

```
In [3]: import pandas as pd
```

```
In [5]: df=pd.read_csv(r"C:\Users\Pooja Shinde\Downloads\covid_19_data.csv")
```

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

```
Out[7]:
```

| | SNo | ObservationDate | Province/State | Country/Region | Last Update | Confirmed | Deaths | Recovered |
|--------|--------|-----------------|-------------------|----------------|---------------------|-----------|--------|-----------|
| 0 | 1 | 01/22/2020 | Anhui | Mainland China | 1/22/2020 17:00 | 1.0 | 0.0 | 0.0 |
| 1 | 2 | 01/22/2020 | Beijing | Mainland China | 1/22/2020 17:00 | 14.0 | 0.0 | 0.0 |
| 2 | 3 | 01/22/2020 | Chongqing | Mainland China | 1/22/2020 17:00 | 6.0 | 0.0 | 0.0 |
| 3 | 4 | 01/22/2020 | Fujian | Mainland China | 1/22/2020 17:00 | 1.0 | 0.0 | 0.0 |
| 4 | 5 | 01/22/2020 | Gansu | Mainland China | 1/22/2020 17:00 | 0.0 | 0.0 | 0.0 |
| ... | ... | ... | ... | ... | ... | ... | ... | ... |
| 306424 | 306425 | 05/29/2021 | Zaporizhia Oblast | Ukraine | 2021-05-30 04:20:55 | 102641.0 | 2335.0 | 95289.0 |
| 306425 | 306426 | 05/29/2021 | Zeeland | Netherlands | 2021-05-30 04:20:55 | 29147.0 | 245.0 | 0.0 |
| 306426 | 306427 | 05/29/2021 | Zhejiang | Mainland China | 2021-05-30 04:20:55 | 1364.0 | 1.0 | 1324.0 |
| 306427 | 306428 | 05/29/2021 | Zhytomyr Oblast | Ukraine | 2021-05-30 04:20:55 | 87550.0 | 1738.0 | 83790.0 |
| 306428 | 306429 | 05/29/2021 | Zuid-Holland | Netherlands | 2021-05-30 04:20:55 | 391559.0 | 4252.0 | 0.0 |

306429 rows × 8 columns

```
In [ ]: ### Python Project :
#Dataset : Corona_virus
#Analyse the data and give the answers of below questions :

#1.What is the total number of confirmed cases worldwide?
#2.How many deaths have been reported globally?
#3.What is the total number of recovered cases worldwide?
#4.How many countries/regions are represented in the dataset?
#5.What is the trend of confirmed cases over time globally?
#6.Which province/state has reported the highest number of confirmed cases?
#7.Which country/region has the highest number of deaths?
#8.How does the number of confirmed cases vary across different provinces/states?
#9.What is the trend of deaths over time globally?
#10.Which country/region has the highest number of recovered cases?
#11.How does the number of recovered cases vary across different countries/regions?
#12.What is the distribution of confirmed cases by country/region?
#13.Is there a correlation between the number of confirmed cases and deaths?
#14.Is there a correlation between the number of confirmed cases and recovered cases?
#15.How does the mortality rate vary across different countries/regions?
#16.How does the recovery rate vary across different countries/regions?
#17.What is the trend of new confirmed cases over time globally?
#18.How does the fatality rate vary across different provinces/states?
#19.How does the recovery rate vary across different provinces/states?
#20.What is the trend of active cases over time globally?
```

```
In [15]: #1.What is the total number of confirmed cases worldwide?
a=df['Confirmed'].sum()
print('the total no. of confirmed cases worldwide are:',a)
```

the total no. of confirmed cases worldwide are: 26252051758.0

```
In [7]: #2.How many deaths have been reported globally?
a=df['Deaths'].sum()
print('the globally deaths reported are:',a)
```

the globally deaths reported are: 624013017.0

```
In [19]: #3.What is the total number of recovered cases worldwide?
a=df['Recovered'].sum()
print('the total no. of recovered cases worldwide are:',a)
```

the total no. of recovered cases worldwide are: 15450237912.0

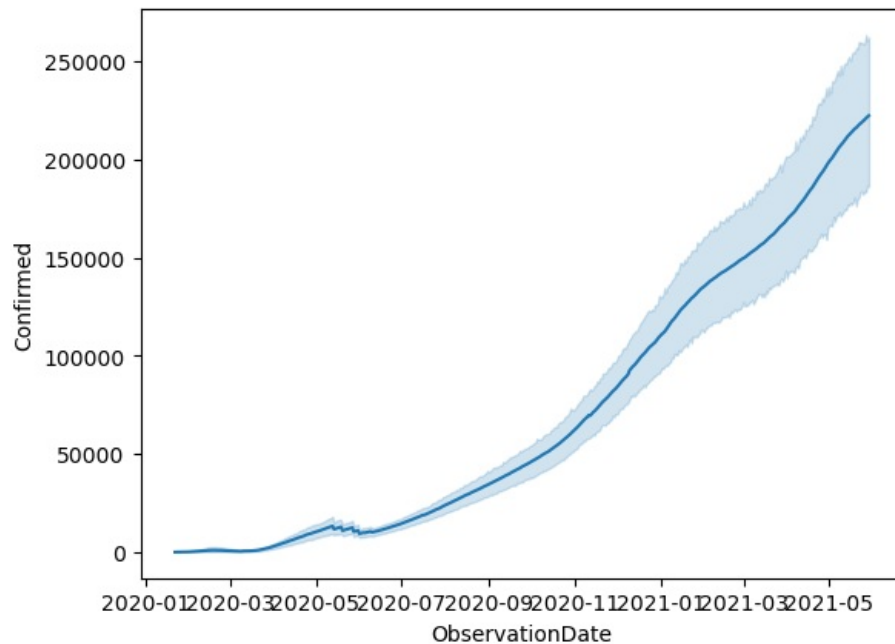
```
In [9]: #4.How many countries/regions are represented in the dataset?
x=df['Country/Region'].nunique()
print('the countries/regions are:',x)
```

the countries/regions are: 229

```
In [15]: import seaborn as sns
#5.What is the trend of confirmed cases over time globally?
df['ObservationDate']=pd.to_datetime(df['ObservationDate'])
global_trend=df.groupby('ObservationDate')['Confirmed'].sum()
```

```
sns.lineplot(data=df,x='ObservationDate',y='Confirmed')
```

```
Out[15]: <Axes: xlabel='ObservationDate', ylabel='Confirmed'>
```



```
In [29]: #6.Which province/state has reported the highest number of confirmed cases?
b=df.groupby('Province/State')['Confirmed'].sum().reset_index()
highest=b.loc[b['Confirmed'].idxmax()]
highest
```

```
Out[29]: Province/State    California
Confirmed      696898013.0
Name: 88, dtype: object
```

```
In [37]: #7.Which country/region has the highest number of deaths?
b=df.groupby('Country/Region')['Deaths'].sum().reset_index()
highest_deaths=b.loc[b['Deaths'].idxmax()]
highest_deaths
```

```
Out[37]: Country/Region    US
Deaths      123303762.0
Name: 214, dtype: object
```

```
In [61]: #8.How does the number of confirmed cases vary across different provinces/states?
statewise_cases=df.groupby('Province/State')['Confirmed'].sum().reset_index()
statewise_cases=statewise_cases.sort_values(by='Confirmed',ascending=False)
statewise_cases
```

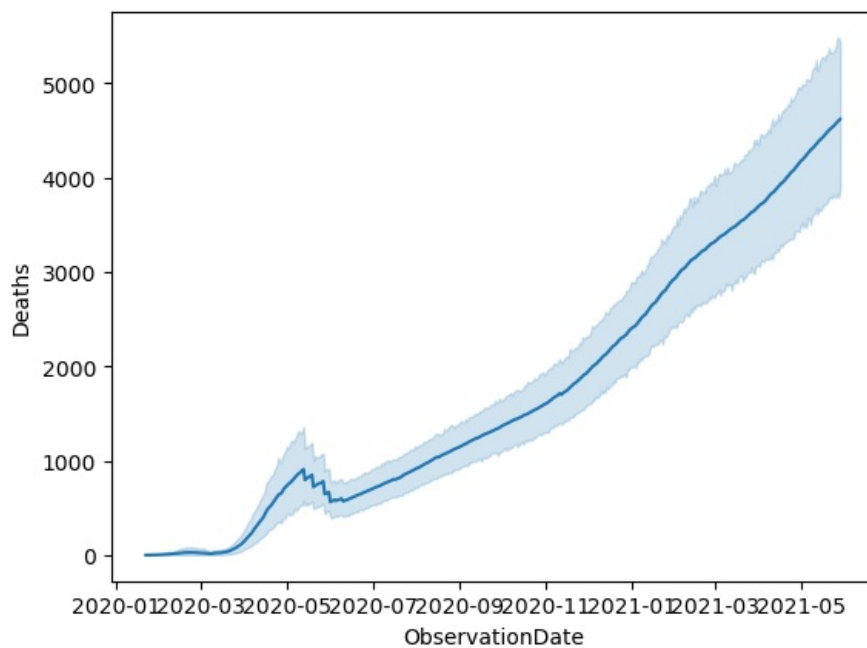
```
Out[61]:
```

| | Province/State | Confirmed |
|-----|-----------------------|-------------|
| 88 | California | 696898013.0 |
| 365 | Maharashtra | 681186928.0 |
| 171 | England | 666227518.0 |
| 630 | Texas | 552039886.0 |
| 570 | Sao Paulo | 521308945.0 |
| ... | ... | ... |
| 17 | American Samoa | 0.0 |
| 278 | Jervis Bay Territory | 0.0 |
| 526 | Recovered | 0.0 |
| 404 | Montgomery County, TX | 0.0 |
| 173 | External territories | 0.0 |

736 rows × 2 columns

```
In [39]: #9.What is the trend of deaths over time globally?
df['ObservationDate']=pd.to_datetime(df['ObservationDate'])
global_trend=df.groupby('ObservationDate')['Deaths'].sum()
sns.lineplot(data=df,x='ObservationDate',y='Deaths')
```

```
Out[39]: <Axes: xlabel='ObservationDate', ylabel='Deaths'>
```



```
In [45]: #10.Which country/region has the highest number of recovered cases?
b=df.groupby(by='Country/Region')['Recovered'].sum().reset_index()
highest_recovered=b.loc[b['Recovered'].idxmax()]
highest_recovered
```

```
Out[45]: Country/Region      India
Recovered      2900589824.0
Name: 96, dtype: object
```

```
In [7]: #11.How does the number of recovered cases vary across different countries/regions?
country_recoveries = df.groupby('Country/Region')['Recovered'].sum().reset_index()
country_recoveries.sort_values(by='Recovered', ascending=False)
```

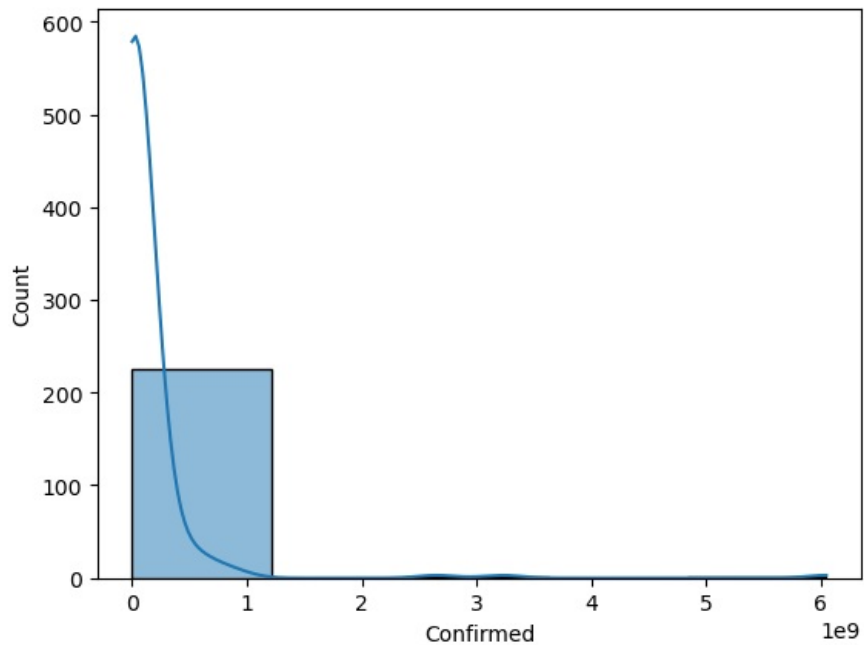
```
Out[7]:
```

| | Country/Region | Recovered |
|-----|--------------------------------|--------------|
| 96 | India | 2.900590e+09 |
| 27 | Brazil | 2.313677e+09 |
| 172 | Russia | 7.907057e+08 |
| 212 | Turkey | 5.641706e+08 |
| 214 | US | 5.033710e+08 |
| ... | ... | ... |
| 166 | Puerto Rico | 0.000000e+00 |
| 168 | Republic of Ireland | 0.000000e+00 |
| 169 | Republic of the Congo | 0.000000e+00 |
| 170 | Reunion | 0.000000e+00 |
| 228 | occupied Palestinian territory | 0.000000e+00 |

229 rows × 2 columns

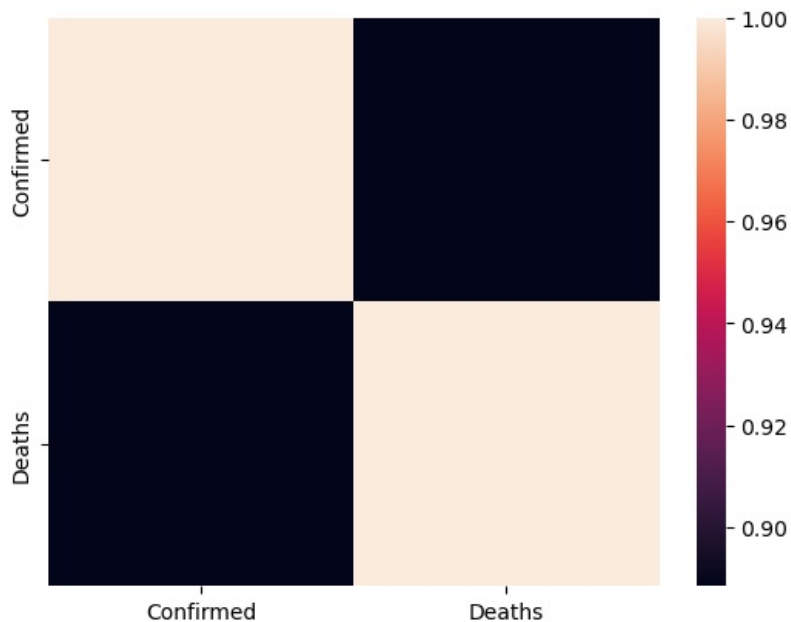
```
In [51]: #12.What is the distribution of confirmed cases by country/region?
country_cases=df.groupby('Country/Region')['Confirmed'].sum().reset_index()
sns.histplot(data=country_cases,x='Confirmed',bins=5,kde=True)
```

```
Out[51]: <Axes: xlabel='Confirmed', ylabel='Count'>
```



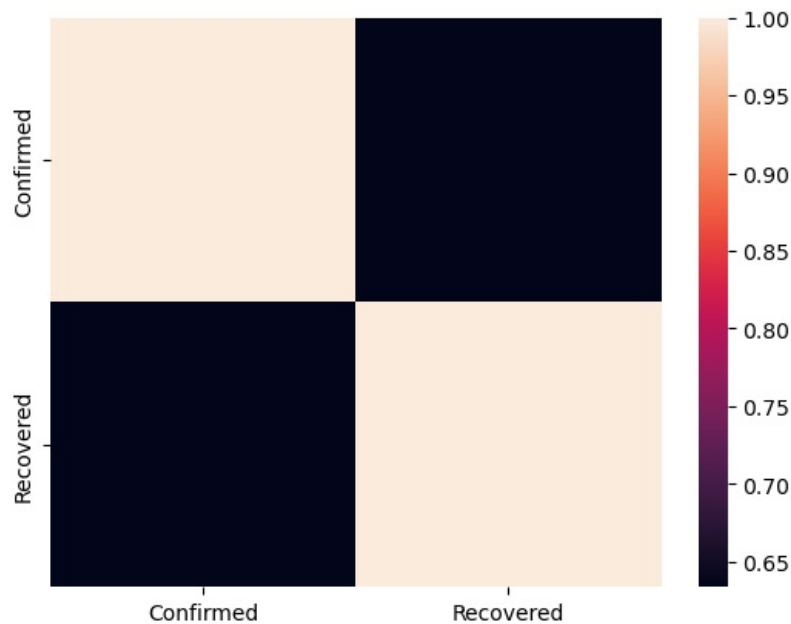
In [63]: *#13.Is there a correlation between the number of confirmed cases and deaths?*
`df[['Confirmed','Deaths']].corr()
sns.heatmap(df[['Confirmed','Deaths']].corr())`

Out[63]: <Axes: >



In [65]: *#14.Is there a correlation between the number of confirmed cases and recovered cases?*
`df[['Confirmed','Recovered']].corr()
sns.heatmap(df[['Confirmed','Recovered']].corr())`

Out[65]: <Axes: >



```
In [9]: #15.How does the mortality rate vary across different countries/regions?
country_data=df.groupby('Country/Region')[['Confirmed','Deaths']].sum().reset_index()
country_data=country_data[country_data['Confirmed']>0]
country_data['Mortality Rate(%)']=(country_data['Deaths']/country_data['Confirmed'])*100
country_data_sorted=country_data.sort_values(by='Mortality Rate(%)',ascending=False)
country_data_sorted.head(100)
```

```
Out[9]:
```

| | Country/Region | Confirmed | Deaths | Mortality Rate(%) |
|-----|----------------|-------------|------------|-------------------|
| 225 | Yemen | 962066.0 | 237613.0 | 24.698202 |
| 123 | MS Zaandam | 3824.0 | 848.0 | 22.175732 |
| 220 | Vanuatu | 406.0 | 39.0 | 9.605911 |
| 137 | Mexico | 460463678.0 | 43005509.0 | 9.339609 |
| 197 | Sudan | 7632455.0 | 488709.0 | 6.403038 |
| ... | ... | ... | ... | ... |
| 143 | Morocco | 104557135.0 | 1823724.0 | 1.744237 |
| 35 | Cameroon | 11346589.0 | 197906.0 | 1.744189 |
| 109 | Kenya | 27728648.0 | 482736.0 | 1.740929 |
| 12 | Austria | 97965875.0 | 1678309.0 | 1.713157 |
| 93 | Hong Kong | 2655935.0 | 45325.0 | 1.706555 |

100 rows × 4 columns

```
In [11]: #16.How does the recovery rate vary across different countries/regions?
country_data=df.groupby('Country/Region')[['Confirmed','Recovered']].sum().reset_index()
country_data=country_data[country_data['Confirmed']>0]
country_data['Recovery Rate(%)']=(country_data['Recovered']/country_data['Confirmed'])*100
country_data_sorted=country_data.sort_values(by='Recovery Rate(%)',ascending=False)
country_data_sorted.head(100)
```

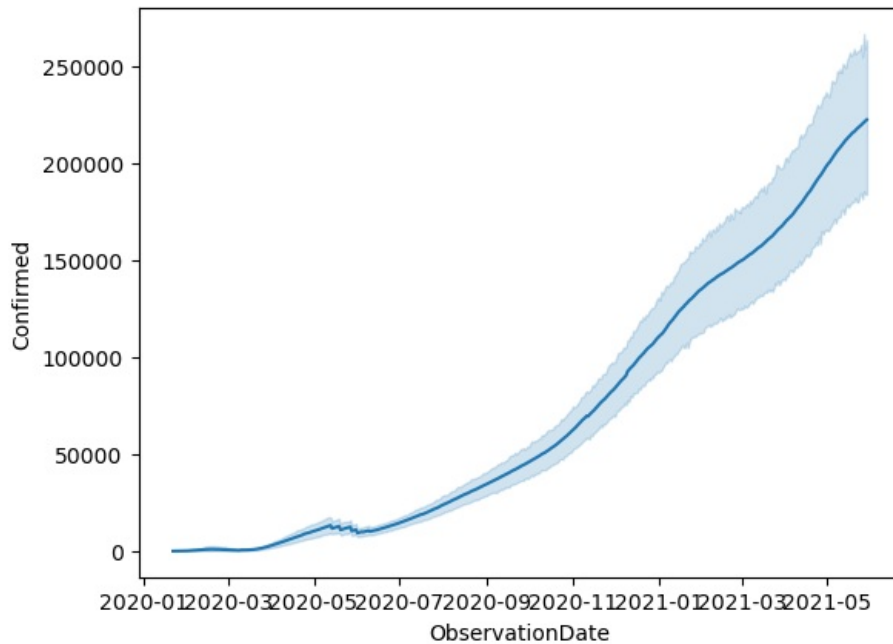
Out[11]:

| | Country/Region | Confirmed | Recovered | Recovery Rate(%) |
|-----|------------------|-------------|-------------|------------------|
| 219 | Uzbekistan | 22207571.0 | 21035683.0 | 94.723025 |
| 78 | Ghana | 20784664.0 | 19586296.0 | 94.234364 |
| 55 | Diamond Princess | 306872.0 | 288580.0 | 94.039209 |
| 138 | Micronesia | 129.0 | 121.0 | 93.798450 |
| 95 | Iceland | 1729527.0 | 1621682.0 | 93.764480 |
| ... | ... | ... | ... | ... |
| 97 | Indonesia | 265186050.0 | 226416174.0 | 85.380122 |
| 151 | Nigeria | 33407947.0 | 28514090.0 | 85.351219 |
| 150 | Niger | 1047041.0 | 892393.0 | 85.229996 |
| 172 | Russia | 930548849.0 | 790705716.0 | 84.971973 |
| 165 | Portugal | 141962632.0 | 120619045.0 | 84.965348 |

100 rows × 4 columns

In [18]: *#17.What is the trend of new confirmed cases over time globally?*
`df['ObservationDate']=pd.to_datetime(df['ObservationDate'])
global_trend=df.groupby('ObservationDate')['Confirmed'].sum().reset_index()
global_trend.sort_values('ObservationDate')
sns.lineplot(data=df,x='ObservationDate',y='Confirmed')`

Out[18]: <Axes: xlabel='ObservationDate', ylabel='Confirmed'>



In [28]: *#18.How does the fatality rate vary across different provinces/states?*
`country_data=df.groupby('Province/State')['Confirmed','Deaths'].sum().reset_index()
country_data=country_data[country_data['Confirmed']>0]
country_data['Mortality Rate(%)']=(country_data['Deaths']/country_data['Confirmed'])*100
country_data_sorted=country_data.sort_values(by='Mortality Rate(%)',ascending=False)
country_data_sorted.head(100)`

Out[28]:

| | Province/State | Confirmed | Deaths | Mortality Rate(%) |
|-----|-----------------------|-------------|------------|-------------------|
| 668 | Unknown | 7804169.0 | 4247616.0 | 54.427525 |
| 568 | Santa Rosa County, FL | 5.0 | 2.0 | 40.000000 |
| 338 | Lee County, FL | 6.0 | 2.0 | 33.333333 |
| 314 | King County, WA | 412.0 | 91.0 | 22.087379 |
| 499 | Placer County, CA | 28.0 | 6.0 | 21.428571 |
| ... | ... | ... | ... | ... |
| 492 | Perm Krai | 9426150.0 | 341592.0 | 3.623876 |
| 601 | Sormland | 3171168.0 | 114559.0 | 3.612518 |
| 171 | England | 666227518.0 | 24042130.0 | 3.608697 |
| 95 | Caqueta | 4000046.0 | 144258.0 | 3.606409 |
| 287 | Junin | 10474345.0 | 372782.0 | 3.559001 |

100 rows × 4 columns

In [22]:

```
#19.How does the recovery rate vary across different provinces/states?
country_data=df.groupby('Province/State')[['Confirmed','Recovered']].sum().reset_index()
country_data=country_data[country_data['Confirmed']>0]
country_data['Recovery Rate(%)']=(country_data['Recovered']/country_data['Confirmed'])*100
country_data_sorted=country_data.sort_values(by='Recovery Rate(%)',ascending=False)
country_data_sorted.head(100)
```

Out[22]:

| | Province/State | Confirmed | Recovered | Recovery Rate(%) |
|-----|--|-------------|-------------|------------------|
| 656 | US | 4.0 | 532.0 | 13300.000000 |
| 668 | Unknown | 7804169.0 | 619474280.0 | 7937.735331 |
| 527 | Repatriated Travellers | 2431.0 | 2431.0 | 100.000000 |
| 549 | Saint Helena, Ascension and Tristan da Cunha | 882.0 | 863.0 | 97.845805 |
| 439 | Ningxia | 35904.0 | 34976.0 | 97.415330 |
| ... | ... | ... | ... | ... |
| 82 | Buryatia Republic | 6975683.0 | 6404564.0 | 91.812716 |
| 626 | Telangana | 84952428.0 | 77992105.0 | 91.806799 |
| 694 | Volgograd Oblast | 10840874.0 | 9952071.0 | 91.801371 |
| 709 | West Bengal | 154070425.0 | 141428487.0 | 91.794702 |
| 97 | Casanare | 2583680.0 | 2371658.0 | 91.793798 |

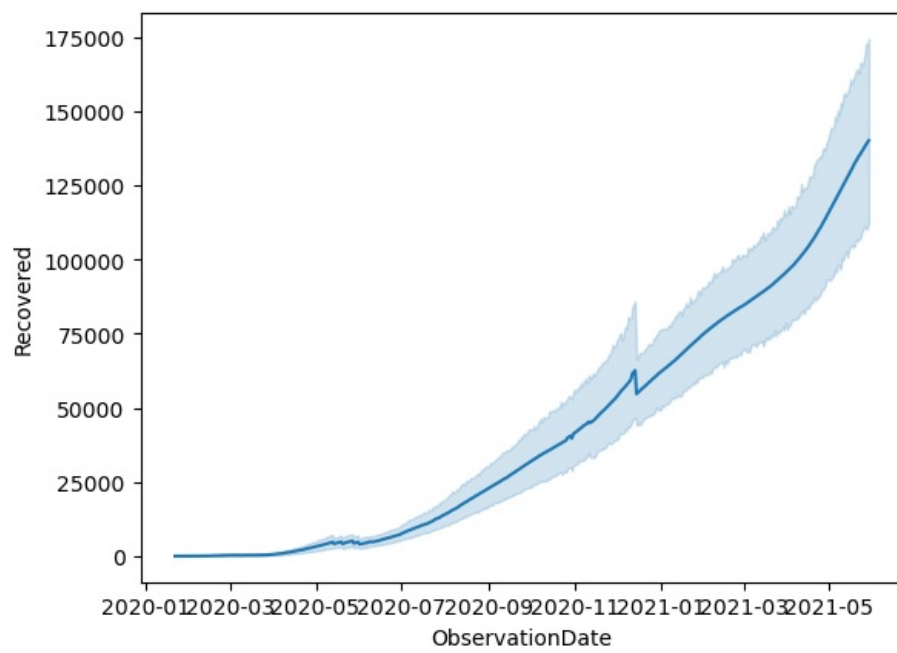
100 rows × 4 columns

In [55]:

```
#20.What is the trend of active cases over time globally?
df['ObservationDate']=pd.to_datetime(df['ObservationDate'])
global_trend=df.groupby('ObservationDate')['Recovered'].sum()
sns.lineplot(data=df,x='ObservationDate',y='Recovered')
```

Out[55]:

<Axes: xlabel='ObservationDate', ylabel='Recovered'>



In []:

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