ProSo02_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo02_best_predictors <- c19ProSo02_significant_predictors[order(abs(c19ProSo02_significant_predictors[, "Pr(>|t|)"])), decreasing = FALSE), ]
c19ProSo02_best_predictors[1:2, ]

##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo01 0.3774541 0.06339993 5.953540 5.703087e-09
## edu        0.1641470 0.04658746 3.523417 4.747146e-04

poland_plot_2 <- qplot(seq_along(poland$c19ProSo02), poland$c19ProSo02, main = "Participant
s Responses for c19ProSo02 for Poland", xlab = "c19ProSo02", ylab = "c19ProSo02") + geom_smo
oth(method = "loess", formula = y ~ x, na.rm = TRUE)

poland_lm_3 <- lm(c19ProSo03 ~., data = poland)
c19ProSo03_coefficients <- summary(poland_lm_3)$coefficients
c19ProSo03_significant_predictors <- c19ProSo03_coefficients[c19ProSo03_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo03_best_predictors <- c19ProSo03_significant_predictors[order(abs(c19ProSo03_significant_predictors[, "Pr(>|t|)"])), decreasing = FALSE), ]
c19ProSo03_best_predictors[1:2, ]

##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo04 0.4643751 0.04670357 9.943032 5.539451e-21
## c19ProSo01 0.3573870 0.05543667 6.446761 3.262814e-10

poland_plot_3 <- qplot(seq_along(poland$c19ProSo03), poland$c19ProSo03, main = "Participant
s Responses for c19ProSo03 for Poland", xlab = "c19ProSo03", ylab = "c19ProSo03") + geom_smo
oth(method = "loess", formula = y ~ x, na.rm = TRUE)

poland_lm_4 <- lm(c19ProSo04 ~., data = poland)
c19ProSo04_coefficients <- summary(poland_lm_4)$coefficients
c19ProSo04_significant_predictors <- c19ProSo04_coefficients[c19ProSo04_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo04_best_predictors <- c19ProSo04_significant_predictors[order(abs(c19ProSo04_significant_predictors[, "Pr(>|t|)"])), decreasing = FALSE), ]
c19ProSo04_best_predictors[1:2, ]

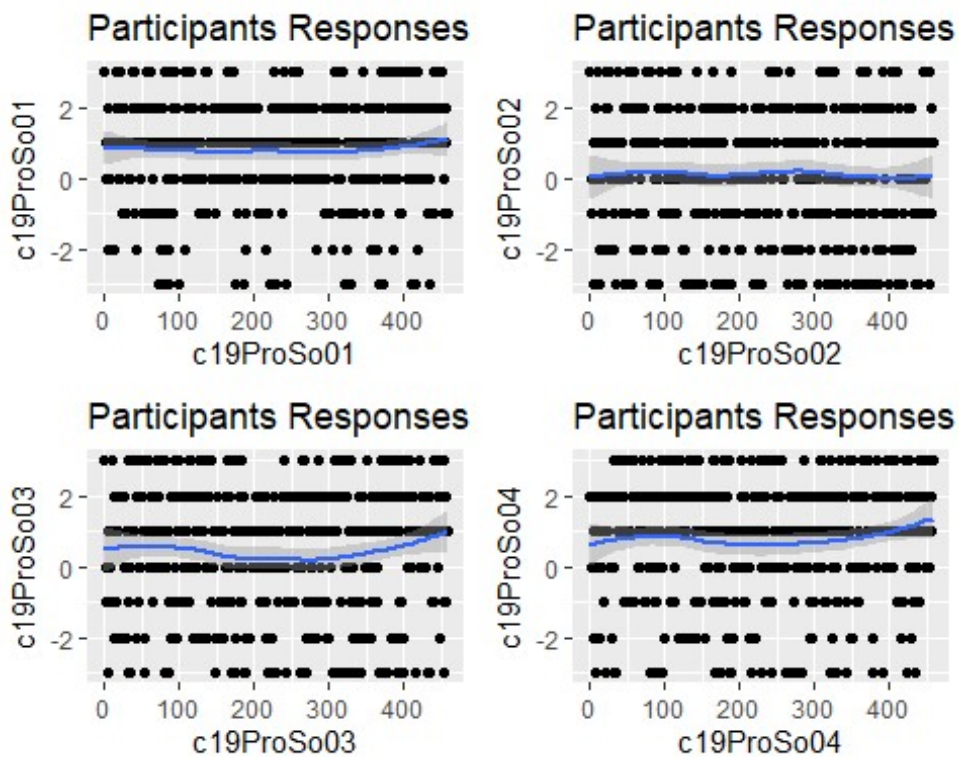
##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo03 0.4233683 0.04257939 9.943032 5.539451e-21
## c19RCA02    0.1762937 0.06408186 2.751070 6.207325e-03

```

```
poland_plot_4 <- qplot(seq_along(poland$c19ProSo04), poland$c19ProSo04, main = "Participant  
s Responses for c19ProSo04 for Poland", xlab = "c19ProSo04", ylab = "c19ProSo04") + geom_s  
mooth(method = "loess", formula = y ~ x, na.rm = TRUE)
```

```
ggarrange(poland_plot_1, poland_plot_2, poland_plot_3, poland_plot_4, ncol = 2, nrow = 2)
```

```
## Warning: Removed 4 rows containing missing values or values outside the scale range  
## (`geom_point()`).  
## Removed 4 rows containing missing values or values outside the scale range  
## (`geom_point()`).  
## Removed 4 rows containing missing values or values outside the scale range  
## (`geom_point()`).  
## Removed 4 rows containing missing values or values outside the scale range  
## (`geom_point()`).
```



For c19ProSo01, participants from Poland responded 1, which means that most participants from Poland “Agree” to the statement that “I am willing to help others who suffer from coronavirus.”

For c19ProSo02, participants from Poland responded 0, which means that most participants from Poland “Neither agree nor disagree” to the statement that “I am willing to make donations to help others that suffer from coronavirus.”

For c19ProSo03, participants from Poland responses fluctuated between 0 and 1 which means that most participants from Poland either “Neither agree nor disagree” or “Somewhat agree” to the statement that “I am willing to protect vulnerable groups from coronavirus even at my own expense.”

For c19ProSo04, participants from Poland responses fluctuated around 1, which means that most participants from Poland “Somewhat agree” to the statement that “I am willing to make personal sacrifices to prevent the spread of coronavirus.”

#### Best Predictors for Poland:

- The best predictor for c19ProSo01 is c19ProSo03 with a p-value of 3.262814e-10 which is less than 0.05.

- The best predictor for c19ProSo02 is c19ProSo01 with a p-value of 5.703087e-09 which is less than 0.05.
- The best predictor for c19ProSo03 is c19ProSo04 with a p-value of 5.539451e-21 which is less than 0.05.
- The best predictor for c19ProSo04 is c19ProSo03 with a p-value of 5.539451e-21 which is less than 0.05.

```
saudi_arabia = cvbase %>% filter(coded_country == "Saudi Arabia")
saudi_arabia <- subset(saudi_arabia, select = -c(rankOrdLife_1, rankOrdLife_2, rankOrdLife_3, rankOrdLife_4, rankOrdLife_5, rankOrdLife_6, coded_country, employment_status, corona_close))

saudi_arabia_lm_1 <- lm(c19ProSo01 ~., data = saudi_arabia)
c19ProSo01_coefficients <- summary(saudi_arabia_lm_1)$coefficients
c19ProSo01_significant_predictors <- c19ProSo01_coefficients[c19ProSo01_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo01_best_predictors <- c19ProSo01_significant_predictors[order(abs(c19ProSo01_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo01_best_predictors[1:2, ]

##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo02 0.3813108 0.04294421 8.879214 4.560679e-18
## c19ProSo04 0.2224086 0.03626050 6.133635 1.363727e-09

saudi_arabia_plot_1 <- qplot(seq_along(saudi_arabia$c19ProSo04), saudi_arabia$c19ProSo04, main = "Participants Responses for c19ProSo04 for Saudi Arabia", xlab = "c19ProSo04", ylab = "c19ProSo04") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

saudi_arabia_lm_2 <- lm(c19ProSo02 ~., data = saudi_arabia)
c19ProSo02_coefficients <- summary(saudi_arabia_lm_2)$coefficients
c19ProSo02_significant_predictors <- c19ProSo02_coefficients[c19ProSo02_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo02_best_predictors <- c19ProSo02_significant_predictors[order(abs(c19ProSo02_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo02_best_predictors[1:2, ]

##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo03 0.2916364 0.02938128 9.925926 5.998479e-22
## c19ProSo01 0.2404651 0.02708180 8.879214 4.560679e-18

saudi_arabia_plot_2 <- qplot(seq_along(saudi_arabia$c19ProSo04), saudi_arabia$c19ProSo04, main = "Participants Responses for c19ProSo04 for Saudi Arabia", xlab = "c19ProSo04", ylab = "c19ProSo04") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

saudi_arabia_lm_3 <- lm(c19ProSo03 ~., data = saudi_arabia)
c19ProSo03_coefficients <- summary(saudi_arabia_lm_3)$coefficients
c19ProSo03_significant_predictors <- c19ProSo03_coefficients[c19ProSo03_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo03_best_predictors <- c19ProSo03_significant_predictors[order(abs(c19ProSo03_significant_predictors[, "Pr(>|t|)"]), decreasing = FALSE), ]
c19ProSo03_best_predictors[1:2, ]

##           Estimate Std. Error  t value    Pr(>|t|)
## c19ProSo04 0.3600643 0.03128383 11.509597 1.997604e-28
## c19ProSo02 0.3841074 0.03869739 9.925926 5.998479e-22
```

```
saudi_arabia_plot_3 <- qplot(seq_along(saudi_arabia$c19ProSo04), saudi_arabia$c19ProSo04, main = "Participants Responses for c19ProSo04 for Saudi Arabia", xlab = "c19ProSo04", ylab = "c19ProSo04") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

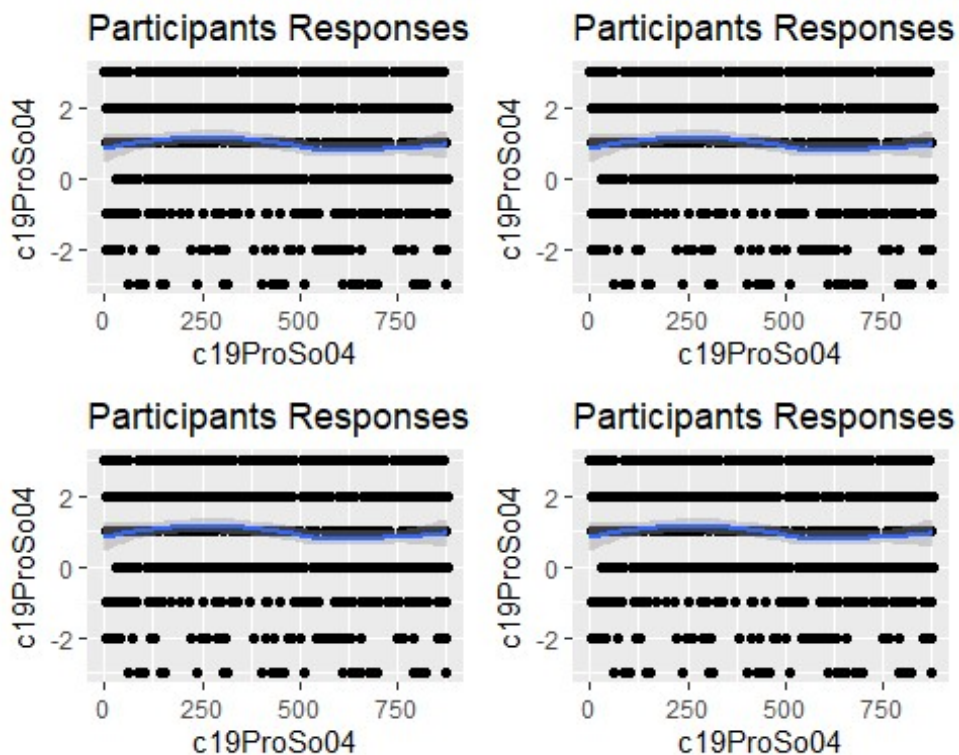
saudi_arabia_lm_4 <- lm(c19ProSo04 ~., data = saudi_arabia)
c19ProSo04_coefficients <- summary(saudi_arabia_lm_4)$coefficients
c19ProSo04_significant_predictors <- c19ProSo04_coefficients[c19ProSo04_coefficients[, "Pr(>|t|)"] < 0.05, ]
c19ProSo04_best_predictors <- c19ProSo04_significant_predictors[order(abs(c19ProSo04_significant_predictors[, "Pr(>|t|)"])), decreasing = FALSE), ]
c19ProSo04_best_predictors[1:2, ]

##           Estimate Std. Error   t value    Pr(>|t|)
## c19ProSo03 0.4027592 0.03499334 11.509597 1.997604e-28
## c19ProSo01 0.2066336 0.03368861  6.133635 1.363727e-09

saudi_arabia_plot_4 <- qplot(seq_along(saudi_arabia$c19ProSo04), saudi_arabia$c19ProSo04, main = "Participants Responses for c19ProSo04 for Saudi Arabia", xlab = "c19ProSo04", ylab = "c19ProSo04") + geom_smooth(method = "loess", formula = y ~ x, na.rm = TRUE)

ggarrange(saudi_arabia_plot_1, saudi_arabia_plot_2, saudi_arabia_plot_3, saudi_arabia_plot_4, ncol = 2, nrow = 2)

## Warning: Removed 4 rows containing missing values or values outside the scale range
## (`geom_point()`).
## Removed 4 rows containing missing values or values outside the scale range
## (`geom_point()`).
## Removed 4 rows containing missing values or values outside the scale range
## (`geom_point()`).
## Removed 4 rows containing missing values or values outside the scale range
## (`geom_point()`).
```



For c19ProSo01, participants from Saudi Arabia responded 1, which means that most participants from Saudi Arabia “Somewhat agree” to the statement that “I am willing to help others who suffer from coronavirus.”

For c19ProSo02, participants from Saudi Arabia responded 1, which means that most participants from Saudi Arabia “Somewhat agree” to the statement that “I am willing to make donations to help others that suffer from coronavirus.”

For c19ProSo03, participants from Saudi Arabia responded 1 which means that most participants from Saudi Arabia “Somewhat agree” to the statement that “I am willing to protect vulnerable groups from coronavirus even at my own expense.”

For c19ProSo04, participants from Saudi Arabia responded 1, which means that most participants from Saudi Arabia “Somewhat agree” to the statement that “I am willing to make personal sacrifices to prevent the spread of coronavirus.”

#### **Best Predictors for Saudi Arabia:**

- The best predictor for c19ProSo01 is c19ProSo02 with a p-value of  $4.560679e-18$  which is less than 0.05.
- The best predictor for c19ProSo02 is c19ProSo03 with a p-value of  $5.998479e-22$  which is less than 0.05.
- The best predictor for c19ProSo03 is c19ProSo04 with a p-value of  $1.997604e-28$  which is less than 0.05.
- The best predictor for c19ProSo04 is c19ProSo03 with a p-value of  $1.997604e-28$  which is less than 0.05.

## Appendix

### Summary of cvbase

summary(cvbase)

```
## isoFriends_inPerson isoOthPpl_inPerson isoFriends_online isoOthPpl_online
## Min. :0.000 Min. :0.000 Min. :0.000 Min. :0.00
## 1st Qu.:0.000 1st Qu.:0.000 1st Qu.:2.000 1st Qu.:0.00
## Median :1.000 Median :1.000 Median :5.000 Median :2.00
## Mean :2.074 Mean :1.952 Mean :4.414 Mean :2.87
## 3rd Qu.:4.000 3rd Qu.:3.000 3rd Qu.:7.000 3rd Qu.:5.00
## Max. :7.000 Max. :7.000 Max. :7.000 Max. :7.00
## NA's :331 NA's :516 NA's :949 NA's :1162
## lone01 lone02 lone03 happy
## Min. :1.000 Min. :1.000 Min. :1.000 Min. : 1.000
## 1st Qu.:1.000 1st Qu.:2.000 1st Qu.:1.000 1st Qu.: 5.000
## Median :2.000 Median :3.000 Median :2.000 Median : 7.000
## Mean :2.422 Mean :2.667 Mean :2.084 Mean : 6.337
## 3rd Qu.:3.000 3rd Qu.:4.000 3rd Qu.:3.000 3rd Qu.: 8.000
## Max. :5.000 Max. :5.000 Max. :5.000 Max. :10.000
## NA's :86 NA's :127 NA's :140 NA's :514
## lifeSat MLQ bor01 bor02
## Min. :1.000 Min. : -3.0000 Min. : -3.0000 Min. : -3.00000
## 1st Qu.:3.000 1st Qu.: 0.0000 1st Qu.: -1.0000 1st Qu.: -2.00000
## Median :4.000 Median : 1.0000 Median : 0.0000 Median : 0.00000
## Mean :4.139 Mean : 0.8472 Mean : 0.3251 Mean : 0.03983
## 3rd Qu.:5.000 3rd Qu.: 2.0000 3rd Qu.: 2.0000 3rd Qu.: 2.00000
## Max. :6.000 Max. : 3.0000 Max. : 3.0000 Max. : 3.00000
## NA's :111 NA's :119 NA's :163 NA's :176
## bor03 consp01 consp02 consp03
## Min. : -3.0000 Min. : 0.000 Min. : 0.000 Min. : 0.000
## 1st Qu.: -1.0000 1st Qu.: 5.000 1st Qu.: 5.000 1st Qu.: 4.000
## Median : 0.0000 Median : 7.000 Median : 8.000 Median : 5.000
## Mean : 0.3145 Mean : 6.839 Mean : 7.163 Mean : 5.591
## 3rd Qu.: 2.0000 3rd Qu.: 9.000 3rd Qu.: 9.000 3rd Qu.: 8.000
## Max. : 3.0000 Max. :10.000 Max. :10.000 Max. :10.000
## NA's :177 NA's :1510 NA's :1535 NA's :1555
## rankOrdLife_1 rankOrdLife_2 rankOrdLife_3 rankOrdLife_4
## Length:40000 Length:40000 Length:40000 Length:40000
## Class :character Class :character Class :character Class :character
## Mode :character Mode :character Mode :character Mode :character
##
##
##
## rankOrdLife_5 rankOrdLife_6 c19perBeh01 c19perBeh02
## Length:40000 Length:40000 Min. : -3.000 Min. : -3.00
## Class :character Class :character 1st Qu.: 2.000 1st Qu.: 2.00
```



```
## Mode :character Mode :character Median : 3.000 Median : 3.00
## Mean : 2.315 Mean : 2.43
## 3rd Qu.: 3.000 3rd Qu.: 3.00
## Max. : 3.000 Max. : 3.00
## NA's :127 NA's :135
## c19perBeh03 c19RCA01 c19RCA02 c19RCA03
## Min. :-3.00 Min. :-3.000 Min. :-3.000 Min. :-3.000
## 1st Qu.: 1.00 1st Qu.: 0.000 1st Qu.: 2.000 1st Qu.: 0.000
## Median : 2.00 Median : 2.000 Median : 3.000 Median : 2.000
## Mean : 1.84 Mean : 1.268 Mean : 2.053 Mean : 1.161
## 3rd Qu.: 3.00 3rd Qu.: 3.000 3rd Qu.: 3.000 3rd Qu.: 3.000
## Max. : 3.00 Max. : 3.000 Max. : 3.000 Max. : 3.000
## NA's :134 NA's :134 NA's :142 NA's :149
## gender age edu coded_country
## Min. :1.000 Min. :1.000 Min. :1.000 Length:40000
## 1st Qu.:1.000 1st Qu.:2.000 1st Qu.:4.000 Class :character
## Median :1.000 Median :3.000 Median :5.000 Mode :character
## Mean :1.389 Mean :2.892 Mean :4.404
## 3rd Qu.:2.000 3rd Qu.:4.000 3rd Qu.:5.000
## Max. :3.000 Max. :8.000 Max. :7.000
## NA's :221 NA's :240 NA's :272
## c19ProSo01 c19ProSo02 c19ProSo03 c19ProSo04
## Min. :-3.0000 Min. :-3.0000 Min. :-3.0000 Min. :-3.000
## 1st Qu.: 0.0000 1st Qu.: 0.0000 1st Qu.: 0.0000 1st Qu.: 0.000
## Median : 1.0000 Median : 1.0000 Median : 1.0000 Median : 2.000
## Mean : 0.9685 Mean : 0.6727 Mean : 0.5469 Mean : 1.282
## 3rd Qu.: 2.0000 3rd Qu.: 2.0000 3rd Qu.: 2.0000 3rd Qu.: 2.000
## Max. : 3.0000 Max. : 3.0000 Max. : 3.0000 Max. : 3.000
## NA's :129 NA's :134 NA's :149 NA's :150
## employment_status corona_close
## Length:40000 Length:40000
## Class :character Class :character
## Mode :character Mode :character
##
##
##
##
```

```
str(cvbase)
```

```
## 'data.frame': 40000 obs. of 38 variables:
## $ isoFriends_inPerson: int 2 3 4 2 4 7 2 7 3 1 ...
## $ isoOthPpl_inPerson : int 0 0 3 0 2 4 3 7 3 0 ...
## $ isoFriends_online : int 7 0 5 4 3 4 5 7 4 7 ...
## $ isoOthPpl_online : int 7 0 0 4 6 0 0 7 0 3 ...
## $ lone01 : int 3 2 1 3 1 2 2 1 3 2 ...
## $ lone02 : int 2 2 1 4 1 4 4 1 3 1 ...
## $ lone03 : int 2 2 1 4 1 3 1 1 2 1 ...
## $ happy : int 1 6 10 7 8 2 7 7 6 8 ...
## $ lifeSat : int 1 4 6 4 6 2 5 4 4 4 ...
## $ MLQ : int 0 2 3 0 3 -2 1 1 -1 -1 ...
## $ bor01 : int 0 2 -3 0 -2 -1 3 2 0 1 ...
## $ bor02 : int -1 1 -3 1 -2 -1 1 2 1 0 ...
## $ bor03 : int -1 -1 3 1 3 -1 2 -1 1 -1 ...
## $ consp01 : int 10 5 8 7 NA 2 3 NA 10 4 ...
## $ consp02 : int 10 10 8 7 NA 2 3 NA 10 6 ...
## $ consp03 : int 0 5 8 7 NA 7 1 NA 9 5 ...
```

```
## $ rankOrdLife_1 : chr "D" "C" "B" "A" ...
## $ rankOrdLife_2 : chr "E" "D" "F" "C" ...
## $ rankOrdLife_3 : chr "C" "E" "C" "D" ...
## $ rankOrdLife_4 : chr "A" "B" "D" "E" ...
## $ rankOrdLife_5 : chr "B" "A" "A" "B" ...
## $ rankOrdLife_6 : chr "F" "F" "E" "F" ...
## $ c19perBeh01 : int 3 2 2 0 3 2 3 3 2 2 ...
## $ c19perBeh02 : int -2 2 3 0 3 3 3 3 2 3 ...
## $ c19perBeh03 : int -2 1 3 1 3 3 3 2 2 3 ...
## $ c19RCA01 : int -3 -2 -3 0 3 1 3 2 -3 1 ...
## $ c19RCA02 : int -1 2 -1 1 3 1 3 3 -2 3 ...
## $ c19RCA03 : int -3 2 -2 0 3 -1 1 2 1 3 ...
## $ gender : int 2 1 2 2 2 1 1 2 2 1 ...
## $ age : int 3 1 2 3 2 2 2 3 2 1 ...
## $ edu : int 3 4 4 5 6 7 3 5 4 4 ...
## $ coded_country : chr "Greece" "Egypt" "Romania" "Italy" ...
## $ c19ProSo01 : int 2 1 3 0 3 2 0 0 1 -2 ...
## $ c19ProSo02 : int 0 1 0 0 3 -2 1 2 1 1 ...
## $ c19ProSo03 : int 2 1 0 -1 3 2 0 2 0 1 ...
## $ c19ProSo04 : int -2 1 3 0 2 3 3 1 2 3 ...
## $ employment_status : chr "4" "9" "9" "3" ...
## $ corona_close : chr "6" "6" "6" "6" ...
```

### Head of the external csv file

*# Take a Look of the dataset*

```
head(corona)
```

```
##          Country Year OVERALL.SCORE
## 1    Afghanistan 2021          28.8
## 2      Albania 2021          45.0
## 3      Algeria 2021          26.2
## 4      Andorra 2021          34.7
## 5      Angola 2021          29.1
## 6 Antigua & Barbuda 2021          30.0
## X1..PREVENTION.OF.THE.EMERGENCE.OR.RELEASE.OF.PATHOGENS
## 1                                     12.0
## 2                                     42.0
## 3                                     15.3
## 4                                     27.1
## 5                                     14.7
## 6                                     16.7
## X1.1..Antimicrobial.resistance..AMR.
## 1                                     16.7
## 2                                     33.3
## 3                                     33.3
## 4                                     0.0
## 5                                     33.3
## 6                                     50.0
## X1.1.1..AMR.surveillance..detection.and.reporting
## 1                                     33.3
## 2                                     16.7
## 3                                     16.7
## 4                                     0.0
## 5                                     16.7
## 6                                     0.0
## X1.1.1a..National.plan.for.AMR.priority.pathogens
## 1                                     50
```



##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X1.1.1b..Capacity.of.national.lab.lab.system.to.test.for.AMR.priority.pathogens	
##	1	50
##	2	50
##	3	50
##	4	0
##	5	50
##	6	0
##	X1.1.1c..National.environmental.surveillance.for.AMR.residues.organisms	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X1.1.2..Antimicrobial.control	
##	1	0
##	2	50
##	3	50
##	4	0
##	5	50
##	6	100
##	X1.1.2a..National.law.s..requiring.prescription.for.antibiotic.use..humans.	
##	1	0
##	2	100
##	3	100
##	4	0
##	5	100
##	6	100
##	X1.1.2b..National.law.s..requiring.prescription.for.antibiotic.use..animals.	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	100
##	X1.2..Zoonotic.disease	
##	1	5.5
##	2	24.6
##	3	8.4
##	4	42.4
##	5	5.1
##	6	0.0
##	X1.2.1..National.planning.for.zoonotic.diseases.pathogens	
##	1	25
##	2	50
##	3	0
##	4	50
##	5	25
##	6	0
##	X1.2.1a..Laws.plans.on.zoonotic.disease	
##	1	100
##	2	100

##	3	0	
##	4	100	
##	5	100	
##	6	0	
##	X1.2.1b..Laws.plans.on.zoonotic.disease.spillover.from.animals.to.humans		
##	1	0	
##	2	0	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##	X1.2.1c..Laws.plans.for.surveillanc...control.of.multiple.zoonotic.pathogens		
##	1	0	
##	2	100	
##	3	0	
##	4	100	
##	5	0	
##	6	0	
##	X1.2.1d..Cross.ministerial.department.agency.unit.for.zoonotic.disease		
##	1	0	
##	2	0	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##	X1.2.2..Surveillance.systems.for.zoonotic.diseases.pathogens		
##	1	0.0	
##	2	66.7	
##	3	33.3	
##	4	33.3	
##	5	0.0	
##	6	0.0	
##	X1.2.2a..Surveillance.reporting.mechanism.for.zoonotic.disease.for.livestock.owners		
##	1	0	
##	2	100	
##	3	100	
##	4	0	
##	5	0	
##	6	0	
##	X1.2.2b..Laws.regulations.on.data.confidentiality.to.protect.livestock.owners		
##	1	0	
##	2	100	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##	X1.2.2c..Wildlife.zoonotic.disease.surveillanc		
##	1	0	
##	2	0	
##	3	0	
##	4	100	
##	5	0	
##	6	0	
##	X1.2.3..International.reporting.of.animal.disease.outbreaks		
##	1	0	
##	2	0	
##	3	0	

##	4		100
##	5		0
##	6		0
##		X1.2.3a..Annual.reporting.to.OIE.on.zoonotic.disease.incidence	
##	1		0
##	2		0
##	3		0
##	4		100
##	5		0
##	6		0
##		X1.2.4..Animal.health.workforce	
##	1		2.7
##	2		6.3
##	3		8.5
##	4		28.5
##	5		0.7
##	6		0.0
##		X1.2.4a..Number.of.veterinarians.per.100.000.people	
##	1		1.3
##	2		12.6
##	3		17.0
##	4		25.1
##	5		0.4
##	6		0.0
##		X1.2.4b..Number.of.veterinary.para.professionals.per.100.000.people	
##	1		4.1
##	2		0.0
##	3		0.0
##	4		32.0
##	5		1.0
##	6		0.0
##		X1.2.5..Private.sector.and.zoonotic.disease	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.2.5a..Inclusion.of.private.sector.in.national.plan.law.on.zoonotic.disease	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.3..Biosecurity X1.3.1..Whole.of.government.biosecurity.systems	
##	1		0
##	2		44
##	3		0
##	4		20
##	5		0
##	6		0
##		X1.3.1a..Updated.national.records.of.especially.dangerous.pathogen.toxin.inventories	
##	1		0
##	2		100
##	3		0
##	4		0

##	5		0
##	6		0
##		X1.3.1b..Biosecurity.laws.on.facility.security.for.especially.dangerous.pathogens	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.3.1c..Agency.for.enforcement.of.biosecurity.laws.regulations	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.3.1d..Consolidation.of.especially.dangerous.pathogens.into.minimum...of.facilities	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.3.1e..Capacity.to.conduct.tests.for.anthrax.Ebola.without.culturing.live.pathogens	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.3.2..Biosecurity.training.and.practices	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.3.2a..Biosecurity.training.using.a.standardised..required.approach	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.3.3..Personnel.vetting..regulating.access.to.sensitive.locations	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.3.3a..Personnel.checks.for.permission.to.access.to.especially.dangerous.pathogens	
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0

##	6		0
##		X1.3.4..Transportation.security	
##	1	0	
##	2	100	
##	3	0	
##	4	100	
##	5	0	
##	6	0	
##		X1.3.4a..National.transport.regulations.for.Category.A.and.B.infectious.substances	
##	1		0
##	2		100
##	3		0
##	4		100
##	5		0
##	6		0
##		X1.3.5..Cross.border.transfer.and.end.user.screening	
##	1	0	
##	2	100	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##		X1.3.5a..Laws.regulations.on.cross.border.transfer.and.end.user.screening	
##	1		0
##	2		100
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.4..Biosafety X1.4.1..Whole.of.government.biosafety.systems	
##	1	0	0
##	2	50	100
##	3	0	0
##	4	0	0
##	5	0	0
##	6	0	0
##		X1.4.1a..Biosafety.laws.regulations	
##	1	0	
##	2	100	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##		X1.4.1b..Agency.for.enforcement.of.biosafety.laws.regulations	
##	1		0
##	2		100
##	3		0
##	4		0
##	5		0
##	6		0
##		X1.4.2..Biosafety.training.and.practices	
##	1	0	
##	2	0	
##	3	0	
##	4	0	
##	5	0	
##	6	0	

```

## X1.4.2a..Biosafety.training.using.a.standardised..required.approach
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X1.5..Dual.use.research.and.culture.of.responsible.science
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X1.5.1..Oversight.of.dual.use.research
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X1.5.1a..Evidence.of.national.assessment.of.dual.use.research
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X1.5.1b..National.law.regulation.on.oversight.of.dual.use.research
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X1.5.1c..Existence.of.agency.responsible.for.oversight.of.dual.use.research
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X1.5.2..Screening.requirements.for.providers.of.genetic.material
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X1.5.2a..Requirement.to.screen.synthesised.DNA.against.list.prior.to.sale
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X1.6..Immunization X1.6.1..Vaccination.rates

```



##	1	50	50
##	2	100	100
##	3	50	50
##	4	100	100
##	5	50	50
##	6	50	50
##	X1.6.1a..Immunization.rate.for.humans..measles.MCV2.		
##	1		0
##	2		100
##	3		0
##	4		100
##	5		0
##	6		100
##	X1.6.1b..Availability.of.vaccination.figures.for.livestock..FMD..through.OIE.database		
##	1		100
##	2		100
##	3		100
##	4		100
##	5		100
##	6		0
##	X2..EARLY.DETECTION...REPORTING.FOR.EPIDEMICS.OF.POTENTIAL.INT.L.CONCERN		
##	1		20.6
##	2		40.0
##	3		12.6
##	4		2.2
##	5		13.3
##	6		5.8
##	X2.1..Laboratory.systems.strength.and.quality		
##	1	12.5	
##	2	50.0	
##	3	25.0	
##	4	0.0	
##	5	25.0	
##	6	0.0	
##	X2.1.1..Lab.capacity.for.detecting.priority.diseases		
##	1	25	
##	2	50	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##	X2.1.1a..Capacity.of.national.lab.system.to.conduct.5.or.more.WHO.core.tests		
##	1		50
##	2		100
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.1.1b..Plan.to.conduct.testing.during.a.public.health.emergency		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.1.2..Laboratory.quality.systems		
##	1		0

##	2	50	
##	3	50	
##	4	0	
##	5	50	
##	6	0	
##	X2.1.2a..Existence.of.an.accredited.national.lab.serving.as.a.reference.facility		
##	1		0
##	2		0
##	3		100
##	4		0
##	5		0
##	6		0
##	X2.1.2b..External.quality.assurance.of.a.national.lab.serving.as.a.reference.facility		
##	1		0
##	2		100
##	3		0
##	4		0
##	5		100
##	6		0
##	X2.2..Laboratory.supply.chains X2.2.1..Specimen.referral.and.transport.system		
##	1	0	0
##	2	0	0
##	3	0	0
##	4	0	0
##	5	0	0
##	6	0	0
##	X2.2.1a..Is.there.a.nationwide.specimen.transport.system.		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.2.2..Laboratory.cooperation.and.coordination		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.2.2a..Plan.to.rapidly.authorize.license.laboratories.to.scale.up.testing.during.an.outbreak		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.3..Real.time.surveillance.and.reporting		
##	1	37.5	
##	2	12.5	

##	3	37.5	
##	4	0.0	
##	5	0.0	
##	6	0.0	
##	X2.3.1..Indicator.and.event.based.surveillance.and.reporting.systems		
##	1	25	
##	2	25	
##	3	75	
##	4	0	
##	5	0	
##	6	0	
##	X2.3.1a..Evidence.of.ongoing.event.based.surveillance.and.analysis		
##	1	50	
##	2	50	
##	3	50	
##	4	0	
##	5	0	
##	6	0	
##	X2.3.1b..Evidence.of.reporting.a.potential.PHEIC.to.the.WHO..last.2.years.		
##	1	0	
##	2	0	
##	3	100	
##	4	0	
##	5	0	
##	6	0	
##	X2.3.2..Interoperable..interconnected..electronic.real.time.reporting.systems		
##	1	50	
##	2	0	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##	X2.3.2a..Electronic.national.and.sub.national.reporting.surveillance.system		
##	1	100	
##	2	0	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##	X2.3.2b..Collection.of.ongoing.real.time.lab.data.by.electronic.surveillance.system		
##	1	0	
##	2	0	
##	3	0	
##	4	0	
##	5	0	
##	6	0	
##	X2.4..Surveillance.data.accessibility.and.transparency		
##	1	23.3	
##	2	40.0	
##	3	13.3	
##	4	13.3	
##	5	30.0	
##	6	10.0	
##	X2.4.1..Coverage.and.use.of.electronic.health.records		
##	1	16.7	
##	2	0.0	
##	3	16.7	

##	4		16.7
##	5		0.0
##	6		0.0
##	X2.4.1a..Common.usage.of.electronic.health.records		
##	1		50
##	2		0
##	3		50
##	4		50
##	5		0
##	6		0
##	X2.4.1b..Public.health.system.access.to.individual.electronic.health.records		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.4.1c..Existence.of.data.standards.for.health.record.data.comparability		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.4.2..Data.integration.between.human..animal.and.environmental.health.sectors		
##	1		100
##	2		100
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.4.2a..Data.sharing.mechanisms X2.4.3..Transparency.of.surveillance.data		
##	1	100	0
##	2	100	0
##	3	0	0
##	4	0	0
##	5	0	0
##	6	0	0
##	X2.4.3a..Availability.of.de.identified.health.surveillance.data.on.disease.outbreaks		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.4.4..Ethical.considerations.during.surveillance		
##	1		0
##	2		100
##	3		50
##	4		50
##	5		50
##	6		50
##	X2.4.4a..Confidentiality.legislation.regulations.for.identifiable.health.information		
##	1		0
##	2		100
##	3		100
##	4		100

##	5		100
##	6		100
##	X2.4.4b..Inclusion.of.cyber.protections.in.health.data.confidentiality.law.regulation		
##	1		0
##	2		100
##	3		0
##	4		0
##	5		0
##	6		0
##	X2.4.5..International.data.sharing		
##	1	0	
##	2	0	
##	3	0	
##	4	0	
##	5	100	
##	6	0	
##	X2.4.5a..Cooperative.commitments.or.agreements.within.regions		
##	1	0	
##	2	0	
##	3	0	
##	4	0	
##	5	100	
##	6	0	
##	X2.5..Case.based.investigation X2.5.1..Case.investigation.and.contact.tracing		
##	1	0.0	0
##	2	37.5	25
##	3	0.0	0
##	4	0.0	0
##	5	0.0	0
##	6	0.0	0
##	X2.5.1a..National.support.to.conduct.contact.tracing.in.the.event.of.a.public.health.emergency		
##	1		
	0		
##	2		
	50		
##	3		
	0		
##	4		
	0		
##	5		
	0		
##	6		
	0		
##	X2.5.1b..Provision.of.wraparound.services.to.enable.self.isolation.quarantine.as.recommended		
##	1		
	0		
##	2		
	0		
##	3		
	0		
##	4		
	0		
##	5		
	0		
##	6		

```

0
## X2.5.2..Point.of.entry.management
## 1 0
## 2 50
## 3 0
## 4 0
## 5 0
## 6 0
## X2.5.2a..Strategy.for.tracing.and.quarantining.international.travelers
## 1 0
## 2 50
## 3 0
## 4 0
## 5 0
## 6 0
## X2.6..Epidemiology.workforce
## 1 50
## 2 100
## 3 0
## 4 0
## 5 25
## 6 25
## X2.6.1..Existence.of.applied.epidemiology.training.program.such.FETP.and.FETPV
## 1 100
## 2 100
## 3 0
## 4 0
## 5 50
## 6 50
## X2.6.1a..Access.to.field.epidemiology.training.program.in.country.and.or.abroad
## 1 100
## 2 100
## 3 0
## 4 0
## 5 100
## 6 100
## X2.6.1b..Existence.of.field.epidemiology.training.for.animal.health.professionals
## 1 100
## 2 100
## 3 0
## 4 0
## 5 0
## 6 0
## X2.6.2..Epidemiology.workforce.capacity
## 1 0
## 2 100
## 3 0
## 4 0
## 5 0
## 6 0
## X2.6.2a..Evidence.of.at.least.1.trained.field.epidemiologist.per.200.000.people
## 1 0
## 2 100
## 3 0
## 4 0
## 5 0
## 6 0

```



##	X3..RAPID.RESPONSE.TO.AND.MITIGATION.OF.THE.SPREAD.OF.AN.EPIDEMIC	
##	1	24.5
##	2	38.1
##	3	25.6
##	4	39.5
##	5	31.6
##	6	32.1
##	X3.1..Emergency.preparedness.and.response.planning	
##	1	20.8
##	2	16.7
##	3	16.7
##	4	20.8
##	5	16.7
##	6	16.7
##	X3.1.1..National.public.health.emergency.preparedness.and.response.plan	
##	1	12.5
##	2	0.0
##	3	0.0
##	4	12.5
##	5	0.0
##	6	0.0
##	X3.1.1a..National.emergency.response.plan.for.diseases.with.pandemic.potential	
##	1	50
##	2	0
##	3	0
##	4	50
##	5	0
##	6	0
##	X3.1.1b..National.public.health.emergency.response.plan.updated.in.past.3.years	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X3.1.1c..Vulnerable.populations.in.national.public.health.emergency.response.plan	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X3.1.1d..Existence.of.public.pandemic.influenza.preparedness.plan.updated.since.2009	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X3.1.2..Private.sector.involvement.in.response.planning	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X3.1.2a..Mechanism.to.engage.private.sector.in.outbreak.preparedness.response	

##	1			0
##	2			0
##	3			0
##	4			0
##	5			0
##	6			0
##	X3.1.3..Non.pharmaceutical.interventions.planning			
##	1		50	
##	2		50	
##	3		50	
##	4		50	
##	5		50	
##	6		50	
##	X3.1.3a..Policy.plan.guidelines.in.place.to.implement.non.pharmaceutical.interventions			
##	..NPIs.			
##	1			
	50			
##	2			
	50			
##	3			
	50			
##	4			
	50			
##	5			
	50			
##	6			
	50			
##	X3.2..Exercising.response.plans X3.2.1..Activating.response.plans			
##	1	25		50
##	2	25		50
##	3	25		50
##	4	25		50
##	5	25		50
##	6	0		0
##	X3.2.1a..Completion.of.biological.focused.IHR.exercise.with.the.WHO.in.past.year			
##	1			100
##	2			100
##	3			100
##	4			100
##	5			100
##	6			0
##	X3.2.1b..Evidence.of.bio.focused.exercise.to.identify.gaps.best.practices			
##	1			0
##	2			0
##	3			0
##	4			0
##	5			0
##	6			0
##	X3.2.2..Private.sector.engagement.in.exercises			
##	1		0	
##	2		0	
##	3		0	
##	4		0	
##	5		0	
##	6		0	
##	X3.2.2a..Evidence.of.national.level.biological.threat.focused.exercise.that.includes.p			
##	rivate.sector			

```

## 1
0
## 2
0
## 3
0
## 4
0
## 5
0
## 6
0
## X3.3..Emergency.response.operation X3.3.1..Emergency.response.operation
## 1 33.3 33.3
## 2 33.3 33.3
## 3 33.3 33.3
## 4 0.0 0.0
## 5 0.0 0.0
## 6 33.3 33.3
## X3.3.1a..Existence.of.Emergency.Operations.Center..EOC.
## 1 100
## 2 100
## 3 100
## 4 0
## 5 0
## 6 100
## X3.3.1b..Requirement.for.EOC.to.conduct.evidence.EOC.conducts.at.least.annual.drills
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X3.3.1c..EOC.activation.within.120.minutes.of.identification.of.emergency.scenario
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X3.4..Linking.public.health.and.security.authorities
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X3.4.1..Public.health.and.security.authorities.linked.for.a.biological.event
## 1 0
## 2 0
## 3 0
## 4 0
## 5 0
## 6 0
## X3.4.1a..Joint.exercise.procedures.for.potential.deliberate.biological.events
## 1 0
## 2 0

```

##	3		0
##	4		0
##	5		0
##	6		0
##	X3.5..Risk.communication X3.5.1..Risk.communication.planning		
##	1	25.0	0
##	2	100.0	100
##	3	37.5	0
##	4	50.0	0
##	5	87.5	100
##	6	37.5	0
##	X3.5.1a..Risk.communication.plan.for.specific.use.during.a.public.health.emergency		
##	1		0
##	2		100
##	3		0
##	4		0
##	5		100
##	6		0
##	X3.5.1b..Inclusion.of.different.population...sector.needs.in.risk.communication.plan		
##	1		0
##	2		100
##	3		0
##	4		0
##	5		100
##	6		0
##	X3.5.1c..Designation.of.a.specific.government.spokesperson.during.a.public.health.emer gency		
##	1		0
##	2		100
##	3		0
##	4		0
##	5		100
##	6		0
##	X3.5.2..Public.health.systems.communication		
##	1	50	
##	2	100	
##	3	75	
##	4	100	
##	5	75	
##	6	75	
##	X3.5.2a..Government.use.of.media.platforms.to.share.info.on.public.health.emergencies		
##	1		0
##	2		100
##	3		50
##	4		100
##	5		50
##	6		50
##	X3.5.2b..Evidence.that.senior.leaders.have.shared.mis.disinformation.on.infectious.dis eases		
##	1		100

```

## 2
100
## 3
100
## 4
100
## 5
100
## 6
100
## X3.6..Access.to.communications.infrastructure X3.6.1..Internet.users
## 1 17.4 13.5
## 2 66.4 69.8
## 3 66.4 59.8
## 4 81.0 91.9
## 5 41.7 14.3
## 6 87.1 76.2
## X3.6.1a..Percentage.of.households.with.Internet X3.6.2..Mobile.subscribers
## 1 13.5 23.9
## 2 69.8 41.1
## 3 59.8 50.9
## 4 91.9 53.3
## 5 14.3 17.0
## 6 76.2 95.8
## X3.6.2a..Mobile.cellular.telephone.subscriptions.per.100.inhabitants
## 1 23.9
## 2 41.1
## 3 50.9
## 4 53.3
## 5 17.0
## 6 95.8
## X3.6.3..Female.access.to.a.mobile.phone
## 1 11.4
## 2 79.5
## 3 84.1
## 4 95.5
## 5 77.3
## 6 93.2
## X3.6.3a..Gender.gap.in.access.to.a.mobile.phone..percentage.points.
## 1 11.4
## 2 79.5
## 3 84.1
## 4 95.5
## 5 77.3
## 6 93.2
## X3.6.4..Female.access.to.the.Internet
## 1 20.8
## 2 75.0
## 3 70.8
## 4 83.3
## 5 58.3
## 6 83.3
## X3.6.4a..Gender.gap.in.access.to.the.Internet..percentage.points.
## 1 20.8
## 2 75.0
## 3 70.8
## 4 83.3

```

## 5		58.3
## 6		83.3
##	X3.7..Trade.and.travel.restrictions	X3.7.1..Trade.restrictions
## 1	50	100
## 2	25	50
## 3	0	0
## 4	100	100
## 5	50	100
## 6	50	100
##	X3.7.1a..Restrictions.on.export.import.of.medical.goods.due.to.an.infectious.disease.o	utbreak
## 1		
100		
## 2		
0		
## 3		
0		
## 4		
100		
## 5		
100		
## 6		
100		
##	X3.7.1b..Restrictions.on.movement.and.or.exports.imports.due.to.disease.outbreak	
## 1		100
## 2		100
## 3		0
## 4		100
## 5		100
## 6		100
##	X3.7.2..Travel.restrictions	
## 1	0	
## 2	0	
## 3	0	
## 4	100	
## 5	0	
## 6	0	
##	X3.7.2a..Evidence.of.travel.ban.due.to.an.infectious.disease.outbreak	
## 1		0
## 2		0
## 3		0
## 4		100
## 5		0
## 6		0
##	X4..SUFFICIENT...ROBUST.HEALTH.SECTOR.TO.TREAT.THE.SICK...PROTECT.HEALTH.WORKERS	
## 1		23.0
## 2		47.4
## 3		15.0
## 4		15.4
## 5		23.1
## 6		16.7
##	X4.1..Health.capacity.in.clinics..hospitals.and.community.care.centers	
## 1		51.0
## 2		42.1
## 3		23.5
## 4		29.4
## 5		18.2



## 6		46.2
##	X4.1.1..Available.human.resources.for.the.broader.healthcare.system	
## 1		34.5
## 2		10.7
## 3		9.2
## 4		19.7
## 5		1.4
## 6		19.0
##	X4.1.1a..Doctors.per.100.000.people	
## 1		3.1
## 2		14.3
## 3		20.3
## 4		39.5
## 5		2.4
## 6		35.0
##	X4.1.1b..Nurses.and.midwives.per.100.000.people	
## 1		0.5
## 2		17.8
## 3		7.3
## 4		19.6
## 5		1.7
## 6		22.1
##	X4.1.1c..Updated.health.workforce.strategy.to.address.human.resource.shortfalls	
## 1		100
## 2		0
## 3		0
## 4		0
## 5		0
## 6		0
##	X4.1.2..Facilities.capacity X4.1.2a..Hospital.beds.per.100.000.people	
## 1	67.4	2.1
## 2	73.5	20.4
## 3	37.7	13.1
## 4	39.2	17.5
## 5	35.0	5.1
## 6	73.5	20.4
##	X4.1.2b..In.country.capacity.to.isolate.patients.with.highly.communicable.diseases	
## 1		100
## 2		100
## 3		0
## 4		100
## 5		0
## 6		100
##	X4.1.2c..Demonstrated.capacity...evidence.of.plan.to.expand.isolation.capacity	
## 1		100
## 2		100
## 3		100
## 4		0
## 5		100
## 6		100
##	X4.2..Supply.chain.for.health.system.and.healthcare.workers	
## 1		0.0
## 2		77.8
## 3		0.0
## 4		33.3
## 5		61.1
## 6		16.7

```

## X4.2.1..Routine.health.care.and.laboratory.system.supply
## 1 0
## 2 100
## 3 0
## 4 100
## 5 100
## 6 50
## X4.2.1a..National.procurement.protocol.for.the.acquisition.of.routine.laboratory.medic
al.supplies
## 1
0
## 2
100
## 3
0
## 4
100
## 5
100
## 6
50
## X4.2.2..Stockpiling.for.emergencies
## 1 0.0
## 2 33.3
## 3 0.0
## 4 0.0
## 5 33.3
## 6 0.0
## X4.2.2a..Stockpile.of.medical.supplies.for.national.use.during.a.public.health.emergen
cy
## 1
0
## 2 1
00
## 3
0
## 4
0
## 5 1
00
## 6
0
## X4.2.2b..Stockpile.of.laboratory.supplies.for.national.use.during.a.public.health.emer
gency
## 1
0
## 2
0
## 3
0
## 4
0
## 5
0
## 6
0
## X4.2.2c..Annual.review.of.national.stockpile.to.ensure.sufficient.supply

```

## 1		0
## 2		0
## 3		0
## 4		0
## 5		0
## 6		0
## X4.2.3..Manufacturing.and.procurement.for.emergencies		
## 1	0	
## 2	100	
## 3	0	
## 4	0	
## 5	50	
## 6	0	
## X4.2.3a..Plan.agreement.to.produce.procure.medical.supplies.during.a.public.health.eme rgency		
## 1		
0		
## 2		
100		
## 3		
0		
## 4		
0		
## 5		
100		
## 6		
0		
## X4.2.3b..Plan.agreement.to.produce.procure.lab.supplies.during.a.public.health.emergen cy		
## 1		
0		
## 2		1
00		
## 3		
0		
## 4		
0		
## 5		
0		
## 6		
0		
## X4.3..Medical.countermeasures.and.personnel.deployment		
## 1	0	
## 2	50	
## 3	0	
## 4	0	
## 5	0	
## 6	0	
## X4.3.1..System.for.dispensing.MCMs.during.a.public.health.emergency		
## 1		0
## 2		0
## 3		0
## 4		0
## 5		0
## 6		0
## X4.3.1a..Plan.program.guidelines.for.dispensing.MCMs.during.a.public.health.emergency		
## 1		0

##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.3.2..System.for.receiving.foreign.health.personnel.during.a.public.health.emergency		
##	1		0
##	2		100
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.3.2a..Plan.to.receive.foreign.health.personnel.during.a.public.health.emergency		
##	1		0
##	2		100
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.4..Healthcare.access X4.4.1..Access.to.healthcare		
##	1	60.0	79.9
##	2	61.8	85.4
##	3	56.4	69.3
##	4	45.2	35.7
##	5	57.4	72.2
##	6	53.8	61.4
##	X4.4.1a..Constitutional.guarantee.of.citizens..right.to.medical.care		
##	1		100
##	2		75
##	3		25
##	4		0
##	5		75
##	6		0
##	X4.4.1b..Access.to.skilled.birth.attendants....of.population.		
##	1		45.4
##	2		99.8
##	3		96.2
##	4		99.0
##	5		44.4
##	6		100.0
##	X4.4.1c..Out.of.pocket.health.expenditures.per.capita..PPP..current.international...		
##	1		94.4
##	2		81.5
##	3		86.7
##	4		8.0
##	5		97.3
##	6		84.3
##	X4.4.2..Paid.medical.leave X4.4.2a..Guaranteed.paid.sick.leave		
##	1	100	100
##	2	100	100
##	3	100	100
##	4	100	100
##	5	100	100
##	6	100	100
##	X4.4.3..Healthcare.worker.access.to.healthcare		
##	1		0
##	2		0

##	3	0	
##	4	0	
##	5	0	
##	6	0	
##	X4.4.3a..Government.prioritisation.of.care.for.healthcare.workers.during.response		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.5..Communications.with.healthcare.workers.during.a.public.health.emergency		
##	1		0
##	2		50
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.5.1..Communication.with.healthcare.workers		
##	1		0
##	2		50
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.5.1a..Existence.of.system.for.communication.during.a.public.health.emergency		
##	1		0
##	2		100
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.5.1b..Inclusion.of.public.and.private.sector.in.healthcare.communication.system		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.6..Infection.control.practices		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.6.1..Healthcare.associated.infection..HCAI..monitoring		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.6.1a..Evidence.of.national.public.health.system.monitoring.and.tracking.of.HCAIs		
##	1		0
##	2		0
##	3		0

##	4		0
##	5		0
##	6		0
##	X4.7..Capacity.to.test.and.approve.new.medical.countermeasures		
##	1	50	
##	2	50	
##	3	25	
##	4	0	
##	5	25	
##	6	0	
##	X4.7.1..Regulatory.process.for.clinical.trials.of.unregistered.interventions		
##	1		50
##	2		50
##	3		50
##	4		0
##	5		0
##	6		0
##	X4.7.1a..Requirement.for.ethical.review.before.beginning.a.clinical.trial		
##	1		100
##	2		100
##	3		100
##	4		0
##	5		0
##	6		0
##	X4.7.1b..Expedited.approval.for.clinical.trials.of.unregistered.MCMs.during.epidemics		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X4.7.2..Regulatory.process.for.approving.medical.countermeasures		
##	1	50	
##	2	50	
##	3	0	
##	4	0	
##	5	50	
##	6	0	
##	X4.7.2a..Existence.of.agency.responsible.for.approving.new.human.MCMs		
##	1		100
##	2		100
##	3		0
##	4		0
##	5		100
##	6		0
##	X4.7.2b..Expedited.approval.for.human.MCMs.during.public.health.emergencies		
##	1		0
##	2		0
##	3		0
##	4		0
##	5		0
##	6		0
##	X5..COMMITMENTS.TO.IMPROVING.NATIONAL.CAPACITY..FINANCING.AND.ADHERENCE.TO.NORMS		
##	1		60.9
##	2		52.1
##	3		38.9
##	4		43.2



##	5		47.7
##	6		45.5
##	X5.1..IHR.reporting.compliance.and.disaster.risk.reduction		
##	1	50	
##	2	50	
##	3	50	
##	4	100	
##	5	50	
##	6	50	
##	X5.1.1..Official.IHR.reporting		
##	1	100	
##	2	100	
##	3	100	
##	4	100	
##	5	100	
##	6	0	
##	X5.1.1a..Submission.of.IHR.reports.to.the.WHO.in.past.year		
##	1	100	
##	2	100	
##	3	100	
##	4	100	
##	5	100	
##	6	0	
##	X5.1.2..Integration.of.health.into.disaster.risk.reduction		
##	1	0	
##	2	0	
##	3	0	
##	4	100	
##	5	0	
##	6	100	
##	X5.1.2a..Existence.of.specific.risk.reduction.strategies.for.epidemics.and.pandemics		
##	1	0	
##	2	0	
##	3	0	
##	4	100	
##	5	0	
##	6	100	
##	X5.2..Cross.border.agreements.on.public.health.and.animal.health.emergency.response		
##	1	50	
##	2	100	
##	3	50	
##	4	50	
##	5	50	
##	6	100	
##	X5.2.1..Cross.border.agreements		
##	1	50	
##	2	100	
##	3	50	
##	4	50	
##	5	50	
##	6	100	
##	X5.2.1a..Existence.of.public.health.emergency.agreements.with.regional.neighbors		
##	1	100	
##	2	100	
##	3	100	
##	4	100	
##	5	100	

##	6		100
##		X5.2.1b..Existence.of.animal.health.emergency.agreements.with.regional.neighbors	
##	1		0
##	2		100
##	3		0
##	4		0
##	5		0
##	6		100
##		X5.3..International.commitments	
##	1	90.6	
##	2	37.5	
##	3	50.0	
##	4	34.4	
##	5	28.1	
##	6	31.3	
##		X5.3.1..Participation.in.international.agreements	
##	1	81.3	
##	2	75.0	
##	3	100.0	
##	4	68.8	
##	5	56.3	
##	6	62.5	
##		X5.3.1a..Biological.and.Toxin.Weapons.Convention.status	
##	1	100	
##	2	100	
##	3	100	
##	4	100	
##	5	100	
##	6	100	
##		X5.3.1b..Submission.of.CBMs.to.the.Biological.and.Toxin.Weapons.Convention	
##	1		100
##	2		0
##	3		100
##	4		0
##	5		0
##	6		0
##		X5.3.1c..Submission.of.UNSCR.1540.reports	
##	1	100	
##	2	100	
##	3	100	
##	4	100	
##	5	100	
##	6	100	
##		X5.3.1d..Extent.of.UNSCR.1540.implementation.on.public.health.emergencies	
##	1		25
##	2		100
##	3		100
##	4		75
##	5		25
##	6		50
##		X5.3.2..Voluntary.memberships	
##	1	100	
##	2	0	
##	3	0	
##	4	0	
##	5	0	
##	6	0	

##	X5.3.2a..Membership.in.global.health.security.and.or.biological.weapons.agreements	
##	1	100
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X5.4..JEE.and.PVS	
##	1	75
##	2	25
##	3	0
##	4	0
##	5	25
##	6	0
##	X5.4.1..Completion.and.publication.of.a.JEE.assessment.and.gap.analysis	
##	1	100
##	2	50
##	3	0
##	4	0
##	5	50
##	6	0
##	X5.4.1a..Completion.and.publication.of.JEE..or.GHSA.pilot.external.assessment..report	
##	1	100
##	2	100
##	3	0
##	4	0
##	5	100
##	6	0
##	X5.4.1b..Completion.and.publication.of.a.NAPHS.or.GHSA.roadmap	
##	1	100
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X5.4.2..Completion.and.publication.of.a.PVS.assessment.and.gap.analysis	
##	1	50
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X5.4.2a..Completion.and.publication.of.PVS.report..past.five.years.	
##	1	100
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X5.4.2b..Completion.and.publication.of.PVS.gap.analysis..past.five.years.	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X5.5..Financing X5.5.1..National.financing.for.epidemic.preparedness	

## 1	33.3	0
## 2	33.3	0
## 3	16.7	0
## 4	8.3	0
## 5	66.7	100
## 6	25.0	0
## X5.5.1a..Evidence.of.allocated.national.funds.to.improve.capacity.to.address.epidemic.threats		
## 1		0
## 2		0
## 3		0
## 4		0
## 5		100
## 6		0
## X5.5.2..Financing.under.JEE.and.PVS.reports.and.gap.analysis		
## 1		0
## 2		0
## 3		0
## 4		0
## 5		0
## 6		0
## X5.5.2a..National.budget.to.address.gaps.identified.in.JEE..NAPHS.or.GHSA.roadmap		
## 1		0
## 2		0
## 3		0
## 4		0
## 5		0
## 6		0
## X5.5.2b..National.budget.to.address.gaps.identified.in.PVS.assessment.or.gap.analysis		
## 1		0
## 2		0
## 3		0
## 4		0
## 5		0
## 6		0
## X5.5.3..Financing.for.emergency.response		
## 1	100	
## 2	100	
## 3	0	
## 4	0	
## 5	100	
## 6	100	
## X5.5.3a..Emergency.public.financing.during.a.public.health.emergency		
## 1		100
## 2		100
## 3		0
## 4		0
## 5		100
## 6		100
## X5.5.4..Accountability.for.international.commitments.to.address.epidemic.threats		
## 1		33.3

##	2	33.3
##	3	66.7
##	4	33.3
##	5	66.7
##	6	0.0
##	X5.5.4a..Commitments.to.improve.domestic.or.foreign.capacity.for.epidemic.threats	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X5.5.4b..Investments.to.improve.domestic.or.foreign.capacity.for.epidemic.threats	
##	1	100
##	2	100
##	3	100
##	4	0
##	5	100
##	6	0
##	X5.5.4c..Evidence.that.the.country.has.fulfilled.its.full.WHO.contribution.within.the.past.two.years	
##	1	0
##	2	0
##	3	100
##	4	100
##	5	100
##	6	0
##	X5.6..Commitment.to.sharing.of.genetic...biological.data...specimens	
##	1	66.7
##	2	66.7
##	3	66.7
##	4	66.7
##	5	66.7
##	6	66.7
##	X5.6.1..Commitment.to.share.data.and.specimens.in.emergency.non.emergency.research	
##	1	66.7
##	2	66.7
##	3	66.7
##	4	66.7
##	5	66.7
##	6	66.7
##	X5.6.1a..Sharing.of.genetic.biological.data.and.materials.beyond.influenza	
##	1	0
##	2	0
##	3	0
##	4	0
##	5	0
##	6	0
##	X5.6.1b..Evidence.of.non.compliance.with.sample.sharing.element.of.PIP.framework	
##	1	100
##	2	100

##	3			100
##	4			100
##	5			100
##	6			100
##	X5.6.1c..Evidence.of.non.sharing.of.pandemic.pathogen.samples.during.an.outbreak			
##	1			100
##	2			100
##	3			100
##	4			100
##	5			100
##	6			100
##	X6..OVERALL.RISK.ENVIRONMENT.AND.COUNTRY.VULNERABILITY.TO.BIOLOGICAL.THREATS			
##	1			31.6
##	2			50.6
##	3			49.7
##	4			80.5
##	5			43.9
##	6			63.2
##	X6.1..Political.and.security.risk X6.1.1..Government.effectiveness			
##	1	5.0		9.9
##	2	62.5		37.3
##	3	41.6		15.9
##	4	92.2		95.3
##	5	60.2		21.7
##	6	83.5		51.1
##	X6.1.1a..Policy.formation X6.1.1b..Quality.of.bureaucracy			
##	1	50		0
##	2	50		25
##	3	25		0
##	4	100		100
##	5	50		25
##	6	50		50
##	X6.1.1c..Excessive.bureaucracy.red.tape X6.1.1d..Vested.interests.cronyism			
##	1	0		0
##	2	25		0
##	3	0		0
##	4	100		100
##	5	0		25
##	6	50		50
##	X6.1.1e..Corruption X6.1.1f..Accountability.of.public.officials			
##	1	19		0
##	2	36		50
##	3	36		25
##	4	67		100
##	5	27		0
##	6	58		50
##	X6.1.1g..Human.rights.risk X6.1.2..Orderly.transfers.of.power			
##	1	0		25
##	2	75		50
##	3	25		0
##	4	100		100
##	5	25		75
##	6	50		75
##	X6.1.2a..Orderly.transfers.of.power X6.1.3..Risk.of.social.unrest			
##	1	25		0
##	2	50		25
##	3	0		0

## 4	100	100
## 5	75	50
## 6	75	75
## X6.1.3a..Risk.of.social.unrest	X6.1.4..Illicit.activities.by.non.state.actors	
## 1	0	0.0
## 2	25	50.0
## 3	0	50.0
## 4	100	75.0
## 5	50	50.0
## 6	75	83.3
## X6.1.4a..Risk.of.terrorism		
## 1	0	
## 2	75	
## 3	50	
## 4	100	
## 5	75	
## 6	100	
## X6.1.4b..Level.of.illicit.arms.flows.within.the.country		
## 1	0	
## 2	50	
## 3	50	
## 4	25	
## 5	0	
## 6	100	
## X6.1.4c..Risk.of.organized.criminal.activity	X6.1.5..Armed.conflict	
## 1	0	0
## 2	25	100
## 3	50	75
## 4	100	100
## 5	75	75
## 6	50	100
## X6.1.5a..Presence.or.risk.of.armed.conflict		
## 1	0	
## 2	100	
## 3	75	
## 4	100	
## 5	75	
## 6	100	
## X6.1.6..Government.territorial.control		
## 1	0	
## 2	100	
## 3	100	
## 4	100	
## 5	100	
## 6	100	
## X6.1.6a..Government.territorial.control	X6.1.7..International.tensions	
## 1	0	0
## 2	100	75
## 3	100	50
## 4	100	75
## 5	100	50
## 6	100	100
## X6.1.7a..International.tensions	X6.2..Socio.economic.resilience	
## 1	0	46.6
## 2	75	62.3
## 3	50	56.5
## 4	75	84.3

## 5	50	40.7
## 6	100	58.2
## X6.2.1..Literacy	X6.2.1a..Adult.literacy.rate..15..years.old..both.sexes.	
## 1	26.6	26.6
## 2	97.6	97.6
## 3	76.1	76.1
## 4	99.9	99.9
## 5	56.2	56.2
## 6	98.7	98.7
## X6.2.2..Gender.equality	X6.2.2a..UNDP.Gender.Inequality.Index.score	
## 1	32.5	32.5
## 2	75.3	75.3
## 3	49.1	49.1
## 4	82.7	82.7
## 5	32.1	32.1
## 6	63.1	63.1
## X6.2.3..Social.inclusion	X6.2.3a..Poverty.gap.at..1.90.a.day..2011.PPP.....	
## 1	30.8	92.4
## 2	72.1	99.5
## 3	72.1	99.5
## 4	99.8	99.5
## 5	24.2	39.2
## 6	60.9	99.2
## X6.2.3b..Share.of.employment.in.the.informal.sector		
## 1	0	
## 2	50	
## 3	50	
## 4	100	
## 5	0	
## 6	50	
## X6.2.3c..Coverage.of.social.insurance.programs....of.population.		
## 1		0.0
## 2		66.7
## 3		66.7
## 4		100.0
## 5		33.3
## 6		33.3
## X6.2.4..Public.confidence.in.government		
## 1	0	
## 2	0	
## 3	0	
## 4	50	
## 5	50	
## 6	50	
## X6.2.4a..Public.confidence.in.government	X6.2.5..Local.media.and.reporting	
## 1	0	100
## 2	0	50
## 3	0	50
## 4	50	100
## 5	50	50
## 6	50	50
## X6.2.5a..Robust..open..diverse.local.media.and.reporting	X6.2.6..Inequality	
## 1	100	89.5
## 2	50	78.9
## 3	50	92.1
## 4	100	73.7
## 5	50	31.6



## 6		50	26.3
##	X6.2.6a..Gini.coefficient	X6.3..Infrastructure.adequacy	
## 1	89.5	0.0	
## 2	78.9	33.3	
## 3	92.1	41.7	
## 4	73.7	100.0	
## 5	31.6	33.3	
## 6	26.3	66.7	
##	X6.3.1..Adequacy.of.road.network	X6.3.1a..Adequacy.of.road.network	
## 1	0	0	
## 2	25	25	
## 3	25	25	
## 4	100	100	
## 5	25	25	
## 6	75	75	
##	X6.3.2..Adequacy.of.airports	X6.3.2a..Adequacy.of.airports	
## 1	0	0	
## 2	50	50	
## 3	50	50	
## 4	100	100	
## 5	50	50	
## 6	75	75	
##	X6.3.3..Adequacy.of.power.network	X6.3.3a..Adequacy.of.power.network	
## 1	0	0	
## 2	25	25	
## 3	50	50	
## 4	100	100	
## 5	25	25	
## 6	50	50	
##	X6.4..Environmental.risks	X6.4.1..Urbanisation	
## 1	61.3	85.6	
## 2	48.4	44.8	
## 3	59.8	30.9	
## 4	62.4	13.8	
## 5	45.6	39.0	
## 6	50.6	87.1	
##	X6.4.1a..Urban.population....of.total.population.	X6.4.2..Land.use	
## 1		85.6	73.4
## 2		44.8	75.3
## 3		30.9	73.5
## 4		13.8	73.4
## 5		39.0	47.7
## 6		87.1	64.7
##	X6.4.2a..Change.in.forest.area..percentage.points.		
## 1		73.4	
## 2		75.3	
## 3		73.5	
## 4		73.4	
## 5		47.7	
## 6		64.7	
##	X6.4.3..Natural.disaster.risk	X6.4.3a..Natural.disaster.risk	
## 1	25	25	
## 2	25	25	
## 3	75	75	
## 4	100	100	
## 5	50	50	
## 6	0	0	

##	X6.5..Public.health.vulnerabilities	X6.5.1..Access.to.quality.healthcare
## 1	45.1	61.9
## 2	46.5	62.1
## 3	49.0	71.0
## 4	63.5	62.1
## 5	39.6	70.0
## 6	57.1	73.1
##	X6.5.1a..Total.life.expectancy..years.	X6.5.1b..NCD.mortality.rate
## 1	31.9	38.2
## 2	70.1	64.9
## 3	65.3	79.8
## 4	82.2	77.8
## 5	21.8	63.1
## 6	65.8	74.6
##	X6.5.1c..Population.aged.65.	X6.5.1d..Tobacco.use....of.adults.
## 1	95.7	49.4
## 2	61.6	47.3
## 3	84.1	68.8
## 4	52.3	37.8
## 5	96.9	78.6
## 6	76.8	77.1
##	X6.5.1e..Level.of.adult.obesity....	
## 1	94.2	
## 2	66.7	
## 3	57.0	
## 4	60.1	
## 5	89.6	
## 6	71.5	
##	X6.5.2..Access.to.potable.water.and.sanitation	
## 1	43.2	
## 2	92.7	
## 3	89.3	
## 4	100.0	
## 5	37.4	
## 6	91.8	
##	X6.5.2a..Access.to.potable.water	
## 1	47.1	
## 2	86.7	
## 3	91.0	
## 4	100.0	
## 5	28.4	
## 6	96.2	
##	X6.5.2b..Access.to.at.least.basic.sanitation.facilities	
## 1	39.4	
## 2	98.6	
## 3	87.6	
## 4	100.0	
## 5	46.5	
## 6	87.5	
##	X6.5.3..Public.healthcare.spending.levels.per.capita	
## 1	0.1	
## 2	6.4	
## 3	10.8	
## 4	42.1	
## 5	1.1	
## 6	13.3	
##	X6.5.3a..Domestic.general.government.health.expenditure.per.capita..PPP.	

```
## 1 0.1
## 2 6.4
## 3 10.8
## 4 42.1
## 5 1.1
## 6 13.3
## X6.5.4..Trust.in.medical.and.health.advice
## 1 75
## 2 25
## 3 25
## 4 50
## 5 50
## 6 50
## X6.5.4a..Trust.medical.and.health.advice.from.the.government
## 1 100
## 2 0
## 3 0
## 4 50
## 5 50
## 6 50
## X6.5.4b..Trust.medical.and.health.advice.from.medical.workers
## 1 50
## 2 50
## 3 50
## 4 50
## 5 50
## 6 50
```

#### All countries and columns names from the external csv file

*# List out all countries*

```
unique(corona$Country)
```

```
## [1] "Afghanistan"
## [3] "Algeria"
## [5] "Angola"
## [7] "Argentina"
## [9] "Australia"
## [11] "Azerbaijan"
## [13] "Bahrain"
## [15] "Barbados"
## [17] "Belgium"
## [19] "Benin"
## [21] "Bolivia"
## [23] "Botswana"
## [25] "Brunei"
## [27] "Burkina Faso"
## [29] "Cabo Verde"
## [31] "Cameroon"
## [33] "Central African Republic"
## [35] "Chile"
## [37] "Colombia"
## [39] "Congo (Brazzaville)"
## [41] "Cook Islands"
## [43] "Côte d'Ivoire"
## [45] "Cuba"
## [47] "Czech Republic"
## [49] "Djibouti"
"Albania"
"Andorra"
"Antigua & Barbuda"
"Armenia"
"Austria"
"Bahamas"
"Bangladesh"
"Belarus"
"Belize"
"Bhutan"
"Bosnia and Hercegovina"
"Brazil"
"Bulgaria"
"Burundi"
"Cambodia"
"Canada"
"Chad"
"China"
"Comoros"
"Congo (Democratic Republic)"
"Costa Rica"
"Croatia"
"Cyprus"
"Denmark"
"Dominica"
```

## [51]	"Dominican Republic"	"Ecuador"
## [53]	"Egypt"	"El Salvador"
## [55]	"Equatorial Guinea"	"Eritrea"
## [57]	"Estonia"	"eSwatini"
## [59]	"Ethiopia"	"Fiji"
## [61]	"Finland"	"France"
## [63]	"Gabon"	"Gambia"
## [65]	"Georgia"	"Germany"
## [67]	"Ghana"	"Greece"
## [69]	"Grenada"	"Guatemala"
## [71]	"Guinea"	"Guinea-Bissau"
## [73]	"Guyana"	"Haiti"
## [75]	"Honduras"	"Hungary"
## [77]	"Iceland"	"India"
## [79]	"Indonesia"	"Iran"
## [81]	"Iraq"	"Ireland"
## [83]	"Israel"	"Italy"
## [85]	"Jamaica"	"Japan"
## [87]	"Jordan"	"Kazakhstan"
## [89]	"Kenya"	"Kiribati"
## [91]	"Kuwait"	"Kyrgyz Republic"
## [93]	"Laos"	"Latvia"
## [95]	"Lebanon"	"Lesotho"
## [97]	"Liberia"	"Libya"
## [99]	"Liechtenstein"	"Lithuania"
## [101]	"Luxembourg"	"Madagascar"
## [103]	"Malawi"	"Malaysia"
## [105]	"Maldives"	"Mali"
## [107]	"Malta"	"Marshall Islands"
## [109]	"Mauritania"	"Mauritius"
## [111]	"Mexico"	"Micronesia, Federated States of"
## [113]	"Moldova"	"Monaco"
## [115]	"Mongolia"	"Montenegro"
## [117]	"Morocco"	"Mozambique"
## [119]	"Myanmar"	"Namibia"
## [121]	"Nauru"	"Nepal"
## [123]	"Netherlands"	"New Zealand"
## [125]	"Nicaragua"	"Niger"
## [127]	"Nigeria"	"Niue"
## [129]	"North Korea"	"North Macedonia"
## [131]	"Norway"	"Oman"
## [133]	"Pakistan"	"Palau"
## [135]	"Panama"	"Papua New Guinea"
## [137]	"Paraguay"	"Peru"
## [139]	"Philippines"	"Poland"
## [141]	"Portugal"	"Qatar"
## [143]	"Romania"	"Russia"
## [145]	"Rwanda"	"Samoa"
## [147]	"San Marino"	"São Tomé and Príncipe"
## [149]	"Saudi Arabia"	"Senegal"
## [151]	"Serbia"	"Seychelles"
## [153]	"Sierra Leone"	"Singapore"
## [155]	"Slovakia"	"Slovenia"
## [157]	"Solomon Islands"	"Somalia"
## [159]	"South Africa"	"South Korea"
## [161]	"South Sudan"	"Spain"
## [163]	"Sri Lanka"	"St Kitts & Nevis"

```
## [165] "St Lucia" "St Vincent & The Grenadines"
## [167] "Sudan" "Suriname"
## [169] "Sweden" "Switzerland"
## [171] "Syria" "Tajikistan"
## [173] "Tanzania" "Thailand"
## [175] "Timor-Leste" "Togo"
## [177] "Tonga" "Trinidad and Tobago"
## [179] "Tunisia" "Turkey"
## [181] "Turkmenistan" "Tuvalu"
## [183] "Uganda" "Ukraine"
## [185] "United Arab Emirates" "United Kingdom"
## [187] "United States of America" "Uruguay"
## [189] "Uzbekistan" "Vanuatu"
## [191] "Venezuela" "Vietnam"
## [193] "Yemen" "Zambia"
## [195] "Zimbabwe"
```

*# List out all column names to help identify potential indicators*  
names(corona)

```
## [1] "Country"
## [2] "Year"
## [3] "OVERALL.SCORE"
## [4] "X1..PREVENTION.OF.THE.EMERGENCE.OR.RELEASE.OF.PATHOGENS"
## [5] "X1.1..Antimicrobial.resistance..AMR."
## [6] "X1.1.1..AMR.surveillance..detection.and.reporting"
## [7] "X1.1.1a..National.plan.for.AMR.priority.pathogens"
## [8] "X1.1.1b..Capacity.of.national.lab.lab.system.to.test.for.AMR.priority.pathogens"
## [9] "X1.1.1c..National.environmental.surveillance.for.AMR.residues.organisms"
## [10] "X1.1.2..Antimicrobial.control"
## [11] "X1.1.2a..National.law.s..requiring.prescription.for.antibiotic.use..humans."
## [12] "X1.1.2b..National.law.s..requiring.prescription.for.antibiotic.use..animals."
## [13] "X1.2..Zoonotic.disease"
## [14] "X1.2.1..National.planning.for.zoonotic.diseases.pathogens"
## [15] "X1.2.1a..Laws.plans.on.zoonotic.disease"
## [16] "X1.2.1b..Laws.plans.on.zoonotic.disease.spillover.from.animals.to.humans"
## [17] "X1.2.1c..Laws.plans.for.surveillance...control.of.multiple.zoonotic.pathogens"
## [18] "X1.2.1d..Cross.ministerial.department.agency.unit.for.zoonotic.disease"
## [19] "X1.2.2..Surveillance.systems.for.zoonotic.diseases.pathogens"
## [20] "X1.2.2a..Surveillance.reporting.mechanism.for.zoonotic.disease.for.livestock.owne
rs"
## [21] "X1.2.2b..Laws.regulations.on.data.confidentiality.to.protect.livestock.owners"
## [22] "X1.2.2c..Wildlife.zoonotic.disease.surveillance"
## [23] "X1.2.3..International.reporting.of.animal.disease.outbreaks"
## [24] "X1.2.3a..Annual.reporting.to.OIE.on.zoonotic.disease.incidence"
## [25] "X1.2.4..Animal.health.workforce"
## [26] "X1.2.4a..Number.of.veterinarians.per.100.000.people"
## [27] "X1.2.4b..Number.of.veterinary.para.professionals.per.100.000.people"
## [28] "X1.2.5..Private.sector.and.zoonotic.disease"
## [29] "X1.2.5a..Inclusion.of.private.sector.in.national.plan.law.on.zoonotic.disease"
## [30] "X1.3..Biosecurity"
## [31] "X1.3.1..Whole.of.government.biosecurity.systems"
## [32] "X1.3.1a..Updated.national.records.of.especially.dangerous.pathogen.toxin.inventor
ies"
## [33] "X1.3.1b..Biosecurity.laws.on.facility.security.for.especially.dangerous.pathogens
"
## [34] "X1.3.1c..Agency.for.enforcement.of.biosecurity.laws.regulations"
```

## [35] "X1.3.1d..Consolidation.of.especially.dangerous.pathogens.into.minimum...of.facilities"

## [36] "X1.3.1e..Capacity.to.conduct.tests.for.anthrax.Ebola.without.culturing.live.pathogens"

## [37] "X1.3.2..Biosecurity.training.and.practices"

## [38] "X1.3.2a..Biosecurity.training.using.a.standardised..required.approach"

## [39] "X1.3.3..Personnel.vetting..regulating.access.to.sensitive.locations"

## [40] "X1.3.3a..Personnel.checks.for.permission.to.access.to.especially.dangerous.pathogens"

## [41] "X1.3.4..Transportation.security"

## [42] "X1.3.4a..National.transport.regulations.for.Category.A.and.B.infectious.substances"

## [43] "X1.3.5..Cross.border.transfer.and.end.user.screening"

## [44] "X1.3.5a..Laws.regulations.on.cross.border.transfer.and.end.user.screening"

## [45] "X1.4..Biosafety"

## [46] "X1.4.1..Whole.of.government.biosafety.systems"

## [47] "X1.4.1a..Biosafety.laws.regulations"

## [48] "X1.4.1b..Agency.for.enforcement.of.biosafety.laws.regulations"

## [49] "X1.4.2..Biosafety.training.and.practices"

## [50] "X1.4.2a..Biosafety.training.using.a.standardised..required.approach"

## [51] "X1.5..Dual.use.research.and.culture.of.responsible.science"

## [52] "X1.5.1..Oversight.of.dual.use.research"

## [53] "X1.5.1a..Evidence.of.national.assessment.of.dual.use.research"

## [54] "X1.5.1b..National.law.regulation.on.oversight.of.dual.use.research"

## [55] "X1.5.1c..Existence.of.agency.responsible.for.oversight.of.dual.use.research"

## [56] "X1.5.2..Screening.requirements.for.providers.of.genetic.material"

## [57] "X1.5.2a..Requirement.to.screen.synthesised.DNA.against.list.prior.to.sale"

## [58] "X1.6..Immunization"

## [59] "X1.6.1..Vaccination.rates"

## [60] "X1.6.1a..Immunization.rate.for.humans..measles.MCV2."

## [61] "X1.6.1b..Availability.of.vaccination.figures.for.livestock..FMD..through.OIE.data base"

## [62] "X2..EARLY.DETECTION...REPORTING.FOR.EPIDEMICS.OF.POTENTIAL.INT.L.CONCERN"

## [63] "X2.1..Laboratory.systems.strength.and.quality"

## [64] "X2.1.1..Lab.capacity.for.detecting.priority.diseases"

## [65] "X2.1.1a..Capacity.of.national.lab.system.to.conduct.5.or.more.WHO.core.tests"

## [66] "X2.1.1b..Plan.to.conduct.testing.during.a.public.health.emergency"

## [67] "X2.1.2..Laboratory.quality.systems"

## [68] "X2.1.2a..Existence.of.an.accredited.national.lab.serving.as.a.reference.facility"

## [69] "X2.1.2b..External.quality.assurance.of.a.national.lab.serving.as.a.reference.facility"

## [70] "X2.2..Laboratory.supply.chains"

## [71] "X2.2.1..Specimen.referral.and.transport.system"

## [72] "X2.2.1a..Is.there.a.nationwide.specimen.transport.system."

## [73] "X2.2.2..Laboratory.cooperation.and.coordination"

## [74] "X2.2.2a..Plan.to.rapidly.authorize.license.laboratories.to.scale.up.testing.during.an.outbreak"

## [75] "X2.3..Real.time.surveillance.and.reporting"

## [76] "X2.3.1..Indicator.and.event.based.surveillance.and.reporting.systems"

## [77] "X2.3.1a..Evidence.of.ongoing.event.based.surveillance.and.analysis"

## [78] "X2.3.1b..Evidence.of.reporting.a.potential.PHEIC.to.the.WHO..last.2.years."

## [79] "X2.3.2..Interoperable..interconnected..electronic.real.time.reporting.systems"

## [80] "X2.3.2a..Electronic.national.and.sub.national.reporting.surveillance.system"

## [81] "X2.3.2b..Collection.of.ongoing.real.time.lab.data.by.electronic.surveillance.system"

## [82] "X2.4..Surveillance.data.accessibility.and.transparency"

## [83] "X2.4.1..Coverage.and.use.of.electronic.health.records"

```
## [84] "X2.4.1a..Common.usage.of.electronic.health.records"
## [85] "X2.4.1b..Public.health.system.access.to.individual.electronic.health.records"
## [86] "X2.4.1c..Existence.of.data.standards.for.health.record.data.comparability"
## [87] "X2.4.2..Data.integration.between.human..animal.and.environmental.health.sectors"
## [88] "X2.4.2a..Data.sharing.mechanisms"
## [89] "X2.4.3..Transparency.of.surveillance.data"
## [90] "X2.4.3a..Availability.of.de.identified.health.surveillance.data.on.disease.outbre
aks"
## [91] "X2.4.4..Ethical.considerations.during.surveillance"
## [92] "X2.4.4a..Confidentiality.legislation.regulations.for.identifiable.health.informat
ion"
## [93] "X2.4.4b..Inclusion.of.cyber.protections.in.health.data.confidentiality.law.regula
tion"
## [94] "X2.4.5..International.data.sharing"
## [95] "X2.4.5a..Cooperative.commitments.or.agreements.within.regions"
## [96] "X2.5..Case.based.investigation"
## [97] "X2.5.1..Case.investigation.and.contact.tracing"
## [98] "X2.5.1a..National.support.to.conduct.contact.tracing.in.the.event.of.a.public.he
alth.emergency"
## [99] "X2.5.1b..Provision.of.wraparound.services.to.enable.self.isolation.quarantine.as.
recommended"
## [100] "X2.5.2..Point.of.entry.management"
## [101] "X2.5.2a..Strategy.for.tracing.and.quarantining.international.travelers"
## [102] "X2.6..Epidemiology.workforce"
## [103] "X2.6.1..Existence.of.applied.epidemiology.training.program.such.FETP.and.FETPV"
## [104] "X2.6.1a..Access.to.field.epidemiology.training.program.in.country.and.or.abroad"
## [105] "X2.6.1b..Existence.of.field.epidemiology.training.for.animal.health.professionals
"
## [106] "X2.6.2..Epidemiology.workforce.capacity"
## [107] "X2.6.2a..Evidence.of.at.least.1.trained.field.epidemiologist.per.200.000.people"
## [108] "X3..RAPID.RESPONSE.TO.AND.MITIGATION.OF.THE.SPREAD.OF.AN.EPIDEMIC"
## [109] "X3.1..Emergency.preparedness.and.response.planning"
## [110] "X3.1.1..National.public.health.emergency.preparedness.and.response.plan"
## [111] "X3.1.1a..National.emergency.response.plan.for.diseases.with.pandemic.potential"
## [112] "X3.1.1b..National.public.health.emergency.response.plan.updated.in.past.3.years"
## [113] "X3.1.1c..Vulnerable.populations.in.national.public.health.emergency.response.plan
"
## [114] "X3.1.1d..Existence.of.public.pandemic.influenza.preparedness.plan.updated.since.2
009"
## [115] "X3.1.2..Private.sector.involvement.in.response.planning"
## [116] "X3.1.2a..Mechanism.to.engage.private.sector.in.outbreak.preparedness.response"
## [117] "X3.1.3..Non.pharmaceutical.interventions.planning"
## [118] "X3.1.3a..Policy.plan.guidelines.in.place.to.implement.non.pharmaceutical.interven
tions..NPIs."
## [119] "X3.2..Exercising.response.plans"
## [120] "X3.2.1..Activating.response.plans"
## [121] "X3.2.1a..Completion.of.biological.focused.IHR.exercise.with.the.WHO.in.past.year"
## [122] "X3.2.1b..Evidence.of.bio.focused.exercise.to.identify.gaps.best.practices"
## [123] "X3.2.2..Private.sector.engagement.in.exercises"
## [124] "X3.2.2a..Evidence.of.national.level.biological.threat.focused.exercise.that.inclu
des.private.sector"
## [125] "X3.3..Emergency.response.operation"
## [126] "X3.3.1..Emergency.response.operation"
## [127] "X3.3.1a..Existence.of.Emergency.Operations.Center..EOC."
## [128] "X3.3.1b..Requirement.for.EOC.to.conduct.evidence.EOC.conducts.at.least.annual.dri
lls"
## [129] "X3.3.1c..EOC.activation.within.120.minutes.of.identification.of.emergency.scenari
```

o"

- ## [130] "X3.4..Linking.public.health.and.security.authorities"
- ## [131] "X3.4.1..Public.health.and.security.authorities.linked.for.a.biological.event"
- ## [132] "X3.4.1a..Joint.exercise.procedures.for.potential.deliberate.biological.events"
- ## [133] "X3.5..Risk.communication"
- ## [134] "X3.5.1..Risk.communication.planning"
- ## [135] "X3.5.1a..Risk.communication.plan.for.specific.use.during.a.public.health.emergency"
- ## [136] "X3.5.1b..Inclusion.of.different.population...sector.needs.in.risk.communication.plan"
- ## [137] "X3.5.1c..Designation.of.a.specific.government.spokesperson.during.a.public.health.emergency"
- ## [138] "X3.5.2..Public.health.systems.communication"
- ## [139] "X3.5.2a..Government.use.of.media.platforms.to.share.info.on.public.health.emergencies"
- ## [140] "X3.5.2b..Evidence.that.senior.leaders.have.shared.mis.disinformation.on.infectious.diseases"
- ## [141] "X3.6..Access.to.communications.infrastructure"
- ## [142] "X3.6.1..Internet.users"
- ## [143] "X3.6.1a..Percentage.of.households.with.Internet"
- ## [144] "X3.6.2..Mobile.subscribers"
- ## [145] "X3.6.2a..Mobile.cellular.telephone.subscriptions.per.100.inhabitants"
- ## [146] "X3.6.3..Female.access.to.a.mobile.phone"
- ## [147] "X3.6.3a..Gender.gap.in.access.to.a.mobile.phone..percentage.points."
- ## [148] "X3.6.4..Female.access.to.the.Internet"
- ## [149] "X3.6.4a..Gender.gap.in.access.to.the.Internet..percentage.points."
- ## [150] "X3.7..Trade.and.travel.restrictions"
- ## [151] "X3.7.1..Trade.restrictions"
- ## [152] "X3.7.1a..Restrictions.on.export.import.of.medical.goods.due.to.an.infectious.disease.outbreak"
- ## [153] "X3.7.1b..Restrictions.on.movement.and.or.exports.imports.due.to.disease.outbreak"
- ## [154] "X3.7.2..Travel.restrictions"
- ## [155] "X3.7.2a..Evidence.of.travel.ban.due.to.an.infectious.disease.outbreak"
- ## [156] "X4..SUFFICIENT...ROBUST.HEALTH.SECTOR.TO.TREAT.THE.SICK...PROTECT.HEALTH.WORKERS"
- ## [157] "X4.1..Health.capacity.in.clinics..hospitals.and.community.care.centers"
- ## [158] "X4.1.1..Available.human.resources.for.the.broader.healthcare.system"
- ## [159] "X4.1.1a..Doctors.per.100.000.people"
- ## [160] "X4.1.1b..Nurses.and.midwives.per.100.000.people"
- ## [161] "X4.1.1c..Updated.health.workforce.strategy.to.address.human.resource.shortfalls"
- ## [162] "X4.1.2..Facilities.capacity"
- ## [163] "X4.1.2a..Hospital.beds.per.100.000.people"
- ## [164] "X4.1.2b..In.country.capacity.to.isolate.patients.with.highly.communicable.diseases"
- ## [165] "X4.1.2c..Demonstrated.capacity...evidence.of.plan.to.expand.isolation.capacity"
- ## [166] "X4.2..Supply.chain.for.health.system.and.healthcare.workers"
- ## [167] "X4.2.1..Routine.health.care.and.laboratory.system.supply"
- ## [168] "X4.2.1a..National.procurement.protocol.for.the.acquisition.of.routine.laboratory.medical.supplies"
- ## [169] "X4.2.2..Stockpiling.for.emergencies"
- ## [170] "X4.2.2a..Stockpile.of.medical.supplies.for.national.use.during.a.public.health.emergency"
- ## [171] "X4.2.2b..Stockpile.of.laboratory.supplies.for.national.use.during.a.public.health.emergency"
- ## [172] "X4.2.2c..Annual.review.of.national.stockpile.to.ensure.sufficient.supply"
- ## [173] "X4.2.3..Manufacturing.and.procurement.for.emergencies"
- ## [174] "X4.2.3a..Plan.agreement.to.produce.procure.medical.supplies.during.a.public.health.emergency"



## [175] "X4.2.3b..Plan.agreement.to.produce.procure.lab.supplies.during.a.public.health.emergency"

## [176] "X4.3..Medical.countermeasures.and.personnel.deployment"

## [177] "X4.3.1..System.for.dispensing.MCMs.during.a.public.health.emergency"

## [178] "X4.3.1a..Plan.program.guidelines.for.dispensing.MCMs.during.a.public.health.emergency"

## [179] "X4.3.2..System.for.receiving.foreign.health.personnel.during.a.public.health.emergency"

## [180] "X4.3.2a..Plan.to.receive.foreign.health.personnel.during.a.public.health.emergency"

## [181] "X4.4..Healthcare.access"

## [182] "X4.4.1..Access.to.healthcare"

## [183] "X4.4.1a..Constitutional.guarantee.of.citizens..right.to.medical.care"

## [184] "X4.4.1b..Access.to.skilled.birth.attendants...of.population."

## [185] "X4.4.1c..Out.of.pocket.health.expenditures.per.capita..PPP..current.international..."

## [186] "X4.4.2..Paid.medical.leave"

## [187] "X4.4.2a..Guaranteed.paid.sick.leave"

## [188] "X4.4.3..Healthcare.worker.access.to.healthcare"

## [189] "X4.4.3a..Government.prioritisation.of.care.for.healthcare.workers.during.response"

## [190] "X4.5..Communications.with.healthcare.workers.during.a.public.health.emergency"

## [191] "X4.5.1..Communication.with.healthcare.workers"

## [192] "X4.5.1a..Existence.of.system.for.communication.during.a.public.health.emergency"

## [193] "X4.5.1b..Inclusion.of.public.and.private.sector.in.healthcare.communication.system"

## [194] "X4.6..Infection.control.practices"

## [195] "X4.6.1..Healthcare.associated.infection..HCAI..monitoring"

## [196] "X4.6.1a..Evidence.of.national.public.health.system.monitoring.and.tracking.of.HCAIs"

## [197] "X4.7..Capacity.to.test.and.approve.new.medical.countermeasures"

## [198] "X4.7.1..Regulatory.process.for.clinical.trials.of.unregistered.interventions"

## [199] "X4.7.1a..Requirement.for.ethical.review.before.beginning.a.clinical.trial"

## [200] "X4.7.1b..Expedited.approval.for.clinical.trials.of.unregistered.MCMs.during.epidemics"

## [201] "X4.7.2..Regulatory.process.for.approving.medical.countermeasures"

## [202] "X4.7.2a..Existence.of.agency.responsible.for.approving.new.human.MCMs"

## [203] "X4.7.2b..Expedited.approval.for.human.MCMs.during.public.health.emergencies"

## [204] "X5..COMMITMENTS.TO.IMPROVING.NATIONAL.CAPACITY..FINANCING.AND.ADHERENCE.TO.NORMS"

## [205] "X5.1..IHR.reporting.compliance.and.disaster.risk.reduction"

## [206] "X5.1.1..Official.IHR.reporting"

## [207] "X5.1.1a..Submission.of.IHR.reports.to.the.WHO.in.past.year"

## [208] "X5.1.2..Integration.of.health.into.disaster.risk.reduction"

## [209] "X5.1.2a..Existence.of.specific.risk.reduction.strategies.for.epidemics.and.pandemics"

## [210] "X5.2..Cross.border.agreements.on.public.health.and.animal.health.emergency.response"

## [211] "X5.2.1..Cross.border.agreements"

## [212] "X5.2.1a..Existence.of.public.health.emergency.agreements.with.regional.neighbors"

## [213] "X5.2.1b..Existence.of.animal.health.emergency.agreements.with.regional.neighbors"

## [214] "X5.3..International.commitments"

## [215] "X5.3.1..Participation.in.international.agreements"

## [216] "X5.3.1a..Biological.and.Toxin.Weapons.Convention.status"

## [217] "X5.3.1b..Submission.of.CBMs.to.the.Biological.and.Toxin.Weapons.Convention"

## [218] "X5.3.1c..Submission.of.UNSCR.1540.reports"

## [219] "X5.3.1d..Extent.of.UNSCR.1540.implementation.on.public.health.emergencies"

## [220] "X5.3.2..Voluntary.memberships"

## [221] "X5.3.2a..Membership.in.global.health.security.and.or.biological.weapons.agreement  
s"

## [222] "X5.4..JEE.and.PVS"

## [223] "X5.4.1..Completion.and.publication.of.a.JEE.assessment.and.gap.analysis"

## [224] "X5.4.1a..Completion.and.publication.of.JEE..or.GHSA.pilot.external.assessment..re  
port"

## [225] "X5.4.1b..Completion.and.publication.of.a.NAPHS.or.GHSA.roadmap"

## [226] "X5.4.2..Completion.and.publication.of.a.PVS.assessment.and.gap.analysis"

## [227] "X5.4.2a..Completion.and.publication.of.PVS.report..past.five.years."

## [228] "X5.4.2b..Completion.and.publication.of.PVS.gap.analysis..past.five.years."

## [229] "X5.5..Financing"

## [230] "X5.5.1..National.financing.for.epidemic.preparedness"

## [231] "X5.5.1a..Evidence.of.allocated.national.funds.to.improve.capacity.to.address.epid  
emic.threats"

## [232] "X5.5.2..Financing.under.JEE.and.PVS.reports.and.gap.analyses"

## [233] "X5.5.2a..National.budget.to.address.gaps.identified.in.JEE..NAPHS.or.GHSA.roadmap  
"

## [234] "X5.5.2b..National.budget.to.address.gaps.identified.in.PVS.assessment.or.gap.anal  
ysis"

## [235] "X5.5.3..Financing.for.emergency.response"

## [236] "X5.5.3a..Emergency.public.financing.during.a.public.health.emergency"

## [237] "X5.5.4..Accountability.for.international.commitments.to.address.epidemic.threats"

## [238] "X5.5.4a..Commitments.to.improve.domestic.or.foreign.capacity.for.epidemic.threats  
"

## [239] "X5.5.4b..Investments.to.improve.domestic.or.foreign.capacity.for.epidemic.threats  
"

## [240] "X5.5.4c..Evidence.that.the.country.has.fulfilled.its.full.WHO.contribution.within  
.the.past.two.years"

## [241] "X5.6..Commitment.to.sharing.of.genetic...biological.data...specimens"

## [242] "X5.6.1..Commitment.to.share.data.and.specimens.in.emergency.non.emergency.research  
h"

## [243] "X5.6.1a..Sharing.of.genetic.biological.data.and.materials.beyond.influenza"

## [244] "X5.6.1b..Evidence.of.non.compliance.with.sample.sharing.element.of.PIP.framework"

## [245] "X5.6.1c..Evidence.of.non.sharing.of.pandemic.pathogen.samples.during.an.outbreak"

## [246] "X6..OVERALL.RISK.ENVIRONMENT.AND.COUNTRY.VULNERABILITY.TO.BIOLOGICAL.THREATS"

## [247] "X6.1..Political.and.security.risk"

## [248] "X6.1.1..Government.effectiveness"

## [249] "X6.1.1a..Policy.formation"

## [250] "X6.1.1b..Quality.of.bureaucracy"

## [251] "X6.1.1c..Excessive.bureaucracy.red.tape"

## [252] "X6.1.1d..Vested.interests.cronyism"

## [253] "X6.1.1e..Corruption"

## [254] "X6.1.1f..Accountability.of.public.officials"

## [255] "X6.1.1g..Human.rights.risk"

## [256] "X6.1.2..Orderly.transfers.of.power"

## [257] "X6.1.2a..Orderly.transfers.of.power"

## [258] "X6.1.3..Risk.of.social.unrest"

## [259] "X6.1.3a..Risk.of.social.unrest"

## [260] "X6.1.4..Illicit.activities.by.non.state.actors"

## [261] "X6.1.4a..Risk.of.terrorism"

## [262] "X6.1.4b..Level.of.illicit.arms.flows.within.the.country"

## [263] "X6.1.4c..Risk.of.organized.criminal.activity"

## [264] "X6.1.5..Armed.conflict"

## [265] "X6.1.5a..Presence.or.risk.of.armed.conflict"

## [266] "X6.1.6..Government.territorial.control"

## [267] "X6.1.6a..Government.territorial.control"

## [268] "X6.1.7..International.tensions"

```

## [269] "X6.1.7a..International.tensions"
## [270] "X6.2..Socio.economic.resilience"
## [271] "X6.2.1..Literacy"
## [272] "X6.2.1a..Adult.literacy.rate..15..years.old..both.sexes."
## [273] "X6.2.2..Gender.equality"
## [274] "X6.2.2a..UNDP.Gender.Inequality.Index.score"
## [275] "X6.2.3..Social.inclusion"
## [276] "X6.2.3a..Poverty.gap.at..1.90.a.day..2011.PPP....."
## [277] "X6.2.3b..Share.of.employment.in.the.informal.sector"
## [278] "X6.2.3c..Coverage.of.social.insurance.programs....of.population."
## [279] "X6.2.4..Public.confidence.in.government"
## [280] "X6.2.4a..Public.confidence.in.government"
## [281] "X6.2.5..Local.media.and.reporting"
## [282] "X6.2.5a..Robust..open..diverse.local.media.and.reporting"
## [283] "X6.2.6..Inequality"
## [284] "X6.2.6a..Gini.coefficient"
## [285] "X6.3..Infrastructure.adequacy"
## [286] "X6.3.1..Adequacy.of.road.network"
## [287] "X6.3.1a..Adequacy.of.road.network"
## [288] "X6.3.2..Adequacy.of.airports"
## [289] "X6.3.2a..Adequacy.of.airports"
## [290] "X6.3.3..Adequacy.of.power.network"
## [291] "X6.3.3a..Adequacy.of.power.network"
## [292] "X6.4..Environmental.risks"
## [293] "X6.4.1..Urbanisation"
## [294] "X6.4.1a..Urban.population....of.total.population."
## [295] "X6.4.2..Land.use"
## [296] "X6.4.2a..Change.in.forest.area..percentage.points."
## [297] "X6.4.3..Natural.disaster.risk"
## [298] "X6.4.3a..Natural.disaster.risk"
## [299] "X6.5..Public.health.vulnerabilities"
## [300] "X6.5.1..Access.to.quality.healthcare"
## [301] "X6.5.1a..Total.life.expectancy..years."
## [302] "X6.5.1b..NCD.mortality.rate"
## [303] "X6.5.1c..Population.aged.65."
## [304] "X6.5.1d..Tobacco.use....of.adults."
## [305] "X6.5.1e..Level.of.adult.obesity...."
## [306] "X6.5.2..Access.to.potable.water.and.sanitation"
## [307] "X6.5.2a..Access.to.potable.water"
## [308] "X6.5.2b..Access.to.at.least.basic.sanitation.facilities"
## [309] "X6.5.3..Public.healthcare.spending.levels.per.capita"
## [310] "X6.5.3a..Domestic.general.government.health.expenditure.per.capita..PPP."
## [311] "X6.5.4..Trust.in.medical.and.health.advice"
## [312] "X6.5.4a..Trust.medical.and.health.advice.from.the.government"
## [313] "X6.5.4b..Trust.medical.and.health.advice.from.medical.workers"

```

## Linear Regression Model Summary for Hungary

```
# c19ProSo01
```

```
summary(hungary_lm_1)
```

```

##
## Call:
## lm(formula = c19ProSo01 ~ ., data = hungary)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.4128 -0.6822  0.0828  0.8160  2.9709

```

```
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.537065   0.744220  -0.722  0.47115
## isoFriends_inPerson -0.003361   0.035583  -0.094  0.92481
## isoOthPpl_inPerson  0.025086   0.044807   0.560  0.57606
## isoFriends_online  -0.001576   0.036624  -0.043  0.96571
## isoOthPpl_online   0.018149   0.031239   0.581  0.56176
## lone01          0.059990   0.085708   0.700  0.48459
## lone02          0.071697   0.080936   0.886  0.37651
## lone03          0.119393   0.066558   1.794  0.07399 .
## happy          0.083205   0.067669   1.230  0.21995
## lifeSat        -0.099637   0.111948  -0.890  0.37427
## MLQ            0.035464   0.055372   0.640  0.52243
## bor01          -0.065040   0.060728  -1.071  0.28515
## bor02          -0.050790   0.057352  -0.886  0.37665
## bor03          -0.017900   0.046996  -0.381  0.70359
## consp01        0.010677   0.041670   0.256  0.79797
## consp02       -0.008935   0.049111  -0.182  0.85577
## consp03       -0.014531   0.033621  -0.432  0.66596
## c19perBeh01     0.067436   0.101904   0.662  0.50871
## c19perBeh02     0.006086   0.114513   0.053  0.95765
## c19perBeh03    -0.130814   0.059468  -2.200  0.02870 *
## c19RCA01        0.054025   0.044564   1.212  0.22649
## c19RCA02       -0.172794   0.075222  -2.297  0.02240 *
## c19RCA03        0.100427   0.045425   2.211  0.02791 *
## gender         0.392598   0.207754   1.890  0.05990 .
## age           -0.010432   0.071696  -0.146  0.88443
## edu           -0.032338   0.050036  -0.646  0.51866
## c19ProSo02      0.368623   0.049268   7.482 1.11e-12 ***
## c19ProSo03      0.251164   0.050652   4.959 1.28e-06 ***
## c19ProSo04      0.140175   0.053379   2.626  0.00915 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.182 on 262 degrees of freedom
## (4 observations deleted due to missingness)
## Multiple R-squared:  0.4288, Adjusted R-squared:  0.3678
## F-statistic: 7.026 on 28 and 262 DF,  p-value: < 2.2e-16
```

# c19ProSo02

```
summary(hungary_lm_2)
```

```
##
## Call:
## lm(formula = c19ProSo02 ~ ., data = hungary)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.8149 -0.8350  0.1270  0.8957  4.0590
##
## Coefficients:
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.268244   0.847776   0.316  0.751944
## isoFriends_inPerson 0.010136   0.040497   0.250  0.802566
## isoOthPpl_inPerson  0.001412   0.051031   0.028  0.977949
## isoFriends_online   0.066677   0.041483   1.607  0.109185
```

```
## isoOthPpl_online      0.024556    0.035548    0.691 0.490316
## lone01                -0.027751    0.097632   -0.284 0.776453
## lone02                -0.042667    0.092224   -0.463 0.644002
## lone03                -0.080953    0.076058   -1.064 0.288143
## happy                 -0.062491    0.077148   -0.810 0.418666
## lifeSat               0.169652    0.127185    1.334 0.183394
## MLQ                   -0.059560    0.062969   -0.946 0.345091
## bor01                 0.073666    0.069124    1.066 0.287533
## bor02                 -0.024513    0.065360   -0.375 0.707926
## bor03                 0.003075    0.053507    0.057 0.954213
## consp01               0.011280    0.047431    0.238 0.812207
## consp02               -0.031994    0.055869   -0.573 0.567361
## consp03               0.039227    0.038206    1.027 0.305499
## c19perBeh01           0.020084    0.116081    0.173 0.862770
## c19perBeh02           -0.097068    0.130205   -0.746 0.456636
## c19perBeh03           0.252495    0.066506    3.797 0.000182 ***
## c19RCA01              0.004715    0.050865    0.093 0.926213
## c19RCA02              0.044861    0.086433    0.519 0.604179
## c19RCA03              -0.004902    0.052183   -0.094 0.925229
## gender                 -0.997763    0.229961   -4.339 2.05e-05 ***
## age                   -0.024762    0.081595   -0.303 0.761769
## edu                    0.038569    0.056949    0.677 0.498841
## c19ProSo01            0.477579    0.063831    7.482 1.11e-12 ***
## c19ProSo03            0.180631    0.059257    3.048 0.002537 **
## c19ProSo04            -0.123236    0.061079   -2.018 0.044648 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.345 on 262 degrees of freedom
## (4 observations deleted due to missingness)
## Multiple R-squared:  0.4203, Adjusted R-squared:  0.3583
## F-statistic: 6.784 on 28 and 262 DF,  p-value: < 2.2e-16
```

```
# c19ProSo03
```

```
summary(hungary_lm_3)
```

```
##
## Call:
## lm(formula = c19ProSo03 ~ ., data = hungary)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.8815 -0.8505  0.0540  0.9075  3.8196
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.403739    0.868416    0.465  0.64238
## isoFriends_inPerson -0.034106    0.041444   -0.823  0.41130
## isoOthPpl_inPerson  0.007776    0.052283    0.149  0.88188
## isoFriends_online  -0.056233    0.042570   -1.321  0.18767
## isoOthPpl_online  -0.051478    0.036316   -1.418  0.15752
## lone01           -0.109270    0.099818   -1.095  0.27466
## lone02            0.090912    0.094362    0.963  0.33621
## lone03           -0.097374    0.077863   -1.251  0.21220
## happy            -0.017392    0.079136   -0.220  0.82622
## lifeSat          -0.060861    0.130697   -0.466  0.64184
## MLQ               0.114031    0.064241    1.775  0.07705 .
```

```
## bor01      0.111589  0.070640  1.580  0.11538
## bor02      0.018868  0.066974  0.282  0.77838
## bor03     -0.026882  0.054797 -0.491  0.62414
## consp01     0.043358  0.048528  0.893  0.37242
## consp02    -0.022835  0.057260 -0.399  0.69036
## consp03    -0.021492  0.039201 -0.548  0.58398
## c19perBeh01  0.145690  0.118599  1.228  0.22039
## c19perBeh02 -0.042119  0.133521 -0.315  0.75267
## c19perBeh03  0.012258  0.069985  0.175  0.86110
## c19RCA01    -0.043692  0.052046 -0.839  0.40196
## c19RCA02     0.026643  0.088587  0.301  0.76384
## c19RCA03    -0.056126  0.053354 -1.052  0.29379
## gender      0.144932  0.243765  0.595  0.55265
## age         0.002143  0.083615  0.026  0.97958
## edu         0.037244  0.058354  0.638  0.52387
## c19ProSo01  0.341590  0.068888  4.959 1.28e-06 ***
## c19ProSo02  0.189617  0.062205  3.048  0.00254 **
## c19ProSo04  0.287364  0.060514  4.749 3.37e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.378 on 262 degrees of freedom
## (4 observations deleted due to missingness)
## Multiple R-squared:  0.3774, Adjusted R-squared:  0.3109
## F-statistic: 5.672 on 28 and 262 DF,  p-value: 4.343e-15
```

# c19ProSo04

```
summary(hungary_lm_4)
```

```
##
## Call:
## lm(formula = c19ProSo04 ~ ., data = hungary)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.5427 -0.7955  0.1450  0.8939  2.6945
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.323534    0.847147  -1.562  0.11941
## isoFriends_inPerson -0.003754    0.040652  -0.092  0.92649
## isoOthPpl_inPerson  0.051247    0.051123   1.002  0.31706
## isoFriends_online  0.034182    0.041788   0.818  0.41412
## isoOthPpl_online  0.052265    0.035566   1.470  0.14289
## lone01         0.047754    0.097965   0.487  0.62634
## lone02        -0.049702    0.092553  -0.537  0.59172
## lone03         0.048399    0.076446   0.633  0.52721
## happy         -0.004001    0.077531  -0.052  0.95888
## lifeSat        0.114235    0.127895   0.893  0.37257
## MLQ           0.013587    0.063305   0.215  0.83023
## bor01         -0.074513    0.069378  -1.074  0.28380
## bor02         0.037899    0.065578   0.578  0.56381
## bor03        -0.047462    0.053626  -0.885  0.37693
## consp01       -0.013393    0.047605  -0.281  0.77867
## consp02       -0.044769    0.056043  -0.799  0.42511
## consp03       0.034519    0.038365   0.900  0.36908
## c19perBeh01    0.151759    0.116140   1.307  0.19247
```

```
## c19perBeh02      0.140805    0.130537    1.079    0.28173
## c19perBeh03      0.139404    0.068021    2.049    0.04142 *
## c19RCA01         0.050217    0.050960    0.985    0.32533
## c19RCA02         0.086538    0.086634    0.999    0.31877
## c19RCA03        -0.078913    0.052150   -1.513    0.13144
## gender           0.069096    0.238924    0.289    0.77266
## age             -0.072376    0.081790   -0.885    0.37703
## edu              0.123457    0.056699    2.177    0.03034 *
## c19ProSo01       0.182957    0.069670    2.626    0.00915 **
## c19ProSo02       -0.124152    0.061533   -2.018    0.04465 *
## c19ProSo03       0.275781    0.058075    4.749    3.37e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.35 on 262 degrees of freedom
## (4 observations deleted due to missingness)
## Multiple R-squared:  0.2899, Adjusted R-squared:  0.2141
## F-statistic: 3.821 on 28 and 262 DF,  p-value: 4.652e-09
```

### Linear Regression Model Summary for Iran

# c19ProSo01

summary(iran\_lm\_1)

```
##
## Call:
## lm(formula = c19ProSo01 ~ ., data = iran)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.5637 -0.6521  0.0318  0.6792  3.2221
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.940052    1.274946  -0.737   0.46226
## isoFriends_inPerson -0.052906    0.073489  -0.720   0.47288
## isoOthPpl_inPerson -0.080835    0.076686  -1.054   0.29381
## isoFriends_online  0.095419    0.059628   1.600   0.11199
## isoOthPpl_online  -0.034935    0.058848  -0.594   0.55378
## lone01          0.122141    0.136719   0.893   0.37332
## lone02         -0.203357    0.151838  -1.339   0.18283
## lone03          0.287910    0.148037   1.945   0.05397 .
## happy           0.156337    0.088735   1.762   0.08046 .
## lifeSat         0.021520    0.134231   0.160   0.87288
## MLQ            -0.222778    0.112597  -1.979   0.05000 *
## bor01           0.172221    0.106753   1.613   0.10913
## bor02          -0.053558    0.113654  -0.471   0.63827
## bor03          -0.115502    0.081769  -1.413   0.16020
## consp01         0.218288    0.091130   2.395   0.01804 *
## consp02        -0.355027    0.105737  -3.358   0.00103 **
## consp03         0.078419    0.044499   1.762   0.08039 .
## c19perBeh01     -0.005881    0.133362  -0.044   0.96489
## c19perBeh02      0.025094    0.206797   0.121   0.90361
## c19perBeh03     -0.072113    0.116577  -0.619   0.53728
## c19RCA01        0.133601    0.137777   0.970   0.33401
## c19RCA02        0.106121    0.119650   0.887   0.37677
## c19RCA03       -0.158816    0.148284  -1.071   0.28616
## gender          0.441586    0.244201   1.808   0.07289 .
```



```
## age          0.248680    0.101230    2.457    0.01536 *
## edu          -0.129005    0.091120   -1.416    0.15925
## c19ProSo02    0.219329    0.123089    1.782    0.07712 .
## c19ProSo03    0.293931    0.125731    2.338    0.02094 *
## c19ProSo04    0.213847    0.087454    2.445    0.01582 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.279 on 129 degrees of freedom
## (35 observations deleted due to missingness)
## Multiple R-squared:  0.585, Adjusted R-squared:  0.495
## F-statistic: 6.495 on 28 and 129 DF,  p-value: 5.065e-14
```

# c19ProSo02

```
summary(iran_lm_2)
```

```
##
## Call:
## lm(formula = c19ProSo02 ~ ., data = iran)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.5325 -0.4252  0.0139  0.3347  4.9936
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.2901908   0.9024814   -0.322  0.74832
## isoFriends_inPerson  0.0506096   0.0518442    0.976  0.33080
## isoOthPpl_inPerson  -0.0734752   0.0540377   -1.360  0.17630
## isoFriends_online  -0.0600332   0.0422227   -1.422  0.15749
## isoOthPpl_online  -0.0417446   0.0414793   -1.006  0.31611
## lone01          0.0750293   0.0966863    0.776  0.43917
## lone02         -0.0524491   0.1079415   -0.486  0.62786
## lone03         -0.0155143   0.1061250   -0.146  0.88400
## happy           0.0265838   0.0634116    0.419  0.67575
## lifeSat        -0.0207892   0.0948470   -0.219  0.82685
## MLQ            0.0001555   0.0807653    0.002  0.99847
## bor01          -0.0967371   0.0757172   -1.278  0.20368
## bor02           0.0926709   0.0799684    1.159  0.24866
## bor03          -0.0461478   0.0580860   -0.794  0.42838
## consp01        -0.1364780   0.0647079   -2.109  0.03687 *
## consp02         0.2008956   0.0758817    2.647  0.00912 **
## consp03        -0.0337677   0.0316825   -1.066  0.28850
## c19perBeh01     -0.0868101   0.0939314   -0.924  0.35711
## c19perBeh02      0.0561446   0.1460586    0.384  0.70132
## c19perBeh03      0.0267954   0.0824678    0.325  0.74577
## c19RCA01         0.0166789   0.0977038    0.171  0.86472
## c19RCA02         0.0867918   0.0844636    1.028  0.30608
## c19RCA03        -0.1410430   0.1045151   -1.349  0.17954
## gender         -0.0040061   0.1747389   -0.023  0.98174
## age            -0.0279067   0.0731477   -0.382  0.70345
## edu             0.0235847   0.0648553    0.364  0.71671
## c19ProSo01       0.1095245   0.0614657    1.782  0.07712 .
## c19ProSo03       0.7540213   0.0618153   12.198 < 2e-16 ***
## c19ProSo04       0.1489997   0.0618393    2.409  0.01739 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```



```
##
## Residual standard error: 0.9037 on 129 degrees of freedom
## (35 observations deleted due to missingness)
## Multiple R-squared: 0.8143, Adjusted R-squared: 0.7739
## F-statistic: 20.2 on 28 and 129 DF, p-value: < 2.2e-16

# c19ProSo03
summary(iran_lm_3)

##
## Call:
## lm(formula = c19ProSo03 ~ ., data = iran)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.5870 -0.3779  0.1175  0.3867  3.0163
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.125569   0.876240   0.143   0.8863
## isoFriends_inPerson -0.055208   0.050272  -1.098   0.2742
## isoOthPpl_inPerson  0.072229   0.052440   1.377   0.1708
## isoFriends_online   0.023618   0.041249   0.573   0.5679
## isoOthPpl_online    0.101045   0.039427   2.563   0.0115 *
## lone01             -0.135270   0.093307  -1.450   0.1496
## lone02              0.020875   0.104849   0.199   0.8425
## lone03              0.083592   0.102751   0.814   0.4174
## happy              -0.117435   0.060716  -1.934   0.0553 .
## lifeSat            0.085255   0.091770   0.929   0.3546
## MLQ                0.090597   0.077985   1.162   0.2475
## bor01              0.055338   0.073795   0.750   0.4547
## bor02             -0.111449   0.077402  -1.440   0.1523
## bor03              0.081732   0.056057   1.458   0.1473
## consp01            0.080052   0.063490   1.261   0.2096
## consp02           -0.083176   0.075271  -1.105   0.2712
## consp03           -0.019182   0.030840  -0.622   0.5351
## c19perBeh01        0.041058   0.091401   0.449   0.6540
## c19perBeh02       -0.029125   0.141824  -0.205   0.8376
## c19perBeh03       -0.053113   0.079940  -0.664   0.5076
## c19RCA01           0.054012   0.094724   0.570   0.5695
## c19RCA02          -0.102143   0.081823  -1.248   0.2142
## c19RCA03           0.199283   0.100639   1.980   0.0498 *
## gender            -0.027756   0.169586  -0.164   0.8702
## age               -0.045716   0.070924  -0.645   0.5203
## edu                0.002974   0.062981   0.047   0.9624
## c19ProSo01         0.138277   0.059149   2.338   0.0209 *
## c19ProSo02         0.710353   0.058235  12.198 <2e-16 ***
## c19ProSo04         0.040912   0.061252   0.668   0.5054
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.8772 on 129 degrees of freedom
## (35 observations deleted due to missingness)
## Multiple R-squared: 0.8236, Adjusted R-squared: 0.7854
## F-statistic: 21.52 on 28 and 129 DF, p-value: < 2.2e-16
```

```
# c19ProSo04
```

```
summary(iran_lm_4)
```

```
##
## Call:
## lm(formula = c19ProSo04 ~ ., data = iran)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.4950 -0.8429  0.0637  0.6127  4.1923
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.1460182   1.2534033   -0.914   0.3623
## isoFriends_inPerson  0.0721899   0.0721943    1.000   0.3192
## isoOthPpl_inPerson -0.0002026   0.0757999   -0.003   0.9979
## isoFriends_online   0.0295725   0.0592086    0.499   0.6183
## isoOthPpl_online   -0.0430766   0.0578735   -0.744   0.4580
## lone01            -0.0138495   0.1349708   -0.103   0.9184
## lone02             0.0536090   0.1504018    0.356   0.7221
## lone03            -0.1359245   0.1473354   -0.923   0.3580
## happy             -0.0548411   0.0882463   -0.621   0.5354
## lifeSat           0.1715235   0.1312593    1.307   0.1936
## MLQ               0.0826761   0.1122522    0.737   0.4628
## bor01            -0.0394522   0.1060653   -0.372   0.7105
## bor02             0.0042714   0.1119556    0.038   0.9696
## bor03             0.0034996   0.0810979    0.043   0.9656
## consp01           0.0844814   0.0913621    0.925   0.3569
## consp02          -0.0908287   0.1082244   -0.839   0.4029
## consp03           0.0241399   0.0442695    0.545   0.5865
## c19perBeh01        0.2055807   0.1300038    1.581   0.1162
## c19perBeh02        0.0644901   0.2034642    0.317   0.7518
## c19perBeh03        0.0916933   0.1146222    0.800   0.4252
## c19RCA01           0.1176021   0.1357004    0.867   0.3878
## c19RCA02          -0.0003563   0.1181193   -0.003   0.9976
## c19RCA03           0.0251999   0.1465732    0.172   0.8638
## gender           -0.1166084   0.2431560   -0.480   0.6324
## age              -0.0183108   0.1019230   -0.180   0.8577
## edu              0.1209002   0.0897461    1.347   0.1803
## c19ProSo01         0.2071479   0.0847140    2.445   0.0158 *
## c19ProSo02         0.2890333   0.1199575    2.409   0.0174 *
## c19ProSo03         0.0842398   0.1261222    0.668   0.5054
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.259 on 129 degrees of freedom
## (35 observations deleted due to missingness)
## Multiple R-squared:  0.5449, Adjusted R-squared:  0.4461
## F-statistic: 5.516 on 28 and 129 DF,  p-value: 7.958e-12
```

### Linear Regression Model Summary for Philippines

```
# c19ProSo01
```

```
summary(philippines_lm_1)
```

```
##
## Call:
## lm(formula = c19ProSo01 ~ ., data = philippines)
```

```
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.0508 -0.4552  0.0883  0.5464  3.2577
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.126312    0.311646  -0.405   0.6854
## isoFriends_inPerson  0.034287    0.013435   2.552   0.0109 *
## isoOthPpl_inPerson  0.014853    0.016149   0.920   0.3580
## isoFriends_online -0.018952    0.014983  -1.265   0.2062
## isoOthPpl_online  0.015074    0.013717   1.099   0.2721
## lone01          0.064874    0.040170   1.615   0.1067
## lone02          0.030553    0.040090   0.762   0.4462
## lone03         -0.093879    0.041910  -2.240   0.0253 *
## happy          -0.005624    0.022611  -0.249   0.8036
## lifeSat        -0.076935    0.039614  -1.942   0.0524 .
## MLQ            0.049915    0.028138   1.774   0.0764 .
## bor01          0.016562    0.024986   0.663   0.5076
## bor02          0.013738    0.024331   0.565   0.5725
## bor03          0.048516    0.023937   2.027   0.0430 *
## consp01        0.037644    0.016675   2.257   0.0242 *
## consp02       -0.010541    0.016834  -0.626   0.5314
## consp03        0.003004    0.013073   0.230   0.8183
## c19perBeh01    -0.010685    0.053131  -0.201   0.8407
## c19perBeh02     0.155613    0.061507   2.530   0.0116 *
## c19perBeh03    -0.040486    0.041809  -0.968   0.3331
## c19RCA01       0.025666    0.031779   0.808   0.4195
## c19RCA02      -0.014016    0.051728  -0.271   0.7865
## c19RCA03       0.058390    0.038264   1.526   0.1274
## gender        0.049688    0.063437   0.783   0.4337
## age          -0.056344    0.024735  -2.278   0.0230 *
## edu           0.035255    0.028811   1.224   0.2214
## c19ProSo02     0.340308    0.033075  10.289 < 2e-16 ***
## c19ProSo03     0.202886    0.028741   7.059 3.39e-12 ***
## c19ProSo04     0.143927    0.027191   5.293 1.52e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9448 on 886 degrees of freedom
## (35 observations deleted due to missingness)
## Multiple R-squared:  0.4566, Adjusted R-squared:  0.4394
## F-statistic: 26.58 on 28 and 886 DF,  p-value: < 2.2e-16
```

```
# c19ProSo02
```

```
summary(philippines_lm_2)
```

```
##
## Call:
## lm(formula = c19ProSo02 ~ ., data = philippines)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.9843 -0.4184  0.0260  0.5432  3.3934
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
```

```
## (Intercept) -0.2814730 0.2990572 -0.941 0.3469
## isoFriends_inPerson -0.0114959 0.0129393 -0.888 0.3745
## isoOthPpl_inPerson -0.0142390 0.0155030 -0.918 0.3586
## isoFriends_online 0.0304160 0.0143602 2.118 0.0344 *
## isoOthPpl_online -0.0050695 0.0131764 -0.385 0.7005
## lone01 0.0054583 0.0386191 0.141 0.8876
## lone02 0.0551865 0.0384540 1.435 0.1516
## lone03 -0.0824222 0.0402518 -2.048 0.0409 *
## happy 0.0335277 0.0216784 1.547 0.1223
## lifeSat 0.0930807 0.0379821 2.451 0.0145 *
## MLQ 0.0254954 0.0270464 0.943 0.3461
## bor01 -0.0391016 0.0239560 -1.632 0.1030
## bor02 0.0134780 0.0233577 0.577 0.5641
## bor03 -0.0014918 0.0230321 -0.065 0.9484
## consp01 -0.0198769 0.0160403 -1.239 0.2156
## consp02 0.0202453 0.0161500 1.254 0.2103
## consp03 -0.0003211 0.0125503 -0.026 0.9796
## c19perBeh01 0.0031294 0.0510063 0.061 0.9511
## c19perBeh02 0.0968775 0.0591697 1.637 0.1019
## c19perBeh03 0.0310043 0.0401440 0.772 0.4401
## c19RCA01 -0.0014693 0.0305187 -0.048 0.9616
## c19RCA02 0.0016777 0.0496603 0.034 0.9731
## c19RCA03 0.0468227 0.0367479 1.274 0.2029
## gender -0.1371032 0.0607455 -2.257 0.0243 *
## age -0.0219086 0.0238032 -0.920 0.3576
## edu 0.0477125 0.0276352 1.727 0.0846 .
## c19ProSo01 0.3136252 0.0304820 10.289 <2e-16 ***
## c19ProSo03 0.2857587 0.0266823 10.710 <2e-16 ***
## c19ProSo04 0.0311163 0.0264923 1.175 0.2405
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.9071 on 886 degrees of freedom
## (35 observations deleted due to missingness)
## Multiple R-squared: 0.4827, Adjusted R-squared: 0.4663
## F-statistic: 29.52 on 28 and 886 DF, p-value: < 2.2e-16
```

```
# c19ProSo03
```

```
summary(philippines_lm_3)
```

```
##
## Call:
## lm(formula = c19ProSo03 ~ ., data = philippines)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.3053 -0.5298  0.1922  0.6276  6.2053
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.280213   0.354359  -0.791  0.42930
## isoFriends_inPerson 0.020487   0.015321   1.337  0.18152
## isoOthPpl_inPerson -0.008171   0.018374  -0.445  0.65665
## isoFriends_online 0.032385   0.017021   1.903  0.05742 .
## isoOthPpl_online -0.008655   0.015609  -0.554  0.57938
## lone01 -0.027060   0.045745  -0.592  0.55431
## lone02 -0.065354   0.045558  -1.435  0.15178
```

```
## lone03          0.138314    0.047574    2.907    0.00374 **
## happy          -0.006101    0.025717   -0.237    0.81254
## lifeSat        -0.006020    0.045151   -0.133    0.89396
## MLQ            0.006841    0.032058    0.213    0.83106
## bor01          0.066559    0.028336    2.349    0.01905 *
## bor02         -0.020404    0.027670   -0.737    0.46105
## bor03         -0.017883    0.027281   -0.656    0.51231
## consp01        0.007578    0.019018    0.398    0.69038
## consp02       -0.035182    0.019114   -1.841    0.06601 .
## consp03        0.016770    0.014858    1.129    0.25933
## c19perBeh01    -0.059865    0.060396   -0.991    0.32185
## c19perBeh02    -0.012007    0.070206   -0.171    0.86424
## c19perBeh03     0.057929    0.047537    1.219    0.22332
## c19RCA01       -0.015740    0.036153   -0.435    0.66339
## c19RCA02       -0.021560    0.058830   -0.366    0.71410
## c19RCA03       0.020148    0.043572    0.462    0.64389
## gender         0.094890    0.072104    1.316    0.18851
## age           -0.035329    0.028189   -1.253    0.21044
## edu           -0.036940    0.032772   -1.127    0.25998
## c19ProSo01     0.262447    0.037179    7.059 3.39e-12 ***
## c19ProSo02     0.401098    0.037452   10.710 < 2e-16 ***
## c19ProSo04     0.234985    0.030403    7.729 2.93e-14 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.075 on 886 degrees of freedom
## (35 observations deleted due to missingness)
## Multiple R-squared:  0.4435, Adjusted R-squared:  0.4259
## F-statistic: 25.22 on 28 and 886 DF,  p-value: < 2.2e-16
```

```
# c19ProSo04
```

```
summary(philippines_lm_4)
```

```
##
## Call:
## lm(formula = c19ProSo04 ~ ., data = philippines)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.5315 -0.5840  0.1037  0.7338  3.5395
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.205584   0.379075  -0.542  0.58773
## isoFriends_inPerson -0.044794   0.016334  -2.742  0.00622 **
## isoOthPpl_inPerson  0.036451   0.019616   1.858  0.06346 .
## isoFriends_online   0.007661   0.018241   0.420  0.67460
## isoOthPpl_online   -0.012909   0.016692  -0.773  0.43952
## lone01          -0.015196   0.048934  -0.311  0.75622
## lone02           0.054726   0.048749   1.123  0.26191
## lone03          -0.003609   0.051125  -0.071  0.94374
## happy           0.026120   0.027493   0.950  0.34233
## lifeSat        -0.029247   0.048282  -0.606  0.54484
## MLQ            -0.009518   0.034287  -0.278  0.78140
## bor01           0.010481   0.030399   0.345  0.73034
## bor02          -0.014513   0.029599  -0.490  0.62402
## bor03           0.070441   0.029089   2.422  0.01565 *
```

```
## consp01      -0.013334    0.020338   -0.656    0.51224
## consp02      0.012062    0.020478    0.589    0.55599
## consp03     -0.011284    0.015898   -0.710    0.47803
## c19perBeh01    0.188895    0.064320    2.937    0.00340 **
## c19perBeh02   -0.173880    0.074862   -2.323    0.02042 *
## c19perBeh03    0.136399    0.050679    2.691    0.00725 **
## c19RCA01      0.045226    0.038642    1.170    0.24216
## c19RCA02      0.155264    0.062710    2.476    0.01348 *
## c19RCA03     -0.071239    0.046546   -1.530    0.12625
## gender        0.036470    0.077184    0.473    0.63668
## age           0.056477    0.030117    1.875    0.06109 .
## edu           0.041264    0.035049    1.177    0.23939
## c19ProSo01    0.212977    0.040236    5.293    1.52e-07 ***
## c19ProSo02    0.049962    0.042537    1.175    0.24049
## c19ProSo03    0.268807    0.034779    7.729    2.93e-14 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.149 on 886 degrees of freedom
## (35 observations deleted due to missingness)
## Multiple R-squared:  0.3143, Adjusted R-squared:  0.2927
## F-statistic: 14.51 on 28 and 886 DF,  p-value: < 2.2e-16
```

### Linear Regression Model Summary for Poland

# c19ProSo01

```
summary(poland_lm_1)
```

```
##
## Call:
## lm(formula = c19ProSo01 ~ ., data = poland)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.4518 -0.6726  0.1044  0.7612  2.7087
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.0619394   0.5294751    0.117   0.9069
## isoFriends_inPerson -0.0001390   0.0247658   -0.006   0.9955
## isoOthPpl_inPerson  0.0583424   0.0270508    2.157   0.0316 *
## isoFriends_online   0.0145846   0.0247714    0.589   0.5563
## isoOthPpl_online   -0.0000155   0.0224665   -0.001   0.9994
## lone01            -0.0148865   0.0696908   -0.214   0.8310
## lone02             0.0412552   0.0592717    0.696   0.4868
## lone03             0.0177875   0.0651293    0.273   0.7849
## happy              0.0236868   0.0460780    0.514   0.6075
## lifeSat            0.0402940   0.0726152    0.555   0.5793
## MLQ                0.0351039   0.0407955    0.860   0.3900
## bor01              0.0085054   0.0417875    0.204   0.8388
## bor02             -0.0485348   0.0432423   -1.122   0.2624
## bor03             -0.0826233   0.0401409   -2.058   0.0402 *
## consp01           -0.0452017   0.0298010   -1.517   0.1301
## consp02            0.0519395   0.0304983    1.703   0.0893 .
## consp03            0.0165675   0.0273899    0.605   0.5456
## c19perBeh01        0.0331601   0.0607165    0.546   0.5853
## c19perBeh02       -0.1458766   0.0644175   -2.265   0.0241 *
## c19perBeh03        0.0291977   0.0335592    0.870   0.3848
```

```
## c19RCA01      -0.0323752  0.0303495  -1.067   0.2867
## c19RCA02      0.0371159  0.0578588   0.641   0.5216
## c19RCA03     -0.0007031  0.0401517  -0.018   0.9860
## gender       -0.0573169  0.1469548  -0.390   0.6967
## age          0.0766131  0.0482090   1.589   0.1128
## edu         -0.0631875  0.0354486  -1.783   0.0754 .
## c19ProSo02    0.2136891  0.0358928   5.954 5.70e-09 ***
## c19ProSo03    0.2609982  0.0404852   6.447 3.26e-10 ***
## c19ProSo04    0.0918444  0.0442930   2.074   0.0388 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.14 on 404 degrees of freedom
## (25 observations deleted due to missingness)
## Multiple R-squared:  0.3718, Adjusted R-squared:  0.3282
## F-statistic: 8.538 on 28 and 404 DF,  p-value: < 2.2e-16
```

# c19ProSo02

summary(poland\_lm\_2)

```
##
## Call:
## lm(formula = c19ProSo02 ~ ., data = poland)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.2255 -1.0782  0.1504  1.0815  4.1301
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -1.717085   0.698506  -2.458 0.014381 *
## isoFriends_inPerson  0.064393   0.032759   1.966 0.050022 .
## isoOthPpl_inPerson -0.048163   0.036079  -1.335 0.182649
## isoFriends_online   0.018442   0.032924   0.560 0.575687
## isoOthPpl_online   -0.007113   0.029857  -0.238 0.811819
## lone01            -0.060677   0.092578  -0.655 0.512574
## lone02             0.024255   0.078813   0.308 0.758426
## lone03             0.026230   0.086558   0.303 0.762022
## happy             0.022666   0.061250   0.370 0.711534
## lifeSat          -0.029367   0.096535  -0.304 0.761122
## MLQ              -0.005506   0.054268  -0.101 0.919241
## bor01             0.035958   0.055512   0.648 0.517514
## bor02             0.040492   0.057525   0.704 0.481901
## bor03             0.047082   0.053577   0.879 0.380047
## consp01           0.011723   0.039715   0.295 0.768003
## consp02          -0.033859   0.040644  -0.833 0.405303
## consp03           0.024010   0.036399   0.660 0.509868
## c19perBeh01       -0.006199   0.080724  -0.077 0.938825
## c19perBeh02        0.219335   0.085462   2.566 0.010634 *
## c19perBeh03       -0.037983   0.044604  -0.852 0.394959
## c19RCA01          0.033790   0.040358   0.837 0.402942
## c19RCA02          0.067836   0.076862   0.883 0.377994
## c19RCA03         -0.043600   0.053320  -0.818 0.414002
## gender           -0.059854   0.195324  -0.306 0.759432
## age              0.074633   0.064165   1.163 0.245456
## edu              0.164147   0.046587   3.523 0.000475 ***
## c19ProSo01        0.377454   0.063400   5.954 5.7e-09 ***
```



```
## c19ProSo03      0.116811    0.056207    2.078 0.038318 *
## c19ProSo04      0.149511    0.058711    2.547 0.011249 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.516 on 404 degrees of freedom
## (25 observations deleted due to missingness)
## Multiple R-squared:  0.3217, Adjusted R-squared:  0.2747
## F-statistic: 6.844 on 28 and 404 DF,  p-value: < 2.2e-16
```

# c19ProSo03

```
summary(poland_lm_3)
```

```
##
## Call:
## lm(formula = c19ProSo03 ~ ., data = poland)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.4694 -0.6472  0.1660  0.7682  4.1094
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    0.3455072   0.6193502   0.558   0.5773
## isoFriends_inPerson  0.0020863   0.0289802   0.072   0.9426
## isoOthPpl_inPerson -0.0020719   0.0318357  -0.065   0.9481
## isoFriends_online  -0.0176009   0.0289861  -0.607   0.5440
## isoOthPpl_online   0.0050068   0.0262885   0.190   0.8490
## lone01            0.1008983   0.0814003   1.240   0.2159
## lone02           -0.0536531   0.0693485  -0.774   0.4396
## lone03           -0.0342548   0.0762006  -0.450   0.6533
## happy             0.0294171   0.0539171   0.546   0.5856
## lifeSat          -0.0821250   0.0849065  -0.967   0.3340
## MLQ              0.0904262   0.0475694   1.901   0.0580 .
## bor01           -0.0556721   0.0488226  -1.140   0.2548
## bor02            0.0461041   0.0506279   0.911   0.3630
## bor03           -0.0263572   0.0471993  -0.558   0.5769
## consp01          0.0000684   0.0349715   0.002   0.9984
## consp02         -0.0562930   0.0357065  -1.577   0.1157
## consp03         -0.0298711   0.0320310  -0.933   0.3516
## c19perBeh01       0.1222752   0.0708143   1.727   0.0850 .
## c19perBeh02      -0.0607079   0.0757964  -0.801   0.4236
## c19perBeh03      -0.0226724   0.0392907  -0.577   0.5642
## c19RCA01          0.0079215   0.0355620   0.223   0.8238
## c19RCA02         -0.0934903   0.0675794  -1.383   0.1673
## c19RCA03         -0.0310457   0.0469591  -0.661   0.5089
## gender           0.0996675   0.1719236   0.580   0.5624
## age             -0.0206052   0.0565797  -0.364   0.7159
## edu             0.0613631   0.0415318   1.477   0.1403
## c19ProSo01       0.3573870   0.0554367   6.447 3.26e-10 ***
## c19ProSo02       0.0905534   0.0435722   2.078   0.0383 *
## c19ProSo04       0.4643751   0.0467036   9.943 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.335 on 404 degrees of freedom
## (25 observations deleted due to missingness)
```



```
## Multiple R-squared:  0.4708, Adjusted R-squared:  0.4341
## F-statistic: 12.83 on 28 and 404 DF,  p-value: < 2.2e-16
```

```
# c19ProSo04
```

```
summary(poland_lm_4)
```

```
##
## Call:
## lm(formula = c19ProSo04 ~ ., data = poland)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.5891 -0.7387  0.1006  0.7533  3.9810
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.070887    0.591590  -0.120   0.90468
## isoFriends_inPerson -0.011728    0.027665  -0.424   0.67185
## isoOthPpl_inPerson  0.012094    0.030392   0.398   0.69088
## isoFriends_online   0.022076    0.027668   0.798   0.42539
## isoOthPpl_online    0.012044    0.025095   0.480   0.63153
## lone01            -0.043093    0.077841  -0.554   0.58016
## lone02            -0.092011    0.066107  -1.392   0.16473
## lone03             0.094330    0.072625   1.299   0.19473
## happy             -0.053850    0.051431  -1.047   0.29571
## lifeSat           0.089626    0.081042   1.106   0.26942
## MLQ               -0.006669    0.045622  -0.146   0.88386
## bor01             -0.017158    0.046684  -0.368   0.71341
## bor02              0.022522    0.048377   0.466   0.64178
## bor03             -0.006898    0.045083  -0.153   0.87847
## consp01            0.016857    0.033381   0.505   0.61386
## consp02           -0.002092    0.034198  -0.061   0.95125
## consp03           -0.011468    0.030612  -0.375   0.70812
## c19perBeh01       -0.003251    0.067864  -0.048   0.96182
## c19perBeh02        0.108276    0.072229   1.499   0.13464
## c19perBeh03        0.094329    0.037237   2.533   0.01168 *
## c19RCA01           0.039733    0.033900   1.172   0.24186
## c19RCA02           0.176294    0.064082   2.751   0.00621 **
## c19RCA03          -0.045351    0.044805  -1.012   0.31206
## gender            0.051491    0.164206   0.314   0.75400
## age              -0.061078    0.053947  -1.132   0.25823
## edu              -0.015232    0.039755  -0.383   0.70182
## c19ProSo01        0.114658    0.055295   2.074   0.03875 *
## c19ProSo02         0.105667    0.041494   2.547   0.01125 *
## c19ProSo03         0.423368    0.042579   9.943   < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.274 on 404 degrees of freedom
## (25 observations deleted due to missingness)
## Multiple R-squared:  0.4453, Adjusted R-squared:  0.4069
## F-statistic: 11.58 on 28 and 404 DF,  p-value: < 2.2e-16
```

### Linear Regression Model Summary for Saudi Arabia

```
# c19ProSo01
```

```
summary(saudi_arabia_lm_1)
```

```
##
## Call:
## lm(formula = c19ProSo01 ~ ., data = saudi_arabia)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -5.8594 -0.5807  0.1711  0.6790  3.9568
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)   -0.4956397   0.3328395  -1.489   0.1369
## isoFriends_inPerson -0.0258141   0.0193290  -1.336   0.1821
## isoOthPpl_inPerson  0.0387136   0.0238568   1.623   0.1050
## isoFriends_online   0.0162002   0.0236766   0.684   0.4940
## isoOthPpl_online    0.0287673   0.0209157   1.375   0.1694
## lone01             0.0923320   0.0502466   1.838   0.0665 .
## lone02            -0.0841890   0.0499563  -1.685   0.0923 .
## lone03             0.0493910   0.0467846   1.056   0.2914
## happy             -0.0371072   0.0265880  -1.396   0.1632
## lifeSat            0.0464417   0.0528960   0.878   0.3802
## MLQ                0.0002888   0.0375085   0.008   0.9939
## bor01             -0.0171767   0.0309849  -0.554   0.5795
## bor02             -0.0272151   0.0244200  -1.114   0.2654
## bor03             0.0552106   0.0292643   1.887   0.0596 .
## consp01           0.0327970   0.0226459   1.448   0.1479
## consp02          -0.0237085   0.0225194  -1.053   0.2928
## consp03           0.0017772   0.0176422   0.101   0.9198
## c19perBeh01        0.0690105   0.0483269   1.428   0.1537
## c19perBeh02        0.1328871   0.0624534   2.128   0.0337 *
## c19perBeh03        0.0182985   0.0449993   0.407   0.6844
## c19RCA01           0.0341827   0.0356236   0.960   0.3376
## c19RCA02           0.0352244   0.0527703   0.668   0.5046
## c19RCA03          -0.0486496   0.0507745  -0.958   0.3383
## gender            0.1217558   0.0930248   1.309   0.1910
## age              -0.0516122   0.0388731  -1.328   0.1847
## edu              -0.0186443   0.0319841  -0.583   0.5601
## c19ProSo02         0.3813108   0.0429442   8.879 < 2e-16 ***
## c19ProSo03         0.0736609   0.0391743   1.880   0.0604 .
## c19ProSo04         0.2224086   0.0362605   6.134 1.36e-09 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.234 on 781 degrees of freedom
## (67 observations deleted due to missingness)
## Multiple R-squared:  0.444, Adjusted R-squared:  0.4241
## F-statistic: 22.27 on 28 and 781 DF,  p-value: < 2.2e-16
```

# c19ProSo02

```
summary(saudi_arabia_lm_2)
```

```
##
## Call:
## lm(formula = c19ProSo02 ~ ., data = saudi_arabia)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -3.5220 -0.5447  0.0272  0.5405  2.7697
```

```
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.173157   0.264617  -0.654  0.51307
## isoFriends_inPerson  0.035419   0.015315   2.313  0.02099 *
## isoOthPpl_inPerson -0.055660   0.018872  -2.949  0.00328 **
## isoFriends_online    0.027872   0.018781   1.484  0.13820
## isoOthPpl_online    -0.009744   0.016626  -0.586  0.55801
## lone01             -0.039931   0.039962  -0.999  0.31801
## lone02             -0.007143   0.039743  -0.180  0.85741
## lone03             -0.036823   0.037156  -0.991  0.32197
## happy              -0.008859   0.021138  -0.419  0.67525
## lifeSat             0.097886   0.041880   2.337  0.01968 *
## MLQ                 -0.015454   0.029781  -0.519  0.60396
## bor01               0.049337   0.024547   2.010  0.04479 *
## bor02              -0.011138   0.019404  -0.574  0.56613
## bor03               0.021560   0.023279   0.926  0.35466
## consp01             0.007114   0.018006   0.395  0.69289
## consp02             0.019131   0.017883   1.070  0.28505
## consp03             0.008731   0.014007   0.623  0.53324
## c19perBeh01         0.051954   0.038382   1.354  0.17626
## c19perBeh02        -0.028415   0.049729  -0.571  0.56789
## c19perBeh03        -0.003205   0.035738  -0.090  0.92856
## c19RCA01            0.026511   0.028290   0.937  0.34899
## c19RCA02           -0.035342   0.041899  -0.844  0.39921
## c19RCA03            0.100498   0.040184   2.501  0.01259 *
## gender             -0.025357   0.073948  -0.343  0.73176
## age                 0.038836   0.030874   1.258  0.20880
## edu                 0.016214   0.025398   0.638  0.52341
## c19ProSo01          0.240465   0.027082   8.879 < 2e-16 ***
## c19ProSo03          0.291636   0.029381   9.926 < 2e-16 ***
## c19ProSo04          0.127409   0.029126   4.374 1.38e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.98 on 781 degrees of freedom
## (67 observations deleted due to missingness)
## Multiple R-squared:  0.5567, Adjusted R-squared:  0.5408
## F-statistic: 35.03 on 28 and 781 DF,  p-value: < 2.2e-16
```

```
# c19ProSo03
```

```
summary(saudi_arabia_lm_3)
```

```
##
## Call:
## lm(formula = c19ProSo03 ~ ., data = saudi_arabia)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.6445 -0.5925  0.2198  0.6174  3.4117
##
## Coefficients:
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept) -0.36108    0.30349  -1.190   0.2345
## isoFriends_inPerson -0.01169    0.01763  -0.663   0.5074
## isoOthPpl_inPerson  0.01521    0.02177   0.699   0.4850
## isoFriends_online  -0.02764    0.02156  -1.282   0.2002
```

```
## isoOthPpl_online      0.03527      0.01904      1.852      0.0644 .
## lone01                0.03690      0.04587      0.804      0.4214
## lone02               -0.04275      0.04559     -0.938      0.3487
## lone03                0.02802      0.04266      0.657      0.5114
## happy                 -0.02051      0.02425     -0.846      0.3979
## lifeSat               0.00628      0.04823      0.130      0.8964
## MLQ                   0.04145      0.03415      1.214      0.2253
## bor01                 -0.03462      0.02822     -1.227      0.2203
## bor02                 0.03522      0.02224      1.584      0.1137
## bor03                 0.04680      0.02668      1.754      0.0798 .
## consp01               0.01483      0.02066      0.718      0.4729
## consp02              -0.01987      0.02053     -0.968      0.3334
## consp03              -0.01234      0.01607     -0.768      0.4427
## c19perBeh01           0.02130      0.04409      0.483      0.6293
## c19perBeh02           0.01427      0.05708      0.250      0.8027
## c19perBeh03           0.09345      0.04088      2.286      0.0225 *
## c19RCA01              0.01542      0.03248      0.475      0.6352
## c19RCA02              0.07976      0.04802      1.661      0.0972 .
## c19RCA03             -0.03867      0.04628     -0.836      0.4037
## gender                0.07530      0.08483      0.888      0.3750
## age                   0.01236      0.03547      0.348      0.7276
## edu                   -0.02636      0.02914     -0.905      0.3659
## c19ProSo01            0.06118      0.03254      1.880      0.0604 .
## c19ProSo02            0.38411      0.03870      9.926      <2e-16 ***
## c19ProSo04            0.36006      0.03128     11.510      <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.125 on 781 degrees of freedom
## (67 observations deleted due to missingness)
## Multiple R-squared:  0.5375, Adjusted R-squared:  0.5209
## F-statistic: 32.42 on 28 and 781 DF,  p-value: < 2.2e-16
```

```
# c19ProSo04
```

```
summary(saudi_arabia_lm_4)
```

```
##
## Call:
## lm(formula = c19ProSo04 ~ ., data = saudi_arabia)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -4.4795 -0.6778  0.0889  0.6539  4.7978
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)  -0.1170321   0.3212464  -0.364   0.7157
## isoFriends_inPerson -0.0142923   0.0186452  -0.767   0.4436
## isoOthPpl_inPerson  0.0499473   0.0229645   2.175   0.0299 *
## isoFriends_online  -0.0102641   0.0228254  -0.450   0.6531
## isoOthPpl_online  -0.0028485   0.0201844  -0.141   0.8878
## lone01         -0.0550581   0.0484964  -1.135   0.2566
## lone02          0.0770319   0.0481607   1.599   0.1101
## lone03         -0.0017794   0.0451271  -0.039   0.9686
## happy           0.0600460   0.0255696   2.348   0.0191 *
## lifeSat        -0.0955239   0.0508961  -1.877   0.0609 .
## MLQ             0.0459903   0.0361164   1.273   0.2033
```

```

## bor01          -0.0120558  0.0298686  -0.404   0.6866
## bor02          -0.0022310  0.0235566  -0.095   0.9246
## bor03          -0.0602672  0.0281892  -2.138   0.0328 *
## consp01        -0.0088067  0.0218550  -0.403   0.6871
## consp02        -0.0103927  0.0217183  -0.479   0.6324
## consp03         0.0030332  0.0170048   0.178   0.8585
## c19perBeh01     -0.0197193  0.0466369  -0.423   0.6725
## c19perBeh02     -0.0489461  0.0603466  -0.811   0.4176
## c19perBeh03     -0.0354441  0.0433601  -0.817   0.4139
## c19RCA01        0.0216052  0.0343486   0.629   0.5295
## c19RCA02        0.0209498  0.0508735   0.412   0.6806
## c19RCA03        0.1100056  0.0488110   2.254   0.0245 *
## gender          0.0284476  0.0897576   0.317   0.7514
## age            -0.0003275  0.0375114  -0.009   0.9930
## edu             0.0486840  0.0307864   1.581   0.1142
## c19ProSo01      0.2066336  0.0336886   6.134 1.36e-09 ***
## c19ProSo02      0.1877049  0.0429098   4.374 1.38e-05 ***
## c19ProSo03      0.4027592  0.0349933  11.510 < 2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.19 on 781 degrees of freedom
## (67 observations deleted due to missingness)
## Multiple R-squared:  0.4809, Adjusted R-squared:  0.4623
## F-statistic: 25.84 on 28 and 781 DF,  p-value: < 2.2e-16

```