

# Marketing Campaign Customer Segmentation

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# Agenda

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2. Problem Definition
3. Solution Approach
4. Data Cleaning, Feature Engineering, and EDA
5. Clustering Methods
6. Profiles From Best Cluster Solution
7. Recommendations and Action Items
8. Potential Benefits and Costs
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# Executive Summary

## What was done

- Addressed null values and removed duplicates
- Categorical variables refined by merging similar categories
- Feature Engineering: Created additional columns e.g. Age, Total Purchases, Average Spend
- Exploratory Analysis: Univariate and Bivariate analysis, data scaling
- Five clustering methods tested, best one determined through silhouette score and cluster profiles

## Outcome

- Identified correlations e.g. purchases negatively correlated with having children
- Found that **K-Means with 3 clusters** provided the optimal clusters
- Determined that income was the biggest driver in behaviour
- Characterised 3 clusters:
  - Higher-Income Spenders
  - Deal-Oriented Average Shoppers
  - Budget-Conscious Families

## Next Steps

- Recommendations
  - White-Label Product Line focusing on family items
  - Refer-a-Friend program
- Identified action items for different teams
  - Marketing
  - Product
  - Sales
  - Data
  - E-Commerce

# Problem Definition

## Context

In today's competitive market, understanding customers is essential for strategic growth. Ignoring differences between groups leads to generic experiences that fail to engage customers, resulting in missed opportunities, lower satisfaction, and reduced revenue, especially for a company with a broad product portfolio.

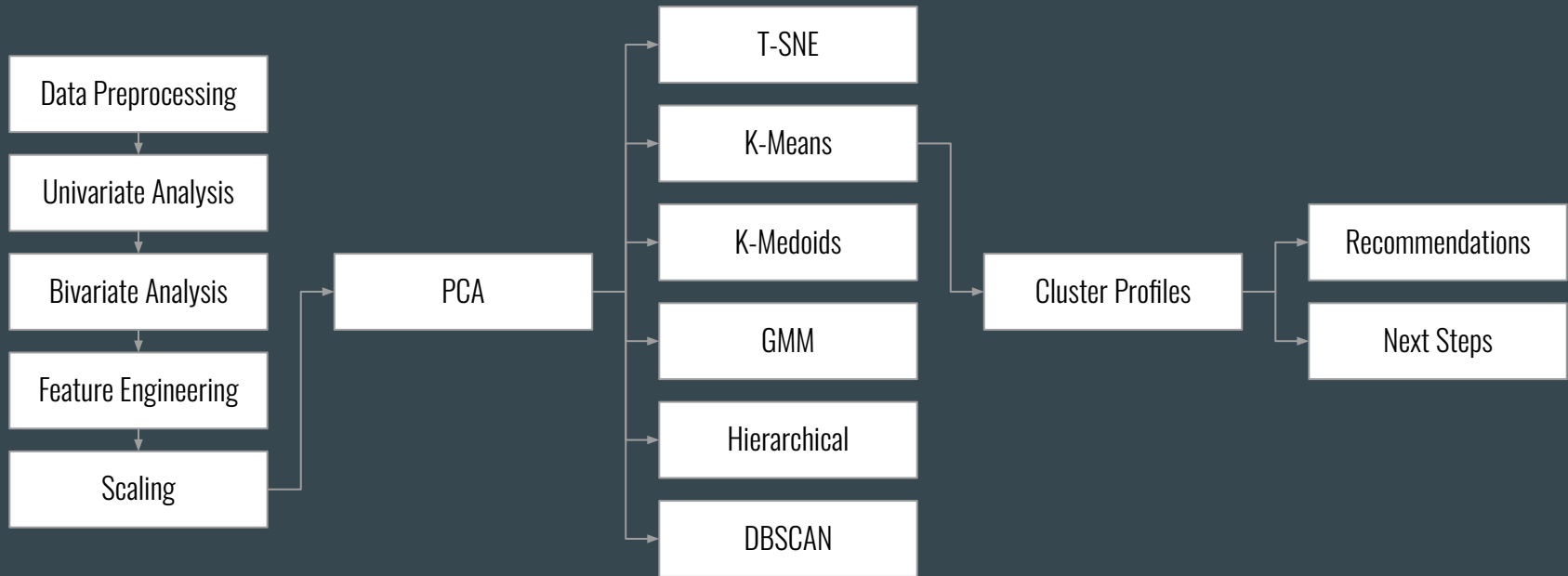
## Objective

Our goal is to segment the customer base using behavioral and demographic data to create a framework for personalized engagement. This segmentation will help us allocate resources effectively to maximize impact. For instance, if we identify a group that responds well to discounts, we can design targeted promotions to boost conversions. For segments less motivated by price, we can create value-based offers, protecting margins.

## Key questions

- How do our customer demographics influence purchasing behaviour?
- What clustering techniques deliver the most actionable customer segments? What are their profiles?
- What are the next steps for stakeholders to improve their approach to these segments?

# Solution Approach



# Data Cleaning, Feature Engineering, and EDA

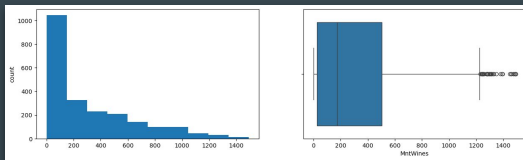
## Data Cleaning + Feature Engineering

- Null values in income: replaced with median
- Removed ID column
- Unnecessary categories in Marital Status and Education: combined with larger categories
- Created new features:
  - Age, Childrenhome, Household size, Total Spend (MntTotal), and Total Purchases

## Exploratory Data Analysis

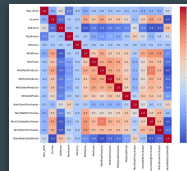
### Univariate Analysis

- Many variables were right skewed
- 15% have responded to campaigns, only 1% have made complaints
- Graduation (bachelors) and being married are the most common education and marital statuses



### Bivariate Analysis

- Channels were negatively correlated with each other
- Income positively correlated with purchasing across categories
- Purchasing negatively correlated with kids

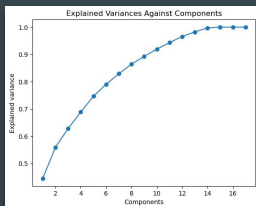


# Clustering Methods

## PCA

### PCA

- 10 components chosen (91.93% explained variance)

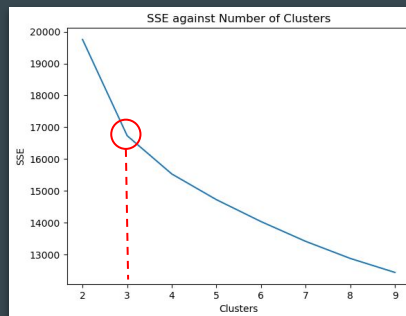


## Chosen Method

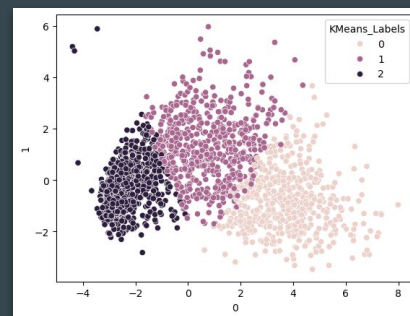
### K-Means

- Elbow identified at 3 clusters
- Highest silhouette score obtained: 0.29
- No cluster overlap in PCA visualization
- Clear cluster characteristics found through group summary statistics

## SSE Plot



## PCA Visualization



## Group Summary Statistics

	Group 0 Mean	Group 1 Mean	Group 2 Mean	Group 0 Median	Group 1 Median	Group 2 Median
Year_Birth	1967.880392	1966.115385	1971.069215	1969.000000	1966.000000	1972.0
Income	76665.962745	57517.082168	36345.072314	76807.500000	57958.000000	35482.0
Kidhome	0.033333	0.272727	0.766529	0.000000	0.000000	1.0
Teenhome	0.219608	0.812937	0.485537	0.000000	1.000000	0.0
Recency	49.954902	48.005245	48.919421	52.500000	48.000000	49.0



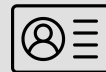
25%

## Group 0



# Higher-Income Spenders

### Profile



- Less kids
- Makes the most web purchases
- Makes more catalog purchases
- Higher response to campaign offers

### Targeting Strategy



- Maintain the current experience
- Create a refer-a-friend program: higher income individuals tend to know other higher income individuals





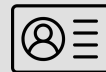
25%

# Group 1



## Deal-Oriented Average Shopper

### Profile



- Highest volume of purchases
- Purchases the most deals
- Makes the most web purchases and web visits
- Spends more on gold relative to income

### Targeting Strategy



- Offer them more deals for more expensive products as their average spend is currently low e.g. when they checkout online, offer them some “You may also be interested in...”



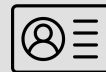
50%

## Group 2



# Budget-Conscious Families

### Profile



- More kids
- Slightly younger
- Less web purchases and but more web visits
- Highest proportion (50%)

### Targeting Strategy



- White-label store product brand focused on family goods
- Use online store to generate interest
- Target for campaigns, remove barriers of entry

# Recommendations and Action Items

## White-Label Product Line



Launch affordable products, focusing on family and kids' items.

### Action Items

- **Product Team:** Identify a white label producer to partner with.
- **Marketing Team:** Create digital campaigns to attract budget-conscious families.
- **E-commerce Team:** Improve visibility of white-label products on the website.
- **Data Team:** Monitor adoption rates and analyze campaign success.

## Refer-a-Friend Program

Leverage their campaign responsiveness by introducing a referral program.

### Action Items

- **Marketing Team:** Develop referral campaigns targeting high-income individuals.
- **Product Team:** Ensure premium products are highlighted.
- **Sales Team:** Track and optimize the referral program.
- **Data Team:** Monitor engagement metrics to refine targeting.

# Potential Benefits and Costs

## Benefits



**Increased Revenue**



**Customer Loyalty**



**Optimized  
Marketing**



**New Customers**

## Costs



**Implementation  
Costs**



**Existing  
Infrastructure**



**Training Needs**



**Marketing Budget**

# Risk and Mitigation Strategies

## Inaccurate Segment Identification

- Risk: Misidentifying segments could lead to ineffective campaigns
- Mitigation: Regularly update segments based on fresh data.

## Too Much Personalization

- Risk: Customers may find excessive targeting invasive.
- Mitigation: Set limits on communication frequency and provide opt-out options.

## Low Adoption of White-Label Products

- Risk: Budget-conscious customers may not adopt the new products
- Mitigation: Test products with small groups and collect feedback

# Questions

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