

PHASE – 5

PROJECT DOCUMENTATION AND SUBMISSION

Date	27-10-2023
Team ID	4570
Project Name	Assessment of Marginal Workers in Tamil Nadu

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1. INTRODUCTION

This project aims to conduct a comprehensive socioeconomic analysis focusing on marginal workers. The study will delve into the demographic attributes of this vital workforce, including age, industrial categorization, and gender distribution. To facilitate a clear understanding, visualizations in the form of bar charts, pie charts, and heat maps will be employed to depict the distribution across these distinct categories.

In this project, we will outline the problem statement, the steps involved in solving it, and the design thinking approach that will guide our project.

2. PROBLEM STATEMENT

Analyze the demographic characteristics of marginal workers based on age, industrial category, and sex. Create visualizations such as bar charts, pie charts, or heat maps to represent the distribution across different categories.

3. OBJECTIVE

The objective of this project is to conduct a comprehensive analysis of marginal workers based on their demographic characteristics, specifically focusing on age, industrial category, and sex. Marginal workers play a crucial role in the labor force, often facing unique challenges and opportunities. Understanding their demographic profile is essential for policy-making and resource allocation.

4. ANALYSIS APPROACH

4.1 Data Collection:

- Identify authoritative sources for demographic data on marginal workers. These may include government reports, labor market surveys, or specialized research studies.
- Obtain data on age, industrial category, and sex of marginal workers. Ensure the data is comprehensive and representative.

4.2 Data Cleaning and Preparation:

- Perform data cleaning to address any missing or erroneous values. This may involve imputation, removal, or correction of outliers.
- Standardize data formats and units to ensure consistency across variables.

4.3 Demographic Segmentation:

- Categorize marginal workers based on age groups (e.g., 18-25, 26-35, 36-45, 46-60, 60+), industrial categories (e.g., agriculture, manufacturing, construction, services), and sex (male/female).

4.4 Descriptive Statistics:

- Calculate summary statistics for each demographic category, including measures like mean, median, mode, and standard deviation. This provides an initial overview of the distribution.

4.5 Visualization:

Create visual representations using appropriate charts:

4.5.1 Age Distribution:

- Bar chart or histogram to display the distribution of marginal workers across different age groups.

4.5.2 Industrial Category Distribution:

- Pie chart or stacked bar chart to show the proportion of workers in each industrial category.

4.5.3 Sex Distribution:

- Pie chart or bar chart to represent the distribution of male, female, and other genders among marginal workers.

4.5.4 Cross-Category Analysis:

- Heat map to visualize the intersection of age groups, industrial categories, and sex.

This analysis approach provides a structured framework to systematically examine the demographic characteristics of marginal workers and generate meaningful insights.

5. LOADING AND PRE – PROCESSING

```
6. from google.colab import files
   uploaded = files.upload()
```

```
7. <IPython.core.display.HTML object>
```

```
8. Saving Dataset.csv to Dataset.csv
```

```
9. import pandas as pd
   import numpy as np
   import seaborn as sns
   import matplotlib.pyplot as plt
   from sklearn.model_selection import train_test_split
   from sklearn.preprocessing import StandardScaler
   from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error
   from sklearn.linear_model import LinearRegression
   from sklearn.linear_model import Lasso
   from sklearn.ensemble import RandomForestRegressor
   from sklearn.svm import SVR
```

```
10. #To know what are the column in the dataset
    dataset.columns
```

```
11. Index(['Table Code', 'State Code', 'District Code', 'Area Name',
          'Total/ Rural/ Urban', 'Age group',
          'Worked for 3 months or more but less than 6 months - Persons',
          'Worked for 3 months or more but less than 6 months - Males',
          'Worked for 3 months or more but less than 6 months - Females',
          'Worked for less than 3 months - Persons',
          'Worked for less than 3 months - Males',
          'Worked for less than 3 months - Females',
          'Industrial Category - A - Cultivators - Persons',
          'Industrial Category - A - Cultivators - Males',
          'Industrial Category - A - Cultivators - Females',
          'Industrial Category - A - Agricultural labourers - Persons',
          'Industrial Category - A - Agricultural labourers - Males',
          'Industrial Category - A - Agricultural labourers - Females',
          'Industrial Category - A - Plantation, Livestock, Forestry, Fishing,
          Hunting and allied activities - Persons',
          'Industrial Category - A - Plantation, Livestock, Forestry, Fishing,
          Hunting and allied activities - Males',
          'Industrial Category - A - Plantation, Livestock, Forestry, Fishing,
          Hunting and allied activities - Females',
          'Industrial Category - B - Persons', 'Industrial Category - B - Males',
          'Industrial Category - B - Females',
          'Industrial Category - C - HHI - Persons',
```

```

'Industrial Category - C - HHI - Males',
'Industrial Category - C - HHI - Females',
'Industrial Category - C - Non HHI - Persons',
'Industrial Category - C - Non HHI - Males',
'Industrial Category - C - Non HHI - Females',
'Industrial Category - D & E - Persons',
'Industrial Category - D & E - Males',
'Industrial Category - D & E - Females',
'Industrial Category - F - Persons', 'Industrial Category - F - Males',
'Industrial Category - F - Females',
'Industrial Category - G - HHI - Persons',
'Industrial Category - G - HHI - Males',
'Industrial Category - G - HHI - Females',
'Industrial Category - G - Non HHI - Persons',
'Industrial Category - G - Non HHI - Males',
'Industrial Category - G - Non HHI - Females',
'Industrial Category - H - Persons', 'Industrial Category - H - Males',
'Industrial Category - H - Females',
'Industrial Category - I - Persons', 'Industrial Category - I - Males',
'Industrial Category - I - Females',
'Industrial Category - J - HHI - Persons',
'Industrial Category - J - HHI - Males',
'Industrial Category - J - HHI - Females',
'Industrial Category - J - Non HHI - Persons',
'Industrial Category - J - Non HHI - Males',
'Industrial Category - J - Non HHI - Females',
'Industrial Category - K to M - Persons',
'Industrial Category - K to M - Males',
'Industrial Category - K to M - Females',
'Industrial Category - N to O - Persons',
'Industrial Category - N to O - Males',
'Industrial Category - N to O - Females',
'Industrial Category - P to Q - Persons',
'Industrial Category - P to Q - Males',
'Industrial Category - P to Q - Females',
'Industrial Category - R to U - HHI - Persons',
'Industrial Category - R to U - HHI - Males',
'Industrial Category - R to U - HHI - Females',
'Industrial Category - R to U - Non HHI - Persons',
'Industrial Category - R to U - Non HHI - Males',
'Industrial Category - R to U - Non HHI - Females'],
dtype='object')

```

12. # TO view TOP Five rows in dataset

```
dataset.head()
```

```

13.  Table Code State Code District Code      Area Name Total/ Rural/ Urban \
0    B0806SC    `33    `000 State - TAMIL NADU              Total
1    B0806SC    `33    `000 State - TAMIL NADU              Total
2    B0806SC    `33    `000 State - TAMIL NADU              Total
3    B0806SC    `33    `000 State - TAMIL NADU              Total
4    B0806SC    `33    `000 State - TAMIL NADU              Total

```

Age group	Worked for 3 months or more but less than 6 months - Persons \
0	Total 1200828
1	5-14 27791
2	15-34 514340
3	35-59 542581
4	60+ 115103

	Worked for 3 months or more but less than 6 months - Males \
0	589003
1	14125
2	259560
3	251957
4	62833

	Worked for 3 months or more but less than 6 months - Females \
0	611825
1	13666
2	254780
3	290624
4	52270

	Worked for less than 3 months - Persons ... \
0	221386 ...
1	2447 ...
2	92423 ...
3	99202 ...
4	27165 ...

	Industrial Category - N to O - Females \
0	3565
1	11
2	1754
3	1619
4	175

	Industrial Category - P to Q - Persons \
0	11080
1	122
2	7536
3	3205
4	211

	Industrial Category - P to Q - Males \
0	4019
1	71
2	2718
3	1131
4	93

	Industrial Category - P to Q - Females \
0	7061
1	51

2	4818
3	2074
4	118

	Industrial Category - R to U - HHI - Persons \
0	16833
1	427
2	8346
3	6591
4	1457

	Industrial Category - R to U - HHI - Males \
0	4266
1	169
2	2127
3	1487
4	483

	Industrial Category - R to U - HHI - Females \
0	12567
1	258
2	6219
3	5104
4	974

	Industrial Category - R to U - Non HHI - Persons \
0	122088
1	19305
2	68929
3	26498
4	7065

	Industrial Category - R to U - Non HHI - Males \
0	55801
1	9774
2	32803
3	9675
4	3394

	Industrial Category - R to U - Non HHI - Females
0	66287
1	9531
2	36126
3	16823
4	3671

[5 rows x 69 columns]

14. # To view *BOTTOM* five rows in dataset
dataset.tail()

15.	Table Code	State Code	District Code	Area Name \
589	B0806SC	`33	`633	District - Tiruppur
590	B0806SC	`33	`633	District - Tiruppur
591	B0806SC	`33	`633	District - Tiruppur
592	B0806SC	`33	`633	District - Tiruppur
593	B0806SC	`33	`633	District - Tiruppur

	Total/ Rural/ Urban	Age group \
589	Urban	`5-14
590	Urban	15-34
591	Urban	35-59
592	Urban	60+
593	Urban	Age not stated

	Worked for 3 months or more but less than 6 months - Persons \
589	272
590	3285
591	3672
592	696
593	2

	Worked for 3 months or more but less than 6 months - Males \
589	129
590	1654
591	1769
592	399
593	1

	Worked for 3 months or more but less than 6 months - Females \
589	143
590	1631
591	1903
592	297
593	1

	Worked for less than 3 months - Persons ... \
589	18 ...
590	473 ...
591	522 ...
592	111 ...
593	0 ...

	Industrial Category - N to O - Females \
589	0
590	20
591	33
592	0
593	0

	Industrial Category - P to Q - Persons \
589	0
590	44

591	35
592	3
593	0

	Industrial Category - P to Q - Males \
589	0
590	15
591	12
592	0
593	0

	Industrial Category - P to Q - Females \
589	0
590	29
591	23
592	3
593	0

	Industrial Category - R to U - HHI - Persons \
589	0
590	62
591	36
592	10
593	0

	Industrial Category - R to U - HHI - Males \
589	0
590	6
591	9
592	3
593	0

	Industrial Category - R to U - HHI - Females \
589	0
590	56
591	27
592	7
593	0

	Industrial Category - R to U - Non HHI - Persons \
589	228
590	675
591	279
592	81
593	0

	Industrial Category - R to U - Non HHI - Males \
589	104
590	247
591	103
592	35
593	0

Industrial Category - R to U - Non HHI - Females	
589	124
590	428
591	176
592	46
593	0

[5 rows x 69 columns]

16. dataset.describe()

17. Worked for 3 months or more but less than 6 months - Persons \

count	5.940000e+02
mean	1.617277e+04
std	7.607172e+04
min	0.000000e+00
25%	2.872500e+02
50%	2.225500e+03
75%	9.628500e+03
max	1.200828e+06

 Worked for 3 months or more but less than 6 months - Males \

count	594.000000
mean	7932.700337
std	36864.822704
min	0.000000
25%	147.250000
50%	1147.000000
75%	4770.500000
max	589003.000000

 Worked for 3 months or more but less than 6 months - Females \

count	594.000000
mean	8240.067340
std	39259.545337
min	0.000000
25%	144.000000
50%	1076.000000
75%	4887.500000
max	611825.000000

 Worked for less than 3 months - Persons \

count	594.000000
mean	2981.629630
std	13909.621137
min	0.000000
25%	27.000000
50%	430.000000
75%	1775.250000
max	221386.000000

	Worked for less than 3 months - Males \
count	594.000000
mean	1338.289562
std	6127.047670
min	0.000000
25%	14.250000
50%	198.500000
75%	774.250000
max	99368.000000

	Worked for less than 3 months - Females \
count	594.000000
mean	1643.340067
std	7808.832522
min	0.000000
25%	13.000000
50%	213.000000
75%	946.500000
max	122018.000000

	Industrial Category - A - Cultivators - Persons \
count	594.000000
mean	865.117845
std	4274.458077
min	0.000000
25%	9.000000
50%	69.500000
75%	466.000000
max	64235.000000

	Industrial Category - A - Cultivators - Males \
count	594.000000
mean	466.424242
std	2298.072295
min	0.000000
25%	5.000000
50%	35.500000
75%	244.250000
max	34632.000000

	Industrial Category - A - Cultivators - Females \
count	594.000000
mean	398.693603
std	1978.682322
min	0.000000
25%	4.000000
50%	32.000000
75%	204.750000
max	29603.000000

	Industrial Category - A - Agricultural labourers - Persons ... \
count	594.000000 ...

mean	12225.616162	...
std	60458.382586	...
min	0.000000	...
25%	79.250000	...
50%	1094.000000	...
75%	6279.750000	...
max	907752.000000	...

	Industrial Category - N to O - Females \
count	594.000000
mean	48.013468
std	222.553500
min	0.000000
25%	0.000000
50%	2.000000
75%	18.000000
max	3565.000000

	Industrial Category - P to Q - Persons \
count	594.000000
mean	149.225589
std	696.553730
min	0.000000
25%	0.000000
50%	14.500000
75%	99.750000
max	11080.000000

	Industrial Category - P to Q - Males \
count	594.000000
mean	54.127946
std	253.067862
min	0.000000
25%	0.000000
50%	6.000000
75%	35.750000
max	4019.000000

	Industrial Category - P to Q - Females \
count	594.000000
mean	95.097643
std	444.011425
min	0.000000
25%	0.000000
50%	6.500000
75%	64.000000
max	7061.000000

	Industrial Category - R to U - HHI - Persons \
count	594.000000
mean	226.707071
std	1039.953069

min	0.000000
25%	0.000000
50%	27.000000
75%	126.750000
max	16833.000000

Industrial Category - R to U - HHI - Males \	
count	594.000000
mean	57.454545
std	265.230865
min	0.000000
25%	0.000000
50%	7.500000
75%	32.000000
max	4266.000000

Industrial Category - R to U - HHI - Females \	
count	594.000000
mean	169.252525
std	776.206806
min	0.000000
25%	0.000000
50%	20.000000
75%	97.500000
max	12567.000000

Industrial Category - R to U - Non HHI - Persons \	
count	594.000000
mean	1644.282828
std	7325.241597
min	0.000000
25%	64.500000
50%	263.500000
75%	994.000000
max	122088.000000

Industrial Category - R to U - Non HHI - Males \	
count	594.000000
mean	751.528620
std	3352.811737
min	0.000000
25%	34.000000
50%	123.000000
75%	447.750000
max	55801.000000

Industrial Category - R to U - Non HHI - Females	
count	594.000000
mean	892.754209
std	3988.125301
min	0.000000
25%	30.500000

50%	135.000000
75%	500.000000
max	66287.000000

[8 rows x 63 columns]

18.Warning: Total number of columns (63) exceeds max_columns (20) limiting to first (20) columns.

19.*#To find how many null values*

```
print(dataset.isnull().sum())
```

```
20.Table Code                                0
   State Code                               0
   District Code                            0
   Area Name                               0
   Total/ Rural/ Urban                      0
   ..
   Industrial Category - R to U - HHI - Males 0
   Industrial Category - R to U - HHI - Females 0
   Industrial Category - R to U - Non HHI - Persons 0
   Industrial Category - R to U - Non HHI - Males 0
   Industrial Category - R to U - Non HHI - Females 0
   Length: 69, dtype: int64
```

21.*#How does the data looks like*

```
print(dataset.shape)      #Total ROWS and COLUMN
print("\n")
print(dataset.dtypes)     #to know the data type each column
```

22. (594, 69)

```
Table Code                                object
State Code                               object
District Code                            object
Area Name                               object
Total/ Rural/ Urban                      object
   ..
   Industrial Category - R to U - HHI - Males  int64
   Industrial Category - R to U - HHI - Females  int64
   Industrial Category - R to U - Non HHI - Persons  int64
   Industrial Category - R to U - Non HHI - Males  int64
   Industrial Category - R to U - Non HHI - Females  int64
   Length: 69, dtype: object
```

23.dataset.info() #This will provide

24.<class 'pandas.core.frame.DataFrame'>

RangeIndex: 594 entries, 0 to 593

Data columns (total 69 columns):

Column

Non-Null Count Dtype

```

---  -----
-----  -----
0    Table Code
594 non-null    object
1    State Code
594 non-null    object
2    District Code
594 non-null    object
3    Area Name
594 non-null    object
4    Total/ Rural/ Urban
594 non-null    object
5    Age group
594 non-null    object
6    Worked for 3 months or more but less than 6 months - Persons
594 non-null    int64
7    Worked for 3 months or more but less than 6 months - Males
594 non-null    int64
8    Worked for 3 months or more but less than 6 months - Females
594 non-null    int64
9    Worked for less than 3 months - Persons
594 non-null    int64
10   Worked for less than 3 months - Males
594 non-null    int64
11   Worked for less than 3 months - Females
594 non-null    int64
12   Industrial Category - A - Cultivators - Persons
594 non-null    int64
13   Industrial Category - A - Cultivators - Males
594 non-null    int64
14   Industrial Category - A - Cultivators - Females
594 non-null    int64
15   Industrial Category - A - Agricultural labourers - Persons
594 non-null    int64
16   Industrial Category - A - Agricultural labourers - Males
594 non-null    int64
17   Industrial Category - A - Agricultural labourers - Females
594 non-null    int64
18   Industrial Category - A - Plantation, Livestock, Forestry, Fishing, Hunting
and allied activities - Persons 594 non-null    int64
19   Industrial Category - A - Plantation, Livestock, Forestry, Fishing, Hunting
and allied activities - Males 594 non-null    int64
20   Industrial Category - A - Plantation, Livestock, Forestry, Fishing, Hunting
and allied activities - Females 594 non-null    int64
21   Industrial Category - B - Persons
594 non-null    int64
22   Industrial Category - B - Males
594 non-null    int64
23   Industrial Category - B - Females
594 non-null    int64
24   Industrial Category - C - HHI - Persons
594 non-null    int64
25   Industrial Category - C - HHI - Males

```

594 non-null int64
 26 Industrial Category - C - HHI - Females
 594 non-null int64
 27 Industrial Category - C - Non HHI - Persons
 594 non-null int64
 28 Industrial Category - C - Non HHI - Males
 594 non-null int64
 29 Industrial Category - C - Non HHI - Females
 594 non-null int64
 30 Industrial Category - D & E - Persons
 594 non-null int64
 31 Industrial Category - D & E - Males
 594 non-null int64
 32 Industrial Category - D & E - Females
 594 non-null int64
 33 Industrial Category - F - Persons
 594 non-null int64
 34 Industrial Category - F - Males
 594 non-null int64
 35 Industrial Category - F - Females
 594 non-null int64
 36 Industrial Category - G - HHI - Persons
 594 non-null int64
 37 Industrial Category - G - HHI - Males
 594 non-null int64
 38 Industrial Category - G - HHI - Females
 594 non-null int64
 39 Industrial Category - G - Non HHI - Persons
 594 non-null int64
 40 Industrial Category - G - Non HHI - Males
 594 non-null int64
 41 Industrial Category - G - Non HHI - Females
 594 non-null int64
 42 Industrial Category - H - Persons
 594 non-null int64
 43 Industrial Category - H - Males
 594 non-null int64
 44 Industrial Category - H - Females
 594 non-null int64
 45 Industrial Category - I - Persons
 594 non-null int64
 46 Industrial Category - I - Males
 594 non-null int64
 47 Industrial Category - I - Females
 594 non-null int64
 48 Industrial Category - J - HHI - Persons
 594 non-null int64
 49 Industrial Category - J - HHI - Males
 594 non-null int64
 50 Industrial Category - J - HHI - Females
 594 non-null int64
 51 Industrial Category - J - Non HHI - Persons
 594 non-null int64


```

52 Industrial Category - J - Non HHI - Males
594 non-null      int64
53 Industrial Category - J - Non HHI - Females
594 non-null      int64
54 Industrial Category - K to M - Persons
594 non-null      int64
55 Industrial Category - K to M - Males
594 non-null      int64
56 Industrial Category - K to M - Females
594 non-null      int64
57 Industrial Category - N to O - Persons
594 non-null      int64
58 Industrial Category - N to O - Males
594 non-null      int64
59 Industrial Category - N to O - Females
594 non-null      int64
60 Industrial Category - P to Q - Persons
594 non-null      int64
61 Industrial Category - P to Q - Males
594 non-null      int64
62 Industrial Category - P to Q - Females
594 non-null      int64
63 Industrial Category - R to U - HHI - Persons
594 non-null      int64
64 Industrial Category - R to U - HHI - Males
594 non-null      int64
65 Industrial Category - R to U - HHI - Females
594 non-null      int64
66 Industrial Category - R to U - Non HHI - Persons
594 non-null      int64
67 Industrial Category - R to U - Non HHI - Males
594 non-null      int64
68 Industrial Category - R to U - Non HHI - Females
594 non-null      int64
dtypes: int64(63), object(6)
memory usage: 320.3+ KB

```

6. DATA VISUALIZATION:

```

import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error
from sklearn.linear_model import LinearRegression
from sklearn.linear_model import Lasso
from sklearn.ensemble import RandomForestRegressor
from sklearn.svm import SVR

```

```

from google.colab import files
uploaded = files.upload()

<IPython.core.display.HTML object>

Saving Dataset.csv to Dataset.csv

df=pd.read_csv('Dataset.csv')

import matplotlib.pyplot as plt

# Assuming you have a DataFrame called df with your data
# You can extract the required columns
age_groups = df["Age group"]
males = df["Worked for 3 months or more but less than 6 months - Males"]
females = df["Worked for 3 months or more but less than 6 months - Females"]

# Create a figure and a set of subplots
fig, ax = plt.subplots()

# Define the width of the bars
width = 0.25

# Set the positions of the bars on the x-axis
x = range(len(age_groups))

# Create the bars
plt.bar([i + width for i in x], males, width, label='Males')
plt.bar([i + width*2 for i in x], females, width, label='Females')

# Set the x-axis ticks and Labels
plt.xlabel('Age Group')
plt.ylabel('Count')
plt.title('Work Duration by Age Group and Gender')
plt.xticks([i + width for i in x], age_groups)

# Add a Legend
plt.legend()

# Show the plot
plt.show()

```



```
import matplotlib.pyplot as plt
```

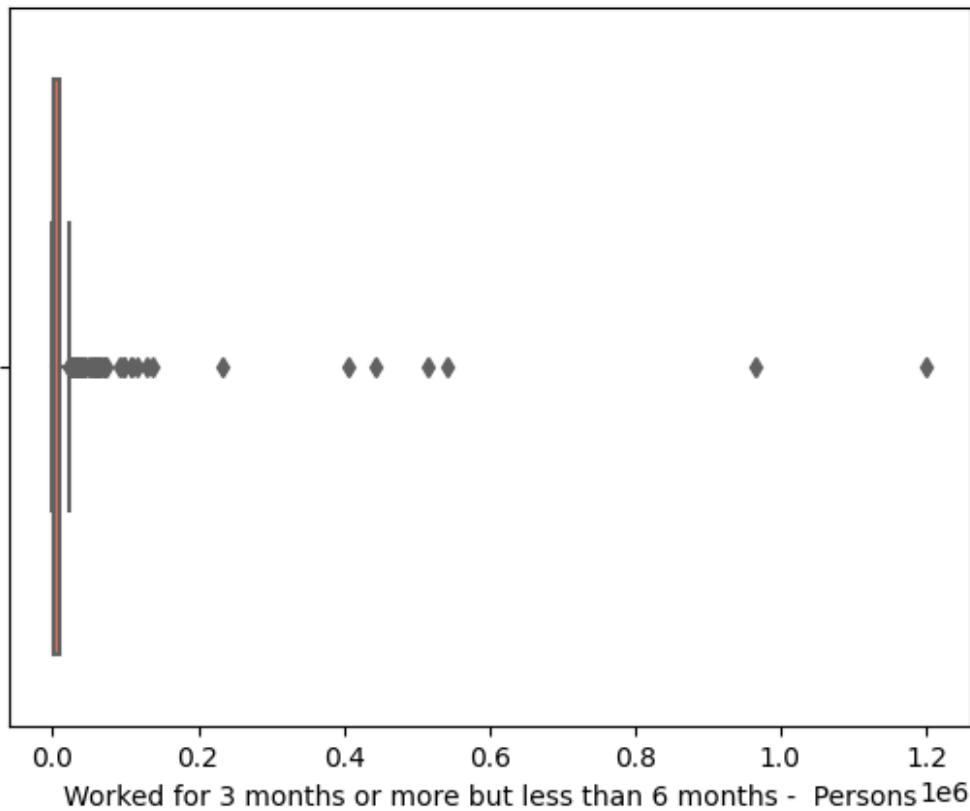
```
# Define the categories and corresponding column names
```

```
categories = ['A - Cultivators', 'A - Agricultural labourers', 'A - Plantation, Livestock, Forestry, Fishing, Hunting and allied activities',
              'B', 'C - HHI', 'C - Non HHI', 'D & E', 'F', 'G - HHI', 'G - Non HHI', 'H', 'I', 'J - HHI', 'J - Non HHI', 'K to M',
              'N to O', 'P to Q', 'R to U - HHI', 'R to U - Non HHI']
```

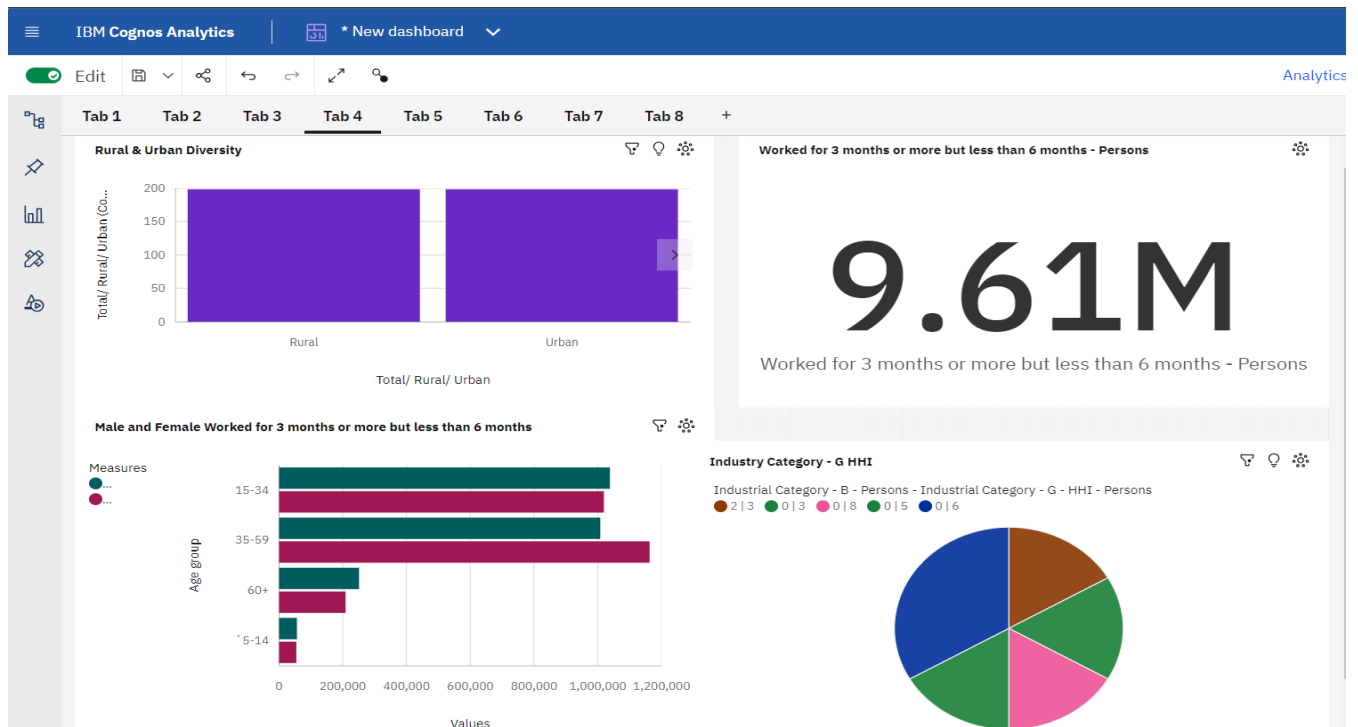
```
column_names = ['Industrial Category - A - Cultivators - Persons', 'Industrial Category - A - Agricultural labourers - Persons', 'Industrial Category - A - Plantation, Livestock, Forestry, Fishing, Hunting and allied activities - Persons',
                'Industrial Category - B - Persons', 'Industrial Category - C - HHI - Persons', 'Industrial Category - C - Non HHI - Persons', 'Industrial Category - D & E - Persons', 'Industrial Category - F - Persons', 'Industrial Category - G - HHI - Persons', 'Industrial Category - G - Non HHI - Persons', 'Industrial Category - H - Persons', 'Industrial Category - I - Persons', 'Industrial Category - J - HHI - Persons', 'Industrial Category - J - Non HHI - Persons', 'Industrial Category - K to M - Persons',
                'Industrial Category - N to O - Persons', 'Industrial Category - P to Q - Persons', 'Industrial Category - R to U - HHI - Persons', 'Industrial Category - R to U - Non HHI - Persons']
```

```
# Extract the data for industrial categories
```

```
category_counts = [df[column].sum() for column in column_names]
```

7. DATA VISUALIZATION USING COGNOS



8. CONCLUSION:

The project concludes by summarizing the data analysis work, emphasizing the use of visualization libraries like Matplotlib and Seaborn, and the application of data-driven techniques for understanding marginal workers. In this project, we embarked on a comprehensive journey to understand and optimize marginal workers through data analysis. By employing a structured approach and leveraging powerful data analysis tools, we've unveiled insights and established a foundation for data-driven decision-making within the socio economic sector. Throughout the project, we've emphasized the importance of data preprocessing as a critical step. It is essential for refining and enhancing the quality of the data, which, in turn, paves the way for more accurate predictions and insights. These insights have the potential to support a wide range of marginal workers improvement initiatives, ultimately benefiting commuters and urban development. Through the fusion of cutting-edge technologies and methodologies, our ultimate goal is to provide a comprehensive and insightful solution for evaluating and enhancing marginal workers efficiency.