





CUSTOMER CHURN PREDICTION USING MACHINE LEARNING

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1. Problem Statement

To develop a machine learning model that can **predict whether a** customer is likely to churn (leave the service)

2. Objectives of the Project

To predict whether a customer is likely to **churn** (leave a company) using historical data, so that the business can take proactive steps to **retain** them.

3. Scope of the Project

The **scope of Customer Churn Prediction using Machine Learning** is broad and highly valuable across industries

4. Data Sources

- i. Customer Demographics
- ii. Account Information
- iii. Payment and Billing History
- iv. Usage and Engagement Data
- v. Customer Support Interactions
- vi. Marketing & Communication History







5. High-Level Methodology

1. Data Collection

Obtain data from CRM systems, logs, billing systems, or public datasets like the Telco Churn Dataset.

2. Data Preprocessing

- ·Handle missing values
- Convert categorical variables using encoding
- Scale numeric features (if needed)
- Clean or drop irrelevant columns (e.g., customer ID)

3. Feature Engineering

- Create new features (e.g., engagement_score, tenure_bucket)
- Transform skewed variables
- Bin continuous features for decision trees

4. Model Selection

Common models used:

- Logistic Regression (baseline)
- Decision Tree / Random Forest

5. Model Evaluation

- Use train-test split or cross-validation
- •Metrics:
 - Accuracy (overall correctness)
 - Precision & Recall (important for imbalanced classes)
 - F1 Score
 - ROC-AUC (to evaluate probability outputs)

6. Interpret Results

- Use SHAP or feature importance to explain predictions
- •Identify top factors that lead to churn (e.g., short tenure, high charges)

7. Actionable Steps

- Target high-risk customers with:
 - Discounts
 - Personalized offers
 - Loyalty programs
 - Improved service quality







6. Tools and Technologies

1. Data Collection & Storage

These tools help gather and store raw data about customers, transactions, and interactions:

- •SQL / MySQL / PostgreSQL For querying customer databases
- Excel / CSV Files For manual or exported data
- CRM Tools (e.g., Salesforce, HubSpot) Customer profiles and behavior
- •Cloud Platforms (e.g., AWS S3, Google BigQuery) Scalable storage
- •APIs To pull in data from external systems or application

2. Data Preprocessing & Analysis

Used for cleaning, exploring, and preparing data before modeling:

- Python Most popular language for ML tasks
- Pandas For data manipulation and preprocessing
- NumPy For numerical operations
- Matplotlib / Seaborn For data visualization

3. Machine Learning Modeling

Tools and libraries for building and evaluating predictive models:

- •Scikit-learn Core ML models (Logistic Regression, Random Forest, etc.)
- XGBoost / LightGBM / CatBoost High-performance gradient boosting libraries
- TensorFlow / Keras / PyTorch For deep learning models

4. Feature Engineering & Selection

Used to create and select the best features to improve model performance:

- •Scikit-learn Has built-in tools for feature selection
- •Featuretools For automated feature engineering
- •SHAP / LIME For interpreting feature importance







5. Model Evaluation & Tuning

Helps fine-tune models and understand their performance:

- •Scikit-learn Cross-validation, confusion matrix, ROC curves
- Optuna / Hyperopt For hyperparameter tuning
- Yellowbrick For visual model diagnostics and performance plots

6. Deployment

Used to deploy churn models into production:

- •Flask / FastAPI Lightweight Python APIs for serving models
- Docker For packaging ML models as containers
- Streamlit / Gradio For building interactive dashboards

7. Reporting & Business Intelligence

Used to present churn insights to non-technical stakeholders: **Tableau / Power BI** – Visual dashboards for business teams **Looker / Metabase** – Data exploration for decision-makers **Excel** – Still common for quick reports and summaries

7. Team Members and Roles

- 1. K.ABIKA (Data collection, cleaning)
- 2. T.R.VIGNESH (tools and technology)
- 3. S.KIRUBANANTHAM (high level methodology)