

IBM NAAN MUDHALVAN

PHASE4

DEVELOPMENT PHASE

COVID 19 VACCINE ANALYSIS

PROBLEM STATEMENT

Forecasting of time taken for completing 100% total vaccinations of particular region over the time period. By this, vaccine manufacturing companies get to know the prior requirements of vaccine which helps to produce the vaccines in large scale and complete the vaccination drive within calculated time.

1.Data Understanding

importing data and libraries

```
import pandas as pd
```

```
import seaborn as sns
```

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
import plotly.express as px
```

```
import io
```

```
import requests
```

```
import warnings
```

```
warnings.filterwarnings('ignore')
```

```
url = "https://raw.githubusercontent.com/owid/covid-19-data/master/public/data/owid-covid-data.csv"
```

```
read_data = requests.get(url).content
```

```
address = pd.read_csv(io.StringIO(read_data.decode('utf-8')))
```

```
address.head()
```

	country	iso_code	date	total_vaccinations	people_vaccinated	people_fully_vaccinated	daily_vaccinations_raw
0	Albania	ALB	2021-01-10	0.0	0.0	NaN	NaN
1	Albania	ALB	2021-01-11	NaN	NaN	NaN	NaN
2	Albania	ALB	2021-01-12	128.0	128.0	NaN	NaN
3	Albania	ALB	2021-01-13	188.0	188.0	NaN	60.0
4	Albania	ALB	2021-01-14	266.0	266.0	NaN	78.0

```
url2= "https://raw.githubusercontent.com/owid/covid-19-  
data/master/public/data/vaccinations/vaccinations-by-manufacturer.csv"
```

```
read_data = requests.get(url2).content
```

```
vaccine=pd.read_csv(io.StringIO(read_data.decode('utf-8')))
```

```
data=address
```

```
data.columns
```

```
Index(['iso_code', 'continent', 'location', 'date', 'total_cases', 'new_cases',  
      'new_cases_smoothed', 'total_deaths', 'new_deaths',  
      'new_deaths_smoothed', 'total_cases_per_million',  
      'new_cases_per_million', 'new_cases_smoothed_per_million',  
      'total_deaths_per_million', 'new_deaths_per_million',  
      'new_deaths_smoothed_per_million', 'reproduction_rate', 'icu_patients',  
      'icu_patients_per_million', 'hosp_patients',  
      'hosp_patients_per_million', 'weekly_icu_admissions',  
      'weekly_icu_admissions_per_million', 'weekly_hosp_admissions',  
      'weekly_hosp_admissions_per_million', 'total_tests', 'new_tests',  
      'total_tests_per_thousand', 'new_tests_per_thousand',  
      'new_tests_smoothed', 'new_tests_smoothed_per_thousand',  
      'positive_rate', 'tests_per_case', 'tests_units', 'total_vaccinations',  
      'people_vaccinated', 'people_fully_vaccinated', 'total_boosters',  
      'new_vaccinations', 'new_vaccinations_smoothed',
```

```

'total_vaccinations_per_hundred', 'people_vaccinated_per_hundred',
'people_fully_vaccinated_per_hundred', 'total_boosters_per_hundred',
'new_vaccinations_smoothed_per_million',
'new_people_vaccinated_smoothed',
'new_people_vaccinated_smoothed_per_hundred', 'stringency_index',
'population', 'population_density', 'median_age', 'aged_65_older',
'aged_70_older', 'gdp_per_capita', 'extreme_poverty',
'cardiovasc_death_rate', 'diabetes_prevalence', 'female_smokers',
'male_smokers', 'handwashing_facilities', 'hospital_beds_per_thousand',
'life_expectancy', 'human_development_index',
'excess_mortality_cumulative_absolute', 'excess_mortality_cumulative',
'excess_mortality', 'excess_mortality_cumulative_per_million'],
dtype='object')

```

```
data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 191376 entries, 0 to 191375
```

```
Data columns (total 67 columns):
```

#	Column	Non-Null Count	Dtype
0	iso_code	191376 non-null	object
1	continent	180250 non-null	object
2	location	191376 non-null	object
3	date	191376 non-null	object
4	total_cases	183834 non-null	float64
5	new_cases	183621 non-null	float64

6	new_cases_smoothed	182447 non-null float64
7	total_deaths	165368 non-null float64
8	new_deaths	165361 non-null float64
9	new_deaths_smoothed	164198 non-null float64
10	total_cases_per_million	182986 non-null float64
11	new_cases_per_million	182773 non-null float64
12	new_cases_smoothed_per_million	181604 non-null float64
13	total_deaths_per_million	164533 non-null float64
14	new_deaths_per_million	164526 non-null float64
15	new_deaths_smoothed_per_million	163368 non-null float64
16	reproduction_rate	140710 non-null float64
17	icu_patients	25496 non-null float64
18	icu_patients_per_million	25496 non-null float64
19	hosp_patients	26747 non-null float64
20	hosp_patients_per_million	26747 non-null float64
21	weekly_icu_admissions	6222 non-null float64
22	weekly_icu_admissions_per_million	6222 non-null float64
23	weekly_hosp_admissions	12397 non-null float64
24	weekly_hosp_admissions_per_million	12397 non-null float64
25	total_tests	77683 non-null float64
26	new_tests	74008 non-null float64
27	total_tests_per_thousand	77683 non-null float64
28	new_tests_per_thousand	74008 non-null float64
29	new_tests_smoothed	101315 non-null float64
30	new_tests_smoothed_per_thousand	101315 non-null float64

31	positive_rate	93441 non-null	float64
32	tests_per_case	91681 non-null	float64
33	tests_units	104079 non-null	object
34	total_vaccinations	52388 non-null	float64
35	people_vaccinated	49909 non-null	float64
36	people_fully_vaccinated	47375 non-null	float64
37	total_boosters	24452 non-null	float64
38	new_vaccinations	42912 non-null	float64
39	new_vaccinations_smoothed	103578 non-null	float64
40	total_vaccinations_per_hundred	52388 non-null	float64
41	people_vaccinated_per_hundred	49909 non-null	float64
42	people_fully_vaccinated_per_hundred	47375 non-null	float64
43	total_boosters_per_hundred	24452 non-null	float64
44	new_vaccinations_smoothed_per_million	103578 non-null	float64
45	new_people_vaccinated_smoothed	102491 non-null	float64
46	new_people_vaccinated_smoothed_per_hundred	102491 non-null	float64
47	stringency_index	148621 non-null	float64
48	population	190211 non-null	float64
49	population_density	170524 non-null	float64
50	median_age	158052 non-null	float64
51	aged_65_older	156377 non-null	float64
52	aged_70_older	157223 non-null	float64
53	gdp_per_capita	157205 non-null	float64
54	extreme_poverty	102625 non-null	float64
55	cardiovasc_death_rate	157692 non-null	float64

56	diabetes_prevalence	165401 non-null float64
57	female_smokers	119268 non-null float64
58	male_smokers	117633 non-null float64
59	handwashing_facilities	77477 non-null float64
60	hospital_beds_per_thousand	139914 non-null float64
61	life_expectancy	178964 non-null float64
62	human_development_index	153621 non-null float64
63	excess_mortality_cumulative_absolute	6553 non-null float64
64	excess_mortality_cumulative	6553 non-null float64
65	excess_mortality	6553 non-null float64
66	excess_mortality_cumulative_per_million	6553 non-null float64

dtypes: float64(62), object(5)

memory usage: 97.8+ MB

data.describe(include='all')

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 4568 entries, 0 to 4567
Data columns (total 15 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   country                               4568 non-null   object
1   iso_code                              4260 non-null   object
2   date                                  4568 non-null   object
3   total_vaccinations                    2988 non-null   float64
4   people_vaccinated                     2541 non-null   float64
5   people_fully_vaccinated                1702 non-null   float64
6   daily_vaccinations_raw                 2523 non-null   float64
7   daily_vaccinations                     4409 non-null   float64
8   total_vaccinations_per_hundred         2988 non-null   float64
9   people_vaccinated_per_hundred          2541 non-null   float64
10  people_fully_vaccinated_per_hundred     1702 non-null   float64
11  daily_vaccinations_per_million          4409 non-null   float64
12  vaccines                               4568 non-null   object
13  source_name                            4568 non-null   object
14  source_website                          4568 non-null   object
dtypes: float64(9), object(6)
memory usage: 535.4+ KB
```

```
vaccine.info()
```

```
<class 'pandas.core.frame.DataFrame'>
```

```
RangeIndex: 42395 entries, 0 to 42394
```

```
Data columns (total 4 columns):
```

#	Column	Non-Null Count	Dtype
0	location	42395 non-null	object
1	date	42395 non-null	object
2	vaccine	42395 non-null	object
3	total_vaccinations	42395 non-null	int64

```
dtypes: int64(1), object(3)
```

```
vaccine.describe()
```

```
total_vaccinations
```

```
count 4.239500e+04
```

```
mean 1.782378e+07
```

```
std 5.733925e+07
```

```
min 0.000000e+00
```

```
25% 1.075450e+05
```

```
50% 1.528400e+06
```

```
75% 9.792642e+06
```

```
max 6.141617e+08
```

```
2.data preprocessing
```

```
data.isnull().sum()
```

```
iso_code 0
```

```
continent 11126
```

```

location          0
date              0
total_cases       7542
...
human_development_index    37755
excess_mortality_cumulative_absolute    184823
excess_mortality_cumulative    184823
excess_mortality    184823
excess_mortality_cumulative_per_million    184823

```

Length: 67, dtype: int64

```
data['date']=pd.to_datetime(data['date'])
```

```
vaccine['date']=pd.to_datetime(data['date'])
```

```

data.drop(['new_cases_smoothed','new_deaths_smoothed', 'new_cases_smoothed_per_million',
          'new_deaths_smoothed_per_million', 'reproduction_rate', 'icu_patients',
          'new_tests_smoothed', 'new_tests_smoothed_per_thousand',
          'new_vaccinations_smoothed',
          'new_vaccinations_smoothed_per_million',
          'new_people_vaccinated_smoothed',
          'new_people_vaccinated_smoothed_per_hundred'], axis=1, inplace=True)

```

```
data.drop(['icu_patients_per_million','hosp_patients','hosp_patients_per_million','weekly_icu_admissions',
```

```

'weekly_icu_admissions_per_million','weekly_hosp_admissions','weekly_hosp_admissions_per_million',

```

```
'new_tests_per_thousand','excess_mortality_cumulative_absolute','excess_mortality_cumulative',
```



```
'excess_mortality','excess_mortality_cumulative_per_million','stringency_index','life_expectancy',  
human_development_index','extreme_poverty',
```

```
'cardiovasc_death_rate',
```

```
'diabetes_prevalence',
```

```
'female_smokers',
```

```
'male_smokers',
```

```
'handwashing_facilities',
```

```
'hospital_beds_per_thousand'],axis= 1,inplace=True)
```

checking for the null values

```
x=data.isnull().sum()*100/len(data)
```

x

iso_code	0.000000
continent	5.813686
location	0.000000
date	0.000000
total_cases	3.940933
new_cases	4.052232
total_deaths	13.590001
new_deaths	13.593659
total_cases_per_million	4.384040
new_cases_per_million	4.495339
total_deaths_per_million	14.026315
new_deaths_per_million	14.029972
total_tests	59.408181
new_tests	61.328484
total_tests_per_thousand	59.408181

positive_rate	51.174128
tests_per_case	52.093784
tests_units	45.615438
total_vaccinations	72.625617
people_vaccinated	73.920972
people_fully_vaccinated	75.245067
total_boosters	87.223058
new_vaccinations	77.577126
total_vaccinations_per_hundred	72.625617
people_vaccinated_per_hundred	73.920972
people_fully_vaccinated_per_hundred	75.245067
total_boosters_per_hundred	87.223058
population	0.608749
population_density	10.895828
median_age	17.412842
aged_65_older	18.288082
aged_70_older	17.846020
gdp_per_capita	17.855426

dtype: float64

checking for duplicate values

duplicate = data[data.duplicated()]

duplicate

iso_code	continent	location	date	total_cases	new_cases	total_deaths
new_deaths	total_cases_per_million		new_cases_per_million		...	
total_vaccinations_per_hundred		people_vaccinated_per_hundred				
people_fully_vaccinated_per_hundred			total_boosters_per_hundred	population		
population_density	median_age	aged_65_older	aged_70_older	gdp_per_capita		

0 rows × 33 columns

```
print(data.isnull().values.any())
```

True

```
data['total_deaths'].mean()
```

64774.858037830774

```
data['total_deaths'].median()
```

917.0

```
data['total_deaths'].replace(np.nan,data['total_deaths'].median()).head(10)
```

0 917.0

1 917.0

2 917.0

3 917.0

4 917.0

5 917.0

6 917.0

7 917.0

8 917.0

9 917.0

Name: total\_deaths, dtype: float64

using bfill method to fill nan cells

```
data.fillna(method="bfill")
```

```
Oxford/AstraZeneca 57
Moderna, Oxford/AstraZeneca, Pfizer/BioNTech 28
Oxford/AstraZeneca, Pfizer/BioNTech 13
Johnson&Johnson, Moderna, Oxford/AstraZeneca, Pfizer/BioNTech 12
Pfizer/BioNTech 12
Oxford/AstraZeneca, Sinopharm/Beijing 8
Sinopharm/Beijing 8
Sputnik V 8
Moderna, Pfizer/BioNTech 6
Oxford/AstraZeneca, Pfizer/BioNTech, Sinovac 6
Name: vaccines, dtype: int64
```

`data.isnull().values.any()` #Checking for nan values in whole dataframe

True

`data.head()`

	iso_code	continent	location	date	total_cases	new_cases	total_deaths	new_deaths	total_cases_per_million	new_cases_per_million	...	total_vaccinations_per_hundred	people_vaccinated_per_hundred	people_fully_vaccinated_per_hundred	total_boosters_per_hundred	population	population_density	median_age	aged_65_older	aged_70_older	gdp_per_capita
0	AFG	Asia	Afghanistan	2020-02-24	5.0	5.0	NaN	NaN	0.126	0.126											
	...	NaN	NaN	NaN	39835428.0	54.422	18.6	2.581	1.337												
					1803.987																
1	AFG	Asia	Afghanistan	2020-02-25	5.0	0.0	NaN	NaN	0.126	0.000											
	...	NaN	NaN	NaN	39835428.0	54.422	18.6	2.581	1.337												
					1803.987																
2	AFG	Asia	Afghanistan	2020-02-26	5.0	0.0	NaN	NaN	0.126	0.000											
	...	NaN	NaN	NaN	39835428.0	54.422	18.6	2.581	1.337												
					1803.987																
3	AFG	Asia	Afghanistan	2020-02-27	5.0	0.0	NaN	NaN	0.126	0.000											
	...	NaN	NaN	NaN	39835428.0	54.422	18.6	2.581	1.337												
					1803.987																
4	AFG	Asia	Afghanistan	2020-02-28	5.0	0.0	NaN	NaN	0.126	0.000											
	...	NaN	NaN	NaN	39835428.0	54.422	18.6	2.581	1.337												
					1803.987																

5 rows × 33 columns

`data.info()`

)

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 191376 entries, 0 to 191375

Data columns (total 33 columns):

#	Column	Non-Null Count	Dtype
0	iso_code	191376 non-null	object
1	continent	180250 non-null	object
2	location	191376 non-null	object
3	date	191376 non-null	datetime64[ns]
4	total_cases	183834 non-null	float64
5	new_cases	183621 non-null	float64
6	total_deaths	165368 non-null	float64
7	new_deaths	165361 non-null	float64
8	total_cases_per_million	182986 non-null	float64
9	new_cases_per_million	182773 non-null	float64
10	total_deaths_per_million	164533 non-null	float64
11	new_deaths_per_million	164526 non-null	float64
12	total_tests	77683 non-null	float64
13	new_tests	74008 non-null	float64
14	total_tests_per_thousand	77683 non-null	float64
15	positive_rate	93441 non-null	float64
16	tests_per_case	91681 non-null	float64
17	tests_units	104079 non-null	object
18	total_vaccinations	52388 non-null	float64
19	people_vaccinated	49909 non-null	float64

20	people_fully_vaccinated	47375 non-null	float64
21	total_boosters	24452 non-null	float64
22	new_vaccinations	42912 non-null	float64
23	total_vaccinations_per_hundred	52388 non-null	float64
24	people_vaccinated_per_hundred	49909 non-null	float64
25	people_fully_vaccinated_per_hundred	47375 non-null	float64
26	total_boosters_per_hundred	24452 non-null	float64
27	population	190211 non-null	float64
28	population_density	170524 non-null	float64
29	median_age	158052 non-null	float64
30	aged_65_older	156377 non-null	float64
31	aged_70_older	157223 non-null	float64
32	gdp_per_capita	157205 non-null	float64

dtypes: datetime64[ns](1), float64(28), object(4)

memory usage: 48.2+ MB

data.drop(['tests\_units'],axis=1,inplace=True)

null\_percentage=data.isna().sum()\*100/len(data)

null\_percentage.head(38)

iso_code	0.000000
continent	5.813686
location	0.000000
date	0.000000
total_cases	3.940933
new_cases	4.052232
total_deaths	13.590001

new_deaths	13.593659
total_cases_per_million	4.384040
new_cases_per_million	4.495339
total_deaths_per_million	14.026315
new_deaths_per_million	14.029972
total_tests	59.408181
new_tests	61.328484
total_tests_per_thousand	59.408181
positive_rate	51.174128
tests_per_case	52.093784
total_vaccinations	72.625617
people_vaccinated	73.920972
people_fully_vaccinated	75.245067
total_boosters	87.223058
new_vaccinations	77.577126
total_vaccinations_per_hundred	72.625617
people_vaccinated_per_hundred	73.920972
people_fully_vaccinated_per_hundred	75.245067
total_boosters_per_hundred	87.223058
population	0.608749
population_density	10.895828
median_age	17.412842
aged_65_older	18.288082
aged_70_older	17.846020
gdp_per_capita	17.855426

dtype: float64

data=data.fillna(method="bfill")

null\_percentage=data.isna().sum()\*100/len(data)

null\_percentage.head(38)

iso_code	0.000000
continent	0.000000
location	0.000000
date	0.000000
total_cases	0.000000
new_cases	0.000000
total_deaths	0.000000
new_deaths	0.000000
total_cases_per_million	0.000000
new_cases_per_million	0.000000
total_deaths_per_million	0.000000
new_deaths_per_million	0.000000
total_tests	0.000523
new_tests	0.007838
total_tests_per_thousand	0.000523
positive_rate	0.000523
tests_per_case	0.000523
total_vaccinations	0.001045
people_vaccinated	0.001045
people_fully_vaccinated	0.001045



```

total_boosters          0.001045
new_vaccinations         0.001045
total_vaccinations_per_hundred  0.001045
people_vaccinated_per_hundred  0.001045
people_fully_vaccinated_per_hundred  0.001045
total_boosters_per_hundred    0.001045
population               0.000000
population_density        0.000000
median_age               0.000000
aged_65_older           0.000000
aged_70_older           0.000000
gdp_per_capita          0.000000

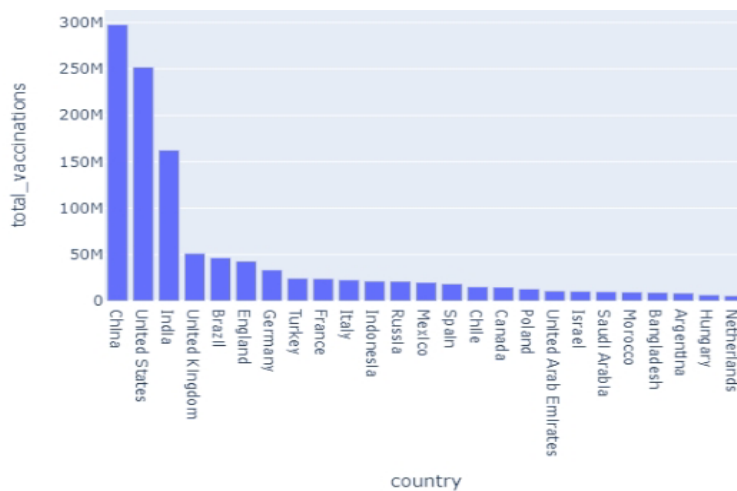
```

```
dtype: float64
```

```
data = new_df[['country','total_vaccinations']].nlargest(25,'total_vaccinations')
```

```
fig = px.bar(data, x = 'country',y = 'total_vaccinations',title="Number of total vaccinations according to countries",)
```

```
fig.show()
```



## Covid-19 Vaccination country wise

```
data = new_df[['country','daily_vaccinations']].nlargest(25,'daily_vaccinations')
```

```
fig = px.bar(data, x = 'country',y = 'daily_vaccinations',title="Number of daily vaccinations  
according to countries",)
```

```
fig.show()
```

