

# Unemployment Analysis using Python

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The COVID-19 pandemic and containment responses have exposed existing social, economic, gender, ethnic and health inequities and are rapidly exacerbating these. The higher risks of infection and mortality recorded for specific population groups can be traced to a legacy of cumulative inequities in the social determinants of health (SDH). Additionally, the necessary COVID-19 response measures applied, while helping to reduce infection and mortality, have placed a disproportionate burden on more disadvantaged populations, widening health inequities.

**Here we will analysis the unemployment rate of India using Python.**

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import cv2
import plotly.express as px

%matplotlib inline
```

## Data Collection

```
In [3]: data = pd.read_csv(r"\\Unemployment_Analysis_data.csv")
data.head()
```

Out[3]:

	Region	Date	Frequency	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Region.1	longitude	latitu
0	Andhra Pradesh	31-01-2020	M	5.48	16635535	41.02	South	15.9129	79.
1	Andhra Pradesh	29-02-2020	M	5.83	16545652	40.90	South	15.9129	79.
2	Andhra Pradesh	31-03-2020	M	5.79	15881197	39.18	South	15.9129	79.
3	Andhra Pradesh	30-04-2020	M	20.51	11336911	33.10	South	15.9129	79.
4	Andhra Pradesh	31-05-2020	M	17.43	12988845	36.46	South	15.9129	79.

```
In [4]: print(data.head())
```

	Region	Date	Frequency	Estimated Unemployment Rate (%)
0	Andhra Pradesh	31-01-2020	M	5.48
1	Andhra Pradesh	29-02-2020	M	5.83
2	Andhra Pradesh	31-03-2020	M	5.79
3	Andhra Pradesh	30-04-2020	M	20.51
4	Andhra Pradesh	31-05-2020	M	17.43

	Estimated Employed	Estimated Labour Participation Rate (%)	Region.1
0	16635535	41.02	South
1	16545652	40.90	South
2	15881197	39.18	South
3	11336911	33.10	South
4	12988845	36.46	South

	longitude	latitude
0	15.9129	79.74
1	15.9129	79.74
2	15.9129	79.74
3	15.9129	79.74
4	15.9129	79.74

## Data Pre-processing

In [5]: data.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 267 entries, 0 to 266
Data columns (total 9 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Region                                267 non-null    object
1   Date                                  267 non-null    object
2   Frequency                             267 non-null    object
3   Estimated Unemployment Rate (%)        267 non-null    float64
4   Estimated Employed                     267 non-null    int64
5   Estimated Labour Participation Rate (%) 267 non-null    float64
6   Region.1                              267 non-null    object
7   longitude                             267 non-null    float64
8   latitude                              267 non-null    float64
dtypes: float64(4), int64(1), object(4)
memory usage: 18.9+ KB
```

In [6]: data.describe()

Out[6]:

	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	longitude	latitude
<b>count</b>	267.000000	2.670000e+02	267.000000	267.000000	267.000000
<b>mean</b>	12.236929	1.396211e+07	41.681573	22.826048	80.532425
<b>std</b>	10.803283	1.336632e+07	7.845419	6.270731	5.831738
<b>min</b>	0.500000	1.175420e+05	16.770000	10.850500	71.192400
<b>25%</b>	4.845000	2.838930e+06	37.265000	18.112400	76.085600
<b>50%</b>	9.650000	9.732417e+06	40.390000	23.610200	79.019300
<b>75%</b>	16.755000	2.187869e+07	44.055000	27.278400	85.279900
<b>max</b>	75.850000	5.943376e+07	69.690000	33.778200	92.937600

In [7]: `print(data.describe())`

	Estimated Unemployment Rate (%)	Estimated Employed \	
count	267.000000	2.670000e+02	
mean	12.236929	1.396211e+07	
std	10.803283	1.336632e+07	
min	0.500000	1.175420e+05	
25%	4.845000	2.838930e+06	
50%	9.650000	9.732417e+06	
75%	16.755000	2.187869e+07	
max	75.850000	5.943376e+07	

	Estimated Labour Participation Rate (%)	longitude	latitude
count	267.000000	267.000000	267.000000
mean	41.681573	22.826048	80.532425
std	7.845419	6.270731	5.831738
min	16.770000	10.850500	71.192400
25%	37.265000	18.112400	76.085600
50%	40.390000	23.610200	79.019300
75%	44.055000	27.278400	85.279900
max	69.690000	33.778200	92.937600

In [8]: `data.columns`

Out[8]: Index(['Region', 'Date', ' Frequency', ' Estimated Unemployment Rate (%)',  
' Estimated Employed', ' Estimated Labour Participation Rate (%)',  
'Region.1', 'longitude', 'latitude'],  
dtype='object')

## Check if the dataset contains null value or not

In [9]: `data.isnull()`

Out[9]:

	Region	Date	Frequency	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Region.1	longitude	lati
0	False	False	False	False	False	False	False	False	f
1	False	False	False	False	False	False	False	False	f
2	False	False	False	False	False	False	False	False	f
3	False	False	False	False	False	False	False	False	f
4	False	False	False	False	False	False	False	False	f
...	...	...	...	...	...	...	...	...	...
262	False	False	False	False	False	False	False	False	f
263	False	False	False	False	False	False	False	False	f
264	False	False	False	False	False	False	False	False	f
265	False	False	False	False	False	False	False	False	f
266	False	False	False	False	False	False	False	False	f

267 rows × 9 columns



In [10]: `print(data.isnull().sum())`

```

Region          0
Date            0
Frequency       0
Estimated Unemployment Rate (%)  0
Estimated Employed  0
Estimated Labour Participation Rate (%)  0
Region.1        0
longitude       0
latitude        0
dtype: int64

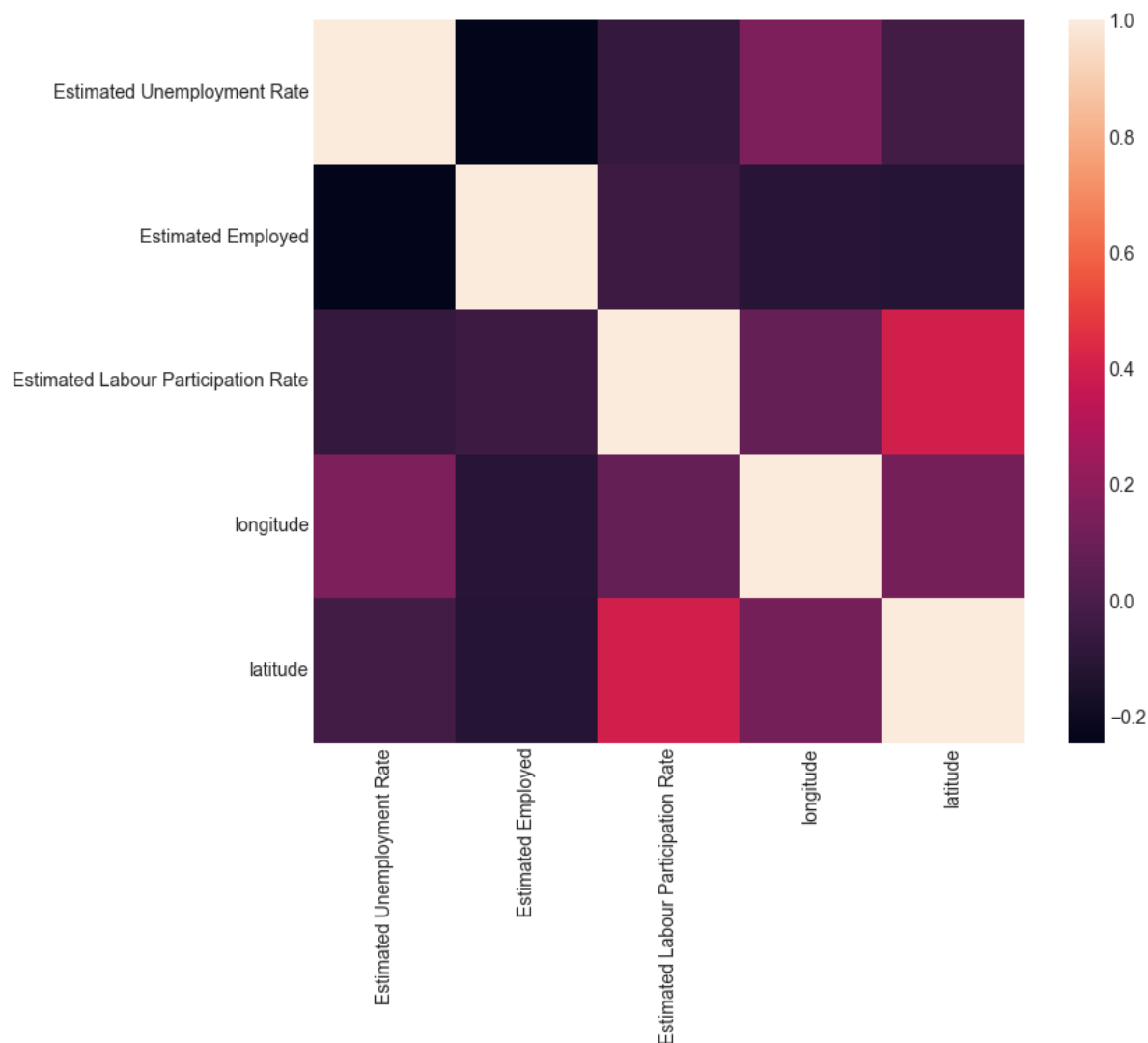
```

**While analyzing the missing values, we found that the column names are not correct. So, for a better understanding of this data, we will rename all the columns**

```
In [11]: data.columns= ["States", "Date", "Frequency",  
                        "Estimated Unemployment Rate",  
                        "Estimated Employed",  
                        "Estimated Labour Participation Rate",  
                        "Region", "longitude", "latitude"]
```

## Correlation between the features of the Dataset

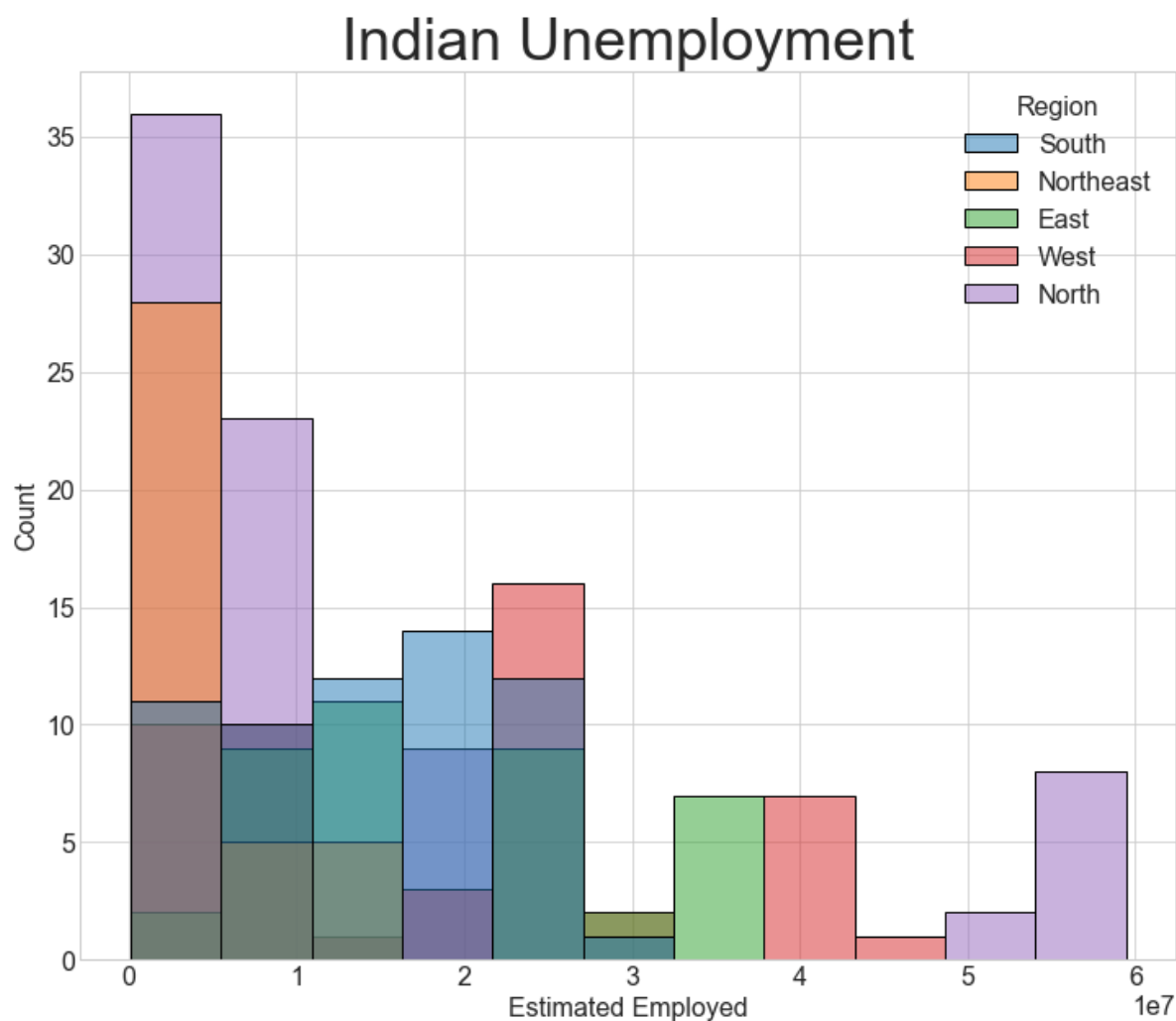
```
In [12]: plt.style.use("seaborn-whitegrid")  
plt.figure(figsize=(12,10))  
plt.rc("font", size=14)  
sns.heatmap(data.corr())  
plt.show()
```



# Unemployment Rate Analysis - Data Visualization

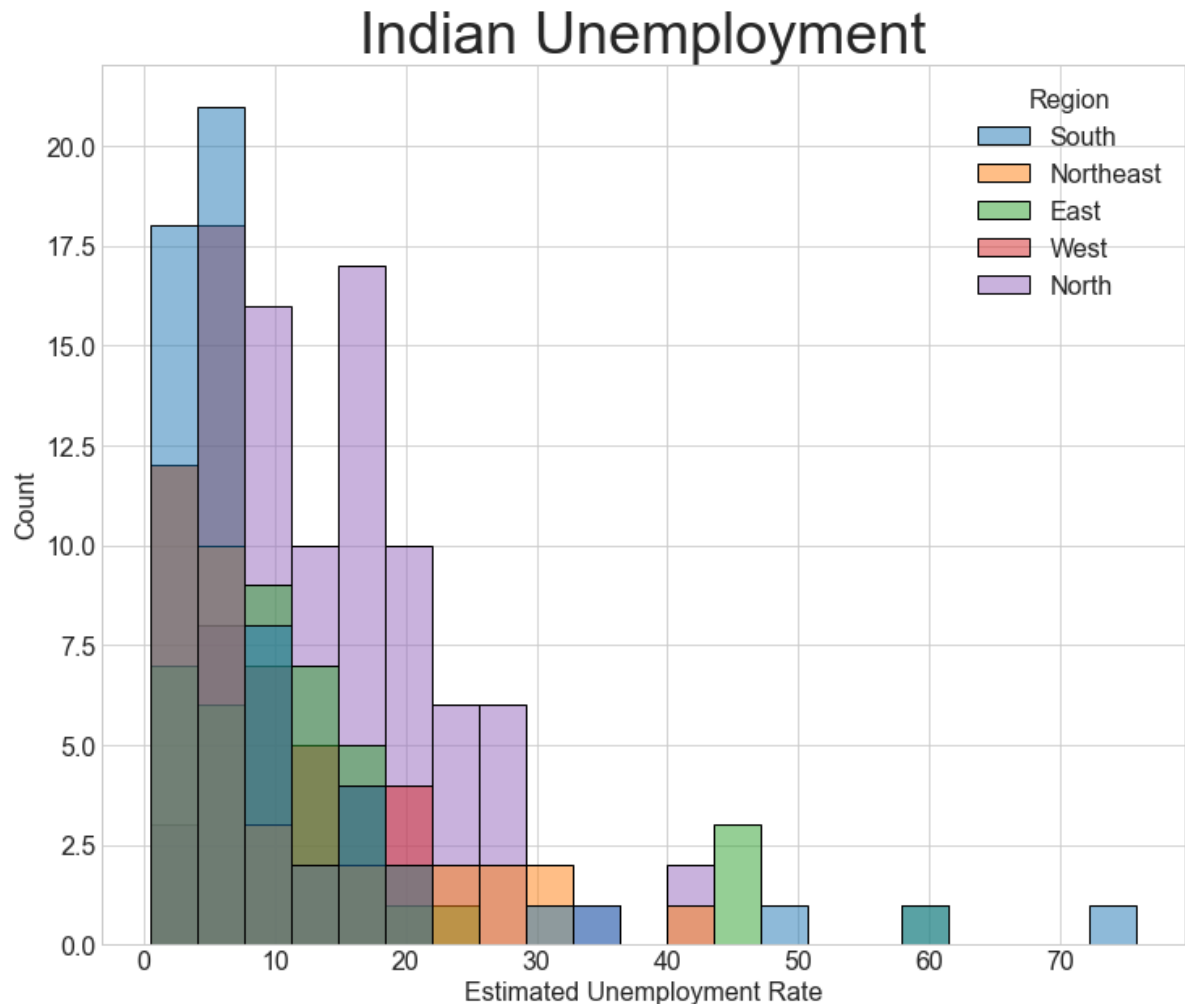
Let visualize the dataset to calculate Unemployment Rate. We will first take a look at the estimated number of employees according to different regions of India

```
In [13]: data.columns = ["States", "Date", "Frequency",
                        "Estimated Unemployment Rate", "Estimated Employed",
                        "Estimated Labour Participation Rate", "Region",
                        "longitude", "latitude"]
plt.figure(figsize=(12,10))
plt.rc("font",size=16)
plt.title("Indian Unemployment", fontsize=36)
sns.histplot(x="Estimated Employed", hue="Region", data=data)
plt.show()
```



Now let's see the unemployment rate according to different regions of India

```
In [15]: plt.figure(figsize=(12, 10))
plt.rc("font", size=16)
plt.title("Indian Unemployment", fontsize=36)
sns.histplot(x="Estimated Unemployment Rate", hue="Region", data=data)
plt.show()
```



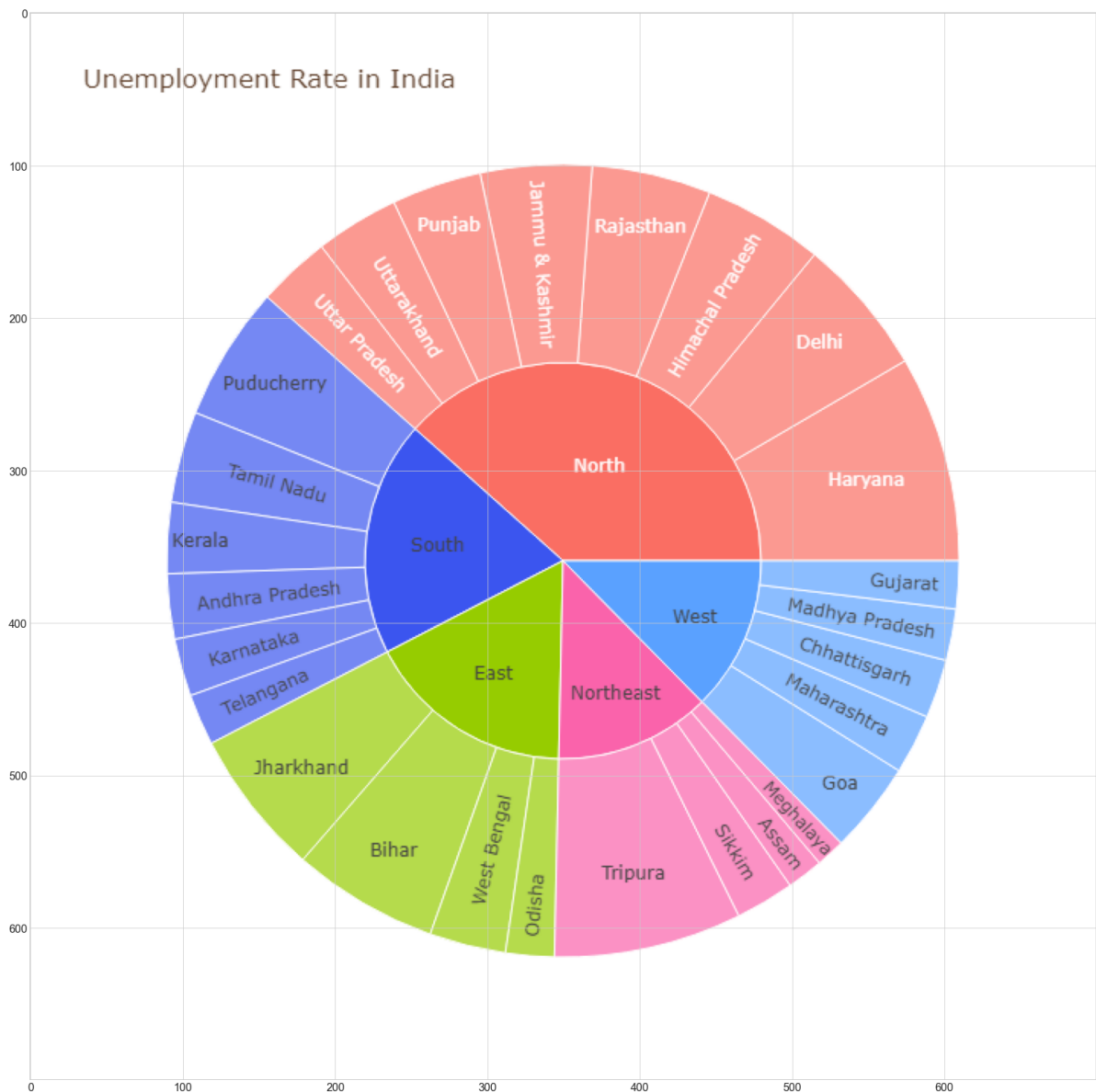
**Create a dashboard to analyze the unemployment rate of each Indian state by region. For this, I'll use a sunburst plot**

```
In [16]: unemployment = data[["States", "Region", "Estimated Unemployment Rate"]]
figure = px.sunburst(unemployment, path=["Region", "States"],
                    values="Estimated Unemployment Rate",
                    width=700, height=700, color_continuous_scale="RdY1Gn",
                    title="Unemployment Rate in India")
figure.show()
```



```
In [17]: img = cv2.imread(".\\unemployment rate.png")
plt.figure(figsize=(35,28))
plt.imshow(img)
```

```
Out[17]: <matplotlib.image.AxesImage at 0x227e48e85b0>
```



```
In [ ]:
```