



**NIC TECHGOV-2020**

**'AI-IDEATHON'**

National level Challenge for NICians



# AI based Tax Fraud Detection

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# Objectives

- Identify suspicious taxpayers likely to evade tax
- Estimate the amount of tax that could be recovered
- Provide the suspicious taxpayer list for scrutiny, assessment and issue of notices and recovery
- Compare the identified suspicious list with the actual list of confirmed tax evaders

# Python and open source AI Libraries and Algorithms

## ■ AI Libraries used

- ▶ Sklearn
- ▶ Pandas
- ▶ Plotly

## ■ AI algorithms used of clustering:

- ▶ DBSan
- ▶ K-Medoids
- ▶ K-Means
- ▶ Silhouette scoring

## ■ Fraud Detection

- ▶ Graph based weighted MAD
- ▶ Benford Analysis

# Base Data for Clustering

- Identified 7 key base data fields from monthly GST return filing for correlation analysis
  - ▷ Total SGST Liability
  - ▷ Total CGST Liability
  - ▷ Total Liability
  - ▷ Total ITC available
  - ▷ Total IGST ITC
  - ▷ Total SGST paid by cash
  - ▷ Total Exempt Sales
- Identified 5 major business sectors for independent study
  - ▷ Gold
  - ▷ Steel
  - ▷ Textile
  - ▷ Timber
  - ▷ Vehicle

# Base Data Sample

	gstin	month	year	total_sales_amount	total_sgst_paid_by_cash
0	ABBBBBB0079N1ZU	7	2,017	53530.0	0.0
1	ABBBBBB0079N1ZU	8	2,017	97376.6	0.0
2	ABBBBBB0079N1ZU	9	2,017	75530.0	21950.0
3	ABBBBBB0079N1ZU	10	2,017	143321.6	0.0
4	ABBBBBB0079N1ZU	11	2,017	540785.0	11035.0

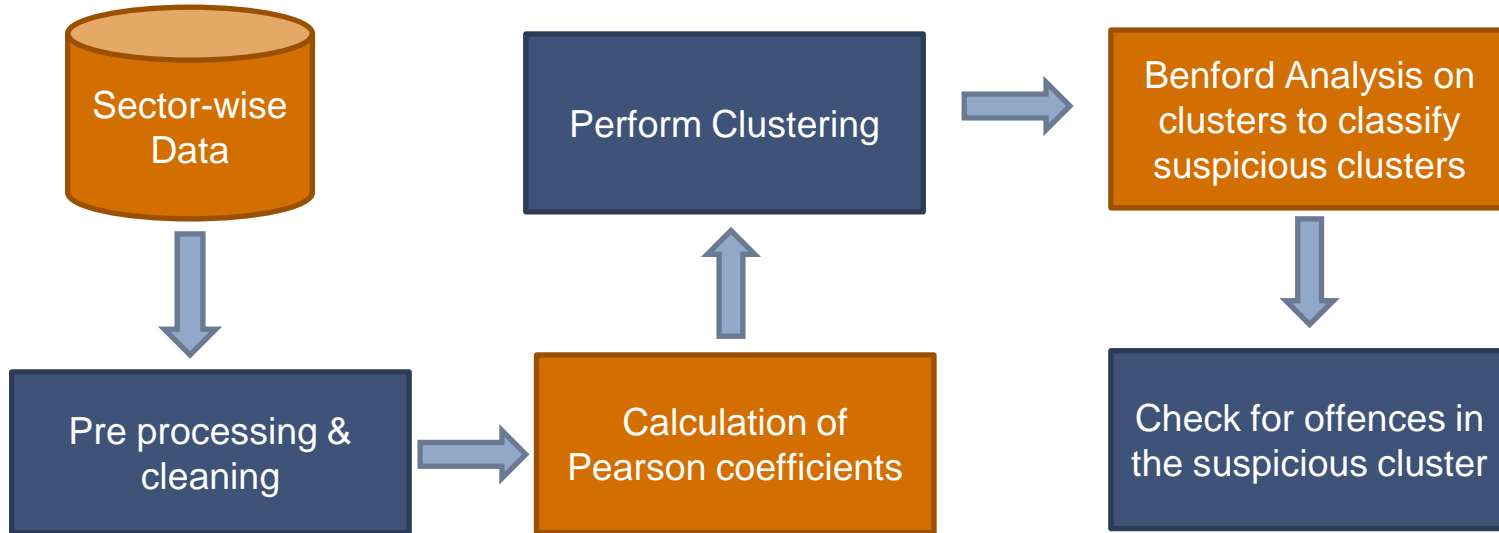
	total_intra_state_sales	total_inter_state_sales	total_gst_paid_by_cash
0	53530.0	0.0	0.0
1	97376.6	0.0	0.0
2	75530.0	0.0	25925.0
3	143321.6	0.0	0.0
4	540785.0	0.0	11035.0

## Base data for Benford Analysis

- Business to business invoice data from GSTR-1 return

	gstin	ctin	invoice_value
0	ABBBBBB070BB1ZB	06BBGCB9961P1ZX	8,054
1	ABBBBBB070BB1ZB	BABDJFS9A71G1ZR	1,07,125
2	ABBBBBB070BB1ZB	B7BBBCR4849R1ZL	4,065
3	ABBBBBB070BB1ZB	B9BNOPL6798H1Z0	73,142
4	ABBBBBB070BB1ZB	ABBBBCS90ABR1Z0	8,21,440

# Approach



# Activities Carried out

## ■ Step 1: Identifying important sectors

- ▷ Gold
- ▷ Timber
- ▷ Vehicle
- ▷ Steel
- ▷ Textiles

## ■ Step 2: Identifying important parameter

- ▷ Total Sales Amount
- ▷ Total SGST liability
- ▷ Total CGST liability
- ▷ Total Liability
- ▷ Total SGST paid in cash
- ▷ Total Exempt sales
- ▷ Total ITC
- ▷ IGST ITC



## Activities Carried out

- Step 3 : Extraction and cleaning of sector-wise data from July-2017 to Dec-2019 (30 months)
- Step 4 : Heat map analysis to identify parameter pairs with high overall correlation (for each sector)

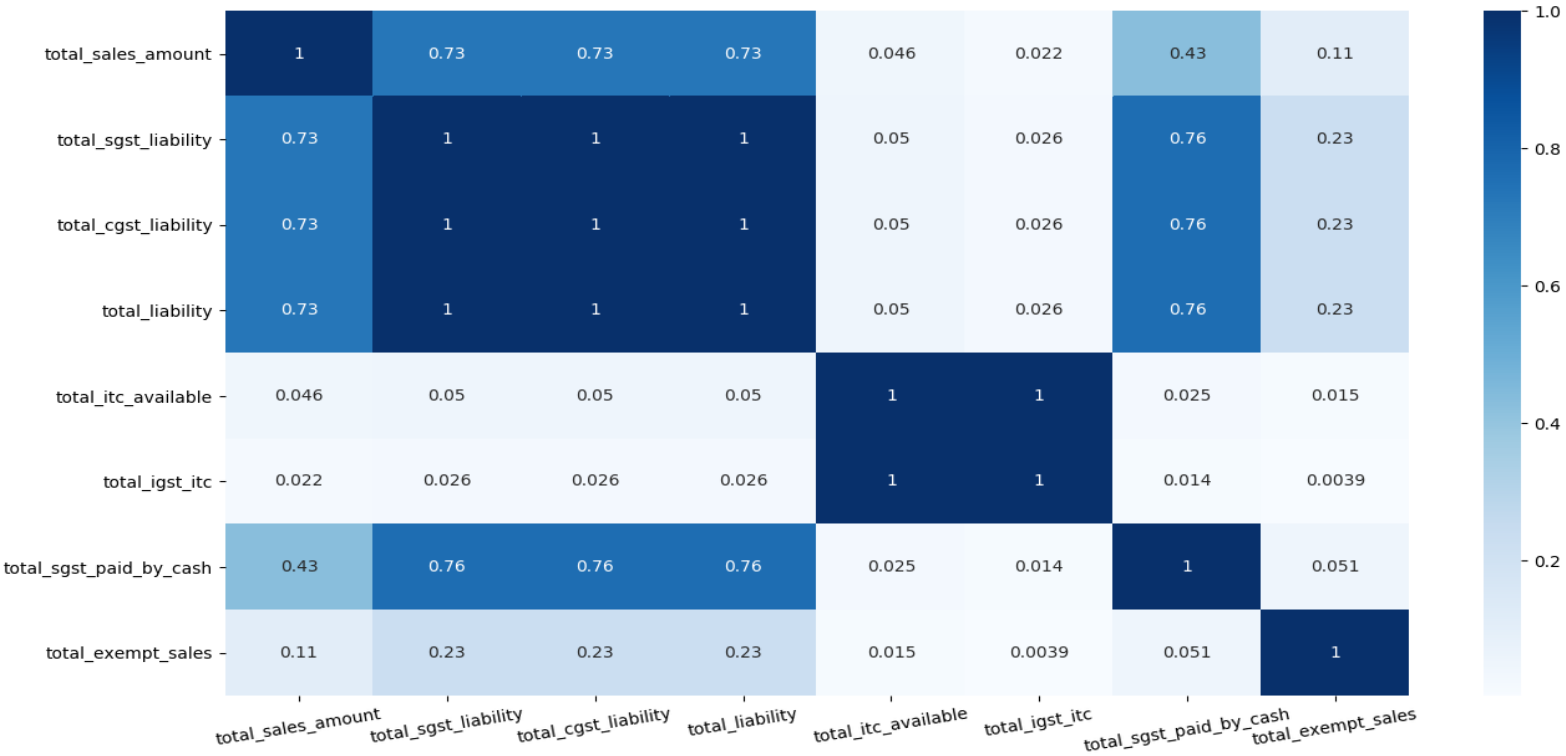
## Activities Carried out

- Step 5 : Calculate Pearson's correlation coefficient between the pairs identified
  - ▷ Calculated for all taxpayers using period wise data for 30 months
  - ▷ The Pearson correlation coefficient measures the linear relationship between two datasets.
  - ▷ It varies between -1 and +1 with 0 implying no correlation. Correlations of -1 or +1 imply an exact linear relationship.
  - ▷ Positive correlations imply that as x increases, so does y. Negative correlations imply that as x increases, y decreases.

## Activities Carried out

- Step 6 : Taxpayer clustering based on the correlation coefficients
  - ▷ DBSan
  - ▷ K-Medoids
  - ▷ K-Means

# Feature Correlation Heatmap



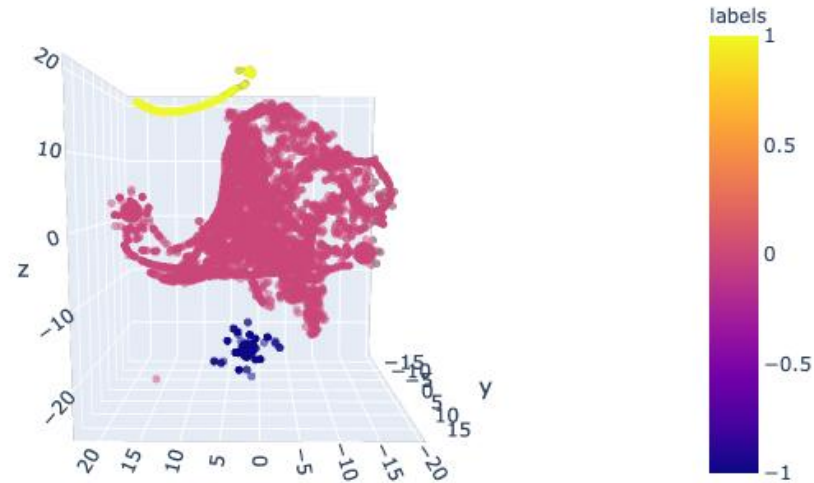
# Data used for clustering

- Pearson coefficients were calculated from base data
  - ▷ Val1 = Total sales amount vs Total sales amount
  - ▷ Val2 = Total GST liability Vs. Total SGST liability
  - ▷ Val3 = Total SGST liability Vs. Total SGST paid in cash
  - ▷ Val4 = Total sales amount Vs. Total SGST paid in cash
  - ▷ Val5 = Total sales amount Vs. Total Exempt sales
  - ▷ Val6 = Total Liability Vs. Total ITC
  - ▷ Val7 = Total ITC Vs. IGST ITC

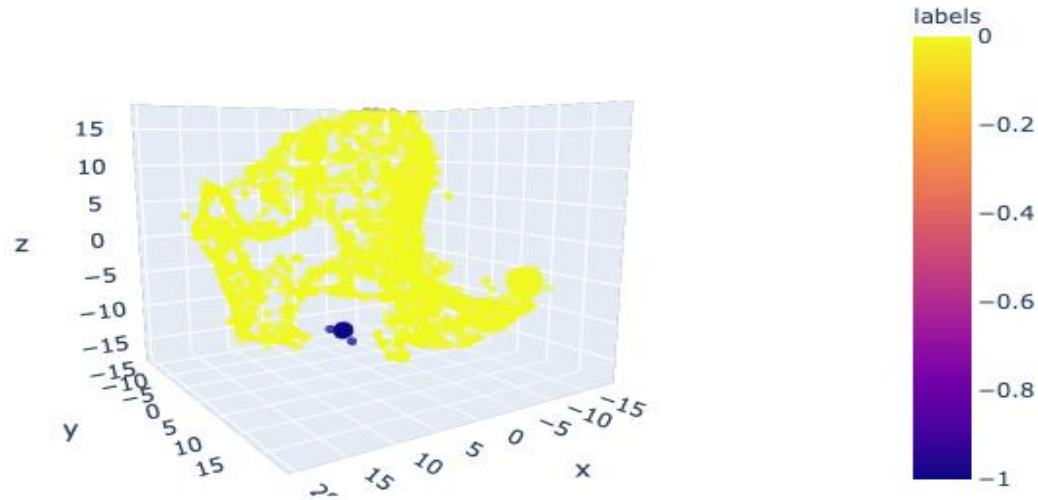
# Sample Pearson Coefficient Data

	ID	val1	val2	val3	val4	val5	val6	val7
0	0.0	1.000000	1.0	0.000000	0.000000	0.0	1.000000	0.962045
1	1.0	0.995825	1.0	0.344852	0.369157	0.0	0.995825	0.851585
2	2.0	0.999997	1.0	0.941790	0.941638	0.0	0.999997	0.623610
3	3.0	0.983462	1.0	0.000000	0.000000	0.0	0.983462	0.084125
4	4.0	0.442941	1.0	0.812662	0.218818	0.0	0.442941	0.284484

# Cluster Analysis on Timber Sector data

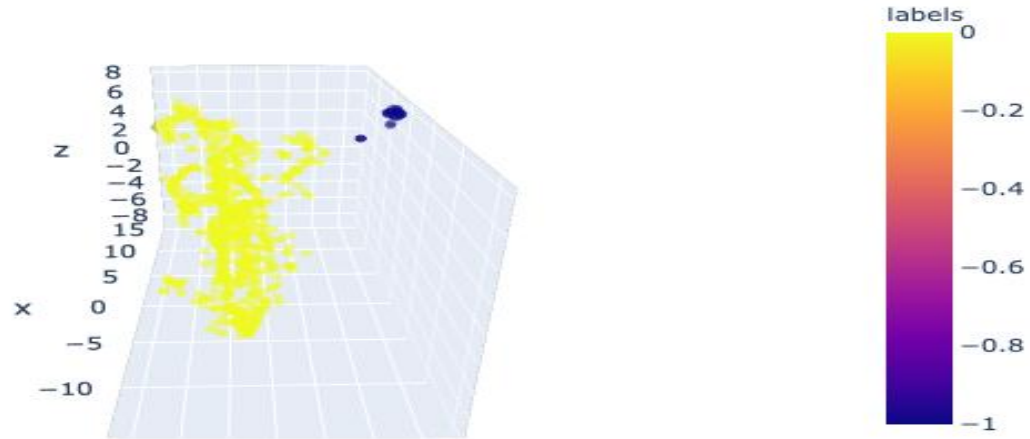


# Cluster Analysis on Vehicle Sector data

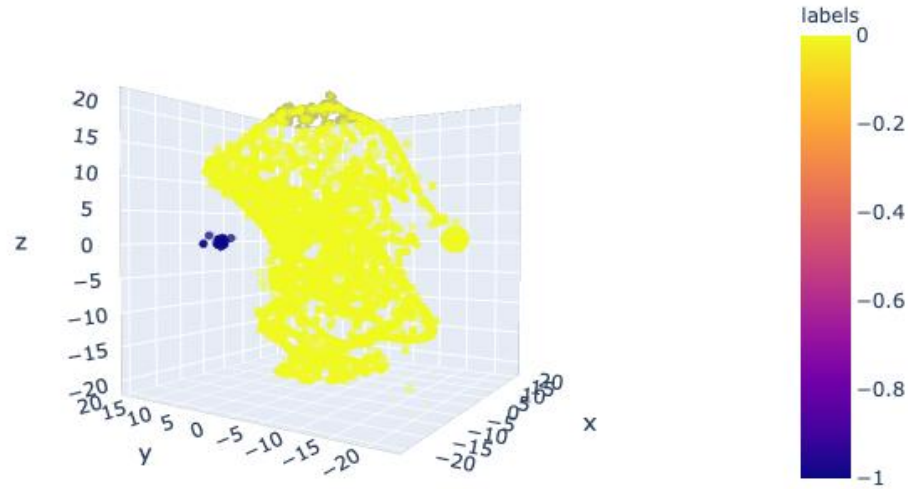




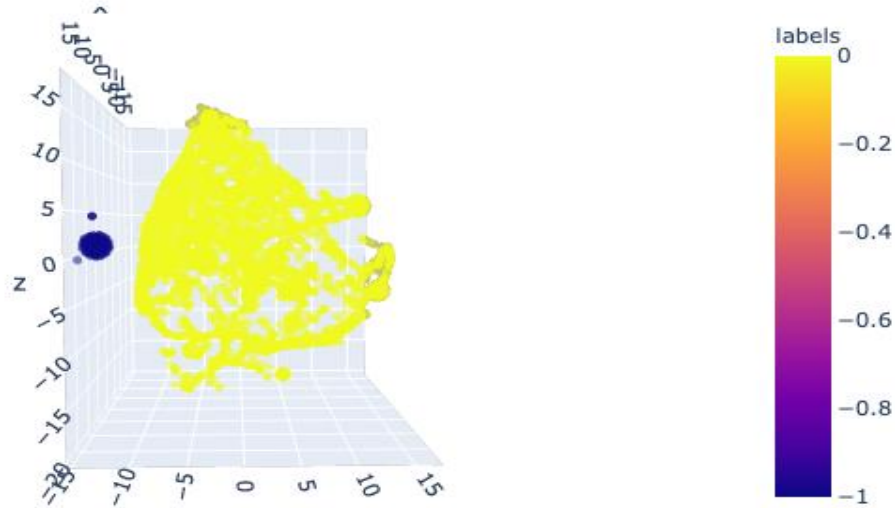
# Cluster Analysis on Textile Sector data



# Cluster Analysis on Steel Sector data



# Cluster Analysis on Gold Sector data



# Cluster Analysis Results

Gold Sector	Cluster 1	Cluster 2	Total
No. of observations	<b>4677</b>	<b>78</b>	<b>4755</b>

Steel Sector	Cluster 1	Cluster 2	Total
No. of observations	<b>5434</b>	<b>51</b>	<b>5485</b>

Vehicle Sector	Cluster 1	Cluster 2	Total
No. of observations	<b>3920</b>	<b>50</b>	<b>3970</b>

Textile Sector	Cluster 1	Cluster 2	Total
No. of observations	861	22	883

Timber Sector	Cluster 1	Cluster 2	Cluster 3	Total
No. of observations	5458	264	418	6140

## Activities Carried out

- Step 6 : Validation of clustering results
  - ▷ Silhouette score was used as a metric for cluster validation
  - ▷ Estimated average distance between clusters
  - ▷ Silhouette Coefficient for a sample is  $(b - a) / \max(a, b)$

Clustering Algorithm		Timber	Gold	Steel	Vehicle	Textile
Silhouette score	DBScan	0.75	0.87	0.90	0.90	0.85
	K-Means	0.68	0.44	0.48	0.57	0.49
	K-Mediods	0.35	0.48	0.37	0.53	0.34

# Activities Carried out

## Step 6 : Benford Analysis

- ▶ Benford's law, also called the Newcomb–Benford law, is an observation about the frequency distribution of leading digits in many real-life sets of numerical data. The law states that in many naturally occurring collections of numbers, the leading significant digit is likely to be small.

$d$	$P(d)$	Relative size of $P(d)$
1	30.1%	
2	17.6%	
3	12.5%	
4	9.7%	
5	7.9%	
6	6.7%	
7	5.8%	
8	5.1%	
9	4.6%	

# Benford Analysis

- Mean Absolute Deviation (MAD) is the measure we used to evaluate the conformity to Benford's Law.

$$\text{Mean Absolute Deviation} = \frac{\sum_{i=1}^K |AP - EP|}{K}$$

- According to research values greater than 0.012 are said to be non-conformal to Benford's law.



# Benford Analysis – Timber Sector

Cluster-1: 5458 (Conformant to Benford)

Cluster-2: 264 (Suspicious)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	904	29.89	910	30.10	-6	-0.21
2	580	19.18	532	17.60	48	1.58
3	349	11.54	378	12.50	-29	-0.96
4	315	10.42	293	9.70	22	0.72
5	247	8.17	239	7.90	8	0.27
6	198	6.55	203	6.70	-5	-0.15
7	132	4.37	175	5.80	-43	-1.43
8	140	4.63	154	5.10	-14	-0.47
9	152	5.03	139	4.60	13	0.43

cluster One: 0.006903880070546738  
cluster members: 5458

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	3	30.00	3	30.10	-0	-0.10
2	3	30.00	2	17.60	1	12.40
3	2	20.00	1	12.50	1	7.50
4	0	0.00	1	9.70	-1	-9.70
5	1	10.00	1	7.90	0	2.10
6	0	0.00	1	6.70	-1	-6.70
7	0	0.00	1	5.80	-1	-5.80
8	0	0.00	1	5.10	-1	-5.10
9	1	10.00	0	4.60	1	5.40

cluster Two: 0.060888888888888895  
cluster members: 264

# Benford Analysis – Timber Sector

## Cluster-2: 418(Suspicious)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	9	32.14	8	30.10	1	2.04
2	4	14.29	5	17.60	-1	-3.31
3	3	10.71	4	12.50	-0	-1.79
4	6	21.43	3	9.70	3	11.73
5	2	7.14	2	7.90	-0	-0.76
6	2	7.14	2	6.70	0	0.44
7	1	3.57	2	5.80	-1	-2.23
8	0	0.00	1	5.10	-1	-5.10
9	1	3.57	1	4.60	-0	-1.03

cluster three: 0.03158730158730159

cluster members: 418

# Benford Analysis – Steel Sector

Cluster-1: 5434 (Conformant to Benford)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	7395	29.94	7434	30.10	-39	-0.16
2	4275	17.31	4347	17.60	-72	-0.29
3	3075	12.45	3087	12.50	-12	-0.05
4	2395	9.70	2396	9.70	-1	-0.00
5	2022	8.19	1951	7.90	71	0.29
6	1649	6.68	1655	6.70	-6	-0.02
7	1472	5.96	1432	5.80	40	0.16
8	1276	5.17	1260	5.10	16	0.07
9	1126	4.56	1136	4.60	-10	-0.04

cluster One: 0.001196402622000869

cluseter members: 5434

Cluster-2: 51 (Suspicious)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	6	27.27	7	30.10	-1	-2.83
2	3	13.64	4	17.60	-1	-3.96
3	2	9.09	3	12.50	-1	-3.41
4	4	18.18	2	9.70	2	8.48
5	4	18.18	2	7.90	2	10.28
6	1	4.55	1	6.70	-0	-2.15
7	1	4.55	1	5.80	-0	-1.25
8	1	4.55	1	5.10	-0	-0.55
9	0	0.00	1	4.60	-1	-4.60

cluster three: 0.04169696969696969

cluster members: 51

# Benford Analysis – Vehicle Sector

Cluster-1: 3920 (Conformant to Benford)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	1389	30.10	1389	30.10	0	0.00
2	810	17.56	812	17.60	-2	-0.04
3	577	12.51	577	12.50	0	0.01
4	419	9.08	448	9.70	-29	-0.62
5	393	8.52	365	7.90	28	0.62
6	269	5.83	309	6.70	-40	-0.87
7	299	6.48	268	5.80	31	0.68
8	259	5.61	235	5.10	24	0.51
9	199	4.31	212	4.60	-13	-0.29

cluster One: 0.0040458507922747225  
 cluser members: 3920

Cluster-2: 50 (Suspicious)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	10	34.48	9	30.10	1	4.38
2	10	34.48	5	17.60	5	16.88
3	0	0.00	4	12.50	-4	-12.50
4	2	6.90	3	9.70	-1	-2.80
5	2	6.90	2	7.90	-0	-1.00
6	0	0.00	2	6.70	-2	-6.70
7	1	3.45	2	5.80	-1	-2.35
8	4	13.79	1	5.10	3	8.69
9	0	0.00	1	4.60	-1	-4.60

cluster three: 0.06657471264367817  
 cluster members: 50

# Benford Analysis – Gold Sector

## Cluster-1: 4677 (Conformant to Benford)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	1496	29.62	1520	30.10	-24	-0.48
2	906	17.94	889	17.60	17	0.34
3	603	11.94	631	12.50	-28	-0.56
4	483	9.56	490	9.70	-7	-0.14
5	471	9.33	399	7.90	72	1.43
6	324	6.42	338	6.70	-14	-0.28
7	295	5.84	293	5.80	2	0.04
8	248	4.91	258	5.10	-10	-0.19
9	223	4.42	232	4.60	-9	-0.18

cluster One: 0.004041804180418043

clusester members: 4677

## Cluster-2: 78 (Suspicious)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	3	21.43	4	30.10	-1	-8.67
2	1	7.14	2	17.60	-1	-10.46
3	1	7.14	2	12.50	-1	-5.36
4	2	14.29	1	9.70	1	4.59
5	2	14.29	1	7.90	1	6.39
6	2	14.29	1	6.70	1	7.59
7	1	7.14	1	5.80	0	1.34
8	2	14.29	1	5.10	1	9.19
9	0	0.00	1	4.60	-1	-4.60

cluster three: 0.06463492063492063

cluster members: 78

# Benford Analysis – Textile Sector

Cluster-1: 861 (Conformant Benford)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	395	30.86	385	30.10	10	0.76
2	244	19.06	225	17.60	19	1.46
3	176	13.75	160	12.50	16	1.25
4	114	8.91	124	9.70	-10	-0.79
5	95	7.42	101	7.90	-6	-0.48
6	80	6.25	86	6.70	-6	-0.45
7	46	3.59	74	5.80	-28	-2.21
8	71	5.55	65	5.10	6	0.45
9	59	4.61	59	4.60	0	0.01

cluster One: 0.008729166666666667  
 cluseter members: 861

Cluster-2: 22 (Suspicious)

n	Data		Benford		Difference	
	Freq	Pct	Freq	Pct	Freq	Pct
1	3	42.86	2	30.10	1	12.76
2	1	14.29	1	17.60	-0	-3.31
3	2	28.57	1	12.50	1	16.07
4	0	0.00	1	9.70	-1	-9.70
5	0	0.00	1	7.90	-1	-7.90
6	0	0.00	0	6.70	-0	-6.70
7	1	14.29	0	5.80	1	8.49
8	0	0.00	0	5.10	-0	-5.10
9	0	0.00	0	4.60	-0	-4.60

cluster three: 0.08292063492063492  
 cluster members: 22

## Usability of Model

- Sector wise suspicious dealer list shall be provided to department
- Prioritization of assessment based on results
- Tuning the model to improve accuracy based on feedback
- Development and integration of the AI module in GST backend for future analysis

# Thank You