

# Schizophrenia Prediction Model Architecture

## 1. Objective

To develop a binary classification model that predicts whether a patient is schizophrenic (1) or not schizophrenic (0) using a dataset of 10,000 patient records with no missing values.

## 2. Data Input

- Dataset consists of 10,000 rows and 18 features:
  - **Patient Information:**
    - Age (18-80)
    - Gender (0: Female, 1: Male)
    - Education Level (1: Primary, 2: Middle School, 3: High School, 4: University, 5: Postgraduate)
    - Marital Status (0: Single, 1: Married, 2: Divorced, 3: Widowed)
    - Occupation (0: Unemployed, 1: Employed, 2: Retired, 3: Student)
    - Income Level (0: Low, 1: Medium, 2: High)
    - Living Area (0: Rural, 1: Urban)
  - **Medical Information:**
    - Disease Duration (1-40 years for schizophrenia patients)
    - Hospitalizations (0-10 for schizophrenia patients)
    - Family History (0: No, 1: Yes)
    - Substance Use (0: No, 1: Yes)
    - Suicide Attempt (0: No, 1: Yes)
    - Positive Symptom Score (0-100)
    - Negative Symptom Score (0-100)
    - GAF Score (0-100)
  - **Social and Behavioral Factors:**
    - Social Support (0: Low, 1: Medium, 2: High)
    - Stress Factors (0: Low, 1: Medium, 2: High)
    - Medication Adherence (0: Poor, 1: Moderate, 2: Good)

### 3. Data Preprocessing

#### 1. Feature Scaling:

- Apply Min-Max Scaling for numerical columns like Age, Symptom Scores, and GAF Score to normalize the data between 0 and 1.

#### 2. Train-Test Split:

- Split the dataset into:
  - 80% Training Data
  - 20% Testing Data

### 4. Model Selection

- **Logistic Regression:**

- Chosen for its simplicity, interpretability, and suitability for binary classification problems.
- Assumes a linear relationship between input features and the log-odds of the target variable.

### 5. Model Training and Prediction

#### 1. Training Phase:

- Fit the logistic regression model using the training dataset.

#### 2. Prediction Phase:

- Predict the probability of a patient being schizophrenic (1) or not (0).

### 6. Model Evaluation

- **Primary Metric: Accuracy**

- Evaluate the model's correctness by calculating the proportion of correct predictions:

- **Confusion Matrix:**

- Display True Positives, True Negatives, False Positives, and False Negatives to analyze performance in detail.

### 7. Pipeline Summary

1. Input patient data (18 features).
2. Preprocess the data (encoding, scaling, train-test split).

3. Train a logistic regression model.
4. Make predictions on new patient data.
5. Evaluate performance using accuracy and a confusion matrix.