ADS-PHASE 4

ASSESSEMENT OF MARGINAL WORKERS IN TAMILNADU-A SOCIOECONIMC ANALYSIS

1. IMPORTING THE REQUIRED LIBRARIES
2. IMPORTING THE DATASET
3. CREATING A MATRIX
4. HANDALING THE MISSING DATAS
5. ENCODING THE CATEGORICAL DATA
6. SPLITTING THE DATA SET
7. FEATURE SCALING

CODING

[3]:
```
#importing the required libraries
import pandas as pd
import numpy as np
#to load the dataset
a=pd.read_csv("C:\\Users\\00AD_
 ↪LAB\\Downloads\\DDW_B06ST_3300_State_TAMIL_NADU-2011.csv")
print(a)
```

```
    Table Code State Code District Code            Area Name  \
0      B0906ST        `33          `000   State - TAMIL NADU
1      B0906ST        `33          `000   State - TAMIL NADU
2      B0906ST        `33          `000   State - TAMIL NADU
3      B0906ST        `33          `000   State - TAMIL NADU
4      B0906ST        `33          `000   State - TAMIL NADU
..         ...         ...           ...                  ...
589    B0906ST        `33          `633  District - Tiruppur
590    B0906ST        `33          `633  District - Tiruppur
591    B0906ST        `33          `633  District - Tiruppur
592    B0906ST        `33          `633  District - Tiruppur
593    B0906ST        `33          `633  District - Tiruppur

    Total/ Rural/ Urban       Age group  \
0                  Total           Total
1                  Total          `5-14
2                  Total          15-34
3                  Total          35-59
```

4	Total	60+
..	...	...
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 \
0	66695
1	2637
2	31370

3	27418
4	5219
..	...
589	4
590	54
591	38
592	4
593	0

	Worked for 3 months or more but less than 6 months - Males \
0	32578
1	1345
2	15374
3	12976
4	2865
..	...
589	4
590	35
591	24
592	2
593	0

	Worked for 3 months or more but less than 6 months - Females \
0	34117
1	1292
2	15996
3	14442
4	2354
..	...
589	0
590	19
591	14
592	2
593	0

	Worked for less than 3 months - Persons ... \	
0	12153	...
1	356	...
2	5714	...
3	4757	...
4	1320	...
..	...	...
589	0	...
590	14	...
591	2	...
592	0	...
593	0	...

	Industrial Category - N to O - Females	\
0	110	
1	0	
2	46	
3	52	
4	12	
..	...	
589	0	
590	0	
591	0	
592	0	
593	0	

	Industrial Category - P to Q - Persons	\
0	278	
1	14	
2	198	
3	60	
4	6	
..	...	
589	0	
590	4	
591	2	
592	0	
593	0	

	Industrial Category - P to Q - Males	\
0	128	
1	6	
2	94	
3	24	
4	4	
..	...	
589	0	
590	2	
591	2	
592	0	
593	0	

	Industrial Category - P to Q - Females	\
0	150	
1	8	
2	104	
3	36	
4	2	
..	...	
589	0	
590	2	

591	0
592	0
593	0

	Industrial Category - R to U - HHI - Persons\
0	978
1	36
2	508
3	356
4	78
..	...
589	0
590	0
591	0
592	0
593	0

	Industrial Category - R to U - HHI - Males \
0	226
1	16
2	114
3	68
4	28
..	...
589	0
590	0
591	0
592	0
593	0

	Industrial Category - R to U - HHI - Females \
0	752
1	20
2	394
3	288
4	50
..	...
589	0
590	0
591	0
592	0
593	0

	Industrial Category - R to U - Non HHI - Persons \
0	5152
1	1156
2	2602
3	1058

```
4                                                    318
..                                                   ...
589                                                    4
590                                                   12
591                                                   28
592                                                    4
593                                                    0

     Industrial Category - R to U - Non HHI - Males\
0                                                 2270
1                                                  586
2                                                 1178
3                                                  348
4                                                  152
..                                                 ...
589                                                  4
590                                                  6
591                                                 16
592                                                  2
593                                                  0

     Industrial Category - R to U - Non HHI - Females
0                                                 2882
1                                                  570
2                                                 1424
3                                                  710
4                                                  166
..                                                 ...
589                                                  0
590                                                  6
591                                                 12
592                                                  2
593                                                  0

[594 rows x 69 columns]
```

[18]:
```
import pandas as pd
import numpy as np
#to convert the dataset into array
a=pd.read_csv("C:\\Users\\00AD_
  ↪LAB\\Downloads\\DDW_B06ST_3300_State_TAMIL_NADU-2011.csv")
array=a.to_numpy()
array1=np.array(a)
print(array)
print(array1)
```

```
[['B0906ST' '`33' '`000' ... 5152 2270 2882]
 ['B0906ST' '`33' '`000' ... 1156 586 570]
```

```
 ['B0906ST' '`33' '`000' … 2602 1178 1424]
 …
 ['B0906ST' '`33' '`633' … 28 16 12]
 ['B0906ST' '`33' '`633' … 4 2 2]
 ['B0906ST' '`33' '`633' … 0 0 0]]
[['B0906ST' '`33' '`000' … 5152 2270 2882]
 ['B0906ST' '`33' '`000' … 1156 586 570]
 ['B0906ST' '`33' '`000' … 2602 1178 1424]
 …
 ['B0906ST' '`33' '`633' … 28 16 12]
 ['B0906ST' '`33' '`633' … 4 2 2]
 ['B0906ST' '`33' '`633' … 0 0 0]]
```

[48]:
```
#to create the matric from the dataset:
a=pd.read_csv("C:\\Users\\00AD␣
↪LAB\\Downloads\\DDW_B06ST_3300_State_TAMIL_NADU-2011.csv")
matrix = np.array([["Worked for less than 3 months - Persons"], ["Industrial␣
↪Category - P to Q - Persons"]])
print(matrix[0, 0])
#print(matrix[1, 1])
print("Shape of the matrix:", matrix.shape)  # Output: (3, 3)

# Matrix transpose
transpose_matrix = array1.T
print("Transposed matrix:")
print(transpose_matrix)
```

```
Worked for less than 3 months - Persons
Shape of the matrix: (2, 1)
Transposed matrix:
[['B0906ST' 'B0906ST' 'B0906ST' ... 'B0906ST' 'B0906ST' 'B0906ST']
 ['`33' '`33' '`33' ... '`33' '`33' '`33']
 ['`000' '`000' '`000' ... '`633' '`633' '`633']
 …
 [5152 1156 2602 ... 28 4 0]
 [2270 586 1178 ... 16 2 0]
 [2882 570 1424 ... 12 2 0]]
```

[11]:
```
from sklearn.impute import SimpleImputer
# to handle the dataset with missing values
data = np.array([[1, 2, np.nan], [4, np.nan, 6], [7, 8, 9]])

# Create an instance of the SimpleImputer class
imputer = SimpleImputer(strategy="mean")  # Other strategies: 'median',␣
↪'most_frequent', 'constant'

# Fit the imputer to the data and transform it
```

```
imputed_data = imputer.fit_transform(data)

# The missing values have been replaced
print("Original Data:")
print(data)

print("\nImputed Data:")
print(imputed_data)
```

```
Original Data:
[[ 1.  2. nan]
 [ 4. nan  6.]
 [ 7.  8.  9.]]

Imputed Data:
[[1.  2.  7.5]
 [4.  5.  6. ]
 [7.  8.  9. ]]
```

[38]:
```
#encoding the categorical dataset:
from sklearn.preprocessing import OneHotEncoder
import numpy as np
import pandas as pd

a=pd.read_csv("C:\\Users\\OOAD_
 ↪LAB\\Downloads\\DDW_B06ST_3300_State_TAMIL_NADU-2011.csv")


# Sample data
data = ['Age group','Worked for less than 3 months - Persons','Industrial_
 ↪Category - P to Q - Females']

# Initialize the OneHotEncoder
onehot_encoder = OneHotEncoder(sparse=False)  # Use sparse=False to get a dense_
 ↪matrix

# Fit and transform the data (reshape is necessary)
encoded_data = onehot_encoder.fit_transform(np.array(data).reshape(-1, 1))

print("Original data:", data)
print("One-Hot Encoded data:")
print(encoded_data)
```

```
Original data: ['Age group', 'Worked for less than 3 months - Persons',
'Industrial Category - P to Q - Females']
One-Hot Encoded data:
[[1. 0. 0.]
```

```
[0. 0. 1.]
[0. 1. 0.]]
```

[37]:
```
#feature scalling process:
from sklearn.preprocessing import StandardScaler
import numpy as np
a=pd.read_csv("C:\\Users\\00AD_
 ↪LAB\\Downloads\\DDW_B06ST_3300_State_TAMIL_NADU-2011.csv")
# Sample data
data= np.array(mydataset["Worked for less than 3 months - Persons"]).
 ↪reshape(-1,1)

# Initialize the StandardScaler
scaler = StandardScaler()

# Fit and transform the data
scaled_data = scaler.fit_transform(data)

# Print the scaled data
print("Original data:")
print(data)
print("Scaled data:")
print(scaled_data)
```

```
Original data:
[[12153]
 [  356]
 [ 5714]
 [ 4757]
 [ 1320]
 [    6]
 [10306]
 [  334]
 [ 4817]
 [ 3995]
 [ 1154]
 [    6]
 [ 1847]
 [   22]
 [  897]
 [  762]
 [  166]
 [    0]
 [ 1366]
 [   41]
 [  636]
 [  556]
 [  131]
```

[2]
[1226]
[39]
[571]
[497]
[117]
[2]
[140]
[2]
[65]
[59]
[14]
[0]
[66]
[2]
[37]
[27]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[66]
[2]
[37]
[27]
[0]
[0]
[998]
[14]
[477]
[413]
[94]
[0]
[764]
[13]
[354]
[324]
[73]
[0]
[234]
[1]
[123]
[89]
[21]

[0]
[2346]
[69]
[1146]
[926]
[203]
[2]
[2210]
[67]
[1082]
[876]
[183]
[2]
[136]
[2]
[64]
[50]
[20]
[0]
[1635]
[57]
[787]
[597]
[194]
[0]
[1605]
[53]
[773]
[587]
[192]
[0]
[30]
[4]
[14]
[10]
[2]
[0]
[1186]
[27]
[573]
[455]
[130]
[1]
[1111]
[27]
[538]
[427]
[118]

[1]
[75]
[0]
[35]
[28]
[12]
[0]
[534]
[17]
[233]
[181]
[103]
[0]
[515]
[17]
[225]
[175]
[98]
[0]
[19]
[0]
[8]
[6]
[5]
[0]
[172]
[15]
[78]
[43]
[36]
[0]
[172]
[15]
[78]
[43]
[36]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[205]
[5]
[83]
[85]
[32]

[0]
[203]
[3]
[83]
[85]
[32]
[0]
[2]
[2]
[0]
[0]
[0]
[0]
[551]
[11]
[291]
[202]
[46]
[1]
[312]
[11]
[148]
[117]
[35]
[1]
[239]
[0]
[143]
[85]
[11]
[0]
[114]
[3]
[53]
[39]
[19]
[0]
[110]
[3]
[51]
[37]
[19]
[0]
[4]
[0]
[2]
[2]
[0]

[0]
[2]
[0]
[0]
[0]
[2]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[2]
[0]
[0]
[0]
[2]
[0]
[44]
[0]
[24]
[10]
[10]
[0]
[28]
[0]
[12]
[6]
[10]
[0]
[16]
[0]
[12]
[4]
[0]
[0]
[77]
[1]
[43]
[31]
[2]
[0]
[22]
[1]
[17]
[3]
[1]

[0]
[55]
[0]
[26]
[28]
[1]
[0]
[211]
[7]
[76]
[91]
[37]
[0]
[207]
[7]
[74]
[91]
[35]
[0]
[4]
[0]
[2]
[0]
[2]
[0]
[457]
[13]
[213]
[205]
[26]
[0]
[259]
[13]
[119]
[110]
[17]
[0]
[198]
[0]
[94]
[95]
[9]
[0]
[63]
[3]
[20]
[36]
[4]

[0]
[24]
[2]
[6]
[13]
[3]
[0]
[39]
[1]
[14]
[23]
[1]
[0]
[24]
[0]
[10]
[13]
[1]
[0]
[17]
[0]
[5]
[11]
[1]
[0]
[7]
[0]
[5]
[2]
[0]
[0]
[18]
[0]
[7]
[8]
[3]
[0]
[14]
[0]
[3]
[8]
[3]
[0]
[4]
[0]
[4]
[0]
[0]

[0]
[51]
[6]
[26]
[17]
[2]
[0]
[51]
[6]
[26]
[17]
[2]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[21]
[1]
[2]
[9]
[9]
[0]
[3]
[1]
[0]
[1]
[1]
[0]
[18]
[0]
[2]
[8]
[8]
[0]
[66]
[0]
[26]
[31]
[9]
[0]
[32]
[0]
[8]
[15]
[9]

```
[  0]
[ 34]
[  0]
[ 18]
[ 16]
[  0]
[  0]
[  2]
[  0]
[  2]
[  0]
[  0]
[  0]
[  2]
[  0]
[  2]
[  0]
[  0]
[  0]
[  0]
[  0]
[  0]
[  0]
[  0]
[  0]
[ 11]
[  0]
[  6]
[  3]
[  2]
[  0]
[  9]
[  0]
[  4]
[  3]
[  2]
[  0]
[  2]
[  0]
[  2]
[  0]
[  0]
[  0]
[ 13]
[  0]
[  6]
[  7]
[  0]
```

[0]
[8]
[0]
[4]
[4]
[0]
[0]
[5]
[0]
[2]
[3]
[0]
[0]
[12]
[2]
[3]
[7]
[0]
[0]
[9]
[2]
[2]
[5]
[0]
[0]
[3]
[0]
[1]
[2]
[0]
[0]
[179]
[3]
[69]
[65]
[42]
[0]
[30]
[1]
[12]
[8]
[9]
[0]
[149]
[2]
[57]
[57]
[33]

[0]
[274]
[4]
[113]
[132]
[25]
[0]
[31]
[0]
[8]
[11]
[12]
[0]
[243]
[4]
[105]
[121]
[13]
[0]
[574]
[30]
[270]
[198]
[76]
[0]
[556]
[30]
[262]
[190]
[74]
[0]
[18]
[0]
[8]
[8]
[2]
[0]
[313]
[14]
[144]
[120]
[35]
[0]
[298]
[14]
[139]
[112]
[33]

```
 [      0]
 [     15]
 [      0]
 [      5]
 [      8]
 [      2]
 [      0]
 [    551]
 [     11]
 [    245]
 [    248]
 [     47]
 [      0]
 [    477]
 [      9]
 [    210]
 [    219]
 [     39]
 [      0]
 [     74]
 [      2]
 [     35]
 [     29]
 [      8]
 [      0]
 [     17]
 [      0]
 [     15]
 [      2]
 [      0]
 [      0]
 [      1]
 [      0]
 [      1]
 [      0]
 [      0]
 [      0]
 [     16]
 [      0]
 [     14]
 [      2]
 [      0]
 [      0]]
Scaled data:
[[ 1.50388006e+01]
 [ 2.41240535e-01]
 [ 6.96204472e+00]
 [ 5.76163233e+00]
```

[1.45043337e+00]
[-1.97781762e-01]
[1.27220173e+01]
[2.13644848e-01]
[5.83689329e+00]
[4.80581807e+00]
[1.24221137e+00]
[-1.97781762e-01]
[2.11147552e+00]
[-1.77712171e-01]
[9.19843570e-01]
[7.50506399e-01]
[2.91414512e-03]
[-2.05307858e-01]
[1.50813345e+00]
[-1.53879532e-01]
[5.92458372e-01]
[4.92110419e-01]
[-4.09880846e-02]
[-2.02799160e-01]
[1.33252453e+00]
[-1.56388231e-01]
[5.10925660e-01]
[4.18103803e-01]
[-5.85489764e-02]
[-2.02799160e-01]
[-2.96989398e-02]
[-2.02799160e-01]
[-1.23775146e-01]
[-1.31301243e-01]
[-1.87746967e-01]
[-2.05307858e-01]
[-1.22520797e-01]
[-2.02799160e-01]
[-1.58896930e-01]
[-1.71440424e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.22520797e-01]
[-2.02799160e-01]
[-1.58896930e-01]
[-1.71440424e-01]

[-2.05307858e-01]
[-2.05307858e-01]
[1.04653286e+00]
[-1.87746967e-01]
[3.93016814e-01]
[3.12738452e-01]
[-8.73990131e-02]
[-2.05307858e-01]
[7.53015098e-01]
[-1.89001316e-01]
[2.38731836e-01]
[2.01101353e-01]
[-1.13740351e-01]
[-2.05307858e-01]
[8.82099056e-02]
[-2.04053509e-01]
[-5.10228799e-02]
[-9.36707602e-02]
[-1.78966521e-01]
[-2.05307858e-01]
[2.73739588e+00]
[-1.18757749e-01]
[1.23217658e+00]
[9.56219704e-01]
[4.93250736e-02]
[-2.02799160e-01]
[2.56680436e+00]
[-1.21266447e-01]
[1.15189821e+00]
[8.93502233e-01]
[2.42380852e-02]
[-2.02799160e-01]
[-3.47163375e-02]
[-2.02799160e-01]
[-1.25029496e-01]
[-1.42590387e-01]
[-1.80220870e-01]
[-2.05307858e-01]
[1.84555344e+00]
[-1.33809942e-01]
[7.81865134e-01]
[5.43538745e-01]
[3.80359288e-02]
[-2.05307858e-01]
[1.80792296e+00]
[-1.38827339e-01]
[7.64304243e-01]
[5.30995251e-01]

[3.55272300e-02]
[-2.05307858e-01]
[-1.67677376e-01]
[-2.00290461e-01]
[-1.87746967e-01]
[-1.92764364e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[1.28235055e+00]
[-1.71440424e-01]
[5.13434359e-01]
[3.65421127e-01]
[-4.22424340e-02]
[-2.04053509e-01]
[1.18827435e+00]
[-1.71440424e-01]
[4.69532129e-01]
[3.30299343e-01]
[-5.72946270e-02]
[-2.04053509e-01]
[-1.11231652e-01]
[-2.05307858e-01]
[-1.61405629e-01]
[-1.70186075e-01]
[-1.90255665e-01]
[-2.05307858e-01]
[4.64514731e-01]
[-1.83983918e-01]
[8.69555562e-02]
[2.17293864e-02]
[-7.61098683e-02]
[-2.05307858e-01]
[4.40682092e-01]
[-1.83983918e-01]
[7.69207608e-02]
[1.42032899e-02]
[-8.23816154e-02]
[-2.05307858e-01]
[-1.81475219e-01]
[-2.05307858e-01]
[-1.95273063e-01]
[-1.97781762e-01]
[-1.99036111e-01]
[-2.05307858e-01]
[1.04402416e-02]
[-1.86492617e-01]
[-1.07468604e-01]
[-1.51370833e-01]

[-1.60151279e-01]
[-2.05307858e-01]
[1.04402416e-02]
[-1.86492617e-01]
[-1.07468604e-01]
[-1.51370833e-01]
[-1.60151279e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[5.18337725e-02]
[-1.99036111e-01]
[-1.01196857e-01]
[-9.86881578e-02]
[-1.65168677e-01]
[-2.05307858e-01]
[4.93250736e-02]
[-2.01544810e-01]
[-1.01196857e-01]
[-9.86881578e-02]
[-1.65168677e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[4.85838671e-01]
[-1.91510015e-01]
[1.59707823e-01]
[4.80707242e-02]
[-1.47607785e-01]
[-2.04053509e-01]
[1.86049160e-01]
[-1.91510015e-01]
[-1.96641444e-02]
[-5.85489764e-02]
[-1.61405629e-01]
[-2.04053509e-01]
[9.44816527e-02]
[-2.05307858e-01]
[-2.59358915e-02]
[-9.86881578e-02]

[-1.91510015e-01]
[-2.05307858e-01]
[-6.23120247e-02]
[-2.01544810e-01]
[-1.38827339e-01]
[-1.56388231e-01]
[-1.81475219e-01]
[-2.05307858e-01]
[-6.73294224e-02]
[-2.01544810e-01]
[-1.41336038e-01]
[-1.58896930e-01]
[-1.81475219e-01]
[-2.05307858e-01]
[-2.00290461e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-1.50116484e-01]
[-2.05307858e-01]
[-1.75203472e-01]
[-1.92764364e-01]
[-1.92764364e-01]
[-2.05307858e-01]
[-1.70186075e-01]
[-2.05307858e-01]
[-1.90255665e-01]
[-1.97781762e-01]

[-1.92764364e-01]
[-2.05307858e-01]
[-1.85238268e-01]
[-2.05307858e-01]
[-1.90255665e-01]
[-2.00290461e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.08722953e-01]
[-2.04053509e-01]
[-1.51370833e-01]
[-1.66423026e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-1.77712171e-01]
[-2.04053509e-01]
[-1.83983918e-01]
[-2.01544810e-01]
[-2.04053509e-01]
[-2.05307858e-01]
[-1.36318640e-01]
[-2.05307858e-01]
[-1.72694774e-01]
[-1.70186075e-01]
[-2.04053509e-01]
[-2.05307858e-01]
[5.93598690e-02]
[-1.96527413e-01]
[-1.09977303e-01]
[-9.11620613e-02]
[-1.58896930e-01]
[-2.05307858e-01]
[5.43424713e-02]
[-1.96527413e-01]
[-1.12486001e-01]
[-9.11620613e-02]
[-1.61405629e-01]
[-2.05307858e-01]
[-2.00290461e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[3.67929826e-01]
[-1.89001316e-01]
[6.18685678e-02]
[5.18337725e-02]

[-1.72694774e-01]
[-2.05307858e-01]
[1.19568641e-01]
[-1.89001316e-01]
[-5.60402776e-02]
[-6.73294224e-02]
[-1.83983918e-01]
[-2.05307858e-01]
[4.30533265e-02]
[-2.05307858e-01]
[-8.73990131e-02]
[-8.61446636e-02]
[-1.94018714e-01]
[-2.05307858e-01]
[-1.26283845e-01]
[-2.01544810e-01]
[-1.80220870e-01]
[-1.60151279e-01]
[-2.00290461e-01]
[-2.05307858e-01]
[-1.75203472e-01]
[-2.02799160e-01]
[-1.97781762e-01]
[-1.89001316e-01]
[-2.01544810e-01]
[-2.05307858e-01]
[-1.56388231e-01]
[-2.04053509e-01]
[-1.87746967e-01]
[-1.76457822e-01]
[-2.04053509e-01]
[-2.05307858e-01]
[-1.75203472e-01]
[-2.05307858e-01]
[-1.92764364e-01]
[-1.89001316e-01]
[-2.04053509e-01]
[-2.05307858e-01]
[-1.83983918e-01]
[-2.05307858e-01]
[-1.99036111e-01]
[-1.91510015e-01]
[-2.04053509e-01]
[-2.05307858e-01]
[-1.96527413e-01]
[-2.05307858e-01]
[-1.99036111e-01]
[-2.02799160e-01]

[-2.05307858e-01]
[-2.05307858e-01]
[-1.82729569e-01]
[-2.05307858e-01]
[-1.96527413e-01]
[-1.95273063e-01]
[-2.01544810e-01]
[-2.05307858e-01]
[-1.87746967e-01]
[-2.05307858e-01]
[-2.01544810e-01]
[-1.95273063e-01]
[-2.01544810e-01]
[-2.05307858e-01]
[-2.00290461e-01]
[-2.05307858e-01]
[-2.00290461e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.41336038e-01]
[-1.97781762e-01]
[-1.72694774e-01]
[-1.83983918e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-1.41336038e-01]
[-1.97781762e-01]
[-1.72694774e-01]
[-1.83983918e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.78966521e-01]
[-2.04053509e-01]
[-2.02799160e-01]
[-1.94018714e-01]
[-1.94018714e-01]
[-2.05307858e-01]
[-2.01544810e-01]
[-2.04053509e-01]
[-2.05307858e-01]
[-2.04053509e-01]

[-2.04053509e-01]
[-2.05307858e-01]
[-1.82729569e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-1.95273063e-01]
[-1.95273063e-01]
[-2.05307858e-01]
[-1.22520797e-01]
[-2.05307858e-01]
[-1.72694774e-01]
[-1.66423026e-01]
[-1.94018714e-01]
[-2.05307858e-01]
[-1.65168677e-01]
[-2.05307858e-01]
[-1.95273063e-01]
[-1.86492617e-01]
[-1.94018714e-01]
[-2.05307858e-01]
[-1.62659978e-01]
[-2.05307858e-01]
[-1.82729569e-01]
[-1.85238268e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.91510015e-01]
[-2.05307858e-01]
[-1.97781762e-01]
[-2.01544810e-01]

[-2.02799160e-01]
[-2.05307858e-01]
[-1.94018714e-01]
[-2.05307858e-01]
[-2.00290461e-01]
[-2.01544810e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.89001316e-01]
[-2.05307858e-01]
[-1.97781762e-01]
[-1.96527413e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.95273063e-01]
[-2.05307858e-01]
[-2.00290461e-01]
[-2.00290461e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.99036111e-01]
[-2.05307858e-01]
[-2.02799160e-01]
[-2.01544810e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.90255665e-01]
[-2.02799160e-01]
[-2.01544810e-01]
[-1.96527413e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.94018714e-01]
[-2.02799160e-01]
[-2.02799160e-01]
[-1.99036111e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.01544810e-01]
[-2.05307858e-01]
[-2.04053509e-01]
[-2.02799160e-01]

[-2.05307858e-01]
[-2.05307858e-01]
[1.92206876e-02]
[-2.01544810e-01]
[-1.18757749e-01]
[-1.23775146e-01]
[-1.52625183e-01]
[-2.05307858e-01]
[-1.67677376e-01]
[-2.04053509e-01]
[-1.90255665e-01]
[-1.95273063e-01]
[-1.94018714e-01]
[-2.05307858e-01]
[-1.84097950e-02]
[-2.02799160e-01]
[-1.33809942e-01]
[-1.33809942e-01]
[-1.63914328e-01]
[-2.05307858e-01]
[1.38383882e-01]
[-2.00290461e-01]
[-6.35663741e-02]
[-3.97337351e-02]
[-1.73949123e-01]
[-2.05307858e-01]
[-1.66423026e-01]
[-2.05307858e-01]
[-1.95273063e-01]
[-1.91510015e-01]
[-1.90255665e-01]
[-2.05307858e-01]
[9.94990504e-02]
[-2.00290461e-01]
[-7.36011694e-02]
[-5.35315787e-02]
[-1.89001316e-01]
[-2.05307858e-01]
[5.14688708e-01]
[-1.67677376e-01]
[1.33366485e-01]
[4.30533265e-02]
[-1.09977303e-01]
[-2.05307858e-01]
[4.92110419e-01]
[-1.67677376e-01]
[1.23331689e-01]
[3.30185312e-02]

[-1.12486001e-01]
[-2.05307858e-01]
[-1.82729569e-01]
[-2.05307858e-01]
[-1.95273063e-01]
[-1.95273063e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[1.87303510e-01]
[-1.87746967e-01]
[-2.46815421e-02]
[-5.47859282e-02]
[-1.61405629e-01]
[-2.05307858e-01]
[1.68488268e-01]
[-1.87746967e-01]
[-3.09532892e-02]
[-6.48207235e-02]
[-1.63914328e-01]
[-2.05307858e-01]
[-1.86492617e-01]
[-2.05307858e-01]
[-1.99036111e-01]
[-1.95273063e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[4.85838671e-01]
[-1.91510015e-01]
[1.02007749e-01]
[1.05770797e-01]
[-1.46353436e-01]
[-2.05307858e-01]
[3.93016814e-01]
[-1.94018714e-01]
[5.81055196e-02]
[6.93946643e-02]
[-1.56388231e-01]
[-2.05307858e-01]
[-1.12486001e-01]
[-2.02799160e-01]
[-1.61405629e-01]
[-1.68931725e-01]
[-1.95273063e-01]
[-2.05307858e-01]
[-1.83983918e-01]
[-2.05307858e-01]
[-1.86492617e-01]
[-2.02799160e-01]

```
[-2.05307858e-01]
[-2.05307858e-01]
[-2.04053509e-01]
[-2.05307858e-01]
[-2.04053509e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-2.05307858e-01]
[-1.85238268e-01]
[-2.05307858e-01]
[-1.87746967e-01]
[-2.02799160e-01]
[-2.05307858e-01]
[-2.05307858e-01]]
```

[21]:
```
#to create the input array for train & test process: x=np.array(mydataset["Worked for less than 3 months - Males"]).reshape(-1,1) print(x)
```

```
[[5414]
 [ 184]
 [2511]
 [2072]
 [ 647]
 [   0]
 [4536]
 [ 173]
 [2085]
 [1734]
 [ 544]
 [   0]
 [ 878]
 [  11]
 [ 426]
 [ 338]
 [ 103]
 [   0]
 [ 647]
 [  21]
 [ 284]
 [ 263]
 [  79]
 [   0]
 [ 565]
 [  19]
 [ 247]
 [ 229]
 [  70]
```

[0]
[82]
[2]
[37]
[34]
[9]
[0]
[38]
[0]
[22]
[16]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[38]
[0]
[22]
[16]
[0]
[0]
[507]
[10]
[229]
[223]
[45]
[0]
[374]
[9]
[162]
[169]
[34]
[0]
[133]
[1]
[67]
[54]
[11]
[0]
[1110]
[28]
[543]
[433]
[106]

[0]
[1032]
[26]
[507]
[408]
[91]
[0]
[78]
[2]
[36]
[25]
[15]
[0]
[652]
[28]
[314]
[240]
[70]
[0]
[644]
[28]
[313]
[235]
[68]
[0]
[8]
[0]
[1]
[5]
[2]
[0]
[470]
[19]
[209]
[175]
[67]
[0]
[446]
[19]
[199]
[167]
[61]
[0]
[24]
[0]
[10]
[8]
[6]

[0]
[241]
[12]
[103]
[77]
[49]
[0]
[231]
[12]
[97]
[77]
[45]
[0]
[10]
[0]
[6]
[0]
[4]
[0]
[61]
[6]
[30]
[10]
[15]
[0]
[61]
[6]
[30]
[10]
[15]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[89]
[3]
[38]
[34]
[14]
[0]
[87]
[1]
[38]
[34]
[14]

[0]
[2]
[2]
[0]
[0]
[0]
[0]
[261]
[8]
[143]
[88]
[22]
[0]
[137]
[8]
[70]
[43]
[16]
[0]
[124]
[0]
[73]
[45]
[6]
[0]
[54]
[2]
[25]
[16]
[11]
[0]
[54]
[2]
[25]
[16]
[11]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]

[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[20]
[0]
[12]
[4]
[4]
[0]
[10]
[0]
[4]
[2]
[4]
[0]
[10]
[0]
[8]
[2]
[0]
[0]
[35]
[0]
[19]
[15]
[1]
[0]
[8]
[0]
[7]
[1]
[0]
[0]
[27]
[0]
[12]
[14]
[1]

[0]
[100]
[7]
[33]
[42]
[18]
[0]
[98]
[7]
[32]
[42]
[17]
[0]
[2]
[0]
[1]
[0]
[1]
[0]
[152]
[3]
[71]
[62]
[16]
[0]
[119]
[3]
[52]
[51]
[13]
[0]
[33]
[0]
[19]
[11]
[3]
[0]
[30]
[3]
[12]
[13]
[2]
[0]
[10]
[2]
[2]
[5]
[1]

[0]
[20]
[1]
[10]
[8]
[1]
[0]
[19]
[0]
[8]
[10]
[1]
[0]
[13]
[0]
[4]
[8]
[1]
[0]
[6]
[0]
[4]
[2]
[0]
[0]
[8]
[0]
[2]
[3]
[3]
[0]
[6]
[0]
[0]
[3]
[3]
[0]
[2]
[0]
[2]
[0]
[0]
[0]
[40]
[6]
[18]
[14]
[2]

[0]
[40]
[6]
[18]
[14]
[2]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[10]
[0]
[0]
[2]
[8]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[10]
[0]
[0]
[2]
[8]
[0]
[19]
[0]
[4]
[12]
[3]
[0]
[13]
[0]
[2]
[8]
[3]
[0]
[6]
[0]
[2]
[4]
[0]

[0]
[2]
[0]
[2]
[0]
[0]
[0]
[2]
[0]
[2]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[6]
[0]
[4]
[1]
[1]
[0]
[4]
[0]
[2]
[1]
[1]
[0]
[2]
[0]
[2]
[0]
[0]
[0]
[6]
[0]
[2]
[4]
[0]
[0]
[4]
[0]
[2]
[2]
[0]

[0]
[2]
[0]
[0]
[2]
[0]
[0]
[5]
[0]
[0]
[5]
[0]
[0]
[5]
[0]
[0]
[5]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[0]
[107]
[0]
[50]
[32]
[25]
[0]
[19]
[0]
[9]
[6]
[4]
[0]
[88]
[0]
[41]
[26]
[21]
[0]
[139]
[2]
[54]
[68]
[15]

[0]
[11]
[0]
[2]
[0]
[9]
[0]
[128]
[2]
[52]
[68]
[6]
[0]
[187]
[14]
[97]
[52]
[24]
[0]
[180]
[14]
[94]
[48]
[24]
[0]
[7]
[0]
[3]
[4]
[0]
[0]
[161]
[7]
[71]
[60]
[23]
[0]
[155]
[7]
[68]
[59]
[21]
[0]
[6]
[0]
[3]
[1]
[2]

```
[    0]
[  237]
[    5]
[  111]
[   98]
[   23]
[    0]
[  207]
[    4]
[   96]
[   91]
[   16]
[    0]
[   30]
[    1]
[   15]
[    7]
[    7]
[    0]
[    1]
[    0]
[    1]
[    0]
[    0]
[    0]
[    1]
[    0]
[    1]
[    0]
[    0]
[    0]
[    0]
[    0]
[    0]
[    0]
[    0]]
```

[22]:
```
x.shape
```

[22]: (594, 1)

[23]:
```
#to create the output array for train & test process:
y=np.array(mydataset["Worked for less than 3 months - Persons"])
print(y)
```

```
[12153   356  5714  4757  1320     6 10306   334  4817  3995  1154     6
  1847    22   897   762   166     0  1366    41   636   556   131     2
```

1226	39	571	497	117	2	140	2	65	59	14	0
66	2	37	27	0	0	0	0	0	0	0	0
66	2	37	27	0	0	998	14	477	413	94	0
764	13	354	324	73	0	234	1	123	89	21	0
2346	69	1146	926	203	2	2210	67	1082	876	183	2
136	2	64	50	20	0	1635	57	787	597	194	0
1605	53	773	587	192	0	30	4	14	10	2	0
1186	27	573	455	130	1	1111	27	538	427	118	1
75	0	35	28	12	0	534	17	233	181	103	0
515	17	225	175	98	0	19	0	8	6	5	0
172	15	78	43	36	0	172	15	78	43	36	0
0	0	0	0	0	0	205	5	83	85	32	0
203	3	83	85	32	0	2	2	0	0	0	0
551	11	291	202	46	1	312	11	148	117	35	1
239	0	143	85	11	0	114	3	53	39	19	0
110	3	51	37	19	0	4	0	2	2	0	0
2	0	0	0	2	0	0	0	0	0	0	0
2	0	0	0	2	0	44	0	24	10	10	0
28	0	12	6	10	0	16	0	12	4	0	0
77	1	43	31	2	0	22	1	17	3	1	0
55	0	26	28	1	0	211	7	76	91	37	0
207	7	74	91	35	0	4	0	2	0	2	0
457	13	213	205	26	0	259	13	119	110	17	0
198	0	94	95	9	0	63	3	20	36	4	0
24	2	6	13	3	0	39	1	14	23	1	0
24	0	10	13	1	0	17	0	5	11	1	0
7	0	5	2	0	0	18	0	7	8	3	0
14	0	3	8	3	0	4	0	4	0	0	0
51	6	26	17	2	0	51	6	26	17	2	0
0	0	0	0	0	0	21	1	2	9	9	0
3	1	0	1	1	0	18	0	2	8	8	0
66	0	26	31	9	0	32	0	8	15	9	0
34	0	18	16	0	0	2	0	2	0	0	0
2	0	2	0	0	0	0	0	0	0	0	0
11	0	6	3	2	0	9	0	4	3	2	0
2	0	2	0	0	0	13	0	6	7	0	0
8	0	4	4	0	0	5	0	2	3	0	0
12	2	3	7	0	0	9	2	2	5	0	0
3	0	1	2	0	0	179	3	69	65	42	0
30	1	12	8	9	0	149	2	57	57	33	0
274	4	113	132	25	0	31	0	8	11	12	0
243	4	105	121	13	0	574	30	270	198	76	0
556	30	262	190	74	0	18	0	8	8	2	0
313	14	144	120	35	0	298	14	139	112	33	0
15	0	5	8	2	0	551	11	245	248	47	0
477	9	210	219	39	0	74	2	35	29	8	0
17	0	15	2	0	0	1	0	1	0	0	0
16	0	14	2	0	0]						

[24]:
```
y.shape
```

[24]: (594,)

[25]:
```
print(type(x))
```

```
<class 'numpy.ndarray'>
```

[26]:
```
print(type(y))
```

```
<class 'numpy.ndarray'>
```

[29]:
```
#splitting the dataset into test set & training set:
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.metrics import mean_squared_error, r2_score
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.20)
x_train.shape
```

[29]: (475, 1)

[30]:
```
x_test.shape
```

[30]: (119, 1)

[31]:
```
y.shape
```

[31]: (594,)

[32]:
```
print(y_train.shape)
```

```
(475,)
```

[33]:
```
print(y_test.shape)
```

```
(119,)
```

[35]:
```
#spliting & training dataset with mean squared error:
# Initialize the model
model = LinearRegression()

# Train the model
model.fit(x_train, y_train)

# Make predictions on the test set
y_pred = model.predict(x_test)

# Evaluate the model
```

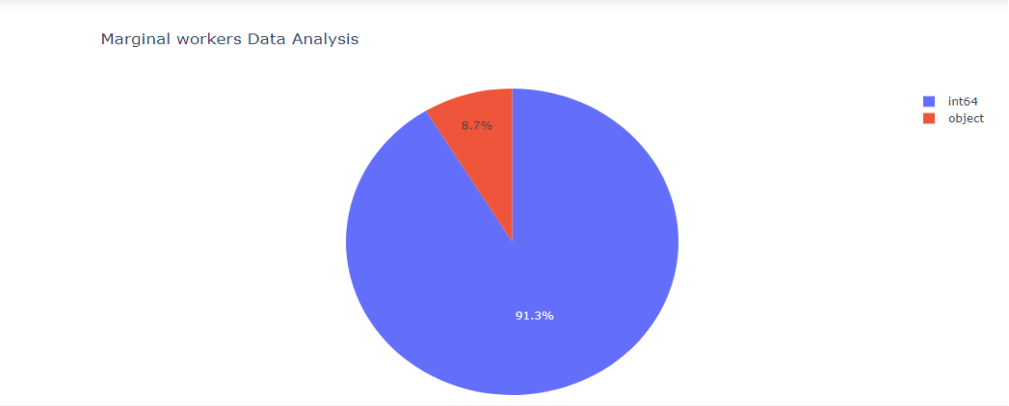
```
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)

print("Mean Squared Error:", mse)
print("R-squared:", r2)
```

```
Mean Squared Error: 796.7282889466414
R-squared: 0.9919039963931663
```

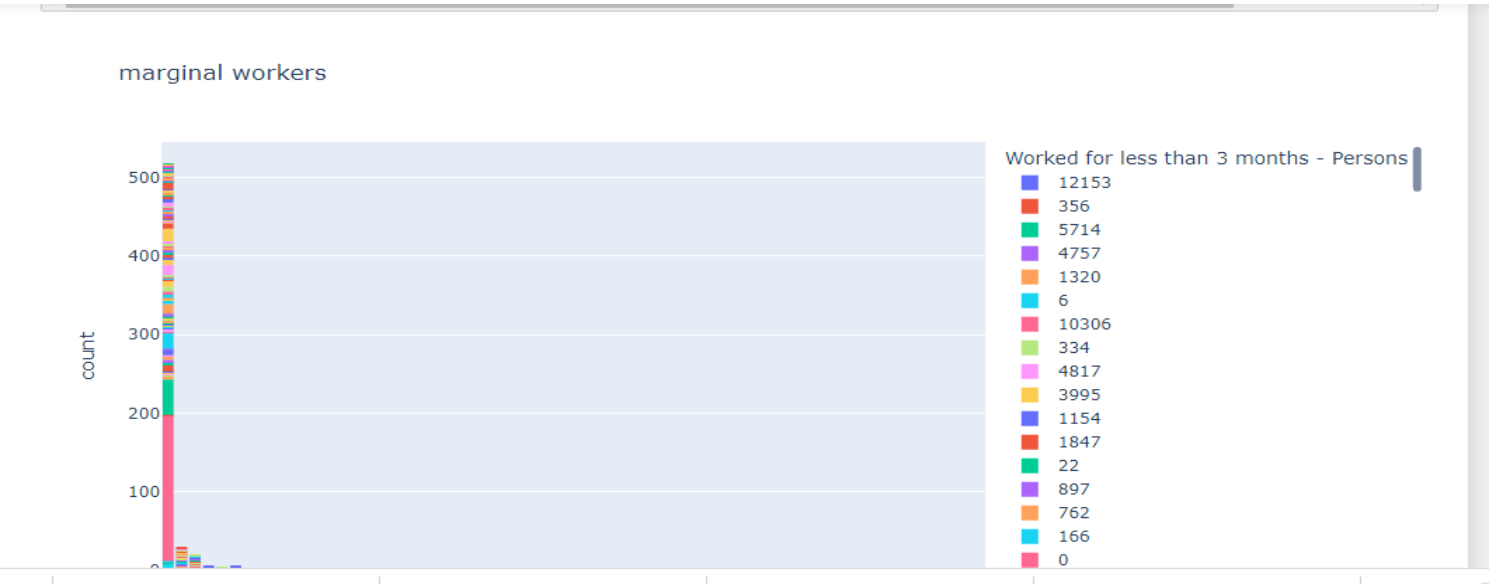
[49]:
```
#pictorical representation for the dataset by using  pie chart:
import plotly.graph_objects as go
data_types_counts = mydataset.dtypes.astype(str).value_counts()
fig = go.Figure(data=[go.Pie(labels=data_types_counts.index,
 ↪values=data_types_counts.values)])
fig.update_layout(title_text=" Marginal workers Data Analysis")
fig.show()
```

Output



[43]:
```
#pictorical representation for the dataset by using  bar chart:
import plotly.express as px
fig = px.histogram(mydataset,x="Worked for less than 3 months - Persons",
 ↪title="marginal workers", color="Worked for less than 3 months - Persons")
# Update the layout and add box plots
fig.update_layout
    (bargap=0.2)

fig.show()
```


marginal workers
count
500
400
300
200
100
Worked for less than 3 months - Persons
12153
356
5714
4757
1320
6
10306
334
4817
3995
1154
1847
22
897
762
166
0