Retail
Customer
Segmentation
Analysis



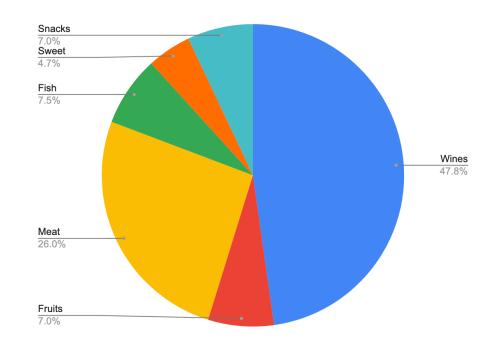


Agenda

- Background, Problem Statement & Objectives
- Approach
- Insights from Analysis
- Strategy Recommendations

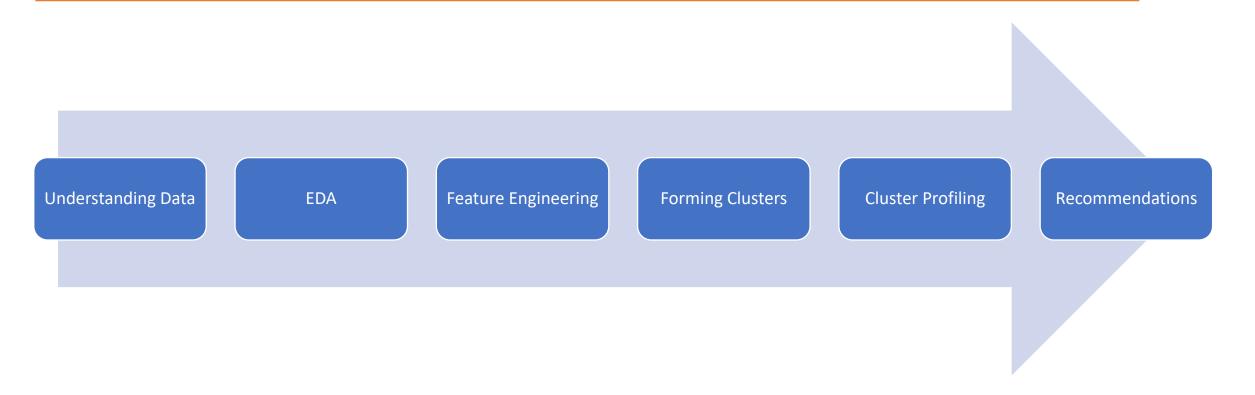
Background, Problem Statement & Objective

- Retailkart.com is a small and medium-scale organization that majorly deals in wine, fruit and meat products, having held around 35% market share.
- The company has been leading the domain offline for a long time.
- The company main challenge is to stay competitive against the rapidly expanding companies offering quick and free deliveries while still accepting online purchases at affordable costs
- The company's near-term objective is to maintain its overall market share, with the increasing competition offering 10-minute delivery, among other benefits.



Identify customer groups which have different purchasing preferences based on their affluence and other attributes and devise an effective marketing plan & preference-based user experience at a cohort/group level.

Approach



Data

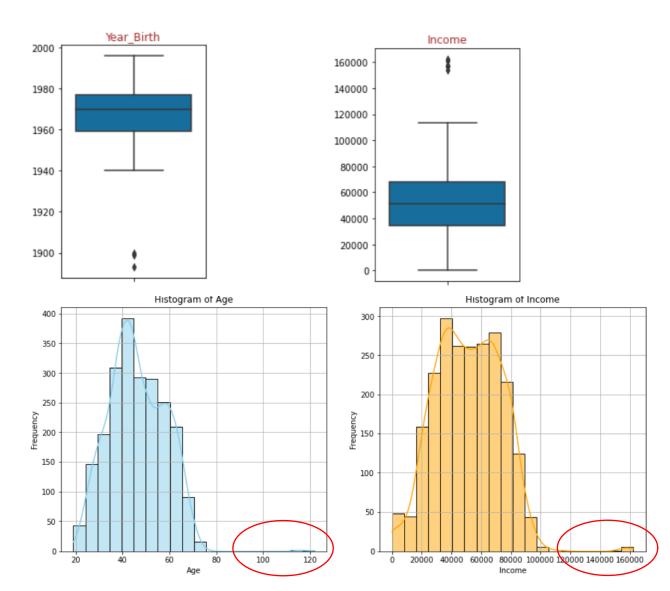
- The data is available at an aggregated level for each customer on the platform from appx from 2012 to 2014
- The data includes their historical transactions in each category, demographics, and reactions to promotions.
- Shape of Data
 - 2239 rows & 29 columns features
 - 26 numerical and 3 categorical features
- Data Quality
 - Very few missing values* (Income 1 record & Responses 23 records)

Category	Column	Description
People	ID	Customer's unique identifier
People	Year_Birth	Customer's birth year
People	Education	Customer's education level
People	Marital_Status	Customer's marital status
People	Income	Customer's yearly household income
People	Kidhome	Number of children in customer's household
People	Teenhome	Number of teenagers in customer's household
People	Dt_Customer	Date of customer's enrollment with the company
People	Recency	Number of days since customer's last purchase
People	Complain	1 if the customer complained in the last 2 years, 0 otherwise
Products	MntWines	Amount spent on wine in last 2 years
Products	MntFruits	Amount spent on fruits in last 2 years
Products	MntMeatProducts	Amount spent on meat in last 2 years
Products	MntFishProducts	Amount spent on fish in last 2 years
Products	MntSweetProducts	Amount spent on sweets in last 2 years
Products	MntSnacksProds	Amount spent on gold in last 2 years
Promotion	NumDealsPurchases	Number of purchases made with a discount
Promotion	AcceptedCmp1	1 if customer accepted the offer in the 1st campaign, 0 otherwise
Promotion	AcceptedCmp2	1 if customer accepted the offer in the 2nd campaign, 0 otherwise
Promotion	AcceptedCmp3	1 if customer accepted the offer in the 3rd campaign, 0 otherwise
Promotion	AcceptedCmp4	1 if customer accepted the offer in the 4th campaign, 0 otherwise
Promotion	AcceptedCmp5	1 if customer accepted the offer in the 5th campaign, 0 otherwise
Promotion	Response	1 if customer accepted the offer in the last campaign, 0 otherwise
Place	NumOnlinePurchases	Number of purchases made through the company's website
Place	NumCatalogPurchases	Number of purchases made using a catalogue
Place	NumStorePurchases	Number of purchases made directly in stores
Place	NumOnlineVisitsMonth	Number of visits to company's website in the last month

^{*} Missing value for Income was imputed by 0. For Responses the null values were replaced with an approximation of whether the customer has accepted any campaigns at a later stage or not.

EDA

- Outlier Analysis was done using Box Plot for all features
- Outliers were observed for Income and Year of Birth
- Outliers were removed for Income and after converting Year_Birth to Age

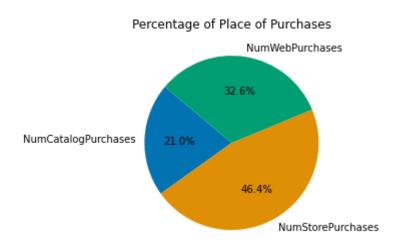


EDA

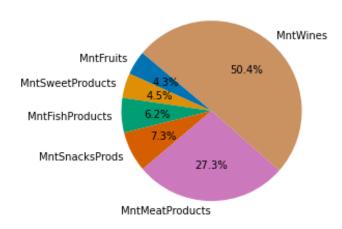
 Appropriate exploratory analysis was conducted to extract useful insights (whether directly useful for business or for eventual modeling/feature engineering).

Insights from EDA:

- Most of the customers are in the age group of appx 35 to 50 yrs with appx 50k as income
- Majority of the customers are Graduates followed by PhD and Masters Degree.
- Customers who are married and staying together are more than divorced, single, etc.
- Maximum customers dont have kids or teens at home.
- Majority of the customers are purcashing from the store followed by web. Which is not that bad.
- Majority of the customers are buying Wines and Meat Products.
- Very few customers are purchasing fruits and sweet products







Feature Engineering

Features Engineered	Method/Approach
Age	Taking 2015 as-of date since data is till end of 2014 and using Year_Birth
Total_Spent	Sum of amount spent on all products purchased by each customer
Member_for	Difference between 01-01-2015 and Dt_Customer
Total_Kids	Sum of Kidhome and Teenhome
Total_Purchases	Sum of purchases made from places – Web, Catalogue, Store
Total_AcceptedCmp	Sum of all accepted campaign columns.
Clubbing Product CategoriesMeat & FishSnacks & FruitsAlcohol	Sum of purchase amount for Meant and Fish Sum of purchase amount for Snacks, Fruits and Sweets Renames Wines
Education: - High_Edu - Basic_Edu	Customers with education - PhD, Cycle, Graduation, Master Customers with education – Basic
Marital_Status - Single - Not-Single	Divorced, Widow, Alone, YOLO, Absurd Married, Together

Correlation Insights

High Income Customers

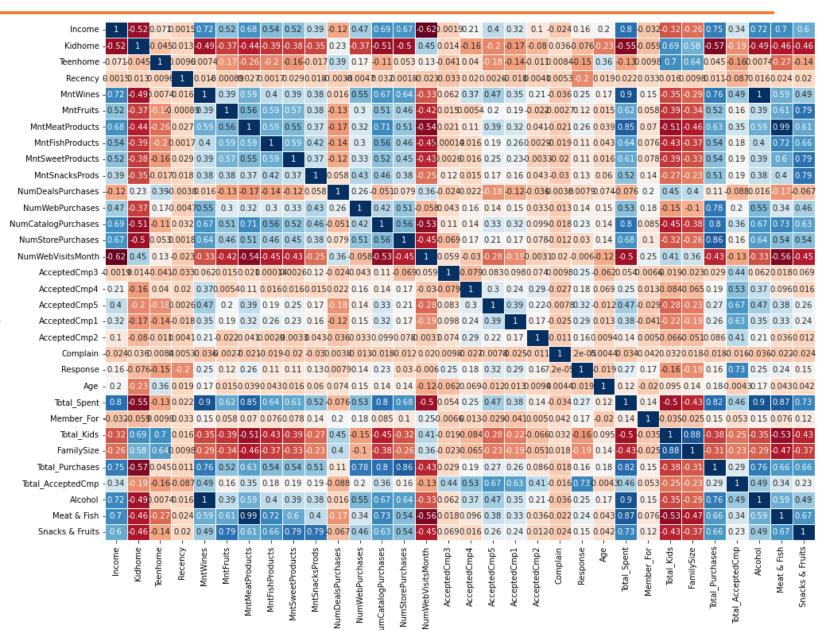
- spend more and purchase more.
- visit the company's website less frequently.
- Have few numbers of purchases made with a discount

Customers with High Spent

- do not to visit the company's website.
- buy more wines and meat products
- make a high number of purchases made using a catalog

Customers with kids

- spend less and purchase less.
- have a high number of purchases made with a discount



Clustering/Segmentation Approach

Encoding

 Label Encoding is used to convert categorical features (nonnumeric) into numerical values.

Feature Scaling

• Standardization done to ensure that all features are on a similar scale and have equal importance during the clustering process.

Check for Cluster Tendency

 Hopkins Test done to check whether the data points tend to cluster together or are randomly distributed

Dealing with Dimensionality

 PCA used in conjunction with k-means clustering for dimensionality reduction and to improve the clustering results.

Derive Clusters

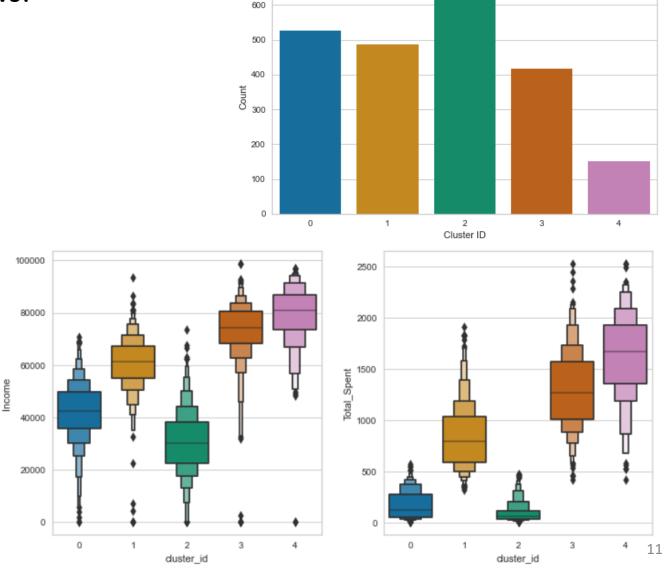
 Elbow curve method to determine the optimal number of clusters

Cluster/Segment Profiles

Total 5 Segments were identified as follows:

- Cluster 4: High Income High Spenders
- Cluster 3: High Income Avg Spenders
- Cluster 1: Avg Income Avg Spenders
- Cluster 0: Avg Income Low Spenders
- Cluster 2: Low Income Low Spenders

Click here to Refer Annexure after Thank You slide for detailed visualization



Count Plot of Cluster IDs

Segment Behavior

High Income High Spending

- Spend a lot of amount in purchasing meat & fish and alcohol
- They make very less Deal Purchases and make most of purchases from store
- Their visit to the website in last month is very low but their response to past campaigns is high
- Majority of them don't have kids at home

High Income Avg Spenders

- This group has sizable number of customers
- They do spend on an average considerable amount of purchases on all products
- Their last month visit to the web is the lowest in all clusters
- Their place of purchase amount is almost similar to high income and high spend segment
- They have smaller family size

Avg Income Avg Spenders

- This segment spends more amount on alcohol and fruits+snacks+sweets than meat+fish
- Their web purchases are comparatively higher than other segments
- Their web visits last month were high and so were the purchases made through deals
- They have a slightly higher family size

Avg Income Low Spenders

- This segment spends low amount on purchases
- They do more deal purchases
- Their visit to web last month was higher than others
- They have the largest family size

Low Income Low Spenders

- This segment is quite sizable in number of segmented customers
- As the name suggests they very less purchases and amount spent is also very low
- Whatever purchases they do is through stores
- However, their last month visit to website is highest among all segments
- They have large family size

Customized Recommendations for Segments

High Income High Spending Segment:

- Enhance the online shopping experience by improving the website's user interface, making it user-friendly and visually appealing.
- Provide online deals and promotions to encourage more web purchases from this segment.
- Implement a loyalty program for frequent online shoppers with discounts or wallet points for future purchases.
- Targeted email marketing and personalized product recommendations based on their past online purchases.

High Income Avg Spenders Segment:

- Develop online-exclusive offers and limited-time promotions to attract this segment to make more web purchases.
- Increase online advertising and promote special deals available only on the website.
- Offer free shipping or discounts on bulk purchases.
- Implement campaigns to remind this segment of products they viewed but did not purchase during their previous online visits.

Avg Income Avg Spenders Segment:

- Create loyalty rewards that encourage repeat purchases and offer special schemes for online shoppers.
- Implement a referral program to encourage satisfied customers to refer friends and family to make online purchases.
- Email marketing to inform this segment about exclusive online deals and upcoming promotions.

Avg Income Low Spenders Segment:

- Create limited-time online flash sales to attract this segment.
- Offer free shipping for online orders to remove potential barriers to online shopping for this segment.
- Flash product descriptions and customer reviews on the landing page of website to build trust and confidence in online purchases.
- Implement a user-friendly checkout process with various payment options to modernize online purchases.
- Improve websites performance for loading and checkout.

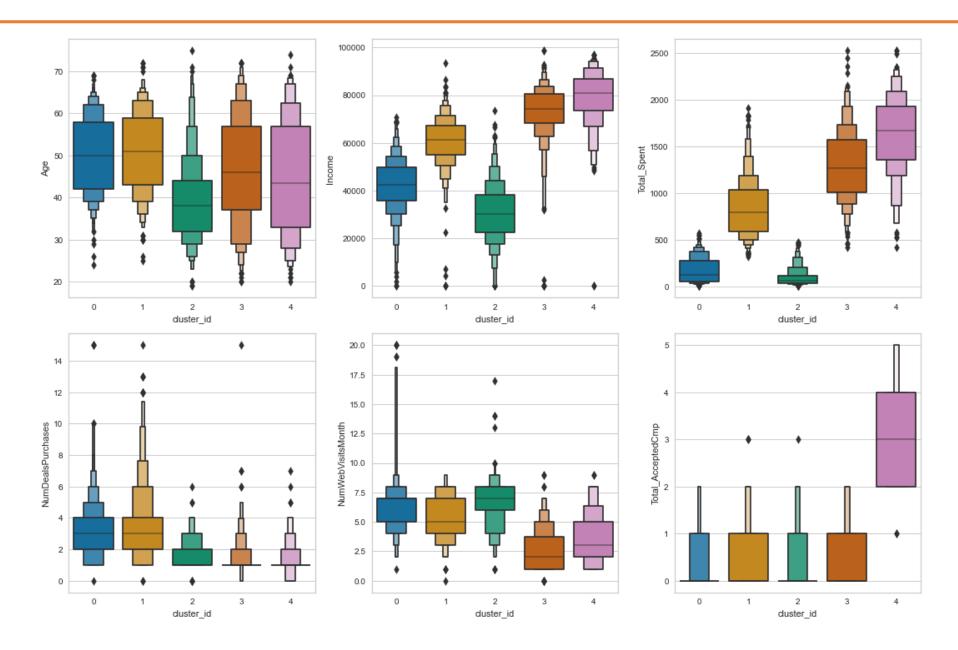
Summary of Recommendations

- Retailkart.com should use web analytics to understand customer/user behavior and identify potential points of improvement on the website.
- Regularly update and refresh Retailkart.com website content to keep it engaging and relevant to the target audience.
- Leverage social media and online advertising to direct traffic to the website during promotional periods.
- Encourage customer feedback and reviews to build trust and credibility among online shoppers.
- Monitor and optimize the website's loading speed to provide a seamless online shopping experience.



Thank you

Cluster Behavior Visualization (1/2)



Cluster Behavior Visualization (2/2)

