

# 320-govt-upd-1-1

March 20, 2025

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2 2211cs010320

3 group 4

## 4 Indra Gandhi Widow Pension Scheme FY 2023-2024

The \*\*Indira Gandhi Widow Pension Scheme (2023-2024) ensures structured tracking of welfare benefits across regions and demographics. LGD codes uniquely identify states and districts, while scheme codes categorize programs. This enhances transparency and accountability, ensuring funds reach the right beneficiaries. Aadhaar and mobile numbers aid in DBT validation, preventing fraud. Demographic analysis (SC, ST, OBC, General) ensures fair distribution. Time-based tracking (financial year, month, last update) allows trend analysis and irregularity detection. Covering multiple districts in Jammu & Kashmir the dataset provides a comprehensive and structured approach to welfare monitoring.

```
[2]: import pandas as pd
import numpy as np
import seaborn as sb
import matplotlib.pyplot as plt
import math
df=pd.read_csv("govt_dataset.csv");
df
```

```
[2]:
```

	fin_year	mnth	lgd_state_code	state_name	lgd_district_code	\
0	2023-2024	2	1	JAMMU AND KASHMIR	13	
1	2023-2024	2	1	JAMMU AND KASHMIR	14	
2	2023-2024	2	1	JAMMU AND KASHMIR	2	
3	2023-2024	2	1	JAMMU AND KASHMIR	4	
4	2023-2024	2	1	JAMMU AND KASHMIR	622	
...	...	...	...	...	...	
14377	2024-2025	2	9	UTTAR PRADESH	172	
14378	2024-2025	2	9	UTTAR PRADESH	173	
14379	2024-2025	2	9	UTTAR PRADESH	176	
14380	2024-2025	2	9	UTTAR PRADESH	177	
14381	2024-2025	2	9	UTTAR PRADESH	181	

	district_name	scheme_code	total_beneficiaries	total_sc	total_st	\
0	SRINAGAR	IGNWPS	546	0	1	
1	UDHAMPUR	IGNWPS	353	85	17	
2	BADGAM	IGNWPS	176	3	1	
3	DODA	IGNWPS	246	9	18	
4	KULGAM	IGNWPS	290	20	0	
...	...	...	...	...	...	
14377	MUZAFFARNAGAR	IGNWPS	7601	1817	13	
14378	PILIBHIT	IGNWPS	8268	1796	0	
14379	RAMPUR	IGNWPS	23245	3169	0	
14380	SAHARANPUR	IGNWPS	10667	3649	143	
14381	SHRAVASTI	IGNWPS	6168	1689	332	

	total_gen	total_obc	total_aadhar	total_mobilenos	lastUpdated
0	544	1	403	127	2024-07-31
1	238	13	250	121	2024-07-31
2	170	2	150	25	2024-07-31
3	219	0	237	33	2024-07-31
4	257	13	290	0	2024-07-31
...	...	...	...	...	...
14377	2719	3052	3680	3177	2025-03-07
14378	1756	4716	4037	4306	2025-03-07
14379	7868	12208	9288	9096	2025-03-07
14380	2940	3935	5896	5587	2025-03-07
14381	2203	1944	1649	2063	2025-03-07

[14382 rows x 15 columns]

[ ]:

**5 This loads necessary libraries and reads “govt\_dataset.csv” into a DataFrame for analysis.**

[3]: df.head(10)

	fin_year	mnth	lgd_state_code	state_name	lgd_district_code	\
0	2023-2024	2	1	JAMMU AND KASHMIR	13	
1	2023-2024	2	1	JAMMU AND KASHMIR	14	
2	2023-2024	2	1	JAMMU AND KASHMIR	2	
3	2023-2024	2	1	JAMMU AND KASHMIR	4	
4	2023-2024	2	1	JAMMU AND KASHMIR	622	
5	2023-2024	2	1	JAMMU AND KASHMIR	624	
6	2023-2024	2	1	JAMMU AND KASHMIR	625	
7	2023-2024	2	10	BIHAR	196	
8	2023-2024	2	10	BIHAR	202	

9	2023-2024	2	10	BIHAR	204
---	-----------	---	----	-------	-----

	district_name	scheme_code	total_beneficiaries	total_sc	total_st	\
0	SRINAGAR	IGNWPS	546	0	1	
1	UDHAMPUR	IGNWPS	353	85	17	
2	BADGAM	IGNWPS	176	3	1	
3	DODA	IGNWPS	246	9	18	
4	KULGAM	IGNWPS	290	20	0	
5	SAMBA	IGNWPS	117	14	4	
6	SHOPIAN	IGNWPS	454	0	47	
7	GAYA	IGNWPS	15236	6590	87	
8	KHAGARIA	IGNWPS	15163	1623	271	
9	LAKHISARAI	IGNWPS	4116	684	34	

	total_gen	total_obc	total_aadhar	total_mobilenos	lastUpdated
0	544	1	403	127	2024-07-31
1	238	13	250	121	2024-07-31
2	170	2	150	25	2024-07-31
3	219	0	237	33	2024-07-31
4	257	13	290	0	2024-07-31
5	98	1	83	2	2024-07-31
6	407	0	454	315	2024-07-31
7	7644	915	11236	11618	2024-07-31
8	11498	1771	10253	12412	2024-07-31
9	1588	1810	2883	3181	2024-07-31

Displays the first 10 rows of the DataFrame to preview the data.

```
[5]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
,RangeIndex: 14382 entries, 0 to 14381
,Data columns (total 15 columns):
, #   Column                Non-Null Count  Dtype
, ---  ---
, 0   fin_year              14382 non-null  object
, 1   mnth                 14382 non-null  int64
, 2   lgd_state_code       14382 non-null  int64
, 3   state_name           14382 non-null  object
, 4   lgd_district_code    14382 non-null  int64
, 5   district_name        14382 non-null  object
, 6   scheme_code          14382 non-null  object
, 7   total_beneficiaries  14382 non-null  int64
, 8   total_sc             14382 non-null  int64
, 9   total_st             14382 non-null  int64
, 10  total_gen            14382 non-null  int64
, 11  total_obc            14382 non-null  int64
, 12  total_aadhar         14382 non-null  int64
```

```
, 13 total_mobilenos      14382 non-null  int64
, 14 lastUpdated          14382 non-null  object
dtypes: int64(10), object(5)
memory usage: 1.6+ MB
```

Shows the DataFrame's structure, including column names, data types, and non-null counts.

```
[7]: print(df.isnull().sum())
```

```
fin_year      0
,mnth         0
,lgd_state_code  0
,state_name    0
,lgd_district_code  0
,district_name  0
,scheme_code    0
,total_beneficiaries  0
,total_sc       0
,total_st       0
,total_gen      0
,total_obc      0
,total_aadhar   0
,total_mobilenos  0
,lastUpdated    0
,dtype: int64
```

Displays the total number of missing (null) values in each column of the DataFrame.

```
[9]: df['state_name'].unique()
```

```
[9]: array(['JAMMU AND KASHMIR', 'BIHAR', 'SIKKIM', 'ARUNACHAL PRADESH',
'NAGALAND', 'MANIPUR', 'MIZORAM', 'TRIPURA', 'MEGHALAYA', 'ASSAM',
'WEST BENGAL', 'HIMACHAL PRADESH', 'JHARKHAND', 'ODISHA',
'CHHATTISGARH', 'MADHYA PRADESH', 'GUJARAT', 'MAHARASHTRA',
'ANDHRA PRADESH', 'KARNATAKA', 'PUNJAB', 'GOA', 'KERALA',
'TAMIL NADU', 'TELANGANA', 'LADAKH',
'THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU', 'UTTARAKHAND',
'HARYANA', 'LAKSHADWEEP', 'CHANDIGARH', 'DELHI', 'RAJASTHAN',
'UTTAR PRADESH', 'PUDUCHERRY', 'ANDAMAN AND NICOBAR'], dtype=object)
```

Lists all unique state names present in the “state\_name” column.

```
[11]: df['state_name'].value_counts()
```

```
[11]: state_name
UTTAR PRADESH      1360
MADHYA PRADESH     1020
BIHAR              760
MAHARASHTRA        680
```

GUJARAT	660
CHHATTISGARH	660
ASSAM	660
TELANGANA	660
RAJASTHAN	660
TAMIL NADU	640
KARNATAKA	600
ODISHA	600
ARUNACHAL PRADESH	500
JHARKHAND	480
WEST BENGAL	460
PUNJAB	440
HARYANA	440
ANDHRA PRADESH	430
JAMMU AND KASHMIR	400
UTTARAKHAND	260
KERALA	252
HIMACHAL PRADESH	240
MIZORAM	220
MEGHALAYA	220
NAGALAND	220
DELHI	200
MANIPUR	180
TRIPURA	160
SIKKIM	80
THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU	60
GOA	40
LADAKH	40
ANDAMAN AND NICOBAR	40
LAKSHADWEEP	20
CHANDIGARH	20
PUDUCHERRY	20

Name: count, dtype: int64

Shows the count of occurrences for each unique state in the “state\_name” column.

```
[13]: union_territories = ['JAMMU AND KASHMIR', 'DELHI', 'PUDUCHERRY', 'LADAKH',
    ↪ 'CHANDIGARH', 'ANDAMAN AND NICOBAR ISLANDS']
union_territories
```

```
[13]: ['JAMMU AND KASHMIR',
    'DELHI',
    'PUDUCHERRY',
    'LADAKH',
    'CHANDIGARH',
    'ANDAMAN AND NICOBAR ISLANDS']
```

Creates a list of union territories for later filtering or classification.

```
[15]: df['region_type'] = df['state_name'].apply(lambda x: 'Union Territory' if x in union_territories else 'State')
```

Adds a new column “region\_type” labeling each row as either “State” or “Union Territory”.

```
[17]: df.head
```

```
[17]: <bound method NDFrame.head of
state_name  lgd_district_code  \
0      2023-2024      2      1  JAMMU AND KASHMIR      13
1      2023-2024      2      1  JAMMU AND KASHMIR      14
2      2023-2024      2      1  JAMMU AND KASHMIR       2
3      2023-2024      2      1  JAMMU AND KASHMIR       4
4      2023-2024      2      1  JAMMU AND KASHMIR     622
...      ...      ...      ...      ...      ...
14377  2024-2025      2      9    UTTAR PRADESH     172
14378  2024-2025      2      9    UTTAR PRADESH     173
14379  2024-2025      2      9    UTTAR PRADESH     176
14380  2024-2025      2      9    UTTAR PRADESH     177
14381  2024-2025      2      9    UTTAR PRADESH     181

      district_name  scheme_code  total_beneficiaries  total_sc  total_st  \
0      SRINAGAR      IGNWPS      546      0      1
1      UDHAMPUR      IGNWPS      353      85     17
2      BADGAM      IGNWPS      176      3      1
3      DODA      IGNWPS      246      9     18
4      KULGAM      IGNWPS      290     20      0
...      ...      ...      ...      ...      ...
14377  MUZAFFARNAGAR      IGNWPS      7601     1817     13
14378      PILIBHIT      IGNWPS      8268     1796      0
14379      RAMPUR      IGNWPS     23245     3169      0
14380  SAHARANPUR      IGNWPS     10667     3649     143
14381  SHRAVASTI      IGNWPS      6168     1689     332

      total_gen  total_obc  total_aadhar  total_mobilenos  lastUpdated  \
0      544      1      403      127  2024-07-31
1      238     13      250     121  2024-07-31
2      170      2      150      25  2024-07-31
3      219      0      237      33  2024-07-31
4      257     13      290      0  2024-07-31
...      ...      ...      ...      ...      ...
14377     2719     3052     3680     3177  2025-03-07
14378     1756     4716     4037     4306  2025-03-07
14379     7868    12208     9288     9096  2025-03-07
14380     2940     3935     5896     5587  2025-03-07
14381     2203     1944     1649     2063  2025-03-07
```

```

      region_type
0      Union Territory
1      Union Territory
2      Union Territory
3      Union Territory
4      Union Territory
...
14377      State
14378      State
14379      State
14380      State
14381      State

```

[14382 rows x 16 columns]>

[18]: df

```

[18]:      fin_year  mnth  lgd_state_code  state_name  lgd_district_code  \
0      2023-2024    2          1  JAMMU AND KASHMIR          13
1      2023-2024    2          1  JAMMU AND KASHMIR          14
2      2023-2024    2          1  JAMMU AND KASHMIR           2
3      2023-2024    2          1  JAMMU AND KASHMIR           4
4      2023-2024    2          1  JAMMU AND KASHMIR        622
...
14377  2024-2025    2          9    UTTAR PRADESH        172
14378  2024-2025    2          9    UTTAR PRADESH        173
14379  2024-2025    2          9    UTTAR PRADESH        176
14380  2024-2025    2          9    UTTAR PRADESH        177
14381  2024-2025    2          9    UTTAR PRADESH        181

      district_name  scheme_code  total_beneficiaries  total_sc  total_st  \
0      SRINAGAR      IGNWPS          546           0           1
1      UDHAMPUR      IGNWPS          353           85          17
2      BADGAM      IGNWPS          176            3           1
3      DODA      IGNWPS          246            9          18
4      KULGAM      IGNWPS          290           20           0
...
14377  MUZAFFARNAGAR      IGNWPS          7601        1817          13
14378      PILIBHIT      IGNWPS          8268        1796           0
14379      RAMPUR      IGNWPS        23245        3169           0
14380  SAHARANPUR      IGNWPS        10667        3649         143
14381  SHRAVASTI      IGNWPS          6168        1689         332

      total_gen  total_obc  total_aadhar  total_mobilenos  lastUpdated  \
0           544           1           403           127  2024-07-31
1           238          13           250           121  2024-07-31
2           170           2           150            25  2024-07-31

```

3	219	0	237	33	2024-07-31
4	257	13	290	0	2024-07-31
...	...	...	...	...	...
14377	2719	3052	3680	3177	2025-03-07
14378	1756	4716	4037	4306	2025-03-07
14379	7868	12208	9288	9096	2025-03-07
14380	2940	3935	5896	5587	2025-03-07
14381	2203	1944	1649	2063	2025-03-07

	region_type
0	Union Territory
1	Union Territory
2	Union Territory
3	Union Territory
4	Union Territory
...	...
14377	State
14378	State
14379	State
14380	State
14381	State

[14382 rows x 16 columns]

```
[19]: df.groupby('state_name')['total_beneficiaries'].sum()
```

```
[19]: state_name
ANDAMAN AND NICOBAR      7970
ANDHRA PRADESH          11729503
ARUNACHAL PRADESH        120200
ASSAM                    2290292
BIHAR                    12684420
CHANDIGARH                49715
CHHATTISGARH             4318904
DELHI                     729357
GOA                       117940
GUJARAT                   7762285
HARYANA                   2657668
HIMACHAL PRADESH          439480
JAMMU AND KASHMIR         138212
JHARKHAND                 5084571
KARNATAKA                 8196100
KERALA                    7940714
LADAKH                     8880
LAKSHADWEEP               1500
MADHYA PRADESH            16521580
MAHARASHTRA               1993335
```



MANIPUR	149261
MEGHALAYA	164478
MIZORAM	46401
NAGALAND	84820
ODISHA	10415590
PUDUCHERRY	199180
PUNJAB	379420
RAJASTHAN	6585102
SIKKIM	29380
TAMIL NADU	10181512
TELANGANA	4017940
THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU	98200
TRIPURA	346542
UTTAR PRADESH	20504720
UTTARAKHAND	477974
WEST BENGAL	14402441

Name: total\_beneficiaries, dtype: int64

Sums the total beneficiaries for each state, grouped by the “state\_name” column.

```
[21]: df['total_beneficiaries'].sum()
```

```
[21]: 150875587
```

Calculates the total number of beneficiaries across the entire dataset.

```
[23]: df.groupby('state_name')['total_beneficiaries'].sum().
      ↪sort_values(ascending=False)
```

```
[23]: state_name
UTTAR PRADESH      20504720
MADHYA PRADESH    16521580
WEST BENGAL       14402441
BIHAR             12684420
ANDHRA PRADESH    11729503
ODISHA            10415590
TAMIL NADU        10181512
KARNATAKA         8196100
KERALA            7940714
GUJARAT           7762285
RAJASTHAN         6585102
JHARKHAND         5084571
CHHATTISGARH      4318904
TELANGANA         4017940
HARYANA           2657668
ASSAM             2290292
MAHARASHTRA       1993335
DELHI             729357
```

UTTARAKHAND	477974
HIMACHAL PRADESH	439480
PUNJAB	379420
TRIPURA	346542
PUDUCHERRY	199180
MEGHALAYA	164478
MANIPUR	149261
JAMMU AND KASHMIR	138212
ARUNACHAL PRADESH	120200
GOA	117940
THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU	98200
NAGALAND	84820
CHANDIGARH	49715
MIZORAM	46401
SIKKIM	29380
LADAKH	8880
ANDAMAN AND NICOBAR	7970
LAKSHADWEEP	1500

Name: total\_beneficiaries, dtype: int64

Displays total beneficiaries per state, sorted in descending order.

```
[25]: df.groupby('state_name')['total_beneficiaries'].sum().
      ↪sort_values(ascending=False).head(10)
```

```
[25]: state_name
      UTTAR PRADESH      20504720
      MADHYA PRADESH    16521580
      WEST BENGAL      14402441
      BIHAR            12684420
      ANDHRA PRADESH    11729503
      ODISHA           10415590
      TAMIL NADU        10181512
      KARNATAKA         8196100
      KERALA            7940714
      GUJARAT           7762285
      Name: total_beneficiaries, dtype: int64
```

Shows the top 10 states with the highest total number of beneficiaries.

```
[27]: df[df['total_mobilenos'] == 0][['state_name', 'district_name',
      ↪'total_beneficiaries']]
```

```
[27]:
```

	state_name	district_name	total_beneficiaries
4	JAMMU AND KASHMIR	KULGAM	290
17	ARUNACHAL PRADESH	UPPER DIBANG VALLEY	78
18	ARUNACHAL PRADESH	KURUNG KUMAY	391
21	ARUNACHAL PRADESH	TIRAP	277

22	ARUNACHAL PRADESH	KRA DAADI	143
...	...	...	...
14255	ARUNACHAL PRADESH	SIANG	123
14256	ARUNACHAL PRADESH	PAKKE KESSANG	74
14257	NAGALAND	PHEK	436
14259	NAGALAND	LONGLENG	111
14278	HIMACHAL PRADESH	CHAMBA	2384

[1116 rows x 3 columns]

Lists states and districts where no beneficiaries have mobile numbers, along with their total beneficiaries.

```
[29]: df[df['state_name']=='JAMMU AND KASHMIR']
```

```
[29]:
```

	fin_year	mnth	lgd_state_code	state_name	lgd_district_code	\
0	2023-2024	2	1	JAMMU AND KASHMIR	13	
1	2023-2024	2	1	JAMMU AND KASHMIR	14	
2	2023-2024	2	1	JAMMU AND KASHMIR	2	
3	2023-2024	2	1	JAMMU AND KASHMIR	4	
4	2023-2024	2	1	JAMMU AND KASHMIR	622	
...	...	...	...	...	...	
14239	2024-2025	2	1	JAMMU AND KASHMIR	14	
14240	2024-2025	2	1	JAMMU AND KASHMIR	3	
14241	2024-2025	2	1	JAMMU AND KASHMIR	622	
14242	2024-2025	2	1	JAMMU AND KASHMIR	623	
14243	2024-2025	2	1	JAMMU AND KASHMIR	8	

	district_name	scheme_code	total_beneficiaries	total_sc	total_st	\
0	SRINAGAR	IGNWPS	546	0	1	
1	UDHAMPUR	IGNWPS	353	85	17	
2	BADGAM	IGNWPS	176	3	1	
3	DODA	IGNWPS	246	9	18	
4	KULGAM	IGNWPS	290	20	0	
...	...	...	...	...	...	
14239	UDHAMPUR	IGNWPS	356	85	17	
14240	BARAMULLA	IGNWPS	674	0	0	
14241	KULGAM	IGNWPS	395	20	0	
14242	BANDIPORA	IGNWPS	477	0	54	
14243	KUPWARA	IGNWPS	492	5	10	

	total_gen	total_obc	total_aadhar	total_mobileno	lastUpdated	\
0	544	1	403	127	2024-07-31	
1	238	13	250	121	2024-07-31	
2	170	2	150	25	2024-07-31	
3	219	0	237	33	2024-07-31	
4	257	13	290	0	2024-07-31	

...	...	...	...	...	...	...
14239	241	13	251	127	2025-03-07	
14240	674	0	671	150	2025-03-07	
14241	361	14	395	366	2025-03-07	
14242	422	1	477	134	2025-03-07	
14243	272	205	492	450	2025-03-07	

	region_type
0	Union Territory
1	Union Territory
2	Union Territory
3	Union Territory
4	Union Territory

...	...
14239	Union Territory
14240	Union Territory
14241	Union Territory
14242	Union Territory
14243	Union Territory

[400 rows x 16 columns]

Filters and displays all rows where the state name is “JAMMU AND KASHMIR”.

```
[31]: df['mnth'] = df['mnth'].replace({
      1: 'Jan', 2: 'Feb', 3: 'Mar', 4: 'Apr',
      5: 'May', 6: 'Jun', 7: 'Jul', 8: 'Aug',
      9: 'Sep', 10: 'Oct', 11: 'Nov', 12: 'Dec'
    })
df
```

[31]:	fin_year	mnth	lgd_state_code	state_name	lgd_district_code	\
0	2023-2024	Feb	1	JAMMU AND KASHMIR	13	
1	2023-2024	Feb	1	JAMMU AND KASHMIR	14	
2	2023-2024	Feb	1	JAMMU AND KASHMIR	2	
3	2023-2024	Feb	1	JAMMU AND KASHMIR	4	
4	2023-2024	Feb	1	JAMMU AND KASHMIR	622	
...	...	...	...	...	...	
14377	2024-2025	Feb	9	UTTAR PRADESH	172	
14378	2024-2025	Feb	9	UTTAR PRADESH	173	
14379	2024-2025	Feb	9	UTTAR PRADESH	176	
14380	2024-2025	Feb	9	UTTAR PRADESH	177	
14381	2024-2025	Feb	9	UTTAR PRADESH	181	

	district_name	scheme_code	total_beneficiaries	total_sc	total_st	\
0	SRINAGAR	IGNWPS	546	0	1	
1	UDHAMPUR	IGNWPS	353	85	17	

2	BADGAM	IGNWPS	176	3	1
3	DODA	IGNWPS	246	9	18
4	KULGAM	IGNWPS	290	20	0
...	...	...	...	...	...
14377	MUZAFFARNAGAR	IGNWPS	7601	1817	13
14378	PILIBHIT	IGNWPS	8268	1796	0
14379	RAMPUR	IGNWPS	23245	3169	0
14380	SAHARANPUR	IGNWPS	10667	3649	143
14381	SHRAVASTI	IGNWPS	6168	1689	332

	total_gen	total_obc	total_aadhar	total_mobilenos	lastUpdated	\
0	544	1	403	127	2024-07-31	
1	238	13	250	121	2024-07-31	
2	170	2	150	25	2024-07-31	
3	219	0	237	33	2024-07-31	
4	257	13	290	0	2024-07-31	
...	...	...	...	...	...	...
14377	2719	3052	3680	3177	2025-03-07	
14378	1756	4716	4037	4306	2025-03-07	
14379	7868	12208	9288	9096	2025-03-07	
14380	2940	3935	5896	5587	2025-03-07	
14381	2203	1944	1649	2063	2025-03-07	

	region_type
0	Union Territory
1	Union Territory
2	Union Territory
3	Union Territory
4	Union Territory
...	...
14377	State
14378	State
14379	State
14380	State
14381	State

[14382 rows x 16 columns]

Replaces month numbers with their short-form names (Jan, Feb, etc.) in the “mnth” column.

```
[33]: df.groupby('mnth')['total_beneficiaries'].sum()
```

```
[33]: mnth
Apr    15332724
Aug    7825327
Dec    7558180
Feb    15028134
```

```

Jan      7558954
Jul      15321763
Jun      15364095
Mar       7539522
May      15379860
Nov      14985987
Oct      14500548
Sep      14480493
Name: total_beneficiaries, dtype: int64

```

Sums the total beneficiaries for each month, grouped by the “mnth” column.

```
[35]: df.groupby('state_name')['total_aadhar'].sum().sort_values(ascending=False)
```

```

[35]: state_name
WEST BENGAL                14065667
ODISHA                    10168479
MADHYA PRADESH             9311717
BIHAR                     8334296
KARNATAKA                  7985203
GUJARAT                   7683545
TAMIL NADU                7587049
ANDHRA PRADESH            7432652
RAJASTHAN                 5535692
JHARKHAND                 5006173
KERALA                    4692877
UTTAR PRADESH             4148635
CHHATTISGARH              4001228
TELANGANA                 3729388
MAHARASHTRA               1743080
ASSAM                    1719339
HARYANA                   1653813
DELHI                     725728
PUNJAB                    216856
UTTARAKHAND              188706
HIMACHAL PRADESH          164740
TRIPURA                  151914
JAMMU AND KASHMIR         112615
MANIPUR                   106508
GOA                       103736
MEGHALAYA                 96931
ARUNACHAL PRADESH         76300
CHANDIGARH               49248
MIZORAM                   34335
PUDUCHERRY               26632
THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU 19068
NAGALAND                  7890

```

ANDAMAN AND NICOBAR	4806
LADAKH	1931
LAKSHADWEEP	1500
SIKKIM	0

Name: total\_aadhar, dtype: int64

Shows the total Aadhaar-linked beneficiaries per state, sorted in descending order.

```
[37]: df.groupby('district_name')['total_beneficiaries'].mean()
```

```
[37]: district_name
24 PARGANAS NORTH    74552.70
24 PARGANAS SOUTH    45113.00
ADILABAD             13699.00
AGAR MALWA           5769.00
AGRA                 4577.00
...
YADADRI              5540.00
YADGIR               1075.70
YAMUNANAGAR          8802.90
YAVATMAL             2044.05
ZUNHEBOTO            434.00
Name: total_beneficiaries, Length: 721, dtype: float64
```

Calculates the average number of beneficiaries for each district.

```
[39]: df.groupby('state_name')['lastUpdated'].max()
```

```
[39]: state_name
ANDAMAN AND NICOBAR    2025-03-07
ANDHRA PRADESH         2025-03-07
ARUNACHAL PRADESH      2025-03-07
ASSAM                  2025-03-07
BIHAR                  2025-03-07
CHANDIGARH             2025-03-07
CHHATTISGARH           2025-03-07
DELHI                  2025-03-07
GOA                    2025-03-07
GUJARAT                2025-03-07
HARYANA                2025-03-07
HIMACHAL PRADESH       2025-03-07
JAMMU AND KASHMIR      2025-03-07
JHARKHAND              2025-03-07
KARNATAKA              2025-03-07
KERALA                 2025-03-07
LADAKH                 2025-03-07
LAKSHADWEEP            2025-03-07
MADHYA PRADESH         2025-03-07
```

MAHARASHTRA	2025-03-07
MANIPUR	2025-03-07
MEGHALAYA	2025-03-07
MIZORAM	2025-03-07
NAGALAND	2025-03-07
ODISHA	2025-03-07
PUDUCHERRY	2025-03-07
PUNJAB	2025-03-07
RAJASTHAN	2025-03-07
SIKKIM	2025-03-07
TAMIL NADU	2025-03-07
TELANGANA	2025-03-07
THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU	2025-03-07
TRIPURA	2025-03-07
UTTAR PRADESH	2025-03-07
UTTARAKHAND	2025-03-07
WEST BENGAL	2025-03-07

Name: lastUpdated, dtype: object

Shows the most recent update date for each state.

```
[41]: df[df['total_mobilenos'] == 0]['state_name'].unique()
```

```
[41]: array(['JAMMU AND KASHMIR', 'ARUNACHAL PRADESH', 'NAGALAND',
        'HIMACHAL PRADESH', 'LADAKH', 'LAKSHADWEEP', 'CHANDIGARH',
        'SIKKIM', 'MANIPUR', 'PUDUCHERRY',
        'THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU', 'TRIPURA'],
        dtype=object)
```

Lists the unique states where no beneficiaries have mobile numbers.

```
[43]: df.
      ↳groupby('state_name')[['total_beneficiaries', 'total_sc', 'total_st', 'total_gen', 'total_obc']]
      ↳sum()
```

```
[43]:
```

state_name	total_beneficiaries	total_sc \
ANDAMAN AND NICOBAR	7970	0
ANDHRA PRADESH	11729503	2187771
ARUNACHAL PRADESH	120200	0
ASSAM	2290292	128180
BIHAR	12684420	2059880
CHANDIGARH	49715	0
CHHATTISGARH	4318904	346288
DELHI	729357	16317
GOA	117940	2080
GUJARAT	7762285	342774
HARYANA	2657668	586454



HIMACHAL PRADESH	439480	128600
JAMMU AND KASHMIR	138212	9577
JHARKHAND	5084571	710744
KARNATAKA	8196100	741149
KERALA	7940714	347804
LADAKH	8880	0
LAKSHADWEEP	1500	0
MADHYA PRADESH	16521580	3171540
MAHARASHTRA	1993335	102123
MANIPUR	149261	3449
MEGHALAYA	164478	1075
MIZORAM	46401	40
NAGALAND	84820	20
ODISHA	10415590	1229465
PUDUCHERRY	199180	30040
PUNJAB	379420	23220
RAJASTHAN	6585102	1538505
SIKKIM	29380	2420
TAMIL NADU	10181512	2472065
TELANGANA	4017940	842420
THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU	98200	3860
TRIPURA	346542	62202
UTTAR PRADESH	20504720	5720240
UTTARAKHAND	477974	97223
WEST BENGAL	14402441	3977488

	total_st	total_gen	total_obc
state_name			
ANDAMAN AND NICOBAR	2270	5700	0
ANDHRA PRADESH	891352	2618222	6032158
ARUNACHAL PRADESH	119080	1100	20
ASSAM	249472	1669274	243366
BIHAR	199880	7953940	2470720
CHANDIGARH	0	49715	0
CHHATTISGARH	1156282	1638101	1178233
DELHI	163	709303	3574
GOA	30380	61400	24080
GUJARAT	1045214	4496532	1876012
HARYANA	0	1552919	518295
HIMACHAL PRADESH	24180	230260	56440
JAMMU AND KASHMIR	8091	112759	7785
JHARKHAND	1407118	1264247	1702462
KARNATAKA	329900	1944622	5180429
KERALA	145506	7447404	0
LADAKH	8858	22	0
LAKSHADWEEP	660	840	0
MADHYA PRADESH	4830140	1702800	6817100

MAHARASHTRA	293757	1427562	169893
MANIPUR	34430	77740	33642
MEGHALAYA	145642	16874	887
MIZORAM	43685	2676	0
NAGALAND	84480	120	200
ODISHA	2304825	5022703	1858597
PUDUCHERRY	740	160920	7480
PUNJAB	160	350620	5420
RAJASTHAN	1574271	1242608	2229718
SIKKIM	12140	8540	6280
TAMIL NADU	105123	2628970	4975354
TELANGANA	545160	277280	2353080
THE DADRA AND NAGAR HAVELI AND DAMAN AND DIU	42460	15320	36560
TRIPURA	88032	138674	57634
UTTAR PRADESH	115640	4700240	9968600
UTTARAKHAND	15240	324606	40905
WEST BENGAL	1027654	9120798	256592

Sums the total beneficiaries and their category-wise breakdown (SC, ST, Gen, OBC) for each state.

```
[45]: df.groupby('fin_year')['total_beneficiaries'].sum().sort_index()
```

```
[45]: fin_year
2023-2024    90896320
2024-2025    59979267
Name: total_beneficiaries, dtype: int64
```

Shows the total beneficiaries for each financial year, sorted in chronological order.

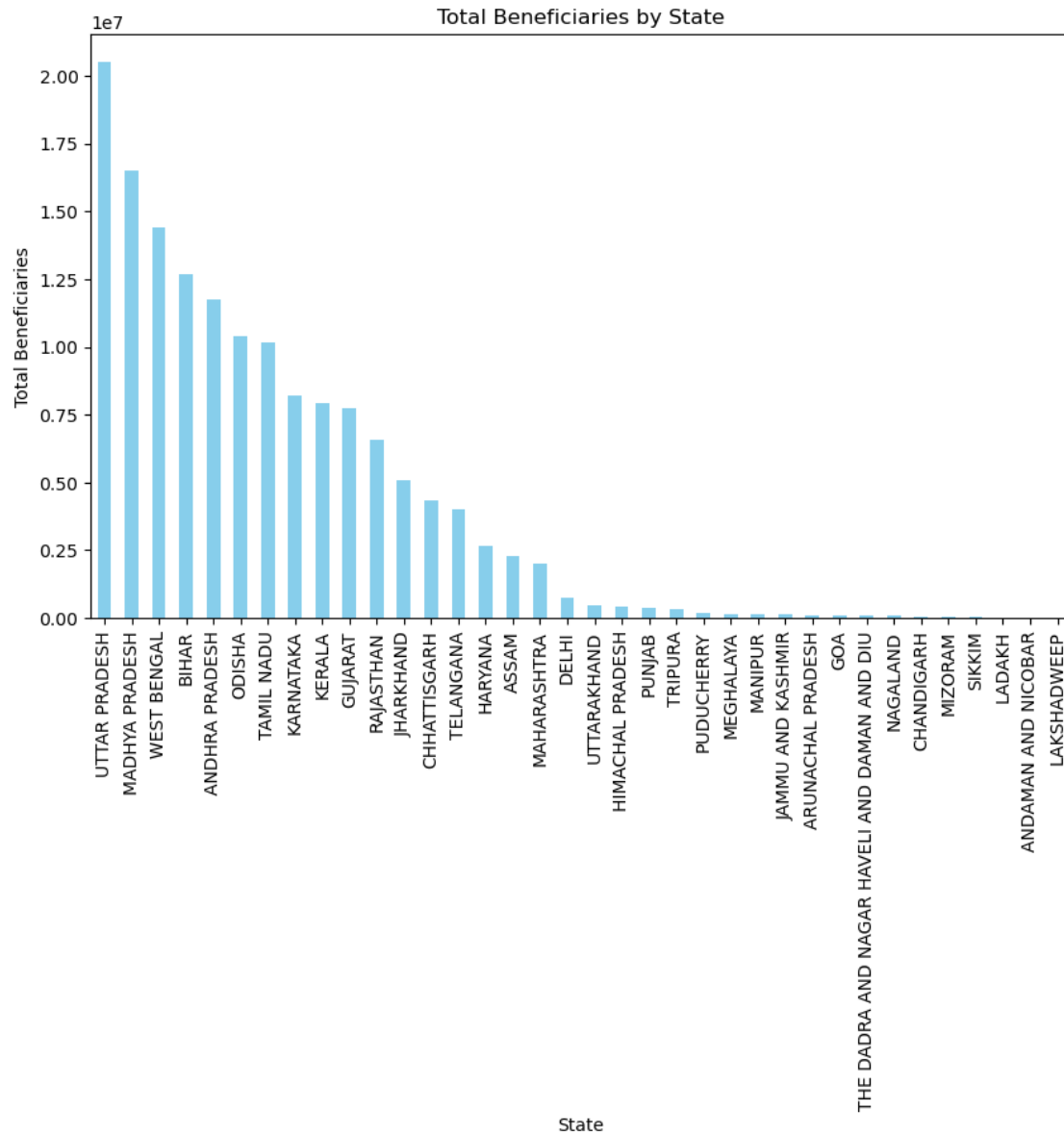
```
[47]: df.groupby('scheme_code')['total_beneficiaries'].sum().
      ↪sort_values(ascending=False).head(5)
```

```
[47]: scheme_code
IGNWPS    150875587
Name: total_beneficiaries, dtype: int64
```

Displays the top 5 schemes with the highest total number of beneficiaries.

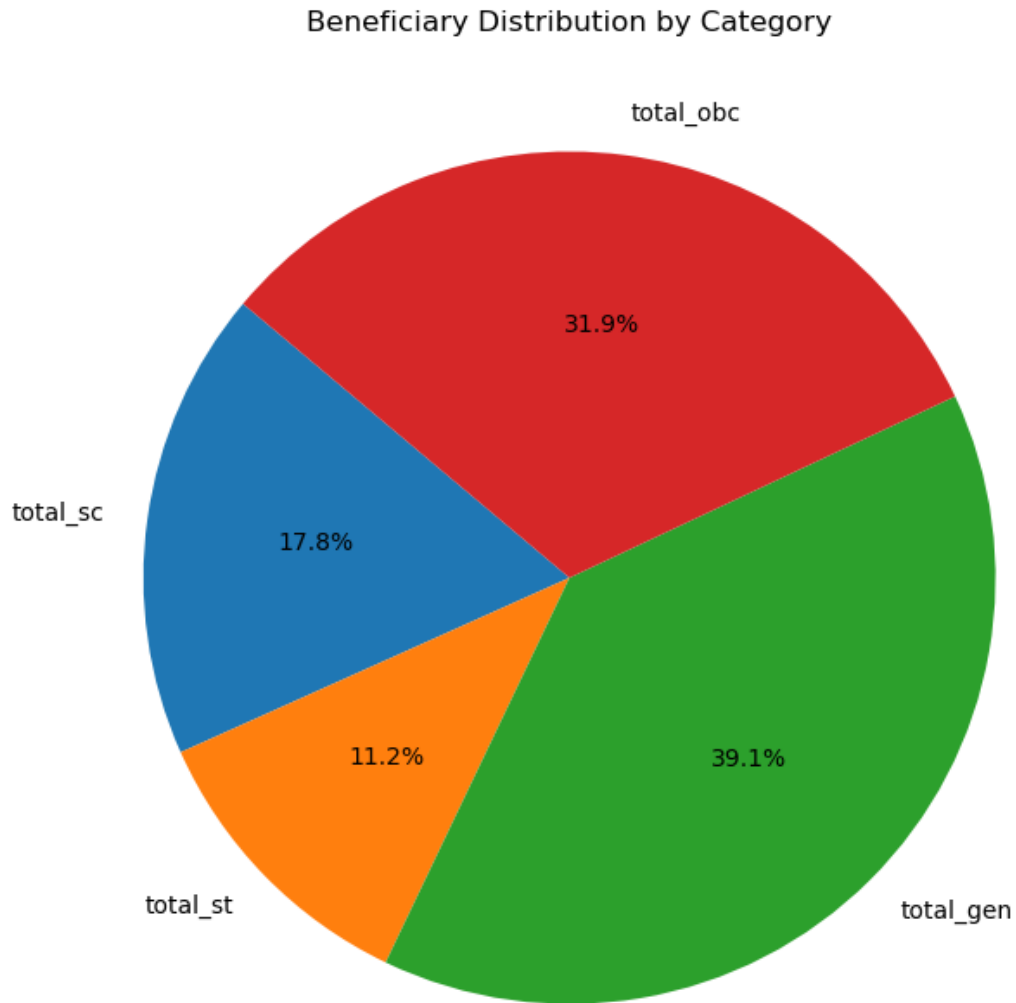
```
[49]: import matplotlib.pyplot as plt

statewise = df.groupby('state_name')['total_beneficiaries'].sum().
      ↪sort_values(ascending=False)
statewise.plot(kind='bar', figsize=(10, 6), title='Total Beneficiaries by_
      ↪State', color='skyblue')
plt.xlabel('State')
plt.ylabel('Total Beneficiaries')
plt.show()
```



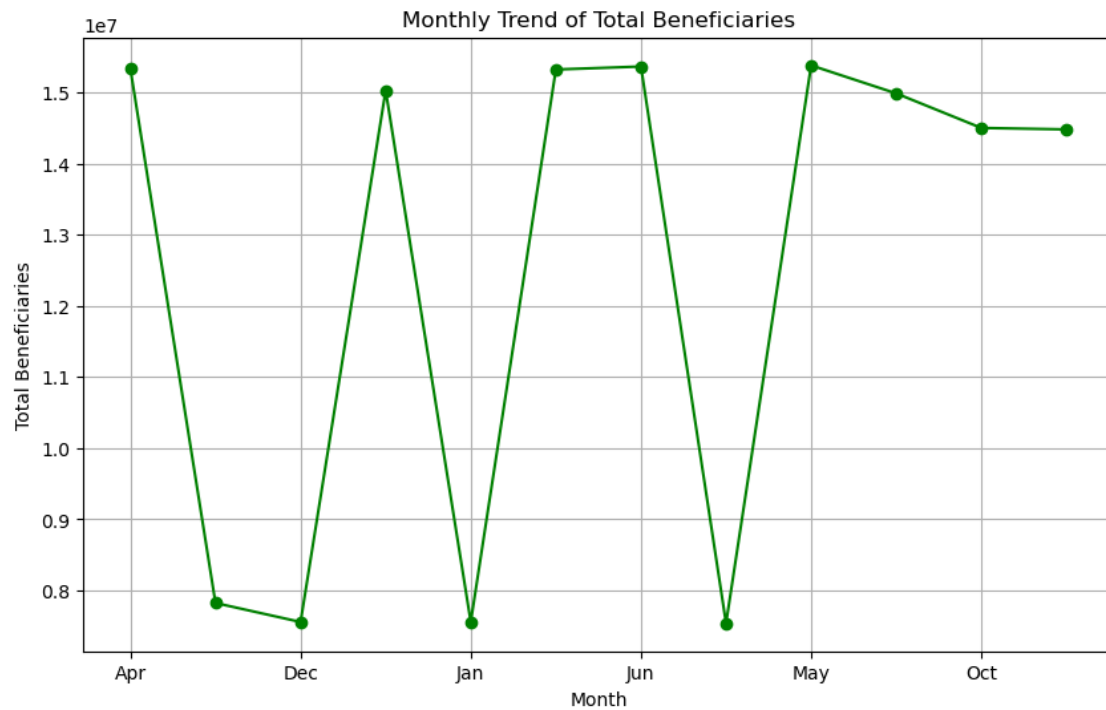
Creates a bar chart showing total beneficiaries by state, sorted in descending order.

```
[51]: category_totals = df[['total_sc', 'total_st', 'total_gen', 'total_obc']].sum()
category_totals.plot(kind='pie', autopct='%1.1f%%', startangle=140, figsize=(8,8), title='Beneficiary Distribution by Category')
plt.show()
```



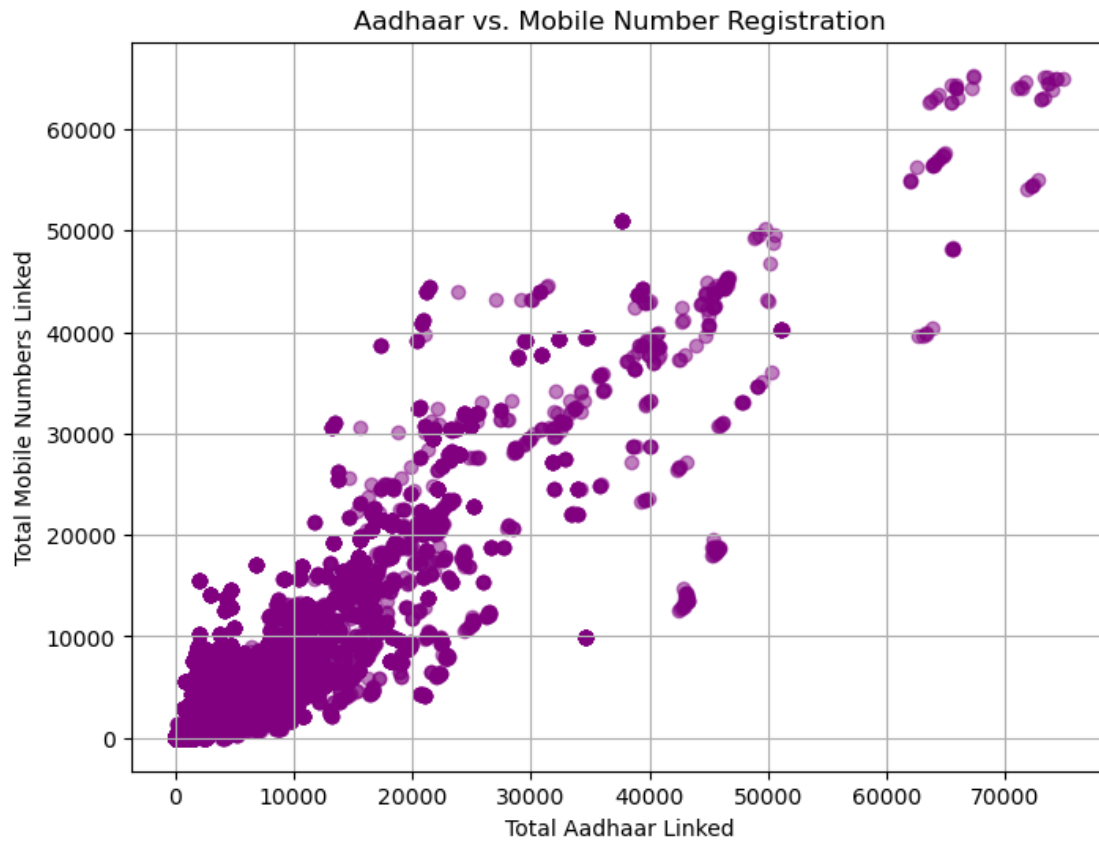
Creates a pie chart showing the percentage distribution of beneficiaries across different categories (SC, ST, Gen, OBC).

```
[53]: monthly_trend = df.groupby('mnth')['total_beneficiaries'].sum()
monthly_trend.plot(kind='line', marker='o', figsize=(10, 6), title='Monthly_
↳Trend of Total Beneficiaries', color='green')
plt.xlabel('Month')
plt.ylabel('Total Beneficiaries')
plt.grid(True)
plt.show()
```



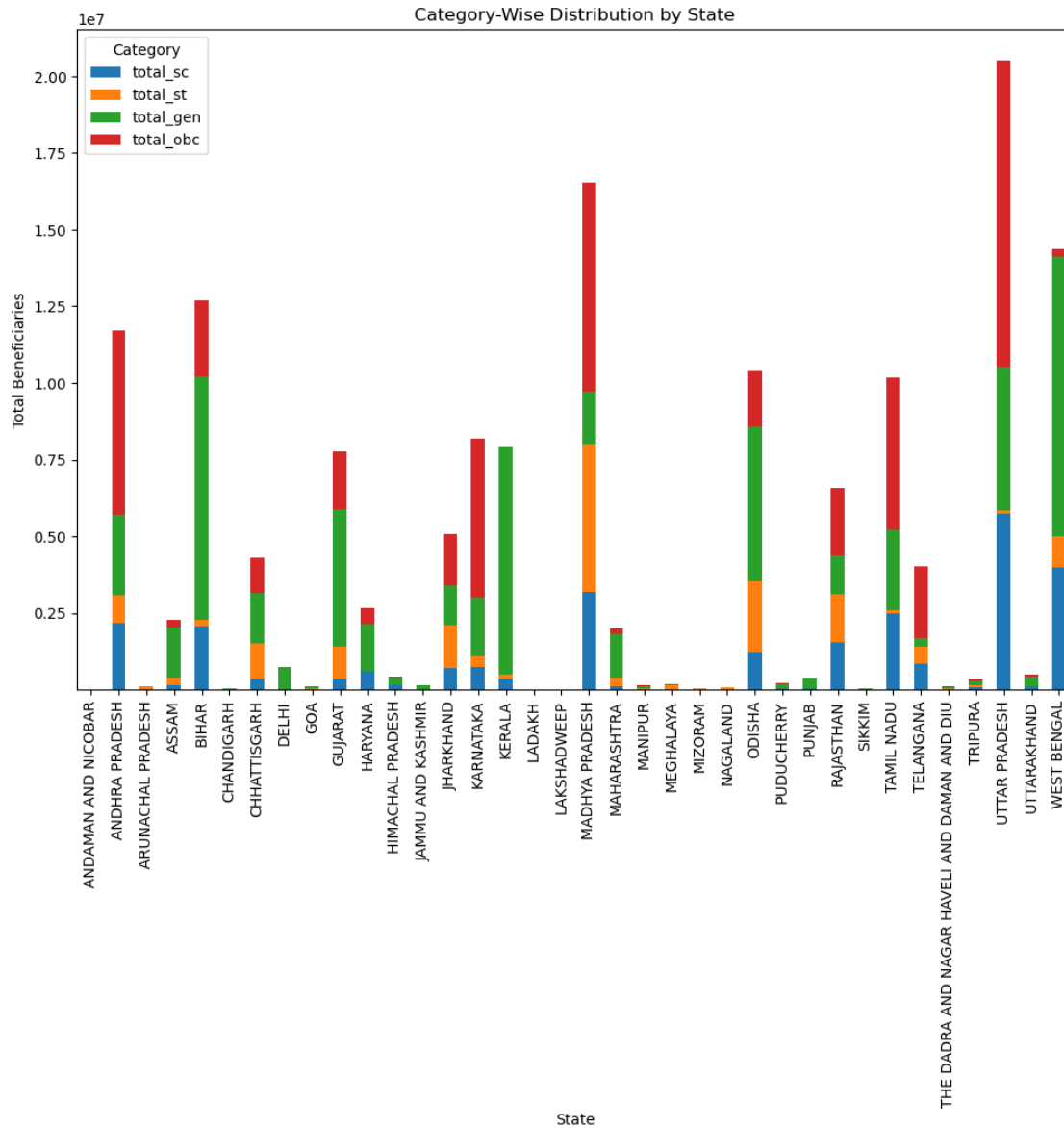
Plots a line chart showing the monthly trend of total beneficiaries, with markers and grid for clarity.

```
[55]: plt.figure(figsize=(8, 6))
plt.scatter(df['total_aadhar'], df['total_mobilenos'], alpha=0.5, color='purple')
plt.xlabel('Total Aadhaar Linked')
plt.ylabel('Total Mobile Numbers Linked')
plt.title('Aadhaar vs. Mobile Number Registration')
plt.grid(True)
plt.show()
```



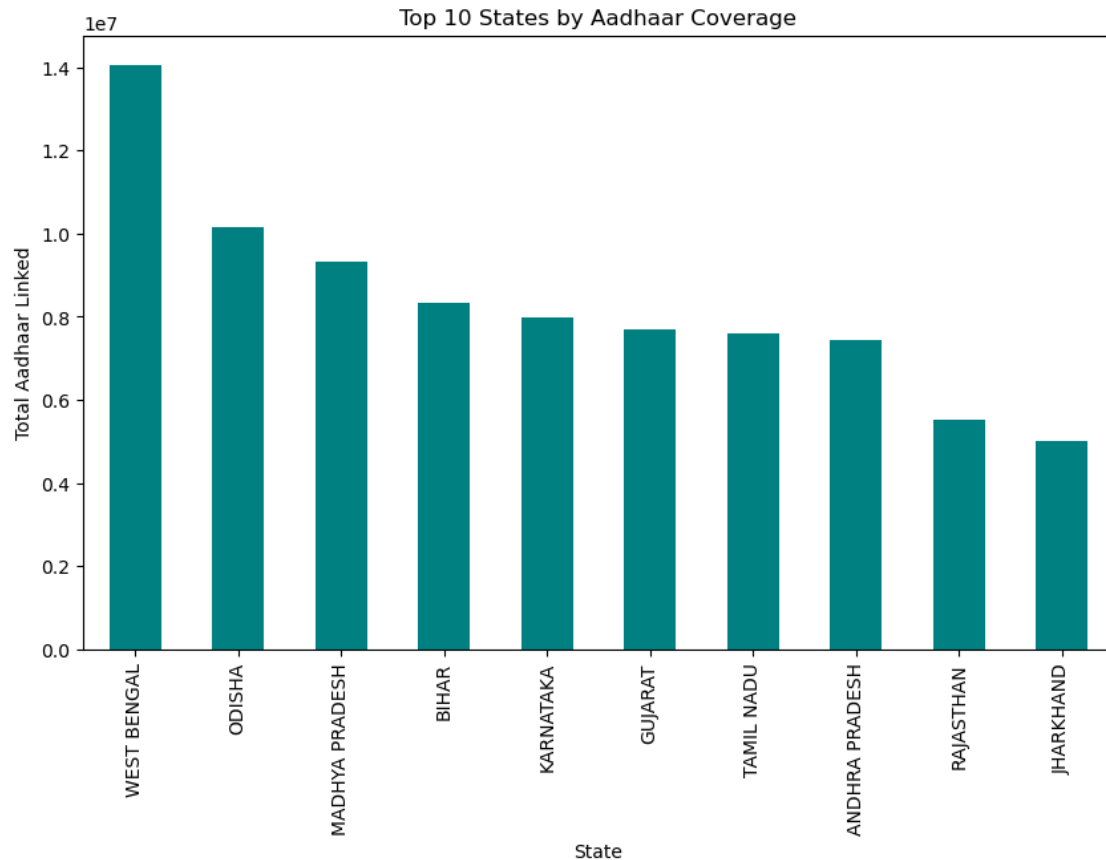
Creates a scatter plot showing the relationship between Aadhaar-linked and mobile-linked beneficiaries.

```
[57]: statewise_category = df.groupby('state_name')[['total_sc', 'total_st',
    ↳ 'total_gen', 'total_obc']].sum()
statewise_category.plot(kind='bar', stacked=True, figsize=(12, 8),
    ↳ title='Category-Wise Distribution by State')
plt.xlabel('State')
plt.ylabel('Total Beneficiaries')
plt.legend(title='Category')
plt.show()
```



Creates a stacked bar chart showing the category-wise distribution of beneficiaries for each state.

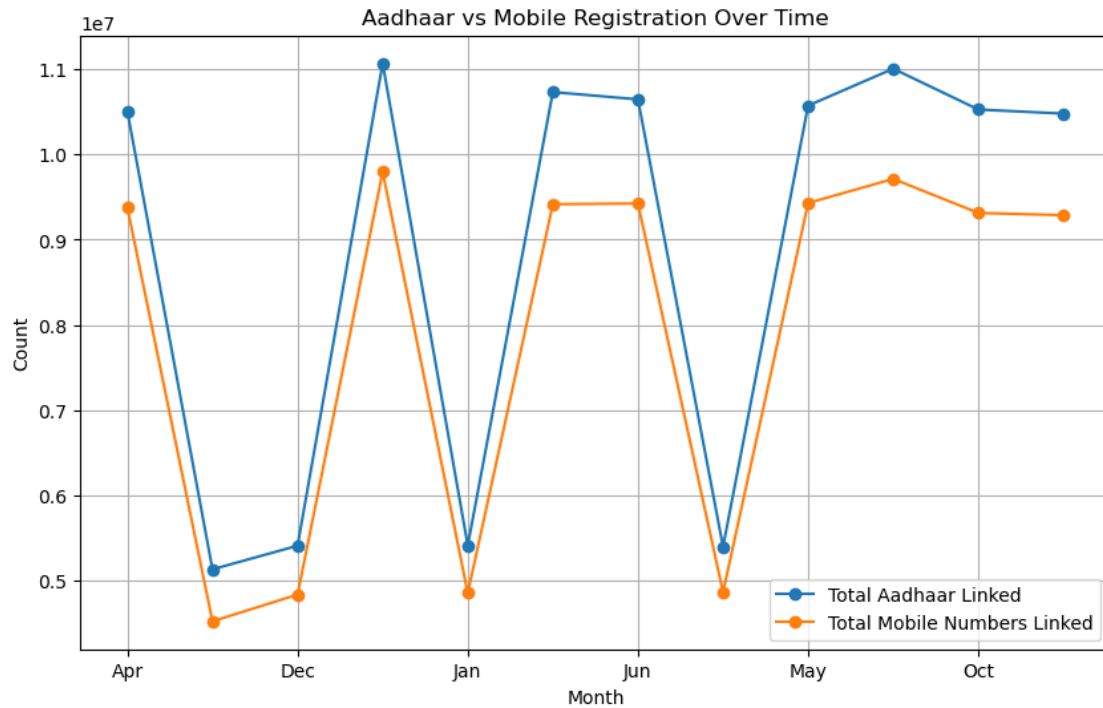
```
[59]: top_aadhar = df.groupby('state_name')['total_aadhar'].sum().
      ↪sort_values(ascending=False).head(10)
top_aadhar.plot(kind='bar', color='teal', figsize=(10, 6), title='Top 10 States_
      ↪by Aadhaar Coverage')
plt.xlabel('State')
plt.ylabel('Total Aadhaar Linked')
plt.show()
```



Displays a bar chart of the top 10 states with the highest total Aadhaar-linked beneficiaries.

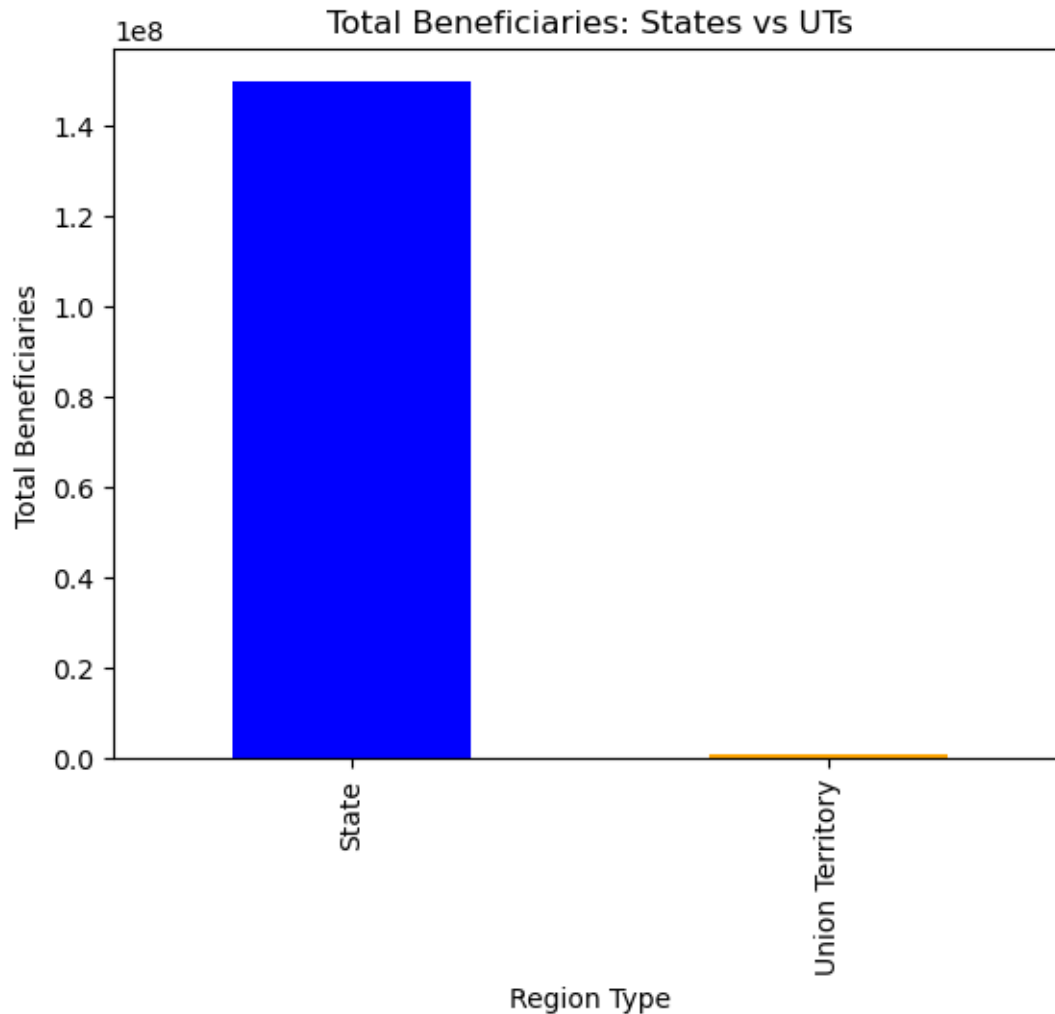
```
[61]: monthly_totals = df.groupby('mnth')[['total_aadhar', 'total_mobileno']].sum()
monthly_totals.plot(kind='line', marker='o', figsize=(10, 6), title='Aadhaar vs Mobile Registration Over Time')
plt.xlabel('Month')
plt.ylabel('Count')
plt.legend(['Total Aadhaar Linked', 'Total Mobile Numbers Linked'])
plt.grid(True)
plt.show()
```





Plots a line chart comparing Aadhaar and mobile registrations over the months, showing trends over time.

```
[63]: state_ut_totals = df.groupby('region_type')['total_beneficiaries'].sum() # Assuming 'region_type' distinguishes State/UT
state_ut_totals.plot(kind='bar', color=['blue', 'orange'], title='Total Beneficiaries: States vs UTs')
plt.xlabel('Region Type')
plt.ylabel('Total Beneficiaries')
plt.show()
```



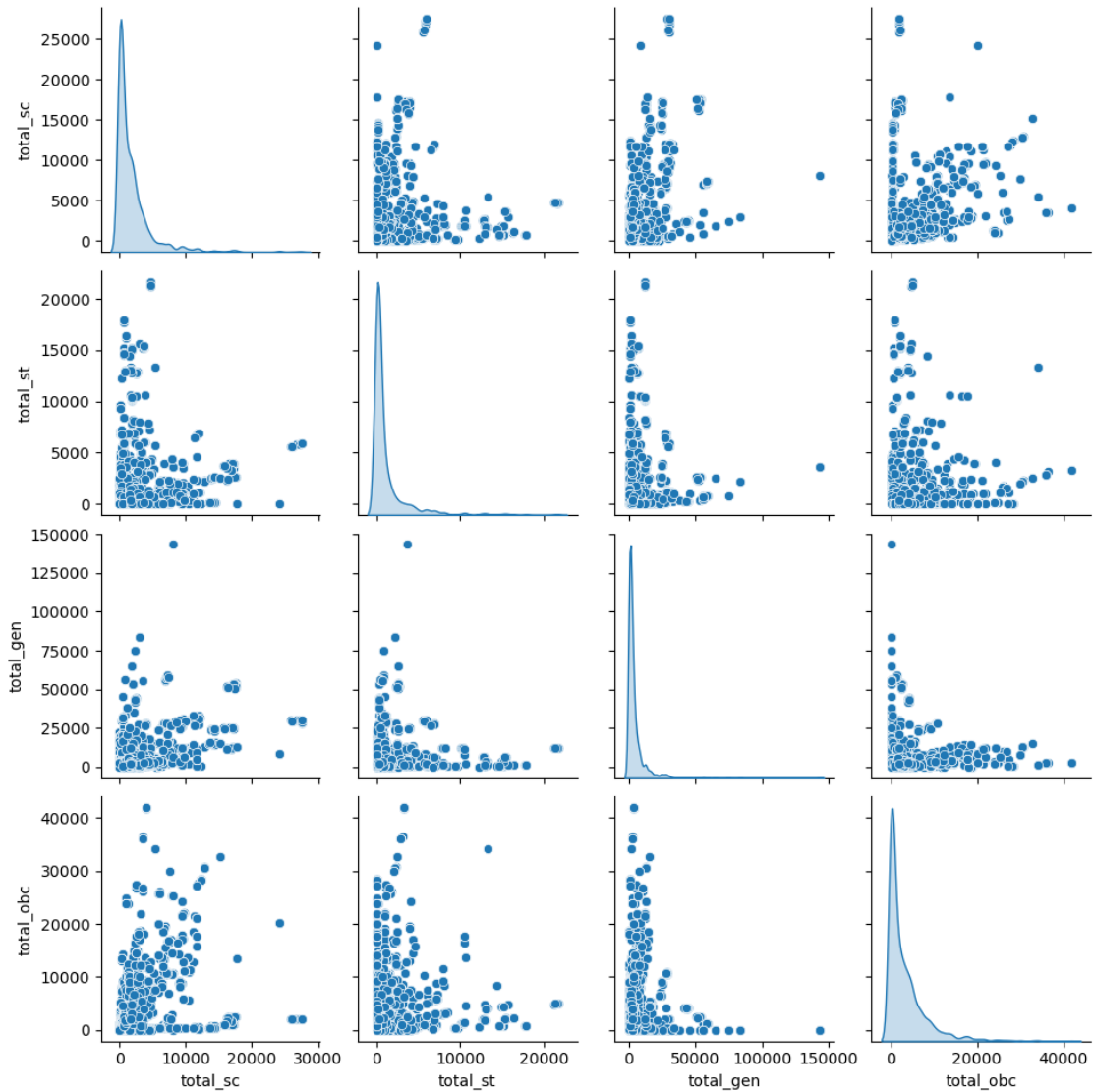
Creates a bar chart comparing total beneficiaries between states and union territories.

```
[70]: import seaborn as sns
sns.pairplot(df[['total_sc', 'total_st', 'total_gen', 'total_obc']],
            diag_kind='kde', palette='husl')
plt.suptitle("Pairwise Relationships Between Beneficiary Categories", y=1.02)
plt.show()
```

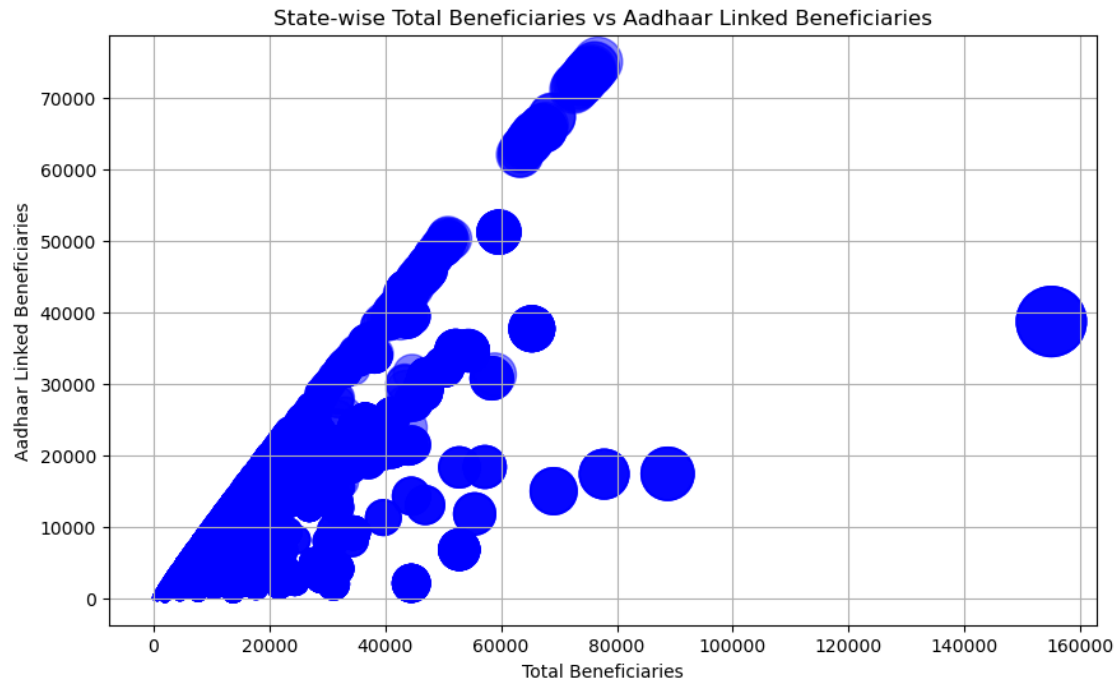
```
C:\Users\Dell\anaconda3\Lib\site-packages\seaborn\axisgrid.py:1513: UserWarning:
Ignoring `palette` because no `hue` variable has been assigned.
, func(x=vector, **plot_kwarg)
,C:\Users\Dell\anaconda3\Lib\site-packages\seaborn\axisgrid.py:1513:
UserWarning: Ignoring `palette` because no `hue` variable has been assigned.
, func(x=vector, **plot_kwarg)
,C:\Users\Dell\anaconda3\Lib\site-packages\seaborn\axisgrid.py:1513:
UserWarning: Ignoring `palette` because no `hue` variable has been assigned.
```



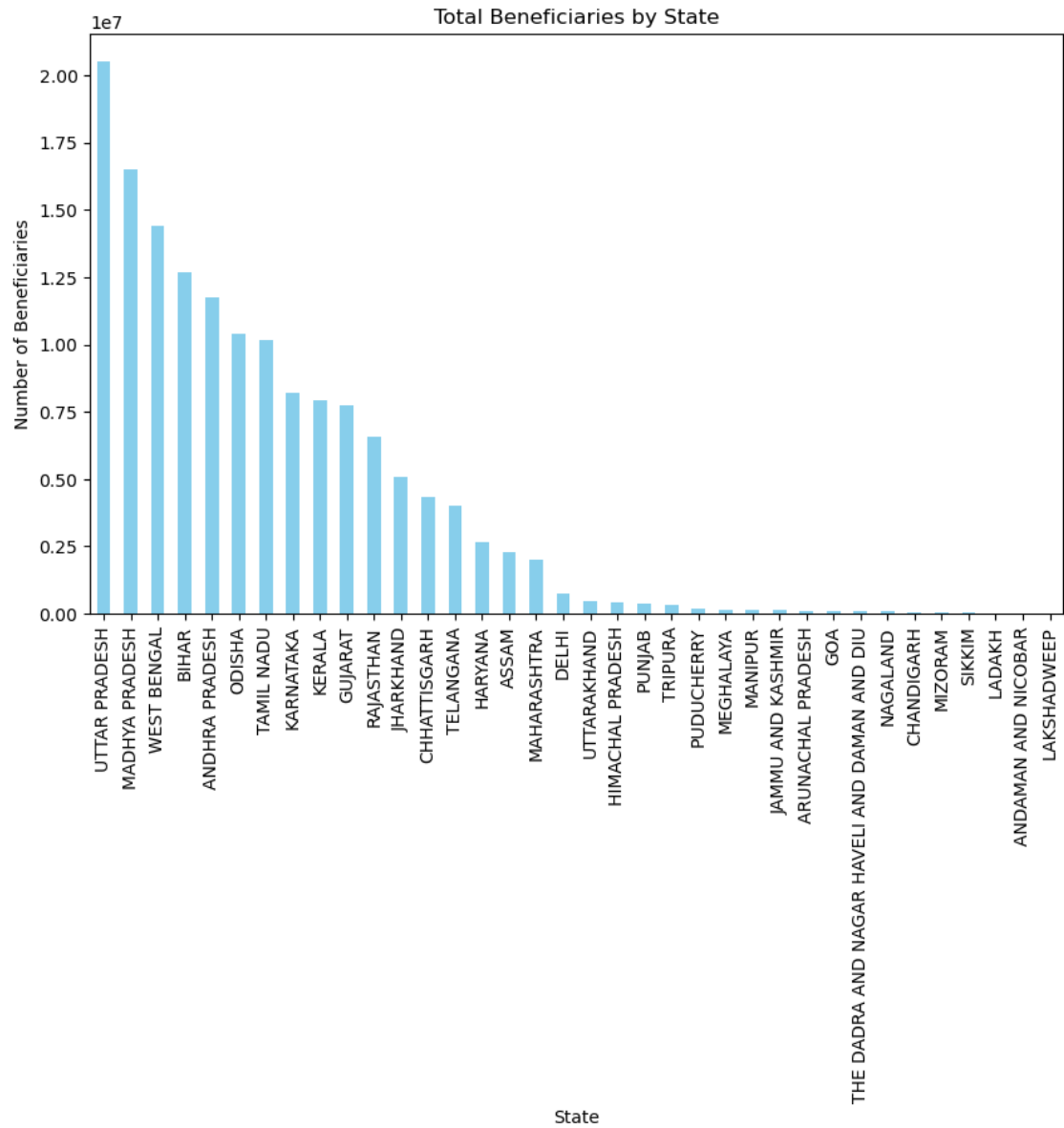
Pairwise Relationships Between Beneficiary Categories



```
[66]: plt.figure(figsize=(10, 6))
plt.scatter(df['total_beneficiaries'], df['total_aadhar'],
            s=df['total_beneficiaries'] / 100, alpha=0.5, color='blue')
plt.title("State-wise Total Beneficiaries vs Aadhaar Linked Beneficiaries")
plt.xlabel("Total Beneficiaries")
plt.ylabel("Aadhaar Linked Beneficiaries")
plt.grid(True)
plt.show()
```



```
[72]: import matplotlib.pyplot as plt
state_beneficiaries = df.groupby('state_name')['total_beneficiaries'].sum()
plt.figure(figsize=(10, 6))
state_beneficiaries.sort_values(ascending=False).plot(kind='bar',
    color='skyblue')
plt.title("Total Beneficiaries by State")
plt.xlabel("State")
plt.ylabel("Number of Beneficiaries")
plt.xticks(rotation=90)
plt.show()
```



[ ]:

- 6 Confirmed 14,382 rows, 15 columns, no missing values.
- 7 Listed the complete set of states, confirming coverage across multiple regions.
- 8 Showed which states had more records, indicating regions with higher data granularity or reporting frequency.
- 9 Highlighted the leading states in terms of outreach and those with lower coverage, pointing to potential gaps.
- 10 Showed that certain states had significantly higher Aadhaar penetration, indicating better digital integration.
- 11 Identified districts with the highest outreach, potentially reflecting higher population density or better program implementation
- 12 Revealed monthly patterns — useful for detecting seasonal impacts or administrative trends.
- 13 Confirmed when each state's data was last refreshed, helping to identify stale data sources.
- 14 Created new column name region type as we divided it into union territories and states.

[ ]: