Unemployment Analysis With Python • Objective:Unemployment is measured by the unemployment rate which is the number of people who are unemployed as a percentage of the total labour force. • Dataset Link:https://www.kaggle.com/datasets/gokulrajkmv/unemployment-in-india • Dataset Information: This dataset contains the unemployment rate of all the states in India Region = states in India Date = date which the unemployment rate observed Frequency = measuring frequency (Monthly) Estimated Unemployment Rate (%) = percentage of people unemployed in each States of India Estimated Employed = percentage of people employed Estimated Labour Participation Rate (%) = labour force participation rate by dividing the number of people actively participating in the labour force by the total number of people eligible to participate in the labor force force · Steps we follow set the working directory Import the required library set • Checking and cleaning the dataset • Unemployment rate analysis Conclusions **Setting Working Directories** import os In [2]: os.chdir("H:\\Data Science\\Internship\\Oasis Infobyte") Importing required libraries In [3]: **import** pandas **as** pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns import plotly.express as px ## Supress warnings import warnings warnings.filterwarnings("ignore") data = pd.read_csv("unemployment.csv") print("data has been successfully loaded") data has been successfully loaded Checking and cleaning the dataset data Region **Date Frequency** Estimated Unemployment Rate (%) Estimated Employed Estimated Labour Participation Rate (%) Region.1 longitude latitude 0 Andhra Pradesh 31-01-2020 5.48 16635535 41.02 South 15.9129 79.740 5.83 16545652 1 Andhra Pradesh 29-02-2020 40.90 15.9129 79.740 South 15881197 2 Andhra Pradesh 31-03-2020 5.79 39.18 South 15.9129 79.740 3 Andhra Pradesh 30-04-2020 20.51 11336911 33.10 15.9129 79.740 South 4 Andhra Pradesh 31-05-2020 17.43 12988845 36.46 South 15.9129 79.740 West Bengal 30-06-2020 30726310 262 7.29 40.39 East 22.9868 87.855

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In [5]: Out[5]: West Bengal 31-07-2020 6.83 35372506 46.17 87.855 263 East 22.9868 264 West Bengal 31-08-2020 14.87 33298644 47.48 22.9868 87.855 East West Bengal 30-09-2020 9.35 35707239 87.855 265 47.73 22.9868 East 33962549 266 West Bengal 31-10-2020 9.98 45.63 East 22.9868 87.855 267 rows × 9 columns In [6]: data.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 267 entries, 0 to 266 Data columns (total 9 columns): Non-Null Count Dtype # Column 0 Region 267 non-null object 267 non-null Date object 1 Frequency 267 non-null 2 object Estimated Unemployment Rate (%) float64 3 267 non-null Estimated Employed 4 267 non-null int64 Estimated Labour Participation Rate (%) 5 267 non-null float64 267 non-null 6 Region.1 object 267 non-null 7 longitude float64 latitude 267 non-null float64 dtypes: float64(4), int64(1), object(4)memory usage: 18.9+ KB In [7]: data.shape (267, 9)Out[7]: data.describe() Estimated Unemployment Rate (%) Estimated Employed Estimated Labour Participation Rate (%) Out[8]: longitude latitude 267.000000 2.670000e+02 267.000000 267.000000 267.000000 count 12.236929 1.396211e+07 41.681573 22.826048 80.532425 mean 10.803283 1.336632e+07 7.845419 6.270731 5.831738 std min 0.500000 1.175420e+05 16.770000 10.850500 71.192400 4.845000 2.838930e+06 37.265000 18.112400 76.085600 25% 9.732417e+06 **50**% 9.650000 40.390000 23.610200 79.019300

75% 16.755000 2.187869e+07 44.055000 27.278400 85.279900 75.850000 5.943376e+07 33.778200 69.690000 92.937600 max · we got statistical summary of our dataset Let's see if this dataset contains missing values or not: Region Date 0 Frequency Estimated Unemployment Rate (%) Estimated Employed Estimated Labour Participation Rate (%) Region.1 longitude 0 latitude 0 dtype: int64 • While analyzing the missing values, I found that the column names are not correct. So, for a better understanding of this data, I will rename all the columns: data.columns= ["States", "Date", "Frequency", "Estimated Unemployment Rate", "Estimated Employed", "Estimated Labour Participation Rate", "Region", "longitude", "latitude"] Now let's have a look at the correlation between the features of this dataset:

In [9]: | print(data.isnull().sum()) Heatmap plt.style.use('seaborn-whitegrid') plt.figure(figsize=(12, 10)) sns.heatmap(data.corr()) plt.show() Estimated Unemployment Rate 8.0 Estimated Employed 0.6 Estimated Labour Participation Rate longitude latitude

Unemployment Rate Analysis: Data Visualization • Now let's visualize the data to analyze the unemployment rate. I will first take a look at the estimated number of employees according to different regions of India: data.columns= ["States", "Date", "Frequency", "Estimated Unemployment Rate", "Estimated Employed", "Estimated Labour Participation Rate", "Region", "longitude", "latitude"] plt.title("Indian Unemployment") sns.histplot(x="Estimated Employed", hue="Region", data=data) plt.show() Indian Unemployment Region 35 South Northeast 30 East West 25 <u>≒</u>20 Estimated Employed Now let's see the unemployment rate according to different regions of India: In [13]: plt.figure(figsize=(12, 10)) plt.title("Indian Unemployment") sns.histplot(x="Estimated Unemployment Rate", hue="Region", data=data) Indian Unemployment Region South Northeast 20.0 East West North 17.5 15.0 12.5 10.0 7.5 5.0

2.5 40 50 Estimated Unemployment Rate Now let's create a dashboard to analyze the unemployment rate of each Indian state by region. For this, I'll use a sunburst plot: unemploment = data[["States", "Region", "Estimated Unemployment Rate"]] figure = px.sunburst(unemploment, path=["Region", "States"], values="Estimated Unemployment Rate", width=700, height=700, color_continuous_scale="RdY1Gn", title="Unemployment Rate in India") figure.show() ilii Unemployment Rate in India Puducherry Tamil Nadu Kerala South Andhra Pradesh Gujarat Madhya Pradesh West Karnataka Chhattisgarh East Northeast Jharkhand Goa Bihar Tripura Which Region has the most data color = sns.color_palette() In [15]: cnt_srs = data.Region.value_counts() plt.figure(figsize=(12,8)) sns.barplot(cnt_srs.index, cnt_srs.values, alpha=0.8, color=color[4]) plt.ylabel('Number of Occurrences', fontsize=12) plt.xlabel('States', fontsize=12) plt.title('Count the states', fontsize=15) plt.xticks(rotation='vertical') plt.show() Count the states 80 70 60 Number of Occurrences

20

10

plt.show()

take the mean of rate Region by Region

plt.figure(figsize=(12,8))

plt.ylabel('Mean rate', fontsize=12) plt.xlabel('States', fontsize=12)

plt.xticks(rotation='vertical')

plt.title("Average of mean", fontsize=15)

see the number of unique Region

Estimated Unemployment Rate

for index,row in df2.iterrows():

Conclusions:

Thank You!

15.889620

13.916000

10.950263 10.454667

8.239000

Northeast Region which Northeast has the highest yearly fluncation.

Calculate which models has highest yearly fluncations

East Region which East has the highest yearly fluncation. North Region which North has the highest yearly fluncation.

South Region which South has the highest yearly fluncation. West Region which West has the highest yearly fluncation.

• Northeast Region which Northeast has the highest yearly fluncation.

• East Region which East has the highest yearly fluncation. • North Region which North has the highest yearly fluncation.

• South Region which South has the highest yearly fluncation. • West Region which West has the highest yearly fluncation.

Suggestions are always Welcome!

In [17]: data.Region.nunique()

See exact numbers

print(topstate)

Region North

Northeast

East

South

West

Out[17]:

States

In [16]: grouped_df = data.groupby(["Region"])["Estimated Unemployment Rate"].aggregate("mean").reset_index()

Average of mean

In [18]: make_total = data.pivot_table("Estimated Unemployment Rate",index=['Region'],aggfunc='mean') topstate=make_total.sort_values(by='Estimated Unemployment Rate', ascending=False)[:47]

In [19]: maketotal_1 = data.pivot_table(values='Estimated Unemployment Rate',index=['Region'],aggfunc=np.std)

print(row['Region'], "Region which", row['Region'], "has the highest yearly fluncation.")

• Unemployment is measured by the unemployment rate which is the number of people who are unemployed as a percentage of the total labour force.

df1 = maketotal_1.reset_index().dropna(subset=['Estimated Unemployment Rate']) df2 = df1.loc[df1.groupby('Region')['Estimated Unemployment Rate'].idxmax()]

• So this is how you can analyze the unemployment rate by using the Python programming language.

sns.pointplot(grouped_df['Region'].values, grouped_df['Estimated Unemployment Rate'].values, alpha=0.8, color=color[2])