

## **MRA PROJECT - 1**

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## **AGENDA**

Problem Statement

Data Analysis

**Exploratory Data Analysis** 

Customer Segmentation using RFM Analysis

RFM Inferences and Identified Segments

#### Problem Statement

An automobile parts manufacturing company has collected data on transactions for 3 years. They do not have any in-house data science team, thus they have hired you as their consultant. Your job is to use your data science skills to find the underlying buying patterns of the customers, provide the company with suitable insights about their customers, and recommend customized marketing strategies for different segments of customers.

## **DATA ANALYSIS**

#### **INFO**

- 1. Dataset is having 20 variables out of which 12 are categorical, 7 Numerical and one date field
- 2. There is no missing values and duplicate values found.
- 3. Total Number of observations are 2747

<class 'pandas.core.frame.DataFrame'> RangeIndex: 2747 entries, 0 to 2746 Data columns (total 20 columns): Column Non-Null Count Dtype ORDERNUMBER 2747 non-null int64 QUANTITYORDERED 2747 non-null PRICEEACH 2747 non-null float64 ORDERLINENUMBER 2747 non-null int64 SALES 2747 non-null float64 ORDERDATE 2747 non-null datetime64[ns] DAYS\_SINCE\_LASTORDER 2747 non-null int64 object STATUS 2747 non-null PRODUCTLINE 2747 non-null object 2747 non-null int64 MSRP object PRODUCTCODE 2747 non-null 11 CUSTOMERNAME 2747 non-null object 2747 non-null object 12 PHONE 13 ADDRESSLINE1 2747 non-null object 2747 non-null 14 CITY object 15 POSTALCODE 2747 non-null object object 16 COUNTRY 2747 non-null CONTACTLASTNAME 2747 non-null object CONTACTFIRSTNAME 2747 non-null object 19 DEALSIZE 2747 non-null object dtypes: datetime64[ns](1), float64(2), int64(5), object(12)

memory usage: 429.3+ KB

#### Data Shape

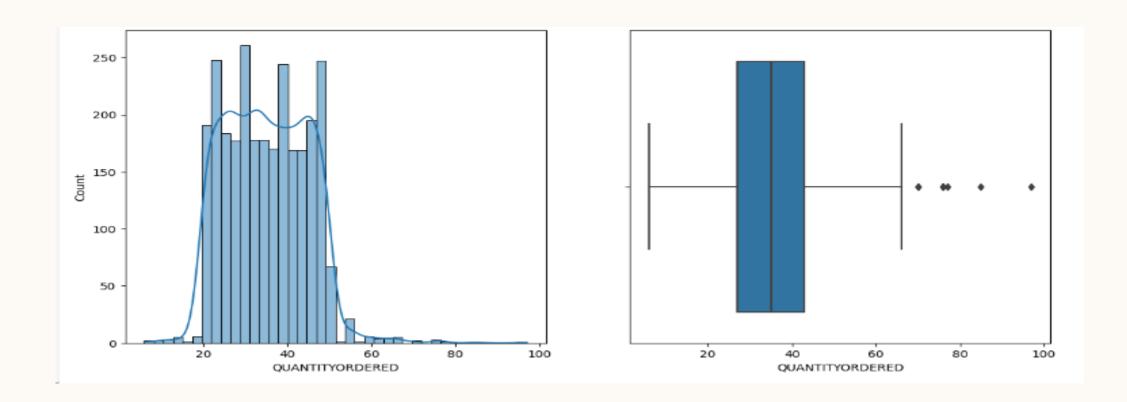
(2747, 20)

#### **Data Summary**

	count	mean	min	25%	50%	75%	max	std
ORDERNUMBER	2747.0	10259.761558	10100.0	10181.0	10264.0	10334.5	10425.0	91.877521
QUANTITYORDERED	2747.0	35.103021	6.0	27.0	35.0	43.0	97.0	9.762135
PRICEEACH	2747.0	101.098951	26.88	68.745	95.55	127.1	252.87	42.042548
ORDERLINENUMBER	2747.0	6.491081	1.0	3.0	6.0	9.0	18.0	4.230544
SALES	2747.0	3553.047583	482.13	2204.35	3184.8	4503.095	14082.8	1838.953901
ORDERDATE	2747	2019-05-13 21:56:17.211503360	2018-01-06 00:00:00	2018-11-08 00:00:00	2019-06-24 00:00:00	2019-11-17 00:00:00	2020-05-31 00:00:00	NaN
DAYS_SINCE_LASTORDER	2747.0	1757.085912	42.0	1077.0	1761.0	2438.5	3562.0	819.280576
MSRP	2747.0	100.691684	33.0	68.0	99.0	124.0	214.0	40.114802

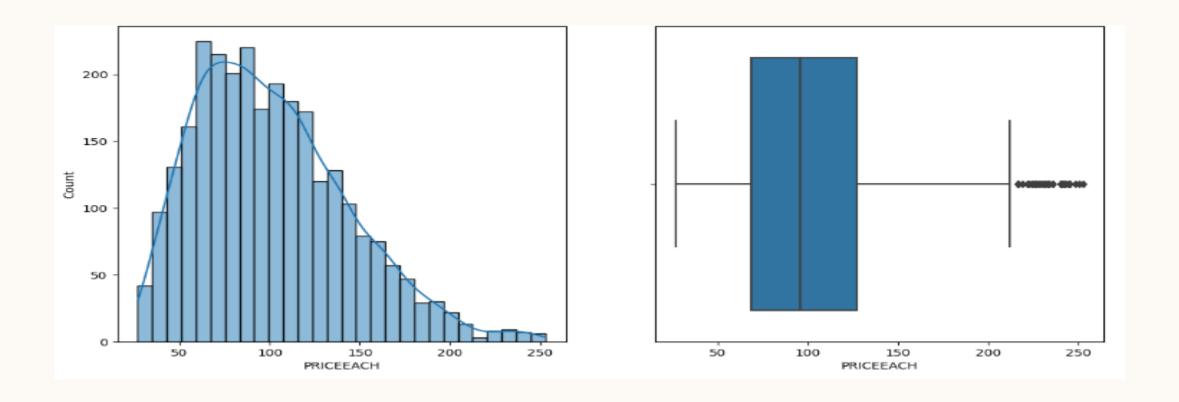
#### **Order Quantity**

After Exploration of the data it is found that there are Outliers present in this variable and data not perfect normally distributed.



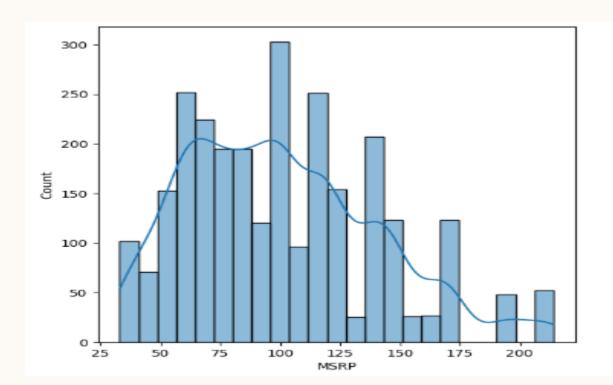
#### **Price of Each Item**

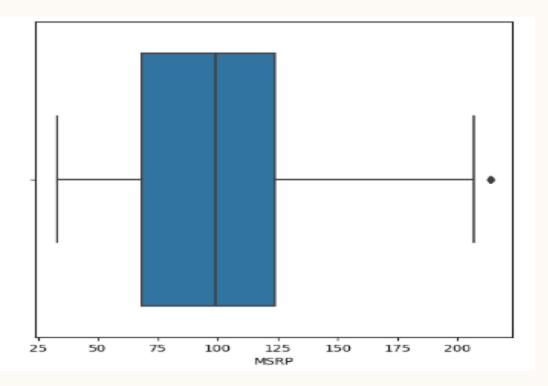
This variable is having approximate normal distribution (slight right skewed) but having many outliers.



#### **Manufacturer's Suggested Retail Price**

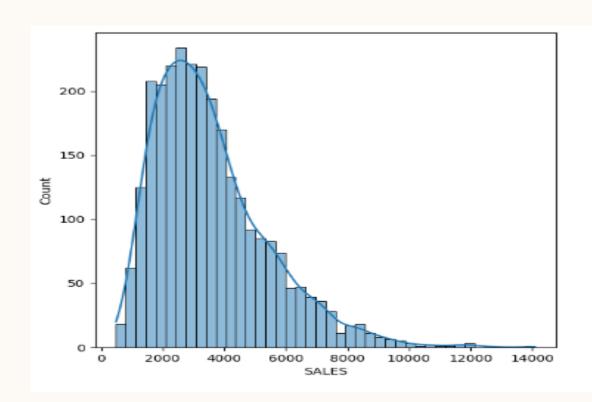
This variable is having very few outliers

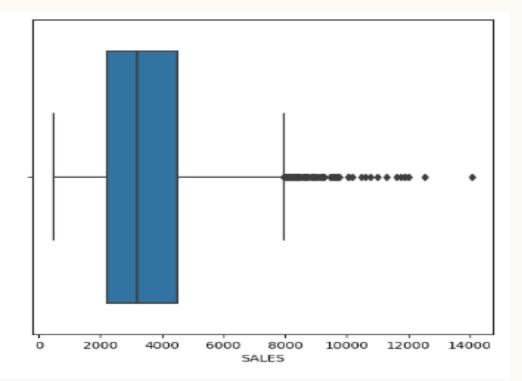




#### **Sales Amount**

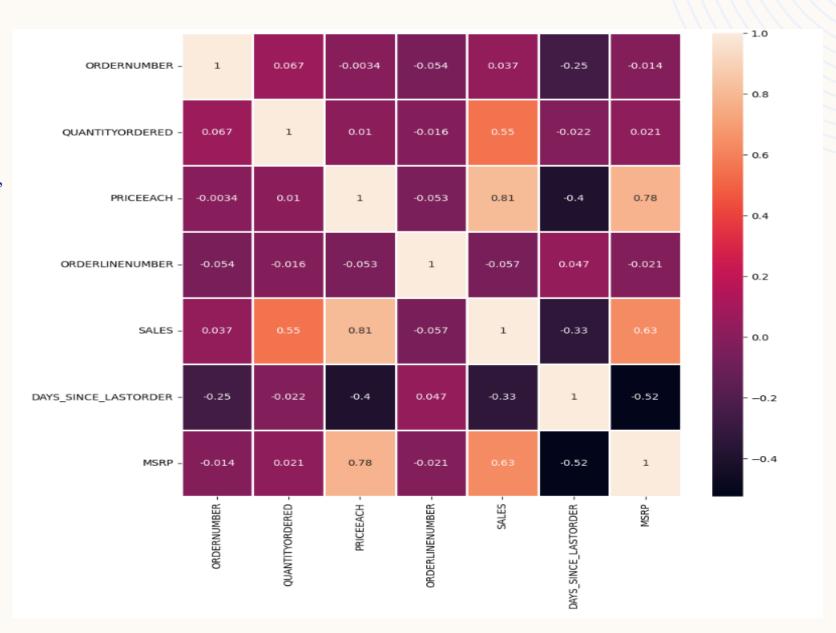
This variable is right skewed with a lot of outliers.





#### **Correlation**

Highly correlated variables are SALES and PRICEEACH, MSRP and PRICEEACH, SALES and QUANTITYORDERED, SALES and MSRP



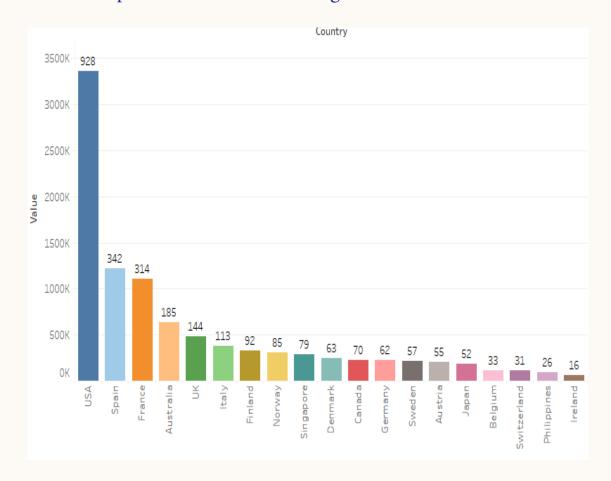
#### **EXPLORATORY DATA ANALYSIS**

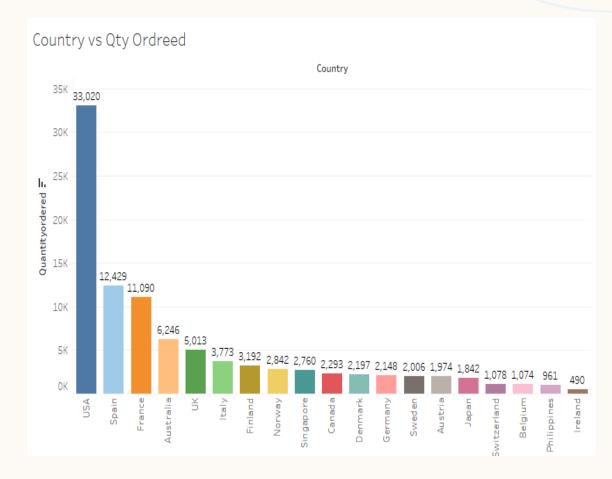
- 1. UNIVARIATE ANALYSIS
- 2. BIVARIATE ANALYSIS
- 3. MULTIVARIATE ANALYSIS

TABLEAU LINK: RFM\_Harikrishnan | Tableau Public

#### **Sales across countries**

USA is top most seller in the among countries.

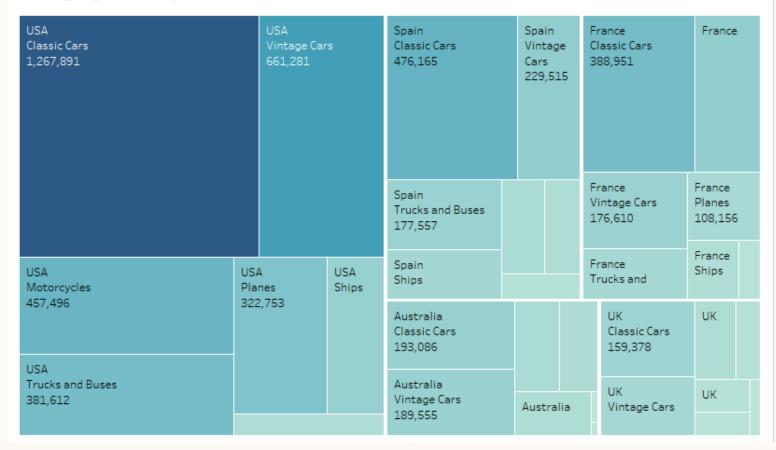




#### **Category, Country Vs Sales**

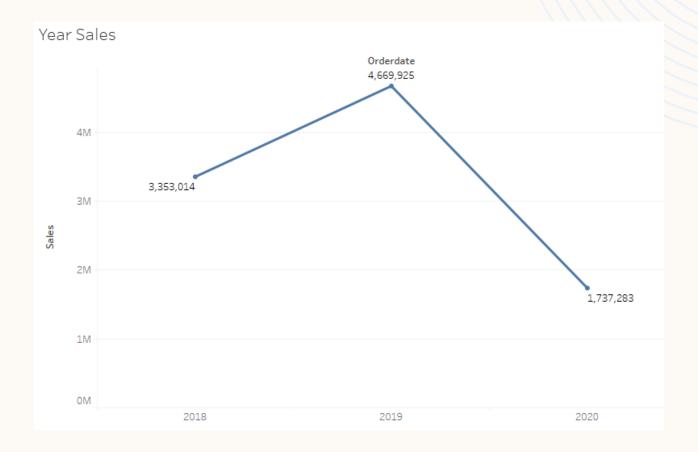
Classic cars have the highest sales, while trains have the lowest.

#### Catergory, Country vs Sales



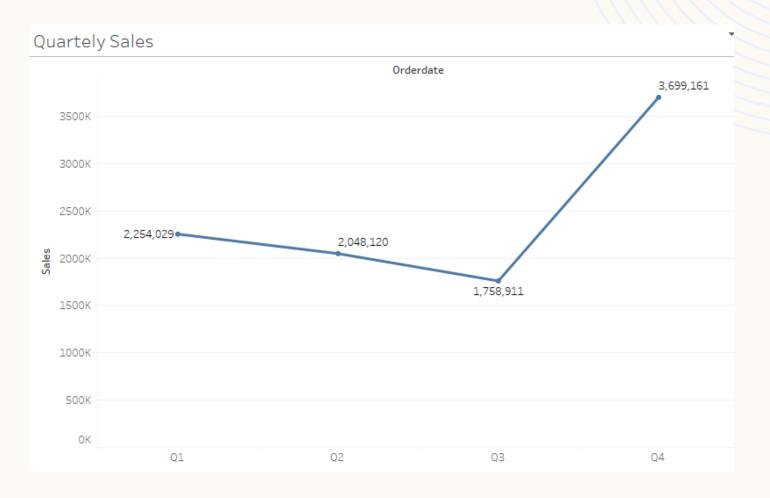
#### **Yearly Sales**

2019 appears to be the best year for sales. The year 2020 has the lowest sales.



#### **Quarterly Sales**

Sales decreased from Q1 to Q3, but increased dramatically in Q4.



#### **Monthly Sales**

The month of November had the biggest sales. June is a month with poor sales.



#### **Day Sales**

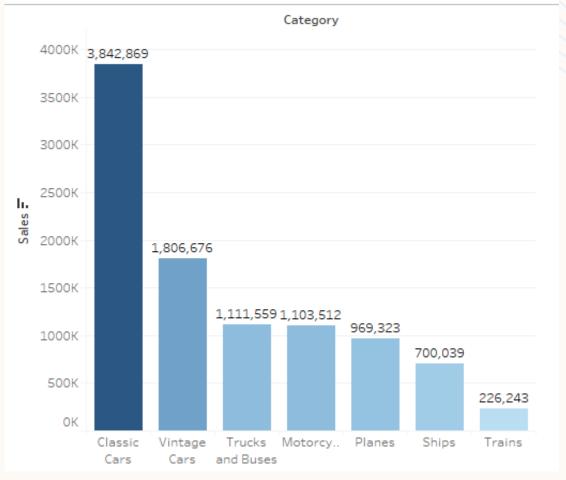
There is a pattern in day sales. People want to order in the first few days of the month. People ordered less at the end of the month.



## **Bivariate Analysis – Category vs Sales**

Classic cars have the highest sales, while trains have the lowest.





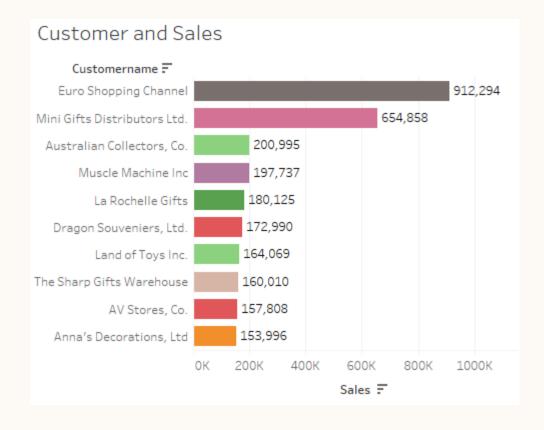
#### Bivariate Analysis – Deal size vs Sales

- 1. When compared to small and large deals, medium-sized deals produce much larger sales.
- 2. Large size deal sales have almost remained flat throughout the years, and it may be assumed that the corporation should focus on acquiring large size chunk projects.



### **Bivariate Analysis – Customer vs Sales**

- 1. In terms of sales volume, the Euro shopping channel appears to be the best customer.
- 2. The company is a customer driver because 4-5 customers account for the majority of their sales. As a result, the corporation should spend more on customer scouting in a strategic manner, because client attrition will have a significant impact on the company's sales.



## **Bivariate Analysis – Country vs Sales**

USA has the highest sales as compared to other countries

#### Country vs Sales

USA 3,355,576	France 1,110,917		ustralia 330,623
	UK 478,880	Finland 329,582	Norway 307,464
Spain 1,215,687	Italy 374,674	Singapore 288,488	Denmark 245,637

#### Bivariate Analysis – Country, City vs Sales

City Madrid in country Spain is delivering the highest sales as compared to other countries

#### Country, City vs Sales

USA San Rafael 654,858	USA New Bedford 207,875		USA San Jose 160,010	USA Boston 154,070	France Paris 268,945		France Reims 135,043		Australia		
	USA Philadelphia	USA	USA	USA	France Nantes 204,305				Australia		
USA	USA				France						
NYC 560,788	Cambridge	USA San Di	USA ego Las	USA New	UK	UK					
	USA Nashua	USA			Manchester 157,808						
	USA	USA	USA		UK						
	Allentown	03/	00/1		London	UK Cowe	Finla	Finland		Norway	
Spain Madrid 1,082,551				Spain	Italy Reggio Emilia			apore apore 488			
				Spain	Italy Bergamo						

#### **SUMMARY ADDON**

- Classic cars are the most popular and generate the most income.
- Planes, trucks, and ships have the longest order lines.
- The top two selling countries are the United States and Spain.
- San Rafael and New York City have the highest sales in the United States, while Madrid is the largest market in Spain
- Euro Shopping Channel and Mini Gifts Distributor are two of the most important consumers.
- Classic cars, as well as Euro shopping and Mini presents, are at the top of the list.
- Annual sales are down, but when broken down, they show a little upward tendency.

#### **RFM**

- > RFM, also known as RFM analysis, is a type of customer segmentation and behavioral targeting used to help businesses rank and segment customers based on the recency, frequency, and monetary value of a transaction.
- ➤ RFM marketing can help marketers and small business owners determine their target audience to use their budget most effectively
- ➤ This method gives customers scores based on 3 factors: recency, frequency, and monetary
- Recency refers to how recent a customer's last purchase was. Customers who have made a recent purchase, typically within the last few weeks, still have the product and brand on their minds and are most likely to make a repeat purchase.
- Frequency is how often the customer makes purchases, which can help you identify repeat customers. For example, many clients make frequent repeat purchases within a set timeframe
- Monetary value refers to how much a customer spends within a given period. It's always important to consider because it can tell you a few things about consumer behavior

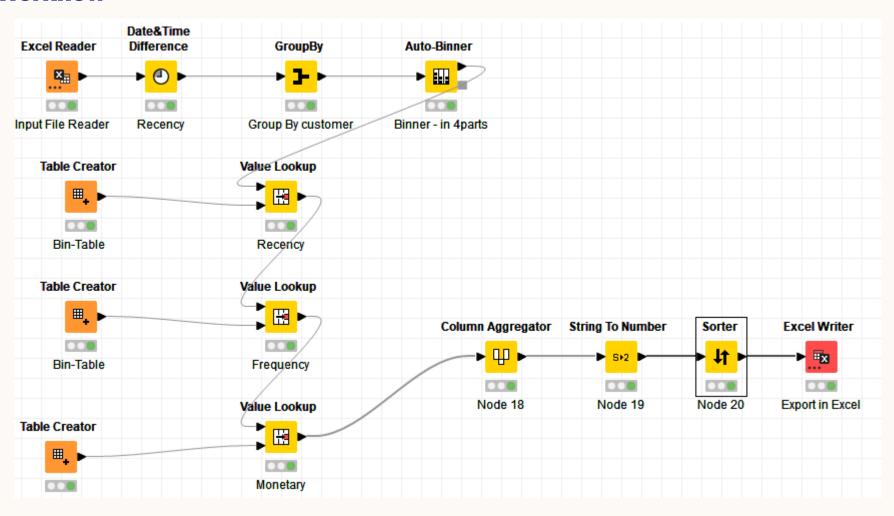
## CUSTOMER SEGMENTATION USING RFM ANALYSIS

- ➤ KNIME tool is used for RFM analysis and customer segmentation
- Customer name, Quantity ordered, Order date, order quantity and Sales are the used parameters.
- **Predictions** 
  - 1. Recency calculated using Max(ORDERDATE)-ORDERDATE
  - 2. Frequency calculated using count of order for a particular customer.
  - 3. Monetary is calculated using sum of sales through particular customer.

TABLEAU LINK: RFM\_Harikrishnan | Tableau Public [https://public.tableau.com/app/profile/harikrishnan.m4921/viz/RFM\_Harikrishnan/Story1?publish=yes]

#### **RFM Analysis**

#### **KNIME Workflow**



#### **RFM Analysis**

#### **Output Table Head**

Row ID	S CUSTOMERNAME	Count*	Sum(Q	D Mean(P	D Sum(SA	Count(	Count(	Count(	L Min*(R	Count(	S Recency	S Freque	S Monetary	RFM Sc
Row32	Euro Shopping Channel	259	9327	97.383	912,294.11	259	259	259	0	259	4	4	4	444
Row53	Mini Gifts Distributors	180	6366	102.696	654,858.06	180	180	180	2	180	4	4	4	444
Row43	La Rochelle Gifts	53	1832	97.046	180,124.9	53	53	53	0	53	4	4	4	444
Row78	The Sharp Gifts Ware	40	1656	93.376	160,010.27	40	40	40	39	40	4	4	4	444
Row72	Souveniers And Thin	46	1601	95.189	151,570.98	46	46	46	2	46	4	4	4	444
Row67	Salzburg Collectables	40	1442	101.398	149,798.63	40	40	40	14	40	4	4	4	444
Row26	Danish Wholesale Im	36	1315	108.038	145,041.6	36	36	36	46	36	4	4	4	444
Row41	L'ordine Souveniers	39	1280	111.147	142,601.33	39	39	39	21	39	4	4	4	444
Row63	Reims Collectables	41	1433	94.343	135,042.94	41	41	41	62	41	4	4	4	444
Row37	Handji Gifts& Co	36	1236	95.593	115,498.73	36	36	36	38	36	4	4	3	443
Row27	Diecast Classics Inc.	31	1111	108.566	122,138.14	31	31	31	1	31	4	3	4	434
Row79	Tokyo Collectables, Ltd	32	1150	101.183	120,562.74	32	32	32	39	32	4	3	3	433
Row83	UK Collectables, Ltd.	29	1046	108.536	118,008.27	29	29	29	53	29	4	3	3	433
Row9	Auto Canal Petit	27	1001	94.255	93,170.66	27	27	27	54	27	4	3	3	433
Row34	Gift Depot Inc.	25	903	108.932	101,894.79	25	25	25	26	25	4	2	3	423
Row77	Tekni Collectables Inc.	21	906	93.571	83,228.19	21	21	21	58	21	4	2	2	422
Row36	Gifts4AllAges.com	26	933	91.564	83,209.88	26	26	26	25	26	4	2	2	422
Row61	Petit Auto	25	796	93.494	74,972.52	25	25	25	1	25	4	2	2	422
Row62	Quebec Home Shoppi	22	717	104.515	74,204.79	22	22	22	30	22	4	2	2	422

## INFERENCE FROM RFM ANALYSIS

**BEST CUSTOMERS** 

CUSTOMERS ON THE VERGE OF CHURNING LOST CUSTOMERS

**LOYAL CUSTOMERS** 

## **BEST CUSTOMERS**

EURO SHOPPING CHANNEL MINI GIFTS
DISTRIBUTORS

LA ROCHELLE GIFTS

THE SHARP GIFTS WAREHOUSE

SOUVENIERS AND THINGS CO.

Total Sales 912294 with RFM 444

Total Sales 654858 with RFM 444

Total Sales 180124 with RFM 444

Total Sales 160010 with RFM 444

Total Sales 151570 with RFM 444

# CUSTOMERS ON THE VERGE OF CHURNING

SAVELEY & HENRIOT, CO.

LAND OF TOYS INC.

**AV STORES, CO** 

**ROVELLI GIFTS** 

ONLINE DIECAST CREATIONS CO.

Total Sales 142874 with RFM 144

Total Sales 164069 with RFM 244

Total Sales 157807 with RFM 244

Total Sales 137955 with RFM 244

Total Sales 131685 with RFM 244

## **LOST CUSTOMERS**

DAEDALUS
DESIGNS IMPORTS

OSAKA SOUVENIERS CO.

AUTO ASSOC. & CIE.

CLOVER COLLECTIONS, CO.

ONLINE MINI COLLECTABLES

Total Sales 69052 with RFM 111

Total Sales 67605 with RFM 111

Total Sales 64834 with RFM 111

Total Sales 57756 with RFM 111

Total Sales 57197 with RFM 111

## LOYAL CUSTOMERS

EURO SHOPPING CHANNEL MINI GIFTS
DISTRIBUTORS

HANDJI GIFTS& CO

DIECAST CLASSICS INC

TOKYO COLLECTABLES.

Total Sales 912294 with RFM 444

Total Sales 654858 with RFM 444

Total Sales 115498 with RFM 443

Total Sales 122138 with RFM 434

Total Sales 120562 with RFM 433

#### **RFM Analysis**

## **SUMMARY**

- We classified our top, loyal, on the point of churning, and lost customers using recency, frequency, and monetary parameters. Customers with a recent purchase have been our top customers, as we have lost customer lists.
- Customers on the edge of leaving can be salvaged and transformed into good buyers.
- The RFM model is used to determine customer kinds such as loyal, top or best, on the verge of churning, and lost consumers.
- Recency, frequency, and monetary criteria were commonly utilized to categorize clients.
- This methodology can assist a firm maintain its sales and customers by focusing on how the company lost customers and taking various actions to attract them back.
- It is critical for the company to convert or keep clients who are on the point of leaving into regular customers.
- It is also possible to identify ways to boost the sales ratio.

## **THANK YOU**

Harikrishnan M