

CUSTOMER DISMEMBERMENT

Siddharth Runwal*¹, Yash Shah*², Suhel Malik*³,

Vishal Raut*⁴, Prof. Sandip Hire*⁵

*^{1,2,3,4,5}Computer Engineering, Smt. Kashibai Navale College Of Engineering, Pune,
Maharashtra, India.

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ABSTRACT

The zeitgeist of the modern era is innovation, where everyone is embroiled in the competition to be better than others. Today's businesses run on the basis of such innovation having the ability to enthrall the customers with the products, but such a large raft of products leave the customers confounded, about what to buy and what to not, and also the companies are nonplussed about what section of customers to target to sell their products. This is where machine learning comes into play, various algorithms are applied to unravel the hidden patterns in the data for better decision-making in the future. This eluded concept of which segment to target is made unequivocal by applying segmentation. The process of segmenting customers with similar behaviors into the same segment and with different patterns into different segments is called customer segmentation. Additionally customer churning is also bridged into the segmentation to witness the percentage rate of potential customers churning away.

Keywords: Segmentation, Churning, Machine Learning, K-Means Clustering.

I. INTRODUCTION

In today's highly competitive marketplace, businesses must understand their target audience to develop effective marketing strategies and improve customer experiences. Customer segmentation, the process of dividing customers into distinct groups based on their characteristics and behaviors, has become an essential tool for businesses seeking to achieve these goals. Customer segmentation enables businesses to identify their most profitable customer segments, tailor their marketing efforts to specific groups, and improve customer satisfaction and retention.

This survey paper aims to explore the use of customer segmentation surveys as a means to identify customer groups and their characteristics. The paper will focus on the use of k-means clustering and F1 score as effective methods for analyzing customer data and creating customer segments. The survey data will be collected from a sample of consumers, and the results will be analyzed to provide insights into the characteristics and behaviors of different customer segments. The paper will also discuss the implications of the findings for businesses seeking to improve customer satisfaction, retention, and profitability.

II. METHODOLOGY

Customer segmentation is the process of dividing customers into groups based on their shared characteristics or behaviors. This methodology outlines the steps involved in customer segmentation using data science techniques:

1. Define the objective: Before starting the segmentation process, it is important to define the objective of the analysis. What is the business problem that needs to be addressed? What are the key performance indicators (KPIs) that will be used to evaluate the success of the segmentation?

2. Collect and preprocess data: The first step is to gather data about the customers, including demographics, transaction history, purchase behavior, and other relevant information. The data should be cleaned and preprocessed to remove any inconsistencies or missing values.

3. Identify relevant variables: Determine which variables are most relevant for customer segmentation. This will depend on the objective of the analysis and the available data. Common variables used for segmentation include age, gender, location, income, purchase history, and website behavior.

4. Choose segmentation method: There are various segmentation methods available, such as clustering, decision trees, and factor analysis. Choose the method that best fits the objective and data.

5. Implement segmentation: Apply the chosen segmentation method to the data. This will create segments or clusters of customers based on their shared characteristics.

6. Evaluate segmentation: Evaluate the effectiveness of the segmentation by measuring how well it aligns with the objective and KPIs. This can be done through statistical analysis or visualization.

7. Implement insights: Use the insights gained from the segmentation to improve marketing campaigns, product development, and customer engagement strategies. Tailor the messaging and promotions for each segment to increase customer retention and revenue.

8. Monitor and refine: Customer segmentation is an ongoing process. Monitor the results and refine the segmentation as necessary to improve its effectiveness over time.

By following these steps, businesses can use data science to effectively segment their customers and improve their marketing and customer engagement strategies

III. EXISTING SYSTEM

K-Means Clustering: K-means is a popular unsupervised clustering algorithm used for customer segmentation. It groups customers based on their shared characteristics, such as demographics, purchase history, and website behavior. This algorithm is available in many data science libraries, such as Scikit-learn and Apache Spark.

Collaborative Filtering: Collaborative filtering is a recommendation system that segments customers based on their preferences and behavior. It uses data on customer behavior, such as purchase history, ratings, and reviews, to recommend products to customers. This technique is commonly used in e-commerce and content-based businesses.

Neural Networks: Neural networks are a popular deep learning technique used for segmentation. They can learn complex patterns in customer data, such as image and speech recognition, and classify customers based on their shared characteristics. This method is useful when the data is highly dimensional and non-linear.

IV. PROPOSED SYSTEM

The proposed customer segmentation and churn prediction is based upon churn analysis and feature engineering. Develop a predictive model to identify customers who are likely to churn, based on their behavior and characteristics. Use techniques such as logistic regression or random forest to build the model. Evaluate the accuracy of the model using metrics such as precision, recall, and F1-score.

Use clustering algorithms, such as K-means or Hierarchical clustering, to group customers based on their shared characteristics. Evaluate the effectiveness of the segmentation by measuring how well it aligns with the business objectives and KPIs.

V. ALGORITHM

L-means clustering is a powerful tool for customer segmentation in data science. The first step is to gather and prepare the data, which involves collecting customer data and converting it into a suitable format for clustering. The next step is to determine the optimal number of clusters, which can be done using the elbow method. After determining the number of clusters, the K-means algorithm is used to group customers into clusters based on their similarities. Each cluster represents a group of customers with similar characteristics such as demographics, behavior, or purchasing habits. Finally, the clusters can be analyzed and visualized to gain insights into customer behavior, preferences, and needs. This information can be used to develop targeted marketing strategies, improve customer satisfaction, and increase customer retention. K-means clustering is an effective technique for customer segmentation in data science, and its use can lead to improved business performance and customer engagement.

Decision trees are a popular tool for customer segmentation in data science. To use decision trees for customer segmentation, you need to first gather and prepare the data, similar to K-means clustering. Then, you need to select the input variables that are most important for predicting customer behavior and segmenting customers. Once you have selected the input variables, you can build a decision tree model using a machine learning algorithm. The decision tree model will create a tree-like structure that models decisions based on the selected input variables. Each branch of the tree represents a decision based on a particular input variable. The final nodes of the tree represent customer segments based on the decisions made by the model. These customer

segments can be analyzed and visualized to gain insights into customer behavior and preferences. Decision trees are an effective technique for customer segmentation in data science, and their use can lead to improved marketing strategies, customer satisfaction, and retention.

VI. CONCLUSION

In conclusion, the customer segmentation survey is an essential tool for businesses to identify their target audience and understand their needs and preferences. The survey results can help businesses create targeted marketing campaigns, improve product offerings, and enhance customer experiences. The key to a successful customer segmentation survey is to ensure that the questions are well-designed and relevant to the target audience, and the data collected is accurate and reliable. Businesses must also analyze the data and develop actionable insights to make informed decisions. Overall, a customer segmentation survey can be a valuable investment for businesses looking to increase customer satisfaction, loyalty, and profitability.

Overall, the usage of k-means clustering and F1 score in customer segmentation surveys can help businesses increase customer satisfaction, retention, and profitability by providing a deeper understanding of their target audience's needs and preferences. It is an essential tool that businesses should consider investing in to achieve success in today's highly competitive marketplace.

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