Oasis Infobyte Internship, Sep-2023 Sonali Kotlapure, Data Science Intern Unemployment Analysivs Untitopy Payt no pasured 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

OASIS

INFOBYTE

41.02

40.90

39.18

33.10

36.46

40.39

46.17

47.48

47.73

45.63

South

South

South

South

South

East

East

East

East

East

15.9129 79.740

15.9129 79.740

15.9129 79.740

15.9129 79.740

79.740

87.855

87.855

87.855

87.855

87.855

15.9129

22.9868

22.9868

22.9868

22.9868

22.9868

In [5]: Out[5]: 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 5.83 16545652 1 Andhra Pradesh 29-02-2020 15881197 2 Andhra Pradesh 31-03-2020 5.79 3 Andhra Pradesh 30-04-2020 20.51 11336911 4 Andhra Pradesh 31-05-2020 17.43 12988845 30726310 262 West Bengal 30-06-2020 7.29

West Bengal 31-07-2020 6.83 35372506 263 264 West Bengal 31-08-2020 14.87 33298644 West Bengal 30-09-2020 9.35 35707239 265 33962549 266 West Bengal 31-10-2020 9.98 267 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 object

267 non-null

object

float64

float64

object

float64

float64

longitude

22.826048

6.270731

10.850500

18.112400

23.610200

27.278400

33.778200

267.000000 267.000000

267.000000

41.681573

7.845419

16.770000

37.265000

40.390000

44.055000

69.690000

latitude

80.532425

5.831738

71.192400

76.085600

79.019300

85.279900

92.937600

8.0

0.6

int64

Date 1 Frequency 2 Estimated Unemployment Rate (%) 3 Estimated Employed 4 Estimated Labour Participation Rate (%) 5 6 Region.1 7 longitude latitude 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]: 267.000000 2.670000e+02 count 12.236929 1.396211e+07 mean 10.803283 1.336632e+07 std min 0.500000 1.175420e+05 4.845000 2.838930e+06 25% 9.732417e+06 **50**% 9.650000

75% 16.755000 2.187869e+07 75.850000 5.943376e+07 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"]

In [9]: | print(data.isnull().sum()) Heatmap plt.style.use('seaborn-whitegrid') plt.figure(figsize=(12, 10)) sns.heatmap(data.corr()) plt.show()

Now let's have a look at the correlation between the features of this dataset: Estimated Unemployment Rate Estimated Employed Estimated Labour Participation Rate

longitude latitude

Unemployment Rate Analysis: Data Visualization

data.columns= ["States", "Date", "Frequency",

plt.title("Indian Unemployment")

plt.show()

35

30

25

ቯ 20

"longitude", "latitude"]

Indian Unemployment

Estimated Employed

sns.histplot(x="Estimated Employed", hue="Region", data=data)

"Estimated Unemployment Rate", "Estimated Employed", "Estimated Labour Participation Rate", "Region",

Region

40

width=700, height=700, color_continuous_scale="RdY1Gn",

values="Estimated Unemployment Rate",

title="Unemployment Rate in India")

50

South Northeast

East West

• 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:

Region South Northeast

ilii

Gujarat

Madhya Pradesh

Chhattisgarh

Goa

West

Northeast

Tripura

Count the states

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])

East West North

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 20.0 17.5 15.0 12.5 10.0 7.5 5.0 2.5 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"],

figure.show()

Unemployment Rate in India

Puducherry

Tamil Nadu

Andhra Pradesh

Jharkhand

Karnataka

Which Region has the most data

cnt_srs = data.Region.value_counts()

plt.xlabel('States', fontsize=12)

plt.xticks(rotation='vertical')

plt.ylabel('Number of Occurrences', fontsize=12)

plt.title('Count the states', fontsize=15)

take the mean of rate Region by Region

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

plt.show()

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

East Northeast

South

West

Out[17]: 5

color = sns.color_palette()

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

plt.show()

80

70

60

Number of Occurrences

20

10

In [15]:

South

Bihar

sns.barplot(cnt_srs.index, cnt_srs.values, alpha=0.8, color=color[4])

East

Kerala