

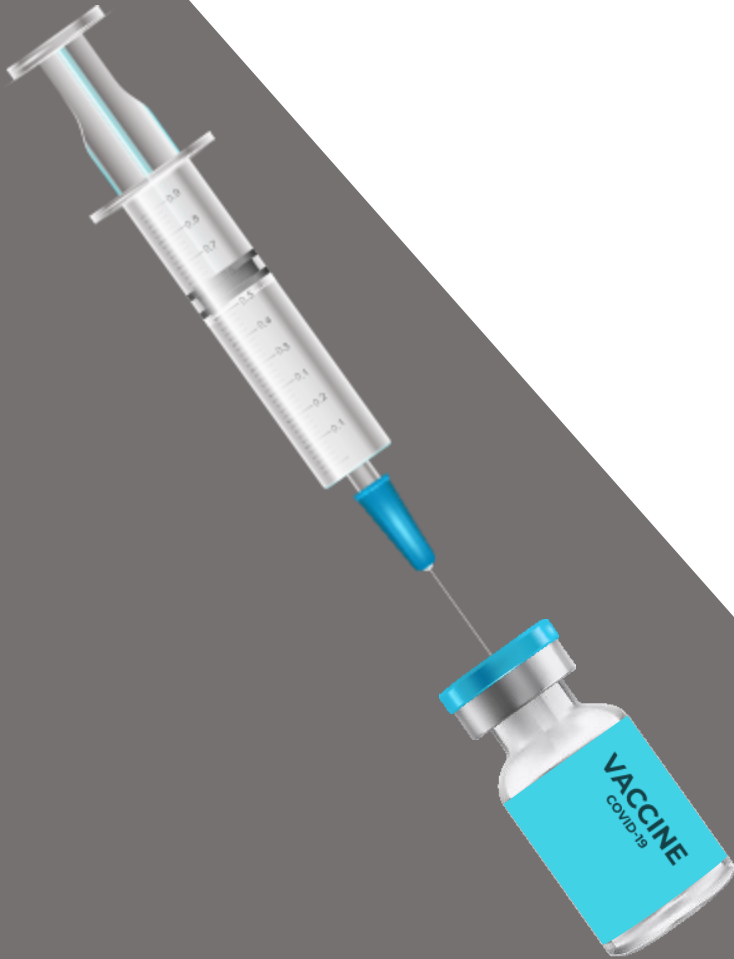
COVID-19

Coronavirus vaccine efficiency

Can we find out from Data Work, is vaccine effective ?

Data Science project

Date:26/6/2021



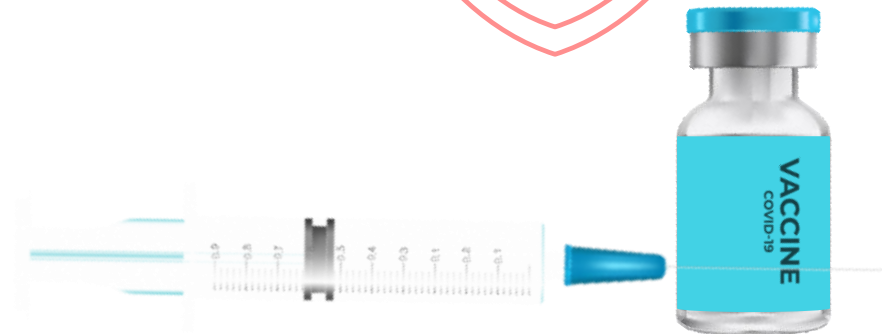
WHAT IS A VACCINE ?

A **vaccine** is a type of medicine that trains the body's immune system so that it can fight a disease it has not come into contact with before.



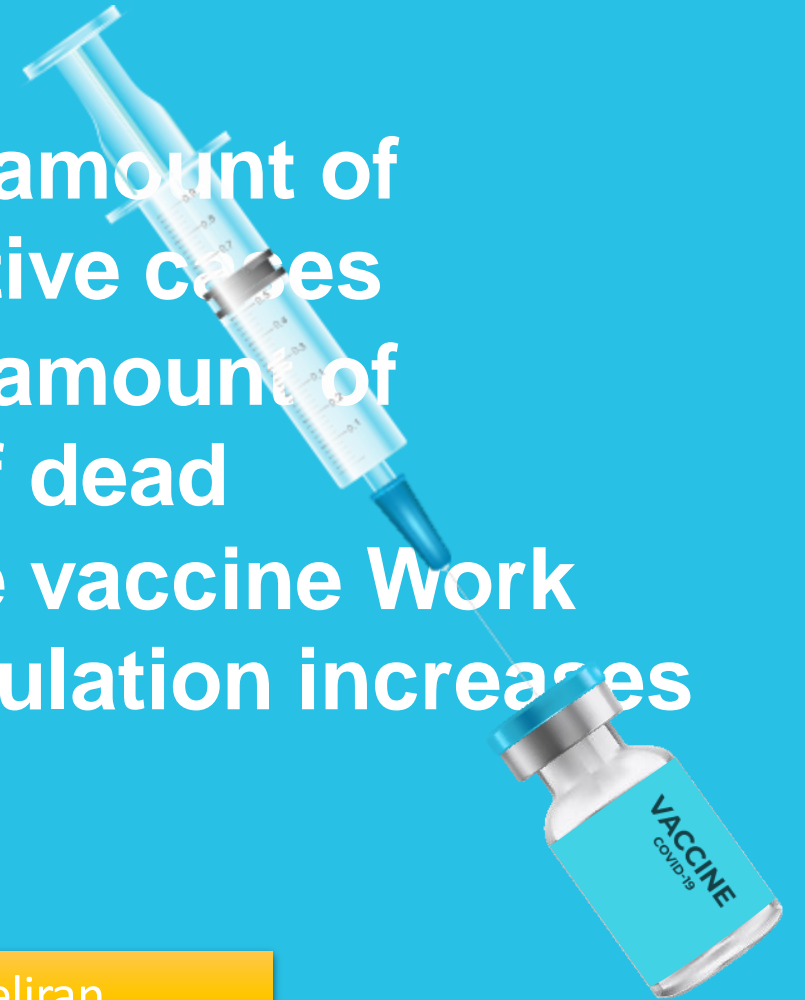
How does a **VACCINE** work?

Vaccines contain weakened or inactive parts of a particular organism (antigen) that triggers an immune response within the body. Newer **vaccines** contain the blueprint for producing antigens rather than the antigen itself.



Our Research Will Check:

- 1) What is the connection between The amount of vaccinated population to amount of active cases
- 2) What is the connection between The amount of vaccinated population To the number of dead
- 3) Will Try to prove that efficiency of the vaccine Work better As the number of vaccinated population increases



Github: <https://github.com/nikaloamashvili/corona-time-nika-eliran>

Data Sources



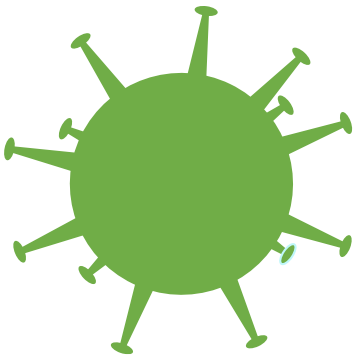
API

- <https://ourworldindata.org/covid-vaccinations>
- <https://www.worldometers.info/coronavirus/#countries>

Data cleaning

Vaccines worldwide

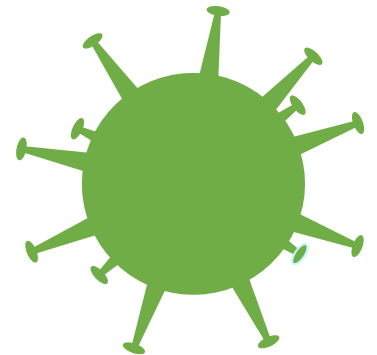
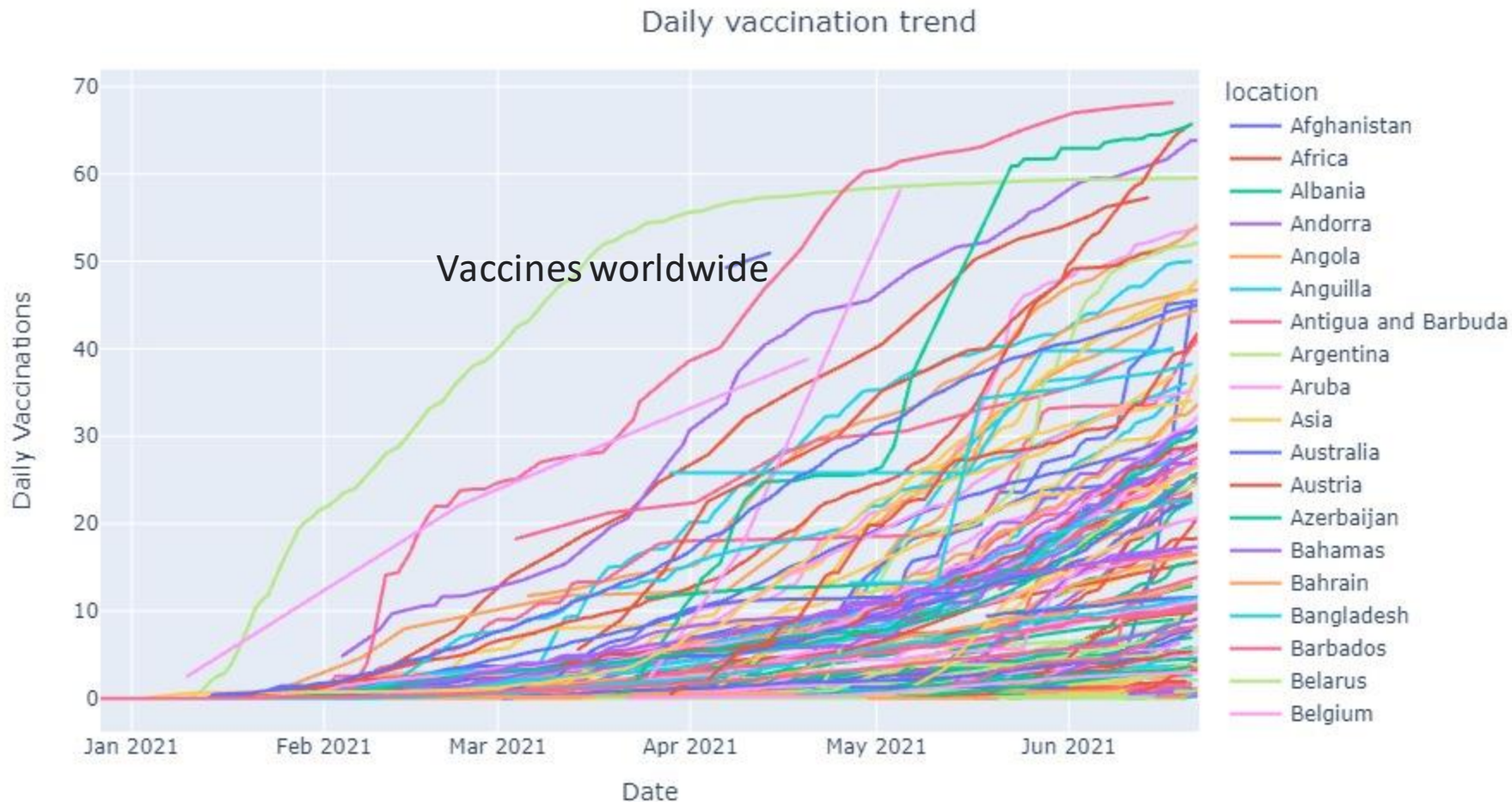
```
rVaccinated.dropna(subset = ["people_fully_vaccinated_per_hundred"], inplace=True)
rVaccinated = rVaccinated[rVaccinated["location"] != 'Gibraltar']
rVaccinated.query("people_fully_vaccinated_per_hundred>0", inplace = True)
```



- We rejected the countries vaccinated with over 100%
- We didn't use information with data that are =0 or not exists(vaccinated %)

Vaccines worldwide

- %of vaccinated population worldwide



GERMANY

ISRAEL

UNITED STATES

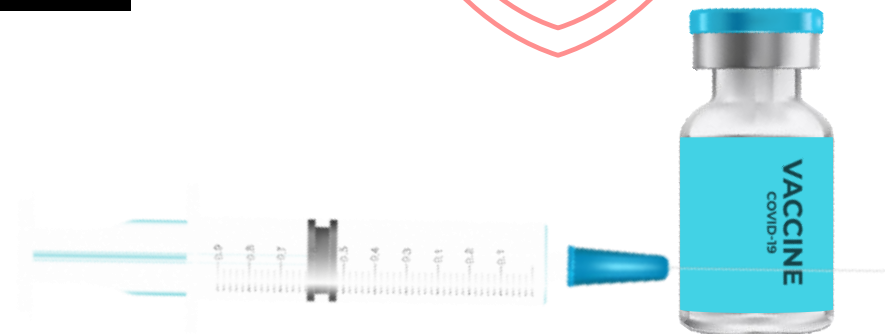
We chose to check 7 countries for our research
6 Countries with high % of vaccinated
1 Country with low % of vaccinated

UNITED KINGDOM

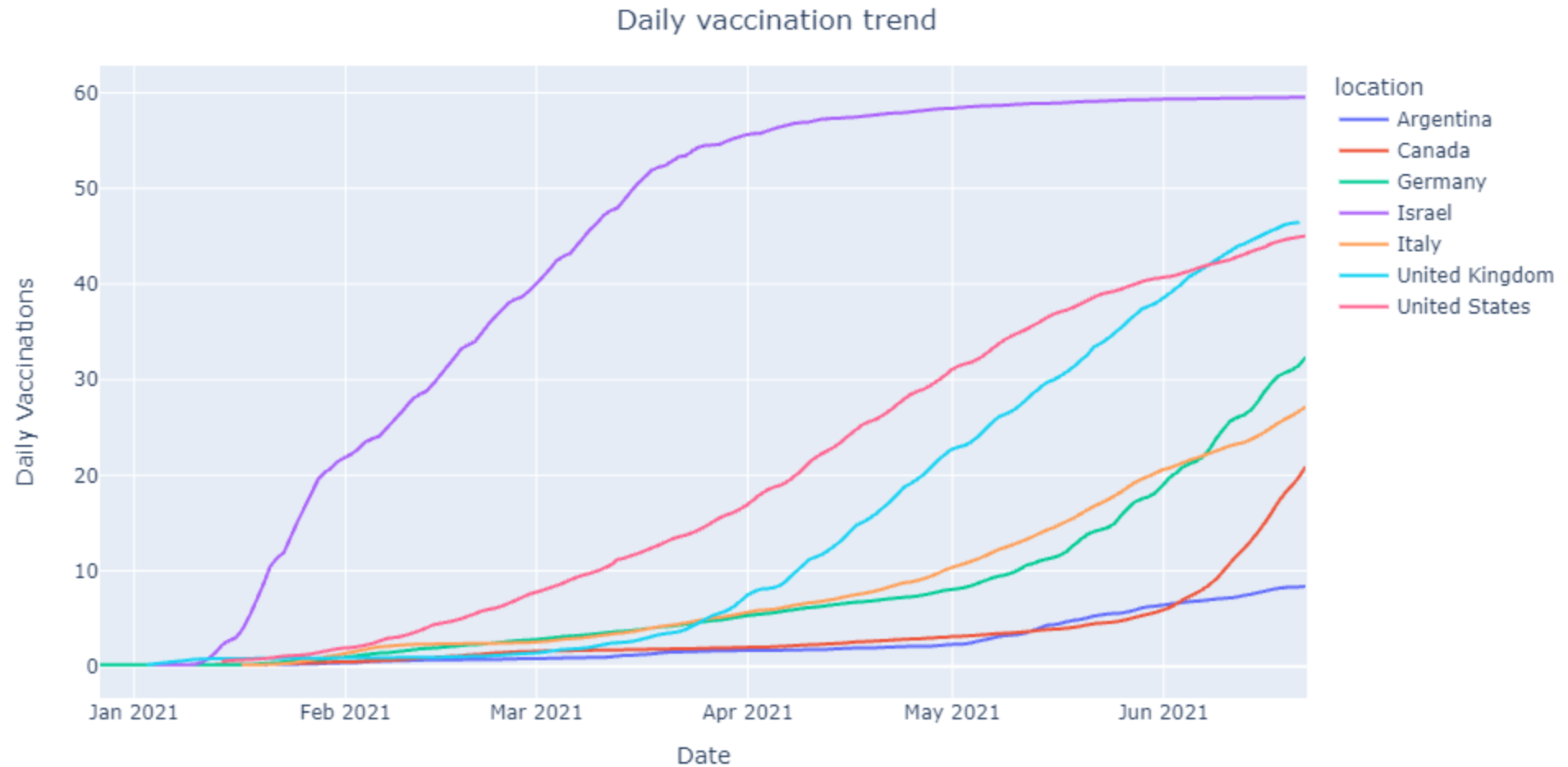
ITALY

Argentina

CANADA



%vaccinated population in selected counries



```

countries=["UK","US","Canada","Israel","Italy","Germany","Argentina"]
# Get Data only for Israel
for c in countries:
    KPI = covid_daily.data(country=c, chart = 'graph-cases-daily', as_json=False)
    Vaccinated=rVaccinated.copy()
    Vaccinated.query("location=='%s'" %c, inplace = True)
    # Start Measuring from the time the vaccines started
    #Vaccinated.query("people_fully_vaccinated_per_hundred>0", inplace =
True)
    # Conver the Day to Datetime Index
    Vaccinated['date'] = pd.DatetimeIndex(Vaccinated['date'])

    # Join the data
    merged = KPI.merge(Vaccinated, how='inner', left_index=True,
right_on='date')
    merged.index=merged.date
    merged.index.name = 'Date'

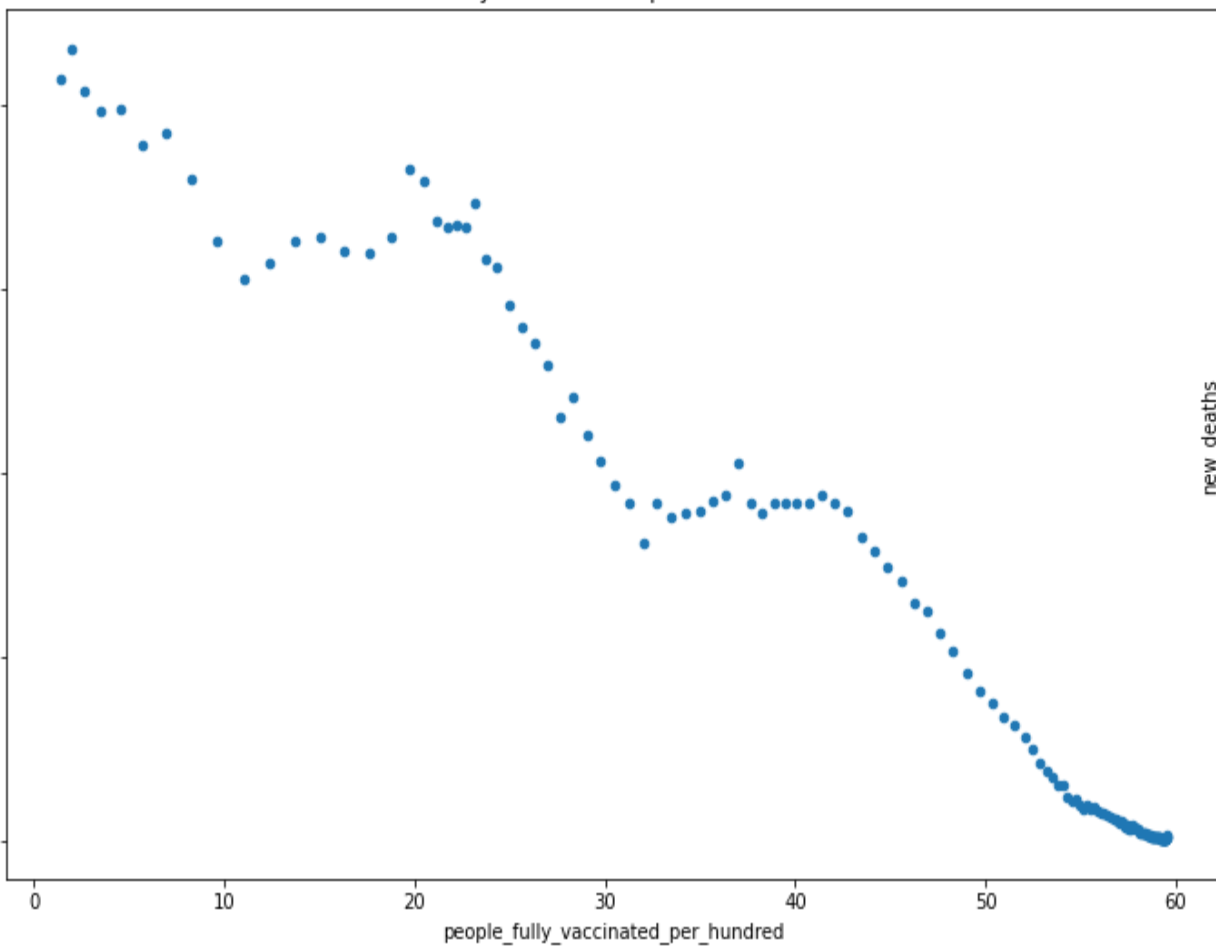
    rolling_merged = merged[['people_fully_vaccinated_per_hundred', 'Novel
Coronavirus Daily Cases']].rolling(7).mean().dropna()

    rolling_merged.plot.scatter('people_fully_vaccinated_per_hundred', 'Novel
Coronavirus Daily Cases',
        figsize=(12,8), title = c+": Daily Cases vs %Population Vaccinated")

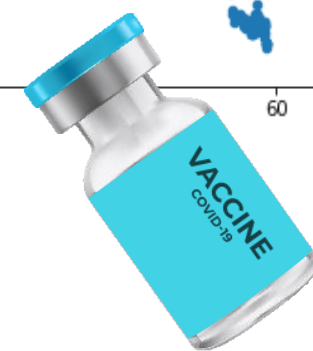
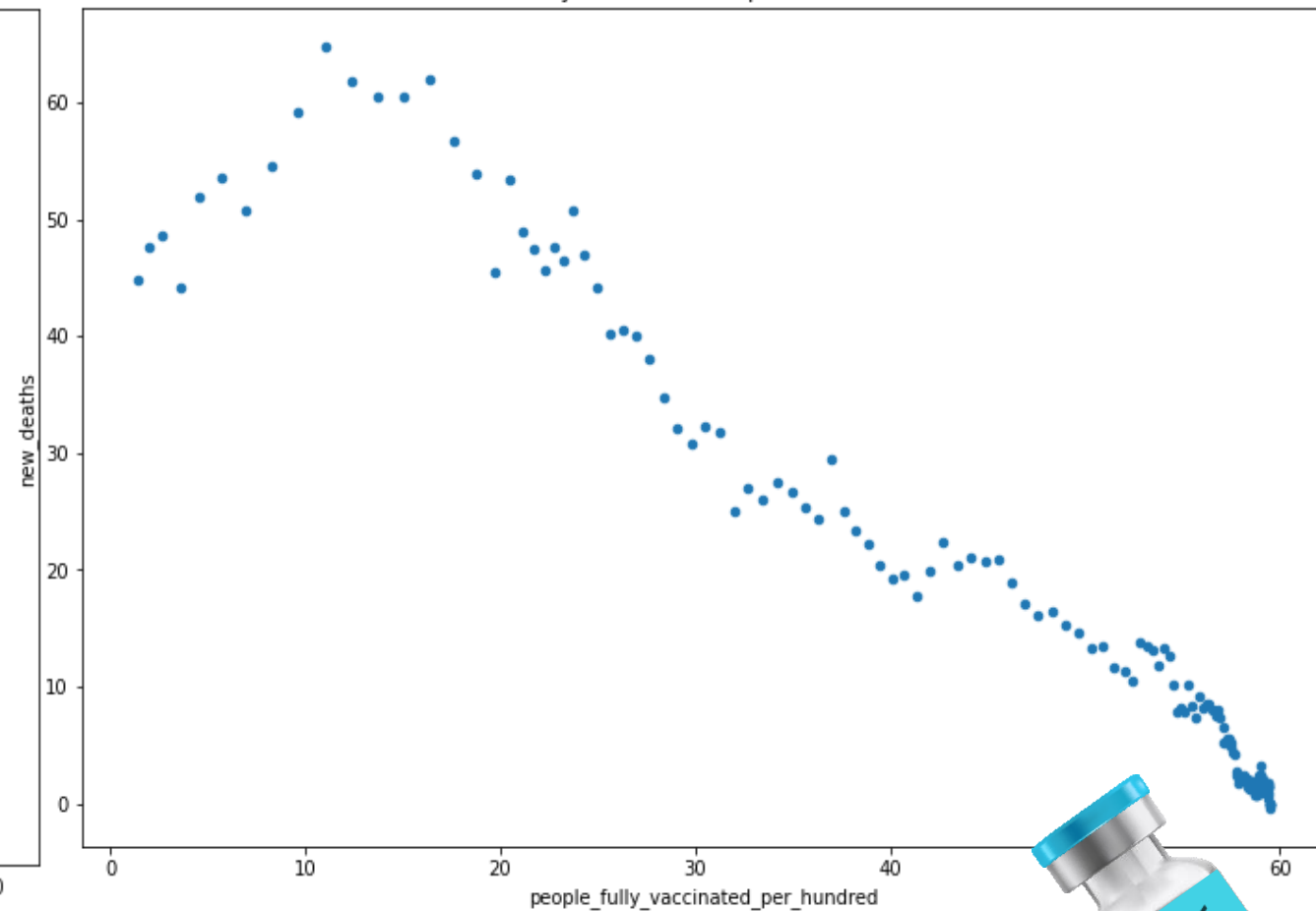
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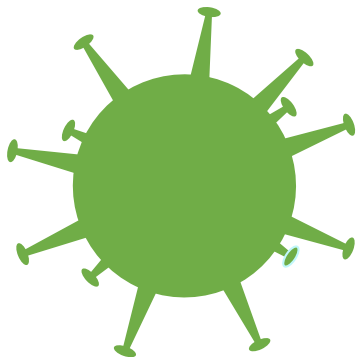
From the data we created 2 graphs for each country that shows the corelation between the % of new cases and the % of vaccinated, and the conection between vaccinated and total deaths

Israel: Daily Cases vs %Population Vaccinated



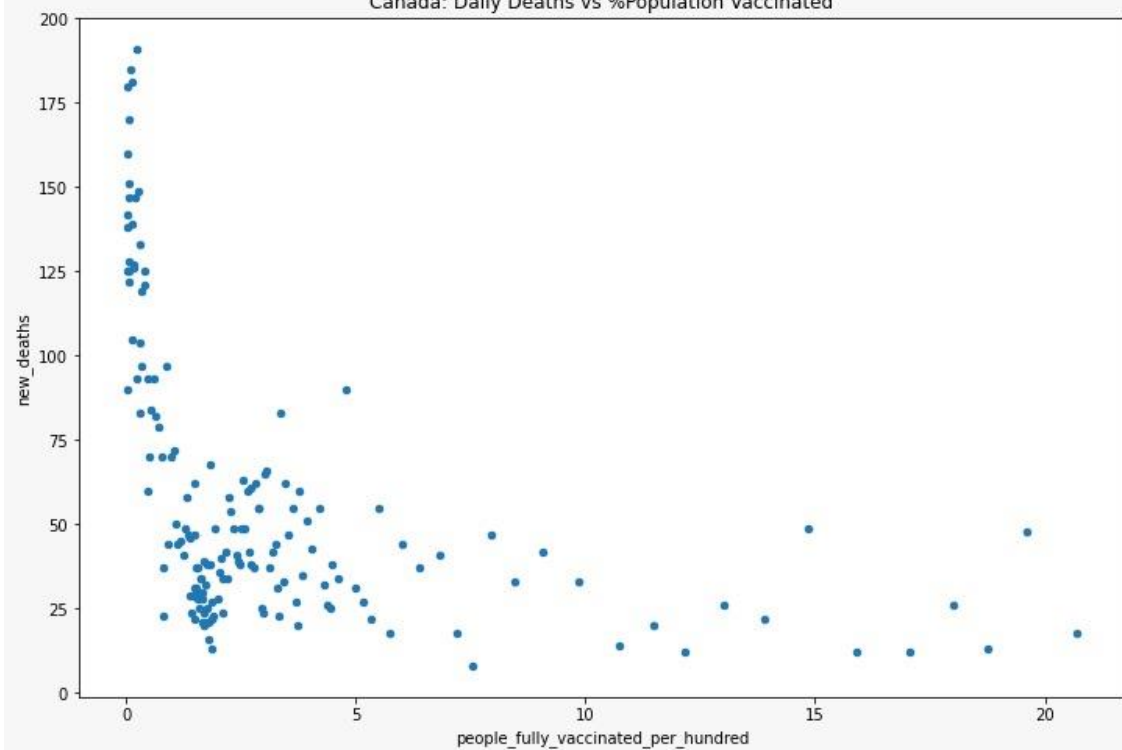
Israel: Daily Deaths vs %Population Vaccinated



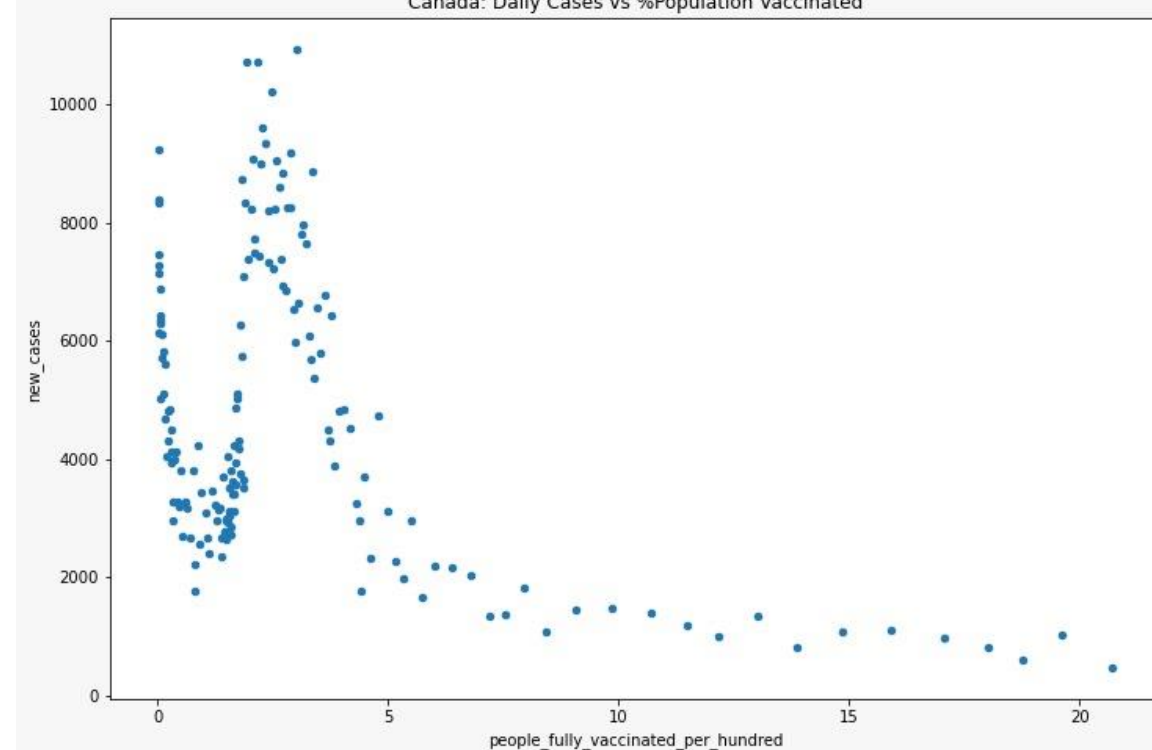


CANADA

Canada: Daily Deaths vs %Population Vaccinated

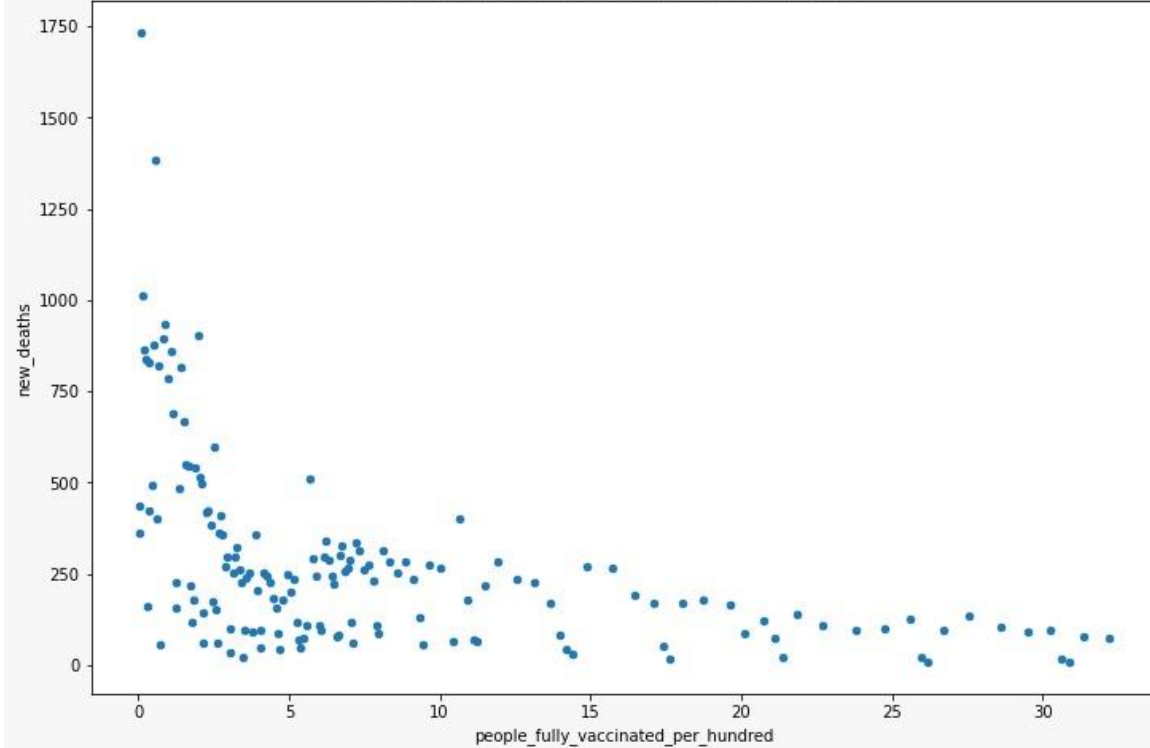


Canada: Daily Cases vs %Population Vaccinated

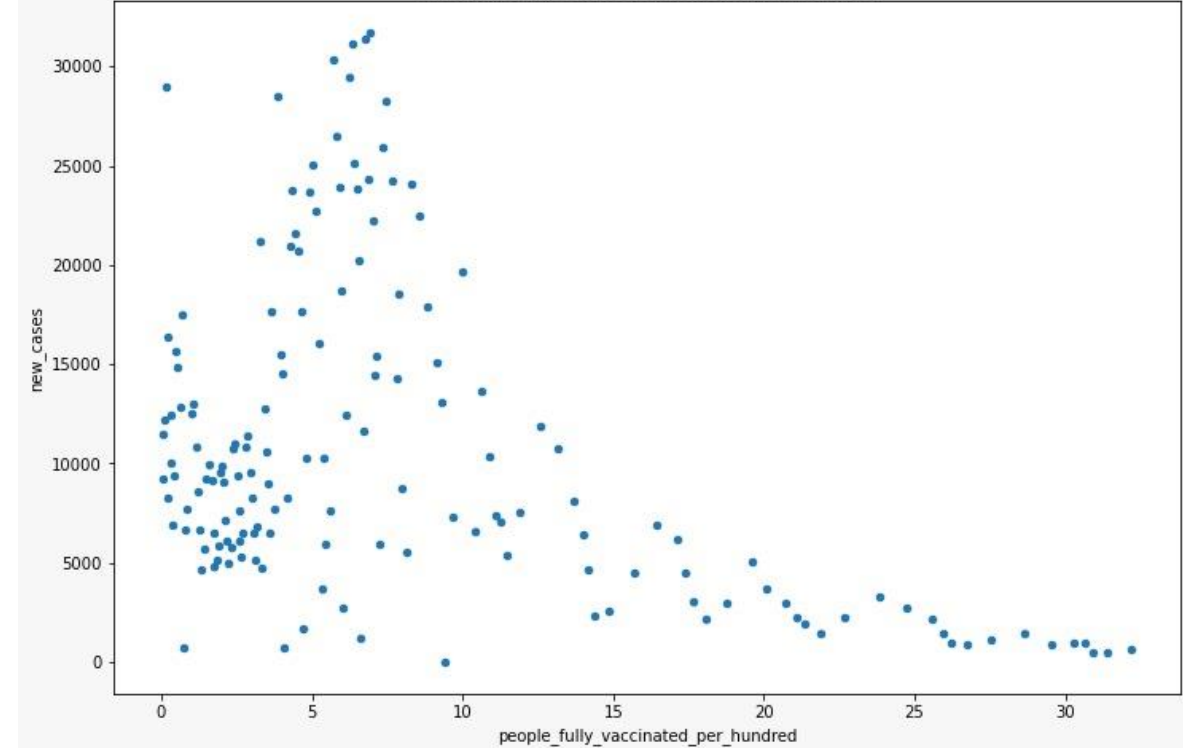


GERMANY

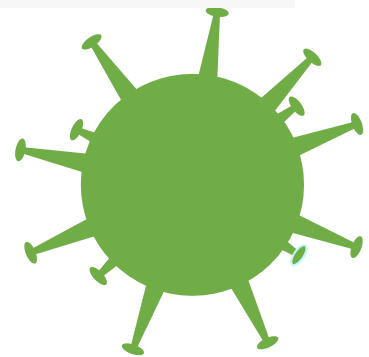
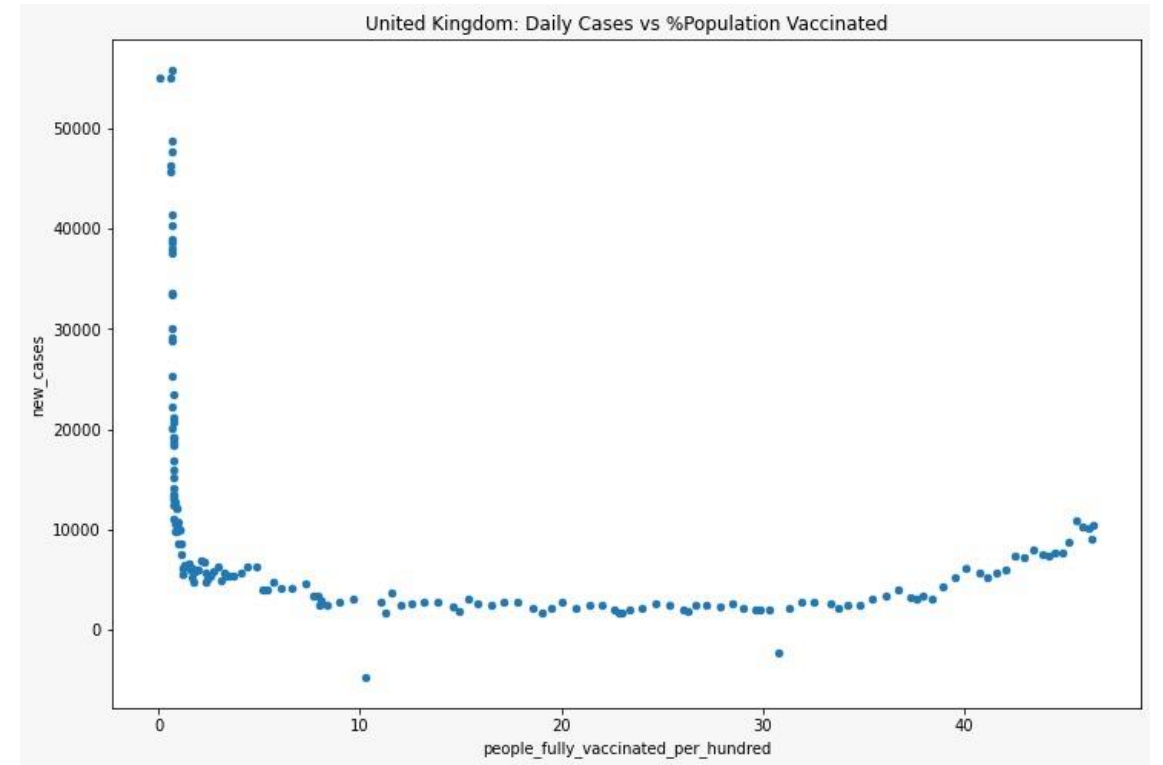
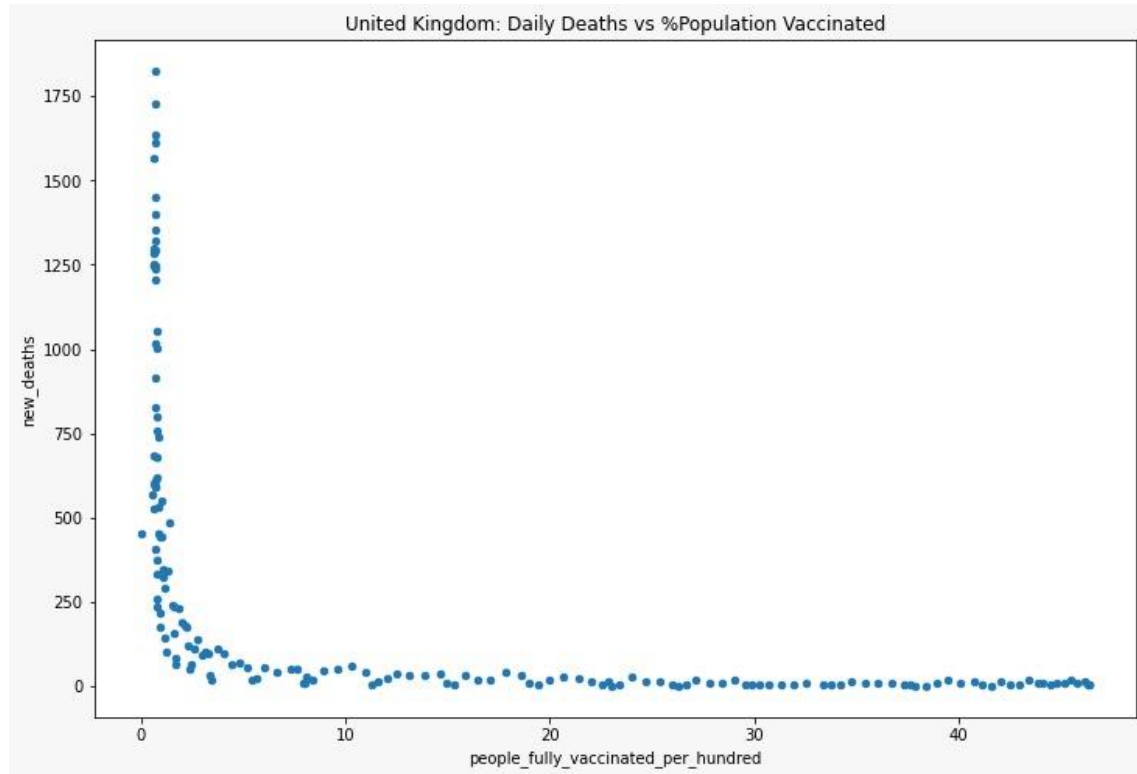
Germany: Daily Deaths vs %Population Vaccinated

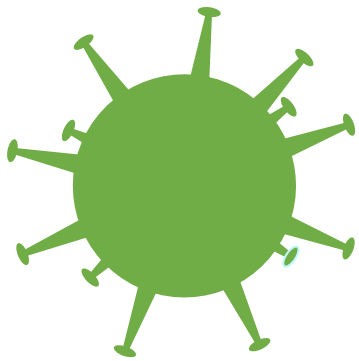


Germany: Daily Cases vs %Population Vaccinated



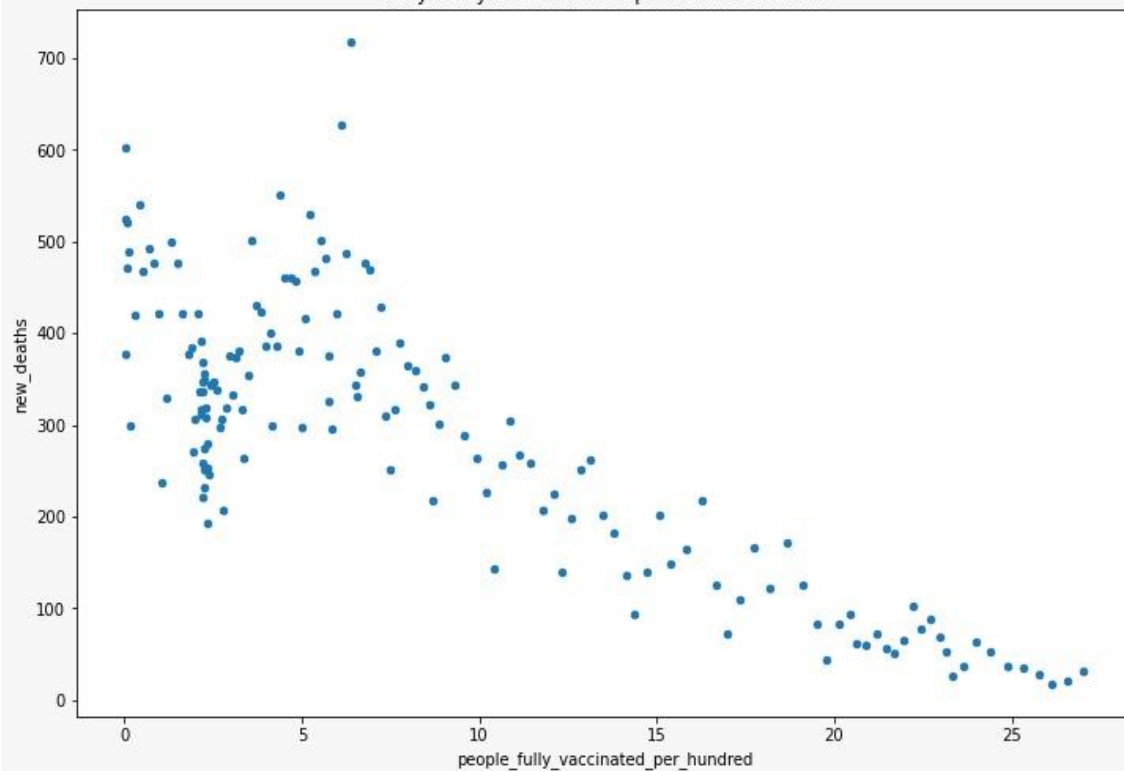
UNITED KINGDOM



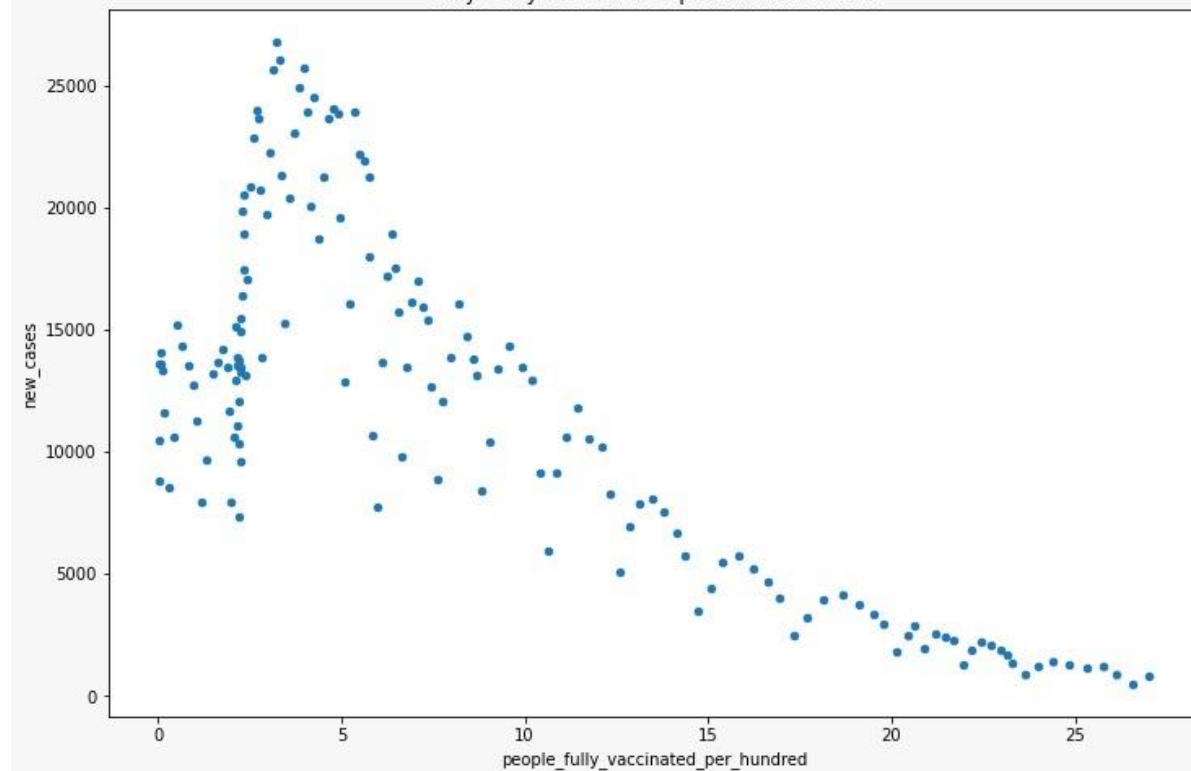


ITALY

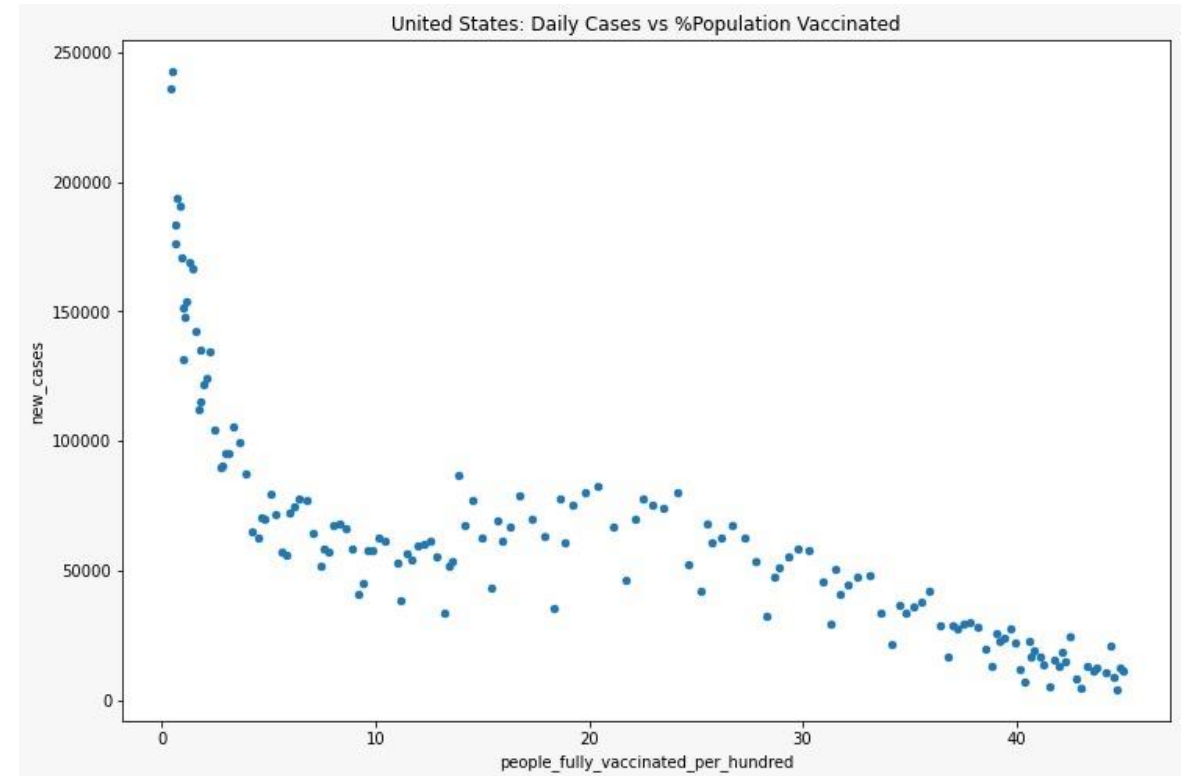
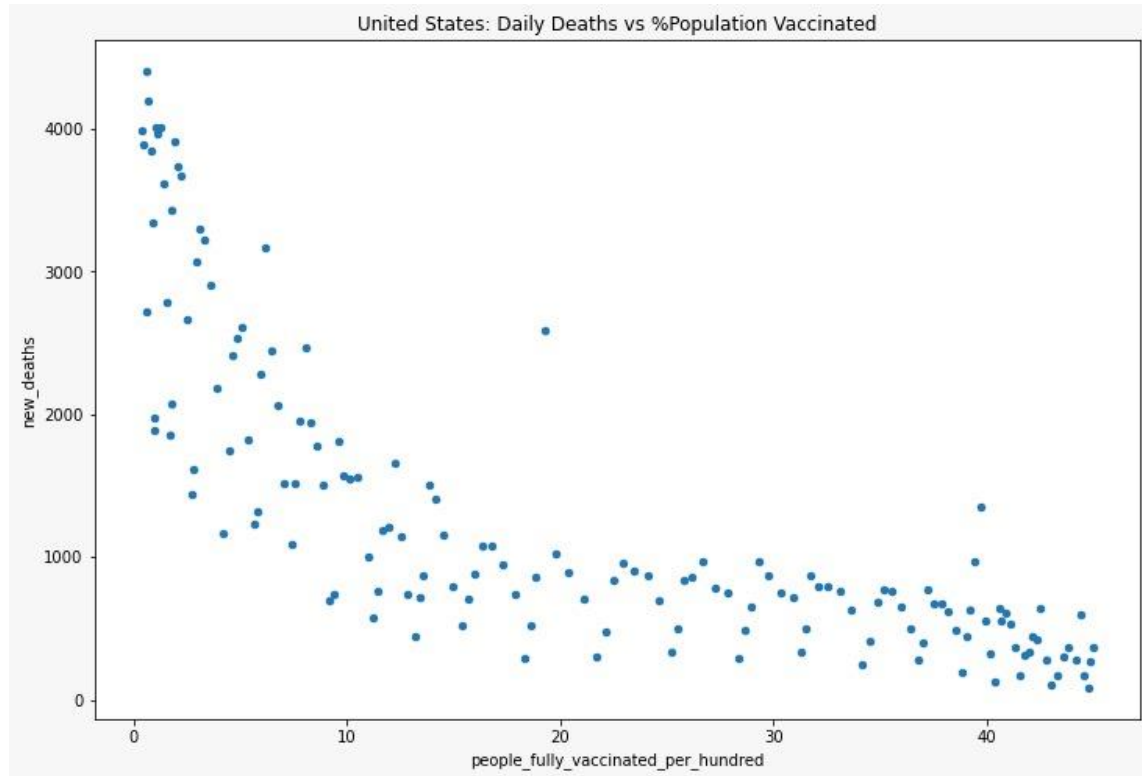
Italy: Daily Deaths vs %Population Vaccinated



Italy: Daily Cases vs %Population Vaccinated

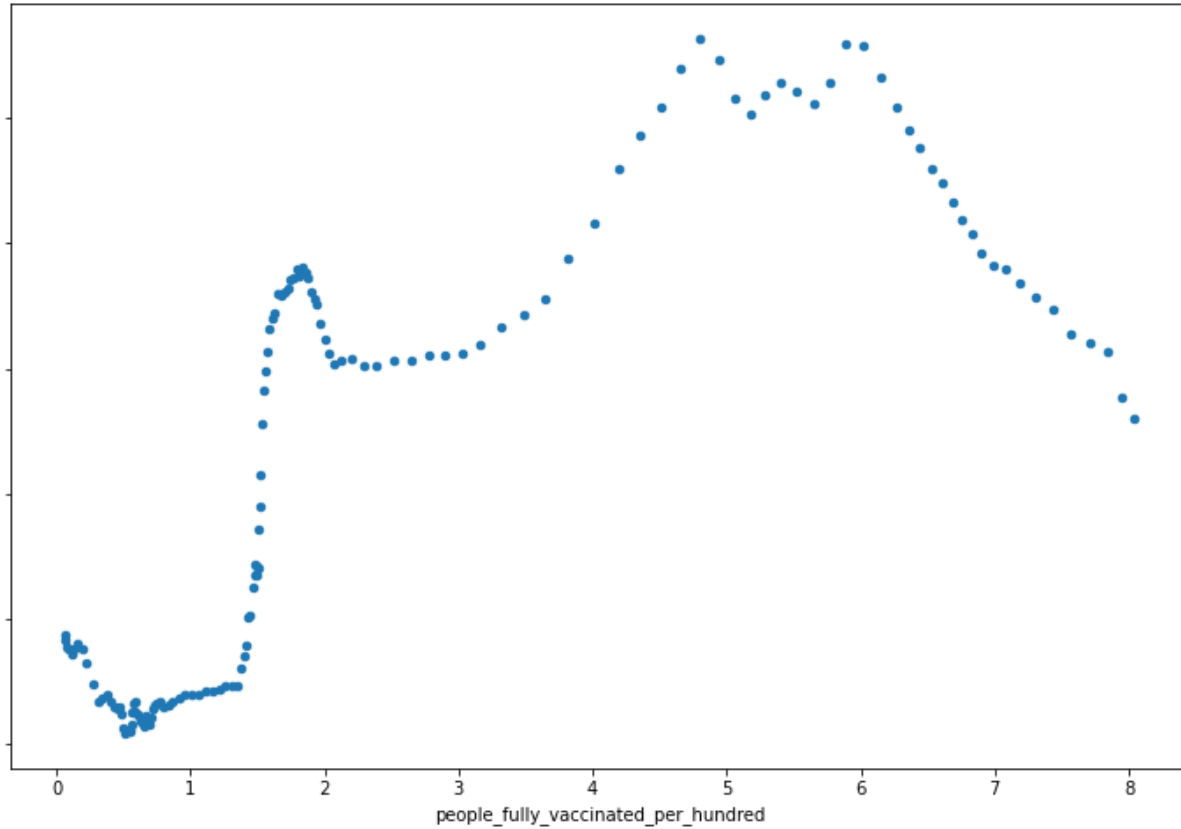


UNITED STATES

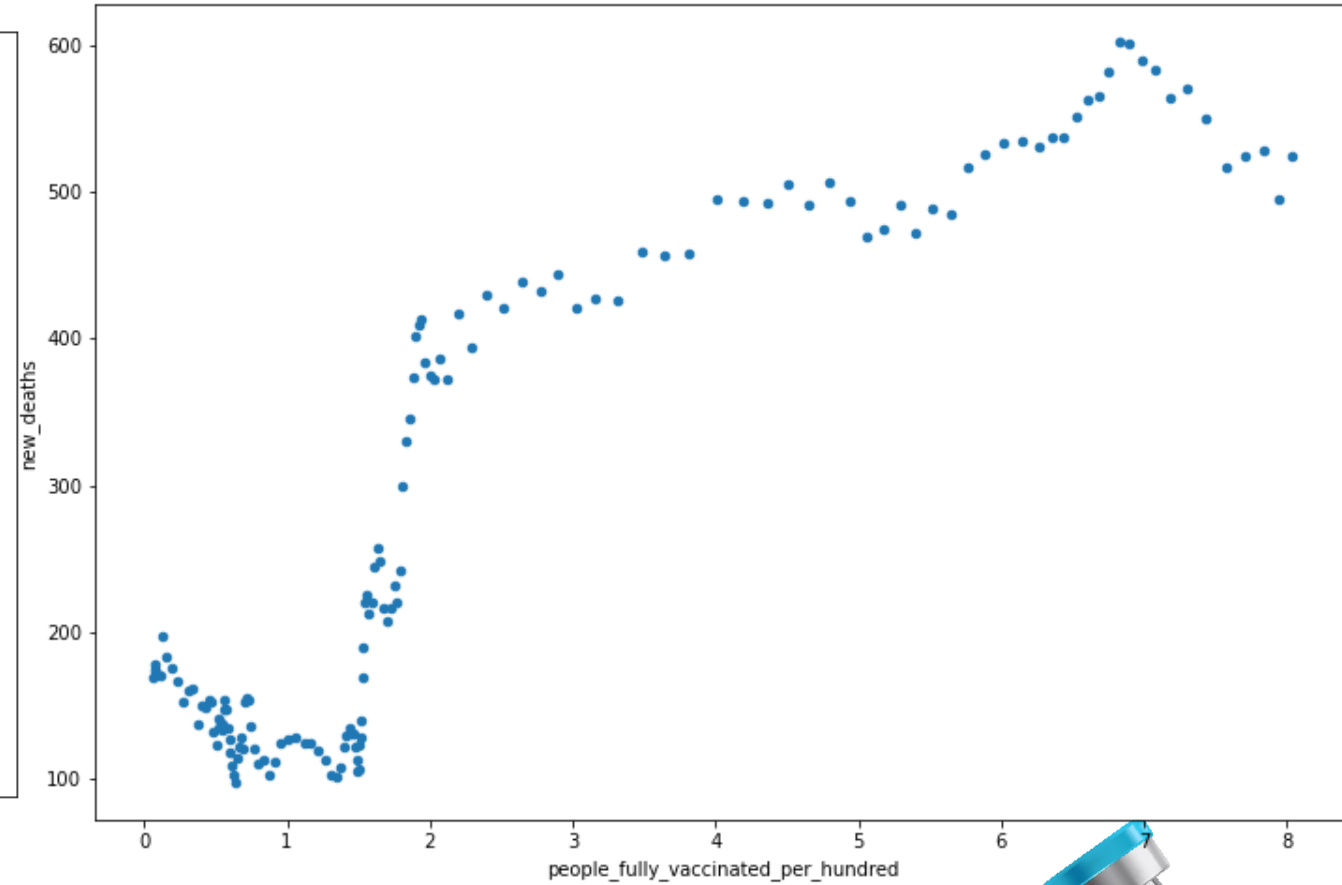


ARGENTINA

Argentina: Daily Cases vs %Population Vaccinated



Argentina: Daily Deaths vs %Population Vaccinated



EDA

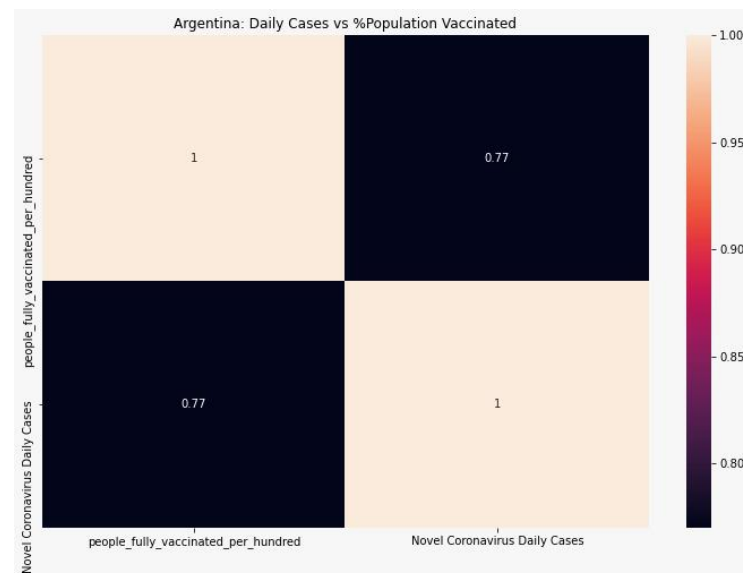
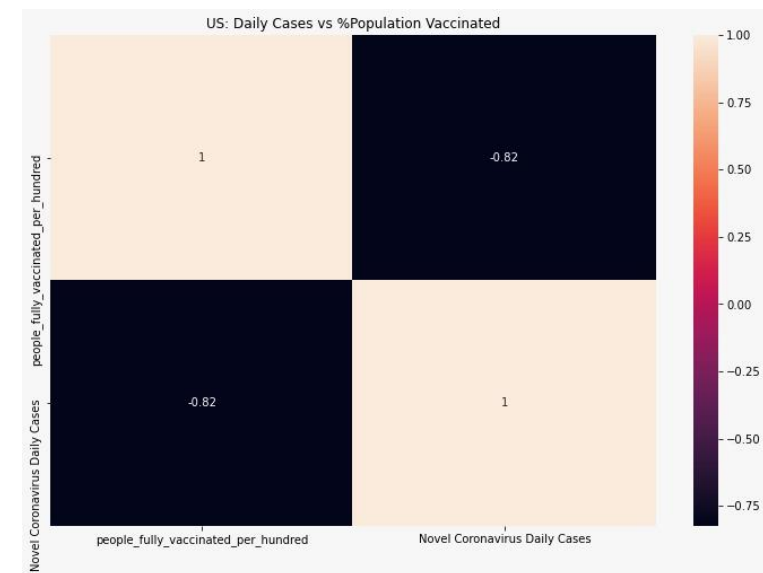
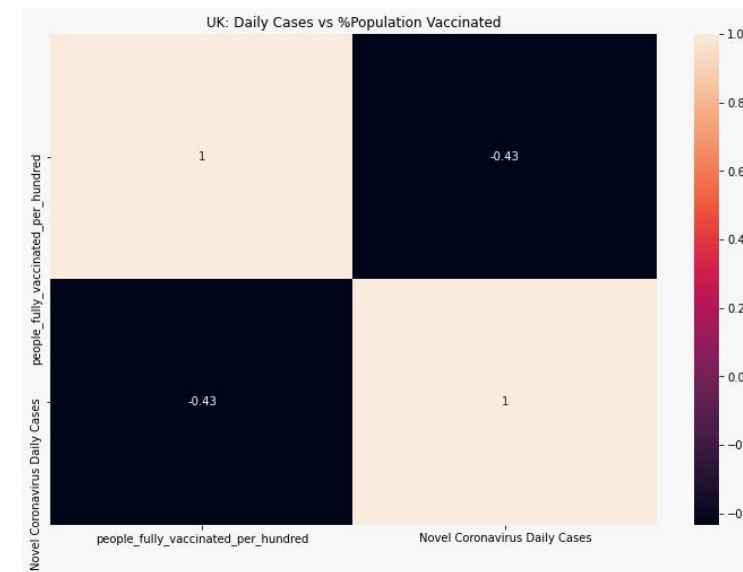
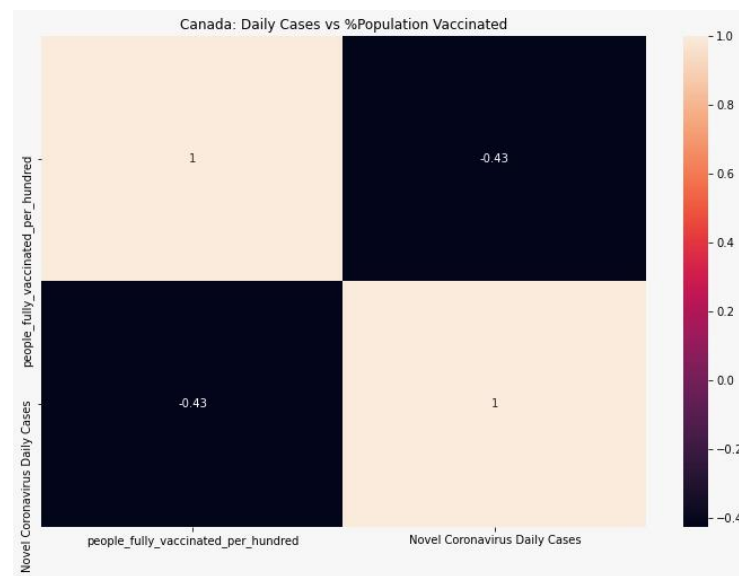
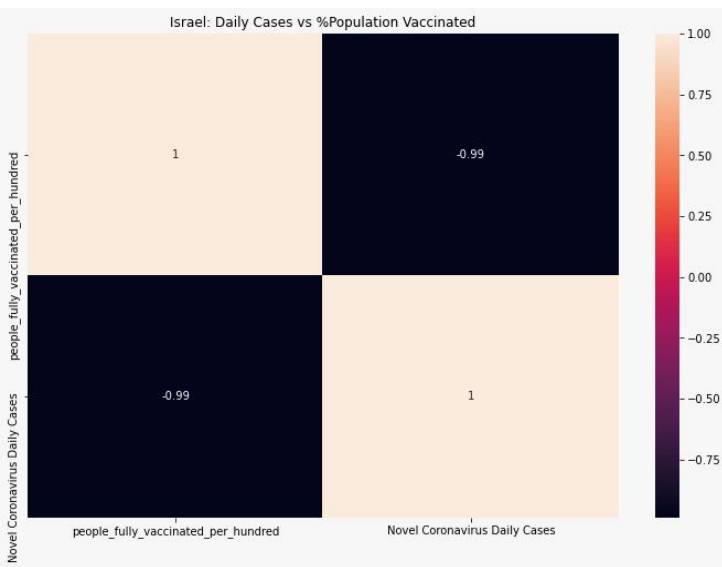
```
countries=["UK","US","Canada","Israel","Italy","Germany","Argentina"]
# Get Data only for Israel
for c in countries:
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    Vaccinated=rVaccinated.copy()
    Vaccinated.query("location=='%s'" %c, inplace = True)
    # Start Measuring from the time the vaccines started
    #Vaccinated.query("people_fully_vaccinated_per_hundred>0" , inplace = True)
    # Conver the Day to Datetime Index
    Vaccinated['date'] = pd.DatetimeIndex(Vaccinated['date'])

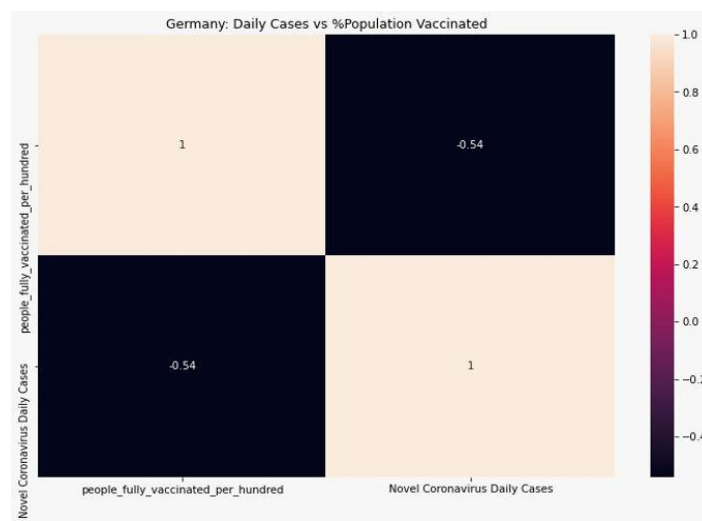
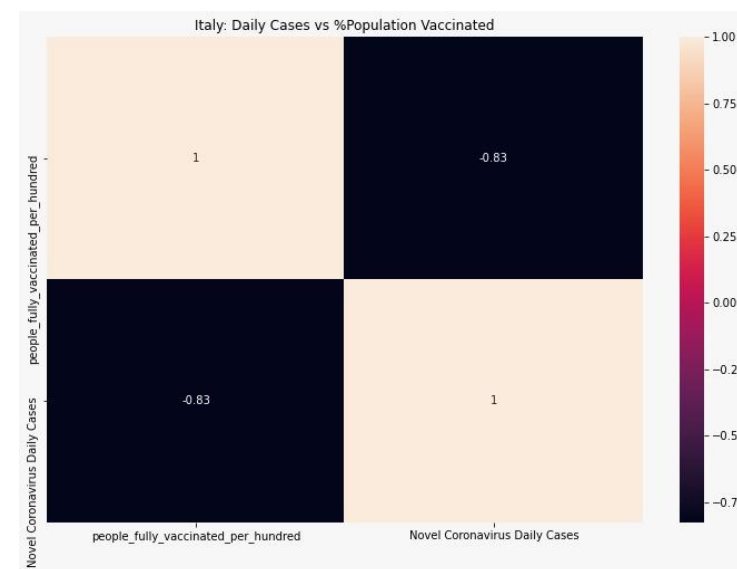
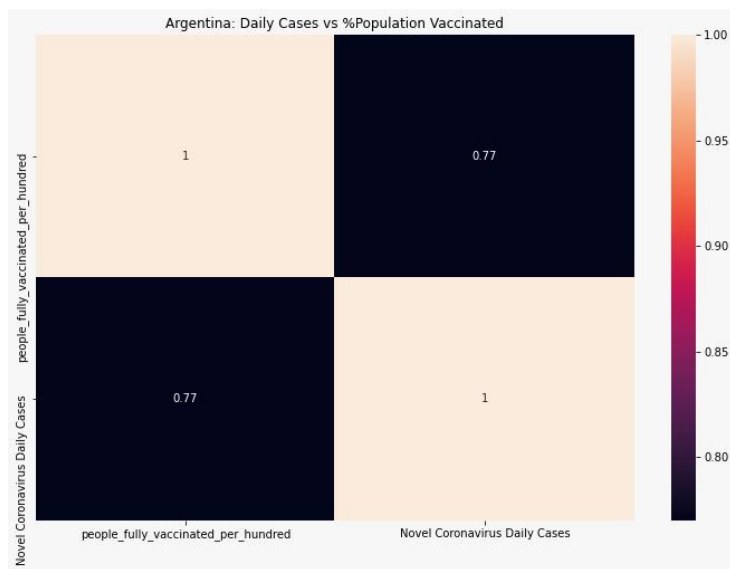
    # Join the data
    merged = KPI.merge(Vaccinated, how='inner', left_index=True, right_on='date')
    merged.index=merged.date
    merged.index.name = 'Date'

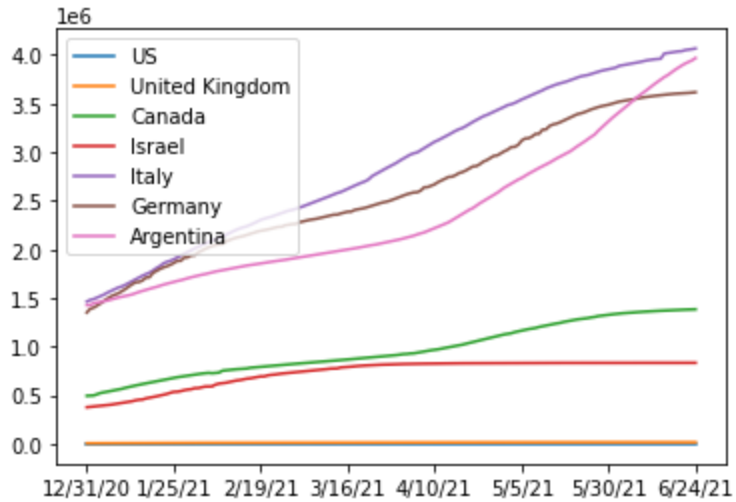
    rolling_merged = merged[['people_fully_vaccinated_per_hundred', 'Novel Coronavirus Daily
Cases']].rolling(7).mean().dropna()
    rolling_merged.plot.scatter('people_fully_vaccinated_per_hundred', 'Novel Coronavirus Daily Cases',
                               figsize=(12,8), title = c+": Daily Cases vs %Population Vaccinated")

corrMatrix = rolling_merged.corr()
sn.heatmap(corrMatrix, annot=True)
```

Correlation –to understand if there is a conection between the % of vaccinated to new cases
If it returns ' – '1there is a conection, if it returns 'noitalerroc on si ereht '0' snruter ti fi 'pihsnoitaler esrever '1–







We can see reverse connection in data from all the countries – as the % of vaccinated population rise, the number of new cases is gating lower

For instance, Israel is standing on 60% vaccinated, and we see a reverse connection 0.99

In USA there is 40% vaccinated' and the reverse connection is 0.82

In Italy, 25% vaccinated and 0.83 reverse connection (because of high number of cases)

While there is more then 4 million of people recovered, that are vaccinated naturally

On the other hand –Argentina has only 8% vaccinated, and so we can see that there is No connection between new cases and the % of vaccinated population



```

counter=0;
esum=0;
country_list=['US','Russia','Germany','France','UK','Japan','India','Italy','Israel','Canada','Spain','South
Korea','Switzerland','Netherlands','Argentina']
for c in country_list:
    KPI = covid_daily.data(country=c, chart = 'graph-cases-daily', as_json=False)
    Vaccinated=rVaccinated.copy()
    Vaccinated.query("location=='%s'" %c, inplace = True)
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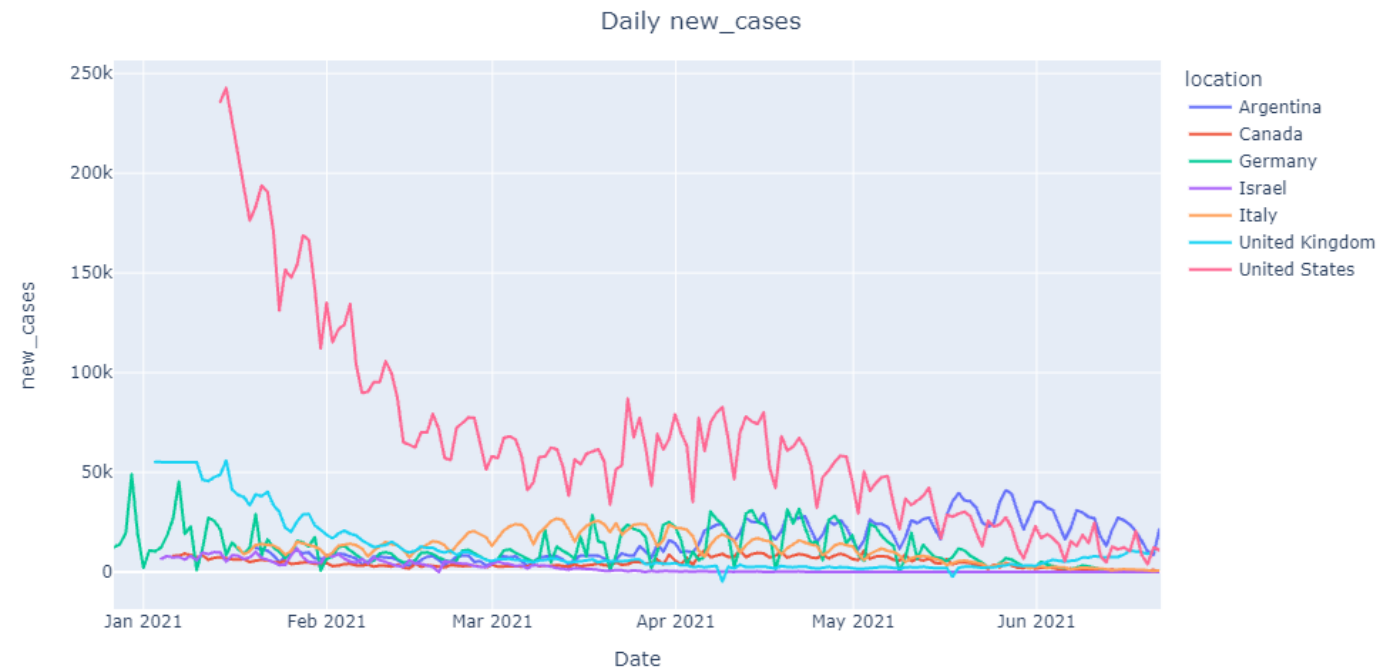
    rolling_merged = merged[['people_fully_vaccinated_per_hundred', 'Novel Coronavirus Daily
Cases']].rolling(7).mean().dropna()
    if(rolling_merged['people_fully_vaccinated_per_hundred'].corr(rolling_merged['Novel Coronavirus Daily
Cases'])<(-0.5)):
        esum+=rolling_merged['people_fully_vaccinated_per_hundred'].iloc[-1]
        counter+=1

print(esum/counter)

```

We ran the test on 15 countries to check what is the connection between % vaccinated and corelation that less then 0.5

The resoult showed that there is a reversed connection between decline of new cases and the % vaccinated
And the average is 33%



Conclusions:

There is a reversed connection between % vaccinated and new cases detected

This connection is intensified when % vaccinated is 33% or higher, then our assumption is working
To answer the question what is the minimum % vaccinated we need- to see decline in new cases, we can't say .

Because the vaccine came after many cases of people recovered or dead, after the winter has passed, and all the countries has been in embargo .

Therefore, we saw the decline in new cases from january, regardless vaccine.

