

K.RAMAKRISHNAN COLLEGE OF TECHNOLOGY (AUTONOMOUS), TRICHY



CLV PULSE – A DYNAMIC CUSTOMER LIFETIME VALUE PREDICTOR

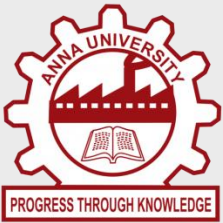
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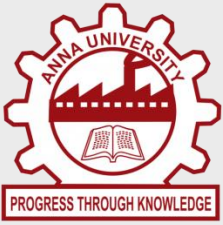


PRESENTATION OVERVIEW



11/5/2024

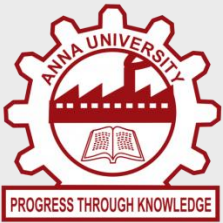
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INTRODUCTION



Customer Lifetime Value (CLV) is an essential metric that helps businesses understand the total revenue expected from a customer over the duration of their relationship. Accurate CLV predictions allow companies to tailor their marketing efforts, optimize customer service, and allocate resources effectively to maximize profitability. CLV Pulse is an innovative tool designed to provide dynamic, real-time CLV predictions using advanced machine learning algorithms and continuous data streams, offering businesses a cutting-edge solution for customer value management.



PROBLEM IDENTIFICATION



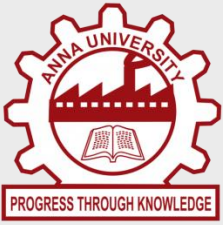
The primary issue with traditional CLV models is their static nature. These models typically rely on historical data, which may not reflect current customer behaviors or market conditions. Consequently, businesses using these models may encounter several problems:

Inaccurate Predictions: Static models can become outdated quickly, leading to erroneous forecasts.

Inefficient Resource Allocation: Without up-to-date information, businesses may misallocate marketing budgets and customer service efforts.

Suboptimal Customer Retention: Failure to adapt to changing customer behaviors can result in decreased retention rates.

Lack of Responsiveness: Traditional models are slow to react to new data, delaying strategic adjustments.



OBJECTIVE



The goal of CLV Pulse is to develop a dynamic, real-time CLV prediction system that continuously updates based on current data. This system aims to:

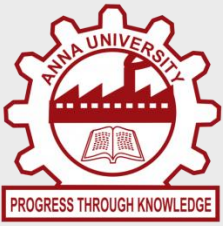
Provide accurate, up-to-date CLV estimates.

Enhance decision-making capabilities.

Improve customer retention strategies.

Optimize marketing and resource allocation.

Adapt quickly to changes in customer behavior and market trends.



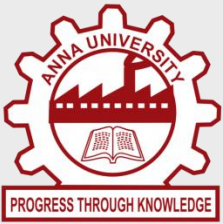
ABSTRACT



Introducing 'CLV Pulse', an innovative machine learning model revolutionizing how businesses understand and maximize Customer Lifetime Value (CLV). Similar to how a heartbeat indicates the health of a person, CLV Pulse provides a 'pulse' on the health of customer relationships. Imagine an e-commerce giant implementing CLV Pulse to analyze its vast reservoir of customer data dynamically. By leveraging advanced predictive analytics, CLV Pulse tracks critical metrics such as purchase frequency, average order values, and customer engagement levels in real-time, unveiling actionable insights into customer segments, their evolving preferences, and potential churn risks. This dynamic predictive model empowers businesses to tailor marketing campaigns, fine-tune product offerings, and implement targeted retention strategies with unprecedented precision, ensuring customer satisfaction and maximizing revenue. With its intuitive dashboard and sophisticated machine learning algorithms, CLV Pulse is not just a data analysis tool; it's a strategic asset for impactful customer relationship management. Audiences seeking transformative solutions applaud CLV Pulse for its ability to deliver actionable insights in a user-friendly interface. Simultaneously, for investors, CLV Pulse represents a forward-thinking technology poised to disrupt industries by enabling businesses to make data-driven decisions, optimize resources, and drive sustainable growth.

LITERATURE SURVEY

S.No	Paper Name	Authors	Insights	Drawbacks
1	Foundational Models for Customer Lifetime Value (CLV)	Reinartz and Kumar	Established basic models for calculating CLV using historical purchase data	None specified
2	Probabilistic Models for Customer Lifetime Value (CLV)	Venkatesan and Kumar	Advanced probabilistic models that consider customer transaction patterns	None specified
3	Inclusion of Retention Rates and Acquisition Costs in CLV	Gupta et al.	Highlighted the importance of including retention rates and acquisition costs in CLV calculations	None specified
4	Application of Machine Learning in CLV Prediction	Predictive Analytics	Explored the application of machine learning algorithms in predicting customer behavior	None specified
5	Real-Time Analytics for Dynamic CLV Calculation	Real-Time Data Processing	Emphasized the advantages of real-time analytics in dynamically adapting to changing business environments	None specified



EXISTING SYSTEM



Base Domain :

Machine Learning

Tools Used:

OLTP, Segmentation, ML algorithms

Merits:

- Faster execution
- Ease of use
- Good accuracy of results

Demerits:

- Not suitable for real time data
- Insight generation is less
- Provides a controlled amount of predictions



PROPOSED SYSTEM



Base Domain:

Machine Learning

Tools Used:

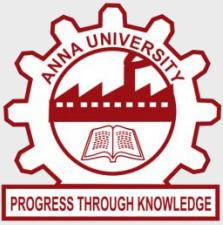
RapidMiner Studio, python, Excel, Database Tools

Advancement of Technology:

- Operates on vast amount of data
- Generates insights in real time
- Self Updating system
- Provides more reliable information

Main Goals:

- To work on continuous data seeding
- To help in business growth especially in e-commerce



DISADVANTAGE OF EXISTING SYSTEM



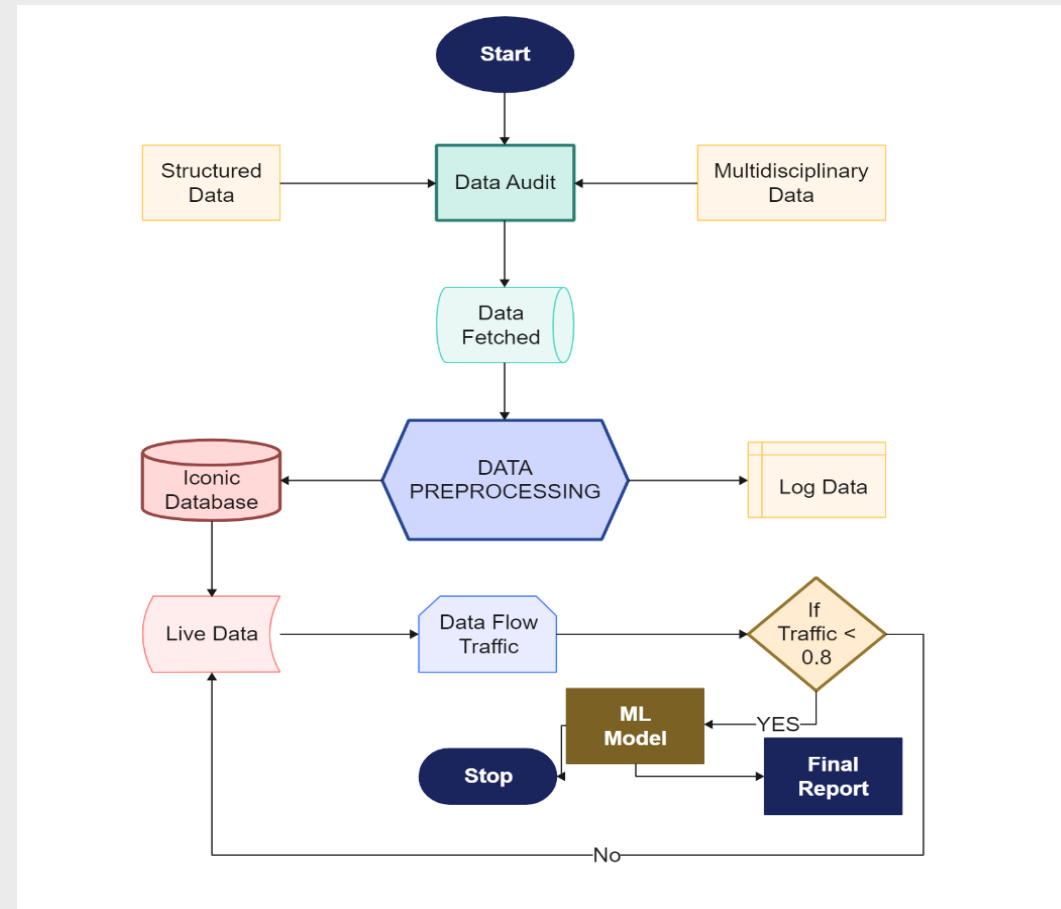
Inflexibility: Static models cannot adjust to real-time changes in customer behavior.

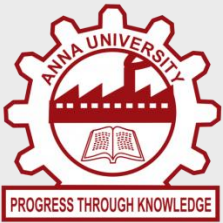
Inaccurate Forecasts: Predictions can quickly become outdated, leading to poor strategic decisions.

High Resource Demand: Significant manual effort is needed to maintain and update models.

Limited Personalization: Generalized predictions fail to capture the unique behaviors and preferences of individual customers.

BLOCK DIAGRAM OF PROPOSED MODEL





EXPLANATION OF PROPOSED MODEL



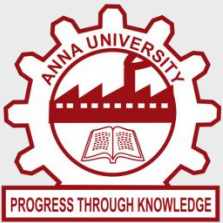
Data Collection: Real-time data from various customer touchpoints, such as e-commerce transactions, website visits, social media interactions, and customer service interactions, is aggregated.

Data Processing: The raw data is cleaned and preprocessed to remove inconsistencies and ensure it is suitable for analysis. This includes normalization, handling missing values, and feature engineering.

Machine Learning Integration: Advanced machine learning algorithms, such as ensemble methods, neural networks, and time series analysis, are employed to identify patterns and predict CLV. The system continuously learns from new data, improving its predictions over time.

Real-Time Dashboard: Provides a comprehensive view of current CLV predictions, historical trends, and future projections. Users can interact with the dashboard to drill down into specific customer segments or individual customer profiles.

Feedback Mechanism: The system incorporates feedback from users and new data inputs to continually refine and update the predictive models, ensuring they remain accurate and relevant.



ADVANTAGES



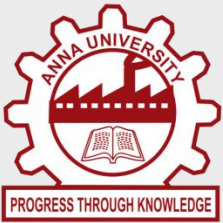
Real-Time Adaptation: The system continuously updates CLV predictions based on the latest data, ensuring they remain accurate and relevant.

Enhanced Accuracy: The use of advanced machine learning algorithms results in more precise predictions compared to traditional models.

Improved Decision Making: Real-time, accurate CLV predictions provide businesses with actionable insights for optimizing marketing efforts, customer retention strategies, and resource allocation.

Resource Efficiency: Reduces the need for manual data processing and intervention, allowing businesses to focus on strategic activities.

Personalization: Tailored predictions that reflect individual customer behaviors and preferences, enabling more personalized marketing and customer service initiatives.



RESULT & DISCUSSION



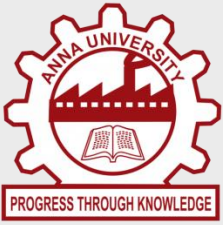
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CONCLUSION



CLV Pulse represents a significant advancement in the field of customer value prediction. By leveraging real-time data and machine learning, it offers businesses a powerful tool for understanding and maximizing the lifetime value of their customers. This dynamic approach ensures that businesses can make informed decisions, optimize resource allocation, and ultimately enhance profitability in an ever-evolving marketplace. CLV Pulse not only addresses the limitations of traditional CLV models but also sets a new standard for predictive analytics in customer relationship management.