<u>Final Internship Project Report</u> <u>Celebal Technologies</u>

Title: Customer Lifetime Value Prediction

Submitted by:

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1. Abstract

This project aims to predict the Customer Lifetime Value (CLV) using historical online retail data. The CLV represents the total monetary value a business expects to earn from a customer throughout their relationship. This analysis helps businesses improve decision-making in marketing, sales, and customer retention strategies. Machine Learning techniques were used for prediction, and visual insights were presented using Tableau and Streamlit.

2. Problem Statement

In the e-commerce domain, understanding customer behavior and predicting their future value is crucial. The project focuses on predicting the future value of a customer to a business based on their past purchase history and behavioral patterns.

Goal: Predict the monetary value (CLV) for each customer using recency and frequency metrics.

3. Dataset Description

- Dataset Source: Kaggle Online Retail II Dataset
- Files Used:
 - online_retail_2009.xlsx
 - online_retail_2010.xlsx
- Key Features:
 - Invoice, StockCode, Description, Quantity, InvoiceDate, UnitPrice, Customer ID, Country

4. Data Preprocessing & Feature Engineering

- Combined 2009 and 2010 data.
- Removed null and duplicate values.
- Excluded cancelled invoices (starting with 'C').
- Calculated RFM (Recency, Frequency, Monetary) metrics:

Recency: Days since last purchase

o **Frequency**: Number of orders

o **Monetary**: Total amount spent

5. Modeling

• Features Used: Recency, Frequency

• Target: Monetary Value (CLV)

• Algorithms Tried:

Linear Regression

Gradient Boosting

Random Forest Regressor

Best Model: Random Forest Regressor

• MSE: ~155,772

• R² Score: ~0.576

The model was saved using pickle as clv model.pkl for deployment.

6. Deployment using Streamlit

A Streamlit web app was developed to make predictions based on user input:

- Users input Recency and Frequency
- The model predicts the CLV instantly

Key Features:

- Clean UI with input form
- Visualization of customer trends
- Top customers by predicted CLV

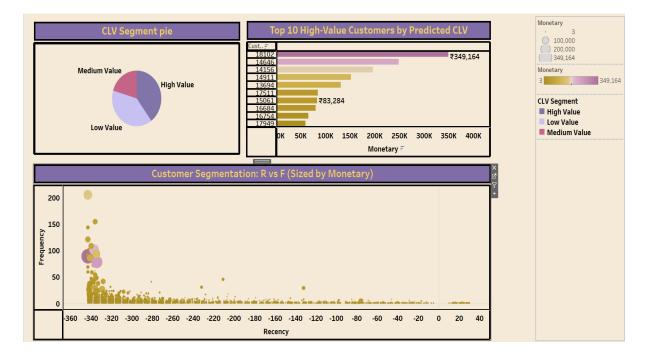
• Recency vs CLV and Frequency vs CLV graphs

7. Tableau Dashboards

To support the analysis, Tableau dashboards were built on the same dataset:

Customer Segmentation Dashboard

- RFM Score Distribution
- Customer categories: Loyal, At Risk, New, etc.



8. Conclusion & Future Work

This project successfully predicts Customer Lifetime Value using Recency and Frequency features. The deployed app and visual dashboards allow business users to make data-driven decisions.

Future Enhancements:

- Include demographic or session data for improved predictions
- Model customer churn and retention scores
- Integrate with live transactional systems

9. References

- Kaggle: https://www.kaggle.com/datasets/lakshmi25npathi/online-retail-dataset
- Scikit-learn Documentation
- Streamlit Docs
- Tableau Public: <u>Dashboard</u>