BUSINESS STRATEGY

Data-driven Decision Making

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Github: https://github.com/NaDub/visu_2025/

Project Overview

A company wants to produce the highest profit for the next direct marketing campaign.

Context:

- A pilot campaign targeted 2,240 customers.
- Customers were randomly selected and contacted by phone.
- Campaign cost: 6,720MU, Revenue: 3,674MU
- Profit: -3,046MU, Profit/Customer: -1.36MU.
- Success rate: 15%.

Objectives

Our Goals

- Develop an accurate predictive model.
- Understand key customer characteristics.
- Maximize the profitability of marketing campaigns.

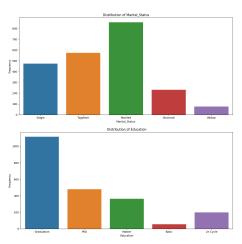
Data Exploration

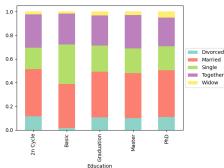
Unveiling Insights for Strategic Decisions

Explanation of Variables

- **AcceptedCmp1**: 1 if the customer accepted the offer in the 1st campaign, 0 otherwise.
- **Response (target)**: 1 if the customer accepted the offer in the last campaign, 0 otherwise.
- Complain: 1 if the customer complained in the last 2 years.
- DtCustomer: Date of the customer's enrolment with the company.
- Education: Customer's level of education.
- Marital: Customer's marital status.
- Kidhome: Number of small children in the customer's household.
- **Teenhome**: Number of teenagers in the customer's household.
- Income: Customer's yearly household income.
- MntFishProducts: Amount spent on fish products in the last 2 years.
- NumDealsPurchases: Number of purchases made with a discount.
- **Recency**: Number of days since the last purchase.
- Customer_Age: Seniority of the consumer.





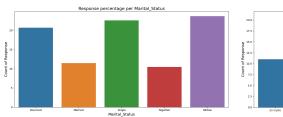


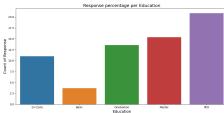
Customer Profile:

- Mostly married and highly educated.
- No significant correlation between these factors.



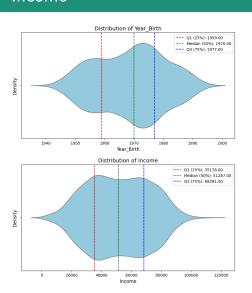
Customer profiles: Percentage of response





- The rate of positive responses in the last campaign appears to be higher if the consumer is not in a relationship.
- The rate of positive responses seems to increase with the level of education.
- A These relationships are not confirmed in previous campaigns.

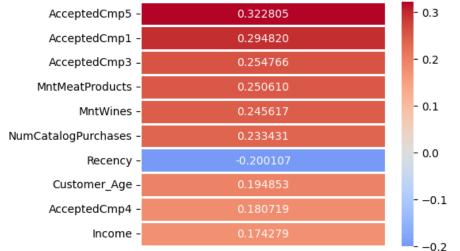
Customer profiles: Distributions of Year birth and Income



- A significant share of consumers were born between 1970 and 1977.
- The income distribution is close to a normal distribution.

W Key Correlations

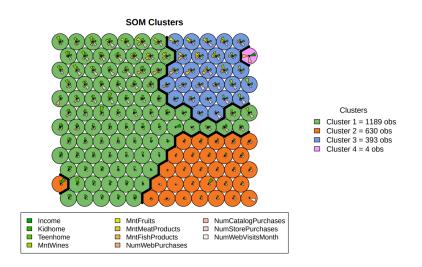
Correlation to Response





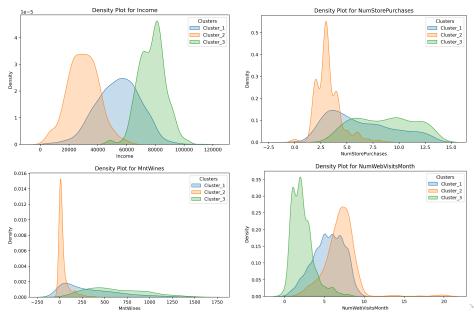
SOM application and analysis of segments

SOM results



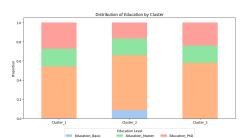


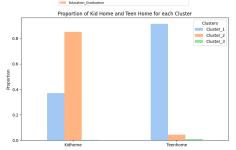
Customer Segmentation Analysis





Customer Segmentation Analysis

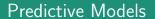




- Cluster 1: Middle-income, balanced spending, moderate wine buyers.
- Cluster 2: Low-income, more children, low spending, frequent web visits.
- Cluster 3: High-income, few children, premium wine buyers.

Classification Models

Gradient and logistic regression

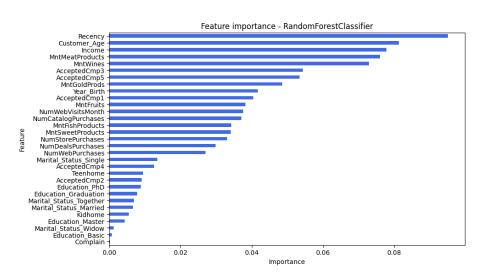


Test on two models (Logistic and Gradient Boosting):

- Accuracy: each $\approx 90\%$
- Key Factors:
 - Past campaign history
 - Customer income
 - Loyalty (Customer_Age and recency)
 - Consumption habits

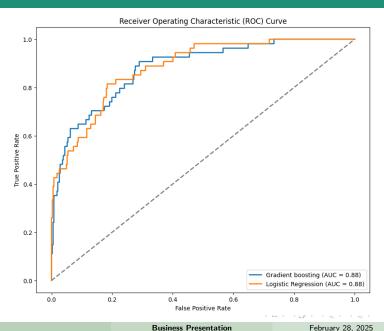


Feature Importance for tree

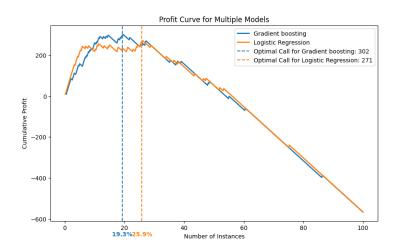




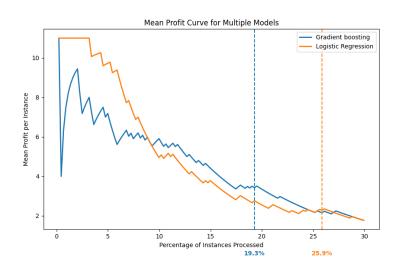
ROC and AUC







€ Profit curve



Conclusion

- We will select the top 19.3% of consumers who are most likely to respond positively to the call, according to the gradient boosting model.
- At the optimal threshold of 19.3%, the average profit is 3.24MU compared to -1.36MU without data-driven decision-making.
- With our model, the total profit of the last campaign would have been 302.