

# Optimizing Customer Purchasing Patterns

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# Project Overview

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- ✓ Objective: Improve the accuracy of customer income level and purchasing pattern estimation to enable more targeted marketing strategies.
- ✓ Challenge: Existing models for estimating customer income levels and purchasing patterns were inaccurate, leading to suboptimal marketing strategies.
- ✓ Expected Outcomes:
  - Enhanced accuracy in estimating customer income and purchasing patterns.
  - Improved ability to target marketing efforts effectively.
  - Increased customer engagement and conversion rates through personalized marketing strategies.

# Project Overview

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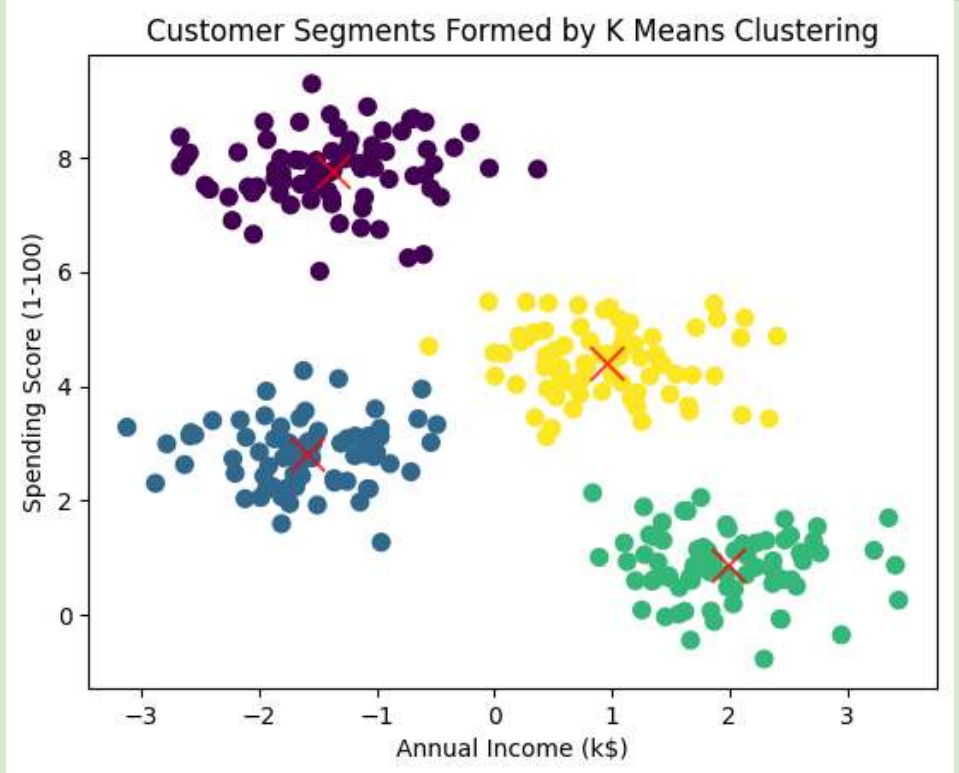
- ✓ Data Collection: Aggregated data from CRM and sales databases.
- ✓ Data Cleaning: Removed duplicates and handled missing values to ensure data integrity.
- ✓ Feature Engineering: Created new features such as average purchase frequency and total spend.
- ✓ Clustering: Used K Means Clustering to segment customers based on income and purchasing patterns.
- ✓ Predictive Modeling: Developed and validated models to forecast future customer behaviors.

# Clustering Methodology

✓ Algorithm: K Means Clustering

✓ Steps:

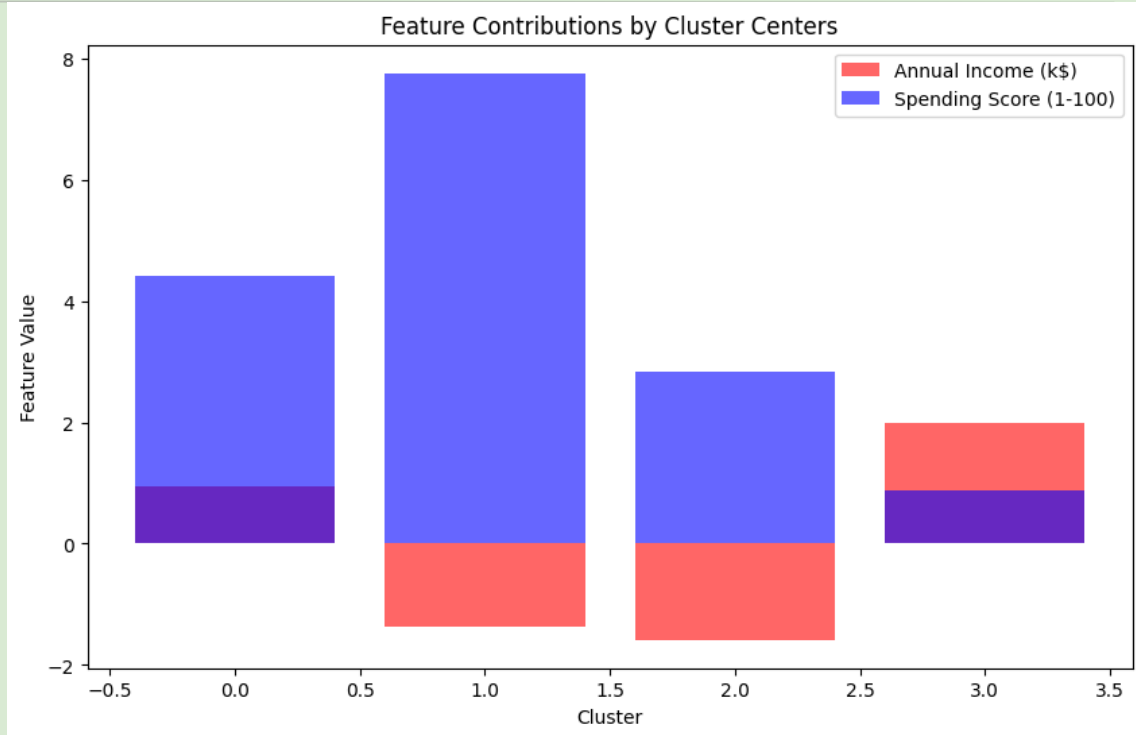
- Standardized the data
- Applied the Elbow Method to determine optimal clusters
- Executed K Means Clustering



# Cluster Centers Analysis

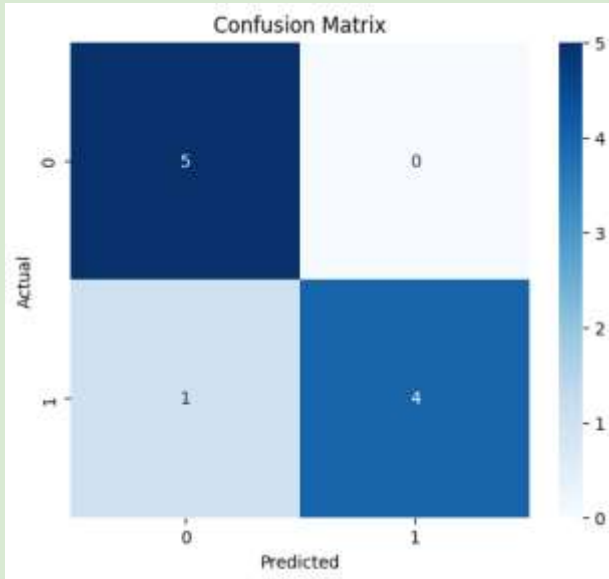
## ✓ Analysis:

- Examined cluster centers to understand feature contributions
- Visualized the importance of features like annual income and spending score in forming clusters

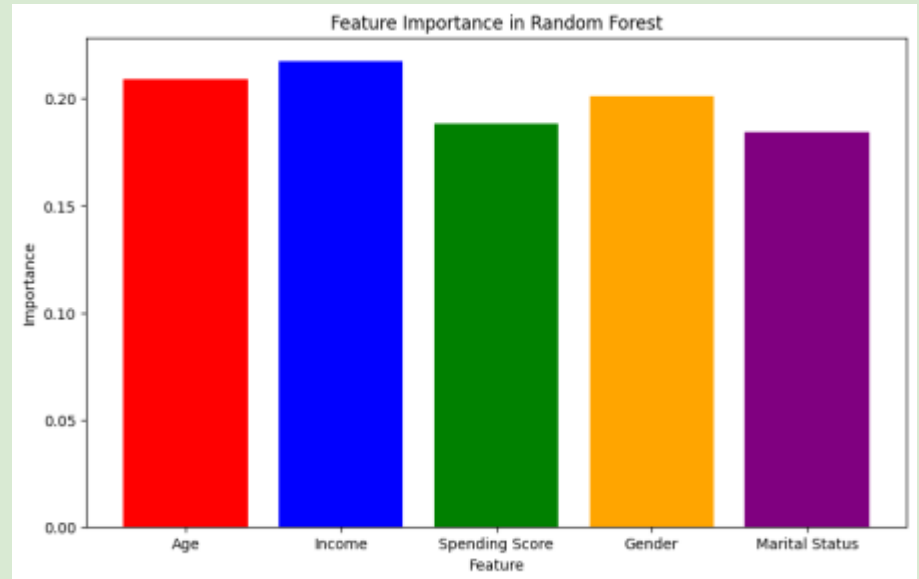


# Predictive Modeling

- ✓ Model: Logistic Regression
- ✓ Purpose: Predict customer responses to marketing campaigns
- ✓ Accuracy: Demonstrated through confusion matrix

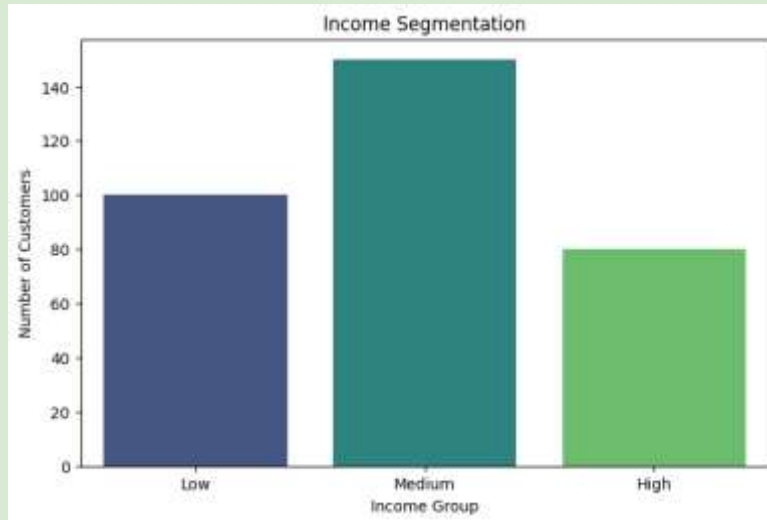


- ✓ Model: Random Forest
- ✓ Purpose: Identify key features influencing predictions



# Visualizations

**Visualization:** Displayed distinct customer segments based on income

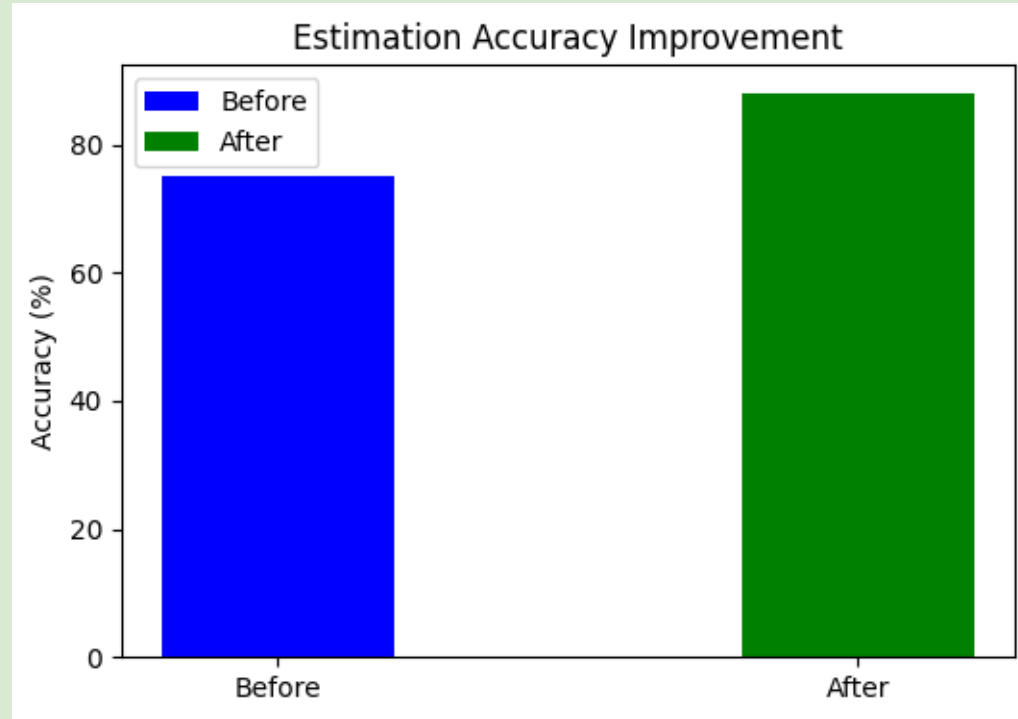


**Visualization:** Displayed distinct customer segments based on income



# Business Impact

- ✓ Before and After Comparison: Showed improvement from 75% to 88% accuracy
- ✓ Impact: Enabled precise and effective marketing campaigns and personalized strategies





# Technical Details and Challenges

## Challenges:

- ✓ Ensuring data accuracy and completeness
- ✓ Improving model performance through feature engineering and parameter tuning

## Solutions:

- ✓ Implemented robust data preprocessing techniques
- ✓ Utilized advanced model tuning methods

