
DATA MINING

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CLUSTERING

ANALYSIS OF DIGITAL ADS DATA

CLUSTERING :

It is an unsupervised learning Technique trying to identify groups of similar objects that are highly dissimilar with other objects.

OVERVIEW OF THE REPORT :

To analyse the digital ad data and to identify trends in clicks, spend, revenue, CPM, CTR, & CPC based on Device Type.

1) DATA ANALYSIS :

Read the data and perform basic analysis such as printing a few rows (head and tail), info, data summary, null values duplicate values, etc.

➤ The dataset contains 23099 Rows and 19 Columns.

➤ **DUPLICATES :**

No duplicates are present.

➤ **NULL VALUES :**

There are

4736 Null values in CPM

4736 Null values in CTR

4736 Null values in CPC

- There are 13 Numerical variables and 6 Categorical variables.

2) Treat missing values in CPC, CTR and CPM using the formula given :

FORMULAS GIVEN :

$$\text{CTR} = \text{Clicks/Impressions} * 100$$

$$\text{CPM} = \text{Spend/Impressions} * 1000$$

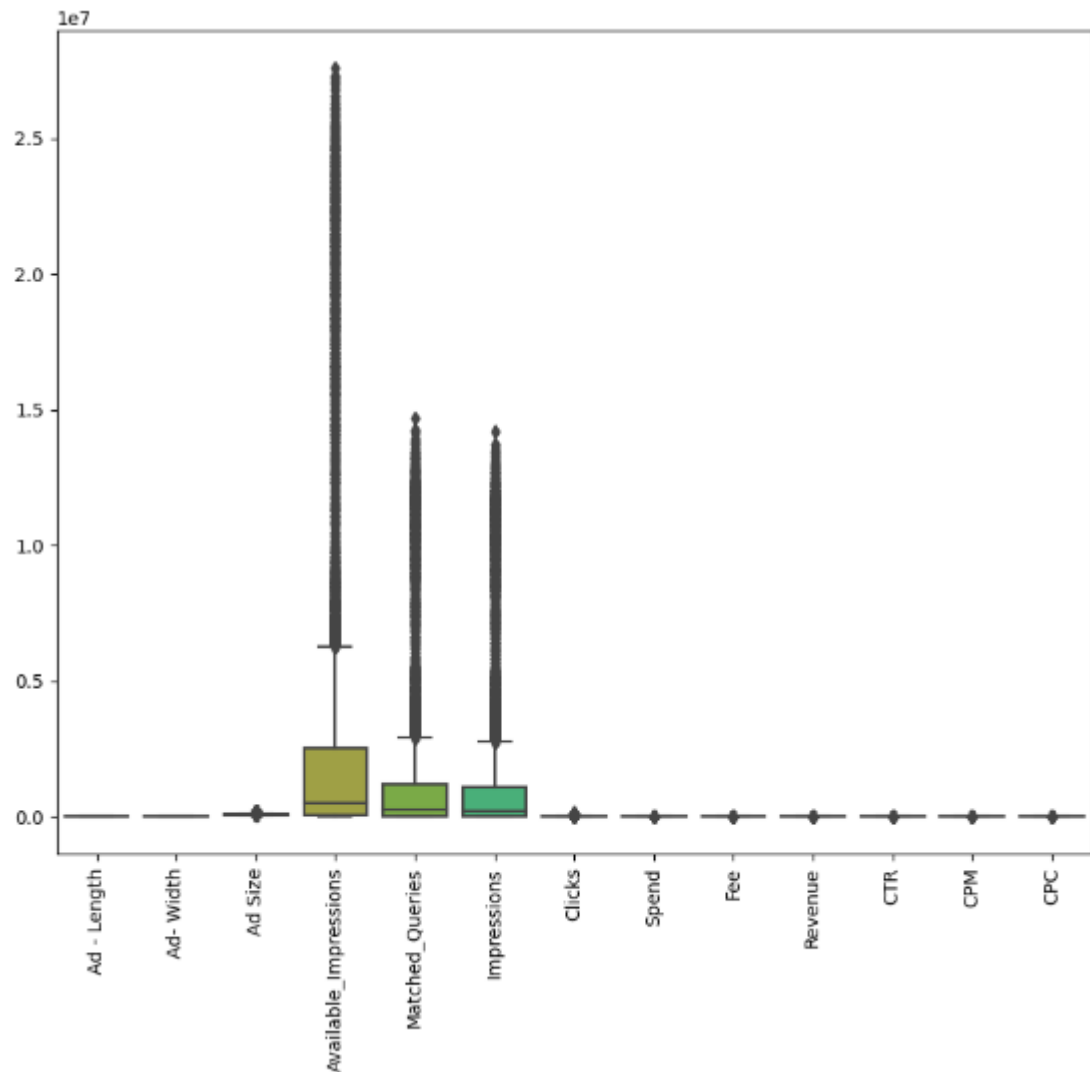
$$\text{CPC} = \text{Spend/Clicks}$$

The Missing values in dataset has to be filled using the above formula.

3) Check if there are any outliers. Do you think treating outliers is necessary for K-Means clustering? Based on your judgement decide whether to treat outliers and if yes, which method to employ. (As an

analyst your judgement may be different from another analyst).

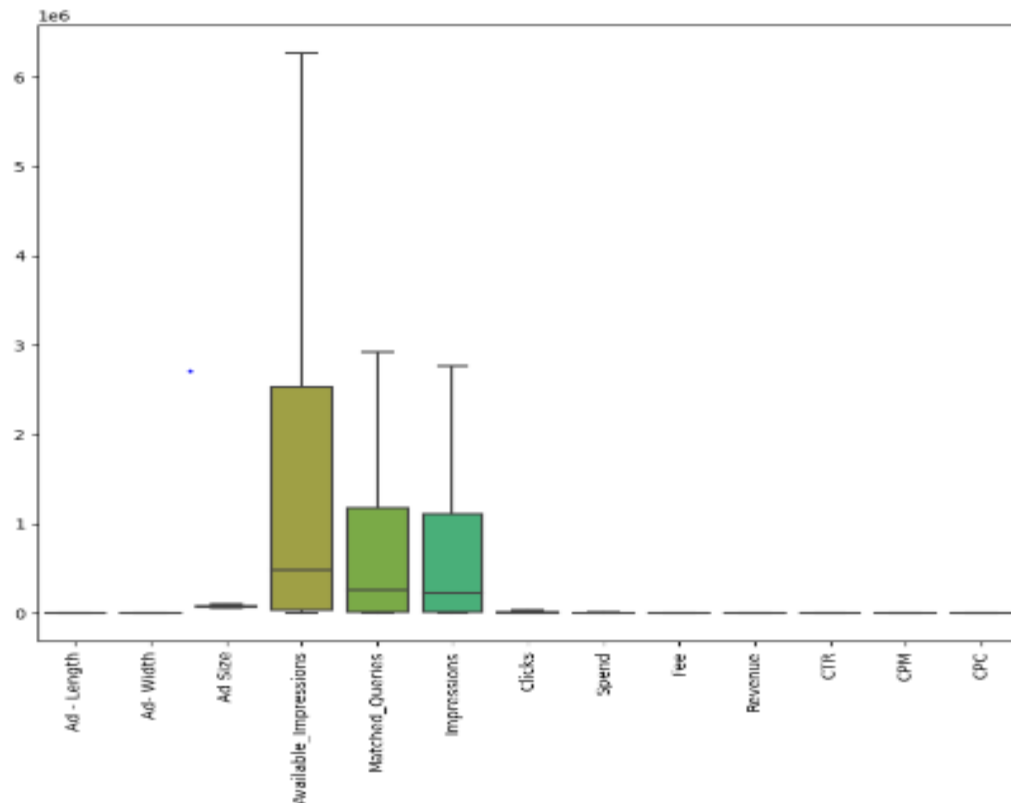
Outliers are present in following features.



- Outlier detection is an important data analysis task
- Treating the outliers from clusters can improve the clustering accuracy. So we have to treat outlier before applying k means algorithm.

- Here we have to use **BOXPLOT(QUARTILE) Method** to treat outliers.

BOXPLOT after treating outliers :



Thus using BOXPLOT Method we can treat outliers to get high accuracy while performing K-Clustering Algorithm.

4) Perform z-score scaling and discuss how it affects the speed of the algorithm.

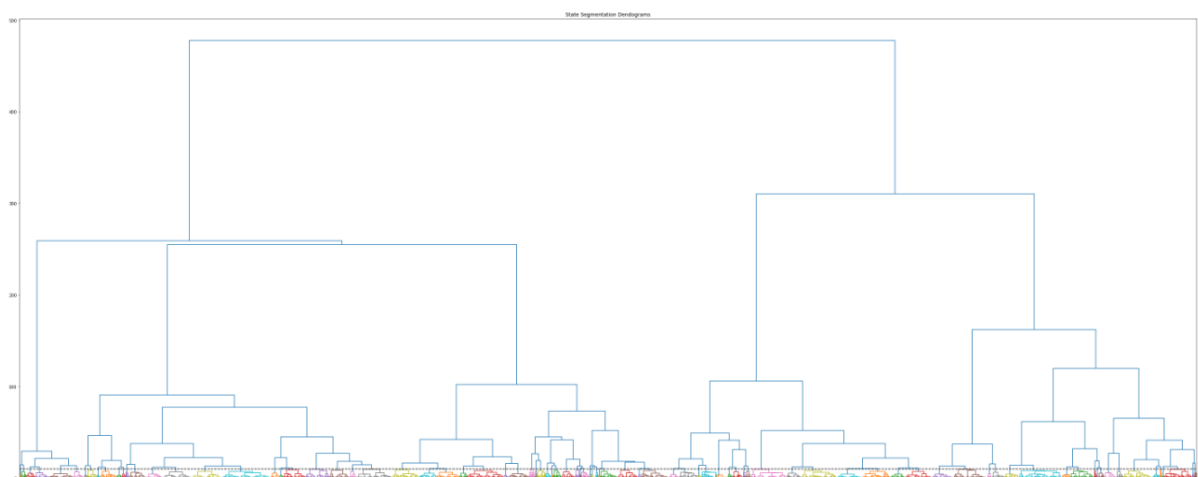
	Ad - Length	Ad - Width	Ad Size	Available_Impressions	Matched_Queries	Impressions	Clicks	Spend	Fee	Revenue	CTR	CPM
0	-0.364496	-0.432797	-0.102518	-0.755333	-0.778949	-0.768478	-0.867488	-0.893170	0.535724	-0.880093	-0.958836	-1.194496
1	-0.364496	-0.432797	-0.102518	-0.755345	-0.778988	-0.768516	-0.867488	-0.893170	0.535724	-0.880093	-0.953835	-1.194496
2	-0.364496	-0.432797	-0.102518	-0.754900	-0.778919	-0.768445	-0.867488	-0.893170	0.535724	-0.880093	-0.962218	-1.194496
3	-0.364496	-0.432797	-0.102518	-0.755040	-0.778781	-0.768302	-0.867488	-0.893170	0.535724	-0.880093	-0.971871	-1.194496
4	-0.364496	-0.432797	-0.102518	-0.755610	-0.779030	-0.768560	-0.867488	-0.893170	0.535724	-0.880093	-0.946281	-1.194496
...
23061	1.433093	-0.186599	1.652896	-0.756182	-0.779265	-0.768806	-0.867488	-0.893141	0.535724	-0.880066	3.035808	3.162718
23062	1.433093	-0.186599	1.652896	-0.756181	-0.779264	-0.768805	-0.867488	-0.893154	0.535724	-0.880078	3.035808	1.712113
23063	1.433093	-0.186599	1.652896	-0.756182	-0.779265	-0.768806	-0.867488	-0.893150	0.535724	-0.880074	3.035808	3.162718
23064	-1.134891	1.290590	-0.297564	-0.756179	-0.779265	-0.768806	-0.867488	-0.893141	0.535724	-0.880066	3.035808	3.162718
23065	1.433093	-0.186599	1.652896	-0.756182	-0.779264	-0.768805	-0.867488	-0.893133	0.535724	-0.880058	3.035808	3.162718

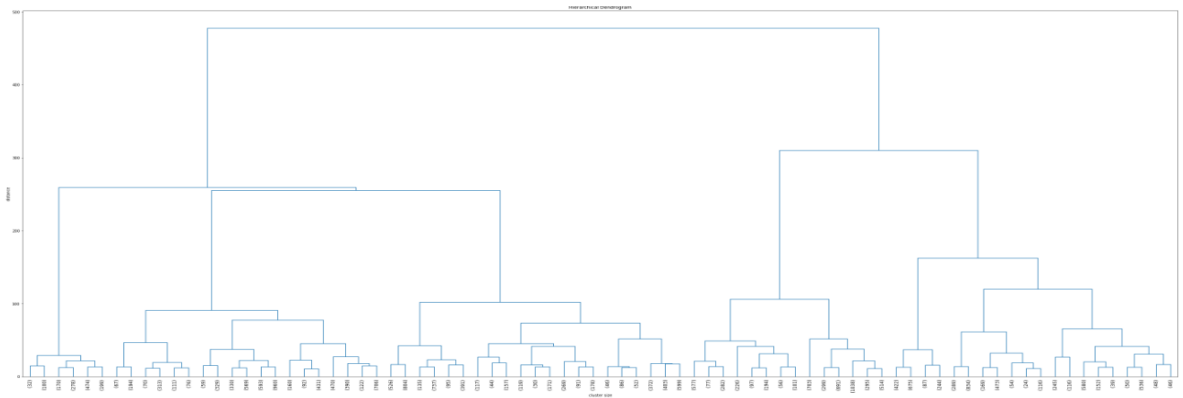
23066 rows x 13 columns

- Z-score is applied to the sorted data points as a measure to improve the selection of initial clusters.
- Because of finding initial cluster centers , it ensures
 - High accuracy
 - Reduced clustering error
 - Less computation time and
 - Less number of iterations.

5) Perform Hierarchical by constructing a Dendrogram using WARD and Euclidean distance.

Dendrogram :

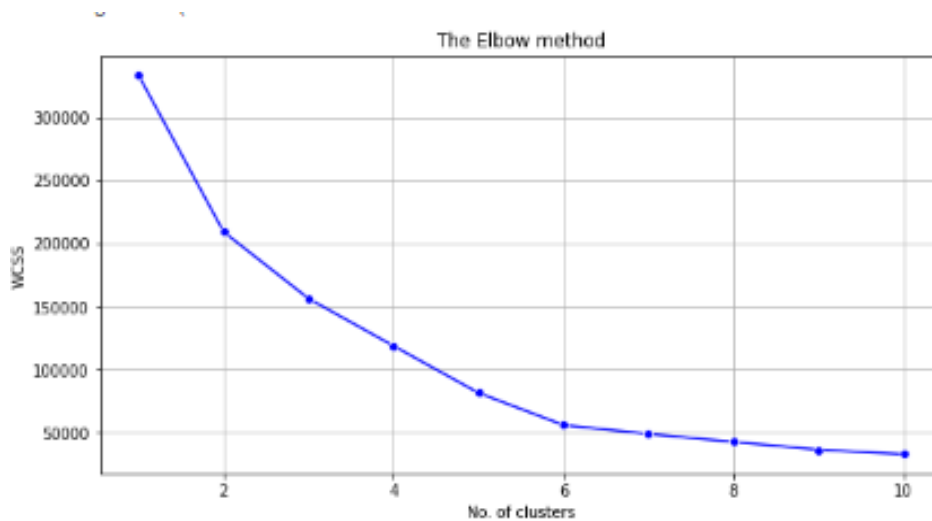




From Hierarchical clustering , we can say that the **optimum number of clusters will be 5.**

6) Clustering: Make Elbow plot (up to n=10) and identify optimum number of clusters for k-means algorithm.

Elbow plot :



FROM THE ABOVE WSS PLOT WE CAN SAY THAT THE **OPTIMUM NUMBER OF CLUSTERS ARE 5** BECAUSE AFTER 5 THE PLOT WILL BE NARROW i.e) NO MUCH DIFFERENCE EXISTS.

7) Print silhouette scores for up to 10 clusters and identify optimum number of clusters.

Silhouette scores

No. of clusters	Silhouette scores
2	0.4241248441233028
3	0.4284360206300488
4	0.49517949988734966
5	0.5947980661900538
6	0.5999254403934559
7	0.6363174847771058
8	0.6288118527264385

9	0.6481558775415522
10	0.6542947087856206

So we can say that the optimum number of clusters will be 5.

8) Clustering: Profile the ads based on optimum number of clusters using silhouette score and your domain understanding [Hint: Group the data by clusters and take sum or mean to identify trends in Clicks, spend, revenue, CPM, CTR, & CPC based on Device Type].

	Ad - Length	Ad- Width	Ad Size	Available_Impressions	Matched_Queries	Impressions	Clicks	Spend	Fee	Revenue	CTR	CPM	CPC	freq
Clus_kmeans5														
0	698.358335	300.734476	101182.728689	2.511686e+05	1.372517e+05	1.164528e+05	14152.788223	1248.942898	0.349542	813.358390	13.328432	11.730688	0.089450	4493
1	482.283056	201.204062	73102.572534	5.652157e+06	2.787940e+06	2.655004e+06	11120.125242	6688.873204	0.313885	3844.073325	0.218220	1.576936	0.747079	4136
2	423.343534	150.447737	64158.488069	1.779724e+06	8.463462e+05	8.083297e+05	3237.675310	1470.694323	0.349472	957.636822	0.420293	1.801644	0.524839	6209
3	147.633044	567.524693	74378.180188	4.114756e+04	2.490452e+04	1.841548e+04	2203.010328	239.357374	0.350000	155.582225	15.716555	14.269791	0.103078	6682
4	141.981889	571.927555	73738.680468	8.032473e+05	5.647976e+05	4.764054e+05	30530.887209	6526.995595	0.305712	4458.294194	13.752640	15.406399	0.112074	1546

TRENDS :

➤ **CLUSTER_0 :**

- Larger Ad size
- More number of clicks
- Lesser amount spending
- Getting low revenue
- Medium amount of CTR , CPM
- Lower CPC

➤ **CLUSTER_1:**

- Smaller Ad size
- Medium number of clicks
- Larger amount spending
- Getting Higher Revenue
- Very Low CTR,CPM
- Higher CPC

➤ **CLUSTER_2 :**

- Very smaller Ad size
- Lesser number of clicks
- Average amount spending
- Average Revenue
- Low CTR,CPM
- Higher CPC

➤ **CLUSTER_3 :**

- Larger Ad size
- Lesser number of clicks
- Lesser amount spending
- Getting Very low revenue
- Higher CTR,CPM
- Lower CPC

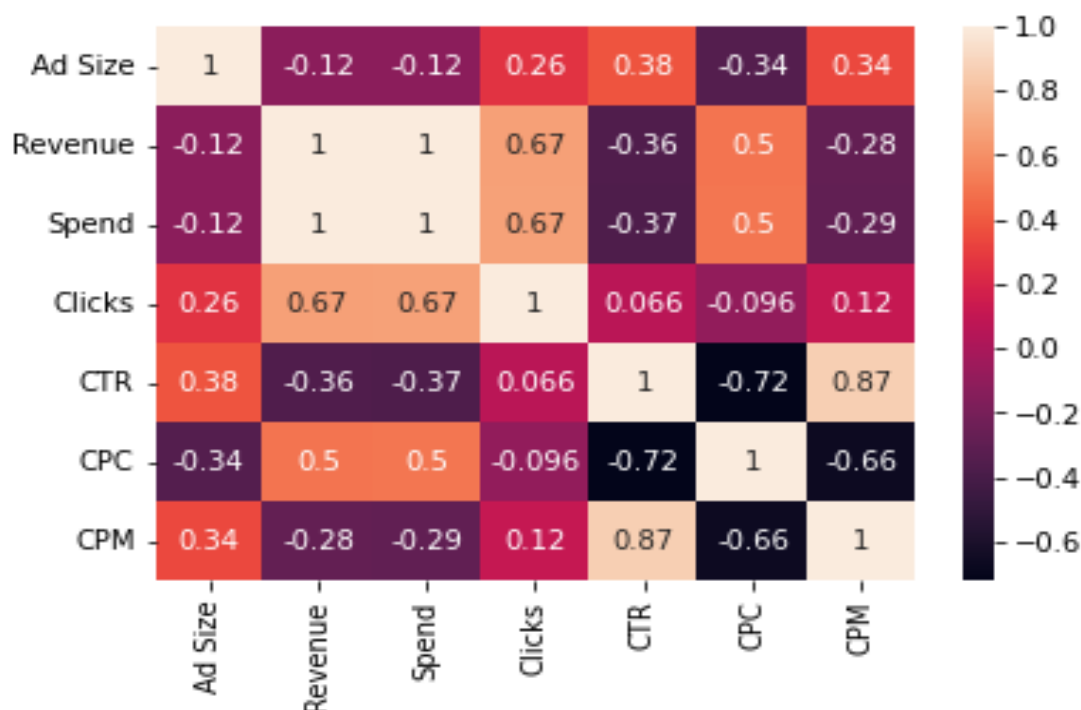
➤ **CLUSTER_4 :**

- Medium Ad size
- Larger number of clicks
- Large amount spending
- Higher Revenue
- Higher CTR,CPM
- Average CPC

9) Conclude the project by providing summary of your learnings:

- ❖ Both Desktop and Mobile (Device type) uses Video and Web as platform
- ❖ App can be uses as a platform only through Mobile
- ❖ When Ad size is small , it yields Higher Revenue.
- ❖ When number of clicks is high , Revenue will be high

- ❖ When spending is more , Revenue will be more
- ❖ When CPC high , Revenue also high
- ❖ When number of clicks high , CPC is low
- ❖ When spending increases , both CPM AND CPC also increases
- ❖ When number of clicks increases , CTR also increases.
- ❖ When Ad size increases , Clicks,CPC ,CPM & CTR also increases.
- ❖ Positive correlation exists between Revenue,Spend,CPC,Clicks.



PRINCIPAL COMPONENT ANALYSIS

POPULATION CENSUS ANALYSIS 2011

PCA :

Principal component analysis (PCA) is a popular unsupervised learning technique for analyzing large datasets containing a high number of dimensions/features per observation, increasing the interpretability of data while preserving the maximum amount of information, and enabling the visualization of multidimensional data. Formally, PCA is a statistical technique for reducing the dimensionality of a dataset.

OBJECTIVE OF PCA :

- DIMENSION REDUCTION
- PATTERN RECOGNITION
- RESOLVE MULTI-COLLINEARITY

OVERVIEW OF THE REPORT :

To analyse the population census data with respect to the features provided such as Literacy Rate , Labour Force , Gender Ratio , Population of scheduled peoples etc. by applying PCA Technique .

1) DATA ANALYSIS :

PCA: Read the data and perform basic checks like checking head, info, summary, nulls, and duplicates, etc.

➤ The dataset contains 641 Rows & 61 Columns .

➤ DUPLICATES :

NO duplicates present.

➤ NULL VALUES :

Absence of Null values in the Data set.

➤ SUMMARY :

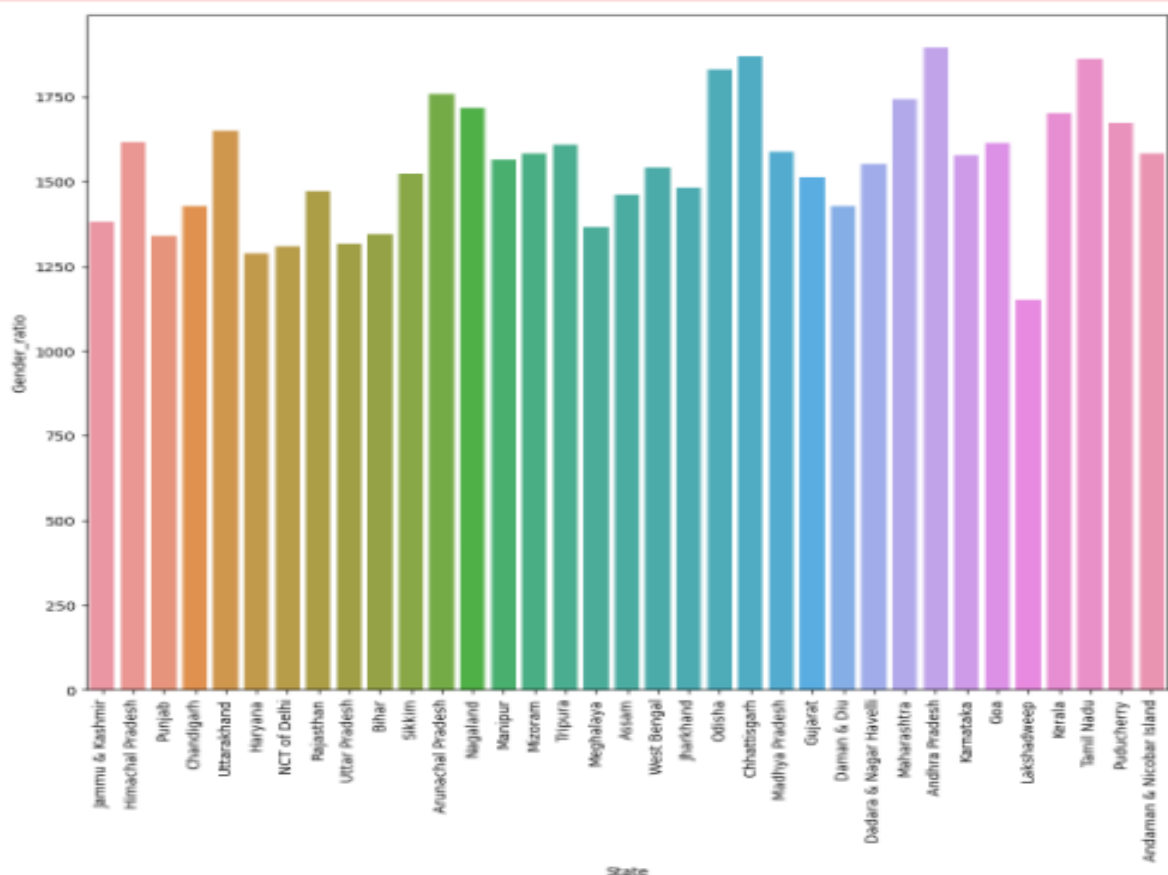
Presence of 59 Numerical Variables and 2 Categorical Variables.

2) EDA :

- i) Which state / UT has highest gender ratio and which has the lowest?

GENDER RATIO defined as number of females per 1000 males in a population

$$\text{GENDER RATIO} = \text{NO. OF FEMALES} / \text{NO. OF MALES} * 1000$$



STATE	GENDER RATIO
ANDRA PRADESH (HIGHEST)	1895 OUT OF 1000 MALES
LAKSHWADEEP (LOWEST)	1152 OUT OF 1000 MALES

ii) Which district has the highest & lowest gender ratio?

```

State      Area Name      Gender_Ratio
Andhra Pradesh  Krishna      2283.249638
Odisha        Koraput      2268.763478
Tamil Nadu    Virudhunagar 2225.428760
Andhra Pradesh West Godavari 2221.848576
Odisha        Baudh        2215.059963
...
Uttar Pradesh Baghpat      1184.830405
Rajasthan     Dhaulpur     1180.761033
Uttar Pradesh Mahamaya Nagar 1180.201612
Jammu & Kashmir Badgam       1179.576206
Lakshadweep   Lakshadweep  1151.992513
Name: Gender_ratio, Length: 640, dtype: float64

```

- **KRISHNA District in ANDRA PRADESH** has highest Gender Ratio .
- **LAKSHADWEEP District in LAKSHADWEEP Island** has lowest Gender Ratio.

iii) Which state / UT has highest & lowest literacy rate ?

FORMULA :

$$\text{LITERACY RATE \%} = (\text{TOTAL MALE LITERATES} + \text{TOTAL FEMALE LITERATES}) / \text{TOTAL POPULATION}$$

State	
Kerala	80.590272
Lakshadweep	79.489038
Mizoram	78.903429
Goa	76.936993
Chandigarh	75.929268
Tripura	74.631148
Puducherry	73.902195
NCT of Delhi	73.816645
Daman & Diu	73.507369
Andaman & Nicobar Island	72.202044
Himachal Pradesh	69.343645
Sikkim	69.279694
Maharashtra	67.432964
Nagaland	66.800108
Uttarakhand	65.660454
Punjab	65.140981
Manipur	64.483249
Tamil Nadu	64.478225
Haryana	63.303344
Gujarat	63.264212
West Bengal	62.439949
Meghalaya	61.256250
Assam	61.237314
Karnataka	60.609735
Dadara & Nagar Haveli	58.535901
Odisha	57.148650
Madhya Pradesh	55.193915
Arunachal Pradesh	55.087266
Uttar Pradesh	54.202634
Andhra Pradesh	53.497705
Jammu & Kashmir	53.034406
Rajasthan	52.892748
Chhattisgarh	52.320579
Jharkhand	51.539596
Bihar	47.988240

Name: literacy_rate%, dtype: float64

- **KERALA** has highest literacy rate of 80.59 %
- **BIHAR** has lowest literacy rate of 47.98%

iv) Which state / UT has highest & lowest female literacy rate ?

FORMULA:

$$\text{FEMALE LITERACY RATE} = \frac{\text{TOTAL FEMALE LITERATES}}{\text{TOTAL FEMALE POPULATION}} * 100$$

State	
Kerala	79.879281
Mizoram	78.705486
Lakshadweep	76.726239
Goa	72.952864
Chandigarh	72.828784
Tripura	70.772460
NCT of Delhi	69.396511
Puducherry	69.057797
Daman & Diu	66.967192
Andaman & Nicobar Island	66.919376
Sikkim	63.418179
Himachal Pradesh	63.115824
Nagaland	62.903650
Maharashtra	61.396842
Meghalaya	60.038426
Uttarakhand	59.643496
Punjab	59.043608
Manipur	58.098332
Tamil Nadu	55.860940
Gujarat	55.581919
West Bengal	55.192171
Assam	54.560051
Haryana	54.534237
Karnataka	51.951959
Arunachal Pradesh	49.042441
Dadara & Nagar Haveli	49.007479
Odisha	48.747607
Madhya Pradesh	46.256606
Uttar Pradesh	45.272077
Jammu & Kashmir	44.797093
Andhra Pradesh	43.223754
Chhattisgarh	43.030413
Jharkhand	42.286003
Rajasthan	41.716589
Bihar	39.752975

Name: Female_literacy_rate, dtype: float64

Among the given states, 79.87 % of female population in **KERALA** are literates whereas only 39.75 % of female population are literates in **BIHAR**.

V) Which state / UT has more scheduled caste population ?

FORMULA :

TOTAL SC POPULATION = TOTAL SC MALES + TOTAL SC FEMALES IN EACH STATE

State	
Uttar Pradesh	4285345
West Bengal	2175971
Tamil Nadu	1811842
Andhra Pradesh	1735314
Karnataka	1678375
Maharashtra	1621300
Bihar	1381929
Punjab	1109952
Rajasthan	954104
Madhya Pradesh	891745
Kerala	709005
Odisha	679142
Haryana	541176
NCT of Delhi	402333
Gujarat	392873
Jharkhand	364310
Chhattisgarh	325179
Himachal Pradesh	277937
Assam	274015
Uttarakhand	263435
Jammu & Kashmir	104653
Tripura	79735
Puducherry	34630
Chandigarh	21654
Manipur	14097
Goa	5857
Sikkim	3370
Meghalaya	2628
Daman & Diu	1231
Dadra & Nagar Haveli	405
Mizoram	35
Andaman & Nicobar Island	0
Nagaland	0
Arunachal Pradesh	0
Lakshadweep	0

Name: Total_sc_pop, dtype: int64

UTTAR PRADESH has more Scheduled caste population.

vi) Which state /UT has more scheduled tribes population ?

FORMULA :

TOTAL ST POPULATION = TOTAL ST MALES + TOTAL ST FEMALES IN EACH STATE

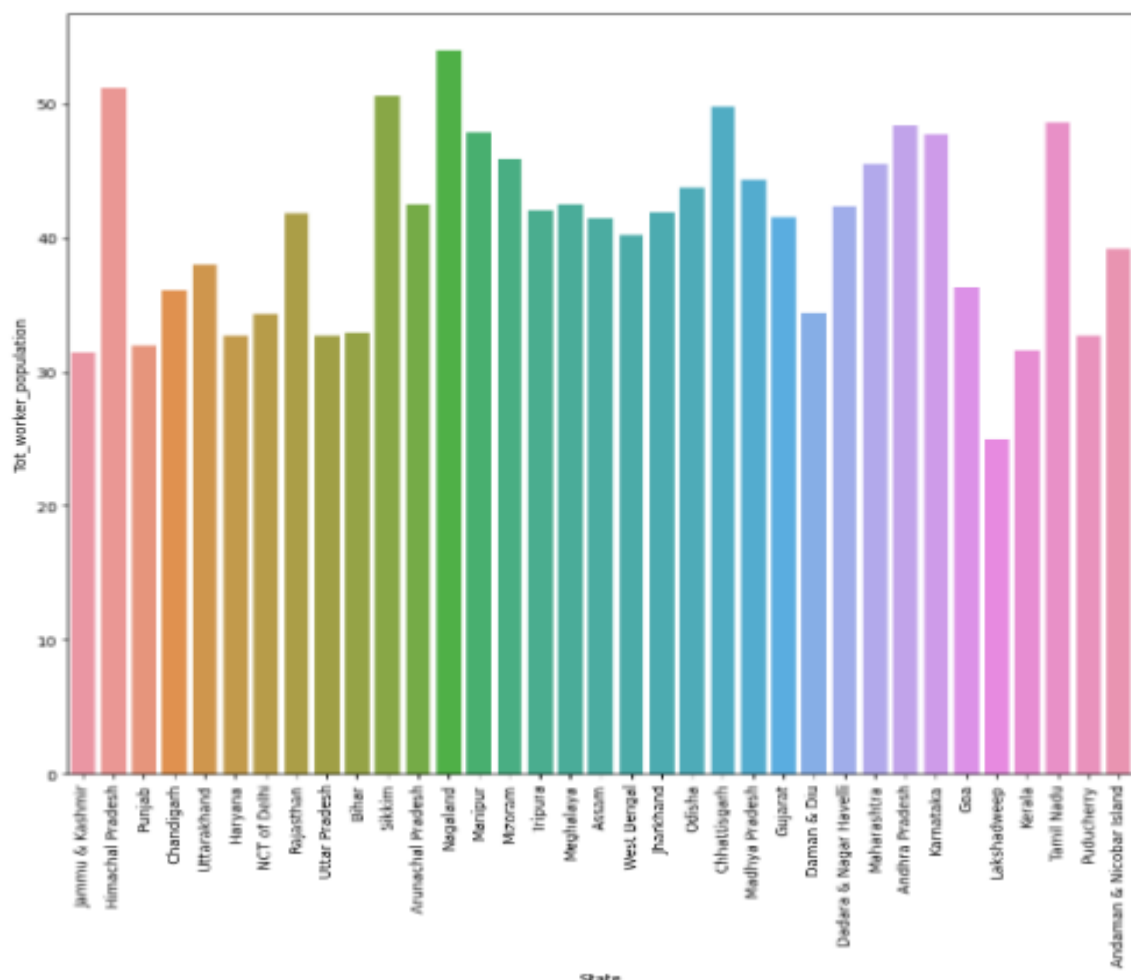
```
State
Madhya Pradesh      1147620
Jharkhand           965988
Maharashtra         956627
Odisha              926650
Gujarat             809454
Chhattisgarh        752040
West Bengal         741507
Karnataka            648253
Andhra Pradesh      614002
Meghalaya           567199
Rajasthan            540420
Assam               384060
Nagaland            179695
Mizoram             150142
Bihar               128241
Manipur             126772
Uttar Pradesh       115649
Jammu & Kashmir      108956
Arunachal Pradesh   107771
Tripura             101494
Kerala              99071
Tamil Nadu          79542
Himachal Pradesh    64362
Goa                 38009
Uttarakhand         35295
Lakshadweep         27244
Sikkim              26464
Dadara & Nagar Haveli 13291
Daman & Diu          3284
Andaman & Nicobar Island 3265
Puducherry          0
Punjab              0
NCT of Delhi        0
Haryana             0
Chandigarh          0
Name: Total_st_pop, dtype: int64
```

MADHYA PRADESH has more Scheduled Tribes population.

vii) Which state / UT has highest and lowest work force ?

FORMULA :

TOTAL WORKER POPULATION = (TOTAL MALE WORKER + TOTAL FEMALE WORKER) / TOTAL POPULATION OF EACH STATE



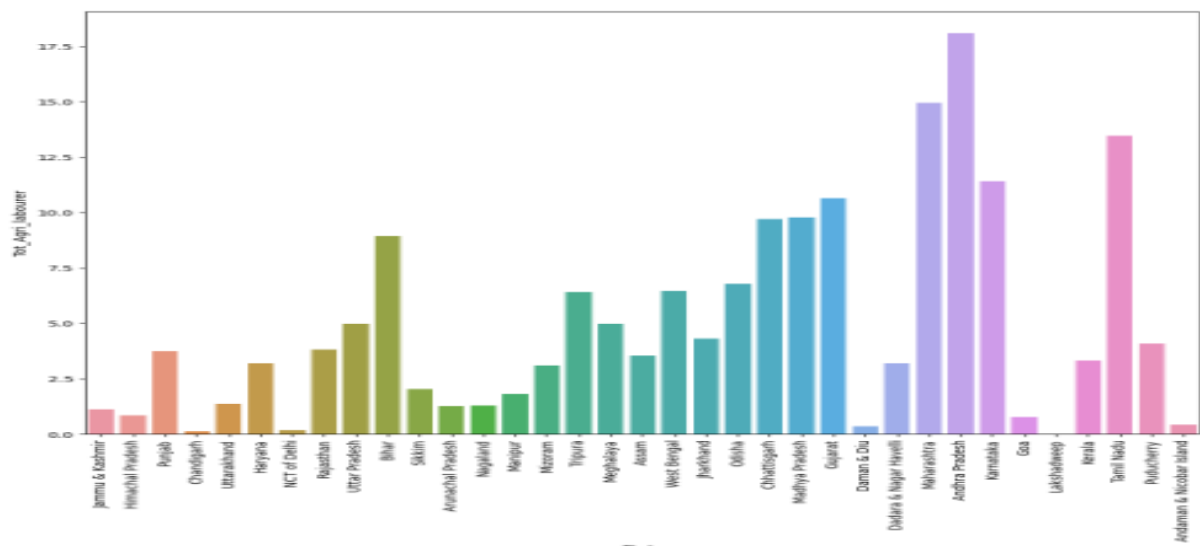
NAGALAND has 53.93 % worker population whereas as **LAKSHADWEEP** has only 24.98 % worker population with respect to their Total Population.

viii) Which state / UT has more agricultural labourers in India?

FORMULA :

$$\text{TOTAL AGRI LABOURERS (\%)} = \left(\frac{\text{TOTAL MALE + TOTAL FEMALE AGRI LABOURERS}}{\text{TOTAL POPULATION OF EACH STATE}} \right) * 100$$

State	
Andhra Pradesh	18.092598
Maharashtra	14.952152
Tamil Nadu	13.462188
Karnataka	11.416412
Gujarat	10.661657
Madhya Pradesh	9.783616
Chhattisgarh	9.712727
Bihar	8.945050
Odisha	6.782081
West Bengal	6.477672
Tripura	6.418161
Uttar Pradesh	4.993301
Meghalaya	4.984457
Jharkhand	4.314657
Puducherry	4.093269
Rajasthan	3.817849
- - - - -	- - - - -



ANDRA PRADESH has 18.09 % Agricultural population who predominantly depends upon Agriculture for their livelihood whereas **LAKSHADWEEP** has 0 % Agricultural population.

ix) Which state/ UT has highest and lowest population ?

FORMULA :

**TOTAL POPULATION OF A STATE = TOTAL MALES +
TOTAL FEMALES IN RESPECTIVE STATE**

State	
Uttar Pradesh	21067854
Maharashtra	11334687
West Bengal	9928671
Bihar	9431081
Andhra Pradesh	9371598
Karnataka	8755157
Tamil Nadu	8684319
Kerala	7776182
Madhya Pradesh	5525353
Rajasthan	5029059
Gujarat	4923157
Odisha	3997011
Punjab	3700830
Assam	3530700
Jharkhand	2966507
Haryana	2666689
Chhattisgarh	2364996
NCT of Delhi	1908680
Uttarakhand	1587071
Himachal Pradesh	1235443
Jammu & Kashmir	994172
Meghalaya	624391
Tripura	416827
Manipur	372487
Goa	310372
Nagaland	199441
Puducherry	189460
Mizoram	154997
Arunachal Pradesh	138648
Chandigarh	101397
Sikkim	68182
Andaman & Nicobar Island	47417
Daman & Diu	31859
Lakshadweep	27595
Dadara & Nagar Haveli	17813

Name: Tot_population, dtype: int64

- **UTTAR PRADESH** has highest population
- **DADRA & NAGAR HAVELLI** has lowest population in India

3) PCA: We choose not to treat outliers for this case. Do you think that treating outliers for this case is necessary?

Because of population data , no need to treat outliers. Some outliers represent natural variations in the population, and they should be left as it is in the dataset.

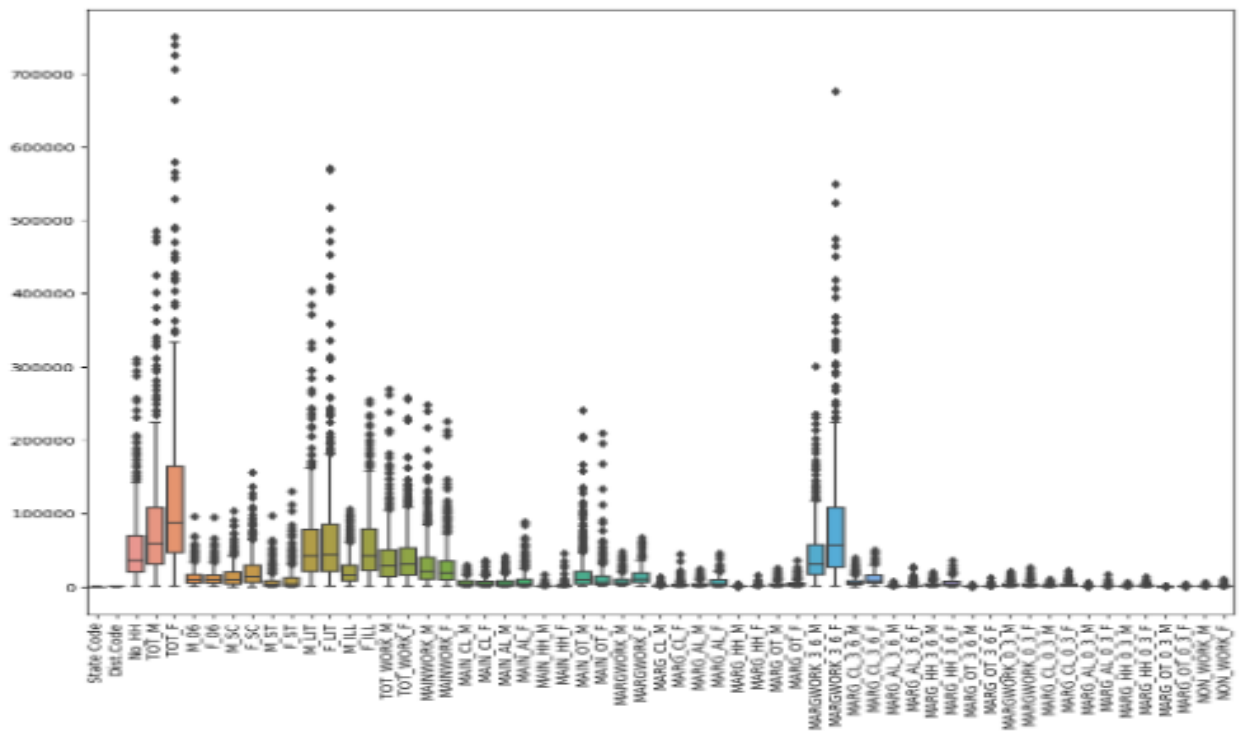
4) PCA: Scale the Data using z-score method. Does scaling have any impact on outliers? Compare boxplots before and after scaling and comment.

z-score scaling :

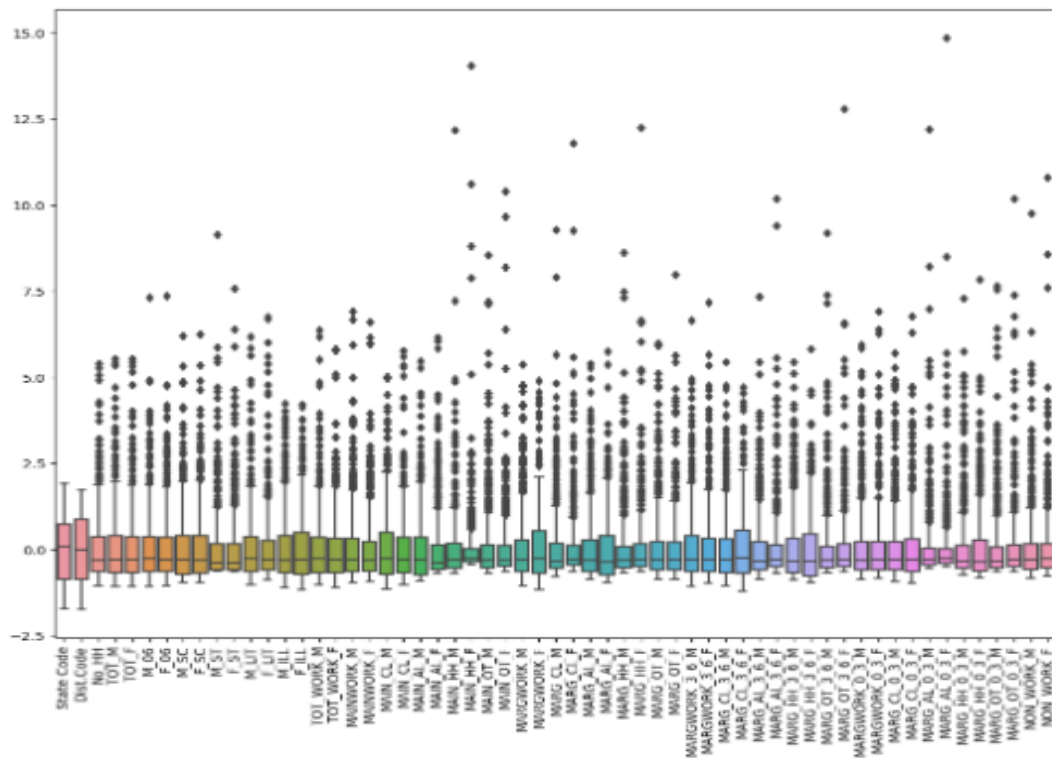
	State Code	Dist.Code	No_HH	TOT_M	TOT_F	M_06	F_06	M_SC	F_SC	M_ST	...	M_F
count	6.400000e+02	640.000000	6.400000e+02	6.400000e+02	6.400000e+02	6.400000e+02	6.400000e+02	6.400000e+02	6.400000e+02	6.400000e+02
mean	8.881784e-17	0.000000	4.440892e-17	-8.881784e-17	-4.440892e-17	-5.551115e-17	6.661338e-17	5.551115e-18	-5.551115e-17	-4.440892e-17
std	1.000782e+00	1.000782	1.000782e+00	1.000782e+00	1.000782e+00	1.000782e+00	1.000782e+00	1.000782e+00	1.000782e+00	1.000782e+00
min	-1.710782e+00	-1.729347	-1.057897e+00	-1.084858e+00	-1.071906e+00	-1.066238e+00	-1.050264e+00	-9.587827e-01	-9.570486e-01	-6.251244e-01
25%	-8.814480e-01	-0.864673	-8.598822e-01	-6.779559e-01	-6.682499e-01	-6.591892e-01	-6.423757e-01	-7.183230e-01	-6.989640e-01	-5.954674e-01
50%	9.405736e-02	0.000000	-3.198873e-01	-2.945918e-01	-3.052330e-01	-2.741142e-01	-2.897563e-01	-2.934040e-01	-3.256148e-01	-3.895344e-01
75%	7.310596e-01	0.864673	3.673585e-01	3.815493e-01	3.889451e-01	3.864446e-01	3.488980e-01	3.890923e-01	3.869764e-01	1.480266e-01
max	1.898897e+00	1.729347	5.389586e+00	5.529890e+00	5.532633e+00	7.301993e+00	7.350309e+00	6.207800e+00	6.248040e+00	9.146281e+00

8 rows x 59 columns

BOXPLOT BEFORE SCALING :



BOXPLOT AFTER SCALING :



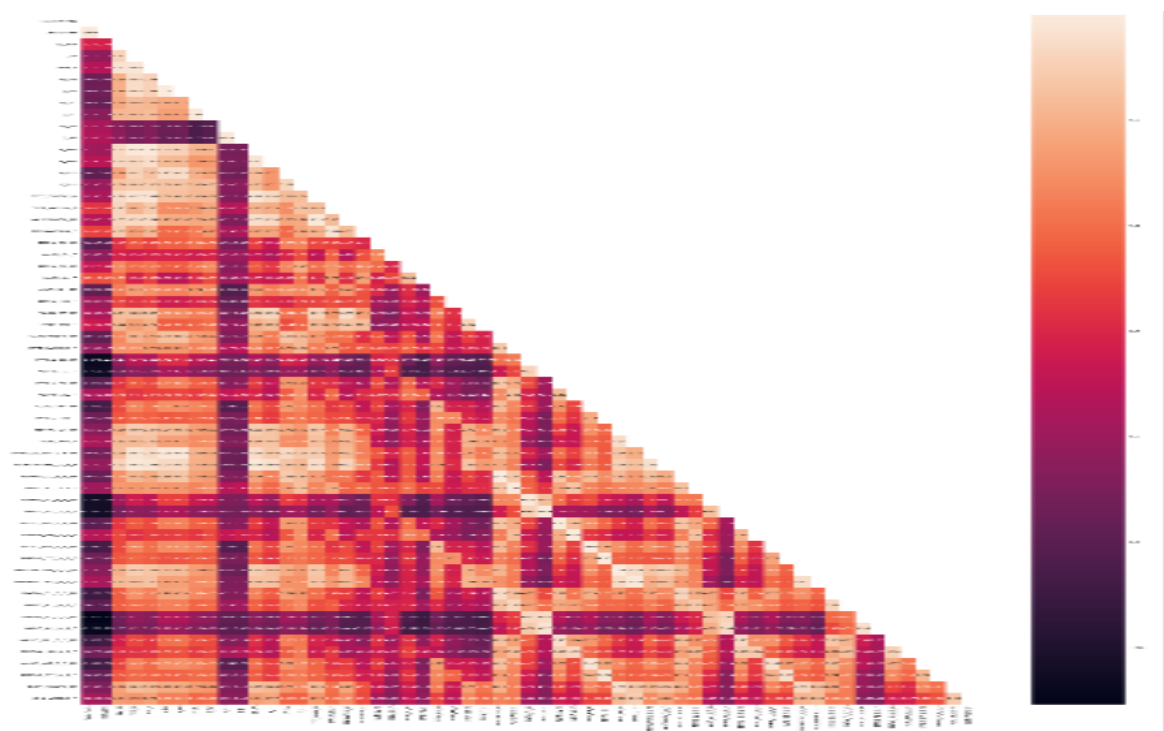
OBSERVATION :

- Scaling shrinks the range of values as shown in the figure while keeping the outliers in.
- However, the outliers have an influence only when computing the empirical mean and standard deviation.
- To compare the boxplot before and after scaling, the only difference is the distances between marginal outliers and inliers are shrunk.

5) PCA: Perform all the required steps for PCA (use sklearn only) Create the covariance Matrix Get eigen values and eigen vector.

STEP 1 : TO FIND CORRELATION

HEATMAP:



Covariance Matrix :

```
covariance matrix /n%s [[1.00156495 0.99457535 0.38502614 ... 0.03409773 0.12572474 0.23208471]
[0.99457535 1.00156495 0.37756089 ... 0.03334295 0.11226784 0.21313518]
[0.38502614 0.37756089 1.00156495 ... 0.53769433 0.76357722 0.73684378]
...
[0.03409773 0.03334295 0.53769433 ... 1.00156495 0.61052325 0.52191235]
[0.12572474 0.11226784 0.76357722 ... 0.61052325 1.00156495 0.88228018]
[0.23208471 0.21313518 0.73684378 ... 0.52191235 0.88228018 1.00156495]]
```

Calculate BARTLETT SPHERICITY :

To confirm the statistical significance of correlation :

H0: All variables in the data are uncorrelated

Ha: At least one pair of variables in the data are correlated

Hence $p_value = 0$,

we can reject H0 because $p_value < 0.05$ and can say that correlations are significant.

Calculate kmo :

To confirm adequacy of sample size :

kmo value is 0.80 which is above 0.7 , so we can say that there are adequate sample size to perform PCA.

STEP 2 : APPLY PCA TECHNIQUE

EIGEN VECTORS :

```
array([[ 3.00700521e-02,  3.00751392e-02,  1.56432451e-01, ...,  
        1.31868671e-01,  1.50219557e-01,  1.31179136e-01],  
       [-1.62782525e-01, -1.58821825e-01, -1.28322211e-01, ...,  
        5.40694563e-02, -5.44095594e-02, -6.94741471e-02],  
       [-2.50129023e-01, -2.59359844e-01, -3.34978669e-02, ...,  
        -1.83333910e-03,  1.28955424e-01,  8.67015734e-02],  
       ...,  
       [-0.00000000e+00, -1.63654316e-17, -1.19546735e-16, ...,  
        1.28535403e-02, -1.52995704e-02, -1.03250104e-02],  
       [ 0.00000000e+00, -1.52601456e-16,  2.34030598e-16, ...,  
        -2.79682185e-02,  1.35923807e-02, -2.86160073e-02],  
       [-0.00000000e+00, -2.88397778e-17, -3.27325786e-16, ...,  
        -3.36689670e-02, -1.14547085e-01, -6.62390944e-02]])
```

EIGEN VALUES :

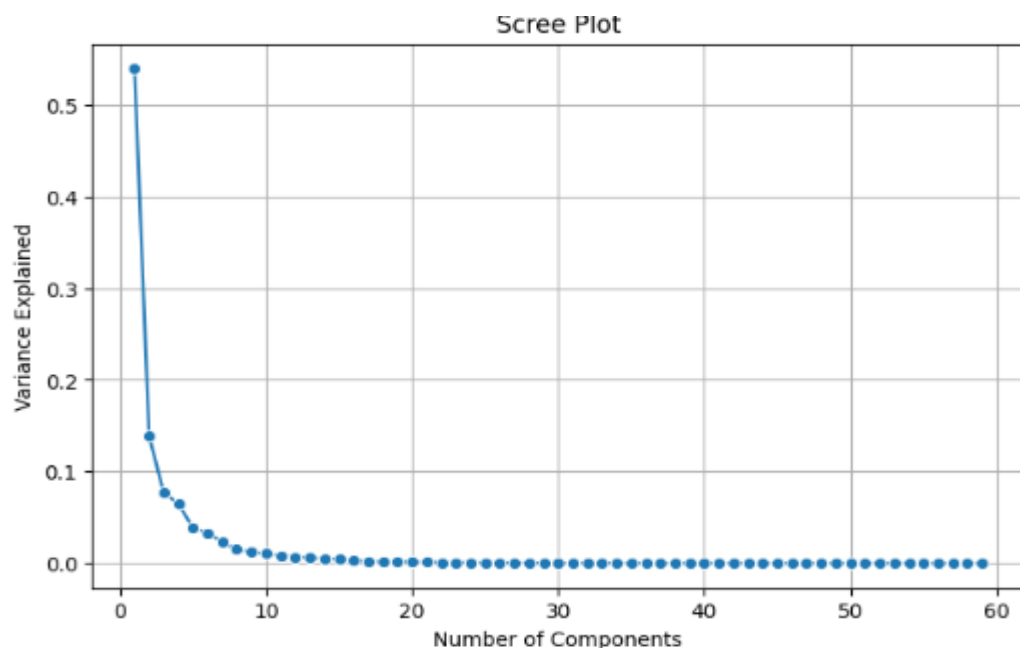
```
array([3.18674263e+01, 8.18907061e+00, 4.54275124e+00, 3.84336785e+00,  
       2.27105793e+00, 1.95992589e+00, 1.37548006e+00, 8.87342674e-01,  
       7.19897963e-01, 6.14059555e-01, 4.94399686e-01, 4.24147991e-01,  
       3.43932360e-01, 2.96118628e-01, 2.75961760e-01, 1.84995268e-01,  
       1.28846861e-01, 1.11536962e-01, 1.03594789e-01, 9.73429345e-02,  
       7.82132546e-02, 5.59614544e-02, 4.44214277e-02, 3.78654873e-02,  
       2.96705436e-02, 2.70572400e-02, 2.34417688e-02, 1.43611558e-02,  
       1.10964929e-02, 9.28775833e-03, 8.27176626e-03, 7.61344489e-03,  
       5.02300148e-03, 4.49943614e-03, 2.51573519e-03, 1.06257176e-03,  
       7.11882677e-04, 6.28474170e-04, 1.09476069e-04, 1.64432752e-31,  
       1.64432752e-31, 1.64432752e-31, 1.64432752e-31, 1.64432752e-31,  
       1.64432752e-31, 1.64432752e-31, 1.64432752e-31, 1.64432752e-31,  
       1.64432752e-31, 1.64432752e-31, 1.64432752e-31, 1.64432752e-31,  
       1.64432752e-31, 1.64432752e-31, 1.64432752e-31, 1.64432752e-31,  
       1.64432752e-31, 1.64432752e-31, 1.64432752e-31, 1.64432752e-31])
```

6) PCA: Identify the optimum number of PCs (for this project, take at least 90% explained variance). Show Scree plot.

```
np.cumsum(pca.explained_variance_ratio_)

array([0.53928192, 0.67786286, 0.75473834, 0.81977838, 0.85821074,
       0.89137792, 0.91465472, 0.92967092, 0.94185352, 0.95224504,
       0.96061161, 0.96778932, 0.97360958, 0.97862069, 0.9832907 ,
       0.98642132, 0.98860175, 0.99048925, 0.99224235, 0.99388966,
       0.99521323, 0.99616025, 0.99691198, 0.99755277, 0.99805487,
       0.99851275, 0.99890945, 0.99915248, 0.99934026, 0.99949743,
       0.99963741, 0.99976625, 0.99985126, 0.9999274 , 0.99996997,
       0.99998795, 1.          , 1.          , 1.          , 1.          ,
       1.          , 1.          , 1.          , 1.          , 1.          ,
       1.          , 1.          , 1.          , 1.          , 1.          ,
       1.          , 1.          , 1.          , 1.          , 1.          ,
       1.          , 1.          , 1.          , 1.          , 1.          ])
```

SCREE PLOT :



7) PCA: Compare PCs with Actual Columns and identify which is explaining most variance. Write inferences about all the Principal components in terms of actual variables.

Abs. loadings of 1

Variable	Abs. Loading
TOT_M	1.0
TOT_F	1.0
F_IL	1.0
MARG_CL_3_6_M	1.0
MARGWORK_3_6_M	1.0
MARGWORK_M	1.0
F_06	1.0
M_LIT	1.0
M_06	1.0
M_IL	1.0
MARGWORK_3_6_F	1.0
TOT_WORK_M	1.0
No_HH	1.0
MARG_CL_3_6_F	1.0
MARGWORK_F	1.0
MARG_OT_M	1.0
MARGWORK_0_3_M	1.0
F_SC	1.0
M_SC	1.0
NON_WORK_M	1.0
MARG_CL_0_3_M	1.0
MARG_OT_F	1.0
F_LIT	1.0
TOT_WORK_F	1.0
MAINWORK_M	1.0
MARGWORK_0_3_F	1.0
MARG_HH_M	1.0
MARG_CL_0_3_F	1.0
MARG_OT_0_3_M	1.0
MARG_OT_3_6_M	1.0
MARG_OT_0_3_F	1.0
MAIN_HH_M	1.0
NON_WORK_F	1.0
MARG_HH_3_6_M	1.0
MARG_AL_M	1.0
MARG_HH_F	1.0
MAINWORK_F	1.0
MARG_OT_3_6_F	1.0
MAIN_OT_M	1.0
MARG_HH_0_3_M	1.0
MARG_HH_0_3_F	1.0
MARG_AL_F	1.0
MAIN_AL_M	1.0
MAIN_OT_F	1.0
MARG_HH_3_6_F	1.0
MAIN_CL_M	1.0
MARG_AL_3_6_M	1.0
MAIN_HH_F	1.0
MARG_CL_M	1.0
MAIN_AL_F	1.0
MAIN_CL_F	1.0
MARG_AL_0_3_M	1.0
MARG_AL_3_6_F	1.0
MARG_CL_F	1.0
MARG_AL_0_3_F	1.0
Dist Code	1.0
State Code	1.0
F_ST	1.0
M_ST	1.0

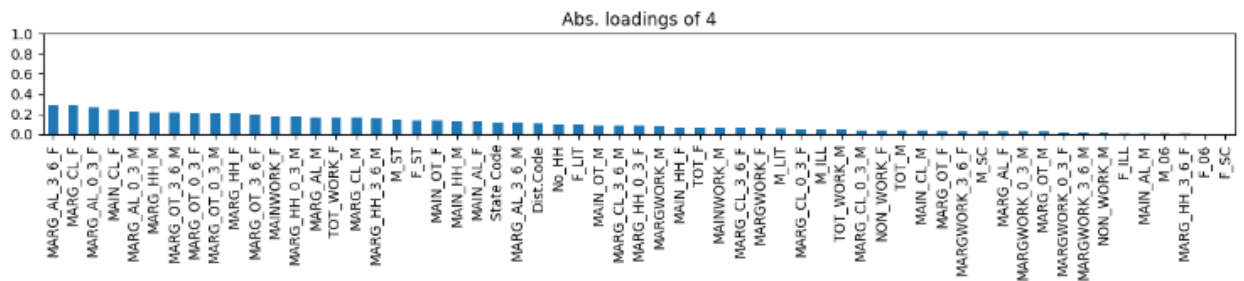
Abs. loadings of 2

Variable	Abs. loading of 2
MARG_CL_M	0.25
MARG_AL_3_6_M	0.25
MARG_AL_0_3_M	0.25
MARG_CL_F	0.25
MARG_AL_3_6_F	0.25
MARG_AL_0_3_F	0.25
MAIN_OT_F	0.25
MAIN_OT_M	0.25
MARG_HH_0_3_M	0.25
MARG_CL_0_3_F	0.25
MAINWORK_M	0.25
MARG_HH_0_3_F	0.25
State Code	0.25
MAINWORK_F	0.25
Dst Code	0.25
MARG_AL_M	0.25
MARG_CL_0_3_M	0.25
MARG_HH_3_6_M	0.25
FLIT	0.25
No_HH	0.25
TOT_WORK_M	0.25
MARG_AL_F	0.25
MARGWORK_0_3_F	0.25
MARGWORK_F	0.25
MARG_OT_F	0.25
MLIT	0.25
TOT_F	0.25
MARG_HH_3_6_F	0.25
TOT_WORK_F	0.25
MARGWORK_M	0.25
MARGWORK_3_6_F	0.25
MARG_OT_0_3_M	0.25
MARG_CL_3_6_F	0.25
MAIN_AL_F	0.25
MAIN_HH_F	0.25
MARGWORK_0_3_M	0.25
TOT_M	0.25
MARG_CL_3_6_M	0.25
MARG_HH_M	0.25
MARG_OT_M	0.25
MAIN_CL_F	0.25
MARG_OT_3_6_M	0.25
NON_WORK_F	0.25
MAIN_HH_M	0.25
MAIN_CL_M	0.25
NON_WORK_M	0.25
MARG_OT_0_3_F	0.25
F_SC	0.25
MAIN_AL_M	0.25
M_SC	0.25
MARGWORK_3_6_M	0.25
MARG_HH_F	0.25
MARG_OT_3_6_F	0.25
M_06	0.25
F_ILL	0.25
F_06	0.25
F_ST	0.25
M_ST	0.25
M_ILL	0.25

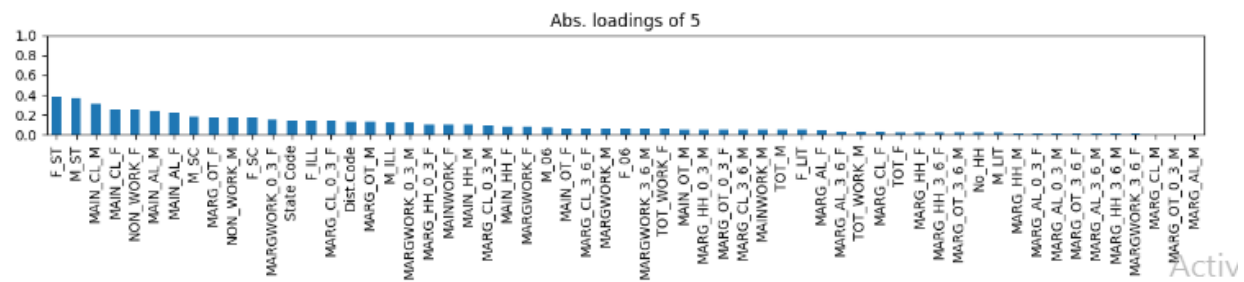
Abs. loadings of 3

Variable	Abs. loading of 3
MAIN_AL_F	0.35
MARG_HH_3_6_F	0.32
MARG_AL_F	0.30
Dist.Code	0.28
State.Code	0.25
MAIN_AL_M	0.22
F_5T	0.20
M_5T	0.18
MARG_HH_0_3_F	0.15
MARG_HH_3_6_M	0.12
MARG_AL_0_3_F	0.10
MARG_AL_0_3_M	0.08
MARG_AL_M	0.07
MARG_OT_M	0.06
MARGWORK_0_3_M	0.05
MAIN_OT_M	0.04
NON_WORK_M	0.03
TOT_WORK_F	0.02
MARG_CL_F	0.01
MARG_CL_3_6_F	0.01
MAIN_CL_F	0.01
MAINWORK_F	0.01
MARG_AL_3_6_F	0.01
MARG_CL_M	0.01
MAIN_HH_M	0.01
MAINWORK_F	0.01
MARG_OT_F	0.01
MARGWORK_0_3_F	0.01
F_LIT	0.01
F_IL	0.01
MARGWORK_3_6_F	0.01
MARG_HH_0_3_M	0.01
NON_WORK_F	0.01
N_LIT	0.01
MARG_AL_3_6_M	0.01
MAIN_CL_M	0.01
MARGWORK_3_6_M	0.01
M_06	0.01
MARG_OT_3_6_M	0.01
MARG_HH_M	0.01
TOT_M	0.01
E_06	0.01
MAIN_OT_F	0.01
MARG_OT_0_3_M	0.01
MARG_CL_0_3_M	0.01
MAINWORK_M	0.01
TOT_WORK_M	0.01
M_SC	0.01
No_HH	0.01
MAIN_HH_F	0.01
TOT_F	0.01
MARG_CL_0_3_F	0.01
MARGWORK_M	0.01
M_IL	0.01
F_SC	0.01
MARG_CL_3_6_M	0.01
MARG_OT_3_6_F	0.01
MARG_HH_F	0.01
MARG_OT_0_3_F	0.01

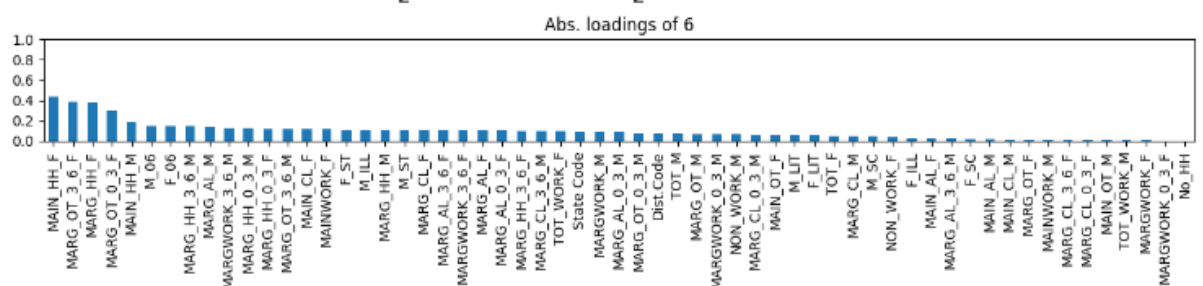
PC4 :



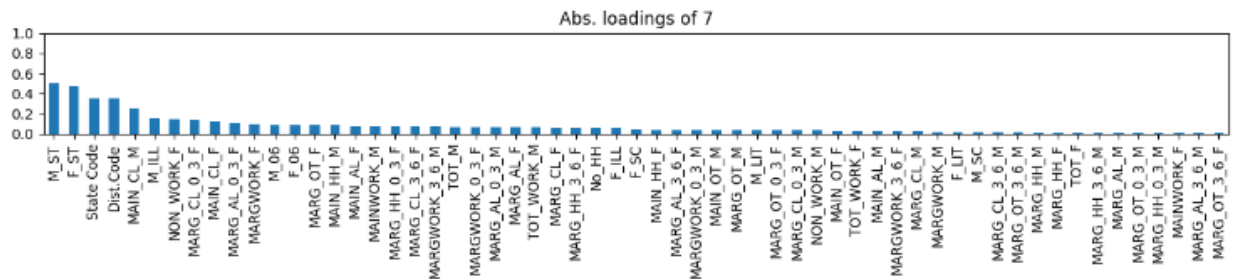
PC5 :



PC6 :



PC7 :



From the above we can say that PC1 exhibit maximum variance when compared to others. Finally we got the heatmap with no correlation exists among the PCs.



8) **PCA: Write linear equation for first PC.**

In General, Linear equation for PC1 :

$$\mathbf{PC1 = a_1x_1 + a_2x_2 + \dots + a_nx_n}$$

a_1, a_2, \dots, a_n : co-efficient / Eigen vectors / Factor loadings

x_1, x_2, \dots, x_n : observed data

$$\mathbf{PC1 = 0.030070 (State\ CODE) + 0.030075 (Dist.CODE) + 0.156432 (NO_HH) + \dots + 0.131179 (NON_WORK_F)}$$

INSIGHTS :

- ❖ **UTTAR PRADESH** has highest population
- ❖ **DADRA & NAGAR HAVELLI** has lowest population in India
- ❖ **ANDRA PRADESH** has highest Gender ratio of 1895 Females per 1000 Males.
- ❖ **LAKSHWADEEP** has lowest Gender ratio of 1152 Females per 1000 Males.
- ❖ **KERALA** has highest literacy rate of 80.59 %
- ❖ **BIHAR** has lowest literacy rate of 47.98%
- ❖ 79.87 % of female population in **KERALA** are literates (highest).
- ❖ 39.75 % of female population are literates in **BIHAR** (lowest).
- ❖ **UTTAR PRADESH** has more Scheduled caste population.
- ❖ **MADHYA PRADESH** has more Scheduled Tribes population.
- ❖ **NAGALAND** has 53.93 % worker population
- ❖ **LAKSHADWEEP** has only 24.98 % worker population with respect to their Total Population.
- ❖ **ANDRA PRADESH** has 18.09 % Agricultural population who predominantly depends upon Agriculture for their livelihood
- ❖ **LAKSHADWEEP** has 0 % Agricultural population.

THANK YOU