## Milestone 2

**Group Team Name:** Group 27 AKA COVID KILLING ANALYZERS

**Team Members:** Natalie Mosqueda, Stephanie Younes, Eric Por

Briefly restate your data analysis goals. Note that if you have substantially changed your
project ideas since submitting milestone 1, please also include a short description of the
reason for the change and discuss your new project ideas using the same format as required in
milestone.

Our goal for this data analysis is to analyze the efficiency of the COVID-19 vaccine in many different methods. One main method we will be using is to compare country's vaccination data to number of day-to-day cases data from March 2021 to October 2021 to analyze how the vaccine impacts covid cases. With this, we can see the difference the vaccination has made on countries with more access to vaccines compared to the ones with less access. Our second method, we will be comparing the vaccination count to the virus' mortality rate of these countries. Lastly, we plan on seeing the impact the vaccine has on the hospitalization rate of countries.

## 2. Describe the data you have

We got our data from the Our World in Data official website (<a href="https://ourworldindata.org/coronavirus-source-data">https://ourworldindata.org/coronavirus-source-data</a>) The general format the data we found looked as such with a couple more columns:

34	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.511
34	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
34	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
34	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
34	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
40	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
45	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
50	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
56	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
61	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
66	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
72	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
72	12.04	39835428	54.422	18.6	2.581	1.337	1803.987	597.029	9.59			37.746	0.5	64.83	0.51
72 72	12.04 30.56	39835428	54.422 54.422	18.6	2.581		1803.987	597.029 507.029	9.59 9.59			37.746	0.5	64.83 64.83	
72	30.56	20835428	54 422	19.6	2 591	1 227	1902 097	507 070	9 59	d 65 oldag	ed 70 oldg	37 746	0.5	EV 83	n 51
72	30.56	20835428	54 422	19.6	2 591	1 227	ir population p	507 070	9 59	d_65_old ag 2.581	ed_70_old g 1.337	37 746	0.5	EV 83	n s 1 ·
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vaccina to	30.56	20835428	54 422	19.6	new_vaccina	stringency	_ir population p 04 39835428 04 39835428	opulation_c media 54.422	o so an_age age	2.581	1.337	37 746 dp_per_cap extre 1803.987	0.5	6/ 83 diovasc_d dia 597.029	0.51 betes_pi 9.5! 9.5!
vaccina to 1367 1367	30.56	20835428	54 422	19.6	new_vaccina	stringency	ir population p 04 39835428 04 39835428 04 39835428	opulation_c media 54.422 54.422	an_age age 18.6 18.6	2.581 2.581	1.337 1.337	27 7/6 dp_per_cap extro 1803.987 1803.987	0.5	diovasc_d dia 597.029 597.029	9.5: 9.5: 9.5: 9.5:
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vaccina to 1367 1367 1367 1367 1367	an se	people_vac	sa app	19.6	new_vaccina 34 34 34 34 34 40	1 227 stringency 12. 12. 12. 12.	1903 097 Lir population p 04 39835428 04 39835428 04 39835428 04 39835428 04 39835428 04 39835428 04 39835428	54.422 54.422 54.422 54.422 54.422 54.422 54.422 54.422 54.422 54.422	18.6 18.6 18.6 18.6 18.6	2.581 2.581 2.581 2.581 2.581	1.337 1.337 1.337 1.337	dp_per_cap extra 1803.987 1803.987 1803.987 1803.987 1803.987	0.5	64 83 rdiovasc_d dia 597.029 597.029 597.029 597.029	9.5 9.5 9.5 9.5 9.5 9.5 9.5 9.5
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But, after pre-processing, we have taken out the data that is not necessary and the data now includes, iso\_code, continent, location, data, total\_cases, new\_cases, total\_deaths, new deaths, new\_vaccinations\_smoothed, population, . Here is a snippet of our data now:

iso_code	continent	location	date	total_cases	new_cases	total_deaths	new_deaths	total_cases_	new_cases_	total_deaths	new_deaths
AFG	Asia	Afghanistan	2/24/21	55664	18	2436	1	1397.349	0.452	61.152	0.025
AFG	Asia	Afghanistan	2/25/21	55680	16	2438	2	1397.751	0.402	61.202	0.05
AFG	Asia	Afghanistan	2/26/21	55696	16	2442	4	1398.152	0.402	61.302	0.1
AFG	Asia	Afghanistan	2/27/21	55707	11	2443	1	1398.429	0.276	61.327	0.025
AFG	Asia	Afghanistan	2/28/21	55714	7	2443	0	1398.604	0.176	61.327	0
AFG	Asia	Afghanistan	3/1/21	55733	19	2444	1	1399.081	0.477	61.352	0.025
AFG	Asia	Afghanistan	3/2/21	55759	26	2446	2	1399.734	0.653	61.403	0.05
AFG	Asia	Afghanistan	3/3/21	55770	11	2446	0	1400.01	0.276	61.403	0
AFG	Asia	Afghanistan	3/4/21	55775	5	2446	0	1400.136	0.126	61.403	0
AFG	Asia	Afghanistan	3/5/21	55827	52	2449	3	1401.441	1.305	61.478	0.075
AFG	Asia	Afghanistan	3/6/21	55840	13	2449	0	1401.767	0.326	61.478	0
AFG	Asia	Afghanistan	3/7/21	55847	7	2449	0	1401.943	0.176	61.478	0
AFG	Asia	Afghanistan	3/8/21	55876	29	2451	2	1402.671	0.728	61.528	0.05
AFG	Asia	Afghanistan	3/9/21	55876	0	2451	0	1402.671	0	61.528	0
AFG	Asia	Afghanistan	3/10/21	55894	18	2451	0	1403.123	0.452	61.528	0
ΔFG	Δsia	Δføhanistan	3/11/21	55917	23	2451	n	1403.7	0 577	61 528	n

NEW DATA

icu_patients_	hosp_patient	hosp_patien	weekly_icu_	weekly_icu_	weekly_hos	p weekly_	hosp new_	vaccina	population	median_age	aged_65_old ag	ed_70_old	d cardiovasc_dea	ath_rate
								1367	39835428	18.6	2.581	1.337	597.029	
								1367	39835428	18.6	2.581	1.337	597.029	
								1367	39835428	18.6	2.581	1.337	597.029	
								1367	39835428	18.6	2.581	1.337	597.029	
								1367	39835428	18.6	2.581	1.337	597.029	
								1580	39835428	18.6	2.581	1.337	597.029	
								1794	39835428	18.6	2.581	1.337	597.029	
								2008	39835428	18.6	2.581	1.337	597.029	
								2221	39835428	18.6	2.581	1.337	597.029	
								2435	39835428	18.6	2.581	1.337	597.029	
								2649	39835428	18.6	2.581	1.337	597.029	
								2862	39835428	18.6	2.581	1.337	597.029	
	icu_patients_	icu_patients, hosp_patient	icu_patients hosp_patien(hosp_patien	icu_patients hosp_patien(hosp_patien(weekly_icu_	icu_patients hosp_patient hosp_patient weekly_icu_i weekly_icu_	icu_patients hosp_patien(hosp_patien(weekly_icu_i weekly_icu_i weekly_hos	icu_patients, hosp_patient hosp_patient weekly_icu_; weekly_icu_; weekly_hosp weekly_	icu_patients hosp_patien( weekly_icu_: weekly_icu_: weekly_hosp weekly_hosp new_	1367 1367 1367 1367 1367 1367 1580 1794 2008 2221 2435 2649	1367 39835428 1367 39835428 1367 39835428 1367 39835428 1367 39835428 1367 39835428 1580 39835428 1580 39835428 2008 39835428 2221 39835428 2435 39835428	1367 39835428 18.6 1367 39835428 18.6 1367 39835428 18.6 1367 39835428 18.6 1367 39835428 18.6 1367 39835428 18.6 1367 39835428 18.6 1367 39835428 18.6 1580 39835428 18.6 1794 39835428 18.6 2008 39835428 18.6 2221 39835428 18.6 2221 39835428 18.6 2243 39835428 18.6	1367 39835428 18.6 2.581 1367 39835428 18.6 2.581 1367 39835428 18.6 2.581 1367 39835428 18.6 2.581 1367 39835428 18.6 2.581 1367 39835428 18.6 2.581 1367 39835428 18.6 2.581 1580 39835428 18.6 2.581 1794 39835428 18.6 2.581 2008 39835428 18.6 2.581 2008 39835428 18.6 2.581 2221 39835428 18.6 2.581 2435 39835428 18.6 2.581 2435 39835428 18.6 2.581	1367 39835428 18.6 2.581 1.337 1367 39835428 18.6 2.581 1.337 1367 39835428 18.6 2.581 1.337 1367 39835428 18.6 2.581 1.337 1367 39835428 18.6 2.581 1.337 1367 39835428 18.6 2.581 1.337 1580 39835428 18.6 2.581 1.337 1580 39835428 18.6 2.581 1.337 2008 39835428 18.6 2.581 1.337 2008 39835428 18.6 2.581 1.337 2008 39835428 18.6 2.581 1.337 2008 39835428 18.6 2.581 1.337 2221 39835428 18.6 2.581 1.337	1367     39835428     18.6     2.581     1.337     597.029       1367     39835428     18.6     2.581     1.337     597.029       1367     39835428     18.6     2.581     1.337     597.029       1367     39835428     18.6     2.581     1.337     597.029       1580     39835428     18.6     2.581     1.337     597.029       1794     39835428     18.6     2.581     1.337     597.029       2008     39835428     18.6     2.581     1.337     597.029       2221     39835428     18.6     2.581     1.337     597.029       2435     39835428     18.6     2.581     1.337     597.029       2649     39835428     18.6     2.581     1.337     597.029

We will of course save a copy of the original data in the case we need to refer to deleted columns. The size of the data is about 55,815 rows and at the moment we have all the data we need but we are open to researching for more data if necessary.

## 3. Show and discuss your preliminary analysis. For example, this could include:

The data we plan on using is split into three different methods we want to use.

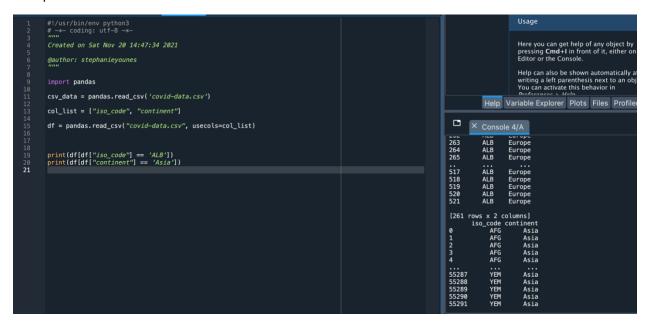
The first method being we want to compare country's vaccination data to number of day-to-day cases data from March 2021 to October 2021. For this method the data needed is location, date, new vaccinations, and total cases per million.

The second method being we want to compare vaccination count to the virus' mortality rate of the countries. For this method the data needed is also location, date, total vaccination, and total deaths per million.

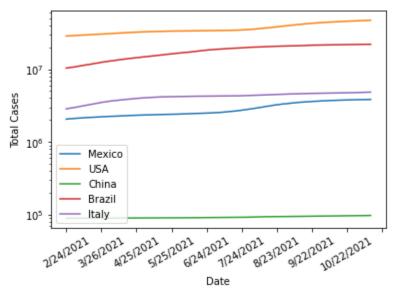
For our last method, we plan on seeing the impact the vaccine has on the hospitalization rate of countries. For this last method, we would need location, date, total vaccination, and icu patients.

To achieve our goal, we do not plan on transforming the data, but we plan on parsing the data by columns. With this, we will be able to read each column to test our methods out.

Below is an example of how we plan on parsing the columns and the result of doing so in order to manipulate the data:



We first sort out the desired data from the file by specifying the ISO\_CODE from the datasets to obtain the total cases and dates for each country. Currently we're working on switching the whole process to have each country's data parsed without specifying the ISO\_CODE. Below, is an example of the code we implemented alongside the graph



```
import pandas as pd
10
          from pandas import Grouper
          import matplotlib.pyplot as plt
          import matplotlib.dates as mdates
          df = pd.read_csv('Covid Data final.csv', header=[0],parse_dates=True)
          mex_cases = df.loc[df["iso_code"] ==
usa_cases = df.loc[df["iso_code"] ==
chn_cases = df.loc[df["iso_code"] ==
bra_cases = df.loc[df["iso_code"] ==
ita_cases = df.loc[df["iso_code"] ==
                                                                  == 'MEX', "total_cases"]
== 'USA', "total_cases"]
== 'CHN', "total_cases"]
== 'BRA', "total_cases"]
== 'ITA', "total_cases"]
          mex_date = df.loc[df["iso_code"] ==
usa_date = df.loc[df["iso_code"] ==
chn_date = df.loc[df["iso_code"] ==
bra_date = df.loc[df["iso_code"] ==
ita_date = df.loc[df["iso_code"] ==
                                                                                 "date"
"date"
                                                                      'USA',
'CHN',
                                                                == 'BRA',
== 'ITA',
                                                                                  "date"
          countries = ['Mexico','USA','China','Brazil','Italy']
          fig = plt.figure()
plt.xlabel("Date")
plt.ylabel("Total Cases")
          plt.plot(mex_date,mex_cases)
          plt.plot(usa_date,usa_cases)
          plt.plot(chn_date,chn_cases)
          plt.plot(bra_date,bra_cases)
          plt.plot(ita_date,ita_cases)
          plt.semilogy()
          plt.legend(countries)
          plt.gca().xaxis.set_major_locator(mdates.DayLocator(interval=30))
          plt.gca().xaxis.set_tick_params(rotation = 30)
          plt.show()
```