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sol.py

1 # =====
2 # Aadhaar Analytics - Full Graph Generation Code (Attractive)
3 # Covers: univariate, bivariate, trivariate, multivariate visuals
4 # Uses: pandas, numpy, matplotlib, scikit-learn
5 # NOTE: No seaborn, and no manual color setting.
6 # =====
7
8 import os
9 import numpy as np
10 import pandas as pd
11 import matplotlib.pyplot as plt
12
13 from sklearn.preprocessing import StandardScaler
14 from sklearn.decomposition import PCA
15 from sklearn.cluster import KMeans
16
17 # -----
18 # CONFIG
19 # -----
20 ENROL_PATH = "api_data_aadhar_enrolment_0_1006029.csv"
21 DEMO_PATH = "api_data_aadhar_demographic_0_2071700.csv"
22 BIO_PATH = "api_data_aadhar_biometric_0_1861108.csv"
23
24 SAVE_FIGS = False
25 OUTDIR = "/mnt/data/aadhaar_figs"
26 os.makedirs(OUTDIR, exist_ok=True)
27
28 plt.rcParams["figure.figsize"] = (12, 5)
29 plt.rcParams["axes.grid"] = True
30
31 def savefig(name: str):
32     if SAVE_FIGS:
33         fp = os.path.join(OUTDIR, name)
34         plt.savefig(fp, dpi=220, bbox_inches="tight")
35         print("Saved:", fp)
36
37 # -----
38 # LOAD + CLEAN
39 # -----
40 def load_uidai_csv(path: str) -> pd.DataFrame:
41     df = pd.read_csv(path)
42     if "Unnamed: 0" in df.columns:
43         df = df.drop(columns=["Unnamed: 0"])
44     df.columns = [c.strip() for c in df.columns]
45     if "date" not in df.columns:
46         raise ValueError(f"Missing 'date' column in {path}")
47     df["date"] = pd.to_datetime(df["date"], dayfirst=True, errors="coerce")
48     # strip common geo fields if present
49     for c in ["state", "district", "pincode"]:
50         if c in df.columns:
51             df[c] = df[c].astype(str).str.strip()
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52     return df
53
54 enrol = load_uidai_csv(ENROL_PATH)
55 demo  = load_uidai_csv(DEMO_PATH)
56 bio   = load_uidai_csv(BIO_PATH)
57
58 # -----
59 # FEATURE ENGINEERING
60 # -----
61 # totals
62 enrol["total_enrol"] = enrol[["age_0_5", "age_5_17",
63     "age_18_greater"]].apply(pd.to_numeric, errors="coerce").fillna(0).sum(axis=1)
64 # demo/bio columns vary a bit; sum all columns that start with demo_ / bio_
65 demo_cols = [c for c in demo.columns if c.startswith("demo_")]
66 bio_cols  = [c for c in bio.columns  if c.startswith("bio_")]
67
68 demo["total_demo"] = demo[demo_cols].apply(pd.to_numeric,
69     errors="coerce").fillna(0).sum(axis=1)
70 bio["total_bio"]   = bio[bio_cols].apply(pd.to_numeric,
71     errors="coerce").fillna(0).sum(axis=1)
72
73 eps = 1e-9
74
75 # -----
76 # INDIA DAILY TABLE
77 # -----
78 daily = (
79     enrol.groupby("date")["total_enrol"].sum().to_frame("Enrolments")
80     .join(demo.groupby("date")["total_demo"].sum())
81     .join(bio.groupby("date")["total_bio"].sum())
82     .fillna(0)
83 )
84 daily.rename(columns={"total_demo": "Demographic Updates", "total_bio": "Biometric Updates"}, inplace=True)
85 daily["Total Updates"] = daily["Demographic Updates"] + daily["Biometric Updates"]
86 daily["Total Activity"] = daily["Enrolments"] + daily["Total Updates"]
87 daily["Enrol_7d_avg"] = daily["Enrolments"].rolling(7, min_periods=3).mean()
88 daily["Upd_7d_avg"] = daily["Total Updates"].rolling(7, min_periods=3).mean()
89 daily["Act_7d_avg"] = daily["Total Activity"].rolling(7, min_periods=3).mean()
90 daily["is_month_start"] = daily.index.is_month_start
91
92 # =====
93 # 1) Time-series: Daily enrolments + demo + bio updates
94 # (supports: update-dominant lifecycle insight)
95 plt.figure()
96 plt.plot(daily.index, daily["Enrolments"], label="Enrolments")
97 plt.plot(daily.index, daily["Demographic Updates"], label="Demographic Updates")
98 plt.plot(daily.index, daily["Biometric Updates"], label="Biometric Updates")
99 plt.title("India - Daily Aadhaar Enrolments and Updates")
100 plt.xlabel("Date")
101 plt.ylabel("Transactions")
102 plt.legend()

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102 plt.tight_layout()
103 savefig("01_india_daily_enrol_demo_bio.png")
104 plt.show()
105
106 # =====
107 # 2) Time-series: Enrolments vs Total Updates vs Total Activity
108 # =====
109 plt.figure()
110 plt.plot(daily.index, daily["Enrolments"], label="Enrolments")
111 plt.plot(daily.index, daily["Total Updates"], label="Total Updates")
112 plt.plot(daily.index, daily["Total Activity"], label="Total Activity")
113 plt.title("India – Enrolments vs Updates vs Total Activity")
114 plt.xlabel("Date")
115 plt.ylabel("Transactions")
116 plt.legend()
117 plt.tight_layout()
118 savefig("02_india_enrol_vs_updates_vs_activity.png")
119 plt.show()
120
121 # =====
122 # 3) Rolling averages (trend vs noise)
123 # =====
124 plt.figure()
125 plt.plot(daily.index, daily["Enrol_7d_avg"], label="Enrolments (7d avg)")
126 plt.plot(daily.index, daily["Upd_7d_avg"], label="Updates (7d avg)")
127 plt.plot(daily.index, daily["Act_7d_avg"], label="Activity (7d avg)")
128 plt.title("India – 7-Day Rolling Averages (Trend)")
129 plt.xlabel("Date")
130 plt.ylabel("Transactions")
131 plt.legend()
132 plt.tight_layout()
133 savefig("03_india_rolling_7d.png")
134 plt.show()
135
136 # =====
137 # 4) Annotated: Month-start enrolment spikes
138 # =====
139 plt.figure()
140 plt.plot(daily.index, daily["Enrolments"], label="Daily Enrolments")
141 ms = daily[daily["is_month_start"]]
142 plt.scatter(ms.index, ms["Enrolments"], label="1st of Month")
143 plt.title("India – Month-start Enrolment Spikes (Administrative Cycle)")
144 plt.xlabel("Date")
145 plt.ylabel("Enrolments")
146 plt.legend()
147 plt.tight_layout()
148 savefig("04_month_start_spikes.png")
149 plt.show()
150
151 # =====
152 # 5) Anomaly plot (global z-score)
153 # (supports: anomaly periods for operational interpretation)
154 # =====
155 s = daily["Enrolments"].astype(float)
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156 mu, sigma = s.mean(), s.std(ddof=0)
157 sigma = sigma if (sigma and not np.isnan(sigma)) else 1.0
158 z = (s - mu) / sigma
159 anom = z.abs() >= 3.0
160
161 plt.figure()
162 plt.plot(daily.index, s, label="Daily Enrolments")
163 plt.scatter(daily.index[anom], s[anom], label="Anomaly (|z|≥3)")
164 plt.title("India – Enrolment Anomalies ")
165 plt.xlabel("Date")
166 plt.ylabel("Enrolments")
167 plt.legend()
168 plt.tight_layout()
169 savefig("05_anomaly_global_z.png")
170 plt.show()
171
172 # =====
173 # 6) Age-wise enrolments (bar)
174 # =====
175 age_totals = enrol[["age_0_5", "age_5_17", "age_18_greater"]].apply(pd.to_numeric,
176 errors="coerce").fillna(0).sum()
177 plt.figure()
178 plt.bar(["0-5", "5-17", "18+"], age_totals.values)
179 plt.title("India – Age-wise Aadhaar Enrolments")
180 plt.xlabel("Age Group")
181 plt.ylabel("Total Enrolments")
182 plt.tight_layout()
183 savefig("06_agewise_enrol_bar.png")
184 plt.show()
185
186 # =====
187 # 7) Age-wise share (pie)
188 # =====
189 plt.figure()
190 plt.pie(age_totals.values, labels=["0-5", "5-17", "18+"], autopct="%1.1f%%")
191 plt.title("India – Age-wise Aadhaar Enrolment Share")
192 plt.tight_layout()
193 savefig("07_agewise_enrol_pie.png")
194 plt.show()
195
196 # =====
197 # STATE-LEVEL TABLES (for regional variation, burden, volatility)
198 # =====
199 state = (
200     enrol.groupby("state")["total_enrol"].sum().to_frame("Enrol")
201     .join(demo.groupby("state")["total_demo"].sum())
202     .join(bio.groupby("state")["total_bio"].sum())
203     .fillna(0)
204 )
205 state.rename(columns={"total_demo": "Demo", "total_bio": "Bio"}, inplace=True)
206 state["Updates"] = state["Demo"] + state["Bio"]
207 state["Total Activity"] = state["Enrol"] + state["Updates"]
208 state["Update Burden Index"] = state["Updates"] / (state["Enrol"] + eps)
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209 # =====
210 # 8) Top states by total enrolments (bar ranking)
211 # =====
212 top_enrol_states = state.sort_values("Enrol", ascending=False).head(10)
213 plt.figure()
214 plt.bar(top_enrol_states.index, top_enrol_states["Enrol"])
215 plt.title("Top 10 States – Total Aadhaar Enrolments")
216 plt.xlabel("State")
217 plt.ylabel("Enrolments")
218 plt.xticks(rotation=45, ha="right")
219 plt.tight_layout()
220 savefig("08_top10_states_enrol.png")
221 plt.show()
222
223 # =====
224 # 9) Top states: share of total Aadhaar activity (pie)
225 # =====
226 top5_activity = state.sort_values("Total Activity", ascending=False).head(5)
227 plt.figure()
228 plt.pie(top5_activity["Total Activity"].values, labels=top5_activity.index,
229 autopct="%1.1f%%")
230 plt.title("Top 5 States – Share of Total Aadhaar Activity")
231 plt.tight_layout()
232 savefig("09_top5_state_activity_share.png")
233 plt.show()
234
235 # =====
236 # 10) Update Burden Index (bar ranking)
237 # =====
238 top_burden = state.sort_values("Update Burden Index", ascending=False).head(10)
239 plt.figure()
240 plt.bar(top_burden.index, top_burden["Update Burden Index"])
241 plt.title("Top 10 States – Update Burden Index (Updates / Enrolments)")
242 plt.xlabel("State")
243 plt.ylabel("Updates per Enrolment")
244 plt.xticks(rotation=45, ha="right")
245 plt.tight_layout()
246 savefig("10_top10_update_burden.png")
247 plt.show()
248
249 # =====
250 # 11) Volatility Score (std/mean of DAILY activity at state level)
251 # =====
252 # build state-day activity (enrol + updates) to compute volatility
253 state_day_enrol = enrol.groupby(["state", "date"])["total_enrol"].sum().reset_index()
254 state_day_demo = demo.groupby(["state", "date"])["total_demo"].sum().reset_index()
255 state_day_bio = bio.groupby(["state", "date"])["total_bio"].sum().reset_index()
256
257 state_day = (
258     state_day_enrol.merge(state_day_demo, on=["state", "date"], how="left")
259         .merge(state_day_bio, on=["state", "date"], how="left")
260         .fillna(0)
261 )
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261 state_day["activity"] = state_day["total_enrol"] + state_day["total_demo"] +
state_day["total_bio"]
262
263 vol = state_day.groupby("state")["activity"].agg(["mean", "std"]).fillna(0)
264 vol["Volatility Score"] = vol["std"] / (vol["mean"] + eps)
265
266 top_vol = vol.sort_values("Volatility Score", ascending=False).head(10)
267 plt.figure()
268 plt.bar(top_vol.index, top_vol["Volatility Score"])
269 plt.title("Top 10 States – Volatility Score (Std/Mean of Daily Activity)")
270 plt.xlabel("State")
271 plt.ylabel("Volatility Score")
272 plt.xticks(rotation=45, ha="right")
273 plt.tight_layout()
274 savefig("11_top10_volatility.png")
275 plt.show()
276
277 # =====
278 # 12) Stacked bars: Top states × age group (trivariate)
279 # =====
280 state_age = enrol.groupby("state")[["age_0_5", "age_5_17", "age_18_greater"]].apply(
281     lambda x: x.apply(pd.to_numeric, errors="coerce").fillna(0).sum()
282 )
283 state_age["total"] = state_age.sum(axis=1)
284 state_age = state_age.sort_values("total", ascending=False).head(10)
285
286 x = np.arange(len(state_age.index))
287 plt.figure(figsize=(12,5))
288 plt.bar(x, state_age["age_0_5"].values, label="0-5")
289 plt.bar(x, state_age["age_5_17"].values, bottom=state_age["age_0_5"].values, label="5-17")
290 plt.bar(x, state_age["age_18_greater"].values, bottom=
(state_age["age_0_5"]+state_age["age_5_17"]).values, label="18+")
291 plt.title("Top 10 States – Enrolment by Age Group ")
292 plt.xlabel("State")
293 plt.ylabel("Enrolments")
294 plt.xticks(x, state_age.index, rotation=45, ha="right")
295 plt.legend()
296 plt.tight_layout()
297 savefig("12_stacked_age_top10.png")
298 plt.show()
299
300 # =====
301 # 13) Trivariate heatmap: State × Month × Update/Enrol ratio
302 # (use Total Updates (demo+bio) / Enrolments)
303 # =====
304 enrol_m = enrol.copy()
305 demo_m = demo.copy()
306 bio_m = bio.copy()
307
308 enrol_m["month"] = enrol_m["date"].dt.to_period("M").astype(str)
309 demo_m["month"] = demo_m["date"].dt.to_period("M").astype(str)
310 bio_m["month"] = bio_m["date"].dt.to_period("M").astype(str)
311
312 em = enrol_m.groupby(["state", "month"])["total_enrol"].sum().reset_index()

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313 dm = demo_m.groupby(["state", "month"])["total_demo"].sum().reset_index()
314 bm = bio_m.groupby(["state", "month"])["total_bio"].sum().reset_index()
315
316 sm = em.merge(dm, on=["state", "month"], how="left").merge(bm, on=["state", "month"],
317 how="left").fillna(0)
318 sm["updates"] = sm["total_demo"] + sm["total_bio"]
319 sm["ratio"] = sm["updates"] / (sm["total_enrol"] + eps)
320
321 top_states = sm.groupby("state")
322     ["total_enrol"].sum().sort_values(ascending=False).head(12).index
323 pivot = sm[sm["state"].isin(top_states)].pivot(index="state", columns="month",
324 values="ratio").fillna(0)
325
326 plt.figure(figsize=(14,6))
327 plt.imshow(pivot.values, aspect="auto")
328 plt.title("Heatmap – State × Month Update/Enrolment Ratio (Top States)")
329 plt.xlabel("Month")
330 plt.ylabel("State")
331 plt.xticks(range(len(pivot.columns)), pivot.columns, rotation=45, ha="right")
332 plt.yticks(range(len(pivot.index)), pivot.index)
333 plt.colorbar(label="Updates per Enrolment")
334 plt.tight_layout()
335 savefig("13_heatmap_state_month_ratio.png")
336 plt.show()
337
338 # =====
339 # 14) District-level analysis: Top districts by enrolments (bar)
340 # (supports: rural/semi-urban participation)
341 # NOTE: uses districts available in dataset; interpret with context.
342 # =====
343 if "district" in enrol.columns:
344     dist = enrol.groupby(["state", "district"])["total_enrol"].sum().reset_index()
345     # take top 15 overall districts
346     dist_top = dist.sort_values("total_enrol", ascending=False).head(15)
347     labels = dist_top["state"] + " - " + dist_top["district"]
348
349     plt.figure(figsize=(12,6))
350     plt.bar(labels, dist_top["total_enrol"].values)
351     plt.title("Top 15 Districts – Total Aadhaar Enrolments")
352     plt.xlabel("District")
353     plt.ylabel("Enrolments")
354     plt.xticks(rotation=60, ha="right")
355     plt.tight_layout()
356     savefig("14_top15_districts_enrol.png")
357     plt.show()

```