

Customer Segmentation Using Supervised Learning



This presentation explores customer segmentation using supervised learning techniques. It covers the project's background, objectives, methodology, and results. We aim to provide a clear understanding of how AI can enhance marketing strategies.

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Introduction

Customer segmentation involves dividing customers into groups. Groups are based on similar characteristics. This allows businesses to better understand their customers. They can create targeted marketing strategies.

Targeted strategies

Segmentation creates customer focused marketing.

Understand customers

Segmentation builds deeper insights.





Background & Motivation

Traditional segmentation methods are manual and inefficient. AI and supervised learning automate the segmentation process. They offer faster and more accurate results.



Automated



Fast



Accurate

Project Objectives

This project focuses on implementing a supervised learning model. The goal is to segment customers effectively. We aim to classify customers into value-based segments.

Automate segmentation

Apply in real-world scenarios.

Visualize important features

Understand key segmentation drivers.



Dataset Overview

The dataset used is the Marketing Campaign Dataset from Kaggle. It contains demographic and behavioral attributes. These include age, education, income, and marital status.

Demographics

- Age
- Education
- Income

Behavioural Data

- Spending on products
- Number of children
- Marital Status

Methodology

The methodology includes data preprocessing, feature engineering, and model selection. We used a Decision Tree Classifier for its interpretability.



Preprocessing



Feature Engineering



Model Selection



Model Evaluation



Code Highlights

The code involves data loading, cleaning, and label encoding. We created a target variable by computing total spending. The DecisionTreeClassifier was used from sklearn.

Data Loading

Model Training

Evaluation

```
# Step 1: Import libraries
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score, classification_report

# Step 2: Load the dataset
df = pd.read_csv('/content/marketing_campaign.csv', sep=';')
print("First 5 rows:\n", df.head())

# Step 3: Drop unnecessary columns (like ID, date fields)
df.drop(['ID', 'Dt_Customer', 'Z_CostContact', 'Z_Revenue'], axis=1, inplace=True)

# Step 4: Handle missing values
df.dropna(inplace=True)

# Step 5: Encode categorical variables
categorical_cols = ['Education', 'Marital_Status']
le = LabelEncoder()
for col in categorical_cols:
    df[col] = le.fit_transform(df[col])

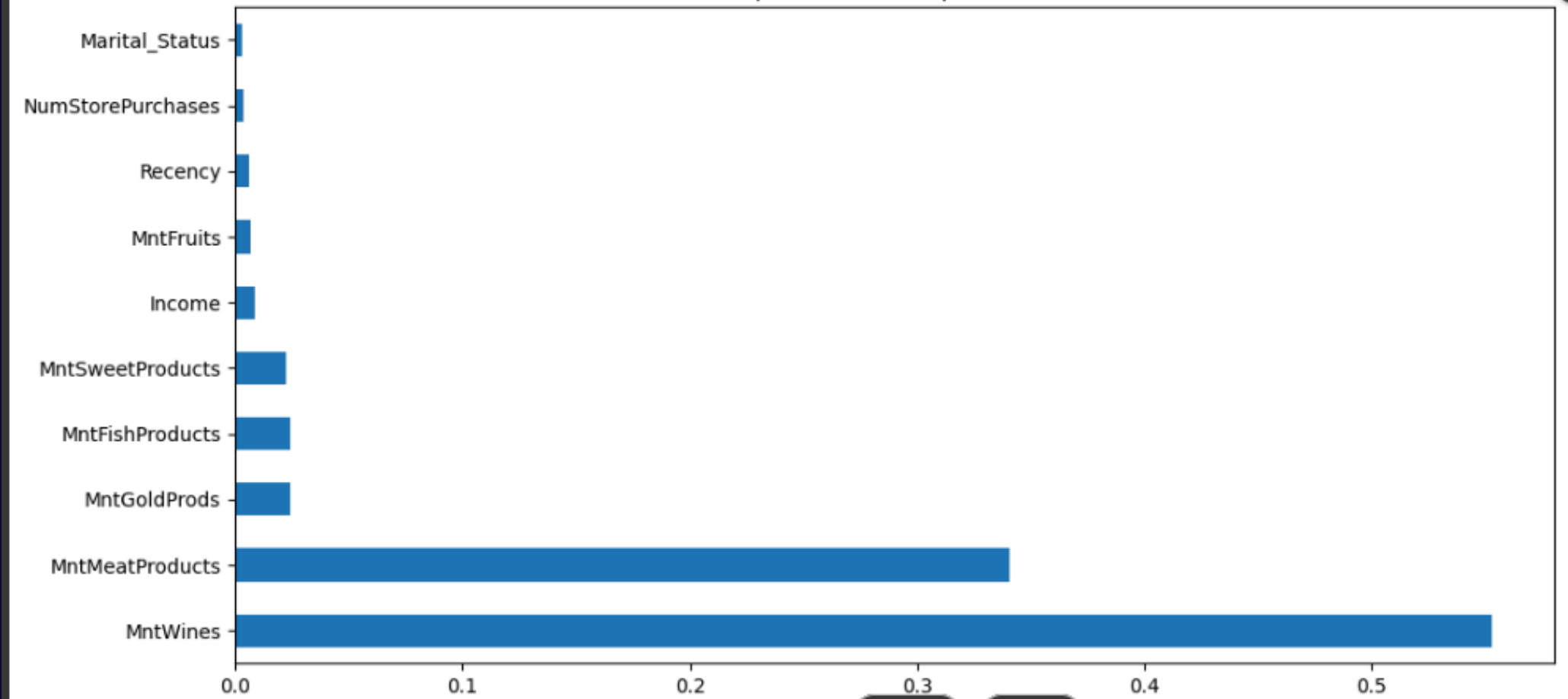
# Step 6: Create a target variable
# Let's define 'High-Value Customers' based on total spending
df['Total_Spend'] = df[['MntWines', 'MntFruits', 'MntMeatProducts', 'MntSeafood', 'MntGroceries', 'MntTotal']].sum(axis=1)
```

Output

First 5 rows:

	ID	Year_Birth	Education	Marital_Status	Income	Kidhome	Teenhome	\
0	5524	1957	Graduation	Single	58138.0	0	0	
1	2174	1954	Graduation	Single	46344.0	1	1	
2	4141	1965	Graduation	Together	71613.0	0	0	
3	6182	1984	Graduation	Together	26646.0	1	0	
4	5324	1981	PhD	Married	58293.0	1	0	
	Dt_Customer	Recency	MntWines	...	NumWebVisitsMonth	AcceptedCmp3	\	
0	04-09-2012	58	635	...	7	0		
1	08-03-2014	38	11	...	5	0		
2	21-08-2013	26	426	...	4	0		
3	10-02-2014	26	11	...	6	0		
4	19-01-2014	94	173	...	5	0		
	AcceptedCmp4	AcceptedCmp5	AcceptedCmp1	AcceptedCmp2	Complain	\		
0	0	0	0	0	0			
1	0	0	0	0	0			
2	0	0	0	0	0			
3	0	0	0	0	0			
4	0	0	0	0	0			
	Z_CostContact	Z_Revenue	Response					
0	3	11	1					
1	3	11	0					
2	3	11	0					
3	3	11	0					
4	3	11	0					

Top 10 Feature Importances



Classification Report:

	precision	recall	f1-score	support
High	0.95	0.93	0.94	221
Low	0.97	1.00	0.98	224
Medium	0.93	0.92	0.92	220
accuracy			0.95	665
macro avg	0.95	0.95	0.95	665
weighted avg	0.95	0.95	0.95	665

Conclusion

The model successfully segmented customers using supervised learning. Key features included income, age, and product spending. The project demonstrates the value of AI in marketing.



Thank You

Thank you for your time and attention. Are there any questions?

