

# Web Marketing & Communication Management

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# Business questions:

## 1) Customer Relationship Valorization:

**How to make the relationship with customers more profitable?**

Exploiting customer base data to understand:

- What are the characteristics of the customer base?
- How can it be segmented?
- Which are the high-value customers to retain and which are the potential churners?
- Which way the customers behavior changed over the last 2 quarters?

## 2) Products Valorization:

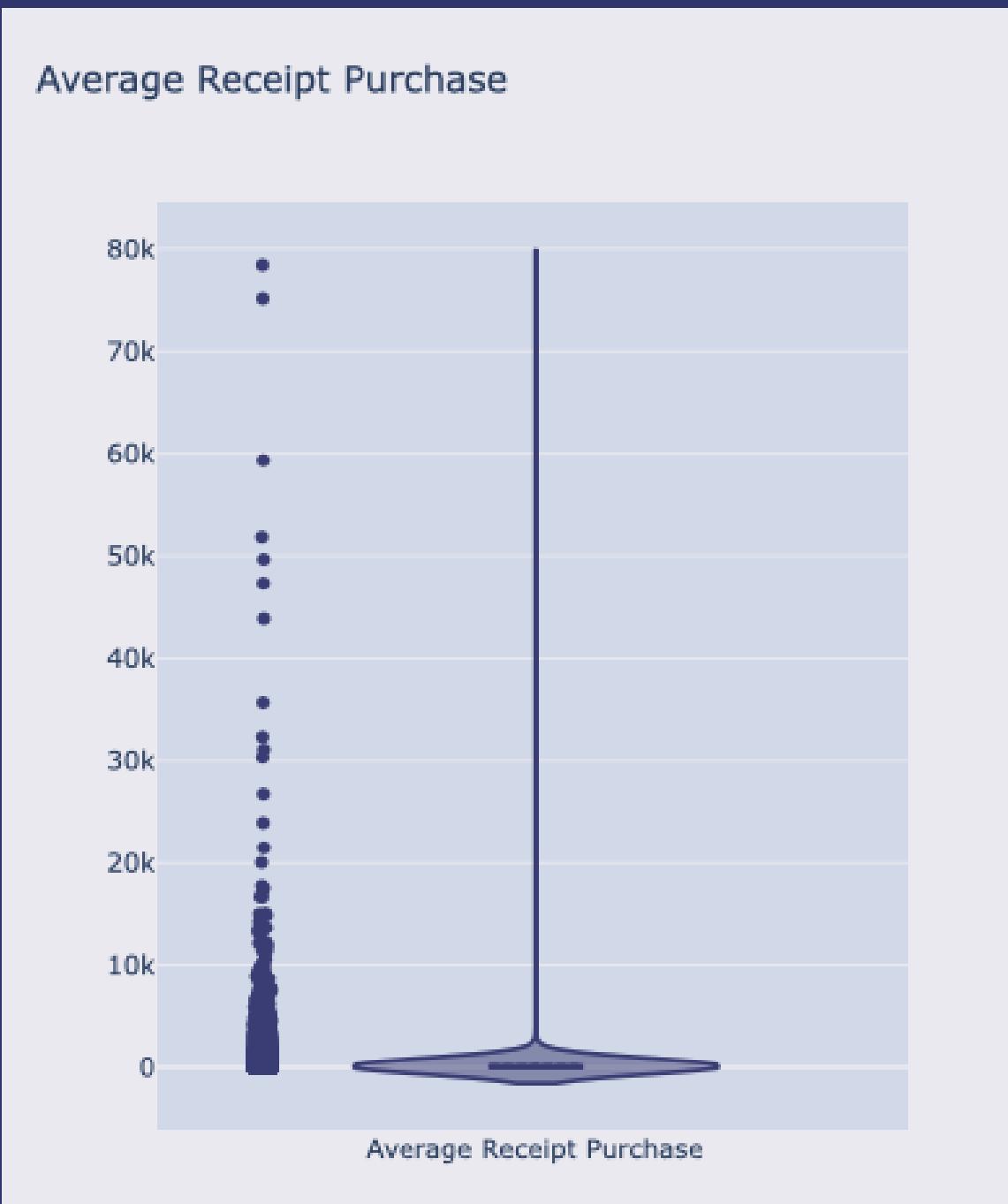
**Which products are more likely to be sold together?**

- Which department they come from and what is their price range?

Let's take a look at data...

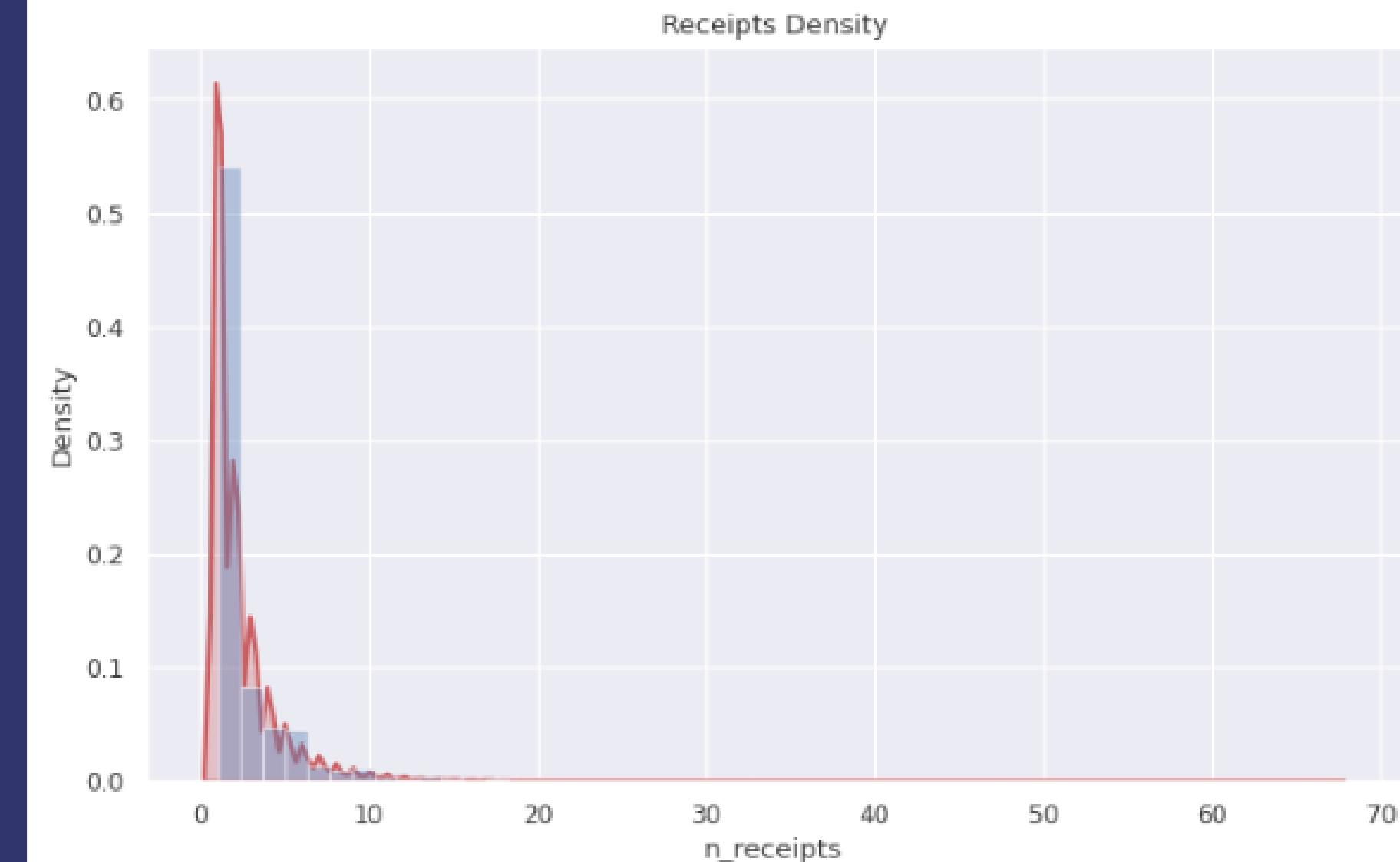
# Data exploration:

Average expenditure is concentrated from 23 to 128 with median receipt close to 55. Outliers from 10k to almost 80k.



mean	142.70
std	679.99
min	0
25%	23.98
50%	54.10
75%	128.89
max	78442.9 6

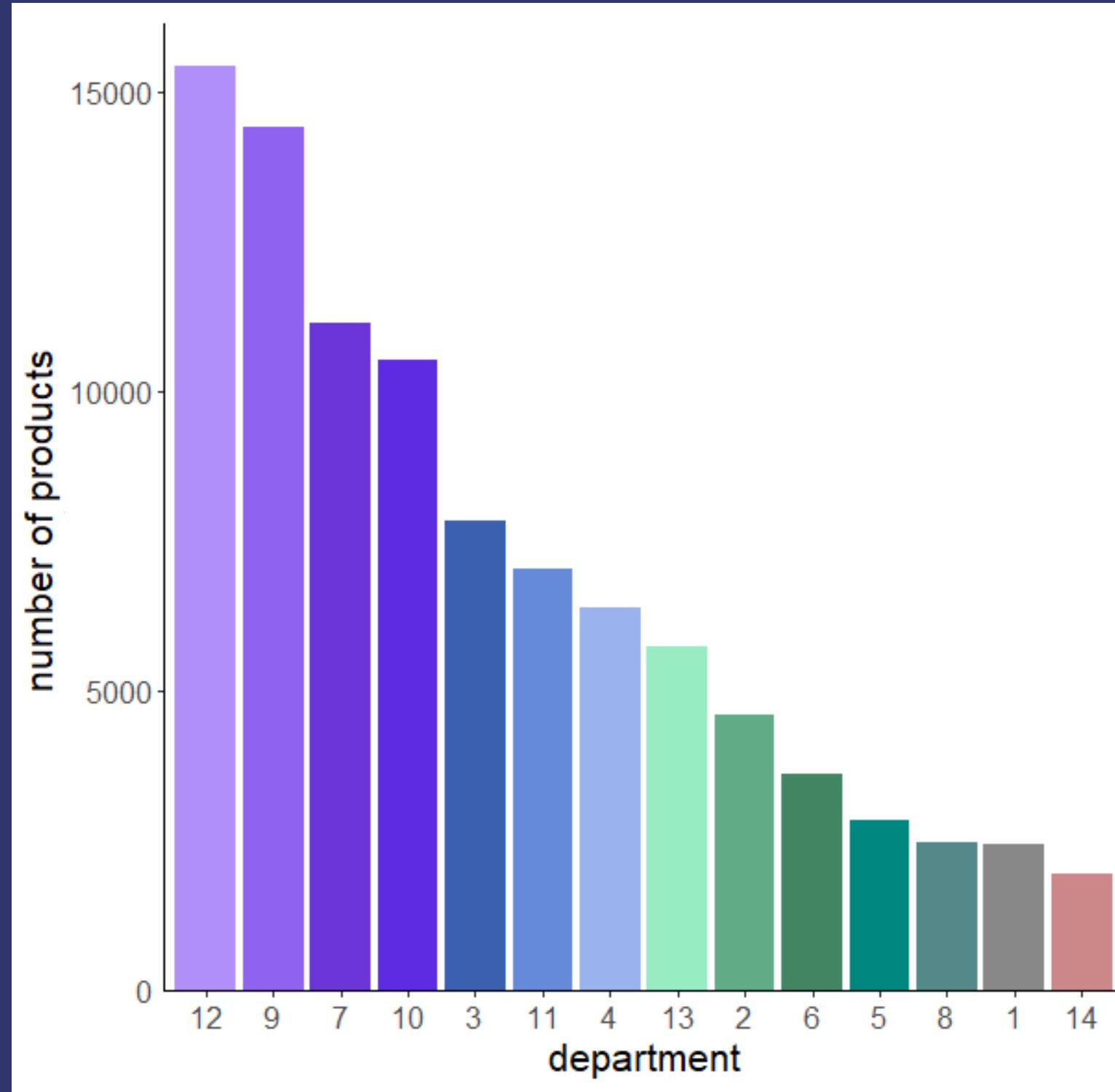
## Customer purchasing behavior



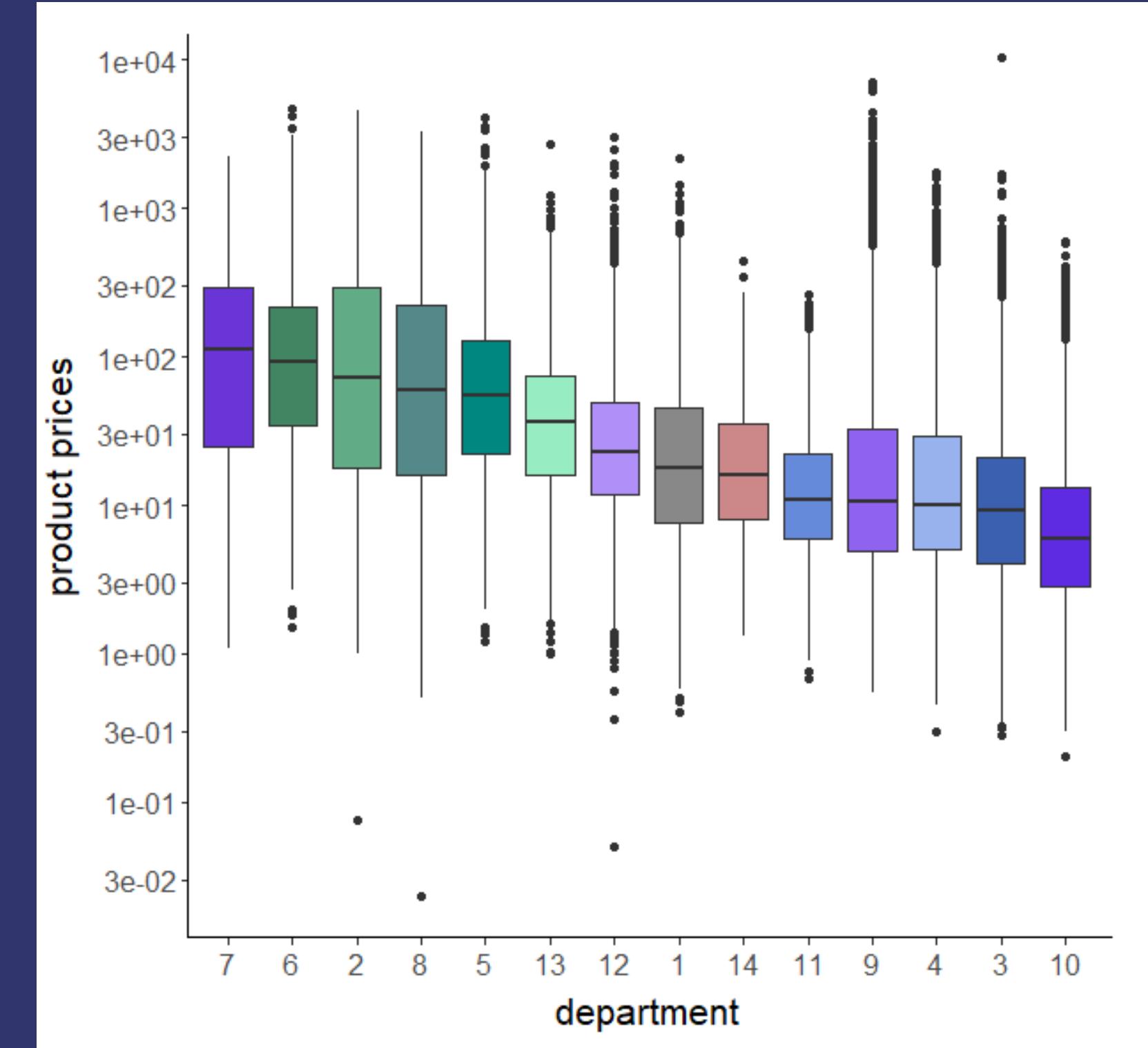
Most of the customers generate from 1 to 10 receipts.  
Only a small portion of the customer base generates a number of receipts between 20 and 65.

# Data Exploration

## Products descriptive analysis

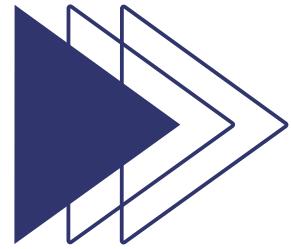
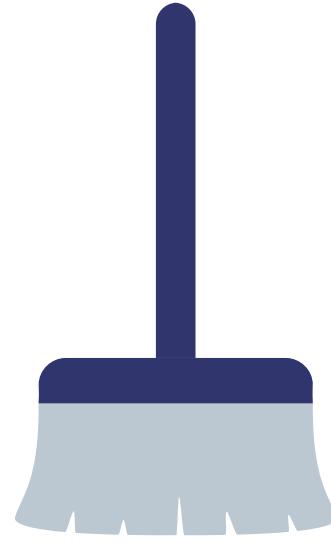


Department #12 is the one with the most product sold.



Whereas Department #7 is, on average, the department with most expensive products.

# Data-driven strategy

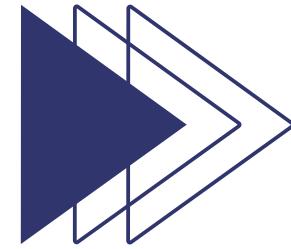


## Cleaning

Missing Values  
handling &  
inconsistency  
resolution

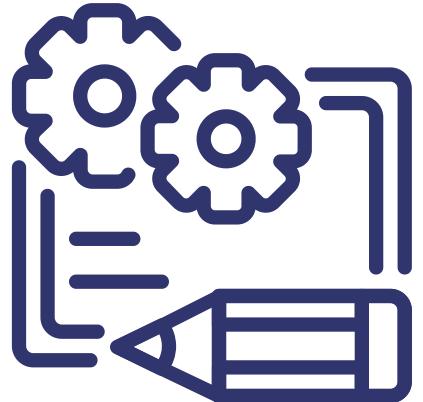
## Preprocessing

Data type  
conversion,  
features  
transformation  
and creation



## Exploration

Descriptive analysis of  
customers purchasing  
behavior & sold  
products



## Modelling

Descriptive and  
predictive modelling  
answering the  
business questions in  
a quantitative and  
qualitative way

# Models

## 1) Customer Relationship Valorization:

(Trimestral analyses using the following approaches)

### Cluster Analysis:

customers behavioural segmentation

### RFM Model:

customers deterministic segmentation & movement between segments

### Churn Model:

prediction of leaving customers

## 2) Products Valorization

(Annual Analyses)

### Market Basket Analysis:

study of product affinities by purchases



# Cluster Analysis

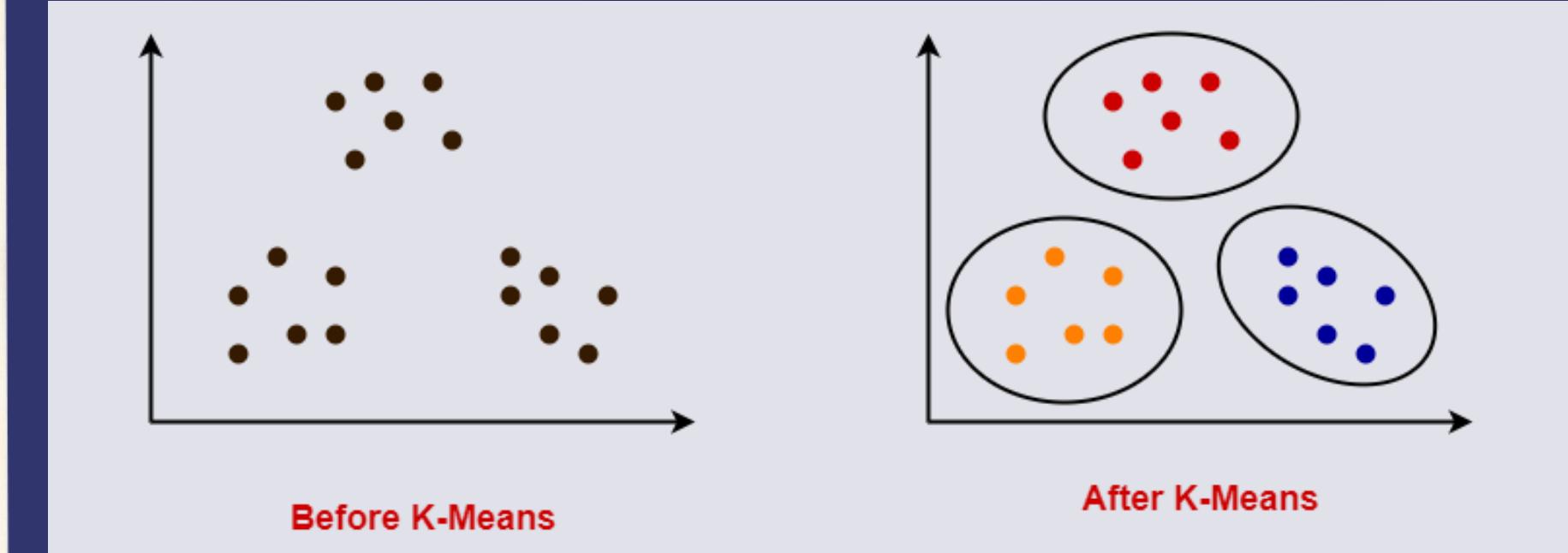
## K - Means Algorithm

Reference period: 01/02/19 - 30/04/19

### How it works:

K-means clustering uses “**centroids**”, K different randomly-initiated points in the data, and assigns every data point to the nearest centroid.

After every point has been assigned, the centroid is moved to the **average of all of the points** assigned to it.



### FEATURES:

- Number of receipts
- Number of products
- Number of clicked mails
- Average expenditure
- Average discount rate

# RFM

Reference period: 01/02/19 - 30/04/19

## Recency

How recently was the customer's last purchase?

## Frequency

How often did this customer make a purchase in a given period?

## Monetary Value

How much money did the customer spend in a given period?

FEATURE → Last purchase date

FEATURE → Number of receipts

FEATURE → Total net import

## Segments:



**Diamond**  
Premium Customers



**Gold**  
High Value Customers



**Silver**  
Medium-High Value  
Customers



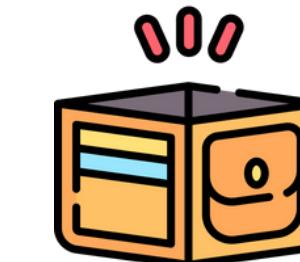
**Bronze**  
Average Customers



**Copper**  
Medium-Low Value  
Customers



**Tin**  
One-Timers



**Cheap**  
Low Value Customers

Icons from <https://www.flaticon.com/>

# Churn Model

Lookback Period -> 01/11/18 - 31/01/19

Holdout Period -> 01/02/19 - 30/04/19

## FEATURES:

- Number of receipts
- Total amount spent
- Average discount rate
- Number of campaigns
- Opened e-mails rate (%)
- Clicked e-mails rate (%)
- Returned item (binary)
- Privacy subscription 1 (binary)
- Privacy subscription 2 (binary)
- Privacy subscription 3 (binary)

Not Churner	Churner	Inactive
Active fidelity card & at least 1 purchase	Active fidelity card but no purchases;  Purchased but Inactive fidelity card	No purchases & Inactive fidelity card;  Fidelity card activated after 01/11/18

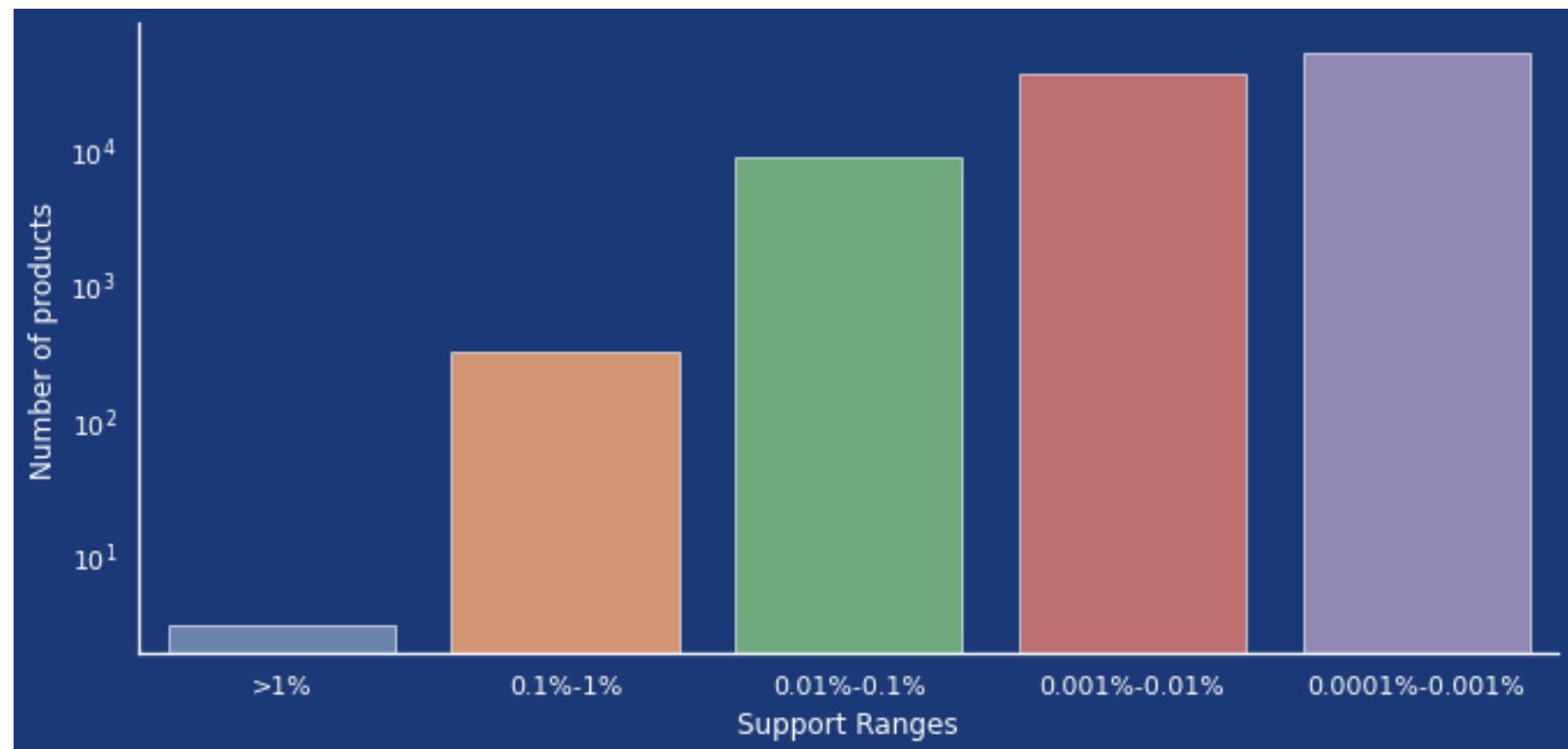
# Market Basket Analysis

Reference period:  
01/05/18 - 30/04/19

Looking for joint purchasing patterns



Affinities among products with high support



Support >0.1% chosen based  
on the above distribution



# Insights

## Results:

### 1) Customer Relationship Valorization:

- Cluster Analysis
- RFM Model
- Churn Model

### 2) Products Valorization

- Market Basket Analysis



# Cluster Analysis - Results



**Loyal Customer  
(1.33%)**

Customers who buy  
very frequently

num. receipts: 16.95  
num. products: 104.8  
avg. discount: 3.7 %  
avg. bill: 177 \$  
num. tot clicks: 0.35



**One-time buyer  
(80.41%)**

Customers who buy on  
average less than 2 times  
per quarter

num. receipts: 1.63  
num. products: 5.81  
avg. discount: 2.5 %  
avg. bill: 89 \$  
num. tot clicks: 0.09



## Clicker (2.53%)

Customers who are very responsive to campaigns

num. receipts: 3.29  
num. products: 13.94  
avg. discount: 3.7 %  
avg. expenditure: 107 \$  
**num. tot clicks: 5.22**



## Premium Customer (2.90%)

Customers who buy high-priced products

num. receipts: 1.75  
num. products: 14.36  
avg. discount: 6.5 %  
**avg. expenditure: 1221 \$**  
num. tot clicks: 0.1



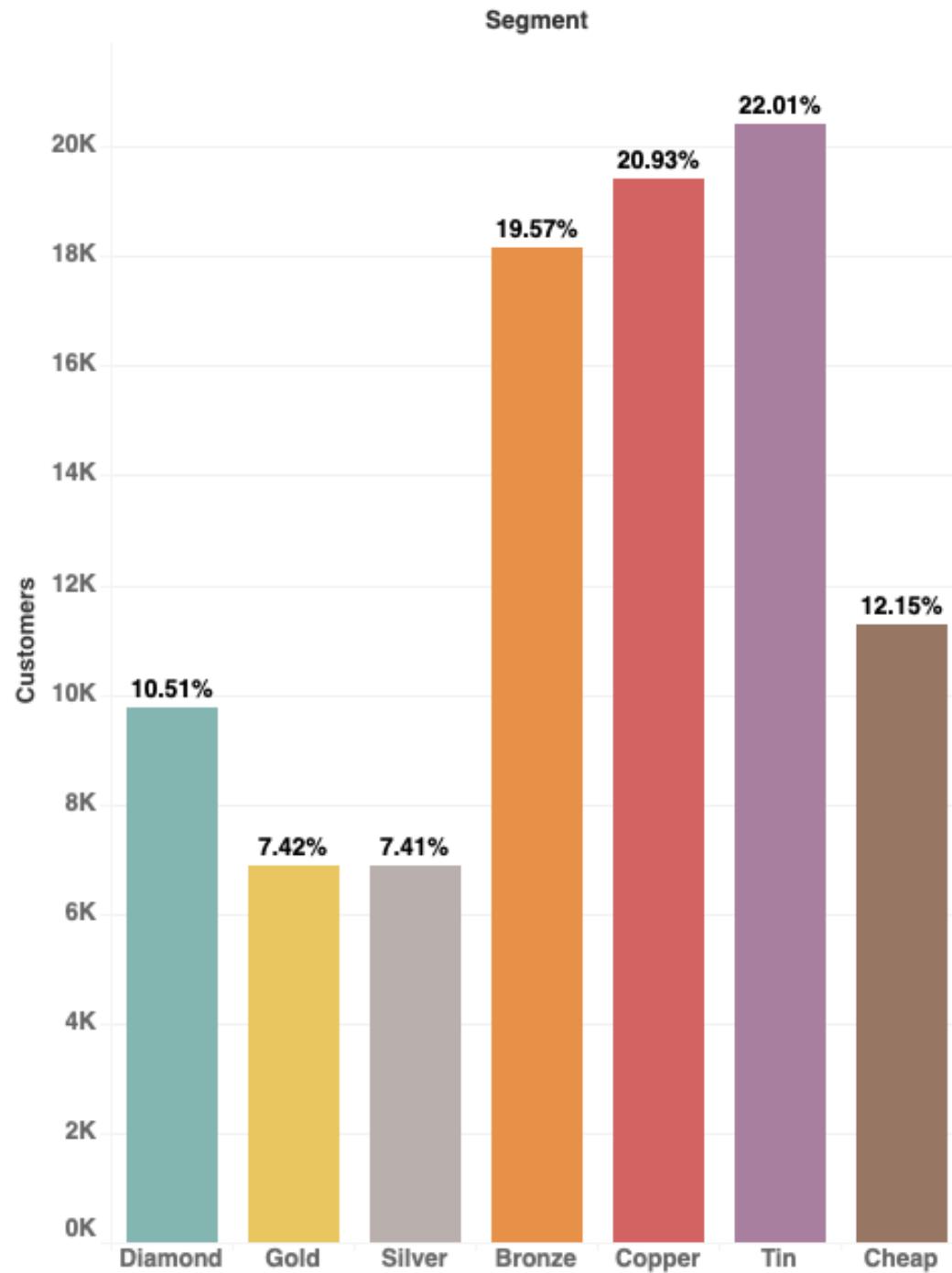
## Average Customer (12.83%)

Customers with average purchasing behavior

num. receipts: 6.16  
num. products: 29.4  
avg. discount: 2.8 %  
avg. expenditure: 128 \$  
num. tot clicks: 0.16

# RFM – Results

Customers Segmentation Nov 2018 to Jan 2019

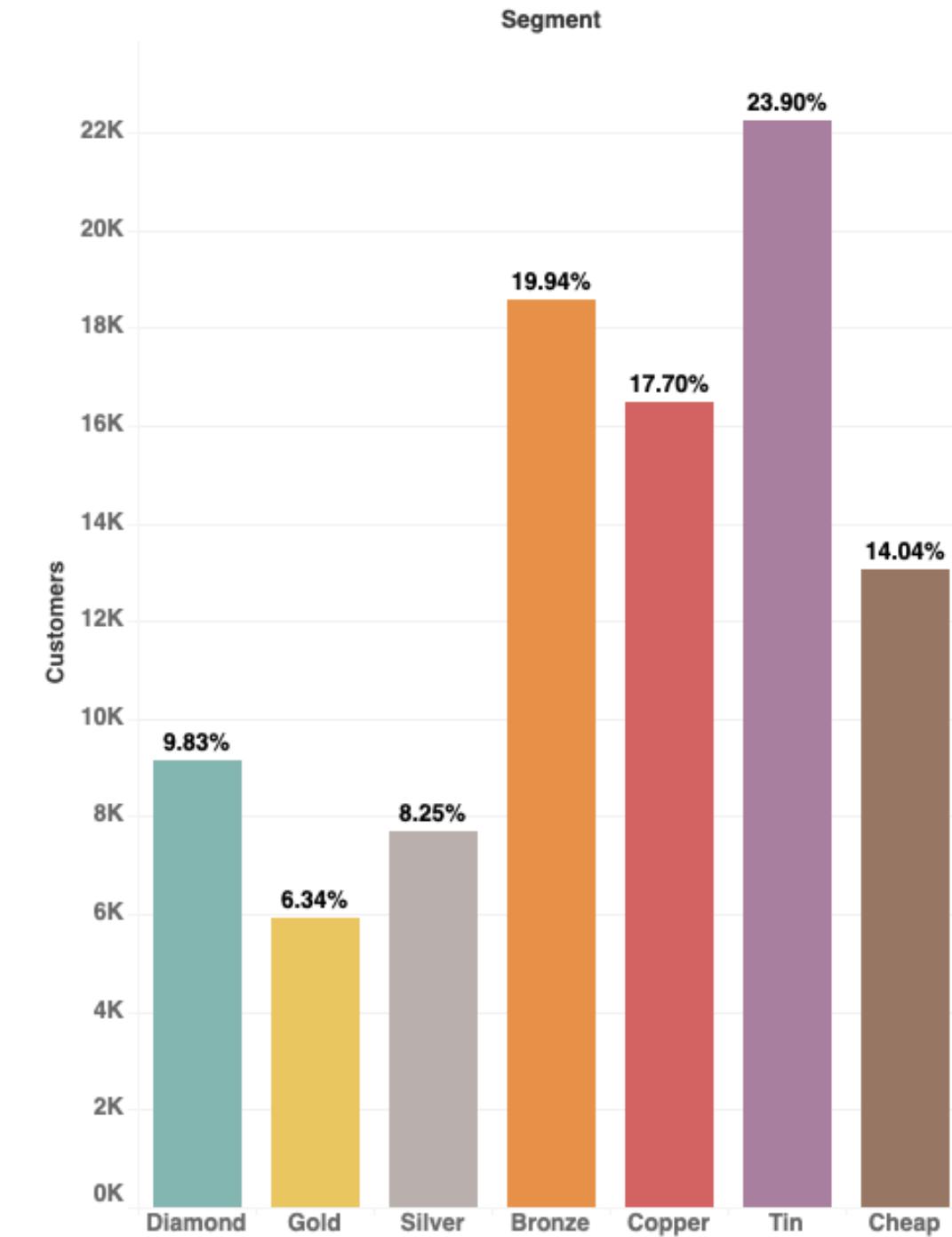


Comparison of segments built over 2 consecutive trimesters

The percentage of customers in the higher-value segments decreased over the 2 periods

Lower-value segments increased over the 2 periods

Customers Segmentation Feb to Apr 2019



Let's take a look closer

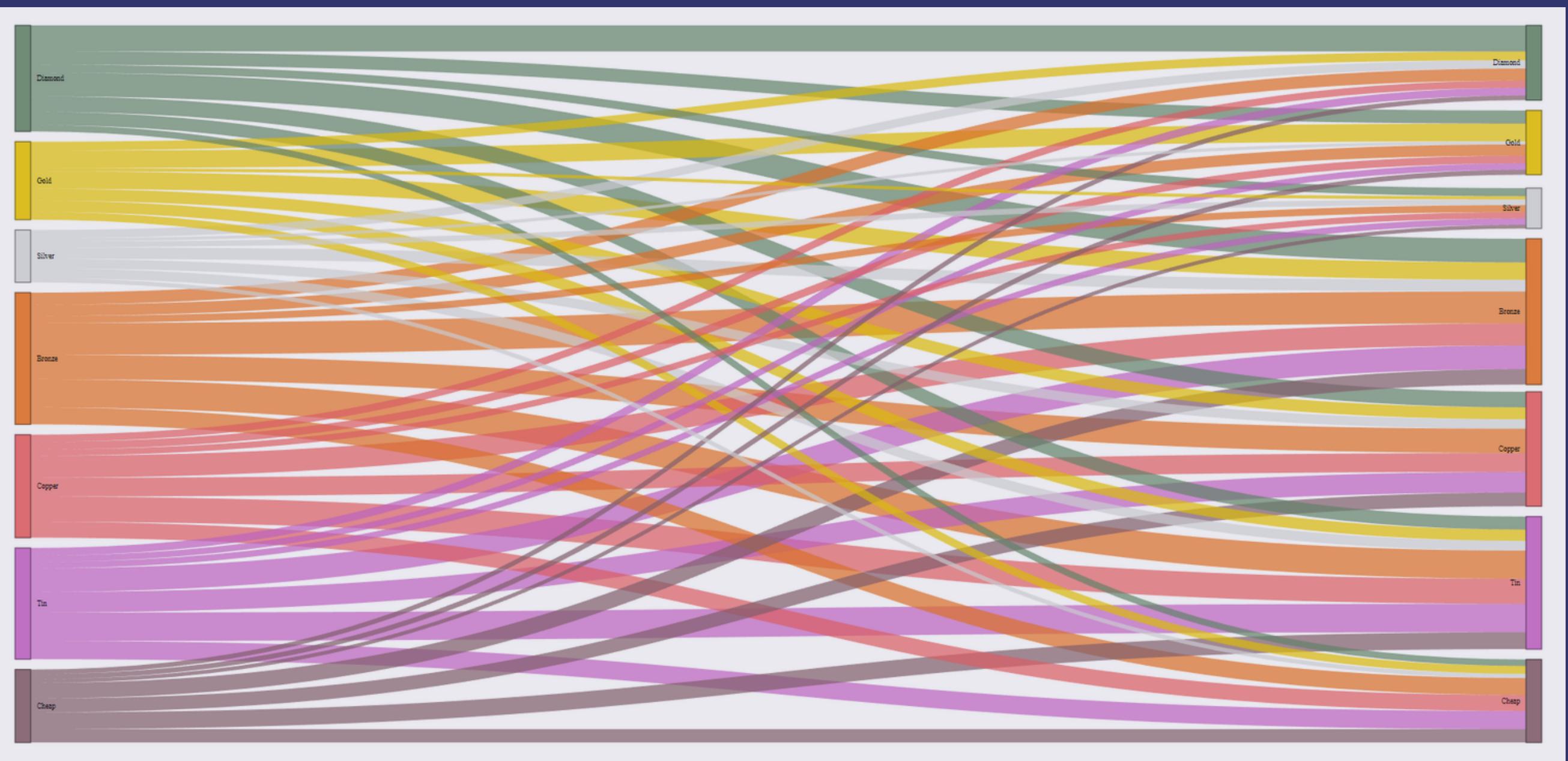


# Customer journey among segments

01/11/2018 to 31/01/2019

01/02/2019 to 30/04/2019

Actionable tips



- Stayed in their high value segment  
--> Upsell higher value products. Ask for reviews. Engage them
- From high to higher value segment  
--> Reward them. Can be early adopters for new products and will promote the shop
- From high to lower value segment  
--> Win them back via renewals or newer products, don't lose them to competition. Reconnect with them
- From medium to higher value segment  
--> Offer membership / loyalty program, recommend other products

For interactive visualization click [here](#)

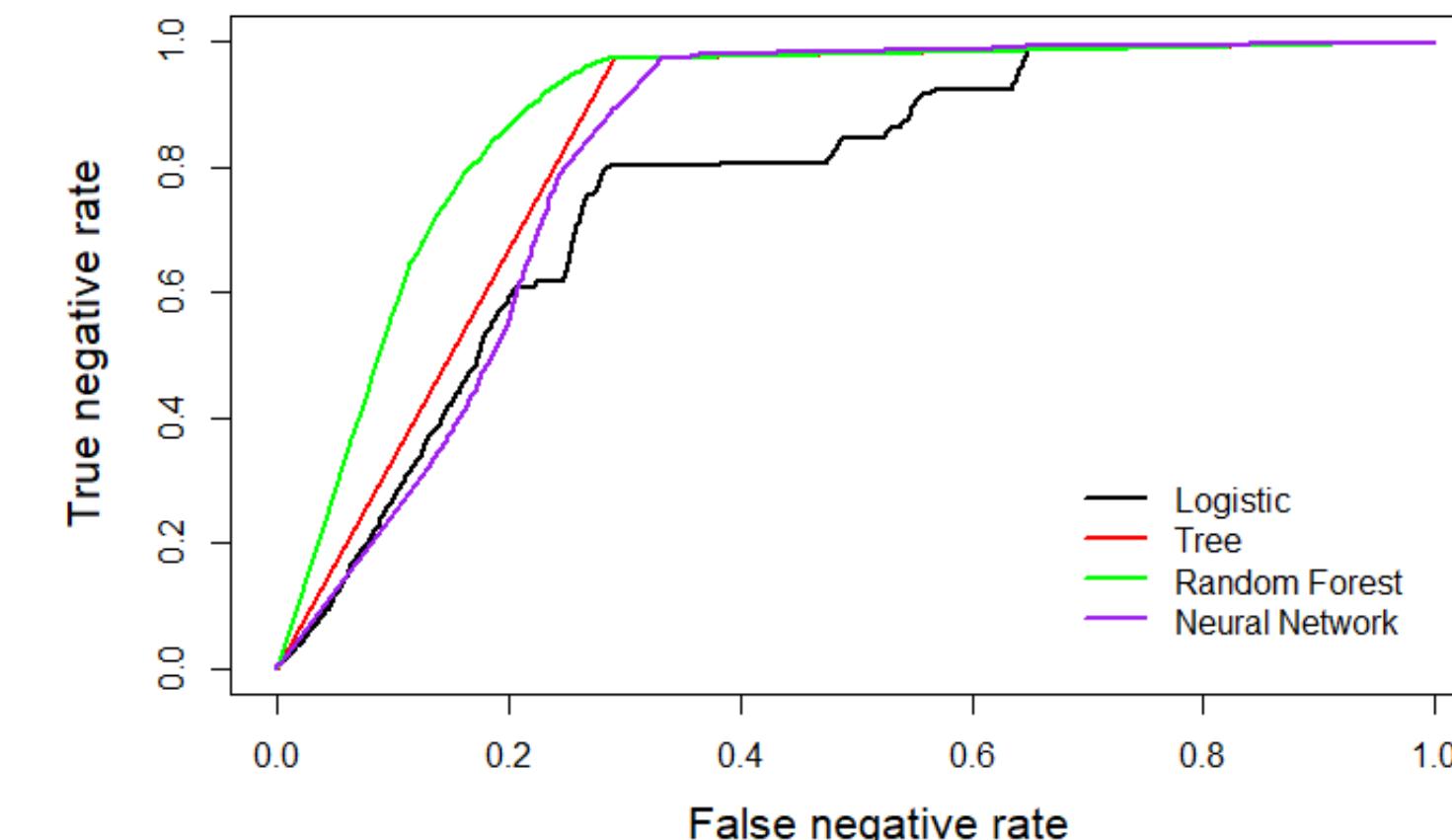
# Churn Model - Results

MODEL	ACCURACY	SPECIFICITY	RECALL
Logistic	0.7389	0.5908	0.8003
Tree	0.7871	0.9739	0.7096
<u>Random Forest</u>	<b>0.8225</b>	0.8369	0.8165
<u>Neural Network</u>	0.7570	<b>0.9747</b>	0.6667

**AREA UNDER THE CURVE (AUC)**  
Logistic: 0.7695  
Tree: 0.8418  
**BEST PERFORMING MODEL** ↘ Random Forest: **0.8874**  
Neural Network: 0.8202

**CHURNER**  
**29.32% (19.162 customers)**

**NOT CHURNER**  
**70.68% (46.184 customers)**

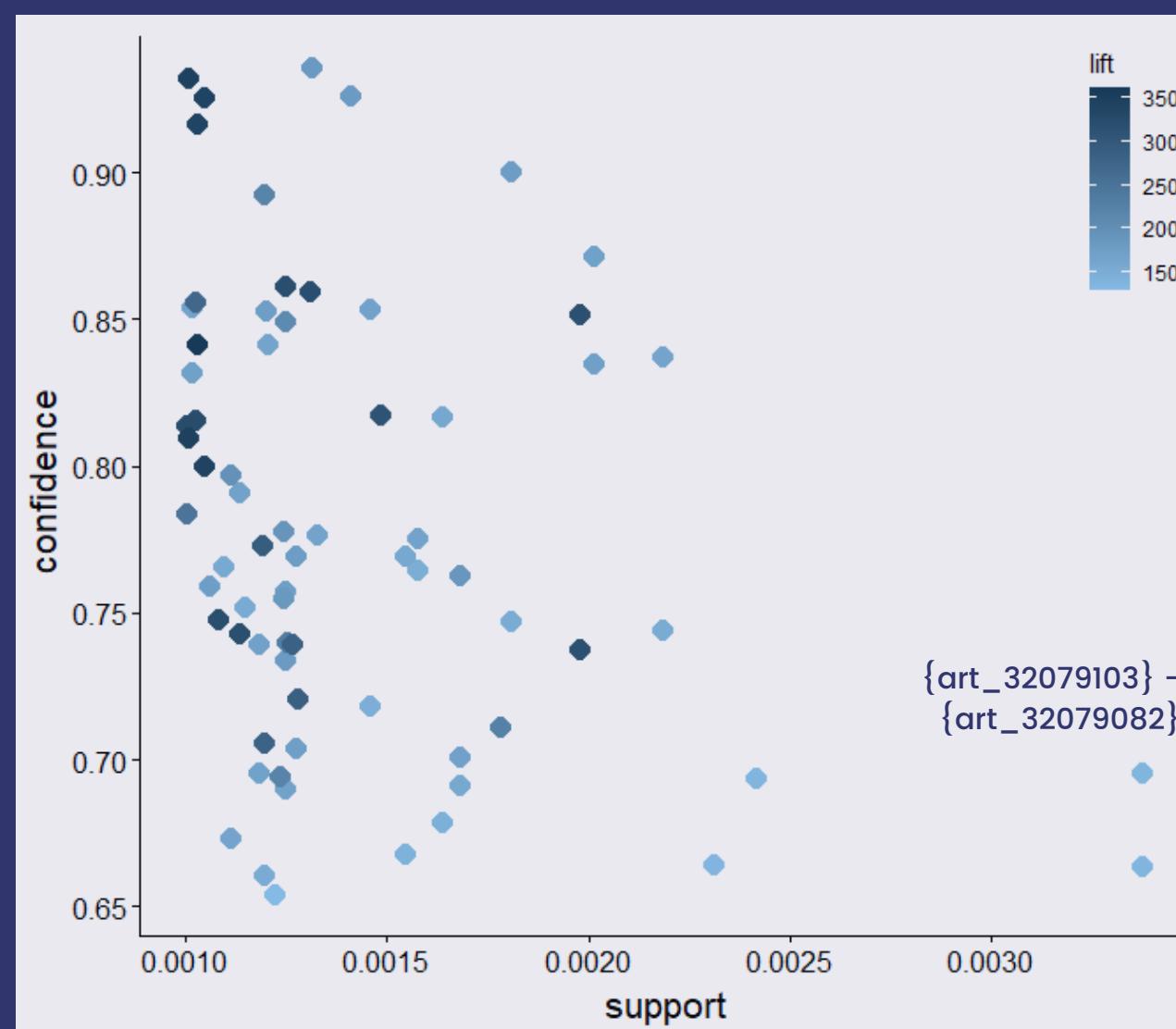


# Market Basket Analysis - Results

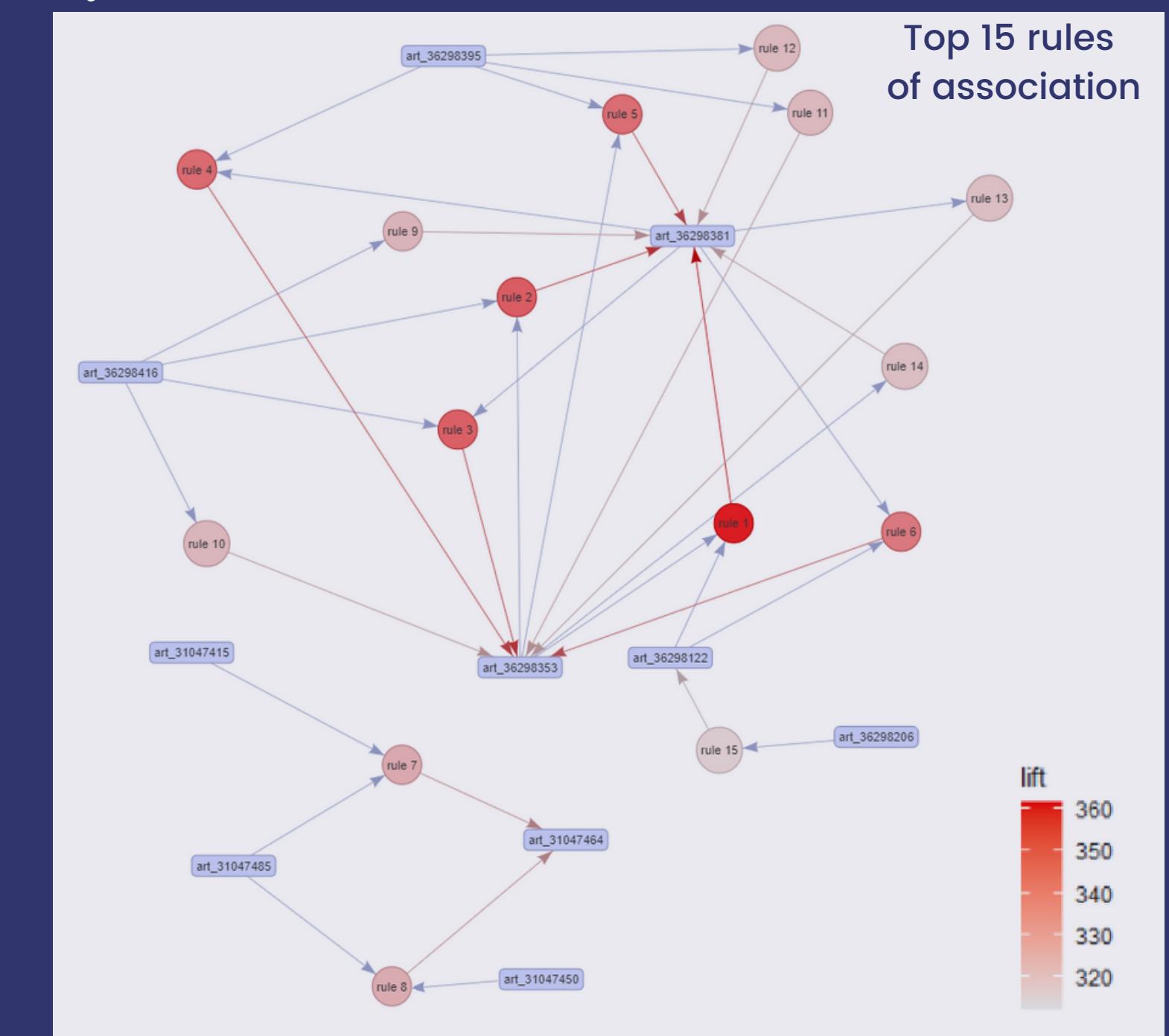
Affinities among products with high support (>0.1%).

Associations with confidence >65% & support >0.1%

71 Non redundant rules between sets of two & three products

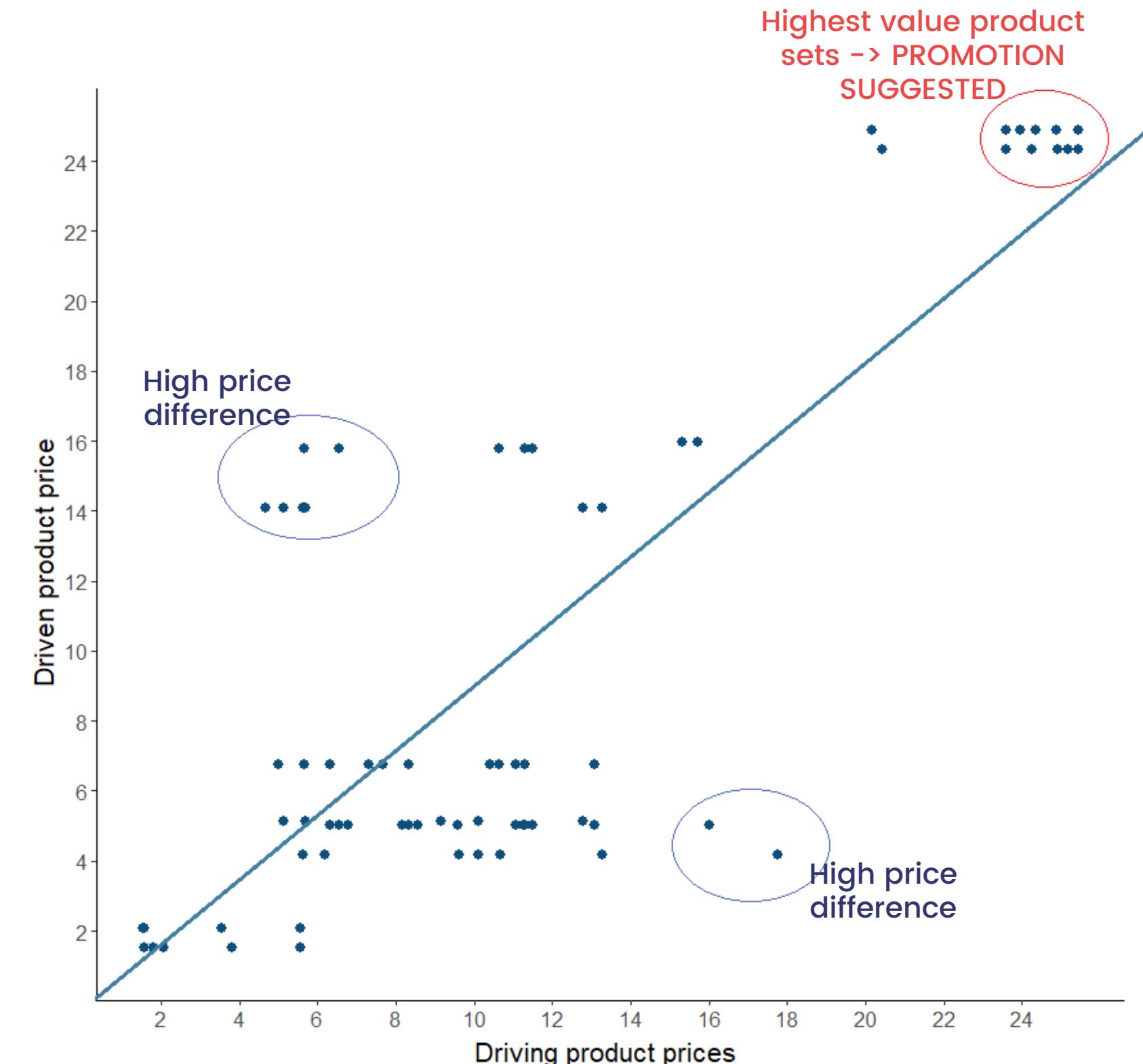


Article 36298381 & 36298353 are the most 'driven' products

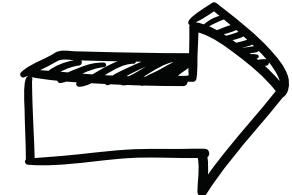


# Market Basket Analysis – Results

- > Meaningful affinities between **department #3** products **only**
- > The majority of the product sets have **driving** and **driven** products **around the same price**
- > Only **few sets** present a **high price rate** between driving and driven products (ex. {art\_31618300, art\_31618321, art\_31618405})
- > **Cross selling actions** focused on **high value product sets** (ex. {art\_36298395, art\_36298381}, having respectively a **25.43\$** and **24.35\$ average price**)
- > **Layout optimization** for strategic collocation of products in **department #3**



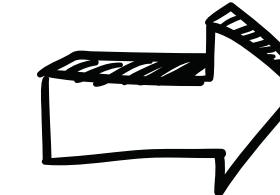
# Summary Marketing Strategies



## Up-selling actions personalized for each high value segment

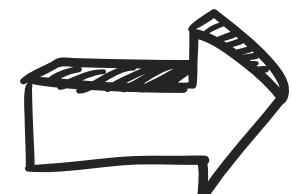
### Customer

Optimization of the relationship with customers can be obtained via **personalized up-selling** actions before and during the purchasing moment. This strategy can be implemented to **increase the value, the life time** and the **satisfaction** of already valuable customers. **Loyal customers** are indeed the **most susceptible** to product recommendations.



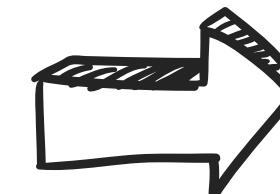
## Potential churner forecasting based on the developed model and consequent caring actions

**Retaining** potential churners is **less expensive** than acquiring new customers -> It is crucial to **implement caring actions** not to lose them. Examples of **potential targeted actions**: addressing special offers, rationalization and diversification of the marketing contact strategy (either via call-centers, e-mails, etc.) according to the risk of churning.



## Level up: from Tin and Copper to Bronze

The **majority** of the customer base resulted to fall in the "Tin" and "Copper" segments. It would be valuable to **shift as many customers as possible** to more **profitable segments**. This can be achieved **engaging** the customers by making **limited time offers** and **recommendations** based on past purchases. Other effective strategies include: providing **on-boarding support**, offering free trials, sending **personalized emails** to reconnect, offering renewals.



## Cross selling actions focused on high value product sets

**Recommendation** of products based on the found association rules, to integrate "next-best-product" to buy actions.

**Cross selling** can be implemented both in the **pysical store** and in the website, by means of specific **e-commerce actions**