Report: COVID-19 Data Analysis with R

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Introduction

This is exactly the TLDR you were looking for. Enjoy!

What is the project about?

The following project is an individual assignment taken from a course on Dataquest. The aim is to analyse a set of data collected from January 20th to June 1st 2020 taken from Kaggle. The research questions are the following:

- Q1: which countries have had the highest number of positive cases against the number of tests?
- **Q2**: which countries have made the best effort in terms of the number of COVID-19 tests conducted related to their population?
- Q3: which countries were ultimately the most and least affected related to their population?

The type of analysis is purely **descriptive**. All the snippets of code in this report use variables defined in the scripts.

Data

Raw data overview

We first analyse the dimensions, the column names and the information provided by each column in the data set:

glimpse(covid19_raw)

```
## Rows: 10,903
## Columns: 14
## $ Date
                      <chr> "2020-01-20", "2020-01-22", "2020-01-22", "202~
                      <chr> "Asia", "North America", "North America", "Nor~
## $ Continent_Name
## $ Two_Letter_Country_Code <chr> "KR", "US", "US", "US", "US", "US", "US", "US", "US",
## $ Country_Region
                      <chr> "South Korea", "United States", "United States~
## $ Province_State
                      <chr> "All States", "All States", "Washington", "All~
                      <int> 1, 1, 1, 1, 1, 2, 1, 1, 4, 0, 3, 0, 0, 0, 1~
## $ positive
## $ hospitalized
                      ## $ recovered
                      ## $ death
```

Looking at the data set repository from Kaggle we have further information on the meaning of each column:

- Date: date of the data collection
- Country_Region: country names
- Province_State: state/province names; value "All States" is put when state/provincial level data is NA
- positive: cumulative number of positive cases reported
- active: number of actively cases on that day
- hospitalized: cumulative number of hospitalized cases reported
- hospitalizedCurr: number of actively hospitalized cases on that day
- recovered: cumulative number of recovered cases reported
- death: cumulative number of deaths reported
- total tested: cumulative number of tests conducted

Note: Not all columns in our version of the data frame were present in the data description on Kaggle at the time of download. This indicates that the downloaded file was not updated to a later version of the data set. Yet, Dataquest gives the information provided by the other columns:

- Continent_Name: continent name
- Two_Letter_Country_Code: country codes
- Country_Region: country names
- daily_tested: number of tests conducted on the day; if daily data is unavailable, daily tested is averaged across number of days in between
- daily_positive: number of positive cases reported on the day; if daily data is unavailable, daily
 positive is averaged across number of days in

Raw Data Cleaning

Data Cleaning Q1

The first clean that we do is removing the Province_State column. Indeed, it might create some unwanted bias, as it also gives information about the specific province. So, to make the data "nationalized", we avoid looking at precise regions, and consider only those rows with "All States" value:

glimpse(covid19_allstates)

```
## Rows: 3,781
## Columns: 14
## $ Date
                    <chr> "2020-01-20", "2020-01-22", "2020-01-23", "202~
                    <chr> "Asia", "North America", "North America", "Asi~
## $ Continent_Name
## $ Two_Letter_Country_Code <chr> "KR", "US", "US", "KR", "US", "AU", "GB", "US"~
                    <chr> "South Korea", "United States", "United States~
## $ Country Region
                    <chr> "All States", "All States", "All States", "All~
## $ Province State
                    <int> 1, 1, 1, 2, 1, 4, 1, 1, 4, 0, 3, 1, 1, 5, 0, 0~
## $ positive
## $ hospitalized
                    ## $ recovered
                    ## $ death
                    <int> 4, 1, 1, 27, 1, 0, 31, 1, 0, 3, 51, 52, 1, 0, ~
## $ total_tested
## $ active
                    ## $ hospitalizedCurr
                    <int> 0, 0, 0, 5, 0, 0, 0, 0, 0, 12, 21, 0, 0, 0,~
## $ daily_tested
## $ daily_positive
```

As we can see we're left with just 30% of the data we previously had.

The second thing that we need to be aware of is the "unit of measurement" adopted by each column. For example, there are factors with cumulative information, others with daily information. To better address our research questions, we only consider the columns providing daily data, that is Date, Country_Region, active, hospitalizedCurr, daily_tested, daily_positive. We get the following refactored data frame:

glimpse(covid19_allstates_daily)

Now data has been cleaned for Q1.

Data Cleaning for Q2

Data Cleaning for Q3

Analysis

Data Analysis for Q1

Recall that with Q1 we are trying to answer the following question: which countries have had the highest number of positive cases against the number of tests?

Based on our cleaned data set, we notice that our data is collected on a daily basis. Therefore we can find the ratios on the overall number of positive cases over the total number of tests performed in each country each day. What We want to do is create a data set that groups Country_Region and aggregates data over all tests made, positive, active and hospitalized cases in the period of time that the data set covers (i.e. from January 20th to June 1st 2020).

Displaying the countries based on those that have done more tests, we get:

##	## # A tibble: 15 x 5					
##		Country_Region	tested	positive	active	hospitalized
##		<chr></chr>	<int></int>	<int></int>	<int></int>	<int></int>
##	1	United States	17282363	1877179	0	0
##	2	Russia	10542266	406368	6924890	0
##	3	Italy	4091291	251710	6202214	1699003
##	4	India	3692851	60959	0	0
##	5	Turkey	2031192	163941	2980960	0
##	6	Canada	1654779	90873	56454	0
##	7	United Kingdom	1473672	166909	0	0
##	8	Australia	1252900	7200	134586	6655
##	9	Peru	976790	59497	0	0
##	10	Poland	928256	23987	538203	0
##	11	South Korea	916276	11493	302633	0
##	12	Israel	546626	16647	0	30653
##	13 Germany		518647	29943	0	0
##	14 Belgium		511055	54209	220744	0
##	15	Czech Republic	446758	9321	0	0

Now, the result can be easily found by dividing the positive cases over the tested column and rank the result for each country. We are going to do it using vectors, extracted from the data frame, as an exercise.

We first extract a vector of the 10 countries that have done more tests, over which we are going to concude the rest of the analysis:

countries

```
## [1] "United States" "Russia" "Italy" "India" ## [5] "Turkey" "Canada" "United Kingdom" "Australia" ## [9] "Peru" "Poland"
```

Note: if we extract the vector on the whole cleaned data set we get that the length of the vector **countries** is 108, which means that we cover more than half of the world with our data. Yet we decide to restrict the study-case on just the top 10 is because we have more data on the tests made, and hence a better chance of having the correct data. As a matter of fact, there might be countries with higher ratios, but more incorrect and non sufficient data, making our data exploration biased (or at least not correct enough).

Then, we extract our two final vectors to run the analysis and assign to each key the name of the country. Thus, we get:

tested_cases

##	United States	Russia	Italy	India	Turkey
##	17282363	10542266	4091291	3692851	2031192
##	Canada	United Kingdom	Australia	Peru	Poland
##	1654779	1473672	1252900	976790	928256
pos	sitive_cases				
##	United States	Russia	Italy	India	Turkey
##	1877179	406368	251710	60959	163941
##	Canada	United Kingdom	Australia	Peru	Poland
##	90873	166909	7200	59497	23987

The conclusion lies in a simple division: positive_cases / tested_cases, which yields the following result:

positive_tested_ratio

##	Australia	India	Poland	Russia	Canada
##	0.005746668	0.016507300	0.025840932	0.038546552	0.054915490
##	Peru	Italy	Turkey	United States	United Kingdom
##	0.060910738	0.061523368	0.080711720	0.108618191	0.113260617

Data Analysis for Q2

Data Analysis for Q3

Conclusions

This paragraph concisely summarizes the findings of each research question:

• Q1: which countries have had the highest number of positive cases against the number of tests?

The following is the final vector we found in the Data Analysis for Q1 section:

##	Australia	India	Poland	Russia	Canada
##	0.005746668	0.016507300	0.025840932	0.038546552	0.054915490
##	Peru	Italy	Turkey	United States	United Kingdom
##	0.060910738	0.061523368	0.080711720	0.108618191	0.113260617

Since there is no precise meaning in the word *highest* that is in question, we decide to choose 4 countries with the highest positive/tests ratio. By doing so, we get:

```
## United Kingdom United States Turkey Italy
## 0.11326062 0.10861819 0.08071172 0.06152337
```

Therefore the top 4 countries were (in order):

- 1. United Kingdom with a ratio of 0.11326062
- 2. United States with a ratio of 0.10861819
- 3. Turkey with a ratio of 0.08071172
- 4. Italy with a ratio of 0.06152337

Their corresponding data is displayed in the next matrix:

```
## Positive/Tested % Tested Positive
## UK 11.326062 1473672 166909
## US 10.861819 17282363 1877179
## Turkey 8.071172 2031192 163941
## Italy 6.152337 4091291 251710
```

- **Q2**: which countries have made the best effort in terms of the number of COVID-19 tests conducted related to their population?
- Q3: which countries were ultimately the most and least affected related to their population?

Improvements and Comments on the Analysis

Q1: which countries have had the highest number of positive cases against the number of tests?

Even though the structure of this analysis was very clear and defined, the results obtained might not properly answer the question. As a matter of fact, the whole data analysis was mainly focused on the countries that had more data available regarding the number tests made. Therefore, if on the one hand we ended up having a pretty good estimate of the ratio for those countries with many swabs, on the other we might have exluded other possible feasible countries, simply because they did not have a sufficient number of tests made in this data set. From a purely statistical point of view, I cannot assess whether this result has a true value or not, as I still don't have sufficient knowledge on this matter. To improve the analysis, we could have tried to find a better assessment method for the countries to include in the analysis.

Q2: which countries have made the best effort in terms of the number of COVID-19 tests conducted related to their population?

Q3: which countries were ultimately the most and least affected related to their population?