# Assignment 2 Group 16(Xinyu Lin and Yujing Jiang)

Due at 11:59pm on October 1.

Github Link:https://github.com/Colin0817/assignment2\_group16.git

You may work in pairs or individually for this assignment. Make sure you join a group in Canvas if you are working in pairs. Turn in this assignment as an HTML or PDF file to ELMS. Make sure to include the R Markdown or Quarto file that was used to generate it.

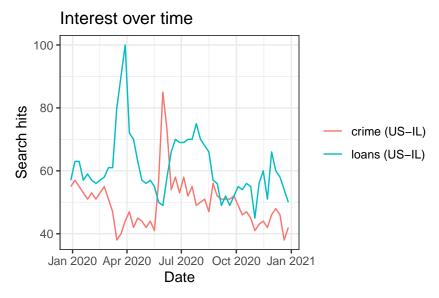
```
library(tidyverse)
library(gtrendsR)
library(censusapi)
```

In this assignment, you will pull from APIs to get data from various data sources and use your data wrangling skills to use them all together. You should turn in a report in PDF or HTML format that addresses all of the questions in this assignment, and describes the data that you pulled and analyzed. You do not need to include full introduction and conclusion sections like a full report, but you should make sure to answer the questions in paragraph form, and include all relevant tables and graphics.

Whenever possible, use piping and dplyr. Avoid hard-coding any numbers within the report as much as possible.

# **Pulling from APIs**

Our first data source is the Google Trends API. Suppose we are interested in the search trends for crime and loans in Illinois in the year 2020. We could find this using the following code:



Answer the following questions for the keywords "crime" and "loans".

• Find the mean, median and variance of the search hits for the keywords.

# Answer:

The mean of the crime is 51.35849 the median is 51, and the variance is 87.00363. The mean of the loans is 62.32075, the median is 59, and the variance is 98.18360.

• Which cities (locations) have the highest search frequency for loans? Note that there might be multiple rows for each city if there were hits for both "crime" and "loans" in that city. It might be easier to answer this question if we had the search hits info for both search terms in two separate variables. That is, each row would represent a unique city.

```
res_location <- res$interest_by_city
head(res_location)</pre>
```

```
location hits keyword
                                   geo gprop
1
         Hampshire
                    100
                           crime US-IL
                                          web
2
              Anna
                      84
                           crime US-IL
                                          web
3 East Saint Louis
                      81
                           crime US-IL
                                          web
4
            Macomb
                     76
                           crime US-IL
                                          web
5
  North Riverside
                     74
                           crime US-IL
                                          web
                      72
6
             Byron
                           crime US-IL
                                          web
```

```
# A tibble: 353 x 5
   location
                  geo
                        gprop crime loans
   <chr>
                  <chr> <chr> <int> <int>
1 Evergreen Park US-IL web
                                  NA
                                       100
2 Long Lake
                  US-IL web
                                  NA
                                        96
3 Rosemont
                  US-IL web
                                  NA
                                        84
4 Peotone
                  US-IL web
                                        77
5 Coal City
                  US-IL web
                                  NA
                                        75
6 Ford Heights
                  US-IL web
                                        74
                                  NA
7 Dolton
                  US-IL web
                                  NA
                                        74
8 Riverdale
                  US-IL web
                                  55
                                        71
9 Oak Lawn
                  US-IL web
                                  NA
                                        70
10 Hazel Crest
                  US-IL web
                                        69
                                  NA
# i 343 more rows
```

From the results, Long Lake have the highest search frequency for loans, which is 100.

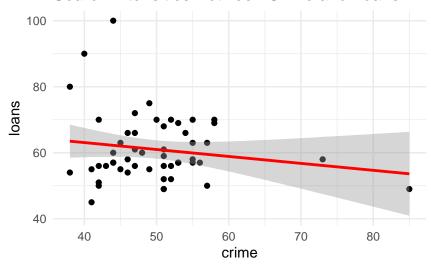
• Is there a relationship between the search intensities between the two keywords we used?

```
res_time1 <- res_time %>%
  select(date, keyword, hits) %>%
  pivot_wider(names_from = keyword, values_from = hits)
cor_time <- cor(res_time1$crime, res_time1$loans)
cor_time</pre>
```

# [1] -0.1652924

```
library(ggplot2)
  ggplot(res_time1, aes(x = crime, y = loans)) +
    geom_point() +
    geom_smooth(method = "lm", color = "red") +
    labs(title = "Search Intensities Between Crime and Loans",
        x = "crime",
        y = "loans") +
    theme_minimal()
```

# Search Intensities Between Crime and Loans



```
model <- lm(loans ~ crime, data = res_time1)
summary(model)</pre>
```

```
Call:
lm(formula = loans ~ crime, data = res_time1)
```

#### Residuals:

```
Min 1Q Median 3Q Max -17.871 -6.451 -2.718 5.862 37.759
```

#### Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 71.4904 8.8589 8.070 1.13e-10 \*\*\*
crime -0.2102 0.1756 -1.197 0.237
---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

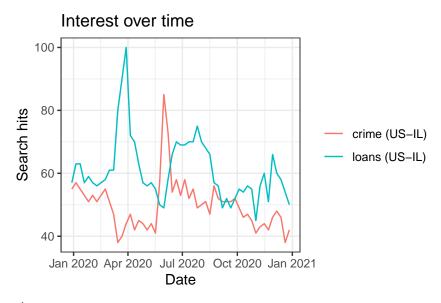
Residual standard error: 10.13 on 51 degrees of freedom Multiple R-squared: 0.02732, Adjusted R-squared: 0.008

F-statistic: 1.433 on 1 and 51 DF, p-value: 0.2369

```
cor(res_time1$crime, res_time1$loans, method = "spearman")
```

# [1] 0.05730665

# plot(res)

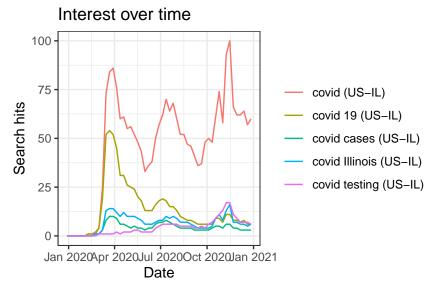


# Answer:

The correlation coefficient is -0.1700124, indicating that there is an almost negligible relationship between the search frequency of these two keywords. Furthermore, the occasional negative correlation is very weak and insufficient to indicate a significant inverse relationship.

The Spearman's correlation coefficient is 0.04702525, indicating that the correlation is not statistically significant when the normal distribution is not taken into account. From the graph "Search Intensities Between Crime and Loans", it is challenging to discern the evident linear correlation between the two variables. However, from the second one, the graph reveals a more pronounced inverse trend between crime and loans in general between March 2020 and October 2020. This indicates that as the search frequency for crime increases, the search frequency for loans decreases. However, before March and after October, the search rates for crime and loans exhibit a similar trend. So the relationship between search rates for loans and search rates for crime needs to be explored further.

Repeat the above for keywords related to covid. Make sure you use multiple keywords like we did above. Try several different combinations and think carefully about words that might make sense within this context.



#### Answer: Why we choose these keywords?

We use "covid testing", "covid", "covid 19", "covid cases", "covid Illinois" as keywords. The reasons for choosing these five keywords are as follows:

covid testing: This keyword reflects people's interest in COVID-19 testing, especially during the pandemic when the availability of testing is directly related to infection rates. It shows the public's demand and attitude toward testing.

covid: As the general name for the coronavirus pandemic, this keyword covers a wide range of information related to the virus. Using this keyword helps capture the overall attention and trends concerning COVID-19.

covid 19: This is the official name of the virus, commonly used in scientific and medical literature. Choosing this keyword helps gather more professional discussions and information.

covid cases: This keyword focuses on the search for confirmed cases, showing the public's concern about the spread of the virus. Analyzing data from this keyword can reveal how people perceive the seriousness of the pandemic.

covid Illinois: This keyword is specific to COVID-19 information in Illinois, helping to analyze the pandemic situation and public reaction in that region. It provides a more localized perspective, suitable for studying specific regional policies and measures.

• Find the mean, median and variance of the search hits for the keywords.

### # A tibble: 5 x 4

	keywor	'd	mean_hits1	median_hits1	var_hits1
	<chr></chr>		<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
1	covid		48.6	52	645.
2	covid	19	14.1	9	177.
3	covid	Illinois	6.88	7	17.1
4	covid	cases	4.16	4	6.45
5	covid	testing	4.49	4	16.8

#### Answer:

covid: The mean number of hits is 48.56, the median number of hits is 52, and the variance of the number of hits is 645.29.

covid 19: The average number of hits was 14.05, the median number of hits was 9, and the variance of the number of hits was 177.30.

covid Illinois: the mean number of hits is 6.88, the median number of hits is 7, and the variance of the number of hits is 17.07.

covid cases: the average number of hits was 4.16, the median number of hits was 4, and the variance of the number of hits was 6.45.

covid testing: mean number of hits is 4.49, median number of hits is 4, variance of hits is 16.77.

• Which cities (locations) have the highest search frequency for covid?

# # A tibble: 551 x 8

	location	geo	gprop	`covid	testing`	covid	`covid 19`	`covid cases`
	<chr></chr>	<chr>&gt;</chr>	<chr></chr>		<dbl></dbl>	<dbl></dbl>	<dbl></dbl>	<dbl></dbl>
1	Geneva	US-IL	web		NA	100	NA	NA
2	Oak Lawn	US-IL	web		NA	100	NA	NA
3	Bartelso	US-IL	web		NA	99	NA	NA
4	Northbrook	US-IL	web		NA	99	NA	NA
5	Evergreen Park	US-IL	web		100	97	NA	NA
6	Beckemeyer	US-IL	web		NA	96	NA	NA
7	Elizabeth	US-IL	web		NA	95	NA	NA
8	Winnetka	US-IL	web		71	94	NA	NA
9	Albany	US-IL	web		NA	94	NA	NA
10	Elkhart	US-IL	web		NA	93	NA	NA

<sup>#</sup> i 541 more rows

#### Answer:

From the results, Geneva and Oak Lawn have the highest search frequency for covid, which is 100.

• Is there a relationship between the search intensities between the five keywords we used?

<sup>#</sup> i 1 more variable: `covid Illinois` <dbl>

```
covid
                           covid.19 covid.testing covid.cases covid.Illinois
covid
               1.0000000 0.5651351
                                        0.6064500
                                                     0.8366279
                                                                    0.9100350
covid 19
               0.5651351 1.0000000
                                       -0.2470571
                                                     0.8268486
                                                                    0.7471191
covid testing 0.6064500 -0.2470571
                                        1.0000000
                                                    0.1977356
                                                                    0.3861644
               0.8366279 0.8268486
                                        0.1977356
                                                    1.0000000
                                                                    0.8875386
covid cases
covid Illinois 0.9100350 0.7471191
                                        0.3861644
                                                     0.8875386
                                                                    1.0000000
```

A correlation coefficient of 0.8366279 was observed between the keyword "covid" and both "covid cases". The resulting correlation coefficient of "covid Illinois" and "covid" is 0.9100350. They are indicating a strong positive correlation. The correlation coefficients with "covid-19" vs "covid" and "covid testing" vs "covid" are comparatively weaker. The keyword "covid 19" exhibits a relatively robust positive correlation with "covid cases" and "covid Illinois," yet displays a comparatively weaker correlation with "covid testing." The keyword "covid testing" demonstrates a relatively weak correlation with all other keywords. The correlation coefficient between the keywords "covid cases" and "covid Illinois" is 0.8875386, indicating a strong positive correlation.

# Google Trends + ACS

Now lets add another data set. The censusapi package provides a nice R interface for communicating with this API. However, before running queries we need an access key. This (easy) process can be completed here:

```
https://api.census.gov/data/key_signup.html
```

Once you have an access key, save it as a text file, then read this key in the cs\_key object. We will use this object in all following API queries. Note that I called my text file census-key.txt – yours might be different!

```
cs_key <- read_file("census-key.txt")</pre>
```

In the following, we request basic socio-demographic information (population, median age, median household income, income per capita) for cities and villages in the state of Illinois. Documentation for the 5-year ACS API can be found here: https://www.census.gov/data/developers/data-sets/acs-5year.html. The information about the variables used here can be found here: https://api.census.gov/data/2022/acs/acs5/variables.html.

	state place		NAME	B01001_001E	B06002_001E	B19013_001E
1	17 15261	Coatsburg village,	Illinois	180	35.6	55714
2	17 15300	Cobden village,	Illinois	1018	44.2	38750
3	17 15352	Coffeen city,	Illinois	640	33.4	35781
4	17 15378	Colchester city,	Illinois	1347	42.2	43942
5	17 15469	Coleta village,	Illinois	230	27.7	56875
6	17 15495	Colfax village,	Illinois	1088	32.5	58889
	B19301_001E					
1	27821					
2	19979					
3	26697					
4	24095					
5	23749					
6	24861					

Convert values that represent missings to NAs.

```
acs_il[acs_il == -666666666] <- NA
```

Now, it might be useful to rename the socio-demographic variables (B01001\_001E etc.) in our data set and assign more meaningful names.

```
acs_il <-
  acs_il %>%
  rename(pop = B01001_001E,
      age = B06002_001E,
      hh_income = B19013_001E,
      income = B19301_001E)
```

It seems like we could try to use this location information listed above to merge this data set with the Google Trends data. However, we first have to clean NAME so that it has the same structure as location in the search interest by city data. Add a new variable location to the ACS data that only includes city names.

```
acs_clean <- acs_il %>%
  mutate(location = sub(",.*", "", NAME))
acs_clean <- acs_clean %>%
  mutate(location = str_replace(location, " village", "")) %>%
  mutate(location = str_replace(location, " city", "")) %>%
  mutate(location = str_replace(location, " CDP", "")) %>%
  mutate(location = str_replace(location, " town", "")) %>%
  mutate(location = trimws(location))
head(acs_clean)
```

```
state place
                                     NAME
                                          pop age hh_income income
                                                                       location
1
     17 15261 Coatsburg village, Illinois
                                          180 35.6
                                                        55714
                                                              27821 Coatsburg
2
    17 15300
                 Cobden village, Illinois 1018 44.2
                                                        38750 19979
                                                                         Cobden
3
    17 15352
                   Coffeen city, Illinois 640 33.4
                                                        35781 26697
                                                                        Coffeen
4
    17 15378
               Colchester city, Illinois 1347 42.2
                                                        43942 24095 Colchester
                Coleta village, Illinois 230 27.7
5
    17 15469
                                                        56875 23749
                                                                         Coleta
                Colfax village, Illinois 1088 32.5
     17 15495
                                                        58889 24861
                                                                         Colfax
```

```
# library(stringr)
# acs_clean <- acs_il %>%
# mutate(location = str_extract(NAME, "^[^,]+"),
# location = str_replace(location, " city|village|town|CDP", ""),
# location = str_trim(location, side="right"))
#head(acs_clean)
```

Answer the following questions with the "crime" and "loans" Google trends data and the ACS data.

• First, check how many cities don't appear in both data sets, i.e. cannot be matched. Then, create a new data set by joining the Google Trends and the ACS data. Keep only cities that appear in both data sets.

```
location_google <- res_location_1$location
location_acs <- acs_clean$location
google_in <- setdiff(location_google, location_acs)
acs_in <- setdiff(location_acs, location_google)
ummatched <- length(google_in) + length(acs_in)
ummatched</pre>
```

[1] 1129

```
merge_ <- inner_join(res_location_1, acs_clean, by = "location")</pre>
```

#### Answer:

There are 1132 cities does not appear in both data sets.

• Compute the mean of the search popularity for both keywords for cities that have an above average median household income and for those that have an below average median household income. When building your pipe, start with creating the grouping variable and then proceed with the remaining tasks. What conclusions might you draw from this?

```
average_income <- mean(merge_$hh_income, na.rm = TRUE)
merge_fil <- merge_ %>%
  mutate(group = ifelse(hh_income > average_income, "above_ave", "below_ave"))
mean_keyword <- merge_fil%>%
  group_by(group)%>%
  summarise(mean(crime, na.rm = TRUE), mean(loans, na.rm = TRUE))
mean_keyword
```

The lower popularity of searches related to "crime" in cities with higher incomes may be attributed to the lower incidence of criminal activity in these cities. The greater prevalence of searches for "loans" may indicate that residents of these cities are more inclined to view loans as a means of investment and are able to improve their quality of life through borrowing.

The higher prevalence of searches for "crime" in cities with lower household incomes may be indicative of a greater focus on crime in these cities. This could reflect the fact that areas with poor socioeconomic conditions may have higher crime rates or residents may be more sensitive to crime perceptions. The search prevalence for "loans" is lower than, but not significantly different from, that observed in higher income cities. The higher search prevalence for loans could imply that residents in these cities are more reliant on loans to make ends meet, reflecting greater economic stress.

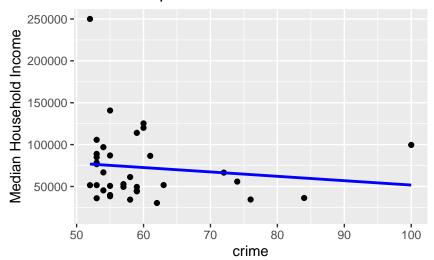
• Is there a relationship between the median household income and the search popularity of the Google trends terms? Describe the relationship and use a scatterplot with qplot().

```
cor_crime <- cor(merge_fil$hh_income, merge_fil$crime, use = "complete.obs")
cor_loans <- cor(merge_fil$hh_income, merge_fil$loans, use = "complete.obs")
cor_crime; cor_loans</pre>
```

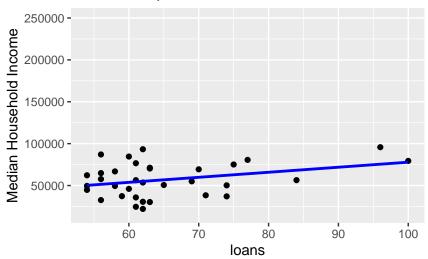
[1] -0.1246437

[1] 0.3299387

# Relationship Between Median Household Incon



# Relationship Between Median Household Incon



#### Answer:

The correlation coefficient between family income and crime search degree is -0.213, indicating that there is a low correlation between the two variables and that the relationship is negative. The correlation coefficient between household income and loan search degree is 0.152, indicating

that the correlation between household income and crime search degree is low and positive. An examination of the scatterplot of CRIME and HOUSEHOLD INCOME reveals that the data points are concentrated in the lower left of the plot, indicating that household income is higher in areas with lower crime search degrees. In general, there seems to be a negative correlation between lower crime search rates and higher household incomes. The concentration of points in the scatterplot LOANS and HOUSEHOLD INCOME suggests that as household income increases, loan search rates increase, and that income growth levels off or increases at a slower rate after a certain point is reached.

Repeat the above steps using the covid data and the ACS data.

• First, check how many cities don't appear in both data sets, i.e. cannot be matched. Then, create a new data set by joining the Google Trends and the ACS data. Keep only cities that appear in both data sets.

```
loc_google <- res_covid_location_1$location
loc_acs <- acs_clean$location
google_in1 <- setdiff(loc_google, loc_acs)
acs_in1 <- setdiff(loc_acs, loc_google)
ummatched <- length(google_in1) + length(acs_in1)
ummatched</pre>
```

[1] 955

• Compute the mean of the search popularity for keywords for cities that have an above average median household income and for those that have an below average median household income. When building your pipe, start with creating the grouping variable and then proceed with the remaining tasks. What conclusions might you draw from this?

```
mena_testing = mean(`covid testing`, na.rm = TRUE),
mean_covid19 = mean(`covid 19`, na.rm = TRUE),
mean_cases = mean(`covid cases`, na.rm = TRUE),
mean_Illinois = mean(`covid Illinois`, na.rm = TRUE))
merge_2
```

```
# A tibble: 3 x 6
 group1
                mean_covid mena_testing mean_covid19 mean_cases mean_Illinois
  <chr>
                      <dbl>
                                    <dbl>
                                                  <dbl>
                                                             <dbl>
                                                                            <dbl>
                                     72.4
                                                   70
                                                              79
1 above average
                       81.3
                                                                             91.9
2 below_average
                       81.1
                                     73.4
                                                  75.2
                                                              84.6
                                                                             91.2
3 <NA>
                      NaN
                                    NaN
                                                 NaN
                                                             NaN
                                                                            NaN
```

The search rate for the "covid" keyword is slightly higher for high-income households (81.51) than for low-income households (81.19), but the difference is not significant.

The search rate is slightly higher for low-income households (74.00) than for high-income households (72.47), suggesting that low-income households may be searching for information related to COVID testing more frequently.

The search rate for low-income households (76.83) is significantly higher than that for high-income households (69.78), indicating that low-income households are investing more search activity in COVID-19-related information.

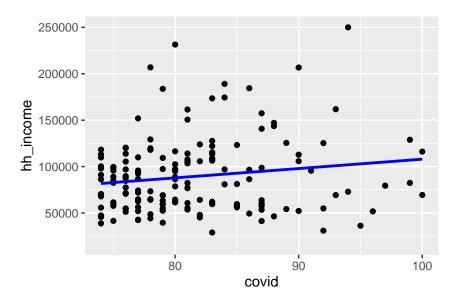
The search rate for low-income households (84.60) is significantly higher than that for high-income households (79.0), which may indicate that low-income households are more likely to pay attention to information about covid cases.

The search rate in covid Illinois is slightly higher for higher-income households (91.86) than for lower-income households (91.17), a relatively small difference.

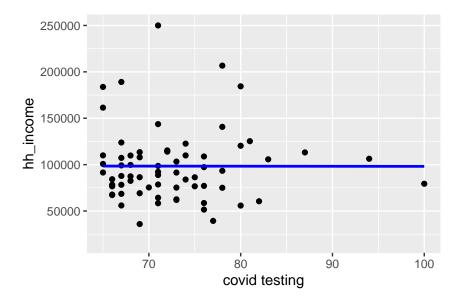
• Is there a relationship between the median household income and the search popularity of the Google trends terms? Describe the relationship and use a scatterplot with qplot().

```
values cor_hhincome
1 covid 0.157081057
2 covid testing -0.001516017
3 covid 19 -0.316655111
4 covid cases -0.374203960
5 covid Illinois -0.141653876
```

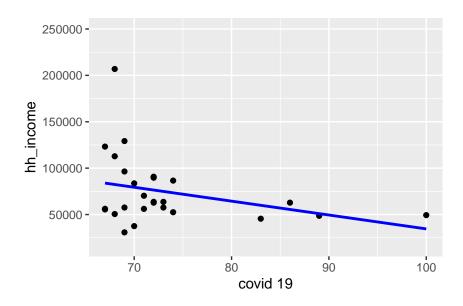
```
qplot(x=covid,y=hh_income,data = merge_1,geom = "point")+
geom_smooth(method = "lm", se = FALSE, color = "blue")
```



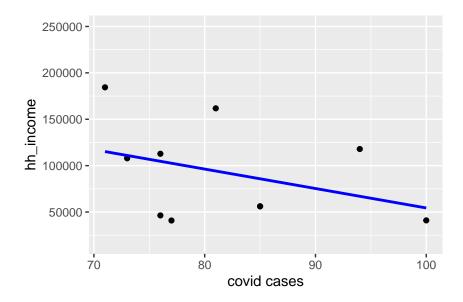
```
qplot(x=`covid testing`,y=hh_income,data = merge_1,geom = "point")+
  geom_smooth(method = "lm", se = FALSE, color = "blue")
```



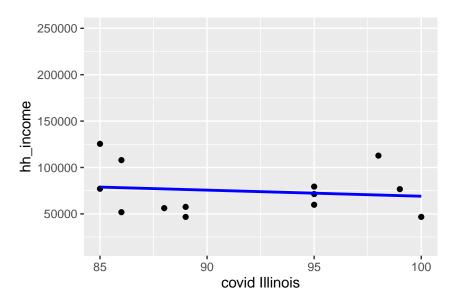
```
qplot(x=`covid 19`,y=hh_income,data = merge_1,geom = "point")+
  geom_smooth(method = "lm", se = FALSE, color = "blue")
```



```
qplot(x=`covid cases`,y=hh_income,data = merge_1,geom = "point")+
  geom_smooth(method = "lm", se = FALSE, color = "blue")
```



```
qplot(x=`covid Illinois`,y=hh_income,data = merge_1,geom = "point")+
  geom_smooth(method = "lm", se = FALSE, color = "blue")
```



The correlation between the keywords "covid" and household income is not statistically significant, yet both variables demonstrate a positive relationship. The correlation between "covid-19", "covid testing", "covid cases" and "covid Illinois" and household income is negative and not statistically significant. This suggests that as income levels rise, individuals may demonstrate greater risk resistance and lower concern for covid related keywords, except for keyword "covid".

The sample size is relatively limited, necessitating an expansion to obtain more robust scientific results.