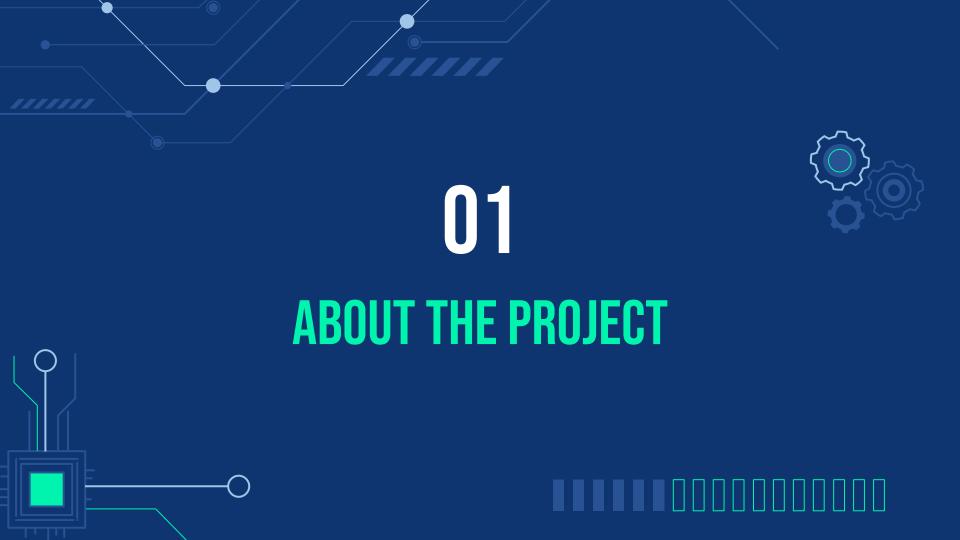


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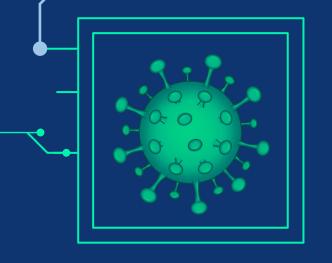




We are examining datasets about the coronavirus pandemic, and trying to predict when the pandemic will come to an end using data about vaccinations, herd immunity, and hospitalizations

### **DATA MINING GOAL**

Our goal is to predict when 70% of the U.S. population will have CV-19 immunity, which will assume herd immunity, using the datasets acquired.





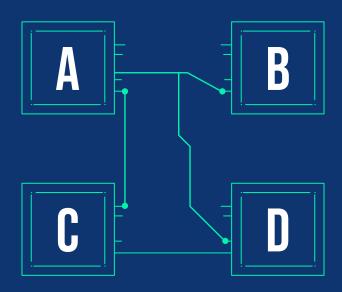
### **OUR DATASETS**

# COVID-19 WORLD VACCINATION PROGRESS

This dataset tells us about: which country is using what vaccine, which country's vaccination programme is more advanced, and where the rate of vaccinated people per day is higher in terms of percent from the entire population.

# COUNTRIES POPULATION BY YEAR 2020

This dataset provides us with the world population and top 20 countries' live clock. It contains population data for the past, present and future.



# NOVEL CORONAVIRUS 2019 DATASET

This dataset has daily level information on the number of affected cases, deaths and recovery from COVID-19. It provides us with data about each country and their cases, deaths and recoveries.

# COVID-19 DAILY VACCINATION

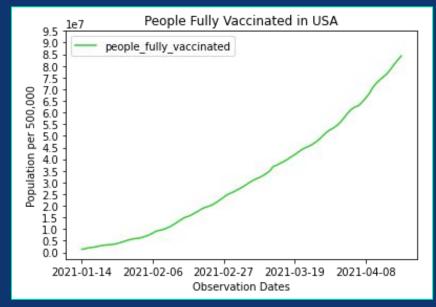
This data set contains vaccination data for countries showing how many people are vaccinated daily.



#### **FULLY VACCINATED INDIVIDUALS IN THE USA**

The mean, standard deviation, min, max, and the quartiles are found through Dataset A

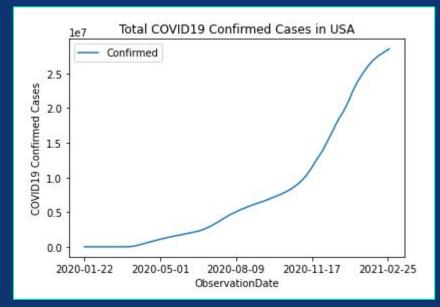
91.00000 count 31929907.90110 mean std 24208365.96282 min 1342086.00000 25% 9679222.00000 50% 27795980.00000 75% 49418470.50000 84263408,00000 max dtype: object



### **USA CONFIRMED COVID-19 CASES**

The mean, standard deviation, min, max, and the quartiles are found through Dataset B

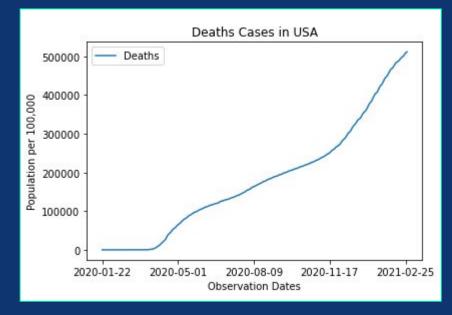
| count  | 21462.00000   |
|--------|---------------|
| mean   | 149160.90355  |
| std    | 309144.87948  |
| min    | 0.00000       |
| 25%    | 2710.25000    |
| 50%    | 35865.50000   |
| 75%    | 156865.25000  |
| max    | 3563578.00000 |
| dtype: | object        |



### **USA DEATHS CAUSED BY COVID-19**

The mean, standard deviation, min, max, and the quartiles are found through Dataset B

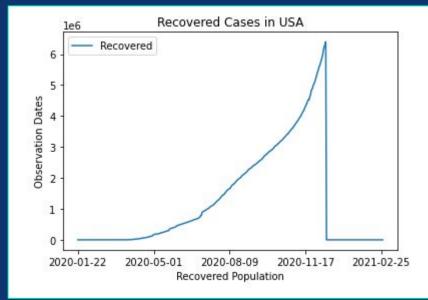
|       | Deaths        |
|-------|---------------|
| count | 403.000000    |
| mean  | 179213.498759 |
| std   | 142776.472362 |
| min   | 0.000000      |
| 25%   | 66075.000000  |
| 50%   | 164041.000000 |
| 75%   | 253826.000000 |
| max   | 511994.000000 |



### **USA COVID-19 RECOVERED CASES**

The mean, standard deviation, min, max, and the quartiles are found through Dataset B

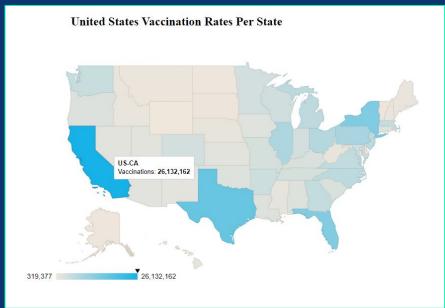
| count  | 403.00000     |
|--------|---------------|
| mean   | 1249059.44417 |
| std    | 1630170.41782 |
| min    | 0.00000       |
| 25%    | 3.00000       |
| 50%    | 391508.00000  |
| 75%    | 2292820.50000 |
| max    | 6399531.00000 |
| dtype: | object        |
|        |               |



### **USA DAILY VACCINE BY STATE**

The mean, standard deviation, min, max, and the quartiles are found through Dataset D

| count  | 50.00000       |
|--------|----------------|
| mean   | 4089885.82000  |
| std    | 4695572.64397  |
| min    | 319377.00000   |
| 25%    | 1110997.00000  |
| 50%    | 2653070.00000  |
| 75%    | 5102783.50000  |
| max    | 26132162.00000 |
| dtype: | object         |

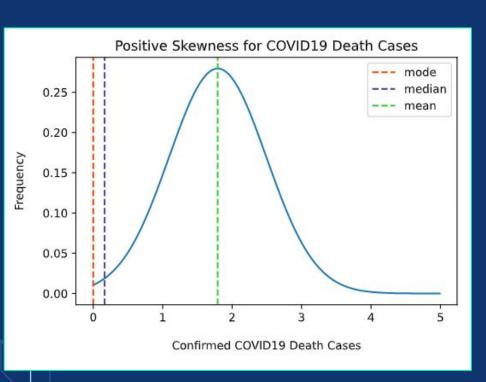


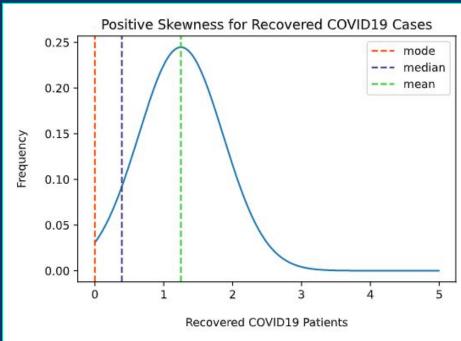


Our group looked at multiple statistical techniques to evaluate the datasets, those techniques are:

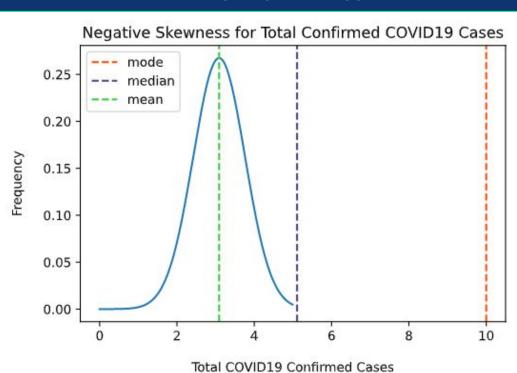
- 1. Feature skewness and looking at dataset characteristics such as mean, mode, median and standard deviation for:
  - a. Confirmed Positive CV-19 cases
  - b. Recovered CV-19 cases
  - c. Death by CV-19 cases
- Correlations between attributes and using heatmaps to model it.
- 3. Kernel Density Estimation to estimate the probability density function of the random variables to make inferences about the population based on the finite data sample that we have.

### **FEATURE SKEWNESS**

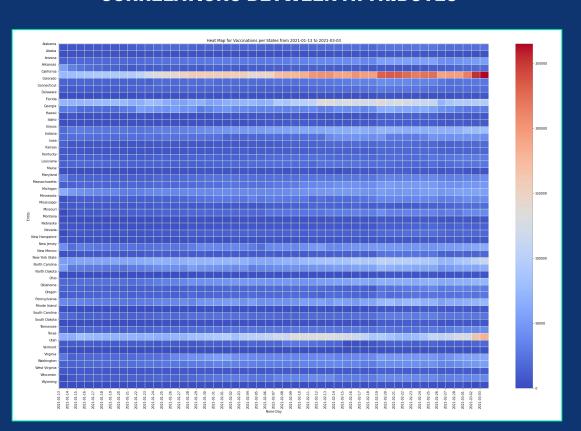




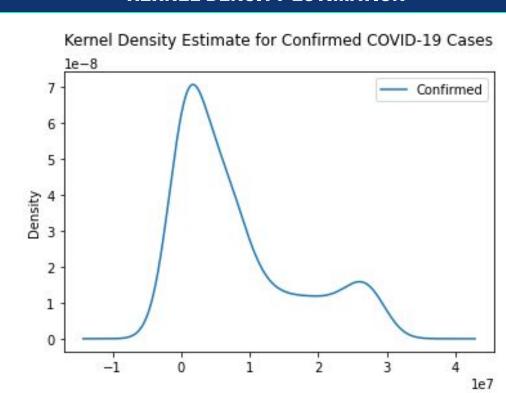
### **FEATURE SKEWNESS**



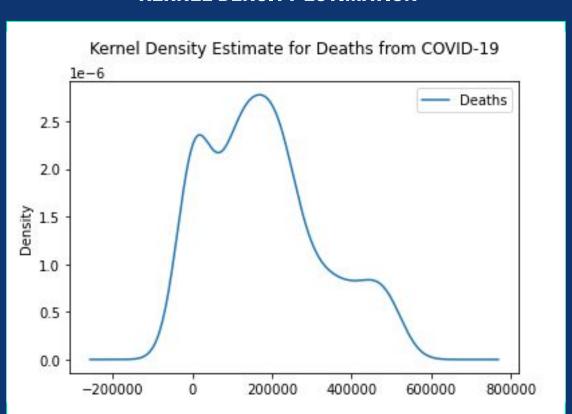
### **CORRELATIONS BETWEEN ATTRIBUTES**



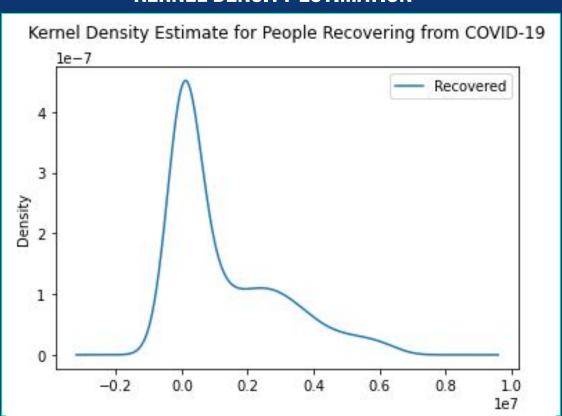
### **KERNEL DENSITY ESTIMATION**

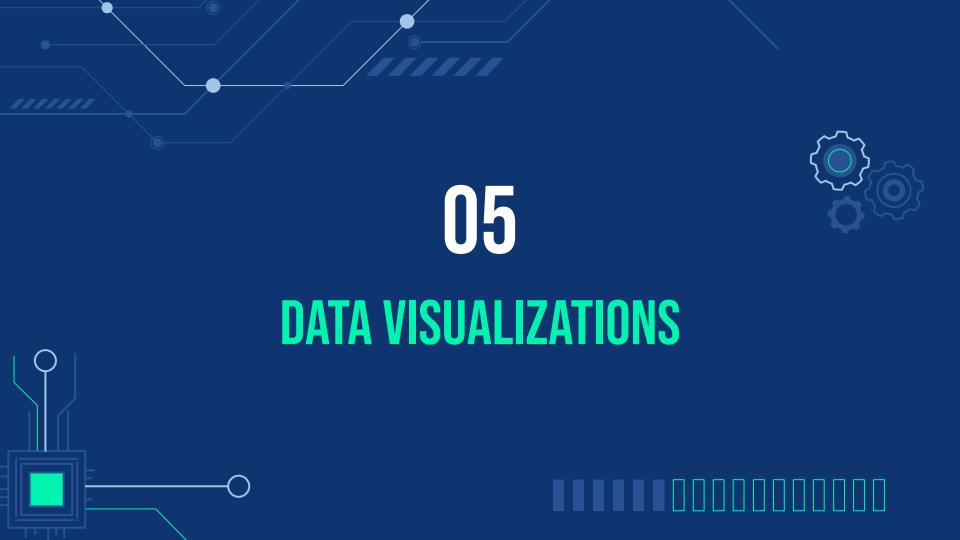


### **KERNEL DENSITY ESTIMATION**



### **KERNEL DENSITY ESTIMATION**





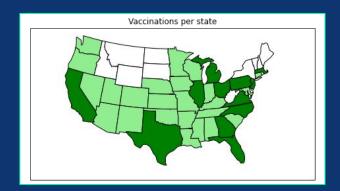
### **DATA VISUALIZATION**

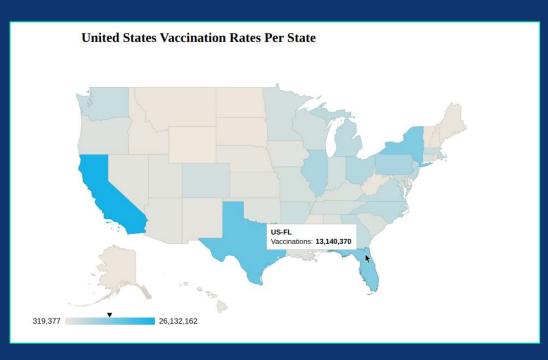
#### **GOOGLE CHARTS**

Google charts was used for creating an interactive map that shows the USA vaccination progress per state.

#### **CARTOPY**

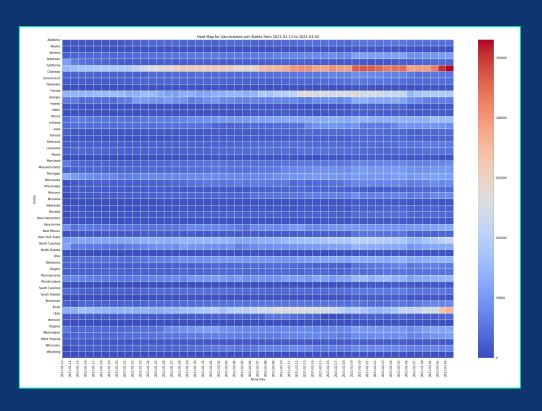
The Cartopy library was used for easy visualization of Geomaps.







## **DATA VISUALIZATION**

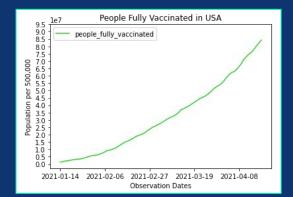


#### **SEABORN**

The Seaborn library was used to create heat maps to show the correlation between the attributes

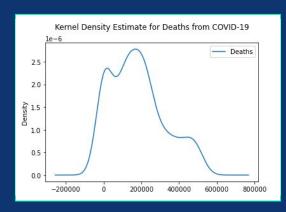


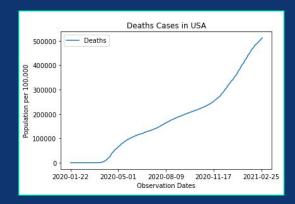
### **DATA VISUALIZATION**



#### **MATPLOT**

The team used the famous matplot library for creating every other 2 dimensional graph such as skewness, rate of fatalities, and density







## DATA CLEANING AND TRANSFORMATION (NEXT STEP)

Here we can see one of datasets, Countries population by year 2020.csv, has N.A. values in addition to NaN. We can choose to drop these rows or fill in those values with other data such as 0 for NaN.

Example of a process we may follow is that we will drop all columns containing N.A., parse percentage values into float types, and fill NaN values as 0. We would also like to note that we will normalize our data as well.

NaN count by attribute for country\_vaccinations.csv

| 230 | Montserrat       | 4991 | 0.06 % | 3   | 50   | 100   | NaN | N.A. | N.A. | 10 % | 0.00 % |
|-----|------------------|------|--------|-----|------|-------|-----|------|------|------|--------|
| 231 | Falkland Islands | 3458 | 3.05 % | 103 | 0    | 12170 | NaN | N.A. | N.A. | 66 % | 0.00 % |
| 232 | Niue             | 1624 | 0.68 % | 11  | 6    | 260   | NaN | N.A. | N.A. | 46 % | 0.00 9 |
| 233 | Tokelau          | 1354 | 1.27 % | 17  | 136  | 10    | NaN | N.A. | N.A. | 0 %  | 0.00 9 |
| 234 | Holy See         | 801  | 0.25 % | 2   | 2003 | 0     | NaN | N.A. | N.A. | N.A. | 0.00   |

| 96 | Aruba               | 106675 | 0.43  | 452  | 593 | 180 | 201.0  | 1.9 | 41 | 44.0 | 0.00 |
|----|---------------------|--------|-------|------|-----|-----|--------|-----|----|------|------|
| 97 | Tonga               | 105449 | 1.15  | 1201 | 147 | 720 | -800.0 | 3.6 | 22 | 24.0 | 0.00 |
| 98 | U.S. Virgin Islands | 104456 | -0.15 | -153 | 298 | 350 | -451.0 | 2.0 | 43 | 96.0 | 0.00 |
| 99 | Seychelles          | 98224  | 0.62  | 608  | 214 | 460 | -200.0 | 2.5 | 34 | 56.0 | 0.0  |
| 00 | Antiqua and Barbuda | 97764  | 0.84  | 811  | 223 | 440 | 0.0    | 2.0 | 34 | 26.0 | 0.0  |

```
NaN value by attribute
country
iso code
date
total vaccinations
people vaccinated
                                        15
people fully vaccinated
                                        29
daily vaccinations raw
                                        24
daily vaccinations
total vaccinations per hundred
                                        14
people vaccinated per hundred
                                        15
                                        29
people_fully_vaccinated_per_hundred
daily_vaccinations_per_million
vaccines
                                         0
source name
source website
dtype: int64
```

