TITLE- UNEMPLOYMENT IN INDIA DURING COVID

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INTRODUCTION

In this project I have taken dataset of UNEMPLOYMENT IN INDIA , time range lying between year 2019 and 2020. We'll clean it, filter out the data, and create meaningful visualizations and insights.

DESCRIPTION OF UNEMPLOYMENT DATASET

This dataset provides a detailed ionformation about the unemployment trends in india across various states , union territories , subdivided into rural and urban aread. The time-range of this dataset is from year 2019 to 2020. The dataset includes 7 key columns

1. Region
2. Date
3. Frequency of data collection
4. Estimated unemployment rate %
5. Estimated employed
6. Labour participation rate %
7. Area

These attributes provide both quantitative and qualitative insights into the employment scenario in the country.

SOURCE OF DATA: The data is sourced from the Centre for Monitoring Indian Economy (CMIE), known for its reliable economic statistics.

KEY FEATURES OF THE DATA SET

1. **Regional Coverage:** The dataset covers a wide range of states and union territories, including Andhra Pradesh, Assam, Bihar, Delhi, Gujarat, Haryana, Karnataka, Kerala, Maharashtra, Punjab, Tamil Nadu, Uttar Pradesh, and West Bengal, among others. This allows for comparative analysis across different regions of India.

2. **Temporal Scope:** The data ranges from May 2019 to June 2020, capturing the immediate impact of the COVID-19 pandemic on employment. The months of April and May 2020, in particular, show significant spikes in unemployment rates due to nationwide lockdowns and economic disruptions.

3. **Frequency:** The data is recorded monthly, providing a precise view of how unemployment fluctuated over time, including seasonal variations and sudden shocks like the pandemic.

4.Rural-Urban Divide: The dataset distinguishes between rural and urban areas, highlighting disparities in employment trends. For instance, rural areas in states like Bihar and Jharkhand experienced extreme unemployment spikes during the lockdown, while urban areas in states like Delhi and Maharashtra also saw significant job losses.

DESCRIPTION ABOUT THE PROJECT

In this project we will be importing the dataset , and will be working on the following objectives with some creative visualizations and a final conclusion which will give more clarity to us about the aim of the project

Objectives are:

1. Comparison of Mean Unemployment rate in 2019 and 2020 using seaborn
2. Rural and urban estimated labour participation using seaborn
3. Using numpy classifying unemployment rate as normal or high comparing it with mean Estimated Unemployment Rate (%) and plotting a geoplot

SO LETS GET STARTED….

TABLE OF UNEMPLOYMENT DATASET

A	B	C	D	E	F	G
Region	Date	Frequency	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate	Area
Andhra Pradesh	31-05-2019	Monthly	3.65	11999139	43.24	Rural
Andhra Pradesh	30-06-2019	Monthly	3.05	11755881	42.05	Rural
Andhra Pradesh	31-07-2019	Monthly	3.75	12086707	43.5	Rural
Andhra Pradesh	31-08-2019	Monthly	3.32	12285693	43.97	Rural
Andhra Pradesh	30-09-2019	Monthly	5.17	12256762	44.68	Rural
Andhra Pradesh	31-10-2019	Monthly	3.52	12017412	43.01	Rural
Andhra Pradesh	30-11-2019	Monthly	4.12	11397681	41	Rural
Andhra Pradesh	31-12-2019	Monthly	4.38	12528395	45.14	Rural
Andhra Pradesh	31-01-2020	Monthly	4.84	12016676	43.46	Rural
Andhra Pradesh	29-02-2020	Monthly	5.91	11723617	42.83	Rural
Andhra Pradesh	31-03-2020	Monthly	4.06	11359660	40.66	Rural
Andhra Pradesh	30-04-2020	Monthly	16.29	8792827	36.03	Rural
Andhra Pradesh	31-05-2020	Monthly	14.46	9526902	38.16	Rural
Andhra Pradesh	30-06-2020	Monthly	0.85	15572975	53.76	Rural
Assam	31-05-2019	Monthly	4.29	11749334	57.39	Rural
Assam	30-06-2019	Monthly	5.08	8923222	43.87	Rural
Assam	31-07-2019	Monthly	4.26	9911534	48.21	Rural
Assam	31-08-2019	Monthly	5.79	9292039	45.83	Rural
Assam	30-09-2019	Monthly	4.46	11468349	55.67	Rural
Assam	31-10-2019	Monthly	4.65	8395906	40.76	Rural
Assam	30-11-2019	Monthly	4.66	9625362	46.64	Rural

Step 1: Import all the libraries

```
#importing all the libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import geopandas as gpd
```

Step 2: Load the dataset

```
#importing the data
df=pd.read_csv("/content/Unemployment in India.csv")
print(df)
```

Output:

	Region	Date	Frequency	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Area
0	Andhra Pradesh	31-05-2019	Monthly	3.65	11999139.0	43.24	Rural
1	Andhra Pradesh	30-06-2019	Monthly	3.05	11755881.0	42.05	Rural
2	Andhra Pradesh	31-07-2019	Monthly	3.75	12086707.0	43.50	Rural
3	Andhra Pradesh	31-08-2019	Monthly	3.32	12285693.0	43.97	Rural
4	Andhra Pradesh	30-09-2019	Monthly	5.17	12256762.0	44.68	Rural
..	...	...	...	...	...	...	...
749	West Bengal	29-02-2020	Monthly	7.55	10871168.0	44.09	Urban
750	West Bengal	31-03-2020	Monthly	6.67	10806105.0	43.34	Urban
751	West Bengal	30-04-2020	Monthly	15.63	9299466.0	41.20	Urban
752	West Bengal	31-05-2020	Monthly	15.22	9240903.0	40.67	Urban
753	West Bengal	30-06-2020	Monthly	9.86	9088931.0	37.57	Urban

[754 rows x 7 columns]

Step 3- Data cleaning and preparation

i : deleting all the extra blank spaces in out text data using str.strip() function

```
[9]  # delete gtrailing spaces
     df.columns = df.columns.str.strip()
```

ii- checking the null values using isnull().sum() function

```
#checking if any value is missing or null
print(df.isnull().sum())
```

Output:

```
Region                                     14
Date                                       14
Frequency                                  14
Estimated Unemployment Rate (%)            14
Estimated Employed                         14
Estimated Labour Participation Rate (%)    14
Area                                       14
dtype: int64
```

iii- Deleting all the null values using dropna()

```
#dropping the empty columns
new_df=df.dropna()
print(new_df)
```

Output:

```
            Region         Date Frequency  Estimated Unemployment Rate (%)  \
0    Andhra Pradesh   31-05-2019   Monthly                              3.65
1    Andhra Pradesh   30-06-2019   Monthly                              3.05
2    Andhra Pradesh   31-07-2019   Monthly                              3.75
3    Andhra Pradesh   31-08-2019   Monthly                              3.32
4    Andhra Pradesh   30-09-2019   Monthly                              5.17
..              ...          ...       ...                               ...
749     West Bengal   29-02-2020   Monthly                              7.55
750     West Bengal   31-03-2020   Monthly                              6.67
751     West Bengal   30-04-2020   Monthly                             15.63
752     West Bengal   31-05-2020   Monthly                             15.22
753     West Bengal   30-06-2020   Monthly                              9.86

     Estimated Employed  Estimated Labour Participation Rate (%)   Area
0             11999139.0                                  43.24  Rural
1             11755881.0                                  42.05  Rural
2             12086707.0                                  43.50  Rural
3             12285693.0                                  43.97  Rural
4             12256762.0                                  44.68  Rural
..                   ...                                    ...    ...
749           10871168.0                                  44.09  Urban
750           10806105.0                                  43.34  Urban
751            9299466.0                                  41.20  Urban
752            9240903.0                                  40.67  Urban
753            9088931.0                                  37.57  Urban

[740 rows x 7 columns]
```

iv- Fixing the wrong date format

Input:

```
#fixing the wrong date format
new_df['Date']=pd.to_datetime(new_df['Date'])
print(new_df)
```

Output:

	Region	Date	Frequency	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Area
0	Andhra Pradesh	2019-05-31	Monthly	3.65	11999139.0	43.24	Rural
1	Andhra Pradesh	2019-06-30	Monthly	3.05	11755881.0	42.05	Rural
2	Andhra Pradesh	2019-07-31	Monthly	3.75	12086707.0	43.50	Rural
3	Andhra Pradesh	2019-08-31	Monthly	3.32	12285693.0	43.97	Rural
4	Andhra Pradesh	2019-09-30	Monthly	5.17	12256762.0	44.68	Rural
..	...	...	...	...	...	...	...
749	West Bengal	2020-02-29	Monthly	7.55	10871168.0	44.09	Urban
750	West Bengal	2020-03-31	Monthly	6.67	10806105.0	43.34	Urban
751	West Bengal	2020-04-30	Monthly	15.63	9299466.0	41.20	Urban
752	West Bengal	2020-05-31	Monthly	15.22	9240903.0	40.67	Urban
753	West Bengal	2020-06-30	Monthly	9.86	9088931.0	37.57	Urban

What we did!

i. Deleted all the extra spaces in our text data using str.strip() method so that while working on our data it dosent give any errors.

ii. Checking for null values using isnull() then doing the sum of all the null values if found using sum().

iii. Deleting all the 14 empty rows.

iv. Fixing the wrong date format from “DD-MM-YYYY” to “YYYY-MM-DD”

Step 4- Data filtering and analysis

- **Filtering data of only 2019:**

```
#data of only 2019
x = new_df[(new_df['Date'] >= pd.to_datetime("2019-01-01")) &
        (new_df['Date'] < pd.to_datetime("2020-01-01"))]
print(x)
```

- **output**

```
                Region       Date Frequency  Estimated Unemployment Rate (%)  \
0    Andhra Pradesh 2019-05-31   Monthly                              3.65
1    Andhra Pradesh 2019-06-30   Monthly                              3.05
2    Andhra Pradesh 2019-07-31   Monthly                              3.75
3    Andhra Pradesh 2019-08-31   Monthly                              3.32
4    Andhra Pradesh 2019-09-30   Monthly                              5.17
..              ...        ...       ...                               ...
743     West Bengal 2019-08-31   Monthly                              7.27
744     West Bengal 2019-09-30   Monthly                              7.79
745     West Bengal 2019-10-31   Monthly                              7.83
746     West Bengal 2019-11-30   Monthly                              6.61
747     West Bengal 2019-12-31   Monthly                              7.24

     Estimated Employed  Estimated Labour Participation Rate (%)   Area
0             11999139.0                                  43.24  Rural
1             11755881.0                                  42.05  Rural
2             12086707.0                                  43.50  Rural
3             12285693.0                                  43.97  Rural
4             12256762.0                                  44.68  Rural
..                   ...                                    ...    ...
743           11456493.0                                  46.77  Urban
744           11158649.0                                  45.74  Urban
745           10563686.0                                  43.25  Urban
746           10768462.0                                  43.44  Urban
747           11335696.0                                  45.97  Urban

[430 rows x 7 columns]
```

Similarly filtering Data of only 2020:

```
#data of only 2020
y = new_df[(new_df['Date'] > pd.to_datetime("2019-12-31")) &
        (new_df['Date'] < pd.to_datetime("2021-01-01"))]
print(y)
```

Output:

```
            Region       Date Frequency  Estimated Unemployment Rate (%)  \
8    Andhra Pradesh 2020-01-31   Monthly                             4.84
9    Andhra Pradesh 2020-02-29   Monthly                             5.91
10   Andhra Pradesh 2020-03-31   Monthly                             4.06
11   Andhra Pradesh 2020-04-30   Monthly                            16.29
12   Andhra Pradesh 2020-05-31   Monthly                            14.46
..              ...        ...       ...                              ...
749     West Bengal 2020-02-29   Monthly                             7.55
750     West Bengal 2020-03-31   Monthly                             6.67
751     West Bengal 2020-04-30   Monthly                            15.63
752     West Bengal 2020-05-31   Monthly                            15.22
753     West Bengal 2020-06-30   Monthly                             9.86

     Estimated Employed  Estimated Labour Participation Rate (%)   Area
8             12016676.0                                  43.46  Rural
9             11723617.0                                  42.83  Rural
10            11359660.0                                  40.66  Rural
11             8792827.0                                  36.03  Rural
12             9526902.0                                  38.16  Rural
..                   ...                                    ...    ...
749           10871168.0                                  44.09  Urban
750           10806105.0                                  43.34  Urban
751            9299466.0                                  41.20  Urban
752            9240903.0                                  40.67  Urban
753            9088931.0                                  37.57  Urban

[310 rows x 7 columns]
```

Key insight:

- Filtering data on the basis of years for the analysis of objective 1.

Objective 1 : avgerage unemployment rate before and during covid

a. Mean unemployment rate before and during covid

```
#barplot of avg unemp rates in 2019 and 2020 using seabor
mean_2019 = x['Estimated Unemployment Rate (%)'].mean()
mean_2020 = y['Estimated Unemployment Rate (%)'].mean()
print('MEAN UNEMPLOMENT RATE IN 2019 =',mean_2019)
print('MEAN UNEMPLOMENT RATE IN 2020 =',mean_2020)
# Prepare data for plotting
```

Output:

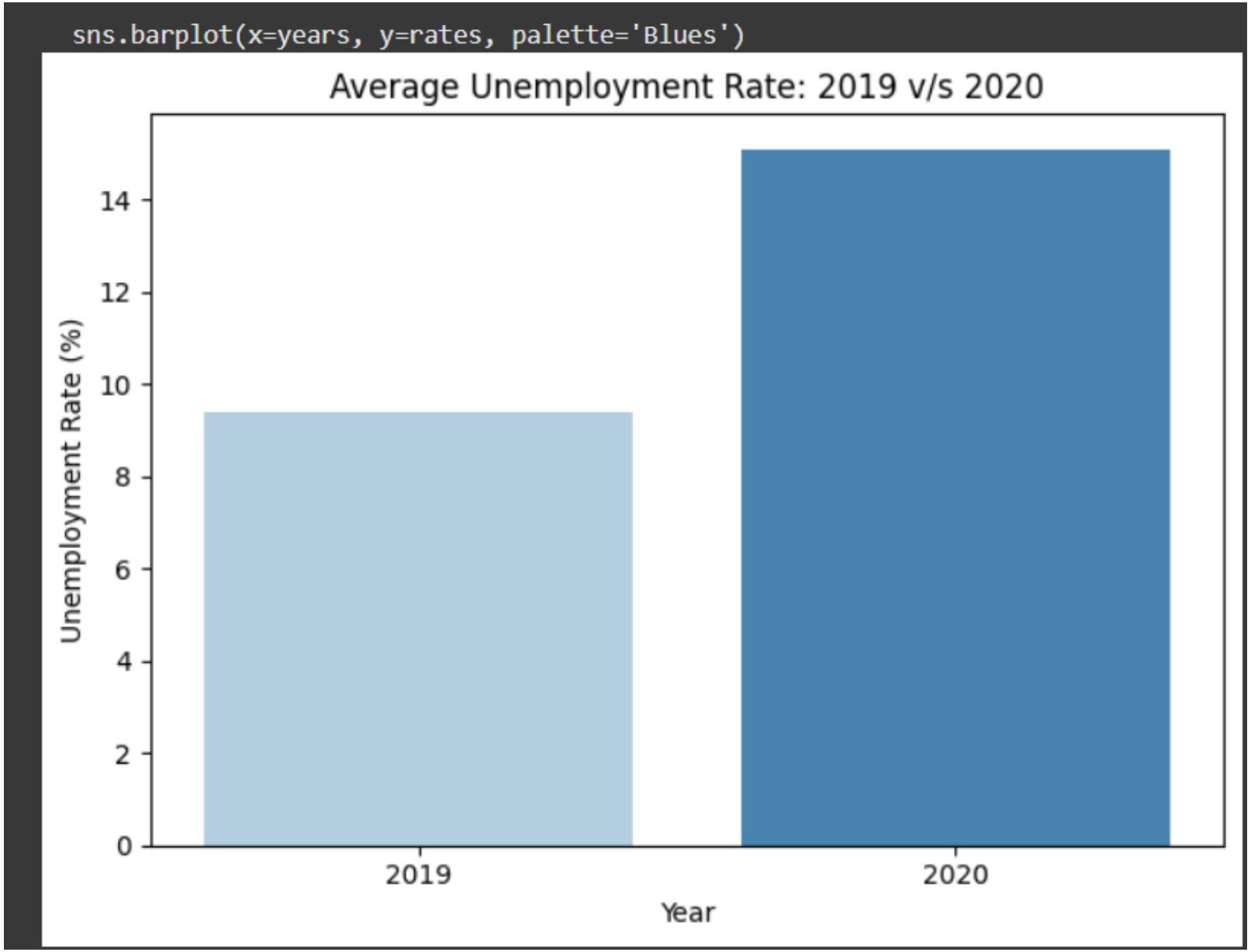
```
MEAN UNEMPLOMENT RATE IN 2019 = 9.3990465511627906
MEAN UNEMPLOMENT RATE IN 2020 = 15.101580645516129
```

- **Visualization of obj 1 using seaborn:**

```
# Prepare data for plotting
years = ['2019', '2020']
rates = [mean_2019, mean_2020]

# Plot
sns.barplot(x=years, y=rates, palette='Blues')
plt.title('Average Unemployment Rate: 2019 v/s 2020')
plt.ylabel('Unemployment Rate (%)')
plt.xlabel('Year')
plt.tight_layout()
plt.show()
```

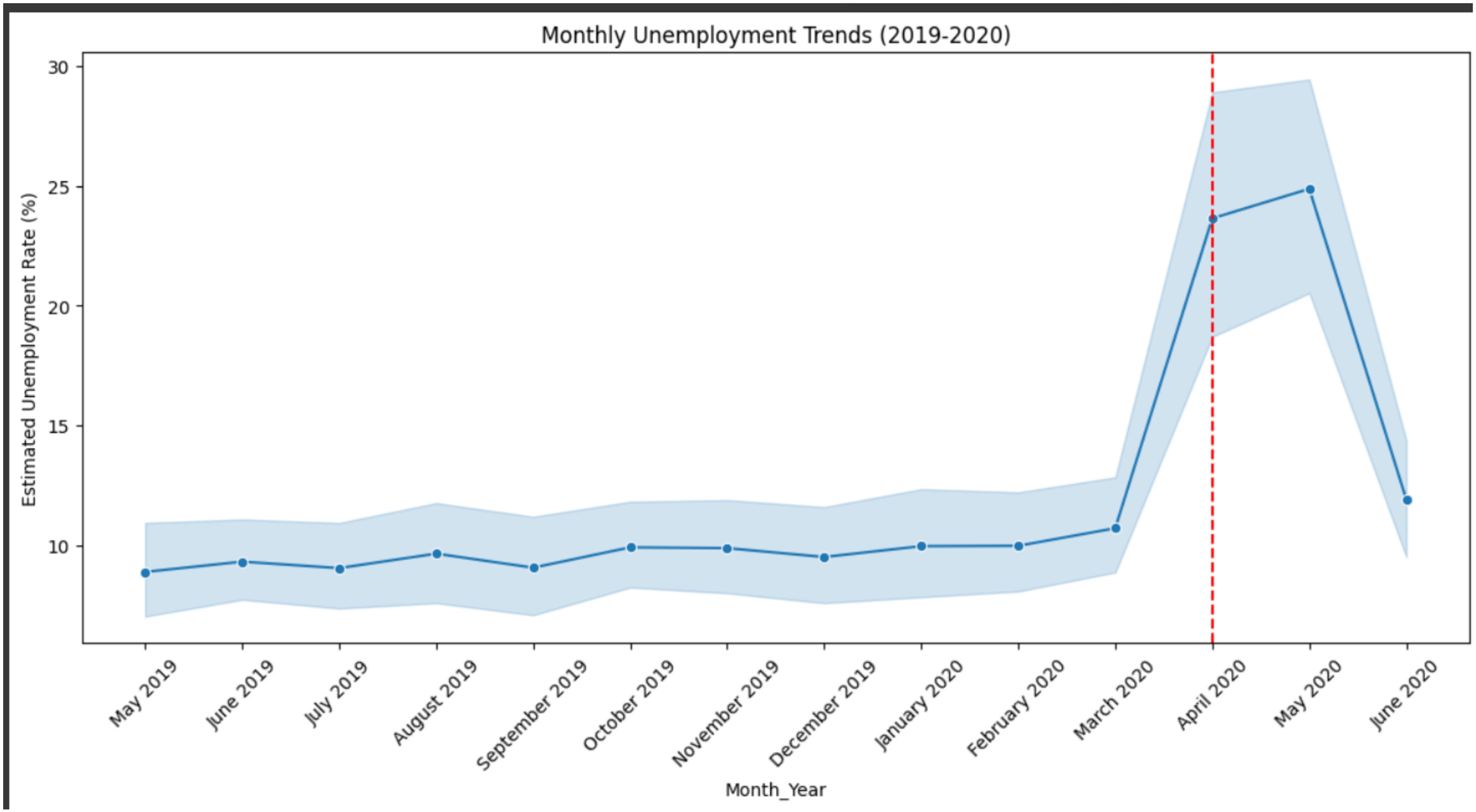
Output:

```
sns.barplot(x=years, y=rates, palette='Blues')
```



- **Monthly estimated unemployment rate analysis from may 2019 to June 2020 using seaborn lineplot**

```
#monthly analysis
new_df['Date'] = pd.to_datetime(new_df['Date'])
new_df['Month'] = new_df['Date'].dt.month_name()  # e.g., "April"
new_df['Year'] = new_df['Date'].dt.year           # e.g., 2020
new_df['Month_Year'] = new_df['Month'] + ' ' + new_df['Year'].astype(str)  # e.g., "April 2020"
plt.figure(figsize=(14, 6))
sns.lineplot(
    x='Month_Year',
    y='Estimated Unemployment Rate (%)',
    data=new_df.sort_values('Date'),  # Ensure chronological order,
    marker='o'
)
plt.xticks(rotation=45)  # Rotate x-labels for readability
plt.axvline(x='April 2020', color='red', linestyle='--', label='Lockdown Start')
plt.title('Monthly Unemployment Trends (2019-2020)')
plt.show()
```


Monthly Unemployment Trends (2019-2020)
Estimated Unemployment Rate (%)
30
25
20
15
10
May 2019
June 2019
July 2019
August 2019
September 2019
October 2019
November 2019
December 2019
January 2020
February 2020
March 2020
April 2020
May 2020
June 2020
Month_Year

Key Insights:

•Compared the average unemployment rates in 2019 (pre-COVID) and 2020 (COVID period) to assess the pandemic's impact.

•Also we performed monthly analysis which gave more clarity in understanding unemployment trend.

•Visualized the data using Seaborn to highlight differences more clearly.

•Observed a significant increase in unemployment during the COVID period, reflecting:

- Economic slowdown
- Job losses due to lockdowns
- Migration of labor from urban to rural areas

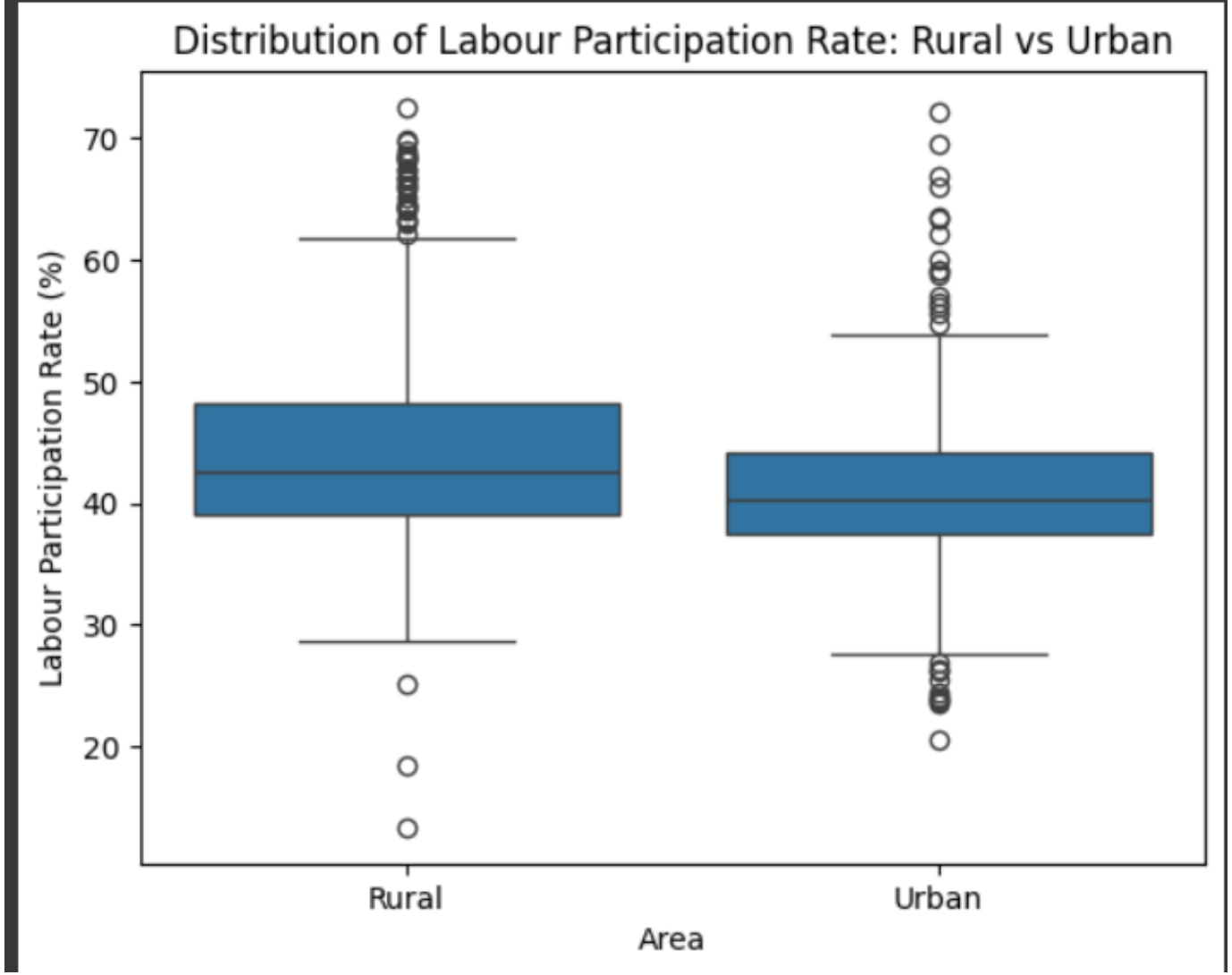
•Helps understand how external shocks like pandemics affect labor markets in India.

Objective 2 : Rural v/s urban labour participation using seaborn

```
grp = new_df.groupby('Area')['Estimated Labour Participation Rate (%)'].mean()
print(grp)
sns.boxplot(x='Area', y='Estimated Labour Participation Rate (%)', data=new_df)
plt.title("Distribution of Labour Participation Rate: Rural vs Urban")
plt.ylabel("Labour Participation Rate (%)")
plt.xlabel("Area")
plt.show()
```

Output:

```
Area
Rural    44.464819
Urban    40.901365
Name: Estimated Labour Participation Rate (%), dtype: float64
```



Key insights:

1. The boxplot visually compares labour participation rates between rural and urban areas.
2. Rural areas generally show a higher median labour participation rate than urban areas.
3. Greater spread (interquartile range) in rural areas indicates more variability in participation, likely due to seasonal jobs , migration of labours , informal sector work, or government employment schemes.
4. Urban participation rates appear more consistent, with fewer fluctuations.
5. Outliers in both areas suggest months or regions with unusually high or low participation.

Objective 3 : Classifying unemployment rate as normal or high using numpy

```
#using numpy classifying unemp rate as normal or high
mean_rate = new_df['Estimated Unemployment Rate (%)'].mean()

new_df['Unemp_Level'] = np.where(
    new_df['Estimated Unemployment Rate (%)'] > mean_rate,
    'High',
    'Normal'
)
print(new_df)
```

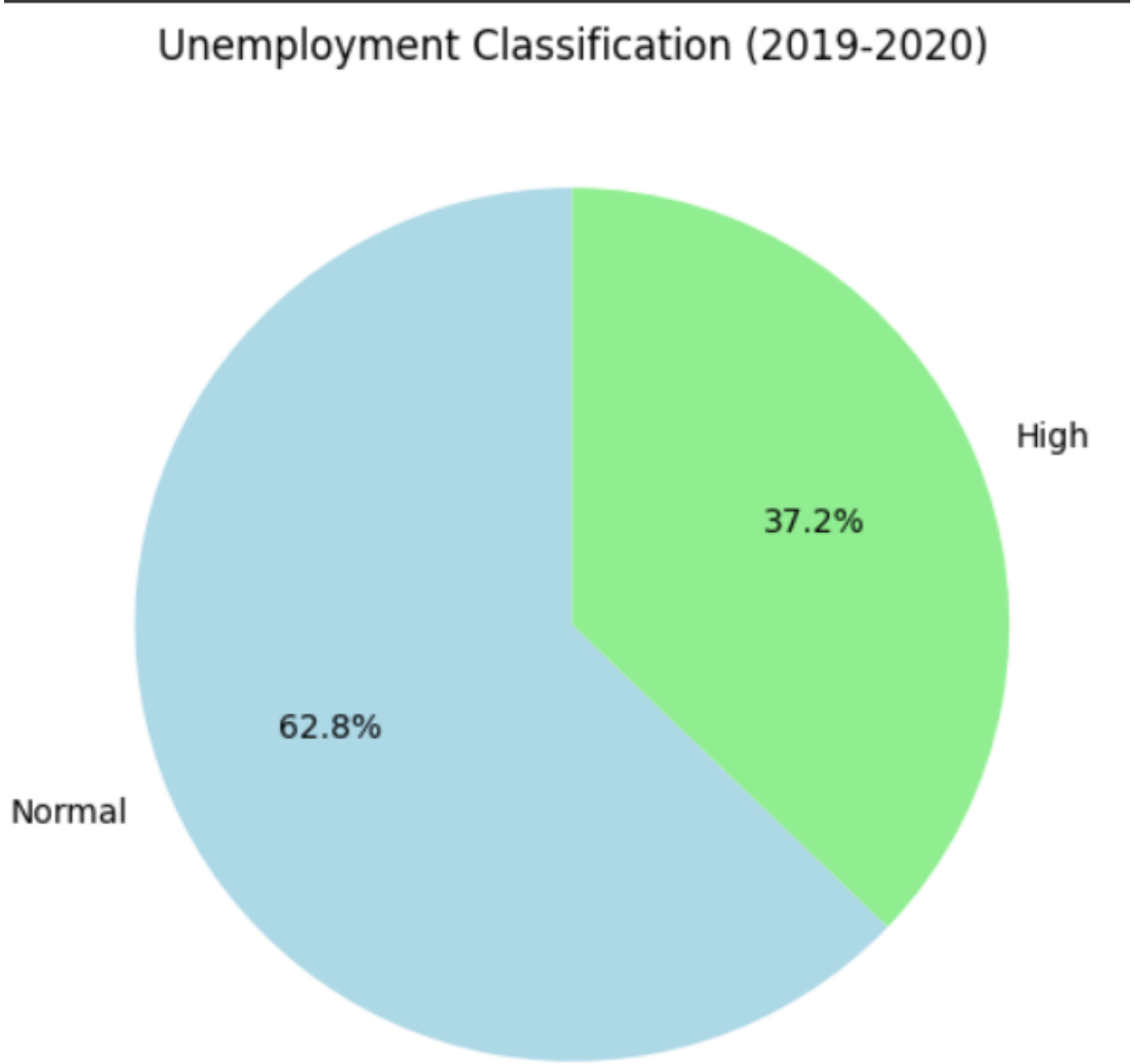
Output:

	Region	Date	Frequency	Unemp_Level
0	Andhra Pradesh	2019-05-31	Monthly	Normal
1	Andhra Pradesh	2019-06-30	Monthly	Normal
2	Andhra Pradesh	2019-07-31	Monthly	Normal
3	Andhra Pradesh	2019-08-31	Monthly	Normal
4	Andhra Pradesh	2019-09-30	Monthly	Normal
..	...	...	...	...
749	West Bengal	2020-02-29	Monthly	Normal
750	West Bengal	2020-03-31	Monthly	Normal
751	West Bengal	2020-04-30	Monthly	High
752	West Bengal	2020-05-31	Monthly	High
753	West Bengal	2020-06-30	Monthly	Normal

Pie chart:

```
# Count High vs Normal unemployment months
counts = new_df['Unemp_Level'].value_counts()

# Creatimg a pie chart
plt.figure(figsize=(6,6))
plt.pie(counts,
        labels=counts.index,
        autopct='%1.1f%%',
        colors=['lightblue', 'lightgreen'],  # Green (Normal) and Red (High)
        startangle=90)
plt.title('Unemployment Classification (2019-2020)')
plt.show()
```

Output:

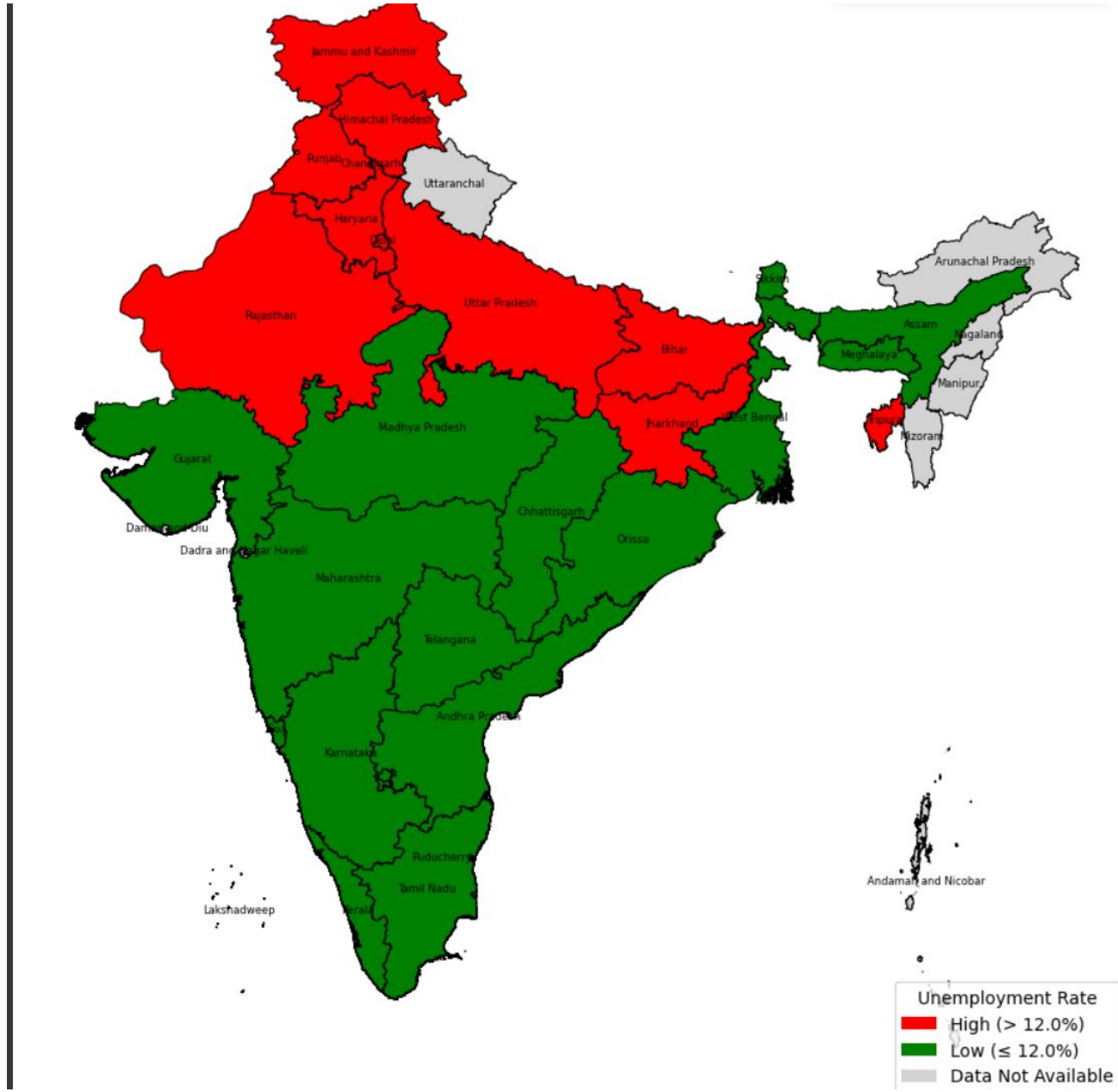

Unemployment Classification (2019-2020)
High
37.2%
62.8%
Normal

- **Geoplot of states with high and normal estimated unemployment rate .**

```
# Load data
india = gpd.read_file('https://github.com/geohacker/india/raw/master/state/india_telengana.geojson')
state_class = new_df.groupby('Region')[' Estimated Unemployment Rate (%)'].mean().reset_index()

# Standardize names in BOTH datasets
def standardize_names(name):
    name = str(name).lower()
    replacements = [
        ('&', 'and'),
        ('nct of ', ''),
        ('dadra and nagar haveli and daman and diu', 'dadra and nagar haveli'),
        ('odisha', 'orissa'),
        ('puducherry', 'pondicherry')
    ]
    for old, new in replacements:
        name = name.replace(old, new)
    return name.strip()

india['clean_name'] = india['NAME_1'].apply(standardize_names)
state_class['clean_name'] = state_class['Region'].apply(standardize_names)

# Merge data
india = india.merge(state_class, on='clean_name', how='left')

# Calculate national mean AFTER merge to include all available states
national_mean = india[' Estimated Unemployment Rate (%)'].mean(skipna=True)
```

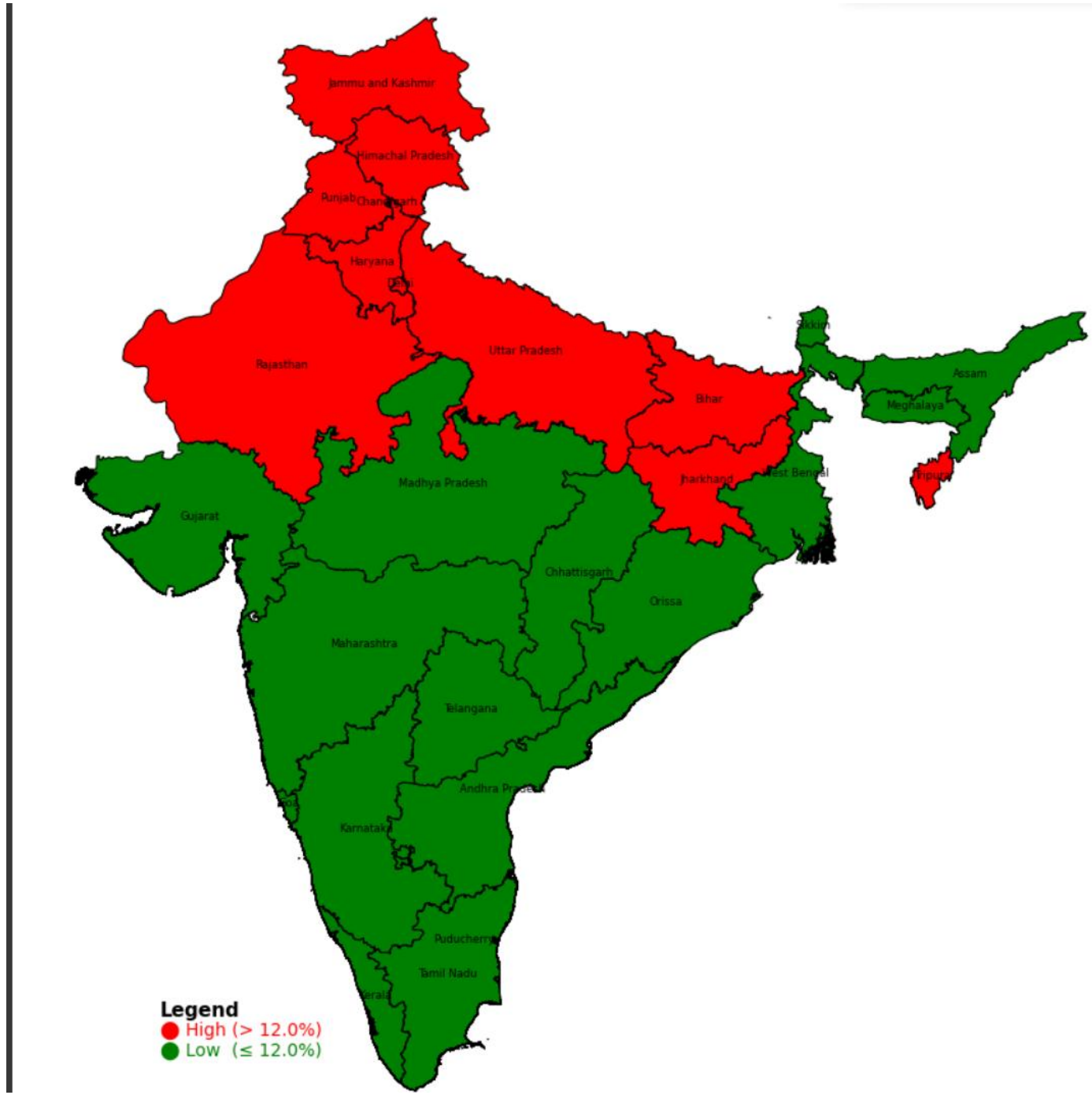
```python
# Categorize
india['Category'] = np.where(
    india[' Estimated Unemployment Rate (%)'] > national_mean,
    'High',
    np.where(~india[' Estimated Unemployment Rate (%)'].isna(), 'Low', 'Not Available')
)

# Assign colors
cmap = {'High': 'red', 'Low': 'green', 'Not Available': 'lightgrey'}
india['color'] = india['Category'].map(cmap)

# Plot
fig, ax = plt.subplots(figsize=(12, 10))
india.plot(color=india['color'], edgecolor='black', linewidth=0.8, ax=ax)

# Manual legend (without Patch)
legend_x, legend_y = 70.5, 8.5  # Adjust if needed
ax.text(legend_x, legend_y + 2.0, 'Legend', fontsize=11, weight='bold',
        bbox=dict(facecolor='white', edgecolor='none', alpha=0.8))

ax.text(legend_x, legend_y + 1.4, f'● High (> {national_mean:.1f}%)', color='red', fontsize=10)
ax.text(legend_x, legend_y + 0.9, f'● Low (≤ {national_mean:.1f}%)', color='green', fontsize=10)
ax.text(legend_x, legend_y + 0.4, '● Not Available', color='grey', fontsize=10)

# Title and formatting
plt.title('State-wise Unemployment Rates (2019–2020)', fontsize=16, pad=20)
plt.axis('off')
```

```python
# Annotate state names (optional)
for idx, row in india.iterrows():
    ax.annotate(text=row['NAME_1'], xy=(row.geometry.centroid.x, row.geometry.centroid.y),
                ha='center', fontsize=6, color='black')

plt.tight_layout()
plt.show()
```


Jammu and Kashmir
Himachal Pradesh
Punjab
Chandigarh
Uttaranchal
Haryana
Delhi
Rajasthan
Uttar Pradesh
Bihar
Sikkim
Arunachal Pradesh
Assam
Nagaland
Meghalaya
Manipur
Tripura
Mizoram
West Bengal
Jharkhand
Madhya Pradesh
Gujarat
Daman and Diu
Dadra and Nagar Haveli
Chhattisgarh
Orissa
Maharashtra
Telangana
Andhra Pradesh
Karnataka
Puducherry
Tamil Nadu
Kerala
Lakshadweep
Andaman and Nicobar
Unemployment Rate
High (> 12.0%)
Low (≤ 12.0%)
Data Not Available

The NAN states in the map are not present in the original dataset so I would be dropping them by adding the following code:

```
# Drop states with missing unemployment data
india = india.dropna(subset=[' Estimated Unemployment Rate (%)'])
```


Jammu and Kashmir
Himachal Pradesh
Punjab
Haryana
Rajasthan
Uttar Pradesh
Bihar
Sikkim
Assam
Meghalaya
Tripura
Jharkhand
West Bengal
Madhya Pradesh
Gujarat
Chhattisgarh
Orissa
Maharashtra
Telangana
Karnataka
Tamil Nadu
Legend
High (> 12.0%)
Low (≤ 12.0%)

Key Insights:

1. Used NumPy to calculate the overall mean of the 'Estimated Unemployment Rate (%)'.

2. Each unemployment value was classified as:

- 'High' if it was greater than the mean.
- Normal' if it was equal to or below the mean.

3. This classification helps:

- Quickly identify states/regions with higher-than-average unemployment.
- Focus further analysis or policy planning on areas marked 'High'.

4. The classification adds a new categorical dimension to the dataset, making it easier to visualize and filter data for reporting or dashboards.
5. Also this classification was presented through a pie chart and a geoplot for a better visualization .

Summary: Unemployment in India During COVID-19

This project analyzed India's unemployment trends from 2019 to 2020 using CMIE data.
Key findings:
1. Unemployment Spike: - Rates surged in 2020, peaking during lockdowns (April–May).

2. Rural vs. Urban:
- Rural areas had higher but unstable labour participation.
- Urban areas saw steady declines in jobs.

3.Regional Hotspots: - States like Rajasthan and Delhi had "high" unemployment (geoplot).

4. Classification: - Over 60% of states exceeded the national average (pie chart).

Conclusion: COVID-19 severely impacted jobs, especially in urban and high-unemployment states. Data visuals (geoplot, line/pie charts) effectively highlighted these trends.

THANK-YOU