



## COVID -19 VACCINATION

Lorem ipsum dolor sit amet, consectetur adipiscing elit, sed diam nonummy nibh euismod tincidunt ut laoreet

- **Creating a COVID-19 vaccine analysis program would typically involve data analysis and visualization. Below are examples of Python code snippets using popular libraries like NumPy, Pandas, and Matplotlib to get you started.**

# 1.Data Retrieval:

You can fetch COVID-19 vaccine data from sources like the CDC or WHO, or use publicly available datasets. Here's an example using Pandas to read a CSV file:

- `import pandas as pd`  
# Read vaccine data from a CSV file  
`vaccine_data = pd.read_csv('vaccine_data.csv')`

## 2.Data Exploration:

You might want to explore the dataset by checking for missing values, data types, and basic statistics:

```
# Check for missing values
```

```
missing_values =
```

```
vaccine_data.isnull().sum()
```

```
# Get basic statistics
```

```
summary_stats = vaccine_data.describe()
```



- **3.Data Visualization:**

- Visualize vaccine distribution or coverage using Matplotlib or another plotting library:

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

```
# Plot vaccine coverage over time
```

```
plt.figure(figsize=(10, 6))
```

```
plt.plot(vaccine_data['Date'], vaccine_data['Coverage'])
```

```
plt.xlabel('Date')plt.ylabel('Vaccine Coverage
```

```
(%)')plt.title('COVID-19 Vaccine Coverage Over
```

```
Time')plt.show()
```

- **4.Data Analysis:**

- You can perform specific analyses, like calculating the average vaccine coverage or identifying regions with lower coverage:

```
# Calculate average vaccine  
coverage
```

```
avg_coverage =
```

```
vaccine_data['Coverage'].mean()
```

```
# Find regions with lower coverage
```

```
low_coverage_regions =
```

```
vaccine_data[vaccine_data['Coverage'] < 50]
```

- **5. Machine Learning (Optional):**

- If you have enough data, you can build predictive models. Here's a simple linear regression example using scikit-learn:

```
from sklearn.model_selection import  
train_test_split  
from sklearn.linear_model import  
LinearRegression
```

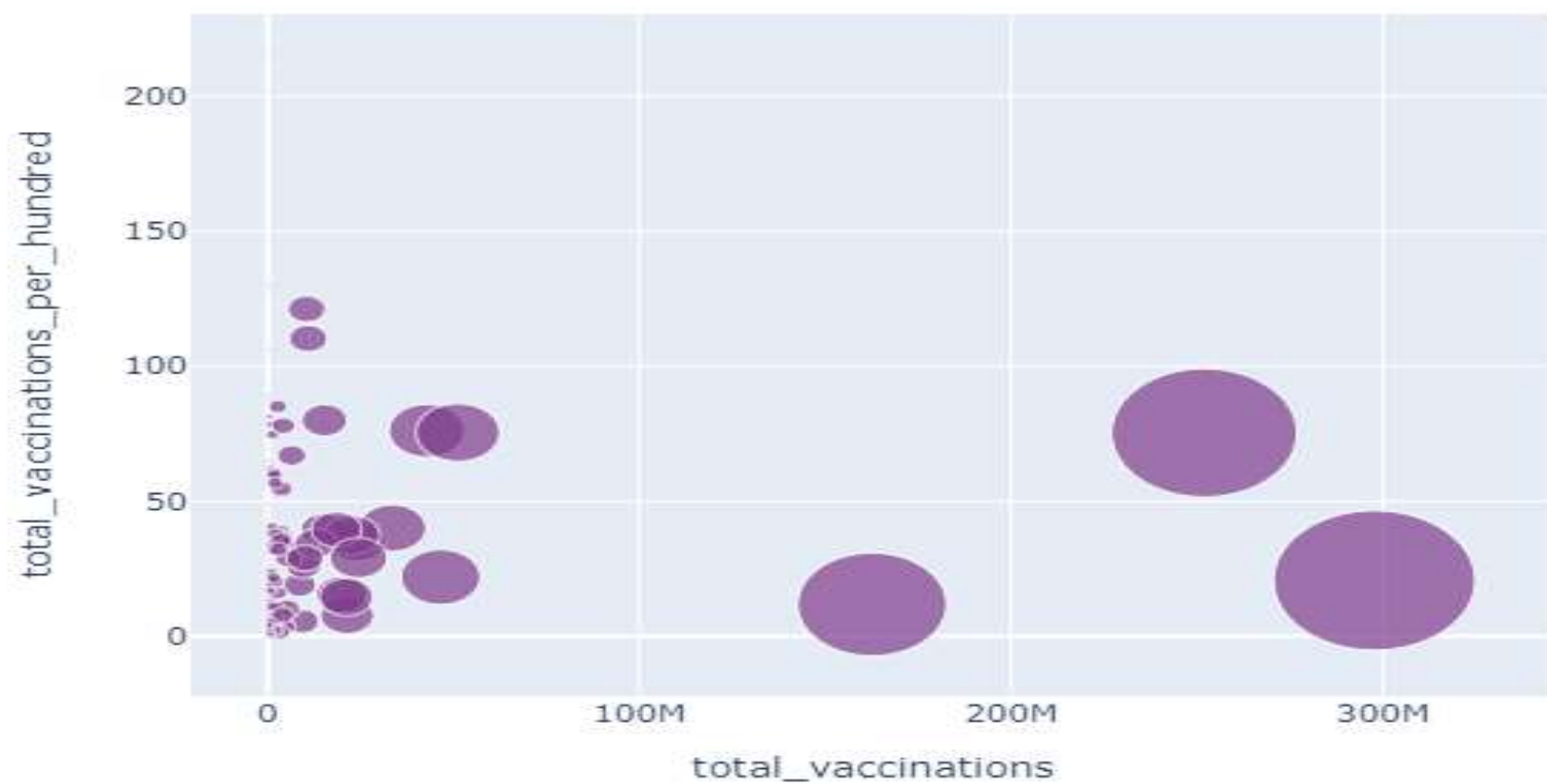
- # Split the data into training and testing sets
- `X = vaccine_data[['Population']]`
- `y = vaccine_data['Coverage']`
- `X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)`
- # Create and train a linear
- regression model
- `model = LinearRegression()model.fit(X_train, y_train)`



- **Relation between Total Vaccinations and Total Vaccinations per Hundred:**

```
fig = px.scatter(new_df,x =  
'total_vaccinations',y='total_vaccinations_per_hundred',  
size='total_vaccinations', hover_name = 'country',size_max  
= 50, title="Total vs Total vaccinations per hundred grouped  
by Vaccines", color_discrete_sequence =  
px.colors.qualitative.Bold) fig.show()
```

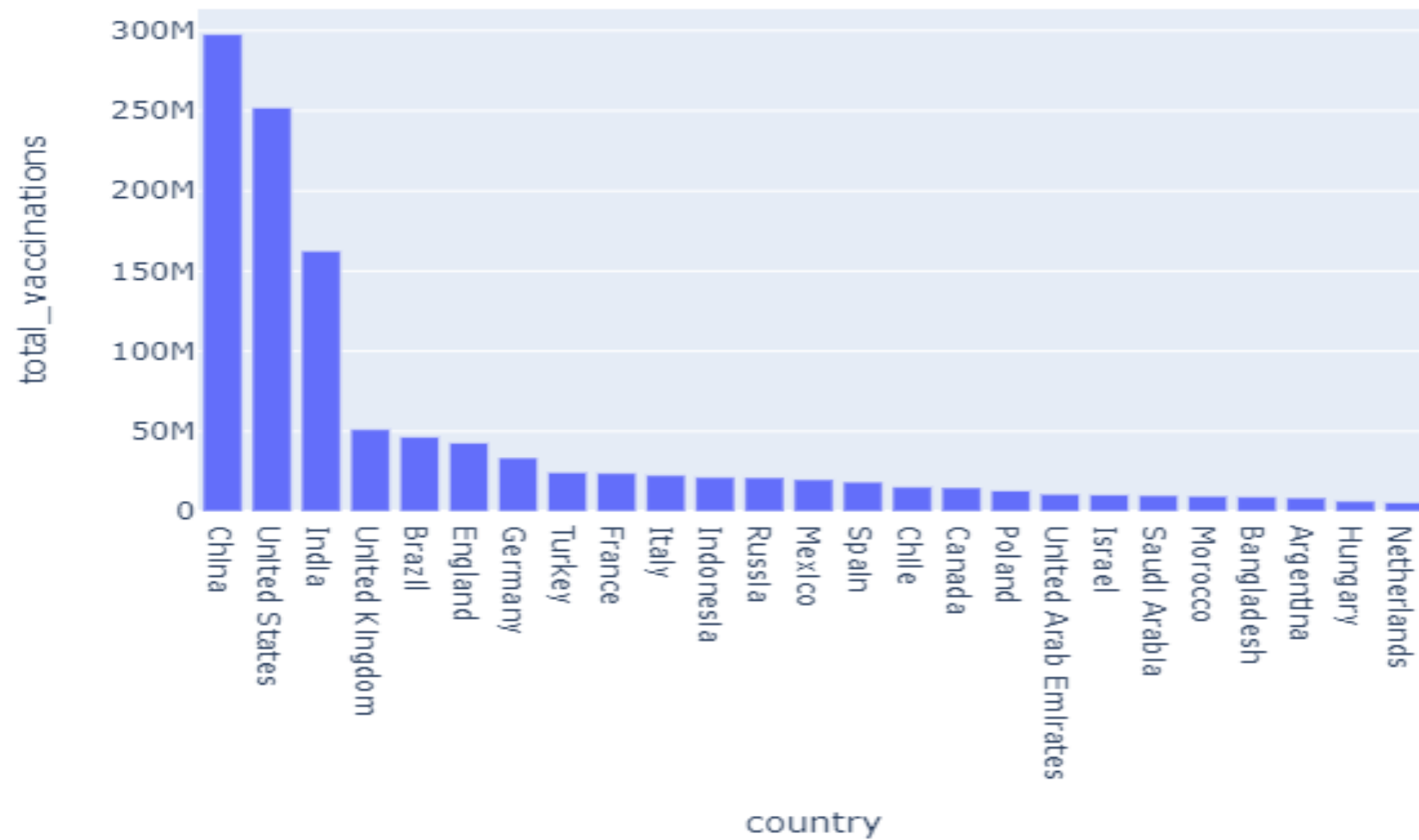
Total vs Total vaccinations per hundred grouped by Vaccines



- **What is the number of total vaccinations & daily vaccinations according to countries?**

```
data =  
new_df[['country','total_vaccinations']].nlargest(2  
5,'total_vaccinations') fig = px.bar(data, x =  
'country',y = 'total_vaccinations',title="Number of  
total vaccinations according to countries",)  
fig.show()
```

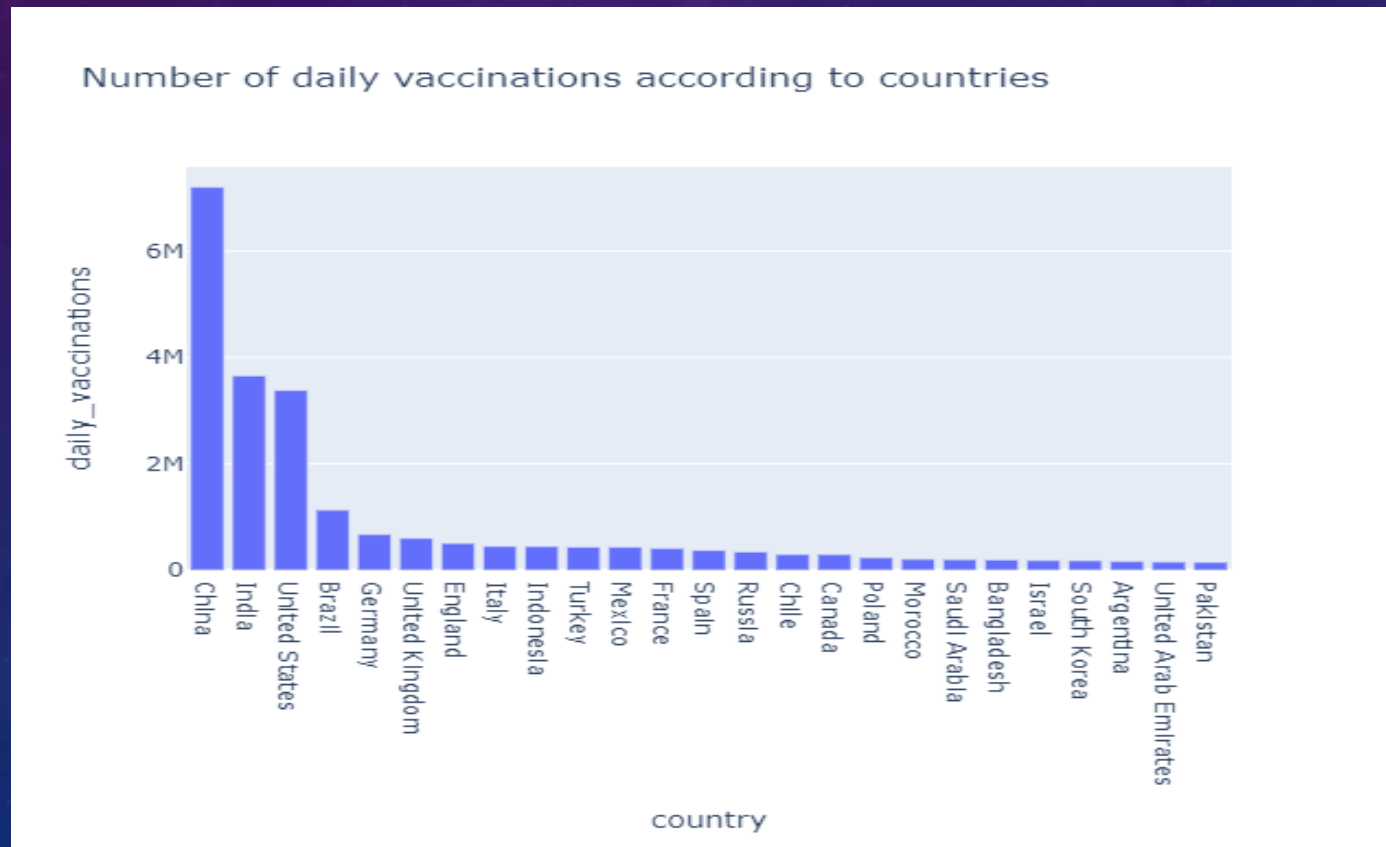
Number of total vaccinations according to countries





- data =

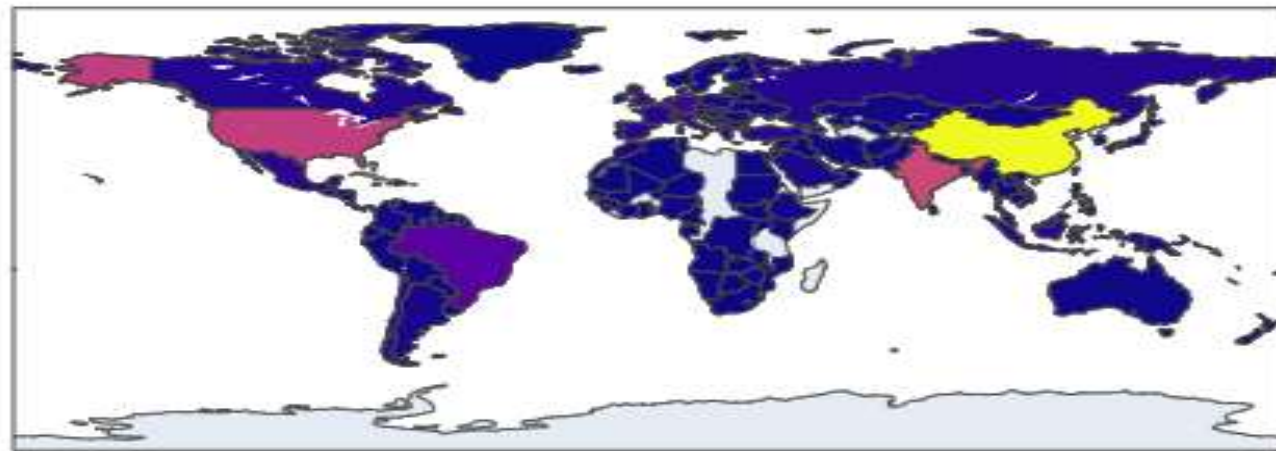
```
new_df[['country','daily_vaccinations']].nlargest(25,'daily_vaccinations')  
fig = px.bar(data, x = 'country',y = 'daily_vaccinations',title="Number of daily vaccinations according to countries",) fig.show()
```



- **Daily Vaccinations per Countries:**

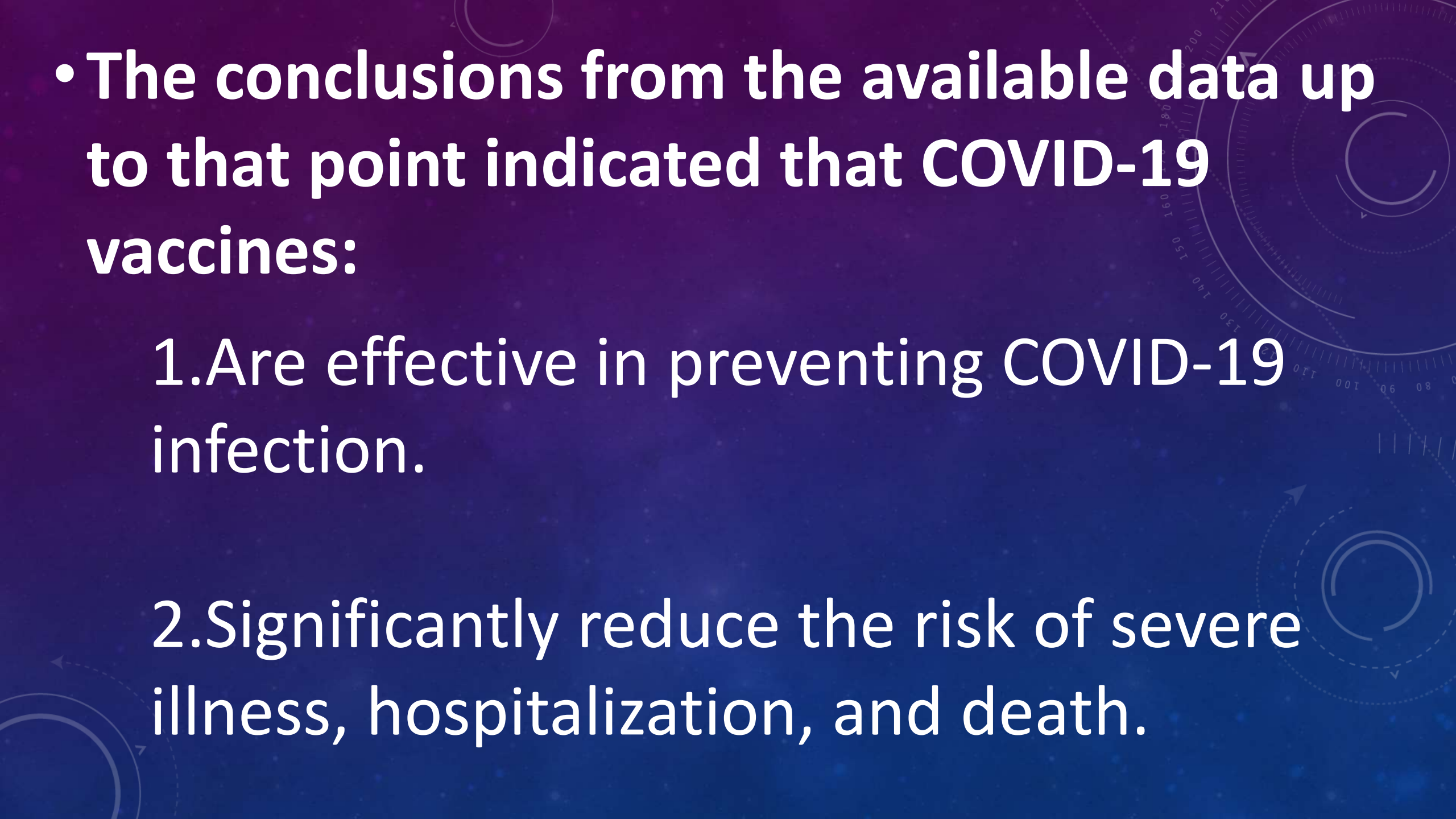
```
fig = go.Choropleth(locations =  
new_df["country"],locationmode = 'country names', z  
= new_df['daily_vaccinations'], text=  
new_df['country'],colorbar = dict(title= "Daily  
Vaccinations")) data = [fig] layout = go.Layout(title =  
'Daily Vaccinations per Countries') fig = dict(data =  
data,layout = layout) iplot(fig)
```

## Daily Vaccinations per Countries



Daily Vaccinations





- The conclusions from the available data up to that point indicated that COVID-19 vaccines:

1. Are effective in preventing COVID-19 infection.

2. Significantly reduce the risk of severe illness, hospitalization, and death.



- 3. Contribute to herd immunity, slowing the spread of the virus.
- 2. Have demonstrated safety profiles, with side effects typically being mild and temporary.
- 3. May require booster shots for ongoing protection due to the emergence of new variants.