

Project – COVID-19 Analysis

Problem Statement:

Given data about COVID-19 patients, write code to visualize the impact and analyze the trend of rate of infection and recovery as well as make predictions about the number of cases expected a week in future based on the current trends.

Dataset:

CSV and Excel files containing data about the number of COVID-19 confirmed deaths and recovered patients both around the world and in India.

Guidelines:

- Use pandas to accumulate data from multiple data files.
- Use plotly (visualization library) to create interactive visualizations.
- Use Facebook prophet library to make time series models.
- Visualize the prediction by combining these technologies.

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import matplotlib
import seaborn as sns
import plotly.express as px
from plotly.subplots import make_subplots
import plotly.graph_objects as go
from prophet import Prophet
from prophet.plot import plot_plotly, plot_components_plotly
```

```
In [2]: import plotly.io as pio
pio.renderers.default = "notebook+pdf"
```

```
In [3]: df = pd.read_csv("covid_19_clean_complete_1_1.csv")
df
```

```
Out[3]:
```

| | Province/State | Country/Region | Lat | Long | Date | Confirmed | Deaths | Recovered | Active | WHO Region |
|-------|----------------|-----------------------|------------|-----------|------------|-----------|--------|-----------|--------|-----------------------|
| 0 | NaN | Afghanistan | 33.939110 | 67.709953 | 2020-01-22 | 0 | 0 | 0 | 0 | Eastern Mediterranean |
| 1 | NaN | Albania | 41.153300 | 20.168300 | 2020-01-22 | 0 | 0 | 0 | 0 | Europe |
| 2 | NaN | Algeria | 28.033900 | 1.699600 | 2020-01-22 | 0 | 0 | 0 | 0 | Africa |
| 3 | NaN | Andorra | 42.506300 | 1.521800 | 2020-01-22 | 0 | 0 | 0 | 0 | Europe |
| 4 | NaN | Angola | -11.202700 | 17.873900 | 2020-01-22 | 0 | 0 | 0 | 0 | Africa |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 49063 | NaN | Sao Tome and Principe | 0.186400 | 6.613100 | 2020-07-27 | 865 | 14 | 734 | 117 | Africa |
| 49064 | NaN | Yemen | 15.552727 | 48.516388 | 2020-07-27 | 1691 | 483 | 833 | 375 | Eastern Mediterranean |
| 49065 | NaN | Comoros | -11.645500 | 43.333300 | 2020-07-27 | 354 | 7 | 328 | 19 | Africa |
| 49066 | NaN | Tajikistan | 38.861000 | 71.276100 | 2020-07-27 | 7235 | 60 | 6028 | 1147 | Europe |
| 49067 | NaN | Lesotho | -29.610000 | 28.233600 | 2020-07-27 | 505 | 12 | 128 | 365 | Africa |

49068 rows × 10 columns

```
In [4]: df['Date'] = pd.to_datetime(df['Date'], format='%Y-%m-%d')
```

```
In [5]: df.drop('Province/State', axis=1, inplace=True)
```

```
In [6]: df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 49068 entries, 0 to 49067
Data columns (total 9 columns):
 #   Column      Non-Null Count  Dtype
---  --
 0   Country/Region  49068 non-null   object
 1   Lat           49068 non-null   float64
 2   Long          49068 non-null   float64
 3   Date          49068 non-null   datetime64[ns]
 4   Confirmed     49068 non-null   int64
 5   Deaths       49068 non-null   int64
 6   Recovered     49068 non-null   int64
 7   Active        49068 non-null   int64
 8   WHO Region    49068 non-null   object
dtypes: datetime64[ns](1), float64(2), int64(4), object(2)
memory usage: 3.4+ MB
```

```
In [7]: df
Out[7]:
```

| | Country/Region | Lat | Long | Date | Confirmed | Deaths | Recovered | Active | WHO Region |
|-------|-----------------------|------------|-----------|------------|-----------|--------|-----------|--------|-----------------------|
| 0 | Afghanistan | 33.939110 | 67.709953 | 2020-01-22 | 0 | 0 | 0 | 0 | Eastern Mediterranean |
| 1 | Albania | 41.153300 | 20.168300 | 2020-01-22 | 0 | 0 | 0 | 0 | Europe |
| 2 | Algeria | 28.033900 | 1.699600 | 2020-01-22 | 0 | 0 | 0 | 0 | Africa |
| 3 | Andorra | 42.506300 | 1.521800 | 2020-01-22 | 0 | 0 | 0 | 0 | Europe |
| 4 | Angola | -11.202700 | 17.873900 | 2020-01-22 | 0 | 0 | 0 | 0 | Africa |
| ... | ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 49063 | Sao Tome and Principe | 0.186400 | 6.613100 | 2020-07-27 | 865 | 14 | 734 | 117 | Africa |
| 49064 | Yemen | 15.552727 | 48.516388 | 2020-07-27 | 1691 | 483 | 833 | 375 | Eastern Mediterranean |
| 49065 | Comoros | -11.645500 | 43.333300 | 2020-07-27 | 354 | 7 | 328 | 19 | Africa |
| 49066 | Tajikistan | 38.861000 | 71.276100 | 2020-07-27 | 7235 | 60 | 6028 | 1147 | Europe |
| 49067 | Lesotho | -29.610000 | 28.233600 | 2020-07-27 | 505 | 12 | 128 | 365 | Africa |

49068 rows × 9 columns

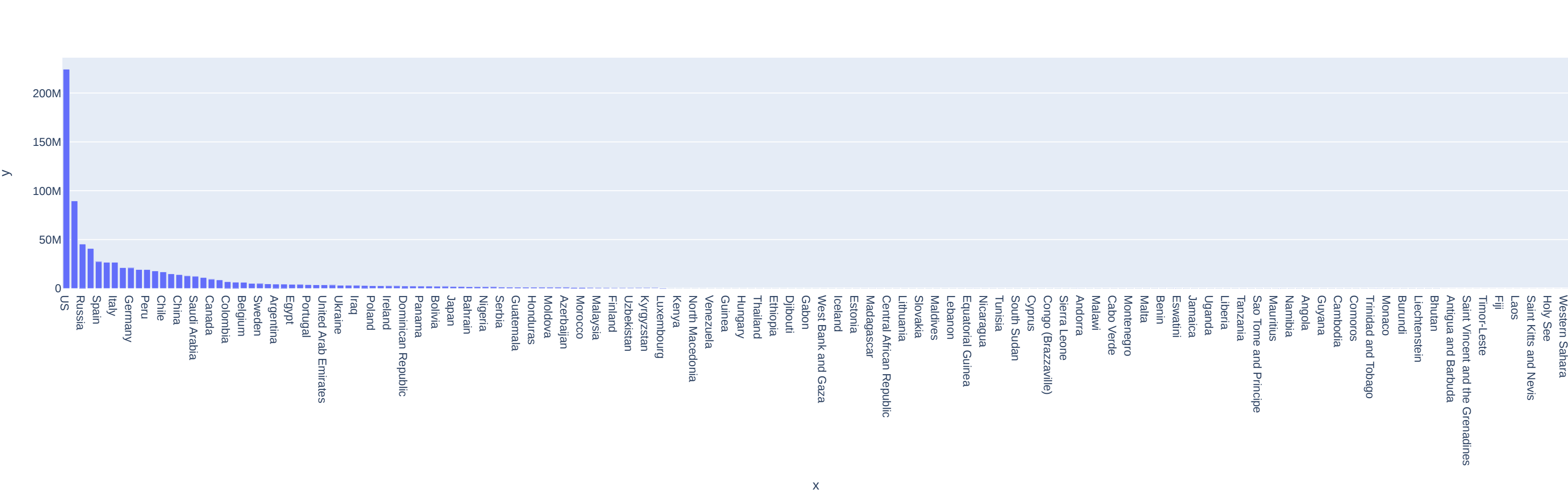
```
In [8]: df_agg_country_sorted = df.groupby('Country/Region').agg(
    Total_Confirmed=('Confirmed', 'sum'),
    Total_Deaths=('Deaths', 'sum'),
    Total_Recovered=('Recovered', 'sum'),
    Total_Active=('Active', 'sum')).sort_values('Total_Confirmed', ascending=False).reset_index()
df_agg_country_sorted
```

```
Out[8]:
```

| | Country/Region | Total_Confirmed | Total_Deaths | Total_Recovered | Total_Active |
|-----|-----------------------|-----------------|--------------|-----------------|--------------|
| 0 | US | 22434948 | 11011411 | 56353416 | 156981121 |
| 1 | Brazil | 89524967 | 3938034 | 54492873 | 31094080 |
| 2 | Russia | 45408411 | 615355 | 25120448 | 19668578 |
| 3 | India | 40883464 | 1111831 | 23783720 | 15987913 |
| 4 | Spain | 27404045 | 3033030 | 15069583 | 9277432 |
| ... | ... | ... | ... | ... | ... |
| 182 | Saint Kitts and Nevis | 1772 | 0 | 1296 | 477 |
| 183 | Greenland | 1507 | 0 | 1372 | 135 |
| 184 | Holy See | 1356 | 0 | 742 | 614 |
| 185 | Papua New Guinea | 1185 | 2 | 695 | 488 |
| 186 | Western Sahara | 901 | 63 | 648 | 190 |

187 rows × 5 columns

```
In [9]: fig = px.bar(x=df_agg_country_sorted['Country/Region'], y=df_agg_country_sorted['Total_Confirmed'])
fig.show()
```

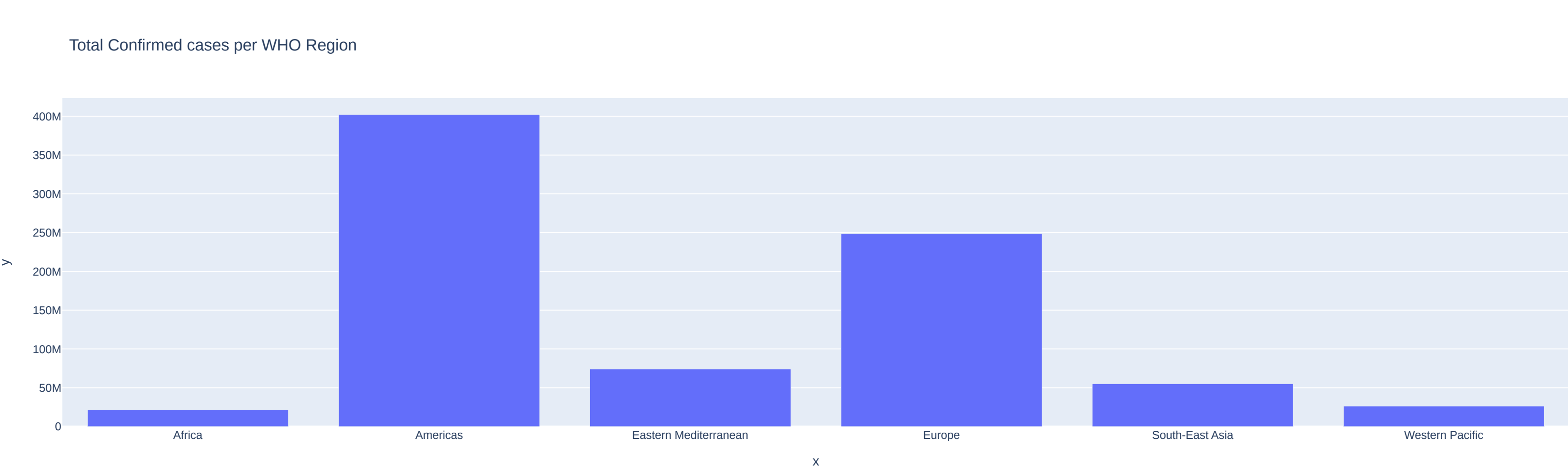


```
In [10]: df_agg_who = df.groupby('WHO Region').agg(
    Total_Confirmed=('Confirmed', 'sum'),
    Total_Deaths=('Deaths', 'sum'),
    Total_Recovered=('Recovered', 'sum'),
    Total_Active=('Active', 'sum')).reset_index()
df_agg_who
```

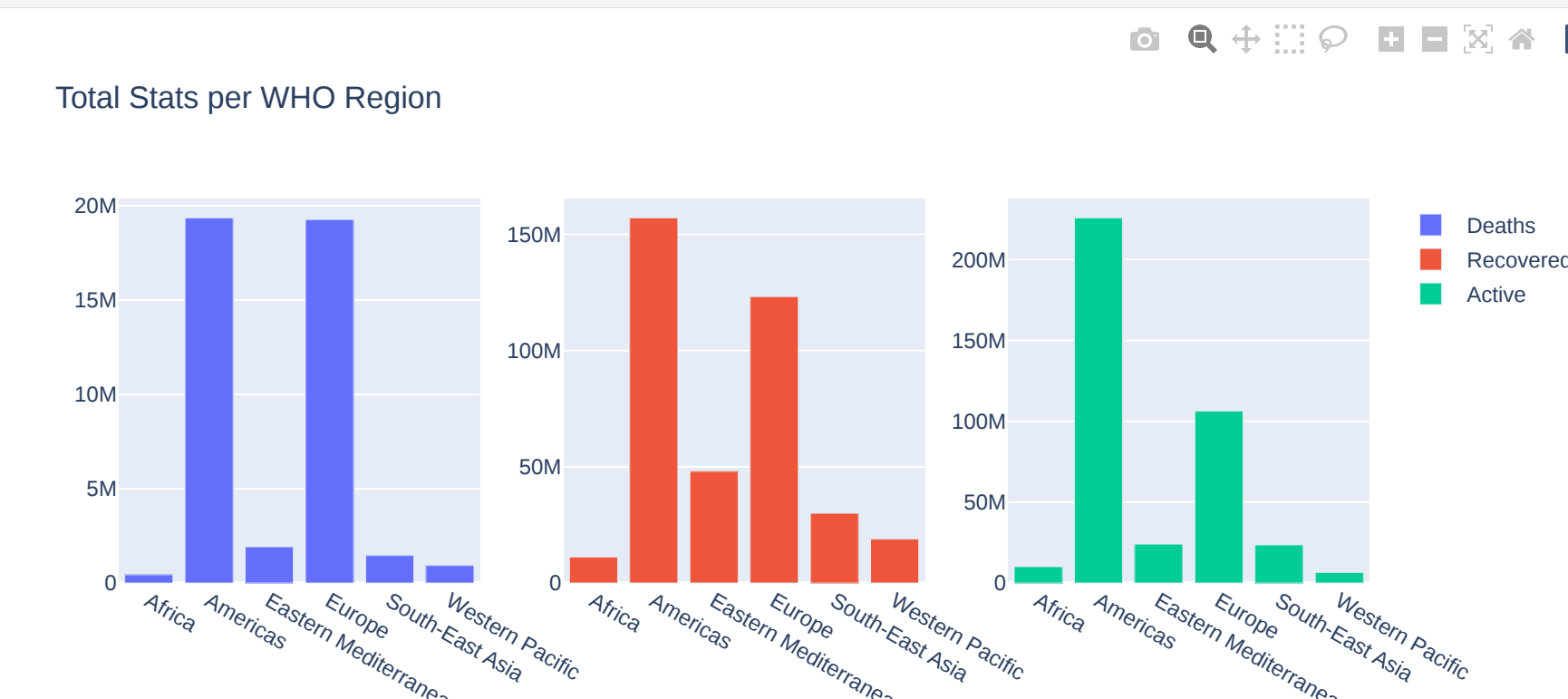
```
Out[10]:
```

| | WHO Region | Total_Confirmed | Total_Deaths | Total_Recovered | Total_Active |
|---|-----------------------|-----------------|--------------|-----------------|--------------|
| 0 | Africa | 21791827 | 439978 | 11193730 | 10158119 |
| 1 | Americas | 402261194 | 19359292 | 157069444 | 225832458 |
| 2 | Eastern Mediterranean | 74082892 | 1924029 | 48050703 | 24108160 |
| 3 | Europe | 248879793 | 19271040 | 123202075 | 106406678 |
| 4 | South-East Asia | 55118365 | 1458134 | 30030327 | 23629904 |
| 5 | Western Pacific | 26374411 | 932430 | 16861950 | 6580031 |

```
In [11]: fig = px.bar(x=df_agg_who['WHO Region'], y=df_agg_who['Total_Confirmed'], title='Total Confirmed cases per WHO Region')
fig.show()
```



```
In [12]: fig = make_subplots(rows=1, cols=3)
fig.add_trace(
    go.Bar(x=df_agg_who['WHO Region'], y=df_agg_who['Total_Deaths'], name='Deaths',
    row=1, col=1)
)
fig.add_trace(
    go.Bar(x=df_agg_who['WHO Region'], y=df_agg_who['Total_Recovered'], name='Recovered',
    row=1, col=2)
)
fig.add_trace(
    go.Bar(x=df_agg_who['WHO Region'], y=df_agg_who['Total_Active'], name='Active',
    row=1, col=3)
)
fig.update_layout(height=498, width=999, title_text='Total Stats per WHO Region')
fig.show()
```



```
In [13]: df_Prophet = pd.date_range(
df_Prophet['ds'] = df[['Date']]
df_Prophet['y'] = df[['Confirmed']]
```

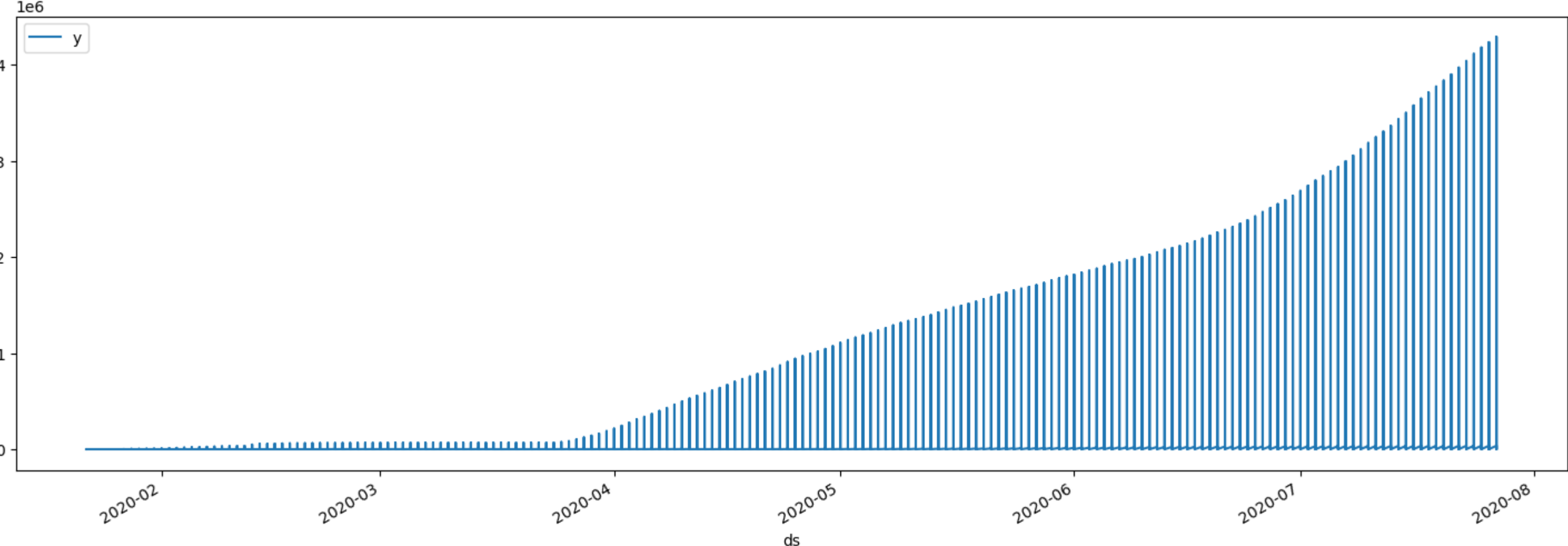
```
In [14]: df_Prophet.tail()
```

```
Out[14]:
```

| | ds | y |
|-------|------------|------|
| 49063 | 2020-07-27 | 865 |
| 49064 | 2020-07-27 | 1691 |
| 49065 | 2020-07-27 | 354 |
| 49066 | 2020-07-27 | 7235 |
| 49067 | 2020-07-27 | 505 |

```
In [15]: df_Prophet.plot(x='ds', y='y', figsize=(18,6))
```

```
Out[15]: <Axes: xlabel='ds'>
```



```
In [16]: m = Prophet()
m.fit(df_Prophet)
18:48:42 - cmdstanpy - INFO - Chain [1] start processing
18:48:52 - cmdstanpy - INFO - Chain [1] done processing
```

```
Out[16]: <prophet.forecaster.Prophet at 8x2266188180>
```

```
In [17]: future = m.make_future_dataframe(periods=60)
future.tail()
```

```
Out[17]:
```

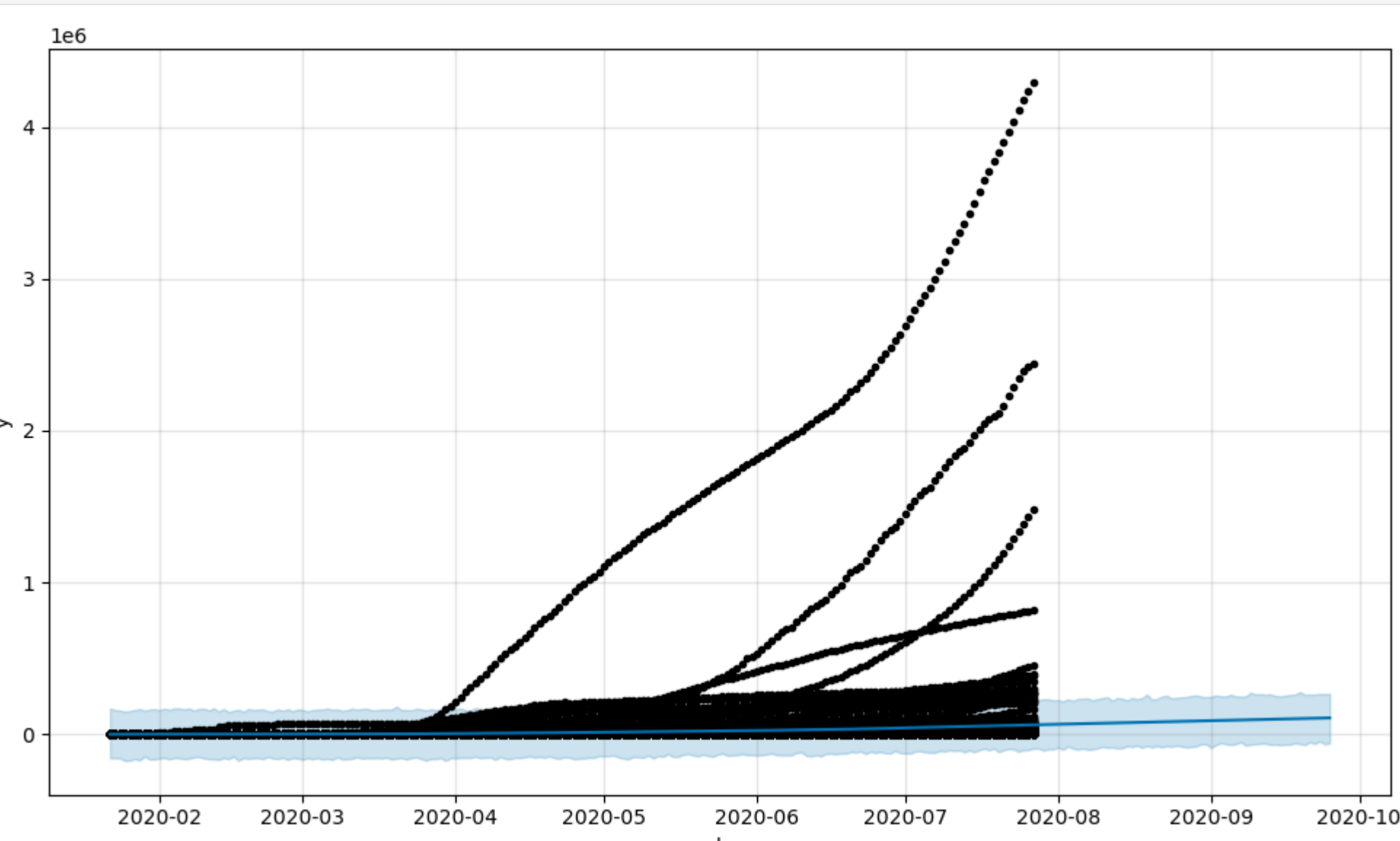
| | ds |
|-----|------------|
| 243 | 2020-09-21 |
| 244 | 2020-09-22 |
| 245 | 2020-09-23 |
| 246 | 2020-09-24 |
| 247 | 2020-09-25 |

```
In [18]: forecast = m.predict(future)
forecast[['ds', 'yhat', 'yhat_lower', 'yhat_upper']].tail()
```

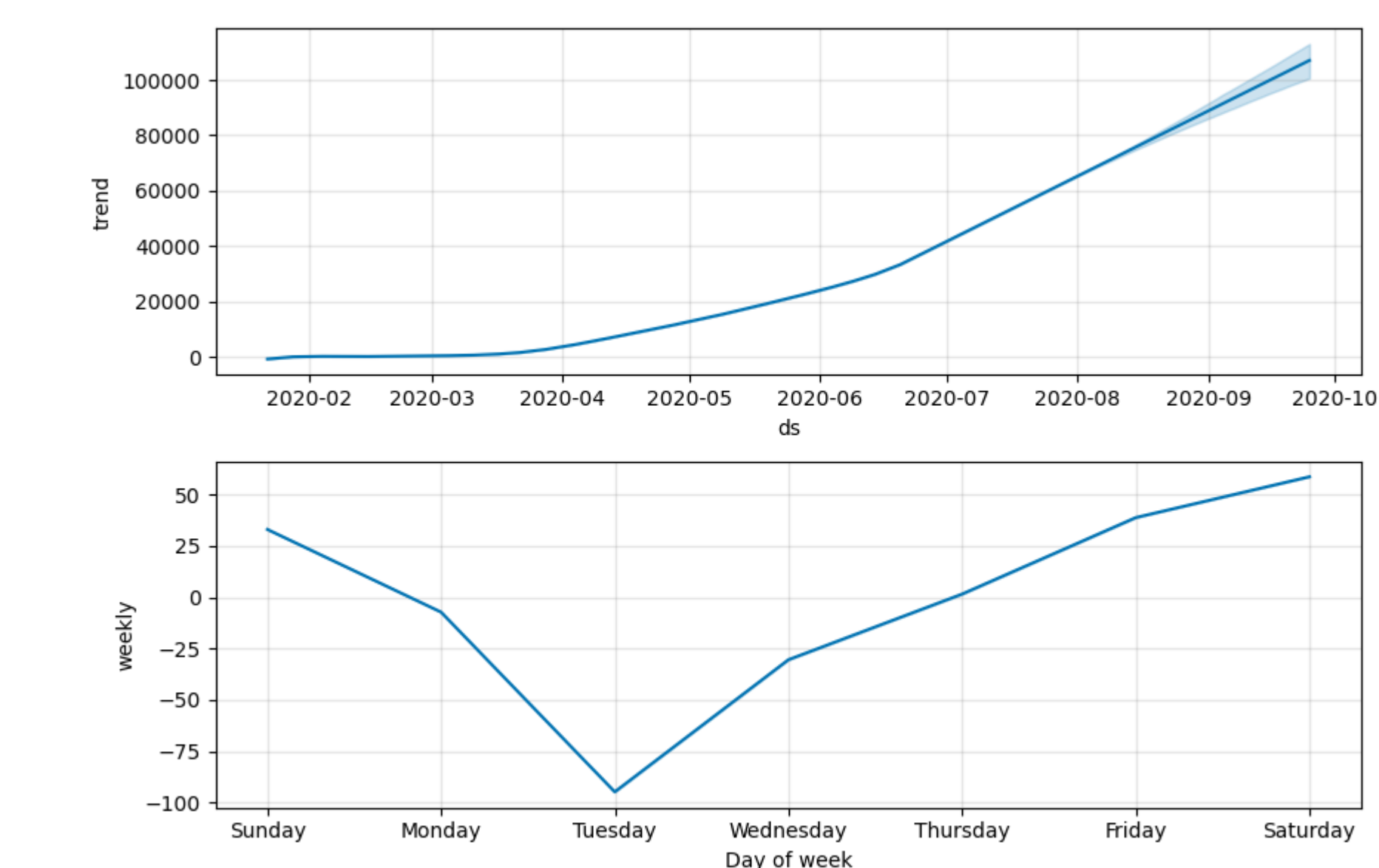
```
Out[18]:
```

| | ds | yhat | yhat_lower | yhat_upper |
|-----|------------|---------------|---------------|---------------|
| 243 | 2020-09-21 | 104009.084667 | -51671.433436 | 261362.803415 |
| 244 | 2020-09-22 | 104681.586670 | -42774.926124 | 261272.054124 |
| 245 | 2020-09-23 | 105506.136445 | -47348.394848 | 265736.526637 |
| 246 | 2020-09-24 | 106298.264285 | -65388.935765 | 265503.745564 |
| 247 | 2020-09-25 | 107095.668541 | -58267.545211 | 266258.406038 |

```
In [19]: fig1 = m.plot(forecast)
```



```
In [20]: fig2 = m.plot_components(forecast)
```



In []: