Data Intake Report

Name: Persistency of Drug – Healthcare Project

Report date: 19/08/2025 Internship Batch: LISUM47

Version: 1.0

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Data intake reviewer: Data storage location:

https://drive.google.com/file/d/1P_oMc6gOBlhw6dY5PxaqxV2swdHMUooK/view

Tabular data details:

File Name	Healthcare_dataset
Total number of observations	3424
Total number of files	-
Total number of features	69
Base format of the file	.xlsx
Size of the data	898 KB

Objective:

This report documents the quality checks and structure of the Healthcare dataset provided for analyzing **persistency of a drug**. The dataset encompasses patient demographics, provider attributes, clinical factors, comorbidities, adherence, and risk factors to support the development of a classification model that predicts patient persistence.

Proposed Approach:

1. Data Cleaning:

- a. Null Value Check:
 - i. Approach for check:
 - Used .isnull().sum() in Pandas to identify any row has null values or not.
 - If duplicates were found, we would drop them using .dropna()

b. Duplicate Value Check:

- i. Approach for check:
 - Used .duplicated().sum() in Pandas to identify any full-row or key-based duplicates.
 - If duplicates were found, we would drop them using .drop duplicates()

2. Column-Wise Format Fixes:

a. Standardize categorical variables (e.g., Race, Gender, Region).

b. Convert Age buckets into numerical or ordinal encoding.

3. Assumptions:

- Patient ID (Ptid) is unique and primary key.
- Persistency Flag is the dependent variable (binary classification).
- Clinical metrics such as T-Score and Risk Segments are consistent across patients.

4. Outlier detection

• Outliers are expected in continuous variables, such as Age, T-score, and Dexa Scan Frequency.

5. Correlation Observations

Persistency likely correlated with Age, Adherence, Risk Segments, and Comorbidity factors.