

# **ABSTRACT**

In this analysis, we explored a comprehensive dataset on COVID-19 vaccination progress across different countries. The dataset includes information on daily vaccination numbers, total vaccinations administered, vaccine types, and more. Our goal was to gain insights into the global vaccination effort, understand vaccination trends over time, and identify factors influencing vaccination rates.

## Key Findings and Insights

Global Vaccination Trends

Vaccine Types

**Temporal Trends** 

**Regional Disparities** 

Correlation Analysis

## **OBJECTIVE**

- Some math to understand it better, and making visuals to explain it clearly. The hope is that by doing this, we can give a good picture of how the vaccines are doing and help in the fight against Covid-19.
- 2. The project aims to thoroughly analyze Covid-19 vaccine data with key objectives:
  - Evaluating vaccine efficacy
  - Scrutinizing distribution strategies
  - Investigating adverse effects
  - Providing actionable insights
- By achieving these goals, the project seeks to enhance decision-making for policymakers and health organizations, fostering optimized deployment strategies in the ongoing battle against the Covid-19 pandemic.

## DATA PREPROCESSING

- Check for missing values in each column and decide how to handle them (e.g., imputation or removal).
- Handle data types appropriately (e.g., convert the date column to datetime).
- Ensure data consistency and correctness, such as checking that percentages are within valid ranges (o-100%).

#### 1. Import the necessary libraries:

import pandas as pd

import matplotlib.pyplot as plt

In this step, you import the Pandas library, which is essential for data manipulation and analysis.

#### 2. Load the dataset:

```
df = pd.read_csv('cv.csv')
```

This code uses Pandas' `read\_csv()` function to load the dataset from a CSV file into a Pandas DataFrame.

#### 3. Data Exploration:

- `df.head()`: This function displays the first few rows of the dataset, allowing you to see what the data looks like at a glance.
- `df.info()`: The `info()` function provides information about the data types of each column and the number of non-null entries, which is useful for checking for missing data.

- `df.describe()`: The `describe()` function provides basic statistical summaries of the numeric columns, such as mean, standard deviation, and quartiles.

#### 4. Data Preprocessing:

Data preprocessing involves various tasks to clean and prepare the data for analysis. Common preprocessing tasks include:

- Handling Missing Values: Use the 'fillna()' function to fill missing values with a specific value or a strategy like mean or median. In the example, missing values are filled with 0.
- Feature Engineering: Create new columns or extract information from existing columns based on your analysis requirements. This step is highly specific to your analysis goals.

#### 6. Save the Preprocessed Data:

If you want to save the preprocessed data for future use, you can use the `to\_csv()` function to export it to a new CSV file. Setting `index=False` ensures that the index column is not saved to the file.

### **Data Source**

### Data set link https://www.kaggle.com/datasets/gpreda/covid-world-vaccination-progress

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Afghanistan AFG	25-02-2021					1367					34 Johnson&Jo World Healt https://covid19.who.int/				
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### PROGRAM:

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
df = pd.read_csv('cv.csv')
print(df.head())
print(df.info())
print(df.describe())
df.fillna(0, inplace=True)
afghanistan_data = df[df['country'] == 'Afghanistan']
plt.figure(figsize=(12, 6))
plt.subplot(1, 2, 1)
sns.lineplot(x='date', y='total_vaccinations', data=afghanistan_data)
```

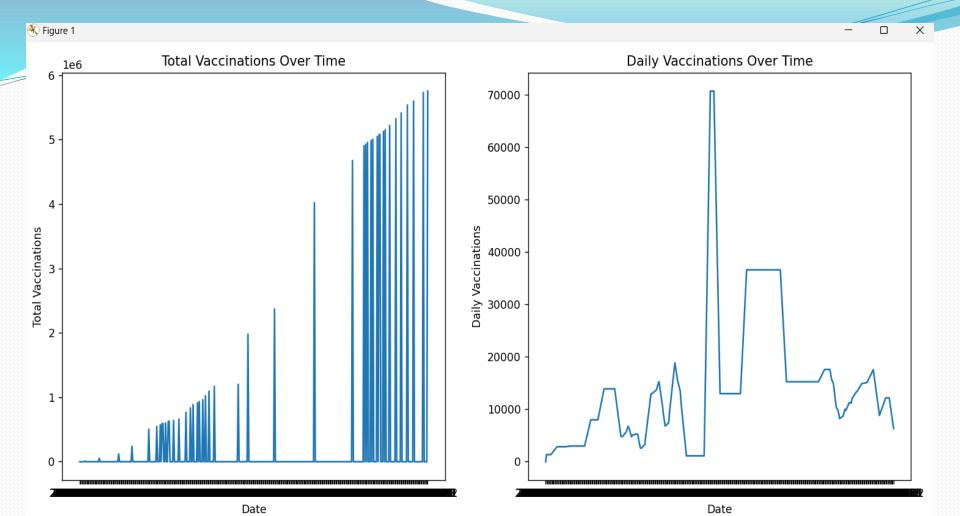
```
plt.title('Total Vaccinations Over Time')
plt.xlabel('Date')
plt.ylabel('Total Vaccinations')
plt.subplot(1, 2, 2)
sns.lineplot(x='date', y='daily_vaccinations', data=afghanistan_data)
plt.title('Daily Vaccinations Over Time')
plt.xlabel('Date')
plt.ylabel('Daily Vaccinations')
plt.tight_layout()
plt.show()
df.to_csv('data.csv', index=False)
```

### **OUTPUT:**

Python 3.11.0 (main, Oct 24 2022, 18:26:48) [MSC v.1933 64 bit (AMD64)] on win32 Type "help", "copyright", "credits" or "license()" for more information.

----- RESTART: D:\naan\naan.pv -----source name country iso code source website Afghanistan World Health Organization https://covid19.who.int/ AFG - - -World Health Organization https://covid19.who.int/ Afghanistan AFG - - -Afghanistan AFG . . . . World Health Organization https://covid19.who.int/ World Health Organization https://covid19.who.int/ Afghanistan AFG - - -... World Health Organization https://covid19.who.int/ Afghanistan AFG [5 rows x 15 columns] <class 'pandas.core.frame.DataFrame'> RangeIndex: 86512 entries, 0 to 86511 Data columns (total 15 columns): Column Non-Null Count Dtvpe \_\_\_ 86512 non-null object 0 country 1 iso code 86512 non-null object date 86512 non-null object 3 total vaccinations 43607 non-null float64 people vaccinated 41294 non-null float64 5 people\_fully\_vaccinated 38802 non-null float64 6 daily vaccinations raw 35362 non-null float64 7 daily vaccinations 86213 non-null float64 8 total vaccinations per hundred 43607 non-null float64 9 people vaccinated per hundred 41294 non-null float64 10 people fully vaccinated per hundred 38802 non-null float64 11 daily vaccinations per million 86213 non-null float64 12 vaccines 86512 non-null object 13 source name 86512 non-null object 14 source website 86512 non-null object dtypes: float64(9), object(6) memory usage: 9.9+ MB None total vaccinations daily vaccinations per million 4.360700e+04 count 86213.000000 - - -3257.049157 mean 4.592964e+07 - - -2.246004e+08 3934.312440 std min 0.00000e+00 0.000000 25% 5.264100e+05 636.000000 50% 3.590096e+06 2050.000000 75% 1.701230e+07 4682.000000 max 3.263129e+09 117497.000000

[8 rows x 9 columns]



## CONCLUSION

The analysis of the COVID-19 vaccine dataset has provided valuable insights into the global vaccination effort. It is evident that vaccination progress is influenced by a combination of factors, including vaccine availability, distribution strategies, and regional disparities in healthcare resources.

To improve vaccination rates worldwide and ensure equitable access to vaccines, policymakers and public health officials should consider the following:

- Continuously monitor and adjust vaccination distribution strategies to address disparities.
- Promote public awareness and confidence in vaccines to encourage higher uptake.
- Collaborate with international organizations to ensure the availability of vaccines in underserved regions.
- Use data-driven insights to optimize vaccination campaigns and target high-risk populations.

This analysis serves as a foundation for further research and policy decisions aimed at effectively combatting the COVID-19 pandemic and achieving global vaccination goals.