

# Customer Churn Analysis and Retention Strategy

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# Project Idea: Addressing Customer Attrition

## The Problem

Businesses consistently face the challenge of customer attrition, commonly known as churn. This phenomenon directly impacts revenue streams, hinders growth, and erodes customer lifetime value. Proactively identifying high-risk customers before they defect is crucial for sustained business success.



## Proposed Solution



### Predictive Model

Development of a robust predictive model designed to analyse customer data.



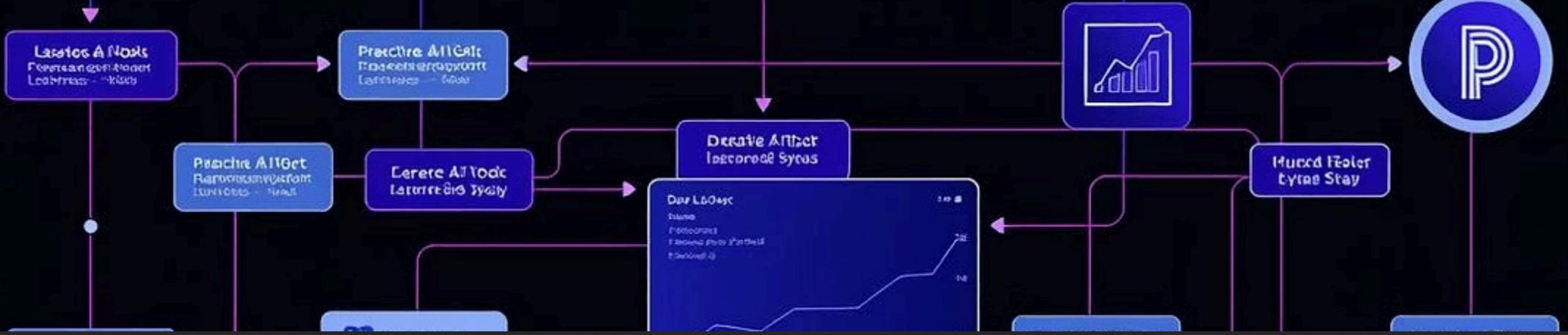
### Churn Forecasting

Accurately forecast churn risk for individual customers based on their behaviour patterns.



### Data-Backed Strategy

Provide a data-driven retention strategy to mitigate identified churn risks effectively.



# Project Wireframe: Conceptual Data Flow & User Interface

Our solution moves beyond mere observation, offering actionable, predictive insights through a deployable application for immediate risk assessment.

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## 1. Input Data

Business users input customer data (tenure, usage, complaints).

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## 2. Process

Deployed model processes the input data in real-time.

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## 3. Output Prediction

Instant churn prediction (High/Low risk) and probability displayed.

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## 4. Action

Users receive recommended retention strategies (e.g., offer discount).

# End Users and Key Features

## Marketing/Retention Team Managers

Need: To identify and prioritise vulnerable customer segments for targeted campaigns.

- **Key Feature:** Comprehensive Model Evaluation Report (Precision/Recall/F1) to ensure reliability for budget allocation decisions.

## Customer Service Representatives (CSRs)

Need: Real-time churn risk assessment during customer interactions to guide conversations.

- **Key Feature:** Intuitive Deployed Model Interface (Streamlit/Flask app) for instant churn predictions.

## Data Analysts/Scientists

Need: To understand the primary drivers of churn and continuously refine predictive capabilities.

- **Key Feature:** Detailed Feature Importance Plots for interpreting customer behaviour patterns and model transparency.



# Data Structure and Flow

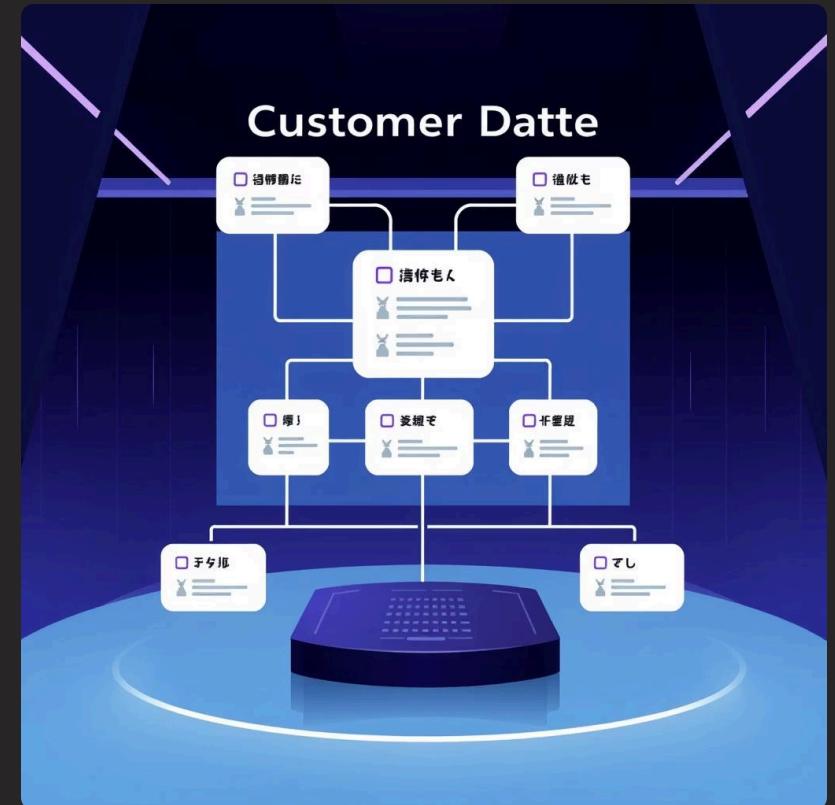
## Database Architecture

Our initial modelling approach assumes a flat-file or CSV-based data architecture, providing flexibility and ease of integration for diverse datasets. This structure is designed for straightforward data handling.

## Key Entities & Relationships

The core of our data model revolves around a single entity: **Customer**. This entity encapsulates critical features such as:

- **customer\_id**: Unique identifier
- **tenure**: Length of customer relationship
- **usage**: Customer activity metrics
- **complaints**: Record of customer issues
- **churn\_status**: The target variable (binary: churned/not churned)

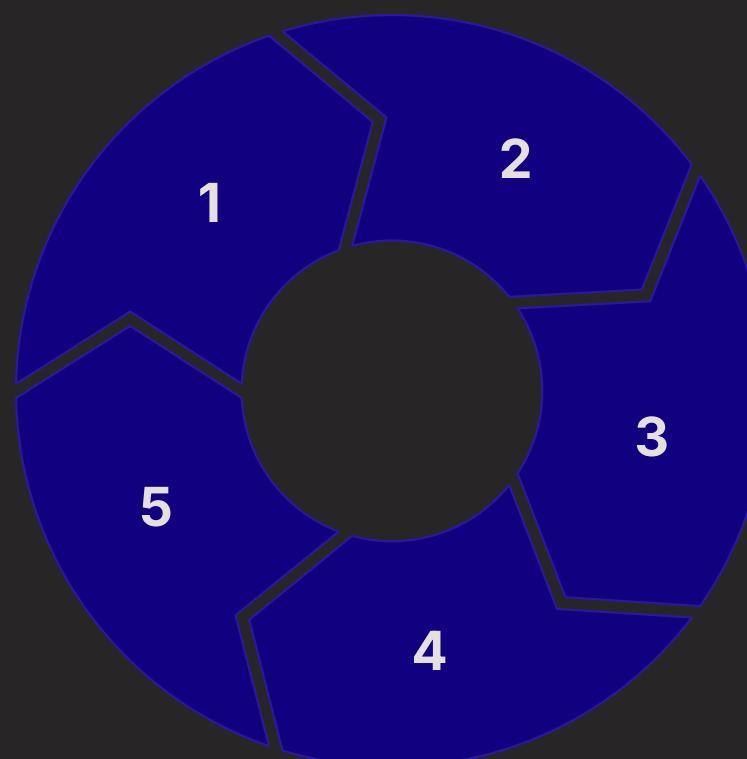


## Data Collection

Utilising public or provided datasets.

## Model Storage

Saving trained models as deployable artifacts.



## Preprocessing

Encoding categorical features, scaling numeric features, handling missing data, and addressing class imbalance.

## Data Split

Dividing data into training and testing sets.

## Model Training

Feeding preprocessed data into chosen algorithms.

# Programming Languages & Frameworks



## Python

The primary language for all ML/Data Science tasks, leveraging its extensive ecosystem.



## Scikit-learn

Essential for implementing foundational ML algorithms such as Logistic Regression and Random Forest.



## XGBoost

Utilised for advanced gradient boosting techniques to enhance model performance and accuracy.



## Streamlit/Flask

Chosen for building a simple, intuitive web application for model deployment and user interaction.



## Google Colab

The collaborative development environment for initial exploration, prototyping, and model training.

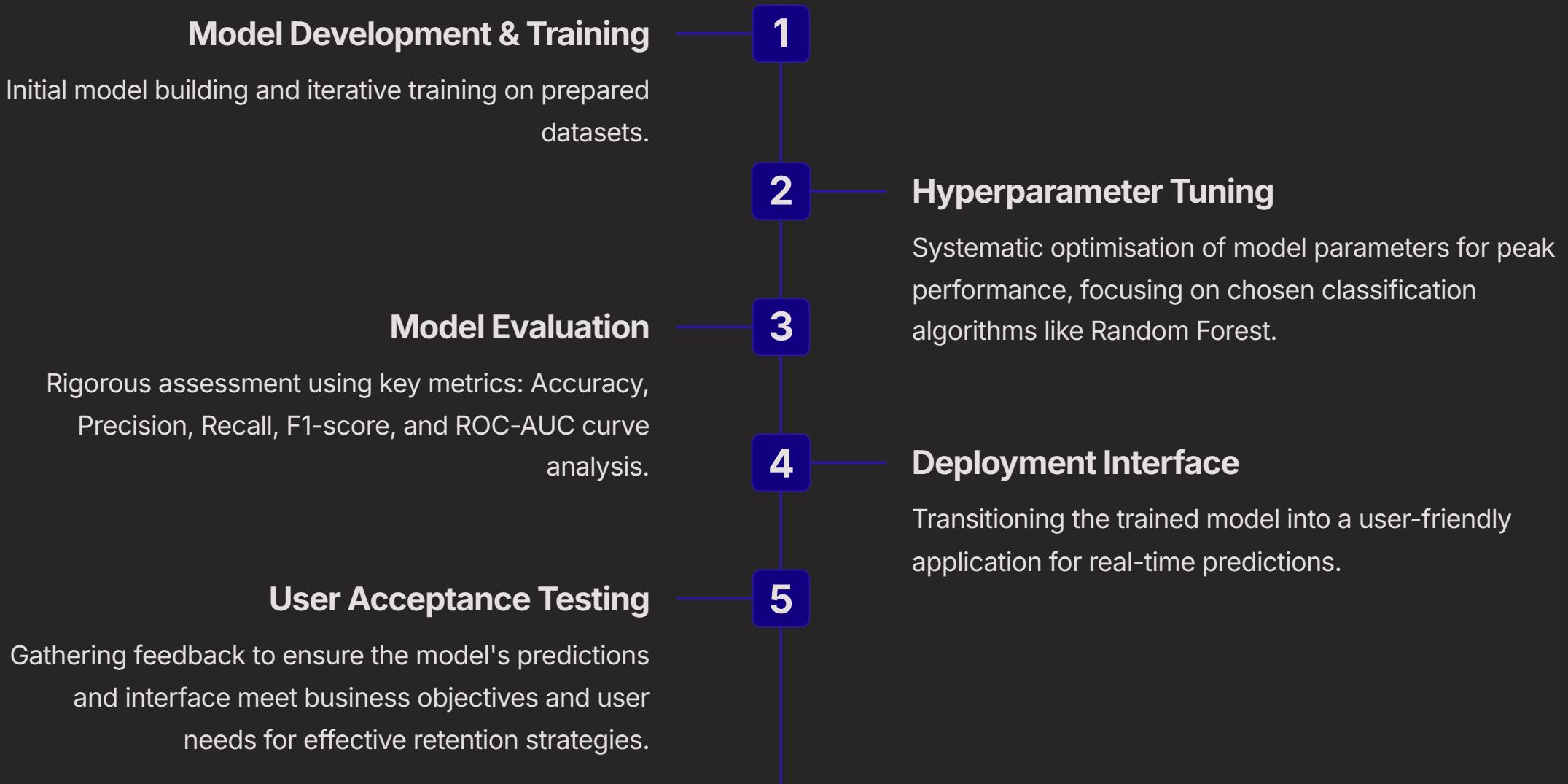


## Supporting Libraries

Including Pandas for data manipulation, and Matplotlib/Seaborn for comprehensive data visualisation.

# Live Application & Testing Phases

Our project is currently in the crucial modelling and deployment phase, with active development on Google Colab. The ultimate goal is a fully functional, deployed model interface.



# Key Deliverables & Project Timeline

## Reports & Documentation

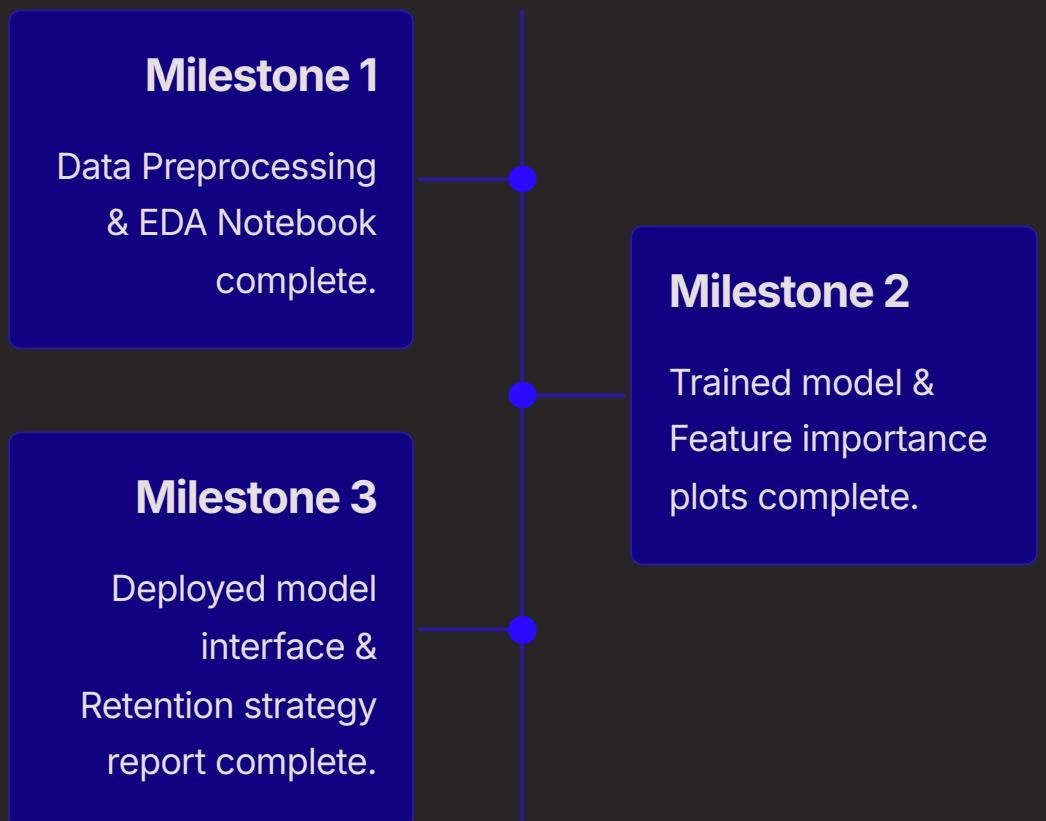
- **EDA Notebook:** Visualisations and insights from exploratory data analysis.
- **Summary Report:** Key customer behaviour insights driving churn.
- **Model Evaluation Report:** Detailed performance metrics and model robustness.
- **Retention Strategy Report:** Data-backed actionable strategies.

## Other Final Products

- Cleaned dataset ready for further modelling.
- Trained model with complete training code.
- Deployed model interface for user interaction.



## Timeline for Deliverables (Milestones)



# Project Team & Roles



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## Key Responsibilities

Roles within the team were dynamically distributed, encompassing Data Collection, Preprocessing, Model Development, Deployment Engineering, and Report Writing, ensuring comprehensive coverage of all project phases.

## Collaboration Methods

Our team leveraged a collaborative work environment, primarily Google Colab, to facilitate seamless integration of efforts and real-time project progression.

# Thank You

We appreciate your time and attention to our project.

## Contact Information

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We welcome any questions or feedback you may have.

