

# Student Engagement and Churn Prediction

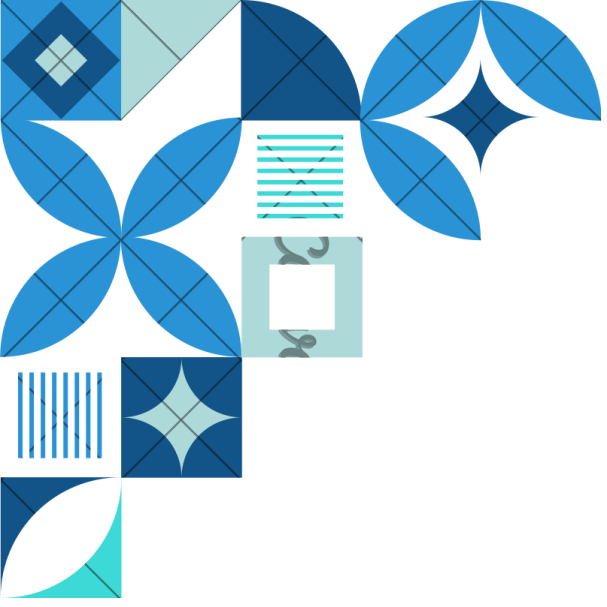
**AI-Powered**

## Team Members:

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# Agenda

- Introduction & Business Problem
- Dataset Overview
- Data Preprocessing
- Exploratory Data Analysis (EDA)
- Churn Analysis
- Model Building
- Model Comparison
- Summary of Strategic Value
- Churn Factors & Recommendations
- Limitations & Future Work
- Conclusion

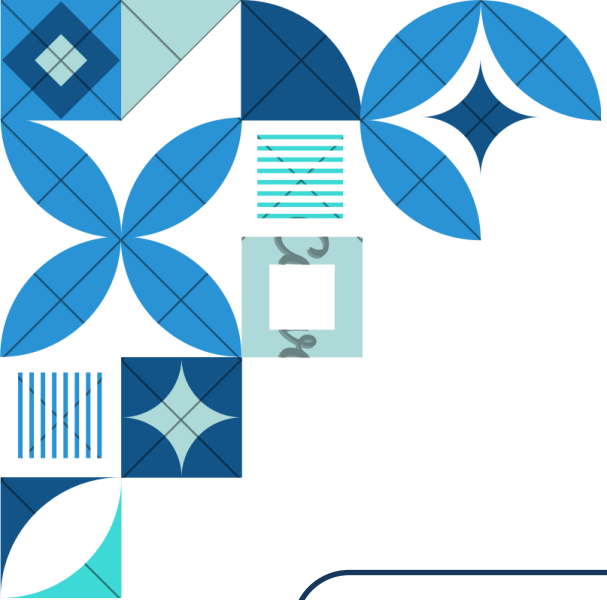




# Introduction & problem Statement

- **Strong Initial Enthusiasm**
  - Students enroll with high motivation to gain skills and career growth.
  - Start with strong expectations, ambition, and promise of opportunities.
- **Unexpected Dropouts**
  - 31.8% of students withdraw silently without warning signals.
  - No clear indicators → difficult to predict or intervene.
  - Dropouts driven by multiple factors (personal, academic, platform-related).
  - Leads to loss of potential and wasted platform resources.
- **Core Problem**
  - High enrollments, but low course completions.
  - Urgent need for predictive solutions to detect risks early and reduce churn.





# Dataset Overview

**Total Records**  
**8558**

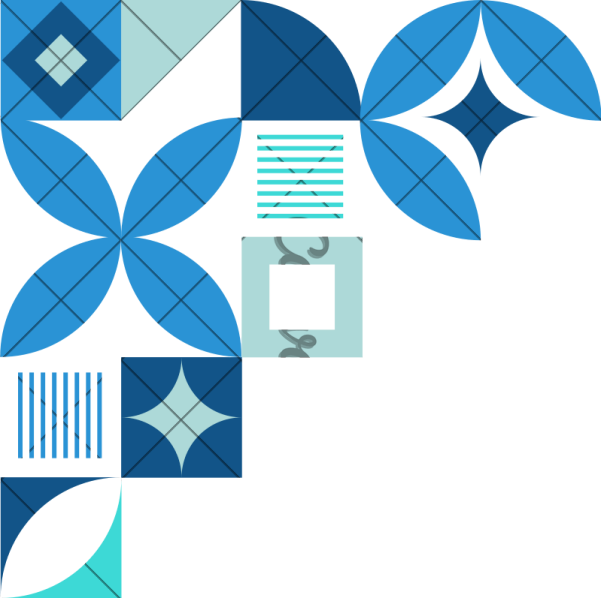
**Total Countries**  
**70**

**Success Rate**  
**47.6%**

**Peak Signups**  
**946**

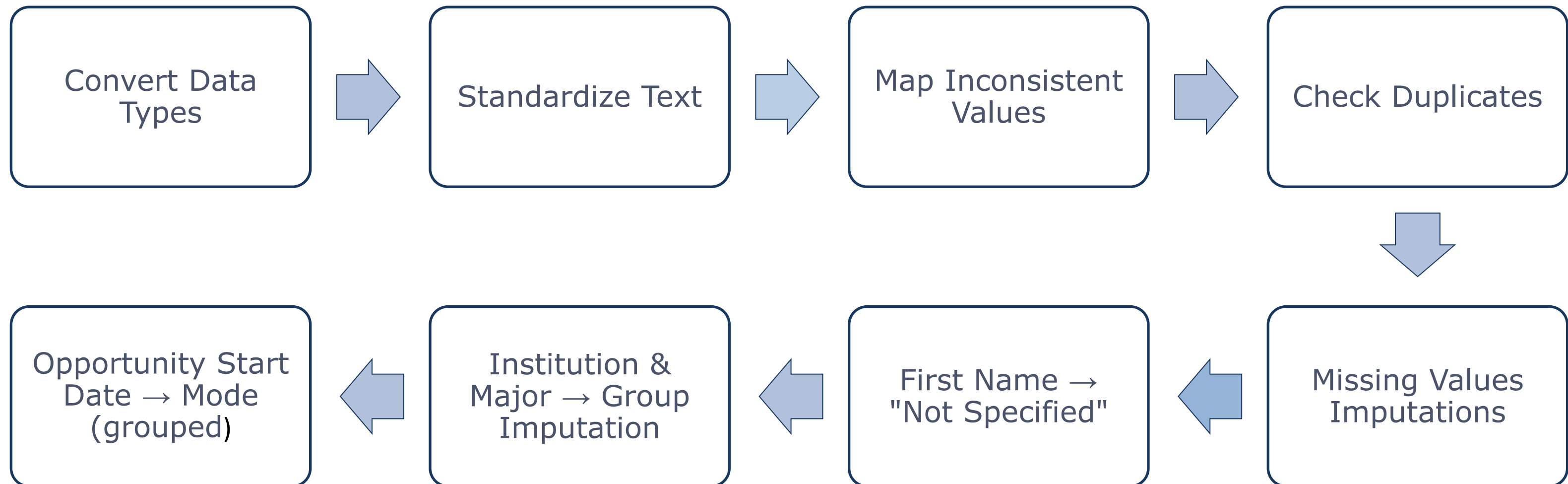
- **Scale:** 8,558 learner applications
- **Coverage:** 70 countries, grouped into 5 regions, 23 programs
- **Success Rate:** 47.6% overall with significant demographic variations
- **Key Insight:** Peak activity was in August 2023 with 946 signups

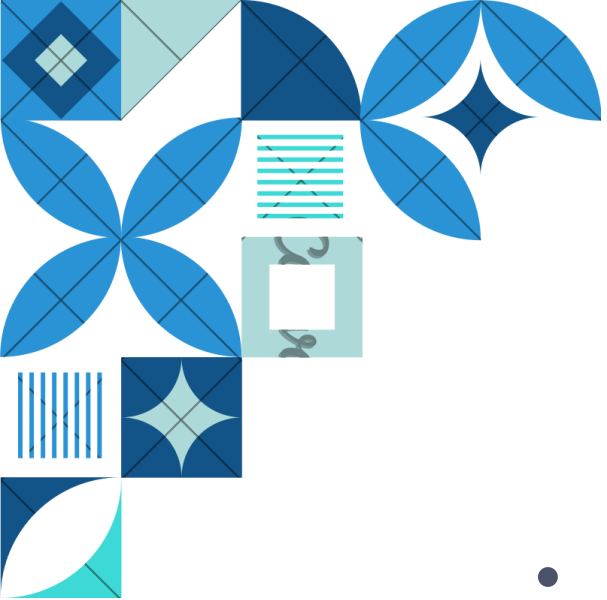




# Data Preprocessing

- **Data Cleaning:**





# Data Preprocessing

- **Feature Engineering:**

## Temporal Features:

Capture time-based patterns

Like: Quick\_Applicant,  
Signup\_to\_Apply\_Days

## Demographic Features:

Classify Learners by background

Like: Age\_Group, Region

## Behavioral Features:

Reflect learner activity & success

Like: Engagement\_Score,  
Opportunity\_Popularity

## Composite Features:

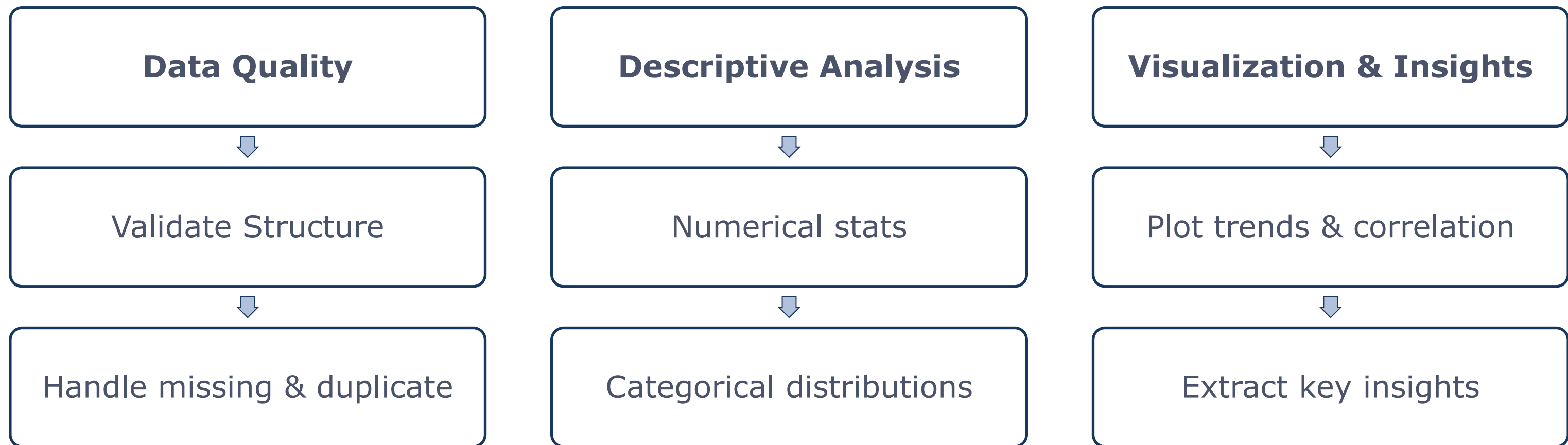
Combine multiple attributes

Like: Age\_Major\_Combo,  
Country\_Gender





# Exploratory Data Analysis

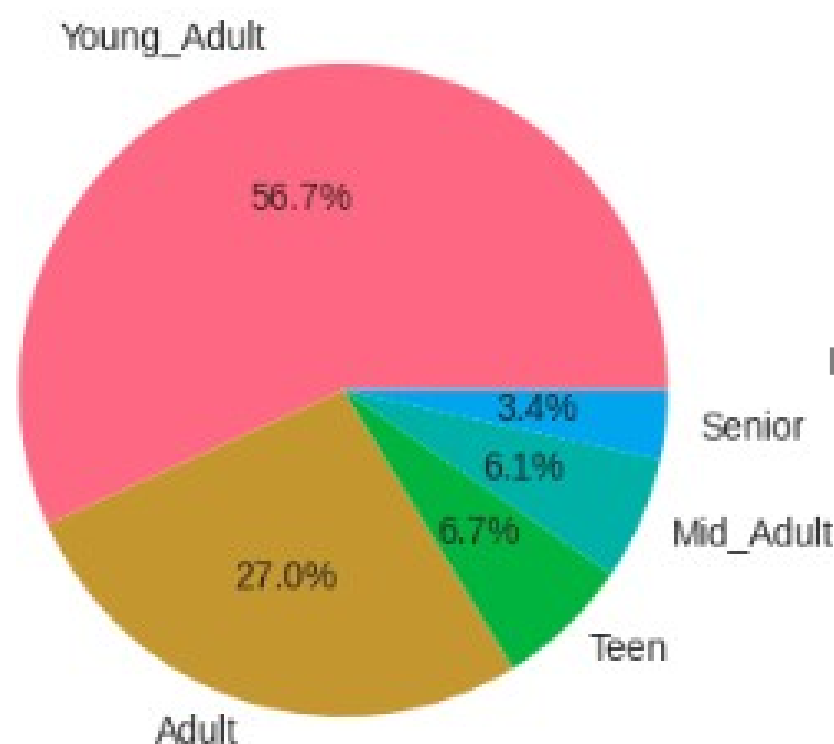






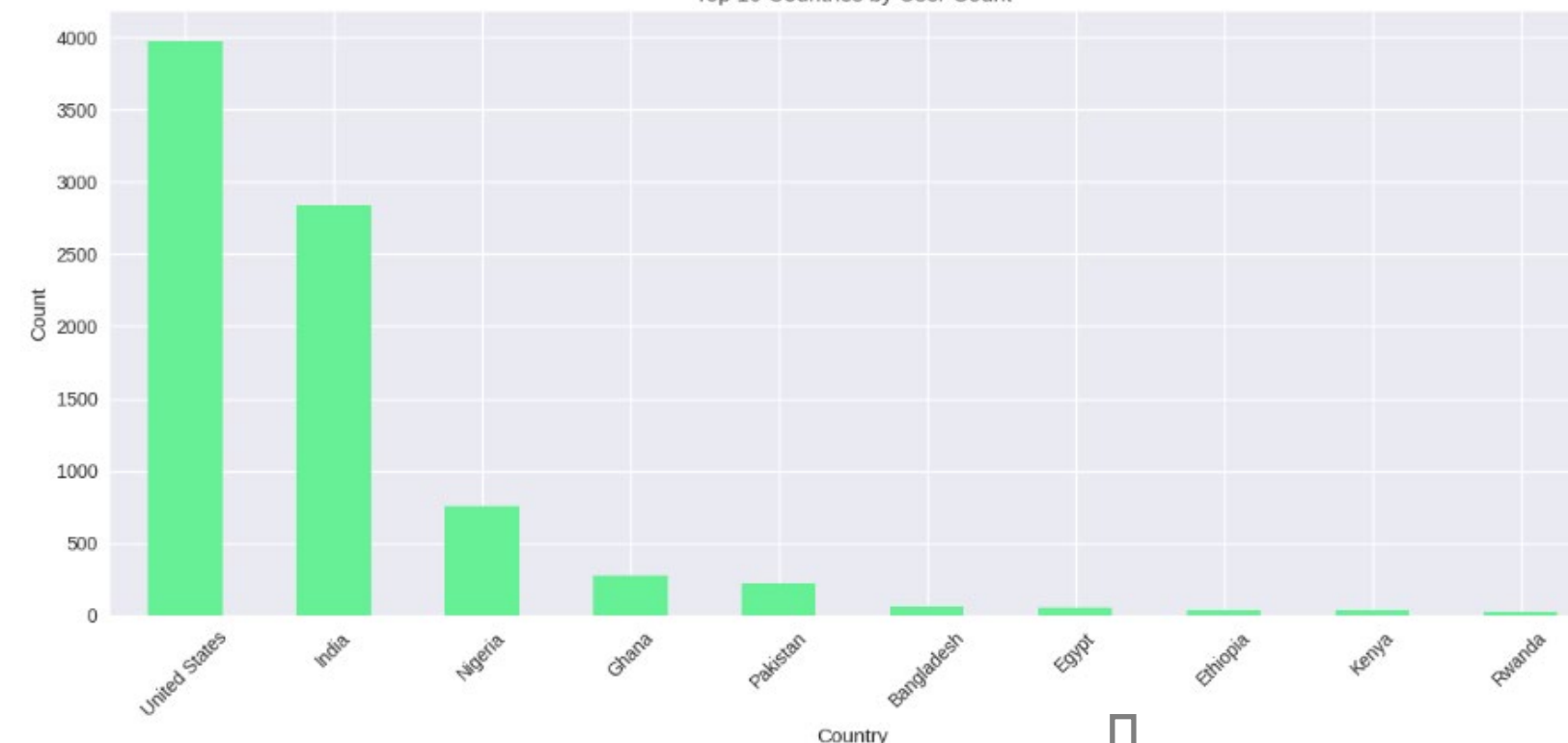
# Demographical Insights

Age Group Distribution

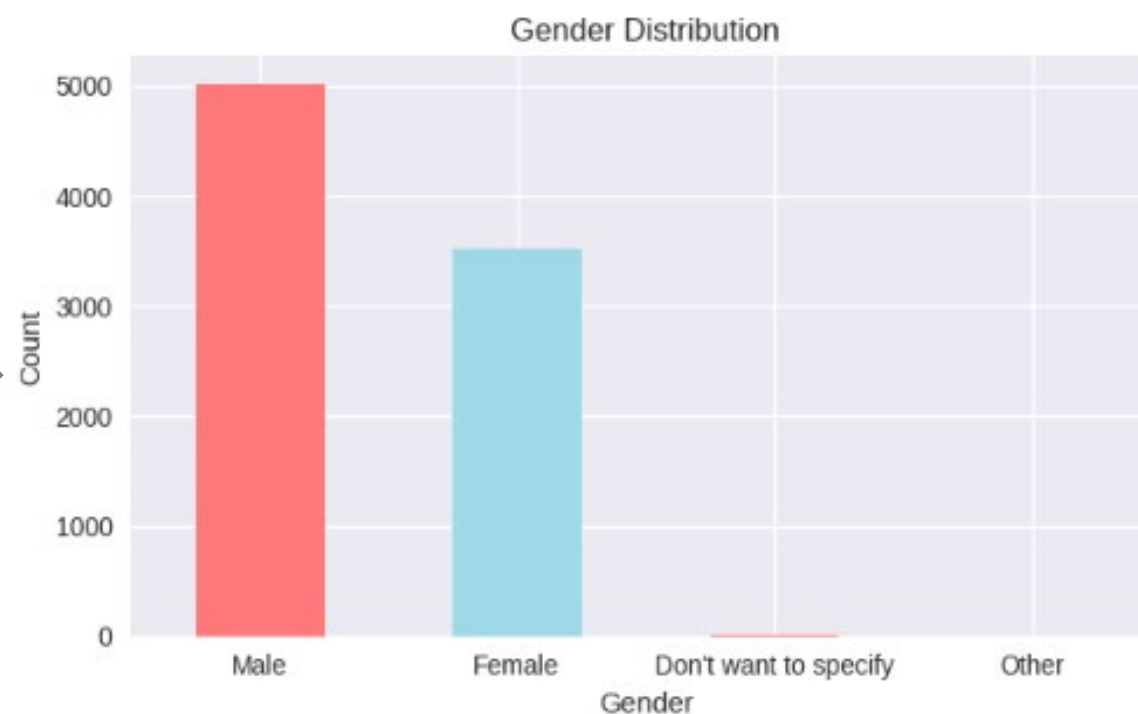


**Insight:** Young Adults dominate (56.7%); Teens and Seniors are underserved (<10%), offering key growth potential.

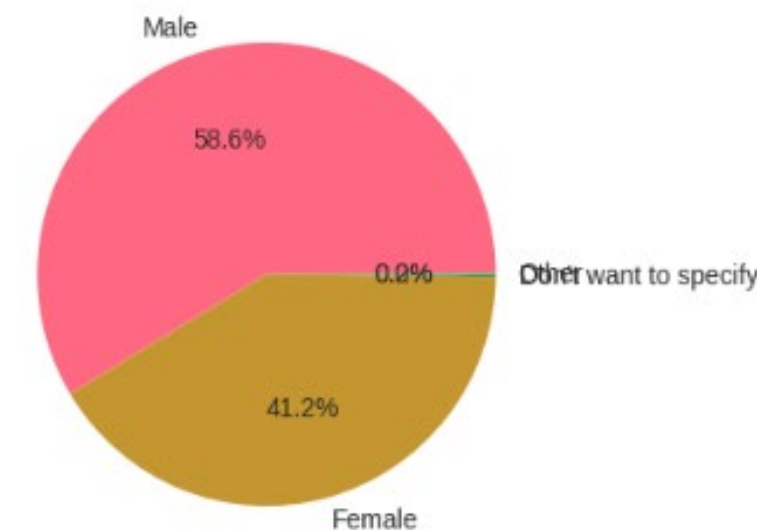
Top 10 Countries by User Count



**Insight:** The majority of users are male, making up over half of the total user base.



Gender Percentage



**Insight:** Enrollment is heavily concentrated in the United States and India, indicating strong market presence.







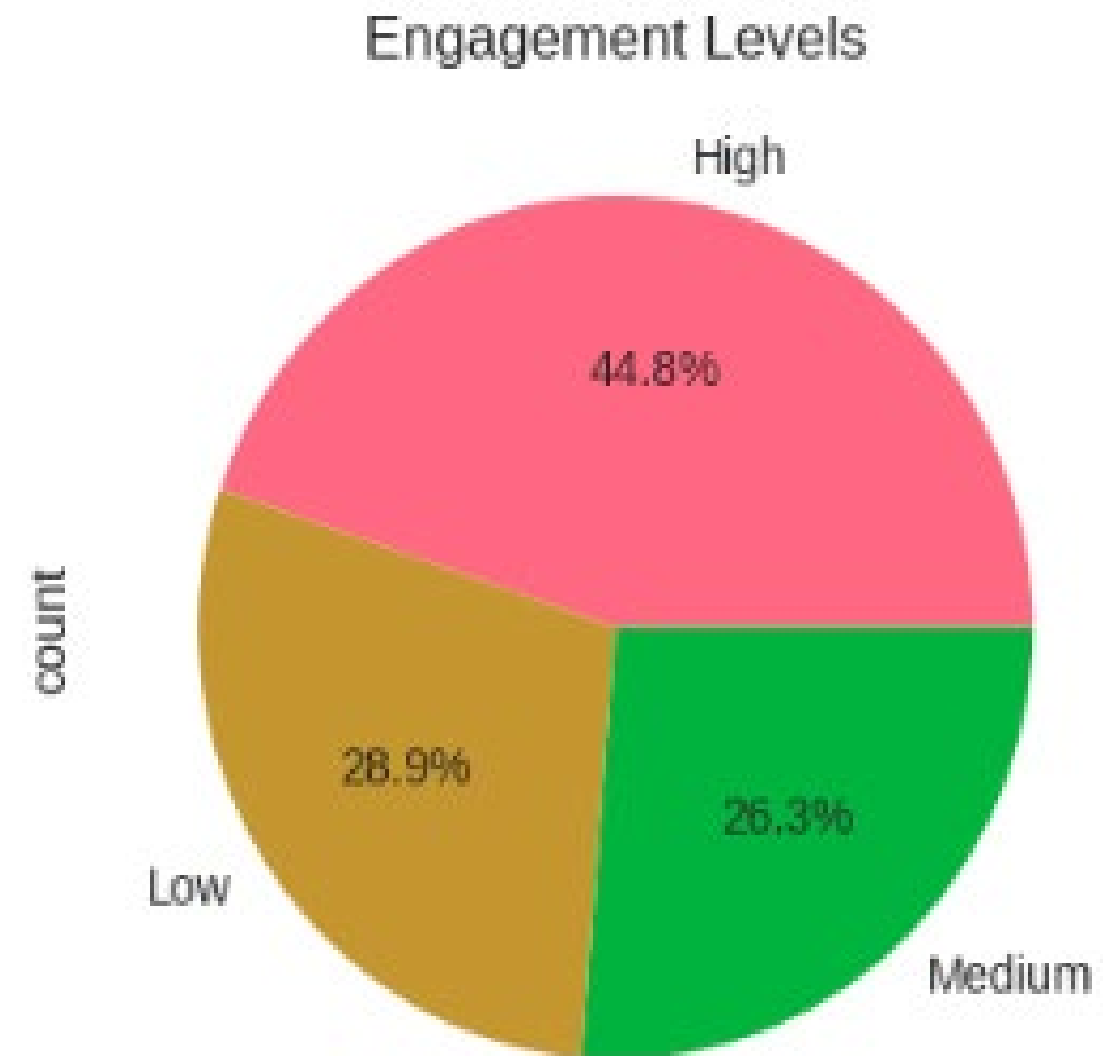
# Engagement Distribution

## Engagement Metrics:

- Average Engagement Score: 0.47
- Engagement Level Distribution:
  - High: 3,835 (44.8%)
  - Low: 2,472 (28.9%)
  - Medium: 2,251 (26.3%)

## Insights:

- Retained learners have significantly higher engagement scores.
- Churned users apply more, but have a lower success rate.





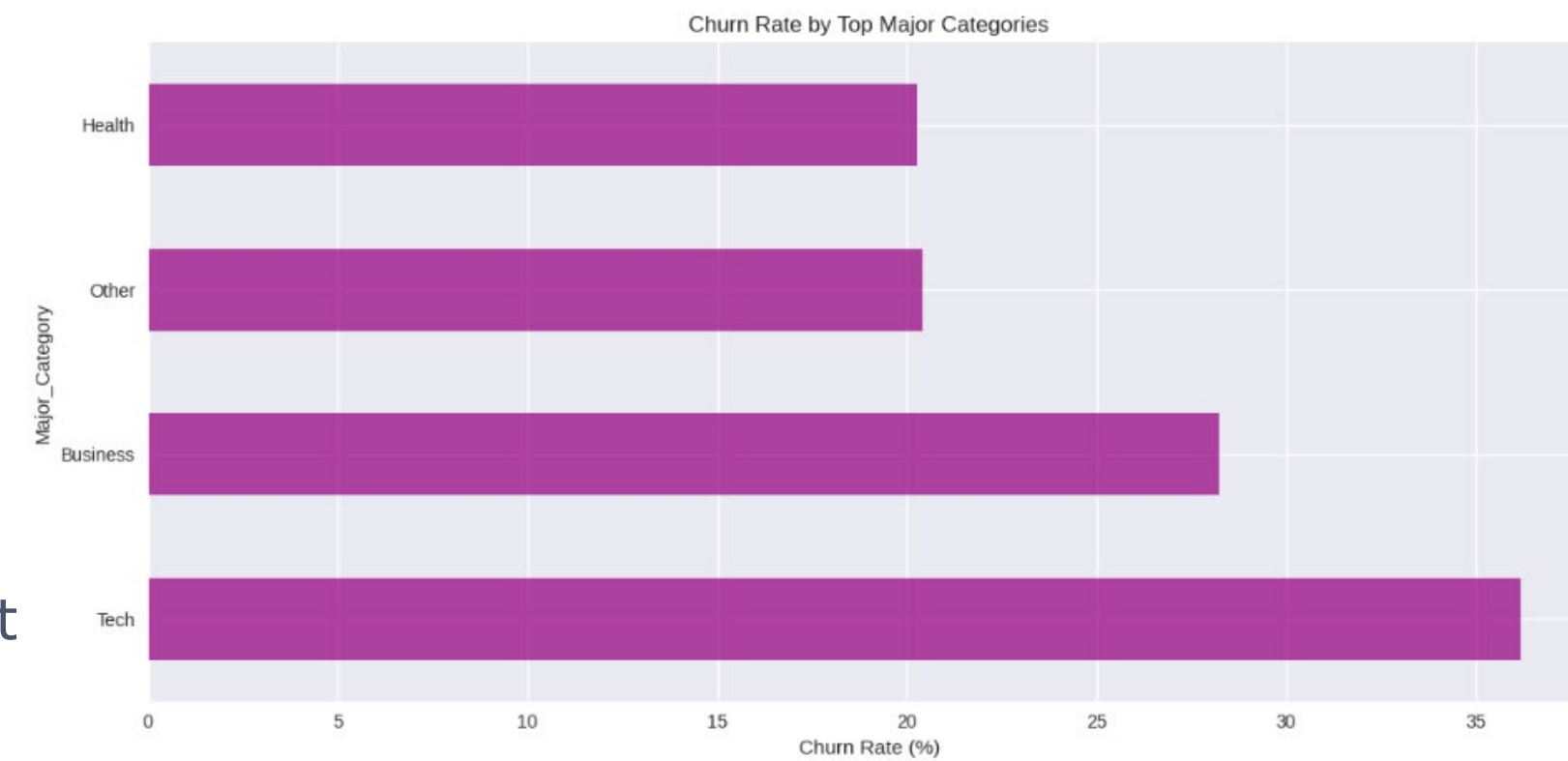
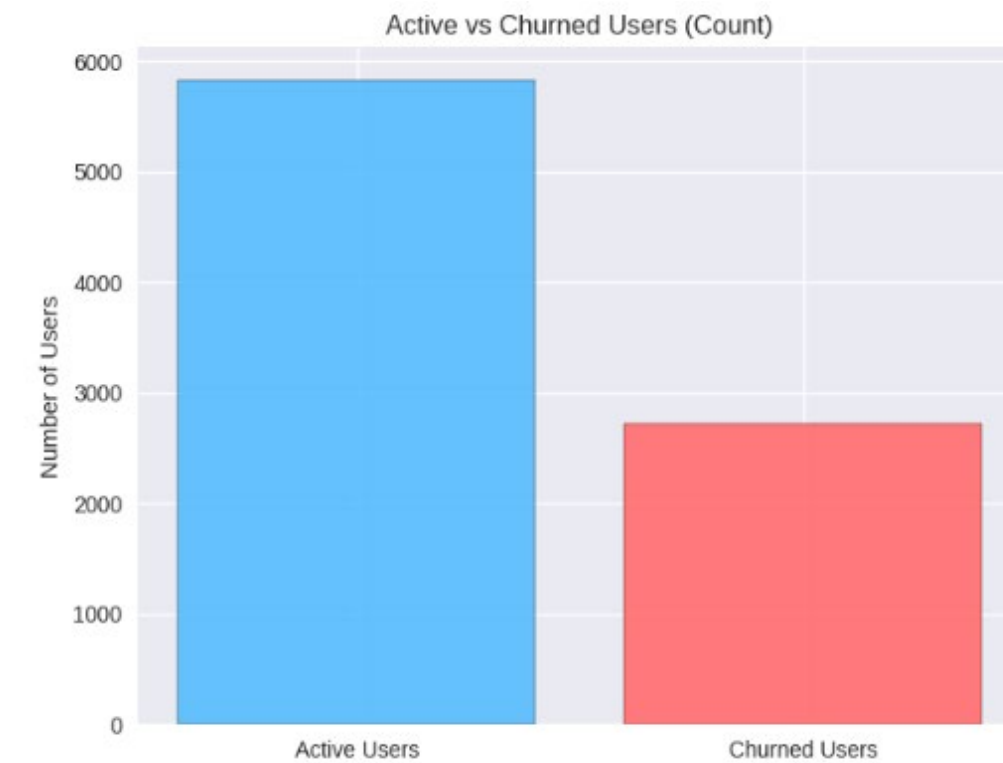
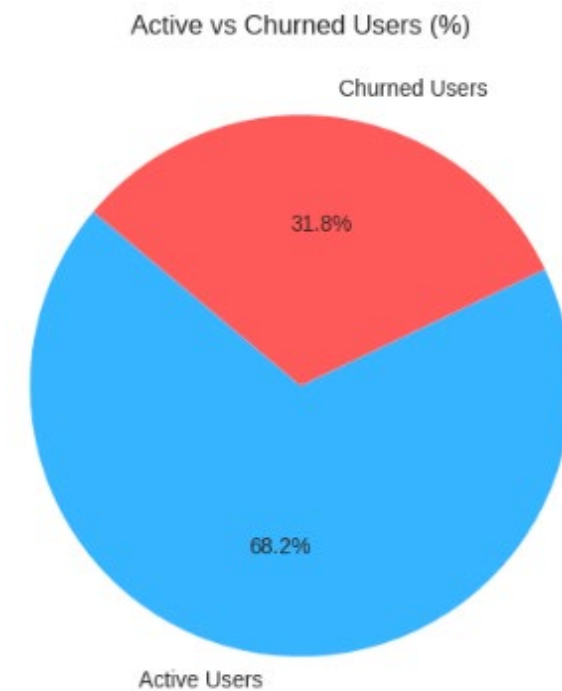
# Churn Analysis

## Understanding & Defining Churn:

- Challenge: Multiple churn definitions (status codes, behaviors, demographics)
- Approach: Rule-based clustering with Kmeans
- Key Features for Thresholds:
  - Engagement\_Score
  - Applications\_per\_User
  - User\_Success\_Rate
  - Apply\_vs\_Start\_Gap
- **Churn Rules:**
  - Low/no engagement
  - Many unsuccessful attempts
  - Dropped before course start
  - Very low success despite many applications

## Insight:

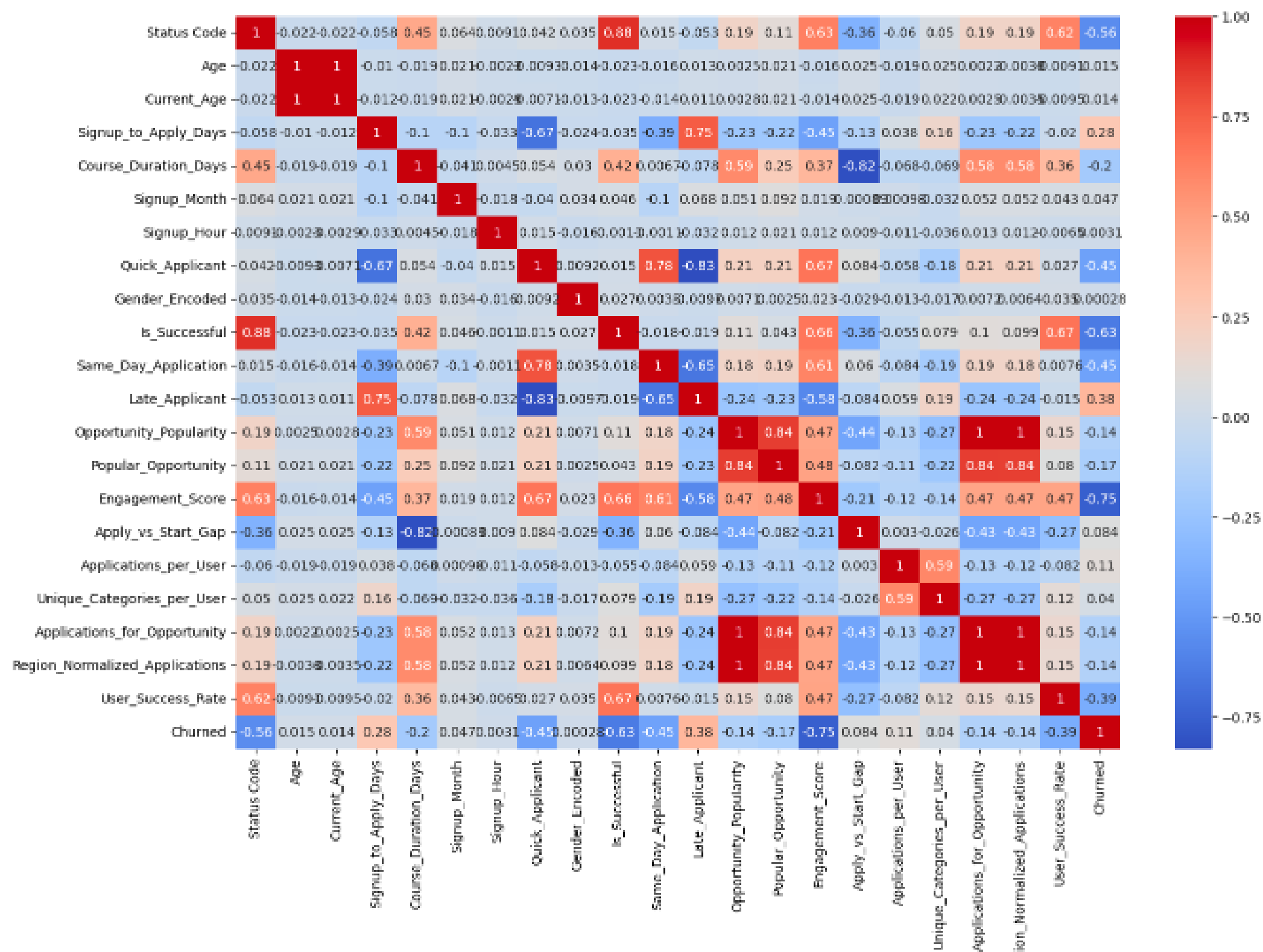
Nearly one-third (31.8%) of users have churned, indicating a significant challenge with user retention that requires immediate attention.





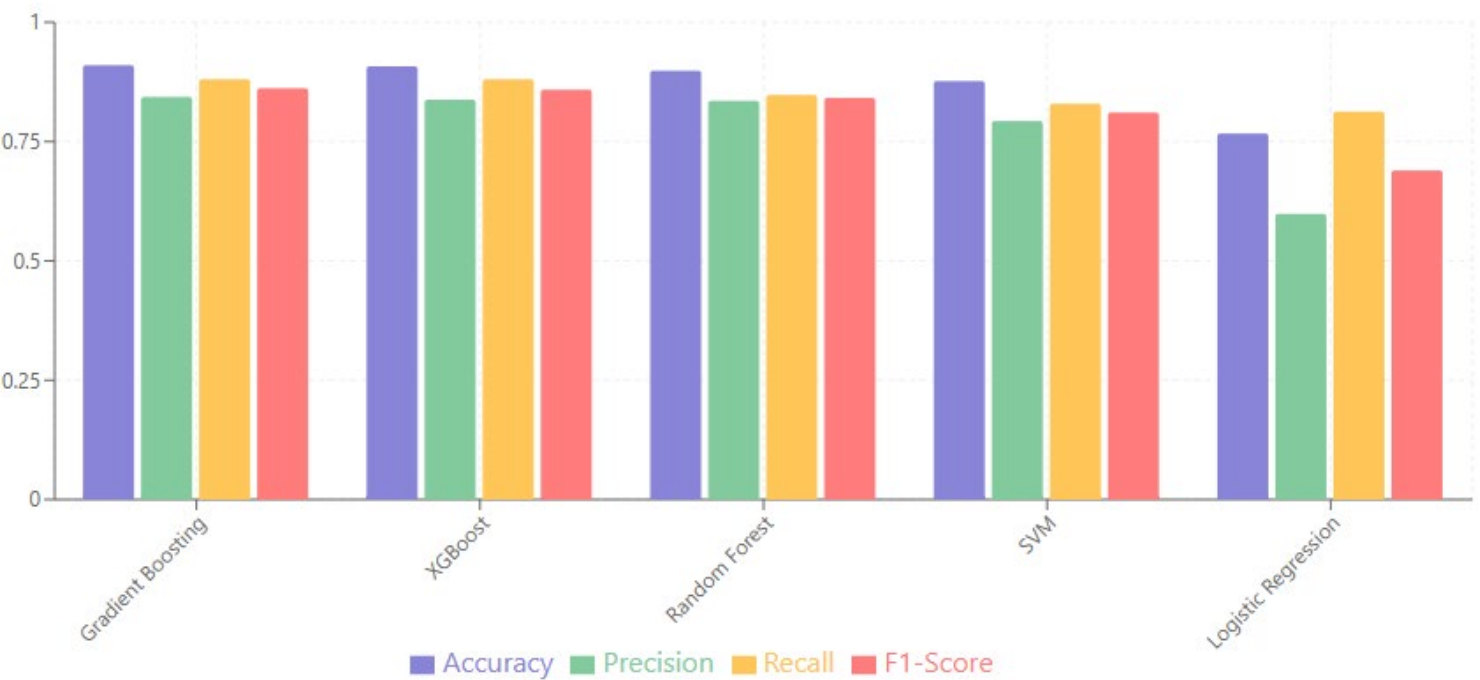
# Model Building for Churn Prediction

- **Feature Selection to Avoid Leakage:**
  - Correlation analysis for churn-related features
  - Removed status-based features (status\_code, status\_description)
  - Checking Circular dependency
  - Retained features providing predictive value without leakage
- **Models Tested:** Logistic Regression, SVM, Random Forest, Gradient Boosting, XGBoost
- **Final Model:** Gradient Boosting Classifier and gives 91% accuracy, 0.86 F1-score



# Model Comparison

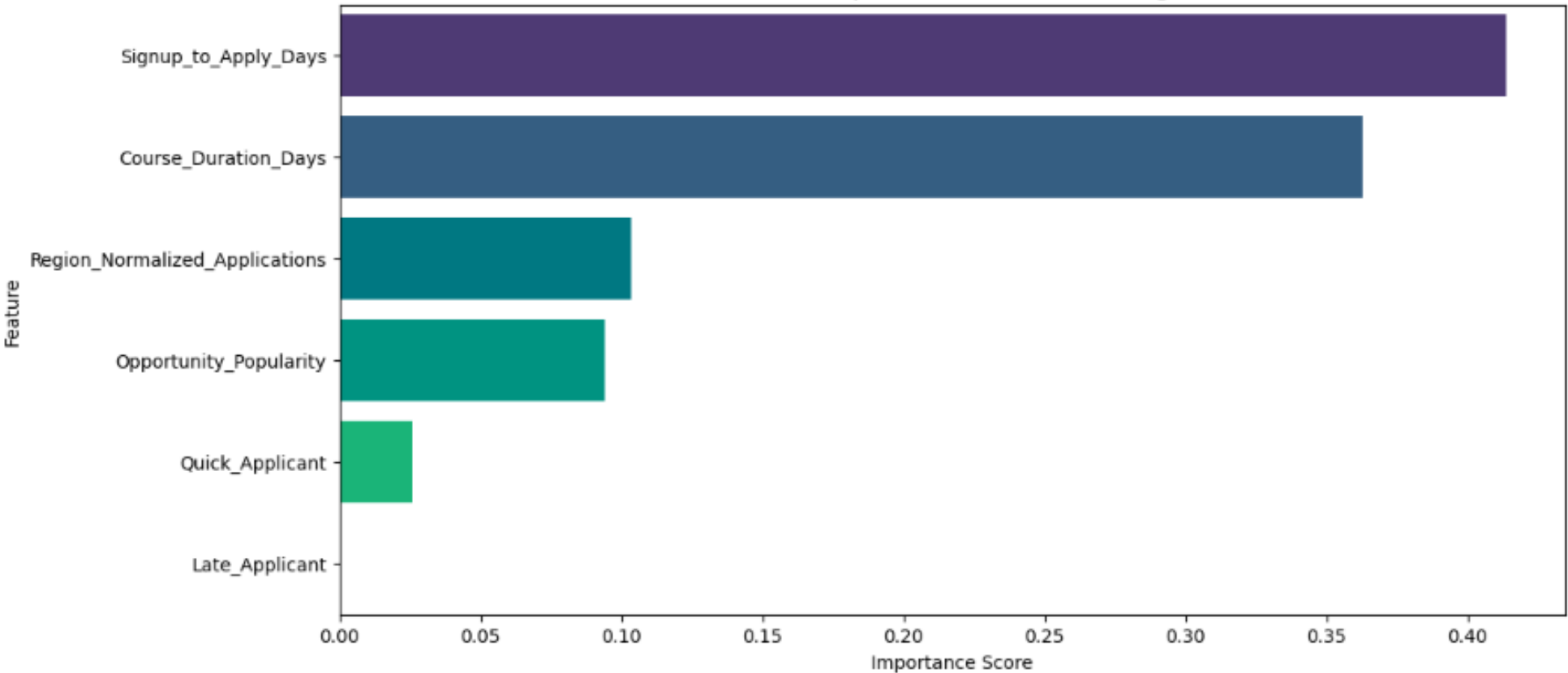
Performance Metrics Comparison



Model Rankings



Feature Importance - Gradient Boosting



# Churn Factors



# Churn Factors & Recommendations

- **Delayed First Application = Early Dropout Risk**

**Strategy:** Automate a multi-email onboarding sequence with a video tour and personalized opportunity suggestions to drive immediate action.

- **Long Course Duration = Higher Overwhelm & Churn**

**Strategy:** Restructure long courses into weekly modules, each with a micro-certificate, to prevent learner overwhelm.

- **Low Regional Application Rate = Localized Disengagement**

**Strategy:** Develop region-specific content and community hubs to increase local relevance and engagement.

- **Chasing Popular Opportunities = Mismatch-Driven Churn**

**Strategy:** Enhance the recommendation algorithm with collaborative filtering to surface trending and peer-approved opportunities.

- **Instant Applications = Rushed Decisions, Higher Regret**

**Strategy:** Improve early onboarding to ensure users understand the platform before applying.

- **Late Applications = Indecision Signals, Likely Churn**

**Strategy:** Use AI-driven prompts to re-engage users who show signs of indecision early in their journey.







# Limitations & Future Plans

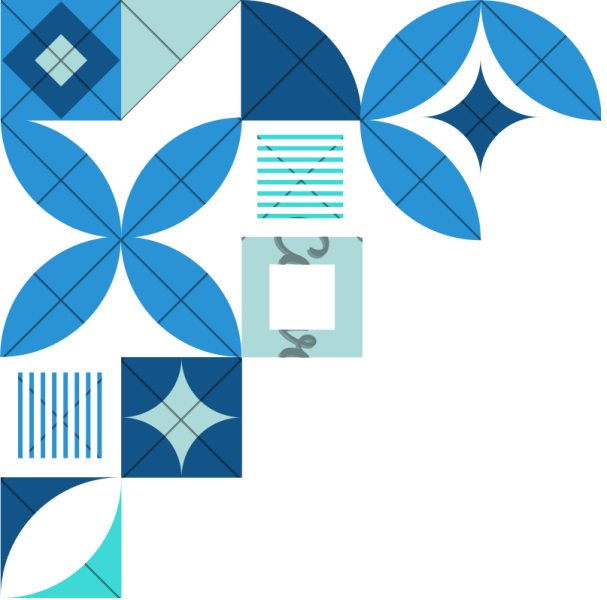
- **Current Gaps:** Missing socioeconomic data, Motivation factors, European representation
- **Future Plans:**
  - Deep learning models
  - Real-time analytics
  - A/B testing
  - Social network analysis

## Conclusion

- **Identified Core Problem:** Mostly churn rate driven by application delays, course length, and engagement gaps.
- **Built Predictive Solution:** Gradient Boosting model achieves 91% accuracy in forecasting student churn.
- **Delivered Actionable Strategies:** Personalized interventions, adaptive content, and targeted campaigns to boost retention.
- **Enabled Proactive Approach:** Shift from reactive support to data-driven, preemptive student success initiatives.







**Thank You!**

