# TASK 2 - UNEMPLOYMENT ANALYSIS WITH PYTHON

Annapoornima task2

Description of the dataset

Importing necessary libraries

```
# data processing
import numpy as np
import pandas as pd

# data visualization
import matplotlib.pyplot as plt
import seaborn as sns

import warnings
warnings.filterwarnings('ignore')
```

 ✓ Loading the dataset

```
df = pd.read_csv('Unemployment in India.csv')
df.head()
```

	Region	Date	Frequency	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Area	11.
0	Andhra Pradesh	31- 05- 2019	Monthly	3.65	11999139.0	43.24	Rural	
1	Andhra Pradesh	30- 06-	Monthly	3.05	11755881.0	42.05	Rural	

✓ Understanding the structure of the dataset

```
df.shape
```

(768, 7)

Renaming the column

```
df.rename(columns={'Region.1': 'Area'}, inplace=True)
```

 ✓ Checking for missing values

```
df.isnull().sum()
```

```
Region 28
Date 28
Frequency 28
Estimated Unemployment Rate (%) 28
Estimated Employed 28
Estimated Labour Participation Rate (%) 28
Area 28
dtype: int64
```

```
df.duplicated().sum()
```

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**~** 

# Summary of the dataframe

# df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype					
0	Region	740 non-null	object					
1	Date	740 non-null	object					
2	Frequency	740 non-null	object					
3	Estimated Unemployment Rate (%)	740 non-null	float64					
4	Estimated Employed	740 non-null	float64					
5	Estimated Labour Participation Rate (%)	740 non-null	float64					
6	Area	740 non-null	object					
dtyp	<pre>dtypes: float64(3), object(4)</pre>							
memo	memory usage: 42.1+ KB							

# Removing unintentional spaces in columns

df.columns = df.columns.str.strip()
df.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 768 entries, 0 to 767
Data columns (total 7 columns):

Duca	cotamins (total / cotamins):							
#	Column	Non-Null Count	Dtype					
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1	Date	740 non-null	object					
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3	Estimated Unemployment Rate (%)	740 non-null	float64					
4	Estimated Employed	740 non-null	float64					
5	Estimated Labour Participation Rate (%)	740 non-null	float64					
6	Area	740 non-null	object					
dtyp	<pre>dtypes: float64(3), object(4)</pre>							
memo	memory usage: 42.1+ KB							

Converting data types

df['Date'] = pd.to\_datetime(df['Date'])
df.dtypes

Region	object
Date	datetime64[ns]
Frequency	object
Estimated Unemployment Rate (%)	float64
Estimated Employed	float64
Estimated Labour Participation Rate (%)	float64
Area	object
dtype: object	

# Summary Statistics

# selecting the categorical variables
categorical\_var = df.select\_dtypes(include='object')
# Obtaining summary statistics for the categorical variables
categorical\_stat = categorical\_var.describe().T
categorical\_stat

	count	unique	top	freq	
Region	740	28	Andhra Pradesh	28	ıl.
Frequency	740	2	Monthly	381	+/
Δrea	740	2	Urhan	381	

# selecting numerical variables
numerical\_var = df.select\_dtypes(exclude='object')
# Obtaining summar statistics for the numerical variables
numerical\_stat = numerical\_var.describe().T
numerical\_stat

	count	mean	std	min	25%	50%
Estimated Unemployment Rate (%)	740.0	1.178795e+01	1.072130e+01	0.00	4.657500e+00	8.35

# ✓ Dropping irrelevent column

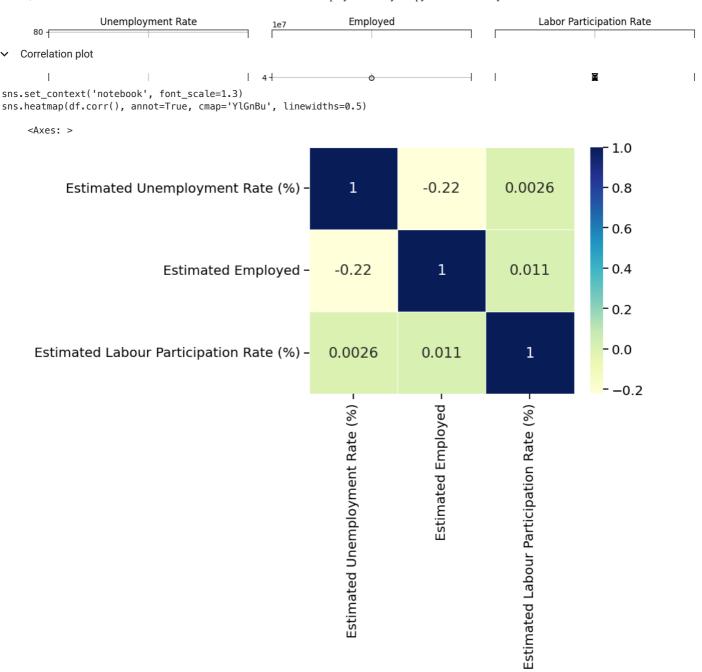
df = df.drop('Frequency', axis=1)
df.head()

	Region	Date	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Area
0	Andhra Pradesh	2019-05- 31	3.65	11999139.0	43.24	Rural
1	Andhra Pradesh	2019-06- 30	3.05	11755881.0	42.05	Rural
2	Andhra Pradesh	2019-07- 31	3.75	12086707.0	43.50	Rural



#### Outlier detection

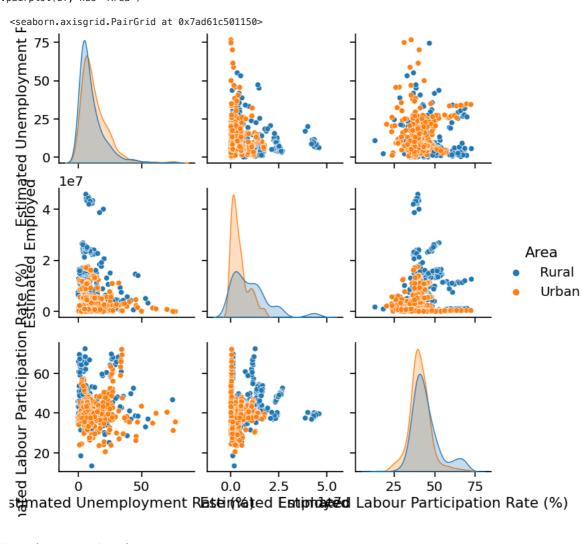
```
colors = ['lightblue', 'lightgreen', 'lightcoral']
# Create a figure with three subplots
plt.figure(figsize=(12, 6))
# Subplot 1: Unemployment Rate
plt.subplot(131)
df.boxplot(column='Estimated Unemployment Rate (%)', patch_artist=True)
plt.gca().get_children()[0].set_facecolor(colors[0])  # Set the color of the first box
plt.title('Unemployment Rate')
# Subplot 2: Employed
plt.subplot(132)
df.boxplot(column='Estimated Employed', patch_artist=True)
\verb|plt.gca().get_children()[0].set_facecolor(colors[1])| # Set the color of the second box|
plt.title('Employed')
# Subplot 3: Labor Participation Rate
plt.subplot(133)
df.boxplot(column='Estimated Labour Participation Rate (%)', patch_artist=True)
\verb|plt.gca().get_children()[0].set_facecolor(colors[2])| # Set the color of the third box|
plt.title('Labor Participation Rate')
plt.tight_layout()
plt.show()
```



✓ Unemployment rate in India during Covid-19

```
plt.figure(figsize=(12, 6))
sns.lineplot(data=df, x='Date', y='Estimated Unemployment Rate (%)')
plt.xticks(rotation=45)
plt.show()
```





# Unemployment rate in each state

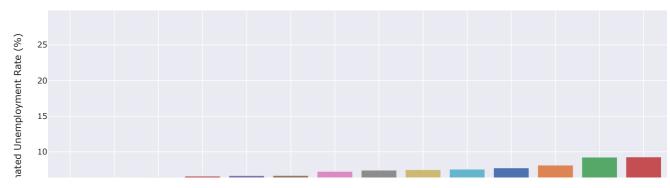
```
import plotly.express as px
plot_unemp = df[['Estimated Unemployment Rate (%)','Region']]
df_unemployed = plot_unemp.groupby('Region').mean().reset_index()

df_unemployed = df_unemployed.sort_values('Estimated Unemployment Rate (%)')

fig = px.bar(df_unemployed, x='Region',y='Estimated Unemployment Rate (%)',color = 'Region',title = 'Average unemployment ratemplate='seaborn')

fig.show()
```

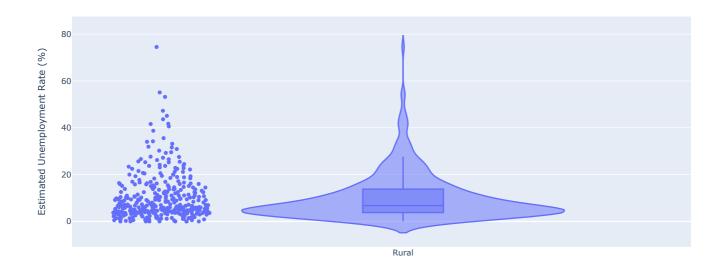
Average unem



Visualizes the distribution of unemployment rates within different areas

```
fig = px.violin(
    df,
    x='Area',
    y='Estimated Unemployment Rate (%)',
    title='Distribution of Unemployment Rates by Areas',
    box=True, # Include box plot inside the violin
    points='all', # Show individual data points
)
fig.show()
```

# Distribution of Unemployment Rates by Areas

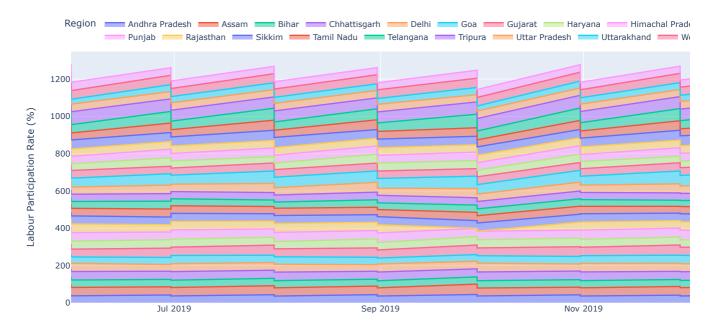


Composition of Labour Participation Rates by Region Over Time

```
fig = px.area(
    df,
    x='Date',
    y='Estimated Labour Participation Rate (%)',
    color='Region',
    labels={'Estimated Labour Participation Rate (%)': 'Labour Participation Rate (%)'},
    category_orders={'Region': df['Region'].unique()} # Preserve the order of regions
)

fig.update_layout(
    xaxis_title='Date',
    yaxis_title='Labour Participation Rate (%)',
    legend_title='Region',
    legend_dict(orientation="h", yanchor="bottom", y=1.02, xanchor="right", x=1),
)
```

fig.show()



## Extracting month from date

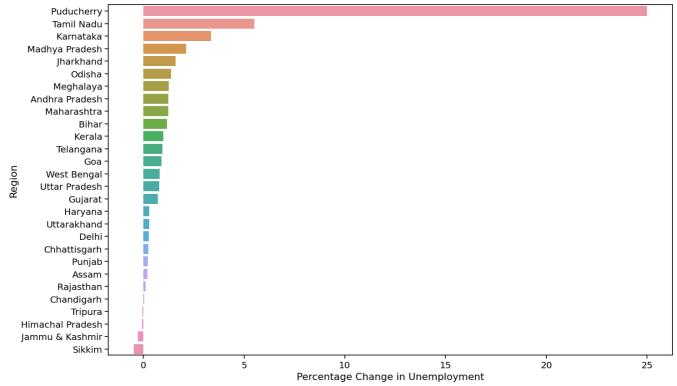
df['Month'] = df['Date'].dt.month
df

	Region	Date	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Area	Month	
0	Andhra Pradesh	2019-05- 31	3.65	11999139.0	43.24	Rural	5.0	11.
1	Andhra Pradesh	2019-06- 30	3.05	11755881.0	42.05	Rural	6.0	
2	Andhra Pradesh	2019-07- 31	3.75	12086707.0	43.50	Rural	7.0	
3	Andhra Pradesh	2019-08- 31	3.32	12285693.0	43.97	Rural	8.0	
4	Andhra Pradesh	2019-09- 30	5.17	12256762.0	44.68	Rural	9.0	
763	NaN	NaT	NaN	NaN	NaN	NaN	NaN	
764	NaN	NaT	NaN	NaN	NaN	NaN	NaN	
765	NaN	NaT	NaN	NaN	NaN	NaN	NaN	

#### Percentage change in unemployment

plot\_df = before\_lock.sort\_values('Percentage Change in Unemployment', ascending=False)
plt.figure(figsize=(16, 10))
sns.barplot(data=plot\_df, y='Region', x='Percentage Change in Unemployment')

<Axes: xlabel='Percentage Change in Unemployment', ylabel='Region'>



If the percentage change is positive (+X%), it means that unemployment has increased by X% compared to the previous period. In other words, more people are unemployed.

If the percentage change is negative (-X%), it means that unemployment has decreased by X% compared to the previous period. Fewer people are unemployed.

The magnitude of the percentage change indicates how significant the change is. A larger percentage change suggests a more substantial shift in unemployment rates compared to a smaller percentage change.

Puducherry's unemployment rate had been seriously impacted by the lock-down.

Sikkim, Jammu & Kashmir Chattisgar ,Himachal Pradesh and Tripur have negetive percentage change. That means these states are not highly impacted by the lock down.

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