



FAKULTÄT FÜR  
INFORMATIK

# Customer Behavioural Analytics in the Retail Sector

Team CuBA: Rutuja Shivraj Pawar, Nadiia Honcharenko, Shivani Jadhav, Sumit Kundu

Under the Guidance of M.Sc. Uli Niemann

`rutuja.pawar@ovgu.de, nadiia.honcharenko@st.ovgu.de,  
shivani.jadhav@st.ovgu.de, sumit.kundu@st.ovgu.de, uli.niemann@ovgu.de`

**Data Science with R (DataSciR) - Final Presentation**

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# Agenda

1. Motivation
2. Project Objective
3. Data Source
4. Research Questions
5. Conclusion and Scope for Future Work

# Motivation

- Customer as a key-centric factor in a business
- Cost of retaining an existing customer is far less than acquiring a new one
- Customer Behavioural Analytics plays an important role in leveraging data analytics to find meaningful behavioural patterns in the customer-specific business data

***"A satisfied customer is the best business strategy of all"***

- Michael LeBoeuf

# Project Objective

- Aim: To understand the purchase behaviour of customers in the retail sector specifically of an Italian retail distribution company *Coop* in a single Italian city
- Intends to discover different analytical insights through answering different formulated Research Questions (RQ)



## Data Source

- Retail market data of one of the largest Italian retail distribution company called *Coop* for a single Italian city [2]
- Contains data aggregated from the original datasets [1] [3] and mapped to new columns
- The dataset contains 40 features with 60,366 instances and is approximately 14.0 MB in size



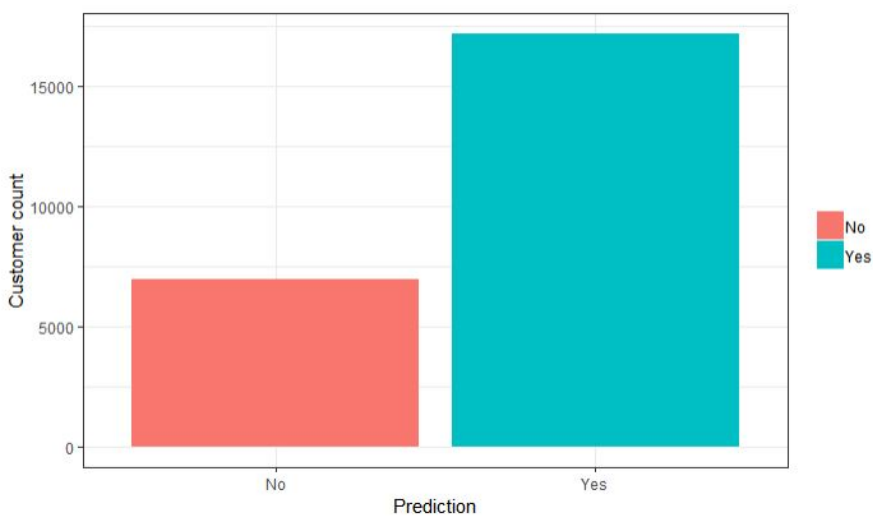
: 603565 obs, of 493 variables)

- customer\_id (int)
- shops\_used\_total (int)
- distance\_shop\_1 (num)
- distance\_shop\_2 (num)
- distance\_shop\_3 (num)
- distance\_shop\_4 (num)
- distance\_shop\_5 (num)
- min\_dist\_to\_cut5ell\_shops (num)
- max\_dist\_to\_cut5ell\_shops (num)
- avg\_distance\_to\_all\_shops (num)
- products\_purchased\_shop\_1 (int)
- products\_purchased\_shop\_2 (int)
- products\_purchased\_shop\_3 (int)
- products\_purchased\_shop\_4 (int)
- products\_purchased\_shop\_5 (int)
- products\_purchased\_total (int)
- unique\_products\_purchased\_shop\_1 (int)
- unique\_products\_purchased\_shop\_2 (int)
- unique\_products\_purchased\_shop\_3 (int)
- unique\_products\_purchased\_shop\_4 (int)
- unique\_products\_purchased\_shop\_5 (int)
- unique\_products\_purchased\_total\_exclCommon (int)
- avg\_product\_price\_shop\_1 (num)
- avg\_product\_price\_shop\_2 (num)
- avg\_product\_price\_shop\_3 (num)
- avg\_product\_price\_shop\_4 (num)
- avg\_product\_price\_shop\_5 (num)
- avg\_purchased\_product\_price\_allShops (num)
- amount\_purchased\_shop\_1 (num)
- amount\_purchased\_shop\_2 (num)
- amount\_purchased\_shop\_3 (num)
- amount\_purchased\_shop\_4 (num)
- amount\_purchased\_shop\_5 (num)
- amount\_purchased\_total (num)
- avg\_purchase\_amount\_shop\_1 (num)
- avg\_purchase\_amount\_shop\_2 (num)
- avg\_purchase\_amount\_shop\_3 (num)
- avg\_purchase\_amount\_shop\_4 (num)
- avg\_purchase\_amount\_shop\_5 (num)
- avg\_purchase\_amount\_allShops (num)

# RQ1: Are customers willing to travel long distances to purchase products?

**Approach:** Classifying the data based on distances and other decision factors

**Algorithms Used:** Support Vector Machine, K-nearest neighbour, Random Forest





# Insights for RQ1

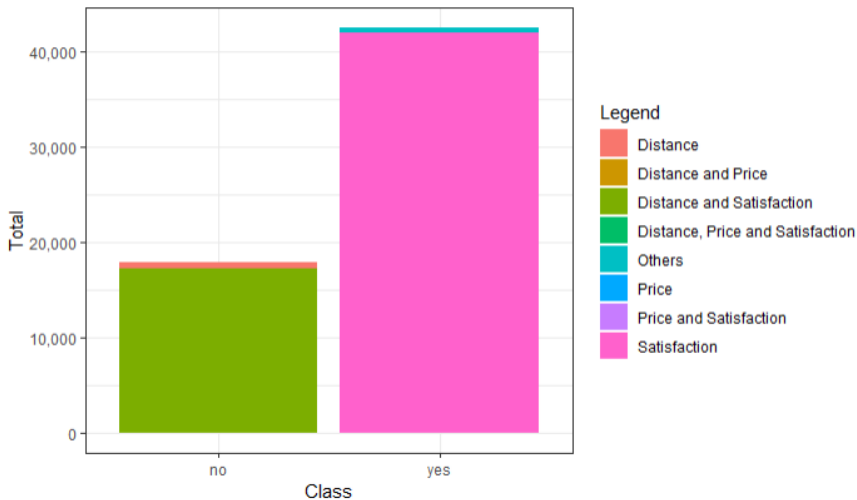
- **Observations:** It was found out that the majority of the customers are ready to travel long distances to purchase products and this decision is affected by certain factors
- **Applications for the Business:**
  - Understand the behaviour trend of the majority of the customers related to long distance travel to purchase products
  - Understand the reasons behind such majority trends
  - Devise strategies in the context of the store locations, enhance the factors influencing such trends, generate more revenue and increase customer satisfaction

## RQ2: What are the factors that contribute towards the long distance travel of the customer to purchase products?

**Approach:** Identifying the factors for the long-distance travel of the majority of the customers

**Algorithm Used:** Custom Algorithm and as a follow-up investigation for RQ1

## Customer Classification



# Insights for RQ2

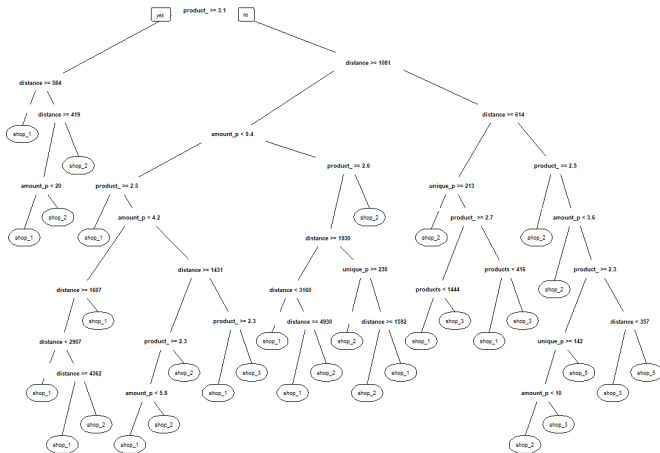
- **Observations:** Responsible factors for the majority of the customer trend towards long-distance travel were determined and It was observed that 'Satisfaction' is a key role factor affecting a customer's decision-making process.
- **Applications for the Business:**
  - Devise strategies to enhance the observed most important factors facilitating customer satisfaction retention
  - Further paves way for a steady business growth

## RQ3: What is the maximum likelihood of a customer to select a particular shop?

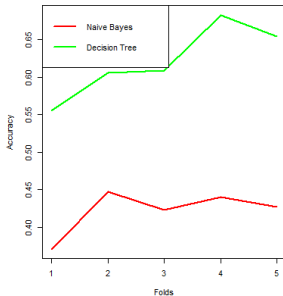
**Approach:** Predicting the shop most likely to be selected by a new customer

**Algorithms Used:** Naive Bayes, Decision Tree

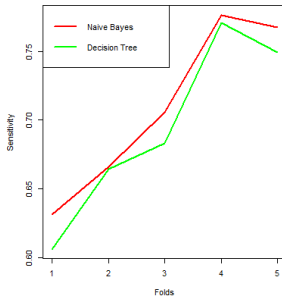
Decision tree



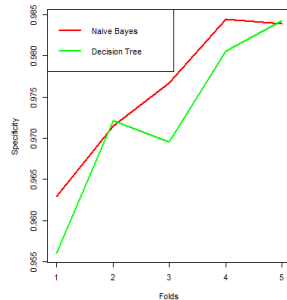
Cross Validation - Accuracy



Cross Validation - Sensitivity



Cross Validation - Specificity



## Insights for RQ3

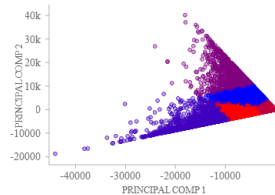
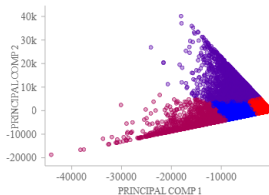
- **Observations:** The predictive model based on Decision tree will output better results in predicting the maximum likelihood of a new customer to select a particular shop.
- **Applications for the Business:**
  - Understand which shops in the retail chain are most likely to be preferred by new customers
  - Facilitates towards better stock management to meet the increasing customer demands
  - Strategies to increase profit and attract new customers in different shops

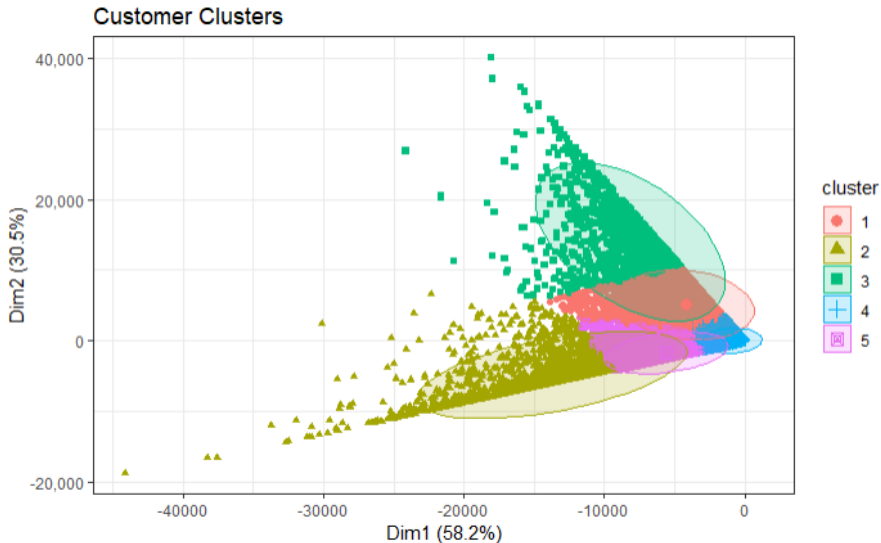


## RQ4: What are the different customer segments based on their purchase behaviour?

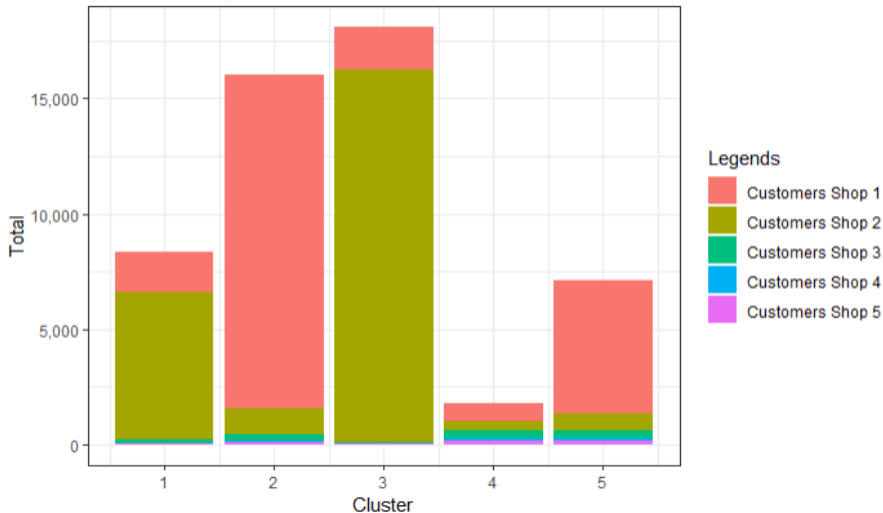
**Approach:** Clustering the data based on the shops customers shop the most

**Algorithms Used:** K-means, Principal Component Analysis





## Customer Segmentation



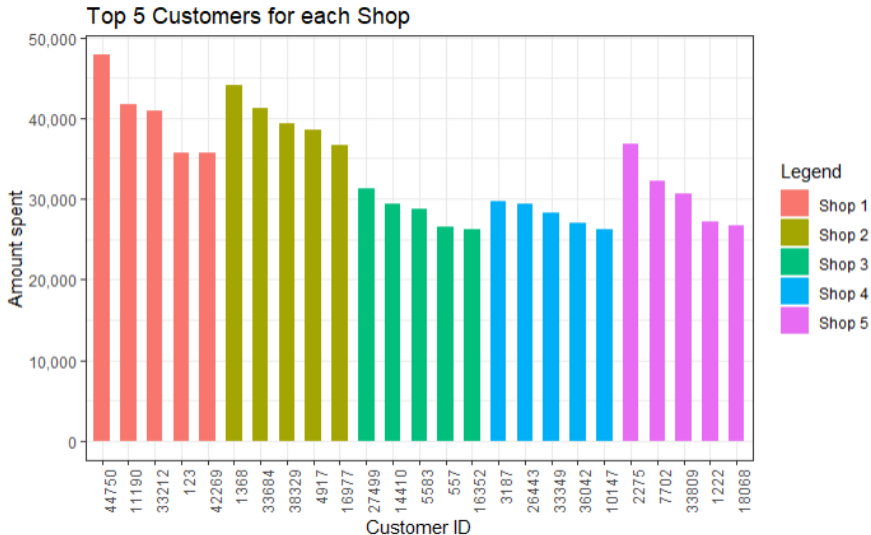
# Insights for RQ4

- **Observations:** Five customer segments were detected and further partitioning these segments revealed the specific customers belonging to the five different shops
- **Applications for the Business:**
  - Develop a specific strategy for each cluster base
  - Understand the purchase behaviour of customers
  - Focus marketing efforts on the right customers
  - Cut-down the marketing costs, generate more revenue and increase customer satisfaction

**RQ5: Which are the Top 100 customers that are most profitable in terms of revenue generation for each shop?**

**Approach:** Ranking the Top 100 profitable customers for each shop

**Algorithms Used:** Custom Ranking algorithm



# Insights for RQ5

- **Observations:** Top 100 customers that spend the most amount of money in each of the 5 shops based on their loyalty score were ranked and determined. Further, Top Ranked 5 customers for each shop were visualized.
- **Applications for the Business:**
  - Identify top profitable customers for each shop
  - Formulate reward schemes to retain the high-value customer base
  - Paves way to transform satisfied loyal customers as advocates for the business

## Conclusion and Scope for Future Work

- Meaningful insights discovered will help the business to implement customer-centric strategies leading to an increased revenue through customer satisfaction
- Long distances travel and its factors, Customer shop selection prediction, Customer segments, Top 100 profitable customers
- Formulation of different RQs, Selection of different data source, Extension of analysis to other business sectors can further help to *decode the customer behaviour through analytics leading to a sustainable business growth*



# Thank you all for your attention!

## Any Questions?

-  Dataset. <https://bigml.com/user/czuriaga/gallery/dataset/5559c2c6200d5a6570000084>
-  Datasets. <http://www.michelecoscia.com/?pageid=379>
-  Project Objective. [Image] *The Importance of Customer Insight*. <https://www.klbdkosher.org/news-and-articles/the-importance-of-customer-insight/>
-  Data Source. [Image] *Coop Logo*. [https://en.wikipedia.org/wiki/Coop\(Italy\)](https://en.wikipedia.org/wiki/Coop(Italy))