unemployment-in-india

May 5, 2024

1 Unemployment Analysis in India during covid pandamic

- Unemployment is measured by the unemployment rate which is the number of people who are unemployed as a percentage of the total labour force.
- During the covid-19 period there was a sharp increase in the unemployment rate.
- The aim is to analyze the unemployment rate using python.

The dataset contains information about unemployment across different states in India during the COVID-19 pandemic. It includes:

States: Various regions within India where unemployment rates were measured. Date: Specific recording dates of unemployment rates. Measuring Frequency: How often measurements were collected (monthly). Estimated Unemployment Rate (%): The percentage of unemployed individuals in each state. Estimated Employed Individuals: The number of people currently engaged in employment. Estimated Labour Participation Rate (%): The percentage of the working-age population actively participating in the job market. This dataset is valuable for understanding how unemployment rates fluctuated throughout the pandemic across different parts of India. It provides insights into the impacts of the pandemic on employment numbers and labor force participation.

The goal of this analysis is to examine the widespread effects of the COVID-19 pandemic on India's employment landscape. By studying this dataset, we aim to gain insights into how unemployment rates, employment figures, and labor participation rates were affected during this challenging period. This analysis will shed light on the socio-economic consequences of the pandemic on India's workforce and labor market dynamics.

[1]:
```
#import required libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import datetime as dt
import calendar
import plotly.graph_objects as go

import warnings
warnings.filterwarnings("ignore")
%matplotlib inline
```

Load the csv file into a pandas dataframe

[2]:
```
df = pd.read_csv("Unemployment_in_india.csv")
df.head()
```

[2]:

	Region	Date	Frequency	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Region.1	longitude	latitude
0	Andhra Pradesh	31-01-2020	M	5.48	16635535	41.02	South	15.9129	79.74
1	Andhra Pradesh	29-02-2020	M	5.83	16545652	40.90	South	15.9129	79.74
2	Andhra Pradesh	31-03-2020	M	5.79	15881197	39.18	South	15.9129	79.74
3	Andhra Pradesh	30-04-2020	M	20.51	11336911	33.10	South	15.9129	79.74
4	Andhra Pradesh	31-05-2020	M	17.43	12988845	36.46	South	15.9129	79.74

[3]:
```
df.tail()
```

[3]:

	Region	Date	Frequency	Estimated Unemployment Rate (%)	Estimated Employed	Estimated Labour Participation Rate (%)	Region.1	longitude	latitude
262	West Bengal	30-06-2020	M	7.29	30726310	40.39	East	22.9868	87.855
263	West Bengal	31-07-2020	M	6.83	35372506	46.17	East	22.9868	87.855
264	West Bengal	31-08-2020	M	14.87	33298644	47.48	East	22.9868	87.855
265	West Bengal	30-09-2020	M	9.35	35707239	47.73	East	22.9868	87.855
266	West Bengal	31-10-2020	M	9.98	33962549	45.63	East	22.9868	87.855

[4]:
```
df.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
```

1.0.1 Renaming the attributes

1. Region = state
2. Date = date
3. Frequency = frequency
4. Estimated Unemployment Rate (%) = estimated unemployment rate
5. Estimated Employed = estimated employment
6. Estimated Labour Participation Rate (%) = estimated labour participation rate
7. Region.1 = region
8. longitude = longitude
9. latitude = latitude

Updating column names:

[5]:
```
df.columns = ['state','date','frequency','estimated unemployment␣
 ↪rate','estimated employed','estimated labour participation␣
 ↪rate','region','longitude','latitude']
df.head()
```

[5]:

	state	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	longitude	\
0	16635535	41.02	South	15.9129	
1	16545652	40.90	South	15.9129	

```
2            15881197                                  39.18  South    15.9129
3            11336911                                  33.10  South    15.9129
4            12988845                                  36.46  South    15.9129

   latitude
0     79.74
1     79.74
2     79.74
3     79.74
4     79.74
```

Revealing basic information of the dataset

[6]:
```
df.shape
```

[6]:
```
(267, 9)
```

[7]:
```
df.columns
```

[7]:
```
Index(['state', 'date', 'frequency', 'estimated unemployment rate',
       'estimated employed', 'estimated labour participation rate', 'region',
       'longitude', 'latitude'],
      dtype='object')
```

[8]:
```
df.describe()
```

[8]:
```
       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
```

[9]:
```
df.isnull().sum()
```

[9]:
```
state                                0
date                                 0
frequency                            0
estimated unemployment rate          0
estimated employed                   0
estimated labour participation rate  0
region                               0
longitude                            0
latitude                             0
dtype: int64
```

[10]:
```
df.duplicated().any()
```

[10]:
```
False
```

[11]:
```
df.state.value_counts()
```

[11]:
```
state
Andhra Pradesh      10
Assam               10
Uttarakhand         10
Uttar Pradesh       10
Tripura             10
Telangana           10
Tamil Nadu          10
Rajasthan           10
Punjab              10
Puducherry          10
Odisha              10
Meghalaya           10
Maharashtra         10
Madhya Pradesh      10
Kerala              10
Karnataka           10
Jharkhand           10
Himachal Pradesh    10
Haryana             10
Gujarat             10
Goa                 10
Delhi               10
Chhattisgarh        10
Bihar               10
West Bengal         10
Jammu & Kashmir      9
Sikkim               8
Name: count, dtype: int64
```

1.0.2 Changing the datatype of 'date' from object to datetime

[12]:
```
df['date'] = pd.to_datetime(df['date'],dayfirst = True)
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 267 entries, 0 to 266
Data columns (total 9 columns):
 #   Column                                   Non-Null Count  Dtype
---  ------                                   --------------  -----
 0   state                                    267 non-null    object
 1   date                                     267 non-null    datetime64[ns]
 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                                   267 non-null    object
 7   longitude                                267 non-null    float64
 8   latitude                                 267 non-null    float64
dtypes: datetime64[ns](1), float64(4), int64(1), object(3)
memory usage: 18.9+ KB
```

1.0.3 Extracting month from date attribute

[13]:
```
df['month_int'] = df['date'].dt.month
df.head()
```

[13]:

	state	date	frequency	estimated unemployment rate	estimated employed	estimated labour participation rate	region	longitude	latitude	month_int
0	Andhra Pradesh	2020-01-31	M	5.48	16635535	41.02	South	15.9129	79.74	1
1	Andhra Pradesh	2020-02-29	M	5.83	16545652	40.90	South	15.9129	79.74	2
2	Andhra Pradesh	2020-03-31	M	5.79	15881197	39.18	South	15.9129	79.74	3
3	Andhra Pradesh	2020-04-30	M	20.51	11336911	33.10	South	15.9129	79.74	4
4	Andhra Pradesh	2020-05-31	M	17.43	12988845	36.46	South	15.9129	79.74	5

The months are in integer datetype. We need to convert the months into words for better analysis,

[14]:
```
df['month'] = df['month_int'].apply(lambda x: calendar.month_abbr[x])
df.head()
```

[14]:

	state	date	frequency	estimated unemployment rate	estimated employed	estimated labour participation rate	region	longitude	latitude	month_int	month
0	Andhra Pradesh	2020-01-31	M	5.48	16635535	41.02	South	15.9129	79.74	1	Jan
1	Andhra Pradesh	2020-02-29	M	5.83	16545652	40.90	South	15.9129	79.74	2	Feb
2	Andhra Pradesh	2020-03-31	M	5.79	15881197	39.18	South	15.9129	79.74	3	Mar
3	Andhra Pradesh	2020-04-30	M	20.51	11336911	33.10	South	15.9129	79.74	4	Apr
4	Andhra Pradesh	2020-05-31	M	17.43	12988845	36.46	South	15.9129	79.74	5	May

Numeric data grouped by months

[15]:
```
data = df.groupby(['month'])[['estimated unemployment rate','estimated␣
 ↪employed','estimated labour participation rate']].mean()
data=pd.DataFrame(data).reset_index()
```

Bar plot of unemployment rate and labour participation rate

[16]:
```
month = data.month
unemployment_rate = data['estimated unemployment rate']
labour_participation_rate = data['estimated labour participation rate']

fig = go.Figure()

fig.add_trace(go.Bar(x = month,y = unemployment_rate,name = 'Unemployment␣
 ↪Rate'))
fig.add_trace(go.Bar(x = month,y = labour_participation_rate,name = 'Labour␣
 ↪Participation Rate'))

fig.update_layout(title = 'Unemployment Rate and Labour Participation',
                  xaxis = {'categoryorder':'array','categoryarray':
 ↪['Jan','Feb','Mar','Apr','May','Jun','Jul','Aug','Sep','Oct']}      )
```

```
fig.show()
```

Bar plot of estimated employed citizen in every month

[17]:
```
import plotly.express as px
```

[18]:
```
fig = px.bar(data,x='month',y='estimated employed',color='month',
             category_orders ={'month':
 ↪['Jan','Feb','Mar','Apr','May','Jun','Jul','Aug','Sep','Oct']},
             title='Estimated employed people from Jan 2020 to Oct 2020')
fig.show()
```

1.1 State wise Analysis

[19]:
```
state =  df.groupby(['state'])[['estimated unemployment rate','estimated␣
 ↪employed','estimated labour participation rate']].mean()
state = pd.DataFrame(state).reset_index()
```

[20]:
```
# Box plot

fig = px.box(data_frame=df,x='state',y='estimated unemployment␣
 ↪rate',color='state',title='Unemployment rate')
fig.update_layout(xaxis={'categoryorder':'total descending'})
fig.show()
```

[21]:
```
# average unemployment rate bar plot

fig = px.bar(state,x='state',y='estimated unemployment␣
 ↪rate',color='state',title='Average unemployment rate (State)')
fig.update_layout(xaxis={'categoryorder':'total descending'})
fig.show()
```

Hariyana and Tripura were having the highest average amount of Unemployment rate

Meghalaya was having the lowest average amount of Unemployment rate

[22]:
```
# Bar plot Unemployment Rate (monthly)

fig = px.bar(df,x='state',y='estimated unemployment␣
 ↪rate',animation_frame='month',color='state',
             title='Unemployment rate from Jan 2020 to Oct 2020(State)')

fig.update_layout(xaxis={'categoryorder':'total descending'})
fig.show()
```

[23]:
```
# Filter data before and during lockdown
before_lockdown = df[df['date'] < '2020-03-25']
```

```
during_lockdown = df[df['date'] >= '2020-03-25']

# Average Unemployment Rate before and during lockdown
avg_unemployment_before = before_lockdown['estimated unemployment rate'].mean()
avg_unemployment_during = during_lockdown['estimated unemployment rate'].mean()

print(f"Average Unemployment Rate before lockdown: {avg_unemployment_before:.
 ↪2f}%")
print(f"Average Unemployment Rate during lockdown: {avg_unemployment_during:.
 ↪2f}%")

# Percentage change in Unemployment Rate
percentage_change = ((avg_unemployment_during - avg_unemployment_before) /␣
 ↪avg_unemployment_before) * 100
print(f"Percentage Change in Unemployment Rate: {percentage_change:.2f}%")
```

```
Average Unemployment Rate before lockdown: 9.23%
Average Unemployment Rate during lockdown: 12.96%
Percentage Change in Unemployment Rate: 40.43%
```

Monthly unemployment rate

[24]:
```
fig=px.scatter_geo(df,'longitude','latitude',color='state',
                   hover_name='state',size='estimated unemployment rate',
                   animation_frame='month',scope='asia',title='Impact of␣
 ↪lockdown on employment in India')

fig.layout.updatemenus[0].buttons[0].args[1]['frame']['duration'] =2000
fig.
 ↪update_geos(lataxis_range=[5,40],lonaxis_range=[65,100],oceancolor='lightblue',
               showocean=True)

fig.show()
```

1.2 Regional Analysis

[25]:
```
df.region.unique()
```

[25]:
```
array(['South', 'Northeast', 'East', 'West', 'North'], dtype=object)
```

[26]:
```
# numeric data grouped by region

region = df.groupby(['region'])[['estimated unemployment rate','estimated␣
 ↪employed','estimated labour participation rate']].mean()
region = pd.DataFrame(region).reset_index()
```

[27]:
```
import plotly.express as px

# Specify dimensions and color parameter for the scatter matrix plot
dimensions = ['estimated unemployment rate', 'estimated employed', 'estimated␣
 ↪labour participation rate']
color_column = 'region'

# Create scatter matrix plot with Plotly Express
fig = px.scatter_matrix(
    df,
    dimensions=dimensions,
    color=color_column,
    title='Scatter Matrix Plot Colored by Region'
)

# Display the plot
fig.show()
```

[28]:
```
import dash
import dash_core_components as dcc
import dash_html_components as html

app = dash.Dash(__name__)

# Define layout
app.layout = html.Div([
    dcc.Graph(
        id='unemployment-trend',
        figure={
            'data': [
                {'x': df['date'], 'y': df['estimated unemployment rate'],␣
 ↪'type': 'line', 'name': 'Unemployment Rate'}
            ],
            'layout': {
                'title': 'Unemployment Rate Over Time'
            }
        }
    )
])

if __name__ == '__main__':
    app.run_server(debug=True)
```

```
<IPython.lib.display.IFrame at 0x1686911ac90>
```

[29]:
```
# Average Unemployment Rate
```

```
fig = px.bar(region,x='region',y='estimated unemployment␣
 ↪rate',color='region',title='Average unemployment rate(region)')
fig.update_layout(xaxis={'categoryorder':'total descending'})
fig.show()
```

[30]:
```
fig = px.bar(df,x='region',y='estimated unemployment␣
 ↪rate',animation_frame='month',color='state',
            title='Unemployment rate from Jan 2020 to Oct 2020')

fig.update_layout(xaxis={'categoryorder':'total descending'})
fig.layout.updatemenus[0].buttons[0].args[1]['frame']['duration'] =2000

fig.show()
```

[31]:
```
unemployment =df.groupby(['region','state'])['estimated unemployment rate'].
 ↪mean().reset_index()
unemployment.head()
```

[31]:

	region	state	estimated unemployment rate
0	East	Bihar	19.471
1	East	Jharkhand	19.539
2	East	Odisha	6.462
3	East	West Bengal	10.192
4	North	Delhi	18.414

[32]:
```
fig = px.sunburst(unemployment,path=['region','state'],values='estimated␣
 ↪unemployment rate',
                  title ='Unemployment rate in state and region',height=600)
fig.show()
```

1.3 Unemployment rate before and after Lockdown

[33]:
```
# data representation before and after lockdown

before_lockdown = df[(df['month_int']>=1) &(df['month_int'] <4)]
after_lockdown = df[(df['month_int'] >=4) & (df['month_int'] <=6)]
```

[34]:
```
af_lockdown = after_lockdown.groupby('state')['estimated unemployment rate'].
 ↪mean().reset_index()

lockdown = before_lockdown.groupby('state')['estimated unemployment rate'].
 ↪mean().reset_index()
lockdown['unemployment rate before lockdown'] = af_lockdown['estimated␣
 ↪unemployment rate']
```

```
lockdown.columns = ['state','unemployment rate before lockdown','unemployment␣
 ↪rate after lockdown']
lockdown.head()
```

[34]:

	state	unemployment rate before lockdown	unemployment rate after lockdown
0	Andhra Pradesh	5.700000	13.750000
1	Assam	4.613333	7.070000
2	Bihar	12.110000	36.806667
3	Chhattisgarh	8.523333	9.380000
4	Delhi	18.036667	25.713333

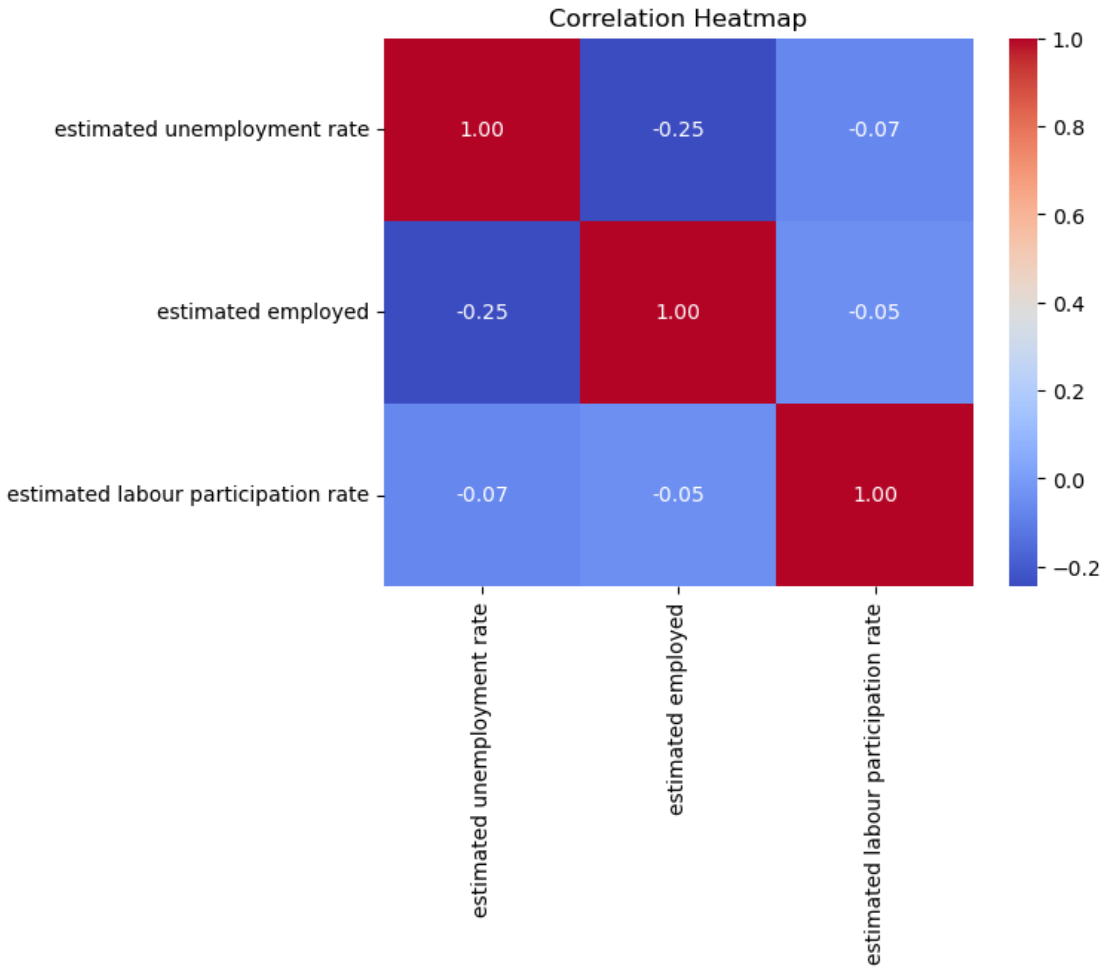
[35]:
```
# unenployment rate change after lockdown

lockdown['rate change in unemployment'] =round(lockdown['unemployment rate␣
 ↪before lockdown']-lockdown['unemployment rate before lockdown']
                                                    /lockdown['unemployment rate␣
 ↪after lockdown'],2)
```

[36]:
```
fig = px.bar(lockdown,x='state',y='rate change in unemployment',color='rate␣
 ↪change in unemployment',
             title='Percentage change in Unemployment rate in each state after␣
 ↪lockdown',template='ggplot2')
fig.update_layout(xaxis={'categoryorder':'total ascending'})
fig.show()
```

[37]:
```
# Calculate correlation matrix
correlation_matrix = df[['estimated unemployment rate', 'estimated employed',␣
 ↪'estimated labour participation rate']].corr()

# Plot correlation heatmap
sns.heatmap(correlation_matrix, annot=True, cmap='coolwarm', fmt='.2f')
plt.title('Correlation Heatmap')
plt.show()
```



[38]:
```
from statsmodels.tsa.arima.model import ARIMA

# Fit ARIMA model
model = ARIMA(df['estimated unemployment rate'], order=(1, 1, 1))
model_fit = model.fit()

# Forecast future unemployment rates
forecast = model_fit.forecast(steps=12)
print(forecast)
```

```
267    11.084691
268    11.640383
269    11.919913
270    12.060524
271    12.131256
272    12.166836
273    12.184734
```

```
274    12.193737
275    12.198266
276    12.200544
277    12.201690
278    12.202267
Name: predicted_mean, dtype: float64
```

[ ]: