

micaQuestionsFinal

August 21, 2022

0.1 First step: we import data and play with it a bit

```
[ ]: # Import and excel file using pandas
import pandas as pd
# Read Excel file
df = pd.read_excel('testData_MICA.xlsx')
# print(df)

df1 = df
df1 = df1.rename({'Migrant Persons' : 'Migrant Persons X 1e6'}, axis=1)
df1 = df1.rename({'Migrant Males' : 'Migrant Males %'}, axis=1)
df1['Migrant Males %'] = df1['Migrant Males %']/df1['Migrant Persons X 1e6']*100
print(df1.head()) # to test that all is properly imported
```

| | Home State Name | Home State ID | Current State \ |
|---|---------------------------|---------------|-----------------|
| 0 | DAMAN & DIU | 25 | JAMMU & KASHMIR |
| 1 | DADRA & NAGAR HAVELI | 26 | JAMMU & KASHMIR |
| 2 | PUDUCHERRY | 34 | JAMMU & KASHMIR |
| 3 | ANDAMAN & NICOBAR ISLANDS | 35 | JAMMU & KASHMIR |
| 4 | MIZORAM | 15 | JAMMU & KASHMIR |

| | Current State ID | Migrant Persons X 1e6 | Migrant Males % | Migrant Females |
|---|------------------|-----------------------|-----------------|-----------------|
| 0 | 1 | 6 | 50.000000 | 3 |
| 1 | 1 | 17 | 23.529412 | 13 |
| 2 | 1 | 18 | 33.333333 | 12 |
| 3 | 1 | 51 | 49.019608 | 26 |
| 4 | 1 | 57 | 38.596491 | 35 |

```
[ ]: print(df['Current State'].unique())
print(df['Current State ID'].unique())
```

```
['JAMMU & KASHMIR' 'HIMACHAL PRADESH' 'PUNJAB' 'CHANDIGARH' 'UTTARAKHAND'
 'HARYANA' 'NCT OF DELHI' 'RAJASTHAN' 'UTTAR PRADESH' 'BIHAR' 'SIKKIM'
 'ARUNACHAL PRADESH' 'NAGALAND' 'MANIPUR' 'MIZORAM' 'TRIPURA' 'MEGHALAYA'
 'ASSAM' 'WEST BENGAL' 'JHARKHAND' 'ODISHA' 'CHHATTISGARH'
 'MADHYA PRADESH' 'GUJARAT' 'DAMAN & DIU' 'DADRA & NAGAR HAVELI'
 'MAHARASHTRA' 'ANDHRA PRADESH' 'KARNATAKA' 'GOA' 'LAKSHADWEEP' 'KERALA'
 'TAMIL NADU' 'PUDUCHERRY' 'ANDAMAN & NICOBAR ISLANDS']
```

```
[ 1  2  3  4  5  6  7  8  9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24
```

25 26 27 28 29 30 31 32 33 34 35]

Let us make an ID file. We will use this dataframe throughout the notebook.

```
[ ]: # initialize list elements
data = {'ID': df['Current State ID'].unique(),
        'State': df['Current State'].unique()}

IDs = pd.DataFrame(data)

# print dataframe.
print(IDs)
```

| | ID | State |
|----|----|---------------------------|
| 0 | 1 | JAMMU & KASHMIR |
| 1 | 2 | HIMACHAL PRADESH |
| 2 | 3 | PUNJAB |
| 3 | 4 | CHANDIGARH |
| 4 | 5 | UTTARAKHAND |
| 5 | 6 | HARYANA |
| 6 | 7 | NCT OF DELHI |
| 7 | 8 | RAJASTHAN |
| 8 | 9 | UTTAR PRADESH |
| 9 | 10 | BIHAR |
| 10 | 11 | SIKKIM |
| 11 | 12 | ARUNACHAL PRADESH |
| 12 | 13 | NAGALAND |
| 13 | 14 | MANIPUR |
| 14 | 15 | MIZORAM |
| 15 | 16 | TRIPURA |
| 16 | 17 | MEGHALAYA |
| 17 | 18 | ASSAM |
| 18 | 19 | WEST BENGAL |
| 19 | 20 | JHARKHAND |
| 20 | 21 | ODISHA |
| 21 | 22 | CHHATTISGARH |
| 22 | 23 | MADHYA PRADESH |
| 23 | 24 | GUJARAT |
| 24 | 25 | DAMAN & DIU |
| 25 | 26 | DADRA & NAGAR HAVELI |
| 26 | 27 | MAHARASHTRA |
| 27 | 28 | ANDHRA PRADESH |
| 28 | 29 | KARNATAKA |
| 29 | 30 | GOA |
| 30 | 31 | LAKSHADWEEP |
| 31 | 32 | KERALA |
| 32 | 33 | TAMIL NADU |
| 33 | 34 | PUDUCHERRY |
| 34 | 35 | ANDAMAN & NICOBAR ISLANDS |

0.1.1 Plotting the global data

Here, we will plot how many workers migrated from a given state and to which state.

```
[ ]: import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
import matplotlib.ticker as ticker

sns.set_theme(style="whitegrid")

# Load the example planets dataset
# planets = sns.load_dataset("planets")
fig, (ax1, ax2) = plt.subplots(1, 2)
fig.set_size_inches(25, 10)

xmin, xmax = min(IDs['ID'])-0.5, max(IDs['ID'])+0.5
ax1.plot([xmin, xmax], [xmin, xmin], 'k-', lw=3)
ax1.plot([xmax, xmax], [xmin, xmax], 'k-', lw=3)
ax1.plot([xmin, xmax], [xmax, xmax], 'k-', lw=3)
ax1.plot([xmin, xmin], [xmin, xmax], 'k-', lw=3)

cmap = sns.cubehelix_palette(rot=-.2, as_cmap=True)

g = sns.scatterplot(
    data=df1,
    x="Home State ID", y="Current State ID",
    size="Migrant Persons X 1e6", hue="Migrant Males %",
    palette=cmap, sizes=(100, 500), ax=ax1
)

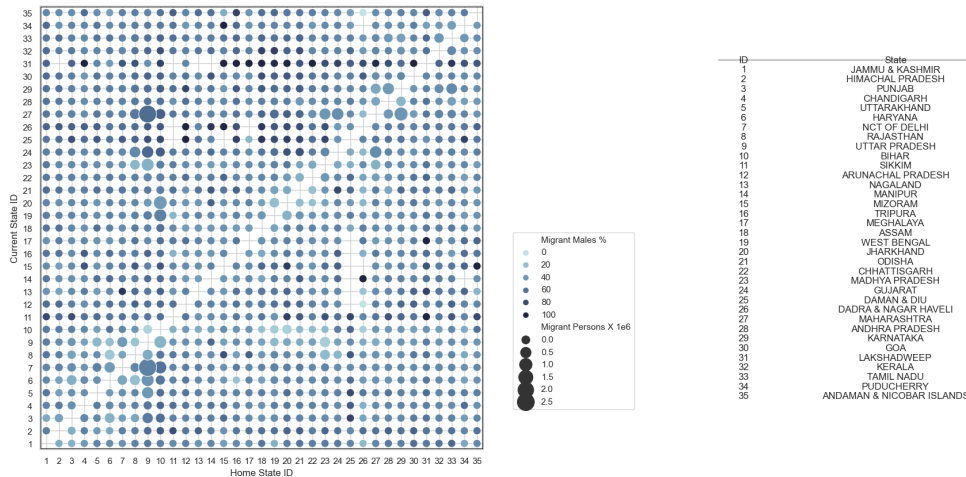
ax1.xaxis.grid(True, "minor", linewidth=.25)
ax1.yaxis.grid(True, "minor", linewidth=.25)
ax1.set_aspect('equal')
ax1.set_xlim(0.5, 35.5)
ax1.set_ylim(0.5, 35.5)
ax1.legend(bbox_to_anchor=(0.85, 0, 0.5, 0.5))

start, end = ax1.get_xlim()
ax1.xaxis.set_ticks(np.arange(start+0.5, end-0.5+1, 1))
ax1.xaxis.set_major_formatter(ticker.FormatStrFormatter('%d'))
ax1.yaxis.set_ticks(np.arange(start+0.5, end-0.5+1, 1))
ax1.yaxis.set_major_formatter(ticker.FormatStrFormatter('%d'))

ax2.axis('off')
ax2.table(cellText=IDs.values, colLabels=IDs.columns, loc='center',
         colWidths=[0.1, 0.5], cellLoc='center', edges='open')
```

```
import matplotlib.lines as lines
yL = 0.88
line = lines.Line2D([0.2, 0.7], [yL, yL],
                    lw=0.5, color='black', axes=ax2)
ax2.add_line(line)
```

```
[ ]: <matplotlib.lines.Line2D at 0x17f397fa0>
```



- As expected, the diagonal is empty. No one migrated to the state they were already in.
- The most common migration is from Uttar Pradesh to Delhi.
- In fact the number of people migrating from UP to all other states is higher in general.
- UP is followed very closely by Bihar.

0.1.2 Shortcoming of this plot:

We do not really see any details because of the large numbers included in this case. In the subsequent questions, we will make use of specific questions to understand the data better.

0.2 Question 1

Now we do question 1. To keep things simple, we will use brute force.

To start, let us first calculate the total number of workers migrating out from a particular state

```
[ ]: # Now, we get a sum of migrants from a particular Home State
df2em = df.groupby(['Home State ID']).sum()

# Plot Migrant Persons against Home State ID
# print(df2em.index)
fig, (ax, ax0) = plt.subplots(1,2)
fig.set_size_inches(25, 10)
```

```

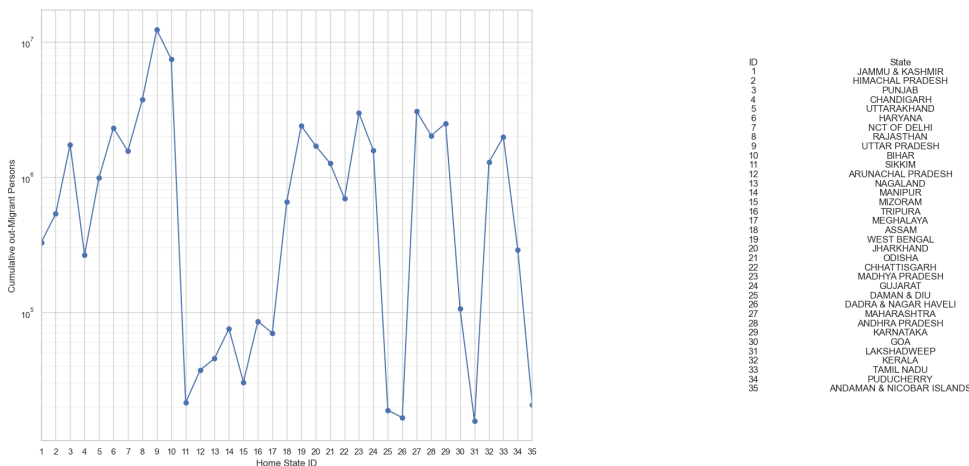
ax.plot(df2em.index, df2em['Migrant Persons'], 'o-')
ax.xaxis.grid(True, "minor", linewidth=.25)
ax.yaxis.grid(True, "minor", linewidth=.25)
ax.set_xlim(1, 35)
start, end = ax1.get_xlim()
ax.xaxis.set_ticks(np.arange(1, 36, 1))
ax.xaxis.set_major_formatter(ticker.FormatStrFormatter('%d'))
ax.set_xlabel('Home State ID')
ax.set_ylabel('Cumulative out-Migrant Persons')

# set yaxis to log scale
ax.set_yscale('log')

ax0.axis('off')
ax0.table(cellText=IDs.values, colLabels=IDs.columns, loc='center',
          colWidths=[0.1, 0.5], cellLoc='center', edges='open')

```

```
[ ]: <matplotlib.table.Table at 0x17f6dc040>
```



- Clearly, we see that by far the most workers are migrating out of Uttar Pradesh. This is followed by Bihar and Rajasthan. This is inline with our previous plot.

For better visualization, we will plot the above line graph onto the map of India. This way, we can keep track of the states and their migration patterns.

```

[ ]: import geopandas as gpd

df6id = df2em.index.to_numpy()
df6value = df2em['Migrant Persons'].to_numpy()

data = {'ID': df6id,
        'Migrant persons': df6value}

```

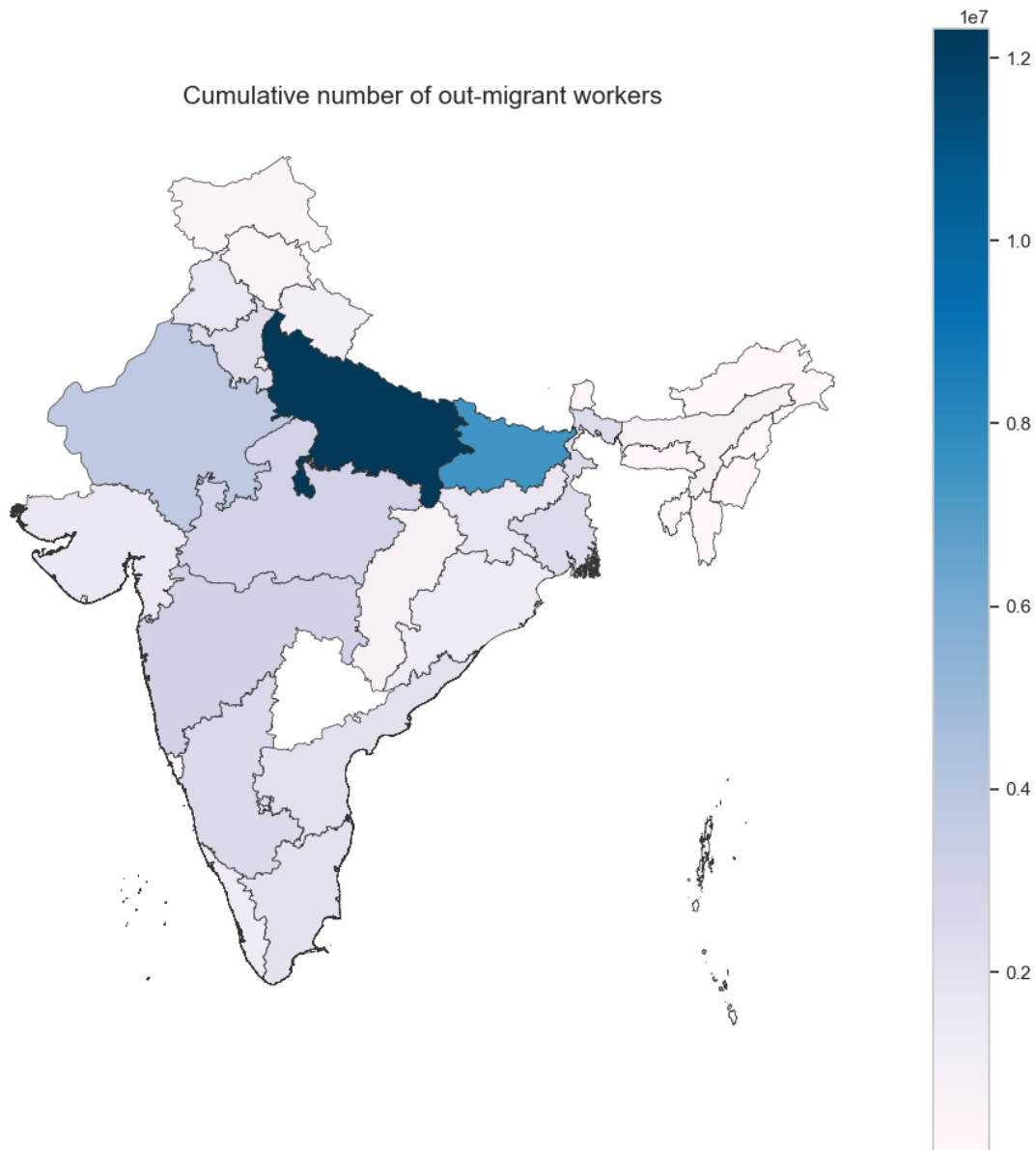
```

df6em = pd.DataFrame(data)

df7em = pd.merge(IDs, df6em, on="ID", how="left")
# print(df7)

map_dfem = gpd.read_file('IND_adm/IND_adm1.shp')
# Rename some states
map_dfem = map_dfem.replace({'Andaman and Nicobar':'ANDAMAN & NICOBAR ISLANDS',
↪ 'Andhra Pradesh':'ANDHRA PRADESH', 'Arunachal Pradesh':'ARUNACHAL PRADESH',
↪ 'Chhattisgarh':'CHHATTISGARH', 'Dadra and Nagar Haveli':'DADRA & NAGAR',
↪ 'HAVELI', 'Daman and Diu':'DAMAN & DIU', 'Delhi':'NCT OF DELHI', 'Jammu and',
↪ 'Kashmir':'JAMMU & KASHMIR', 'Lakshadweep':'LAKSHADWEEP', 'Madhya Pradesh':
↪ 'MADHYA PRADESH', 'Maharashtra':'MAHARASHTRA', 'Bihar':'BIHAR', 'Assam':
↪ 'ASSAM', 'Chandigarh':'CHANDIGARH', 'Goa':'GOA', 'Gujarat':'GUJARAT', 'Haryana':
↪ 'HARYANA', 'Himachal Pradesh':'HIMACHAL PRADESH', 'Jharkhand':
↪ 'JHARKHAND', 'Karnataka':'KARNATAKA', 'Kerala':'KERALA', 'Mizoram':
↪ 'MIZORAM', 'Nagaland':'NAGALAND', 'Orissa':'ODISHA', 'Puducherry':
↪ 'PUDUCHERRY', 'Punjab':'PUNJAB', 'Rajasthan':'RAJASTHAN', 'Sikkim':
↪ 'SIKKIM', 'Tamil Nadu':'TAMIL NADU', 'Telangana':'TELANGANA', 'Tripura':
↪ 'TRIPURA', 'Uttar Pradesh':'UTTAR PRADESH', 'Uttarakhand':'UTTARAKHAND', 'West',
↪ 'Bengal':'WEST BENGAL', 'Uttaranchal':'UTTARAKHAND', 'Manipur':
↪ 'MANIPUR', 'Meghalaya':'MEGHALAYA'})
# print(map_df)
dfMergeem = pd.merge(map_dfem, df7em, left_on='NAME_1', right_on='State',
↪ how='left')
# print(dfMerge)
fig, ax = plt.subplots(1, figsize=(12, 12))
ax.axis('off')
ax.set_title('Cumulative number of out-migrant workers',
             fontdict={'fontsize': '15', 'fontweight' : '3'})
fig = dfMergeem.plot(column='Migrant persons', cmap='PuBu', linewidth=0.5,
↪ ax=ax, edgecolor='0.2', legend=True)

```



Dark blue states are those with the highest number of workers migrating out. Light blue states are those with the lowest number of workers migrating out.

- Clearly, once again UP stands out.
- Unfortunately, there was no Telengana in the datafile. So, we have skipped it.

0.2.1 Finding the neighbours of a given state

Here, we will use geopandas to find the neighbours of a given state. We will use this to find the states that are most likely to be migrated to from a given state.

```
[ ]: from sre_parse import State

map_dfem2 = gpd.read_file('IND_adm/IND_adm1.shp')
# Rename some states
map_dfem2 = map_dfem2.replace({'Andaman and Nicobar':'ANDAMAN & NICOBAR',
    ↳ISLANDS', 'Andhra Pradesh':'ANDHRA PRADESH', 'Arunachal Pradesh':'ARUNACHAL',
    ↳PRADESH', 'Chhattisgarh':'CHHATTISGARH', 'Dadra and Nagar Haveli':'DADRA &',
    ↳NAGAR HAVELI', 'Daman and Diu':'DAMAN & DIU', 'Delhi':'NCT OF DELHI', 'Jammu',
    ↳and Kashmir':'JAMMU & KASHMIR', 'Lakshadweep':'LAKSHADWEEP', 'Madhya',
    ↳Pradesh':'MADHYA PRADESH', 'Maharashtra':'MAHARASHTRA', 'Bihar':'BIHAR',
    ↳'Assam':'ASSAM', 'Chandigarh':'CHANDIGARH', 'Goa':'GOA', 'Gujarat':
    ↳'GUJARAT', 'Haryana':'HARYANA', 'Himachal Pradesh':'HIMACHAL',
    ↳PRADESH', 'Jharkhand':'JHARKHAND', 'Karnataka':'KARNATAKA', 'Kerala':
    ↳'KERALA', 'Mizoram':'MIZORAM', 'Nagaland':'NAGALAND', 'Orissa':
    ↳'ODISHA', 'Puducherry':'PUDUCHERRY', 'Punjab':'PUNJAB', 'Rajasthan':
    ↳'RAJASTHAN', 'Sikkim':'SIKKIM', 'Tamil Nadu':'TAMIL NADU', 'Telangana':
    ↳'TELANGANA', 'Tripura':'TRIPURA', 'Uttar Pradesh':'UTTAR',
    ↳PRADESH', 'Uttarakhand':'UTTARAKHAND', 'West Bengal':'WEST',
    ↳BENGAL', 'Uttaranchal':'UTTARAKHAND', 'Manipur':'MANIPUR', 'Meghalaya':
    ↳'MEGHALAYA'})

# map_dfem2.plot()
# add NEIGHBORS column
map_dfem2["NEIGHBORS"] = None

for index, State in map_dfem2.iterrows():

    # get 'not disjoint' countries
    neighbors = map_dfem2[~map_dfem2.geometry.disjoint(State.geometry)].NAME_1.
    ↳tolist()

    # remove own name of the country from the list
    neighbors = [ name for name in neighbors if State.NAME_1 != name ]

    # add names of neighbors as NEIGHBORS value
    map_dfem2.at[index, "NEIGHBORS"] = ", ".join(neighbors)

# print(map_dfem2)

map_dfem3StatesTemp = map_dfem2['NAME_1'].to_numpy()
map_dfem3NeighboursTemp = map_dfem2['NEIGHBORS'].to_numpy()

data = {'Home State Name': map_dfem3StatesTemp,
        'Neighbouring states': map_dfem3NeighboursTemp}

map_dfem3 = pd.DataFrame(data)
```



```
# print(map_dfem3)
```

df is our base dataframe. map_dfem3 is the dataframe that has all the states and their corresponding neighbours.

```
[ ]: print(map_dfem3.head())
```

```

      Home State Name \
0  ANDAMAN & NICOBAR ISLANDS
1          ANDHRA PRADESH
2      ARUNACHAL PRADESH
3              ASSAM
4              BIHAR

      Neighbouring states
0
1  KARNATAKA, ODISHA, PUDUCHERRY, TAMIL NADU, TEL...
2              ASSAM, NAGALAND
3  ARUNACHAL PRADESH, MANIPUR, MEGHALAYA, MIZORAM...
4      JHARKHAND, UTTAR PRADESH, WEST BENGAL

```

```
[ ]: print(df.head())
```

```

      Home State Name  Home State ID  Current State \
0      DAMAN & DIU             25  JAMMU & KASHMIR
1  DADRA & NAGAR HAVELI          26  JAMMU & KASHMIR
2      PUDUCHERRY             34  JAMMU & KASHMIR
3  ANDAMAN & NICOBAR ISLANDS        35  JAMMU & KASHMIR
4      MIZORAM                 15  JAMMU & KASHMIR

      Current State ID  Migrant Persons  Migrant Males  Migrant Females
0              1              6              3              3
1              1             17              4             13
2              1             18              6             12
3              1             51             25             26
4              1             57             22             35

```

0.2.2 The most important part of the code to calculate the number of migrations to the neighbouring states

Here, we loop over all the states to find the percentage of workers migrating to the neighbouring states.

```
[ ]: # sort based on Home State Name
df9em = df.sort_values(by=['Home State Name'])
# print(df9em)

neighborPercent = []
```

```

counter = 0
for state in map_dfem3['Home State Name']:
    # print("Doing %s" % state)
    df10em = df9em[df9em['Home State Name'] == state]
    TotalMigrants = df10em['Migrant Persons'].sum()
    # print(df10em)

    neighbors = map_dfem3['Neighbouring states'][counter]
    neighbors = neighbors.split(", ")
    # print(df10em)
    NeighborMigrant = 0
    for neighbor in neighbors:
        df11em = df10em[df10em['Current State'] == neighbor]
        NeighborMigrant += df11em['Migrant Persons'].sum()
        # df9em = df9em[df9em['Home State Name'] != neighbor]
    # % of migrants going to neighboring states
    if TotalMigrants != 0:
        NeighborMigrant = (NeighborMigrant/TotalMigrants)*100
    else:
        NeighborMigrant = 0
    print("Percentage Migrants from %s to neighbours is %d percent" % (state, NeighborMigrant))
    neighborPercent.append(NeighborMigrant)

    # if counter == 1:
    #     break
    counter = counter + 1

```

Percentage Migrants from ANDAMAN & NICOBAR ISLANDS to neighbours is 0 percent
 Percentage Migrants from ANDHRA PRADESH to neighbours is 65 percent
 Percentage Migrants from ARUNACHAL PRADESH to neighbours is 54 percent
 Percentage Migrants from ASSAM to neighbours is 63 percent
 Percentage Migrants from BIHAR to neighbours is 47 percent
 Percentage Migrants from CHANDIGARH to neighbours is 83 percent
 Percentage Migrants from CHHATTISGARH to neighbours is 81 percent
 Percentage Migrants from DADRA & NAGAR HAVELI to neighbours is 83 percent
 Percentage Migrants from DAMAN & DIU to neighbours is 76 percent
 Percentage Migrants from NCT OF DELHI to neighbours is 66 percent
 Percentage Migrants from GOA to neighbours is 85 percent
 Percentage Migrants from GUJARAT to neighbours is 84 percent
 Percentage Migrants from HARYANA to neighbours is 90 percent
 Percentage Migrants from HIMACHAL PRADESH to neighbours is 61 percent
 Percentage Migrants from JAMMU & KASHMIR to neighbours is 26 percent
 Percentage Migrants from JHARKHAND to neighbours is 75 percent
 Percentage Migrants from KARNATAKA to neighbours is 95 percent
 Percentage Migrants from KERALA to neighbours is 62 percent
 Percentage Migrants from LAKSHADWEEP to neighbours is 0 percent
 Percentage Migrants from MADHYA PRADESH to neighbours is 87 percent

Percentage Migrants from MAHARASHTRA to neighbours is 73 percent
 Percentage Migrants from MANIPUR to neighbours is 50 percent
 Percentage Migrants from MEGHALAYA to neighbours is 57 percent
 Percentage Migrants from MIZORAM to neighbours is 74 percent
 Percentage Migrants from NAGALAND to neighbours is 57 percent
 Percentage Migrants from ODISHA to neighbours is 54 percent
 Percentage Migrants from PUDUCHERRY to neighbours is 97 percent
 Percentage Migrants from PUNJAB to neighbours is 61 percent
 Percentage Migrants from RAJASTHAN to neighbours is 62 percent
 Percentage Migrants from SIKKIM to neighbours is 53 percent
 Percentage Migrants from TAMIL NADU to neighbours is 79 percent
 Percentage Migrants from TELANGANA to neighbours is 0 percent
 Percentage Migrants from TRIPURA to neighbours is 53 percent
 Percentage Migrants from UTTAR PRADESH to neighbours is 58 percent
 Percentage Migrants from UTTARAKHAND to neighbours is 41 percent
 Percentage Migrants from WEST BENGAL to neighbours is 42 percent

neighborPercent variable stores the percentage of workers migrating to the neighbouring states.

```
[ ]: print(neighborPercent)
```

```
[0.0, 65.61292490064059, 54.5546992078784, 63.66300133092009, 47.12418613692903,
83.37480472058574, 81.38551854585717, 83.22813345356177, 76.70051835396171,
66.4719322910375, 85.26498173189198, 84.38215314067011, 90.91028815824414,
61.32043603951305, 26.104907287204448, 75.20950806152179, 95.22908113446661,
62.201769500319436, 0.0, 87.2134914273977, 73.93318821170897,
50.873255798603324, 57.999373825923605, 74.4574345463527, 57.974373551405954,
54.81578858346295, 97.28286836037309, 61.745085953803745, 62.42973916580332,
53.76298988769281, 79.88224608935212, 0, 53.005986350189836, 58.42205650966363,
41.89740028382499, 42.19454239038346]
```

Let us make neighborPercent into a panda dataframe

```
[ ]: data = {'Home State Name': map_dfem3['Home State Name'],
            'Neighbor Percent migration': neighborPercent}

df12em = pd.DataFrame(data)
print(df12em.head())
```

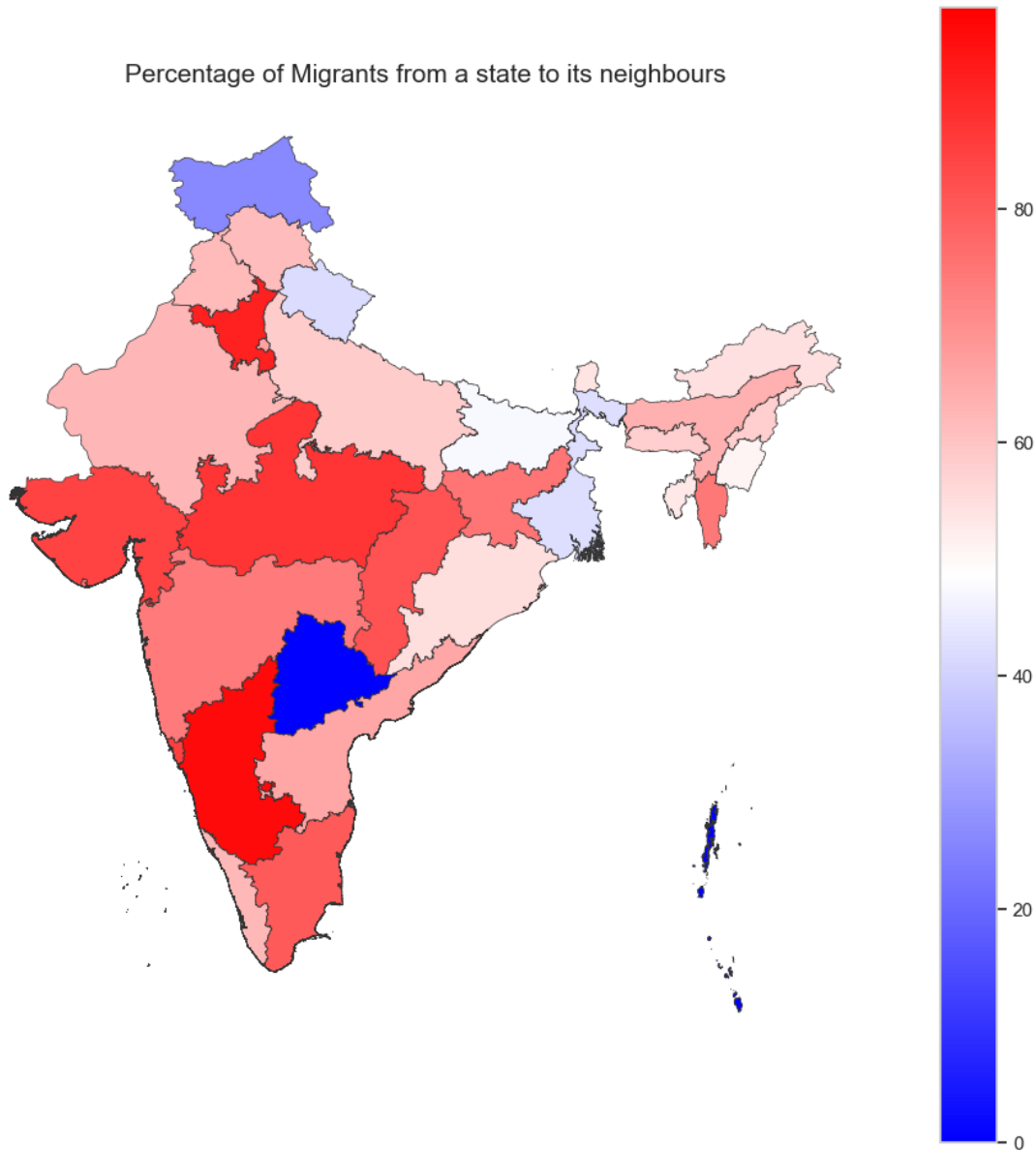
| | Home State Name | Neighbor Percent migration |
|---|---------------------------|----------------------------|
| 0 | ANDAMAN & NICOBAR ISLANDS | 0.000000 |
| 1 | ANDHRA PRADESH | 65.612925 |
| 2 | ARUNACHAL PRADESH | 54.554699 |
| 3 | ASSAM | 63.663001 |
| 4 | BIHAR | 47.124186 |

```
[ ]: map_dfNeighbour = gpd.read_file('IND_adm/IND_adm1.shp')
# Rename some states
```

```

map_dfNeighbour = map_dfNeighbour.replace({'Andaman and Nicobar':'ANDAMAN &
↳NICOBAR ISLANDS', 'Andhra Pradesh':'ANDHRA PRADESH', 'Arunachal Pradesh':
↳'ARUNACHAL PRADESH', 'Chhattisgarh':'CHHATTISGARH', 'Dadra and Nagar Haveli':
↳'DADRA & NAGAR HAVELI', 'Daman and Diu':'DAMAN & DIU', 'Delhi':'NCT OF
↳DELHI', 'Jammu and Kashmir':'JAMMU & KASHMIR', 'Lakshadweep':'LAKSHADWEEP',
↳'Madhya Pradesh':'MADHYA PRADESH', 'Maharashtra':'MAHARASHTRA', 'Bihar':
↳'BIHAR', 'Assam':'ASSAM', 'Chandigarh':'CHANDIGARH', 'Goa':'GOA', 'Gujarat':
↳'GUJARAT', 'Haryana':'HARYANA', 'Himachal Pradesh':'HIMACHAL
↳PRADESH', 'Jharkhand':'JHARKHAND', 'Karnataka':'KARNATAKA', 'Kerala':
↳'KERALA', 'Mizoram':'MIZORAM', 'Nagaland':'NAGALAND', 'Orissa':
↳'ODISHA', 'Puducherry':'PUDUCHERRY', 'Punjab':'PUNJAB', 'Rajasthan':
↳'RAJASTHAN', 'Sikkim':'SIKKIM', 'Tamil Nadu':'TAMIL NADU', 'Telangana':
↳'TELANGANA', 'Tripura':'TRIPURA', 'Uttar Pradesh':'UTTAR
↳PRADESH', 'Uttarakhand':'UTTARAKHAND', 'West Bengal':'WEST
↳BENGAL', 'Uttaranchal':'UTTARAKHAND', 'Manipur':'MANIPUR', 'Meghalaya':
↳'MEGHALAYA'})
# print(map_df)
dfMergeem = pd.merge(map_dfNeighbour, df12em, left_on='NAME_1', right_on='Home
↳State Name', how='left')
# print(dfMerge)
fig, ax = plt.subplots(1, figsize=(12, 12))
ax.axis('off')
ax.set_title('Percentage of Migrants from a state to its neighbours',
fontdict={'fontsize': '15', 'fontweight' : '3'})
fig = dfMergeem.plot(column='Neighbor Percent migration', cmap='bwr',
↳linewidth=0.5, ax=ax, edgecolor='0.2', legend=True)

```



Red denotes the states with highest number of workers migrating to the neighbouring states. Blue denotes the states with the lowest number of workers migrating to the neighbouring states.

In the above map of India, we have plotted the percentage of workers migrating to the neighbouring states. Clearly, the southern states have the highest percentage of workers migrating to the neighbouring states.

We also see that percentage of migrant workers from North-East to their neighbours is very high.

Not surprising, % of workers migrating from UP, Bihar, and other northern to their neighbours is very low. This is due to the fact that people from UP and Bihar go to Delhi, Maharashtra, and other southern states for jobs.

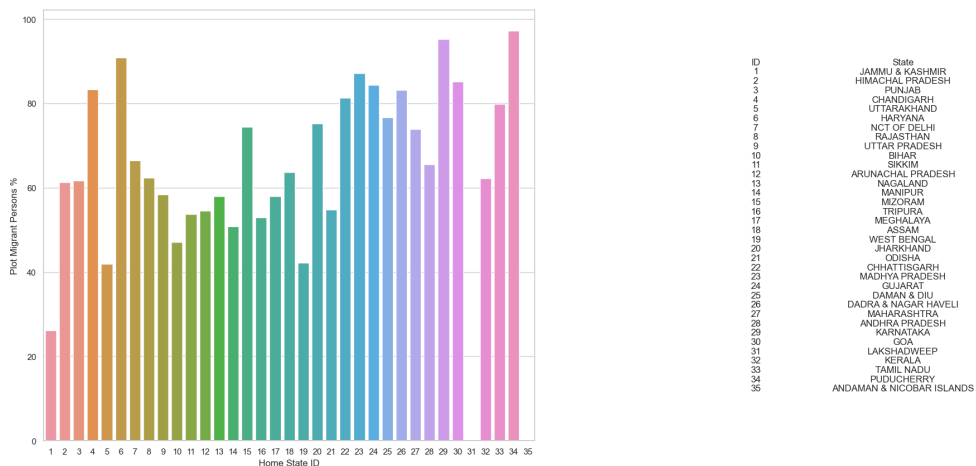
To further understand the data, we can make bar plots of the states with the highest number of workers migrating to the neighbouring states.

```
[ ]: sns.set_theme(style="whitegrid")

df13em = pd.merge(IDs, df12em, left_on='State', right_on='Home State Name',
                  how='left')
# print(df13em)
# Plot Migrant Persons % to neighbour against Home State ID
fig, (ax10, ax12) = plt.subplots(1,2)
fig.set_size_inches(25, 10)
sns.barplot(x="ID", y="Neighbor Percent migration", data=df13em,ax=ax10)
ax10.xaxis.grid(True, "minor", linewidth=.25)
ax10.yaxis.grid(True, "minor", linewidth=.25)
ax10.set_xlabel('Home State ID')
ax10.set_ylabel('Plot Migrant Persons %')

ax12.axis('off')
ax12.table(cellText=IDs.values, colLabels=IDs.columns, loc='center',
          colWidths=[0.1, 0.5],cellLoc='center',edges='open')
```

```
[ ]: <matplotlib.table.Table at 0x18b921930>
```



0.3 End of Question 1

0.4 Question (2)

In this question, we need to find the top 10 employing states.

Idea is to first group and sum the data based on this group. Then plot it using a line plot to visualize the data.

After that, we can sort this group based on highest number of migrant persons.

```
[ ]: # Now, we get a sum of migrants to a particular Current State

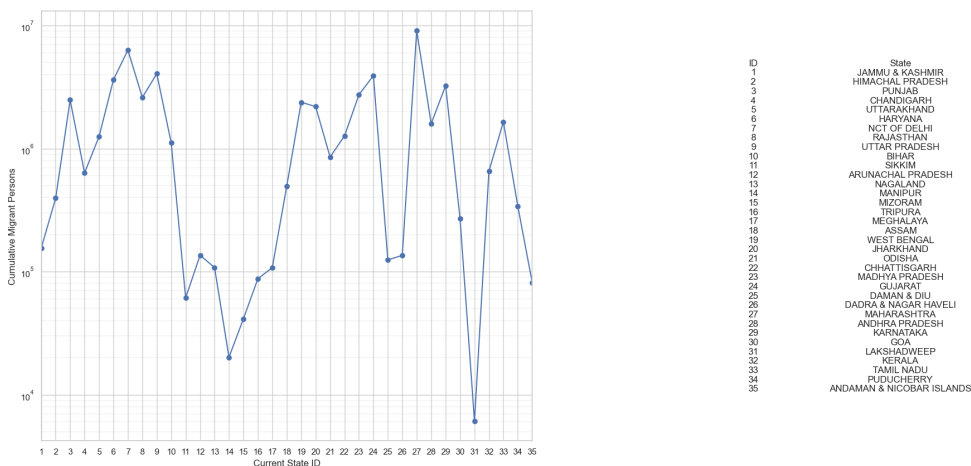
df2 = df.groupby(['Current State ID']).sum()

# Plot Migrant Persons against Current State ID
# print(df2.index)
fig, (ax, ax0) = plt.subplots(1,2)
fig.set_size_inches(25, 10)
ax.plot(df2.index, df2['Migrant Persons'], 'o-')
ax.xaxis.grid(True, "minor", linewidth=.25)
ax.yaxis.grid(True, "minor", linewidth=.25)
ax.set_xlim(1, 35)
start, end = ax1.get_xlim()
ax.xaxis.set_ticks(np.arange(1, 36, 1))
ax.xaxis.set_major_formatter(ticker.FormatStrFormatter('%d'))
ax.set_xlabel('Current State ID')
ax.set_ylabel('Cumulative Migrant Persons')

# set yaxis to log scale
ax.set_yscale('log')

ax0.axis('off')
ax0.table(cellText=IDs.values, colLabels=IDs.columns, loc='center',
         colWidths=[0.1, 0.5], cellLoc='center', edges='open')
```

```
[ ]: <matplotlib.table.Table at 0x18bd58310>
```



As we saw in the previous plots, Maharashtra and Delhi are the top 2 states with the highest number of migrant workers.

Surprisingly, UP is third in this list. This implies that UP does not only have the highest emigration but also has one of the highest immigration.

To further enhance the interpretation of this data, let us plot on the map of India

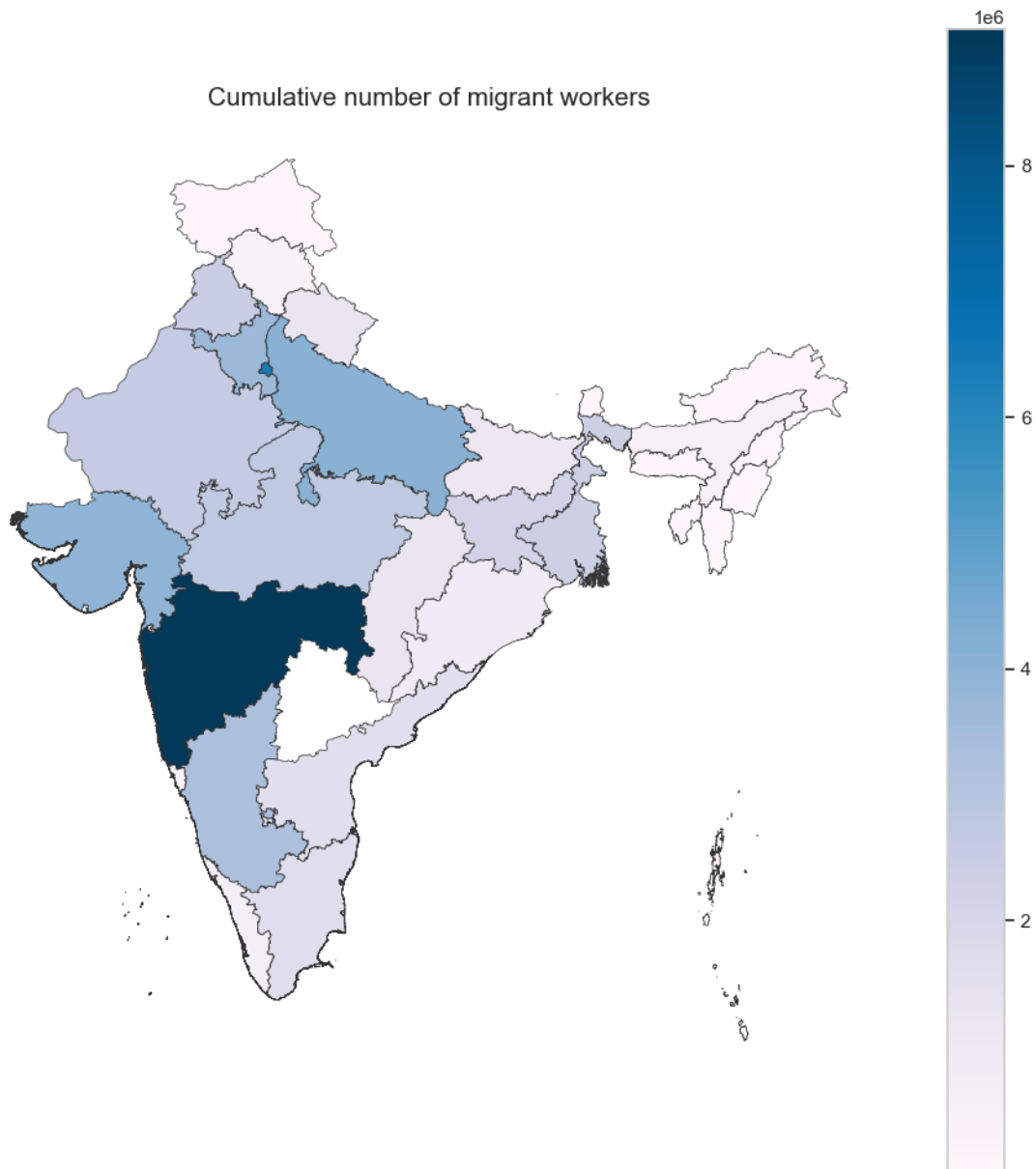
```
[ ]: df6id = df2.index.to_numpy()
df6value = df2['Migrant Persons'].to_numpy()

data = {'ID': df6id,
        'Migrant persons': df6value}

df6 = pd.DataFrame(data)

df7 = pd.merge(IDs, df6, on="ID", how="left")
# print(df7)

import geopandas as gpd
# Read shapefile using Geopandas
map_df = gpd.read_file('IND_adm/IND_adm1.shp')
# Rename some states
map_df = map_df.replace({'Andaman and Nicobar': 'ANDAMAN & NICOBAR ISLANDS',
    ↪ 'Andhra Pradesh': 'ANDHRA PRADESH', 'Arunachal Pradesh': 'ARUNACHAL PRADESH',
    ↪ 'Chhattisgarh': 'CHHATTISGARH', 'Dadra and Nagar Haveli': 'DADRA & NAGAR
    ↪ HAVELI', 'Daman and Diu': 'DAMAN & DIU', 'Delhi': 'NCT OF DELHI', 'Jammu and
    ↪ Kashmir': 'JAMMU & KASHMIR', 'Lakshadweep': 'LAKSHADWEEP', 'Madhya Pradesh':
    ↪ 'MADHYA PRADESH', 'Maharashtra': 'MAHARASHTRA', 'Bihar': 'BIHAR', 'Assam':
    ↪ 'ASSAM', 'Chandigarh': 'CHANDIGARH', 'Goa': 'GOA', 'Gujarat': 'GUJARAT', 'Haryana':
    ↪ 'HARYANA', 'Himachal Pradesh': 'HIMACHAL PRADESH', 'Jharkhand':
    ↪ 'JHARKHAND', 'Karnataka': 'KARNATAKA', 'Kerala': 'KERALA', 'Mizoram':
    ↪ 'MIZORAM', 'Nagaland': 'NAGALAND', 'Orissa': 'ODISHA', 'Puducherry':
    ↪ 'PUDUCHERRY', 'Punjab': 'PUNJAB', 'Rajasthan': 'RAJASTHAN', 'Sikkim':
    ↪ 'SIKKIM', 'Tamil Nadu': 'TAMIL NADU', 'Telangana': 'TELANGANA', 'Tripura':
    ↪ 'TRIPURA', 'Uttar Pradesh': 'UTTAR PRADESH', 'Uttarakhand': 'UTTARAKHAND', 'West
    ↪ Bengal': 'WEST BENGAL', 'Uttaranchal': 'UTTARAKHAND', 'Manipur':
    ↪ 'MANIPUR', 'Meghalaya': 'MEGHALAYA'})
# print(map_df)
dfMerge = pd.merge(map_df, df7, left_on='NAME_1', right_on='State', how='left')
# print(dfMerge)
fig, ax = plt.subplots(1, figsize=(12, 12))
ax.axis('off')
ax.set_title('Cumulative number of migrant workers',
             fontdict={'fontsize': '15', 'fontweight' : '3'})
fig = dfMerge.plot(column='Migrant persons', cmap='PuBu', linewidth=0.5, ax=ax,
    ↪ edgecolor='0.2', legend=True)
```

There is no Telengana in the data. So, I have removed it from the plot.

```
[ ]: # sort the data frame by Migrant Persons

df3 = df2.sort_values(by=['Migrant Persons'], ascending=False)
# top 10 states with highest number of migrants
df4 = df3.head(10)
# print(df4)

df5id = df4.index.to_numpy()
```

```

df5value = df4['Migrant Persons'].to_numpy()

data = {'ID': df5id,
        'Migrant persons': df5value}

df5 = pd.DataFrame(data)

print(df5)

```

| | ID | Migrant persons |
|---|----|-----------------|
| 0 | 27 | 9087380 |
| 1 | 7 | 6330065 |
| 2 | 9 | 4061933 |
| 3 | 24 | 3916075 |
| 4 | 6 | 3626318 |
| 5 | 29 | 3247660 |
| 6 | 23 | 2744332 |
| 7 | 8 | 2604298 |
| 8 | 3 | 2488299 |
| 9 | 19 | 2381045 |

So, we have the id of the top ten employing states. Now, we need to find their names. For this, we will use the IDs dataframe that we created earlier.

```

[ ]: # find index of top 10 states from IDs
IDs3 = IDs[IDs['ID'].isin(df4.index)]
# print(IDs3)

Tempid = IDs3['ID'].to_numpy()
Tempvalue = IDs3['State'].to_numpy()
Tempdata = {'ID': Tempid,
            'State': Tempvalue}

IDs4 = pd.DataFrame(Tempdata)
print(IDs4)

```

| | ID | State |
|---|----|----------------|
| 0 | 3 | PUNJAB |
| 1 | 6 | HARYANA |
| 2 | 7 | NCT OF DELHI |
| 3 | 8 | RAJASTHAN |
| 4 | 9 | UTTAR PRADESH |
| 5 | 19 | WEST BENGAL |
| 6 | 23 | MADHYA PRADESH |
| 7 | 24 | GUJARAT |
| 8 | 27 | MAHARASHTRA |
| 9 | 29 | KARNATAKA |

Of course, we need to arrange these states based on the migrant population as available in df5

```
[ ]: IDs4 = IDs4.set_index('ID')
      IDs4 = IDs4.reindex(index=df5['ID'])
      IDs4 = IDs4.reset_index()
      print(IDs4)
```

| | ID | State |
|---|----|----------------|
| 0 | 27 | MAHARASHTRA |
| 1 | 7 | NCT OF DELHI |
| 2 | 9 | UTTAR PRADESH |
| 3 | 24 | GUJARAT |
| 4 | 6 | HARYANA |
| 5 | 29 | KARNATAKA |
| 6 | 23 | MADHYA PRADESH |
| 7 | 8 | RAJASTHAN |
| 8 | 3 | PUNJAB |
| 9 | 19 | WEST BENGAL |

Finally we can merge the two datasets to get the top 10 employing states

```
[ ]: pd.merge(df5, IDs4, on="ID", how="left")
```

```
[ ]:   ID  Migrant persons      State
0  27          9087380  MAHARASHTRA
1   7          6330065  NCT OF DELHI
2   9          4061933  UTTAR PRADESH
3  24          3916075    GUJARAT
4   6          3626318    HARYANA
5  29          3247660    KARNATAKA
6  23          2744332  MADHYA PRADESH
7   8          2604298    RAJASTHAN
8   3          2488299    PUNJAB
9  19          2381045  WEST BENGAL
```

Above, we have plotted the top ten states with the highest number of migrant workers.

0.4.1 End of Question (2)

0.5 Question 3

Here, we want to look at the gender ratio of workers migrating to Maharashtra from all states over India.

To do so, we will categorize data into dfMaharashtra

```
[ ]: dfMaharashtra = df[df['Current State ID'] == 27]
      dfMaharashtra = dfMaharashtra.rename({'Migrant Males' : 'Migrant Males %'},
      ↪axis=1)
```

```

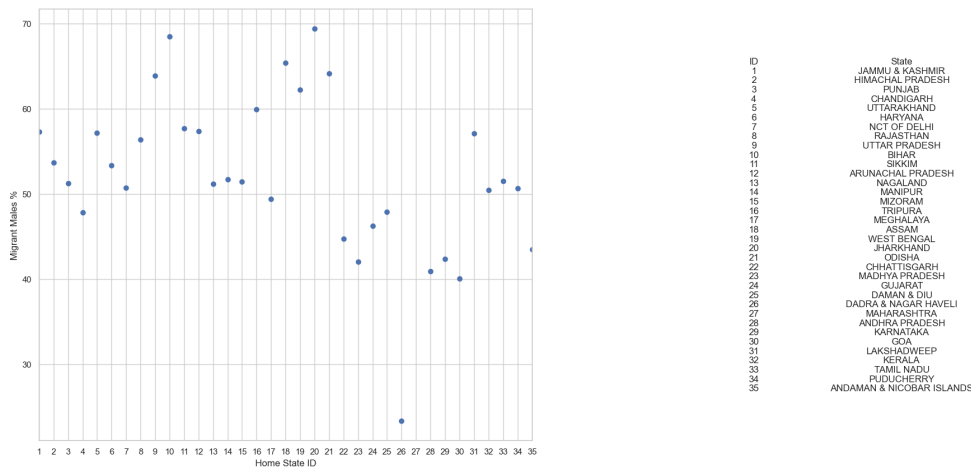
dfMaharashtra['Migrant Males %'] = dfMaharashtra['Migrant Males %']/
↳dfMaharashtra['Migrant Persons']*100
# print(dfMaharashtra)

# plot Migrant Males % against Home State ID
fig, (ax3, ax4) = plt.subplots(1,2)
fig.set_size_inches(25, 10)
ax3.plot(dfMaharashtra['Home State ID'], dfMaharashtra['Migrant Males %'], 'o')
ax3.xaxis.grid(True, "minor", linewidth=.25)
ax3.yaxis.grid(True, "minor", linewidth=.25)
ax3.set_xlim(1, 35)
start, end = ax1.get_xlim()
ax3.xaxis.set_ticks(np.arange(1, 36, 1))
ax3.xaxis.set_major_formatter(ticker.FormatStrFormatter('%d'))
ax3.set_xlabel('Home State ID')
ax3.set_ylabel('Migrant Males %')

ax4.axis('off')
ax4.table(cellText=IDs.values, colLabels=IDs.columns, loc='center',
↳colWidths=[0.1, 0.5], cellLoc='center', edges='open')

```

```
[ ]: <matplotlib.table.Table at 0x18ba83250>
```



To further understand this data about percentage of males migrating to Maharashtra, we once again plot it on the map of india

```

[ ]: map_dfNeighbour = gpd.read_file('IND_adm/IND_adm1.shp')
# Rename some states

```

```

map_dfNeighbour = map_dfNeighbour.replace({'Andaman and Nicobar':'ANDAMAN &
↳NICOBAR ISLANDS', 'Andhra Pradesh':'ANDHRA PRADESH', 'Arunachal Pradesh':
↳'ARUNACHAL PRADESH', 'Chhattisgarh':'CHHATTISGARH', 'Dadra and Nagar Haveli':
↳'DADRA & NAGAR HAVELI', 'Daman and Diu':'DAMAN & DIU', 'Delhi':'NCT OF
↳DELHI', 'Jammu and Kashmir':'JAMMU & KASHMIR', 'Lakshadweep':'LAKSHADWEEP',
↳'Madhya Pradesh':'MADHYA PRADESH', 'Maharashtra':'MAHARASHTRA', 'Bihar':
↳'BIHAR', 'Assam':'ASSAM', 'Chandigarh':'CHANDIGARH', 'Goa':'GOA', 'Gujarat':
↳'GUJARAT', 'Haryana':'HARYANA', 'Himachal Pradesh':'HIMACHAL
↳PRADESH', 'Jharkhand':'JHARKHAND', 'Karnataka':'KARNATAKA', 'Kerala':
↳'KERALA', 'Mizoram':'MIZORAM', 'Nagaland':'NAGALAND', 'Orissa':
↳'ODISHA', 'Puducherry':'PUDUCHERRY', 'Punjab':'PUNJAB', 'Rajasthan':
↳'RAJASTHAN', 'Sikkim':'SIKKIM', 'Tamil Nadu':'TAMIL NADU', 'Telangana':
↳'TELANGANA', 'Tripura':'TRIPURA', 'Uttar Pradesh':'UTTAR
↳PRADESH', 'Uttarakhand':'UTTARAKHAND', 'West Bengal':'WEST
↳BENGAL', 'Uttaranchal':'UTTARAKHAND', 'Manipur':'MANIPUR', 'Meghalaya':
↳'MEGHALAYA'})
# print(map_df)
dfMergeem = pd.merge(map_dfNeighbour, dfMaharashtra, left_on='NAME_1',
↳right_on='Home State Name', how='left')
print(dfMergeem)
fig, ax = plt.subplots(1, figsize=(12, 12))
ax.axis('off')
ax.set_title('Percentage of Males Migrants from other states to Maharashtra',
fontdict={'fontsize': '15', 'fontweight' : '3'})
fig = dfMergeem.plot(column='Migrant Males %', cmap='hot_r', linewidth=0.5,
↳ax=ax, edgecolor='0.2', legend=True)

```

| | ID_0 | ISO | NAME_0 | ID_1 | NAME_1 | TYPE_1 | \ |
|----|------|-----|--------|------|---------------------------|----------------|---|
| 0 | 105 | IND | India | 1 | ANDAMAN & NICOBAR ISLANDS | Union Territor | |
| 1 | 105 | IND | India | 2 | ANDHRA PRADESH | State | |
| 2 | 105 | IND | India | 3 | ARUNACHAL PRADESH | State | |
| 3 | 105 | IND | India | 4 | ASSAM | State | |
| 4 | 105 | IND | India | 5 | BIHAR | State | |
| 5 | 105 | IND | India | 6 | CHANDIGARH | Union Territor | |
| 6 | 105 | IND | India | 7 | CHHATTISGARH | State | |
| 7 | 105 | IND | India | 8 | DADRA & NAGAR HAVELI | Union Territor | |
| 8 | 105 | IND | India | 9 | DAMAN & DIU | Union Territor | |
| 9 | 105 | IND | India | 10 | NCT OF DELHI | Union Territor | |
| 10 | 105 | IND | India | 11 | GOA | State | |
| 11 | 105 | IND | India | 12 | GUJARAT | State | |
| 12 | 105 | IND | India | 13 | HARYANA | State | |
| 13 | 105 | IND | India | 14 | HIMACHAL PRADESH | Union Territor | |
| 14 | 105 | IND | India | 15 | JAMMU & KASHMIR | State | |
| 15 | 105 | IND | India | 16 | JHARKHAND | State | |
| 16 | 105 | IND | India | 17 | KARNATAKA | State | |
| 17 | 105 | IND | India | 18 | KERALA | State | |
| 18 | 105 | IND | India | 19 | LAKSHADWEEP | Union Territor | |

| | | | | | | |
|----|-----|-----|-------|----|----------------|----------------|
| 19 | 105 | IND | India | 20 | MADHYA PRADESH | State |
| 20 | 105 | IND | India | 21 | MAHARASHTRA | State |
| 21 | 105 | IND | India | 22 | MANIPUR | State |
| 22 | 105 | IND | India | 23 | MEGHALAYA | State |
| 23 | 105 | IND | India | 24 | MIZORAM | State |
| 24 | 105 | IND | India | 25 | NAGALAND | State |
| 25 | 105 | IND | India | 26 | ODISHA | State |
| 26 | 105 | IND | India | 27 | PUDUCHERRY | Union Territor |
| 27 | 105 | IND | India | 28 | PUNJAB | State |
| 28 | 105 | IND | India | 29 | RAJASTHAN | State |
| 29 | 105 | IND | India | 30 | SIKKIM | State |
| 30 | 105 | IND | India | 31 | TAMIL NADU | State |
| 31 | 105 | IND | India | 32 | TELANGANA | State |
| 32 | 105 | IND | India | 33 | TRIPURA | State |
| 33 | 105 | IND | India | 34 | UTTAR PRADESH | State |
| 34 | 105 | IND | India | 35 | UTTARAKHAND | State |
| 35 | 105 | IND | India | 36 | WEST BENGAL | State |

| | ENGTYPE_1 | NL_NAME_1 | \ |
|----|-----------------|-----------|---|
| 0 | Union Territory | None | |
| 1 | State | None | |
| 2 | State | None | |
| 3 | State | None | |
| 4 | State | None | |
| 5 | Union Territory | None | |
| 6 | State | None | |
| 7 | Union Territory | None | |
| 8 | Union Territory | None | |
| 9 | Union Territory | None | |
| 10 | State | None | |
| 11 | State | None | |
| 12 | State | None | |
| 13 | Union Territory | None | |
| 14 | State | None | |
| 15 | State | None | |
| 16 | State | None | |
| 17 | State | None | |
| 18 | Union Territory | None | |
| 19 | State | None | |
| 20 | State | None | |
| 21 | State | None | |
| 22 | State | None | |
| 23 | State | None | |
| 24 | State | None | |
| 25 | State | None | |
| 26 | Union Territory | None | |
| 27 | State | None | |
| 28 | State | None | |

| | | |
|----|-------|------|
| 29 | State | None |
| 30 | State | None |
| 31 | State | None |
| 32 | State | None |
| 33 | State | None |
| 34 | State | None |
| 35 | State | None |

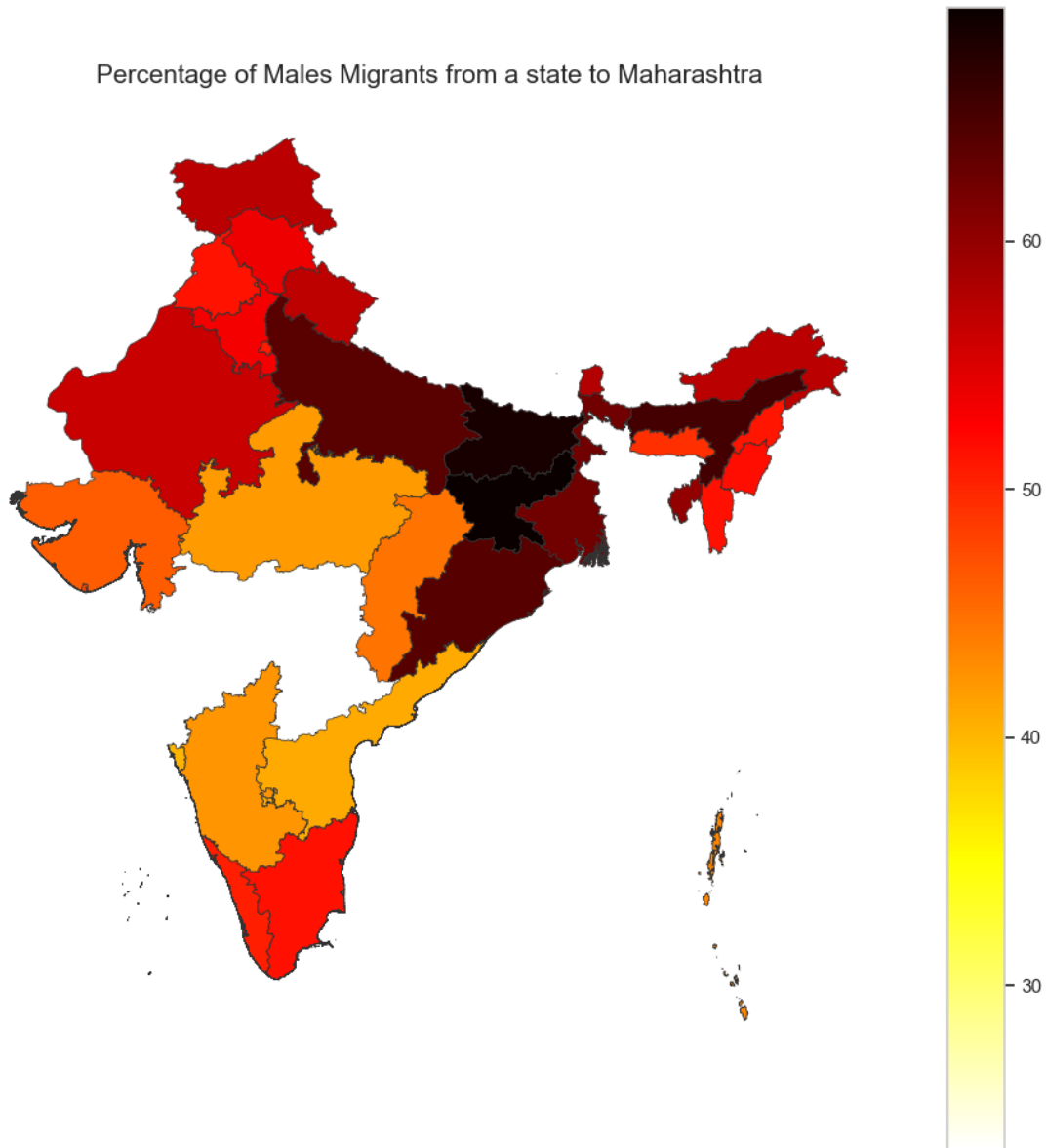
| | | |
|----|---|------|
| | VARNAME_1 | \ |
| 0 | Andaman & Nicobar Islands Andaman et Nicobar I... | |
| 1 | | None |
| 2 | Agence de la Frontière du Nord-Est(French-obso... | |
| 3 | | None |
| 4 | | None |
| 5 | | None |
| 6 | | None |
| 7 | Dadra et Nagar Haveli Dadra e Nagar Haveli | |
| 8 | | None |
| 9 | | None |
| 10 | | Gôa |
| 11 | Goudjerate Gujerat Gujerate | |
| 12 | | None |
| 13 | | None |
| 14 | | None |
| 15 | Vananchal | |
| 16 | Maisur Mysore | |
| 17 | | None |
| 18 | Îles Laquedives Laccadive Minicoy and Amindivi... | |
| 19 | | None |
| 20 | | None |
| 21 | | None |
| 22 | | None |
| 23 | | None |
| 24 | | None |
| 25 | | None |
| 26 | Pondicherry Puduchcheri Pondichéry | |
| 27 | | None |
| 28 | Greater Rajasthan Rajputana | |
| 29 | | None |
| 30 | Madras Tamilnad | |
| 31 | | None |
| 32 | | None |
| 33 | United Provinces | |
| 34 | UTTARAKHAND | |
| 35 | Bangla Bengala Occidentale Bengala Ocidental B... | |

| | | | |
|---|--|----|---|
| | geometry | ID | \ |
| 0 | MULTIPOLYGON (((93.78773 6.85264, 93.78849 6.8... 35.0 | | |

| | | |
|----|---|------|
| 1 | MULTIPOLYGON (((80.27458 13.45958, 80.27458 13... | 28.0 |
| 2 | POLYGON ((96.15778 29.38310, 96.16380 29.37668... | 12.0 |
| 3 | MULTIPOLYGON (((89.87145 25.53730, 89.87118 25... | 18.0 |
| 4 | MULTIPOLYGON (((88.10548 26.53904, 88.10505 26... | 10.0 |
| 5 | POLYGON ((76.80293 30.67548, 76.79437 30.66932... | 4.0 |
| 6 | POLYGON ((83.32760 24.09965, 83.34575 24.09707... | 22.0 |
| 7 | POLYGON ((73.02468 20.09630, 73.01955 20.10502... | 26.0 |
| 8 | MULTIPOLYGON (((72.86014 20.47096, 72.86340 20... | 25.0 |
| 9 | POLYGON ((77.32713 28.68516, 77.32539 28.68250... | 7.0 |
| 10 | MULTIPOLYGON (((73.78181 15.35569, 73.78181 15... | 30.0 |
| 11 | MULTIPOLYGON (((70.86097 20.75292, 70.86097 20... | 24.0 |
| 12 | POLYGON ((76.83715 30.87887, 76.85243 30.87069... | 6.0 |
| 13 | POLYGON ((76.80276 33.23666, 76.80630 33.23623... | 2.0 |
| 14 | POLYGON ((77.89957 35.42789, 77.90297 35.42759... | 1.0 |
| 15 | POLYGON ((87.59989 25.31466, 87.60688 25.31138... | 20.0 |
| 16 | MULTIPOLYGON (((74.67097 13.19986, 74.67097 13... | 29.0 |
| 17 | MULTIPOLYGON (((76.46736 9.54097, 76.46736 9.5... | 32.0 |
| 18 | MULTIPOLYGON (((73.01014 8.28042, 73.01014 8.2... | 31.0 |
| 19 | POLYGON ((78.36465 26.86884, 78.36688 26.86259... | 23.0 |
| 20 | MULTIPOLYGON (((73.45597 15.88986, 73.45597 15... | 27.0 |
| 21 | POLYGON ((94.57723 25.64833, 94.57609 25.64470... | 14.0 |
| 22 | POLYGON ((91.85384 26.10479, 91.86470 26.10035... | 17.0 |
| 23 | POLYGON ((92.80080 24.41905, 92.80370 24.41879... | 15.0 |
| 24 | POLYGON ((95.21445 26.93695, 95.21706 26.93420... | 13.0 |
| 25 | MULTIPOLYGON (((84.76986 19.10597, 84.76986 19... | 21.0 |
| 26 | MULTIPOLYGON (((79.84486 10.82653, 79.84486 10... | 34.0 |
| 27 | POLYGON ((75.86877 32.48868, 75.88712 32.47203... | 3.0 |
| 28 | POLYGON ((73.88944 29.97761, 73.89118 29.97007... | 8.0 |
| 29 | POLYGON ((88.64526 28.09912, 88.65411 28.08984... | 11.0 |
| 30 | MULTIPOLYGON (((77.55596 8.07903, 77.55596 8.0... | 33.0 |
| 31 | POLYGON ((78.33625 19.88319, 78.34669 19.88140... | NaN |
| 32 | POLYGON ((92.18520 24.52287, 92.18896 24.52019... | 16.0 |
| 33 | POLYGON ((77.58468 30.40878, 77.58639 30.40801... | 9.0 |
| 34 | POLYGON ((79.19478 31.35362, 79.19817 31.35196... | 5.0 |
| 35 | MULTIPOLYGON (((88.01861 21.57278, 88.01889 21... | 19.0 |

| | State | Migrant persons |
|----|---------------------------|-----------------|
| 0 | ANDAMAN & NICOBAR ISLANDS | 81267.0 |
| 1 | ANDHRA PRADESH | 1591890.0 |
| 2 | ARUNACHAL PRADESH | 136010.0 |
| 3 | ASSAM | 495699.0 |
| 4 | BIHAR | 1111954.0 |
| 5 | CHANDIGARH | 633966.0 |
| 6 | CHHATTISGARH | 1267668.0 |
| 7 | DADRA & NAGAR HAVELI | 135602.0 |
| 8 | DAMAN & DIU | 124522.0 |
| 9 | NCT OF DELHI | 6330065.0 |
| 10 | GOA | 269689.0 |

| | | |
|----|------------------|-----------|
| 11 | GUJARAT | 3916075.0 |
| 12 | HARYANA | 3626318.0 |
| 13 | HIMACHAL PRADESH | 395504.0 |
| 14 | JAMMU & KASHMIR | 155187.0 |
| 15 | JHARKHAND | 2195521.0 |
| 16 | KARNATAKA | 3247660.0 |
| 17 | KERALA | 654423.0 |
| 18 | LAKSHADWEEP | 6077.0 |
| 19 | MADHYA PRADESH | 2744332.0 |
| 20 | MAHARASHTRA | 9087380.0 |
| 21 | MANIPUR | 20100.0 |
| 22 | MEGHALAYA | 107915.0 |
| 23 | MIZORAM | 41380.0 |
| 24 | NAGALAND | 108020.0 |
| 25 | ODISHA | 855096.0 |
| 26 | PUDUCHERRY | 339967.0 |
| 27 | PUNJAB | 2488299.0 |
| 28 | RAJASTHAN | 2604298.0 |
| 29 | SIKKIM | 61163.0 |
| 30 | TAMIL NADU | 1650771.0 |
| 31 | NaN | NaN |
| 32 | TRIPURA | 87378.0 |
| 33 | UTTAR PRADESH | 4061933.0 |
| 34 | UTTARAKHAND | 1250575.0 |
| 35 | WEST BENGAL | 2381045.0 |



We notice that Bihar and Jharkhand present one of the highest % of migrant males to Maharashtra. Clearly, the trend implies a mass gendered biased migration of people from these states to Maharashtra.

0.6 End of Question 3
