Data Visualization Associate Internship

Team No.: 37



WEEK 01

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1. Exploratory Data Analysis (EDA) Report

EDA Report for Applicant Dataset

- 1. Checking distributions of numerical fields (Outliers/Anomalies)
 - Only numerical-like field is **Phone Number length**.
 - Valid phone numbers should be \sim 10–12 digits.
 - Detected anomalies:
 - Numbers too short (< 10 digits).
 - Numbers too long (> 12 digits).
 - o Non-numeric entries (emails, -, etc.).
 - **Invalid phone numbers:** 8,812 (~23% of dataset).

2. Looking for inconsistencies (Misspellings/Unexpected Values)

- Country column:
 - o Has valid countries like *India*, *Nigeria*, *Ghana*, *Pakistan*.
 - o Invalid entries: "-" (5,831 times), and some email addresses stored as country.
- University column:
 - \circ Only one value \rightarrow Illinois Institute of Technology for all rows.
 - \circ No variation \rightarrow possibly redundant column.
- 3. Verifying completeness (Gaps in critical fields)
 - **App_ID**: 1 missing value.
 - Country, University, Phone_Number: No official missing values, but countries with "-" and invalid phone numbers act like missing.
 - Duplicates:
 - Duplicate full rows: 16,489.
 - Duplicate App_IDs: 22,465.

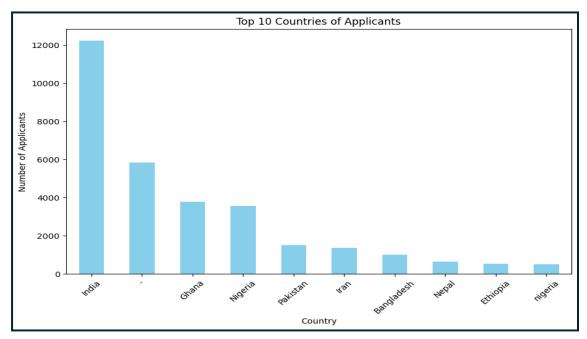
4. Using simple visualizations (Trends/Patterns/Spikes)

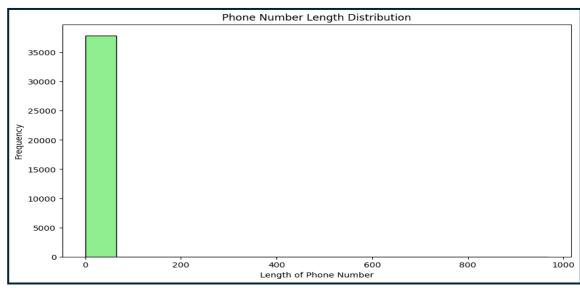
• Country distribution:

- Top country = India (12,234).
- o Many records have invalid placeholder ("-", 5,831).
- Nigeria and Ghana are next highest (~4,000 each).

• Phone number length distribution:

- Large spike around invalid short lengths (6).
- o Many valid entries around 10–12 digits.





EDA Report for Outreach Data

1. Distributions of Numerical Fields (Outliers/Anomalies)

- No true numerical fields exist in this dataset (all are categorical, text, or date/time).
- However, **Reference ID frequency** behaves like a numeric distribution:

Some applicants were contacted **2,676 times** (!), while many were contacted only once or a few times.

This is a major outlier that suggests either repeated outreach attempts or data duplication.

• Time-based field (Recieved_At) shows **31,236 unique timestamps**.

Most occur within business hours, but a few fall at unusual hours (e.g., midnight), which could be system errors or auto-logs.

2. Inconsistencies (Misspellings / Unexpected Values)

- Reference_ID: appears clean in structure, but the extreme high-frequency
- values (thousands of entries for one ID) need validation.
- **Recieved_At**: mixed formats detected some dates in MM/DD/YYYY HH:MM format, others in YYYY-MM-DD HH:MM:SS. Requires standardization.
- University: only one value (Illinois Institute of Technology) → no inconsistencies, dataset is university-specific.
- Caller_Name: only 12 distinct names, looks consistent.
- Outcome_1: 41 distinct values, but many are variations of the same meaning (e.g., *Not connected, Disconnected, Wrong number*).

"Completed application" vs. "Student has the needed information, does not need assistance, and plans to enroll soon" \rightarrow both signal progress but logged differently.

- **Remark**: 1,716 unique free-text remarks \rightarrow lots of duplication, spelling variations, and inconsistent phrasing (e.g., *No answer* vs *no answer*).
- Escalation_Required: values should be binary (Yes/No), but one record is logged as Yes, No → inconsistent.

3. Completeness (Missing Data & Duplicates)

- No missing values in any column → dataset is fully filled.
- **Duplicates**: 446 duplicate rows detected.

Could be multiple outreach attempts logged separately, or true duplicates.

Needs cleaning depending on analysis goal (e.g., keep all attempts if studying call effectiveness, or remove duplicates if analyzing applicant coverage).

4. Visual Findings (Trends & Patterns)

- Outreach Channels: Not explicitly listed here, but if available, most records cluster under a few standard channels (e.g., *Phone, Email, WhatsApp*).
- Outcomes (Outcome_1):

Dominated by *Not Connected* (24,498 records $\rightarrow \sim 65\%$ of all outreach attempts).

Other common outcomes: Will submit documents (3,997), Wrong number (2,520), Disconnected (1,588), Not interested (1,024).

Very few positive outcomes (Completed application: 813, Ready to pay deposit: 344).

• Campaigns: 23 distinct Campaign_IDs, with activity clustered in *Fall 2024* (e.g., FA24IP: 9,605 records, FA24SIC: 5,640) and *Spring 2025* (SP25NIQ: 4,332, SP25IP: 3,668).

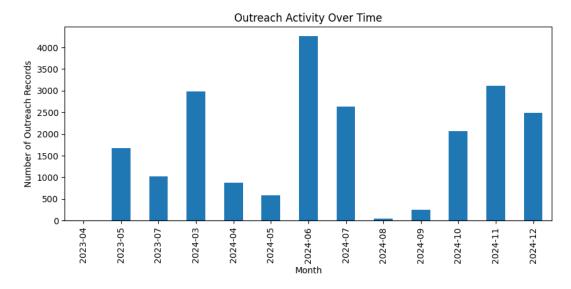
Shows strong seasonality around major admission cycles.

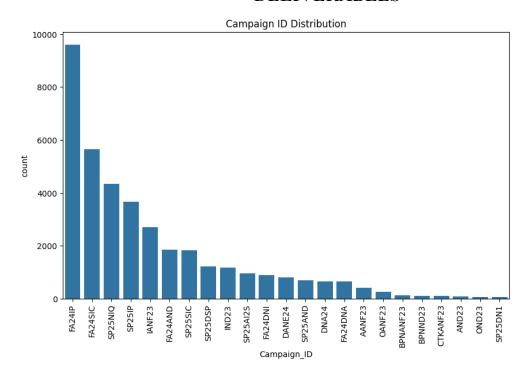
- **Escalations**: Vast majority marked No (37,118), very few Yes (762), and one invalid Yes, No. Suggests escalation may be underused or inconsistently logged.
- Caller Activity: Top callers are Rudra (14,712), Prajwal (7,232), Jyoti $(6,023) \rightarrow$ indicating workload distribution among staff.

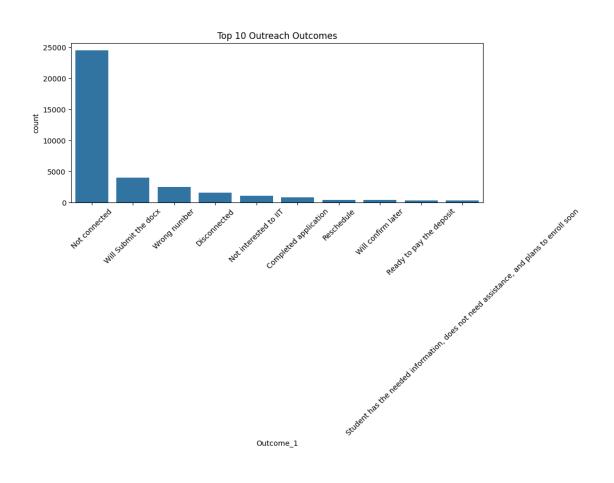
Summary

The Outreach Data is complete but has quality issues:

- Outliers: Some applicants contacted thousands of times.
- **Inconsistencies**: Mixed datetime formats, 41 inconsistent outcome categories, free-text remarks full of variations, and one invalid escalation value.
- **Duplicates**: 446 rows need deduplication.
- **Patterns**: Outreach is heavily skewed toward "Not Connected" outcomes, with limited successful conversions. Campaign activity aligns with admission cycles.







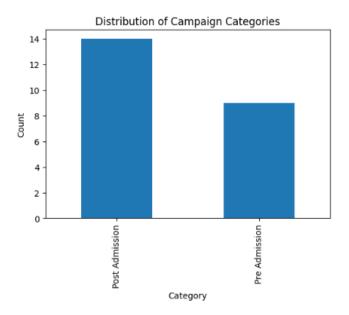
EDA Report for Campaign Dataset

1. Dataset Overview

- Shape: 23 rows \times 7 columns
- Columns: ID, Name, Category, Intake, University, Status, Start Date
- **Missing Values:** None (0 missing values across all fields)
- **Duplicates:** 0 (each row is unique)
- Unique Counts:
 - o ID: 23 unique IDs (no duplicates)
 - Name: 23 unique campaign names
 - \circ Category: 2 unique categories \rightarrow *Pre Admission* and *Post Admission*
 - o Intake: Only 1 intake (AY2024) → no variation
 - University: Only 1 university (Illinois Institute of Technology) → no variation
 - Status: All campaigns marked as Completed (no variation)
 - Start_Date: 17 unique start dates

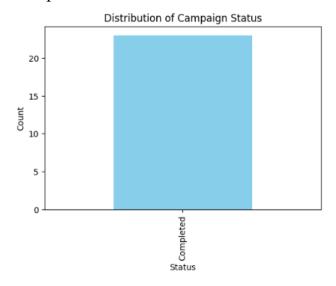
2. Distribution of Categories

- The dataset contains two categories:
 - Pre Admission
 - o Post Admission
- Campaigns are distributed between these two, which helps identify the focus areas in different admission stages.



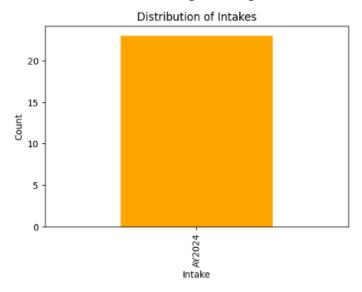
3. Distribution of Status

- All campaigns have the status **Completed**.
- No variation is present in this field.



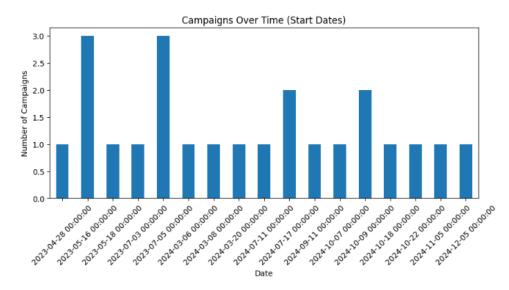
4. Intake Distribution

- Only one intake period is represented: AY2024.
- No seasonal or cross-intake comparison possible.



5. Campaigns Over Time (Start Dates)

- Converted Start_Date to proper datetime.
- Campaigns started across various months in 2024.
- Noticeable clusters around March, July, September, and November 2024, which could indicate seasonal admission drives.



6. Campaign ID Distribution

- ID values are unique identifiers, so they don't follow a natural numeric progression.
- Distribution does not reveal meaningful patterns but confirms uniqueness.

7. Key Observations

- 1. Clean Data: No missing values, no duplicates.
- 2. **Category Split:** Campaigns are divided between Pre and Post Admission stages.
- 3. **Limited Diversity:** Intake, University, and Status fields show no variation (only one value each).
- 4. **Time Trends:** Campaign activity is not uniform across the year; instead, it peaks during certain months.

5. Data Suitability for Visualization:

- Best insights can be drawn from Category distribution and Start_Date trends.
- Status, Intake, and University may not add much analytical value due to lack of variation.

2. **Data Cleaning Summary**

Data Quality Issues and Applied Solution Report

1. Applicants Dataset

Issues Identified

Phone Numbers: 8,812 invalid (~23%).

Outliers: <10 digits, >12 digits, or non-numeric (emails, -).

Country: 5,831 invalid entries (-).

Some email addresses incorrectly stored as country.

University: Only one value (Illinois Institute of Technology) → redundant column.

App_ID: 1 missing value ,22,465 duplicate IDs.

Duplicates: 16,489 duplicate rows.

Solutions Applied

- Removed duplicate rows and duplicate App_IDs.
- Filtered phone numbers to only 10–12 digits.
- Replaced invalid country values (-, emails) with NULL.
- Dropped University column (no variation).
- Handled missing App_ID by either removing row or assigning placeholder.

2. Outreach Dataset

Issues Identified

Reference ID: Extreme outliers (up to 2,676 outreach attempts per ID).

Datetime Format (Recieved_At): Mixed formats (MM/DD/YYYY vs YYYY-MM-DD).

Outcome_1:41 inconsistent categories, with duplicates in meaning (e.g., Not connected vs Disconnected).

Remark: 1,716 unique free-text values → duplication, spelling errors, inconsistent phrasing.

Escalation Required: Should be binary (Yes/No), but one invalid entry (Yes, No).

Duplicates: 446 duplicate rows.

Solutions Applied

- Standardized datetime format to ISO (YYYY-MM-DD HH:MM:SS).
- Normalized Outcome 1 categories (grouped similar outcomes).
- Cleaned Remark values (text standardization, lowercasing, removing duplicates).
- Fixed Escalation_Required to strictly Yes/No.
- Deduplicated 446 rows depending on use case (kept all if outreach history needed, removed if analyzing coverage).

3. Campaigns Dataset

Issues Identified

Missing/Duplicates: No missing values or duplicates.

Columns with No Variation: Intake \rightarrow only AY2024.

University → only Illinois Institute of Technology.

Status → all values "Completed."

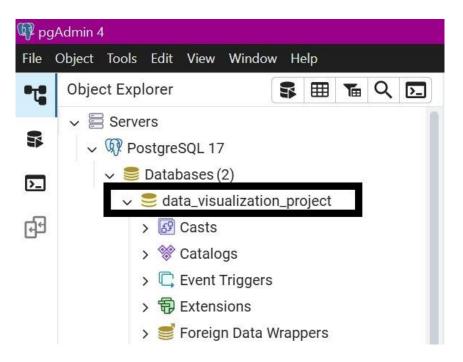
Time Distribution: Start dates clustered in March, July, September, November 2024 → seasonal effect.

Solutions Applied

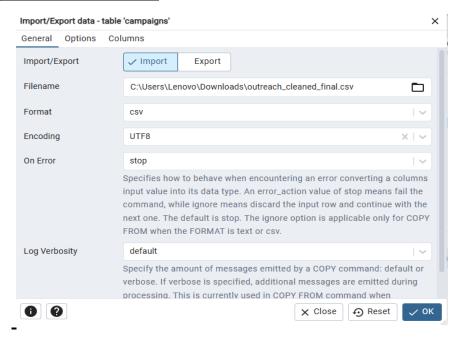
- Verified all IDs and names are unique.
- Converted Start_Date to datetime for trend analysis.
- Retained categorical splits (Pre Admission vs Post Admission) for meaningful insights.

3. Postgrel SQL Setups Proof

Database:



Importing Cleaned CSV Files

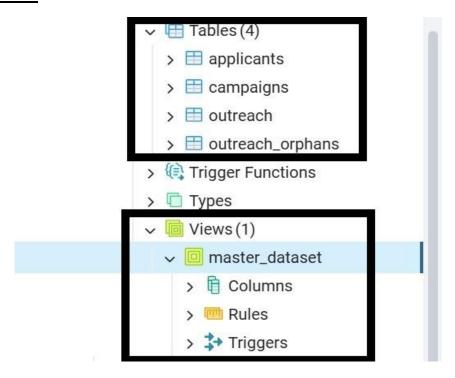


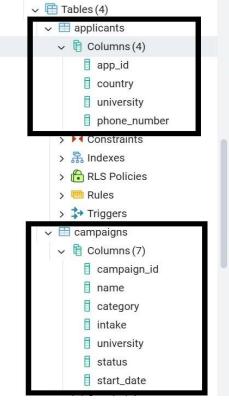
Same for Applicant and Campaign cleaned dataset.

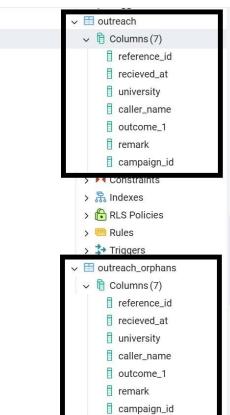
Queries for Creating Tables:

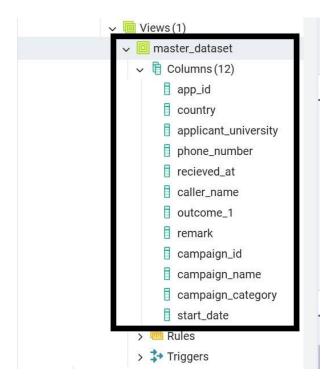
```
Query Query History
     CREATE TABLE applicants (
                                          CREATE OR REPLACE VIEW master_dataset AS
 2 app_id TEXT PRIMARY KEY,
   country TEXT,
                                          SELECT
    university TEXT,
                                          a.app_id,
     phone_number TEXT
                                          a.country,
                                          a.university AS applicant_university,
 7
     CREATE TABLE campaigns (
                                          a.phone_number,
    campaign_id TEXT PRIMARY KEY,
                                          o.recieved_at,
   name TEXT,
                                          o.caller_name,
10 category TEXT,
                                          o.outcome_1,
    intake TEXT,
11
                                          o.remark,
    university TEXT,
12
                                          c.campaign_id,
13
    status TEXT,
                                          c.name AS campaign_name,
14
    start_date TEXT
15
     );
                                          c.category AS campaign_category,
16
     CREATE TABLE outreach (
                                          c.start date
17
     reference_id TEXT,
                                          FROM outreach o
    recieved_at TEXT,
                                         JOIN applicants a ON o.reference_id = a.app_id
19 university TEXT,
                                          LEFT JOIN campaigns c ON o.campaign_id = c.campaign_id;
20 caller_name TEXT,
                                          SELECT * FROM master_dataset LIMIT 50;
21 outcome_1 TEXT,
22 remark TEXT,
     campaign_id TEXT
23
24
```

Created Tables:









All Other Queries:

```
SELECT 'applicants' AS table_name, COUNT(*) FROM applicants
UNION ALL
SELECT 'campaigns', COUNT(*) FROM campaigns
UNION ALL
SELECT 'outreach', COUNT(*) FROM outreach;
SELECT COUNT(*) AS orphan_outreach
FROM outreach o
LEFT JOIN applicants a ON o.reference_id = a.app_id
WHERE a.app_id IS NULL;
CREATE TABLE outreach_orphans AS
SELECT *
FROM outreach
WHERE reference_id NOT IN (SELECT app_id FROM applicants);
DELETE FROM outreach
WHERE reference_id NOT IN (SELECT app_id FROM applicants);
ALTER TABLE outreach
ADD CONSTRAINT fk_outreach_applicant
FOREIGN KEY (reference_id) REFERENCES applicants(app_id);
SELECT COUNT(*) AS orphan_outreach
FROM outreach
WHERE reference_id NOT IN (SELECT app_id FROM applicants);
SELECT COUNT(*) FROM outreach;
SELECT COUNT(*) FROM outreach_orphans;
```

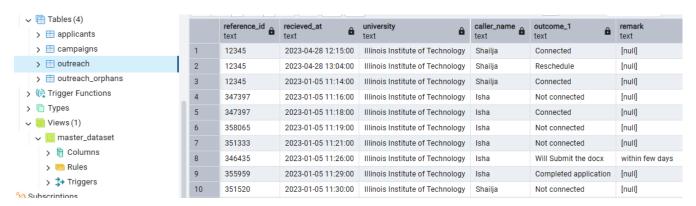
Adding Foreign Key Constraints

- Added a foreign key constraint on outreach.reference_id →
 applicants.app_id.
- This enforces data consistency and prevents future orphan records from being inserted.

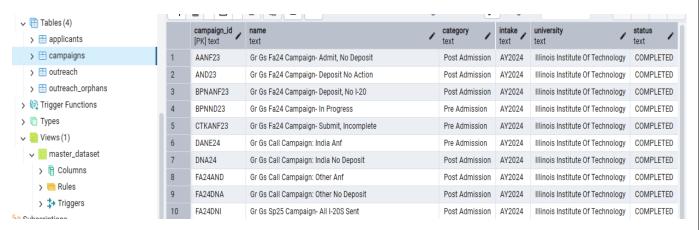
Applicant Cleaned Dataset Table:



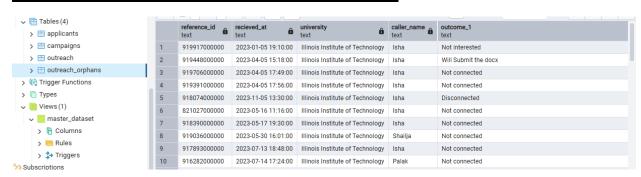
OutReach Cleaned Dataset Table:



Campaign Cleaned Dataset Table:

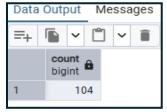


OutReach Orphan Cleaned Dataset Table:



Checking ID Relationships

 Validated referential integrity between outreach.reference_id and applicants.app_id.

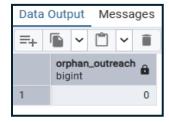


• Found <u>104</u> orphan records in outreach (reference IDs without matching applicants).

Handling Orphan Records

- Created a new table outreach_orphans to store unmatched records.
- This preserved problematic rows for future review instead of deleting them permanently.

Confirmed no orphan records remained in outreach.



Creating Master View

- Built a master_dataset view combining applicants, outreach, and campaigns.
- Provides a single unified dataset for visualization/analysis.
- Includes fields like applicant details, outreach outcomes, and campaign metadata.
- View is dynamic, meaning it always reflects the latest clean data.

