

E-commerce Cart Abandonment Analysis

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1. Overview

This project addresses the challenge of cart abandonment in e-commerce using machine learning. By analyzing customer session behavior, we build and deploy a predictive model that estimates whether a user is likely to abandon their cart or complete the purchase. The model is integrated into a Streamlit web application for real-time decision-making, enhancing user engagement and boosting conversion rates.

2. Introduction

In today's digital economy, online retailers face a major hurdle in converting potential buyers into customers. A significant percentage of users abandon their shopping carts before completing a purchase. Predictive analytics powered by machine learning can provide insights into user behavior and allow for preemptive engagement strategies. This project aims to build a complete AI/ML pipeline to detect and respond to potential cart abandonment.

3. Problem Statement

High cart abandonment rates on the e-commerce platform result in lost sales and reduced customer lifetime value.

The objective is to leverage behavioral session data to build a predictive model capable of identifying abandonment-prone sessions.

4. Business Value

- Improve overall conversion rate and revenue
 - Trigger timely interventions like discounts or reminders
 - Automate marketing and retention workflows
 - Understand user behavior patterns to optimize UX
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5. Dataset Description

The dataset used consists of anonymized e-commerce session logs with the following fields:

Feature	Description
session_id	Unique identifier for each session
pages_visited	Number of pages visited
time_on_site	Time spent on site (in seconds)
cart_value	Total value of items in the cart (USD)
abandoned	Target variable (1 = abandoned, 0 = purchase)

There are no missing values. Class imbalance was handled during model training using `class_weight='balanced'`.

6. Methodology

6.1 Data Preprocessing

- Verified data types, ranges, and nulls
 - Scaled continuous features for clustering
 - Used class weighting for model imbalance
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6.2 Exploratory Data Analysis (EDA)

- Visualized relationships between time, cart value, and abandonment
 - Used histograms, boxplots, and pairplots
 - Identified lower engagement as a leading cause of abandonment
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6.3 Model Building

- Models used:
 - Logistic Regression
 - Random Forest Classifier
 - Split: 80% train / 20% test
 - Used sklearn pipeline for reproducibility
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6.4 Model Evaluation

- Evaluation Metrics:
 - Accuracy, F1-score, Precision, Recall
 - ROC Curve and AUC
 - Feature Importance:
 - time_on_site > pages_visited > cart_value
 - Random Forest outperformed Logistic Regression with ~69% accuracy
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6.5 Unsupervised Learning

- Applied KMeans to identify user clusters
- Used PCA to reduce dimensions to 2D
- Helped visualize session behavior groups for segmentation

7. Deployment

The trained Random Forest model was saved as `final_model.pkl` and used in a **Streamlit app** for live predictions.

App Features:

- Inputs: pages visited, time on site, cart value
 - Outputs:
 - "Likely to Abandon Cart" or "Likely to Complete Purchase"
 - Probability Score of prediction
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8. Results and Observations

- Random Forest performed best on evaluation metrics
 - Sessions with low engagement had significantly higher abandonment risk
 - Clustering revealed distinct behavior segments
 - Model predictions aligned with expected patterns from EDA
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9. Conclusion

This project demonstrates an end-to-end ML workflow to solve a real-world business problem. The deployed app allows stakeholders to identify and act on at-risk sessions in real-time. The model can be further improved by incorporating features like user device, traffic source, or browsing history.

10. Tools and Technologies Used

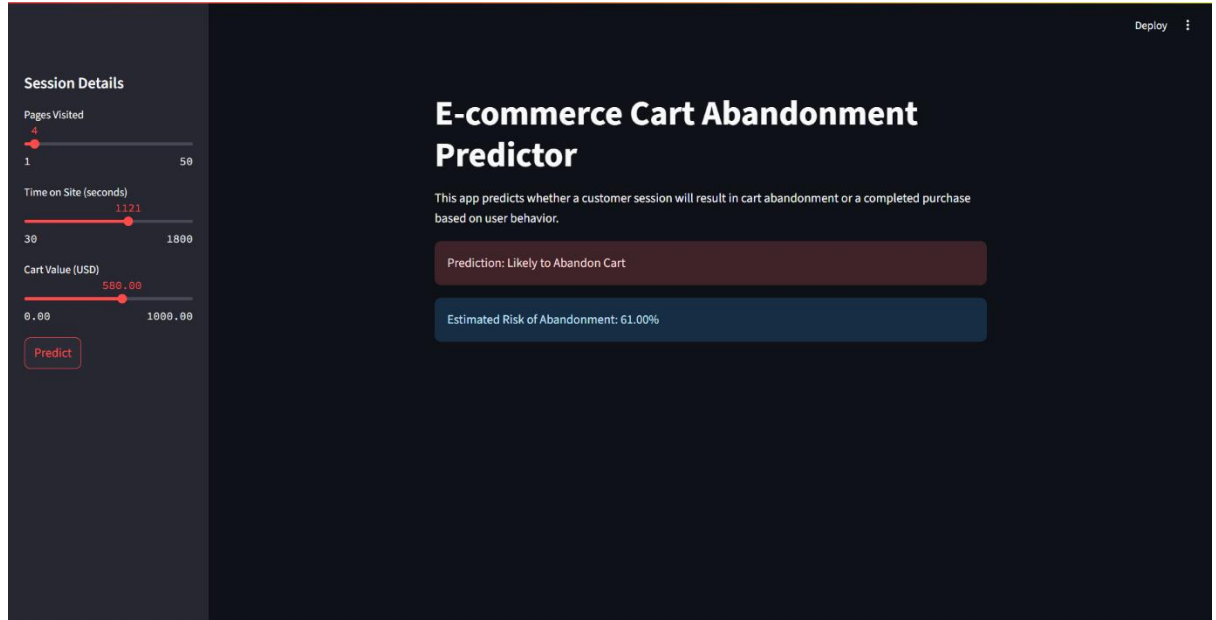
- **Language:** Python
 - **Libraries:** pandas, numpy, matplotlib, seaborn, scikit-learn, joblib
 - **Deployment:** Streamlit
 - **Environment:** Jupyter Notebook, VS Code
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11. References

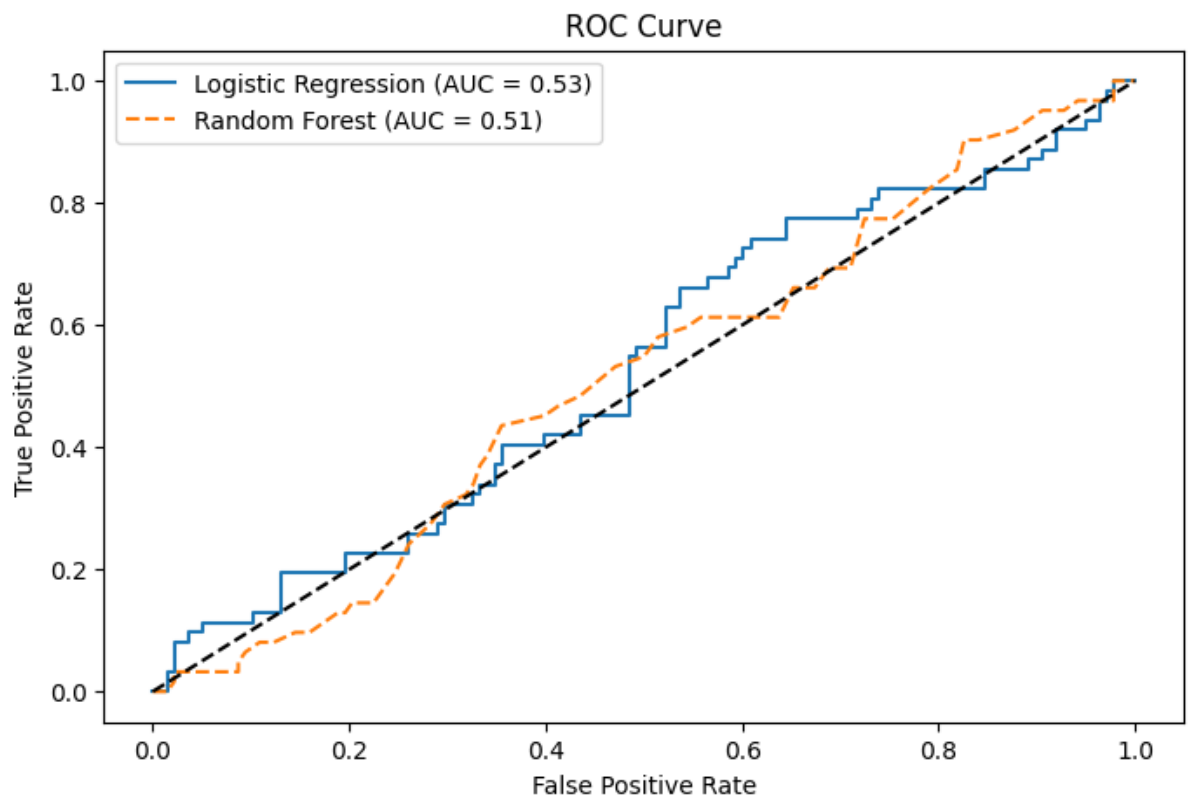
- scikit-learn Documentation: <https://scikit-learn.org/>
- Streamlit Docs: <https://docs.streamlit.io/>
- Kaggle e-commerce behavior datasets
- Research blogs on cart abandonment prediction

12. Appendix

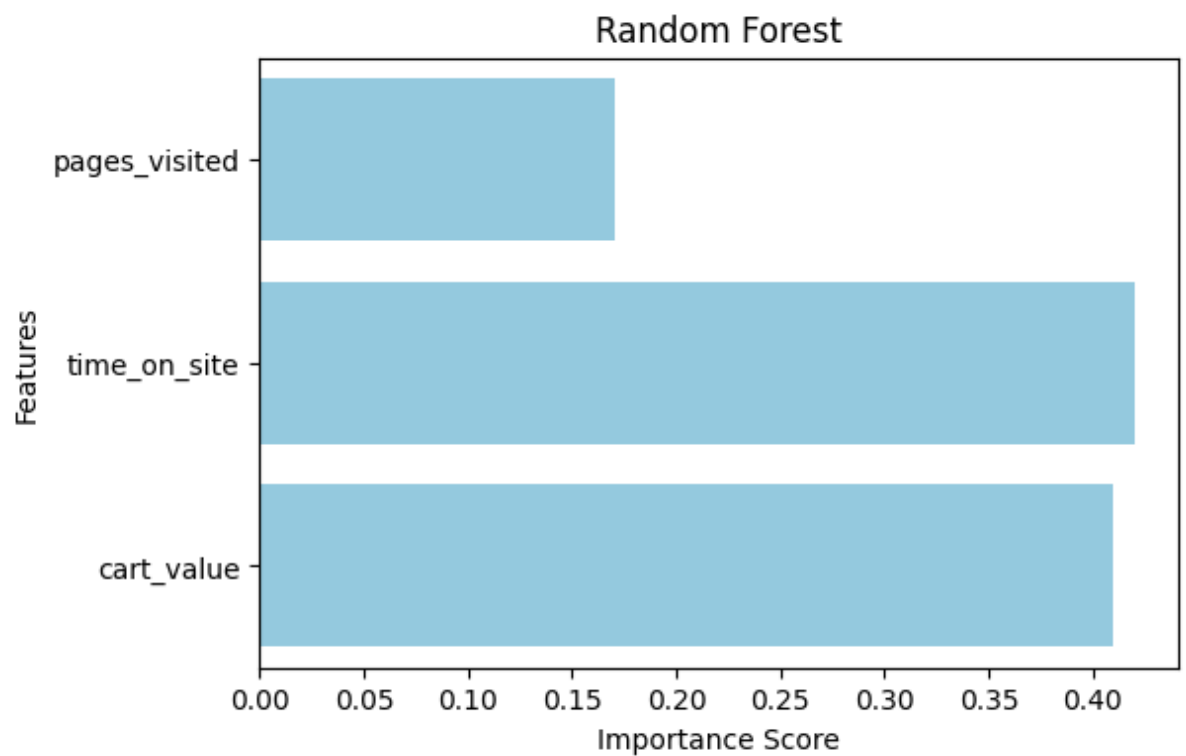
- *Streamlit App Interface*



- *ROC Curve*



- *Feature Importance Plot*



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