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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.

What's in our data

9

 The dataset comprises 2747 entries and 20 columns, representing 89 unique customers. None of the columns have

miccina valuac Non-Null Count Dtype ORDERNUMBER 2747 non-null int64 0 int64 1 QUANTITYORDERED 2747 non-null 2 2747 non-null float64 PRICEEACH 2747 non-null int64 ORDERLINENUMBER float64 2747 non-null 4 SALES datetime64[ns] 5 ORDERDATE 2747 non-null int64 DAYS SINCE LASTORDER 2747 non-null STATUS object 7 2747 non-null object PRODUCTLINE 2747 non-null 8 int64 **MSRP** 2747 non-null object 2747 non-null 10 PRODUCTCODE 2747 non-null object 11 CUSTOMERNAME object 12 PHONE 2747 non-null object 2747 non-null 13 ADDRESSLINE1 14 2747 non-null object CITY object 15 POSTALCODE 2747 non-null 16 2747 non-null object COUNTRY object 17 2747 non-null CONTACTLASTNAME object 2747 non-null 18 CONTACTFIRSTNAME object 2747 non-null DEALSIZE 19

What's in our data(Contd.)



Given data has no duplicates.

df.shape (2747, 20)

Below is the five point summary of numeric columns.

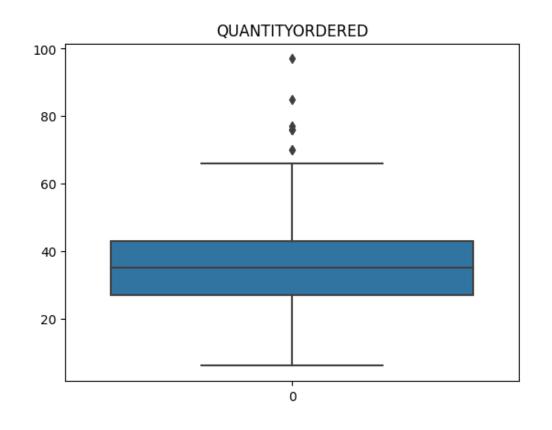
	count	mean	std	min	25%	50%	75%	max
ORDERNUMBER	2747.0	10259.76	91.88	10100.00	10181.00	10264.00	10334.50	10425.00
QUANTITYORDERED	2747.0	35.10	9.76	6.00	27.00	35.00	43.00	97.00
PRICEEACH	2747.0	101.10	42.04	26.88	68.74	95.55	127.10	252.87
ORDERLINENUMBER	2747.0	6.49	4.23	1.00	3.00	6.00	9.00	18.00
SALES	2747.0	3553.05	1838.95	482.13	2204.35	3184.80	4503.09	14082.80
DAYS_SINCE_LASTORDER	2747.0	1757.09	819.28	42.00	1077.00	1761.00	2436.50	3562.00
MSRP	2747.0	100.69	40.11	33.00	68.00	99.00	124.00	214.00

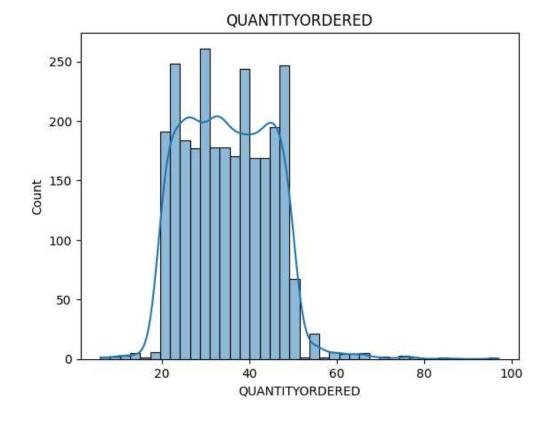
Exploratory Data Analysis



Univariate Analysis(QUANTITYORDERED):

- On an average, 30 items have been ordered.
- Distribution of orders is almost uniform with outliers due to very few high order quantities.

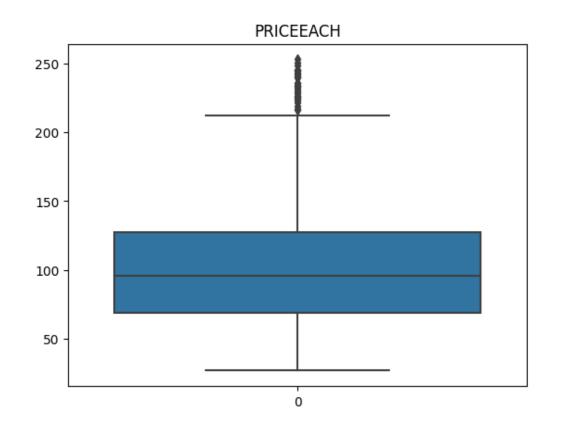


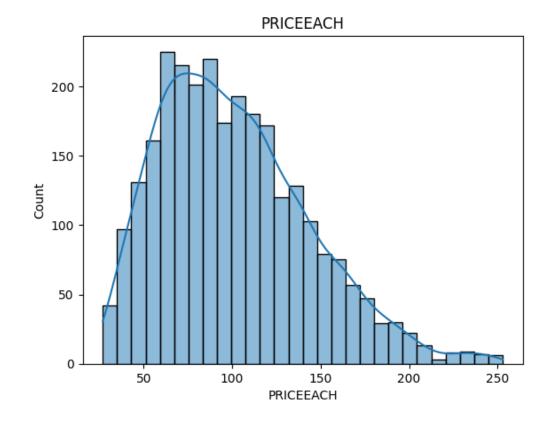




Univariate Analysis(PRICEEACH):

- Average price of all of our products is around 101.
- Distribution of prices is almost normal with outliers due to some premium priced products.

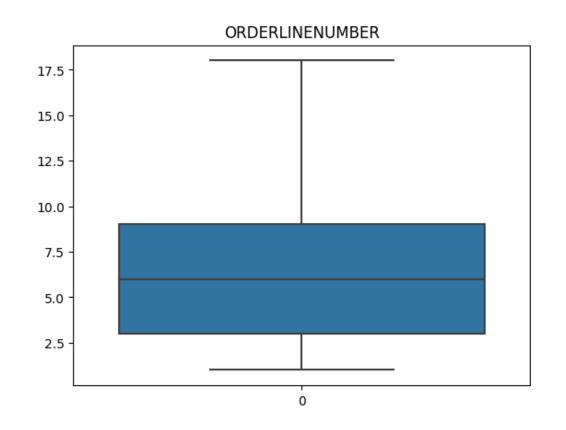


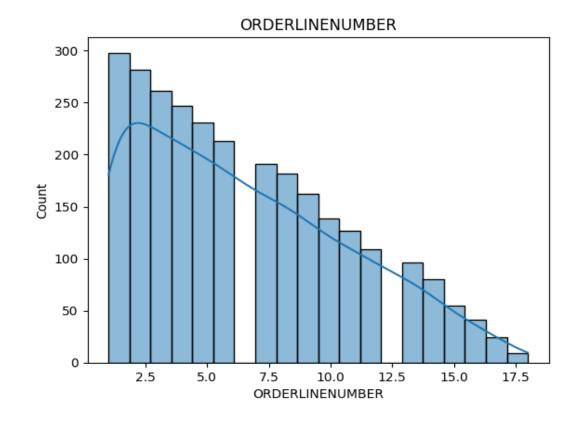




Univariate Analysis(ORDERONLINENUMBER):

It represents the line number of each item within an order.





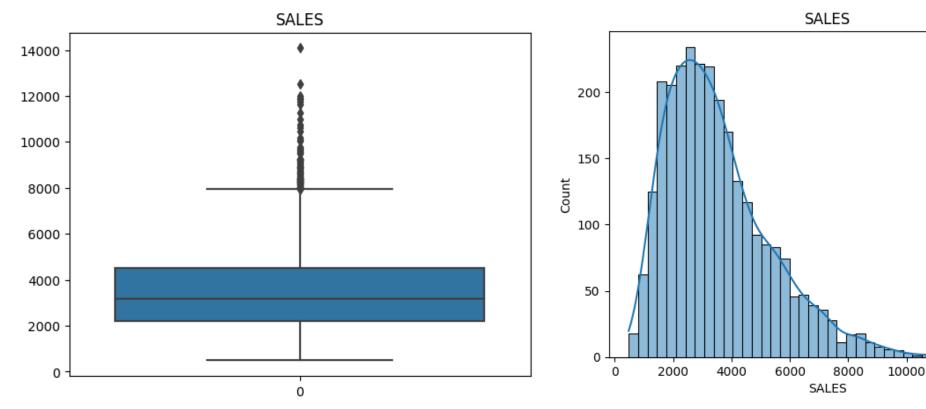


12000

14000

Univariate Analysis(SALES):

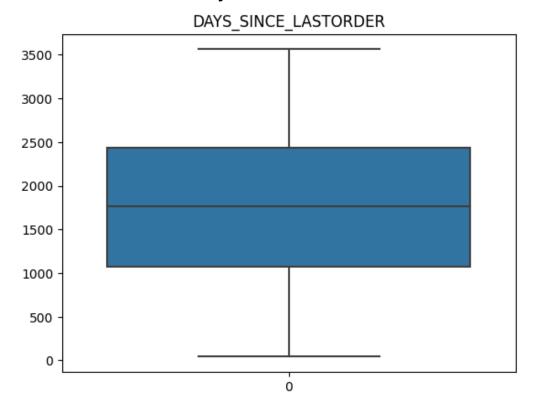
- Majority of our sales amount lies between 2000 and 7000 currency units.
- Yet there are very few sales crossing 8000 currency units. Every sale thats crossing 8000 is observed to be a Large Order.

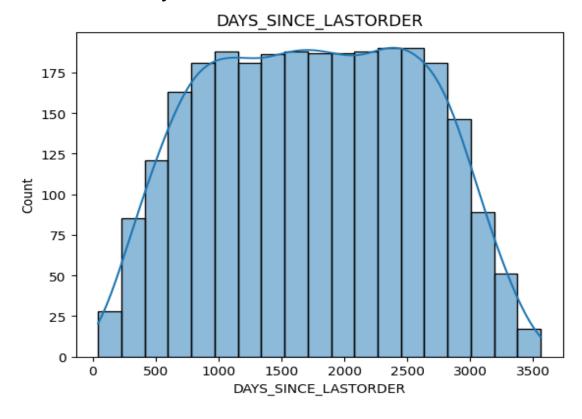




Univariate Analysis(DAYS_SINCE_LASTORDER):

- On an average, customer returns to the store to buy again after almost 1757 days, which is very concerning.
- There are very few customers that shows up before 500 days.

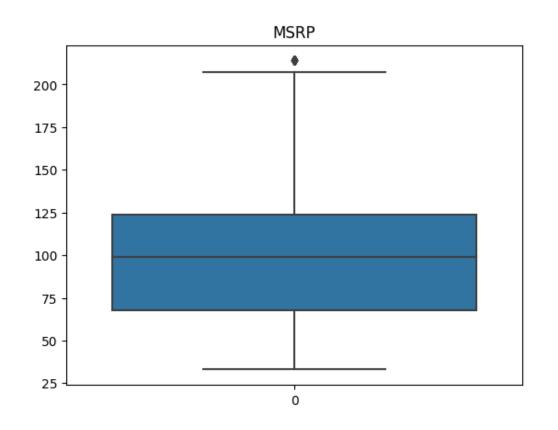


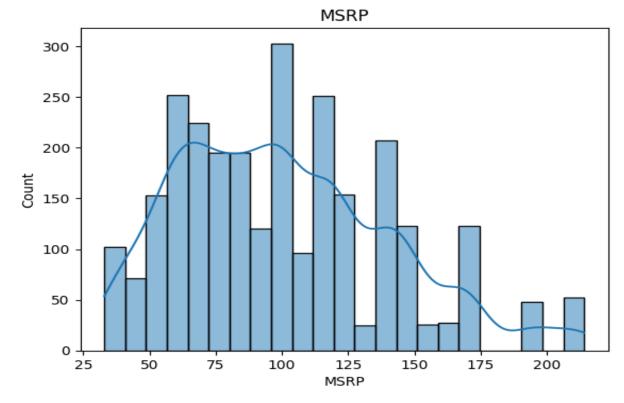




Univariate Analysis(MSRP):

 It stands for Manufacturer's Suggested Retail Price and represents the suggested selling price for each item.

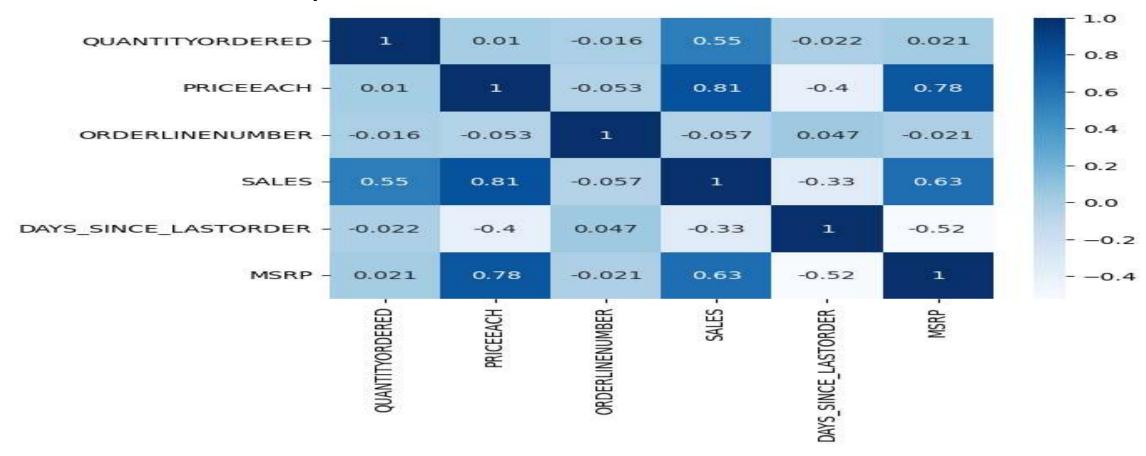






Correlation between above discussed varaibles:

 There is high amount of positive correlation between sales and item's price. None of the other attributes has any useful correlations.



Sales across Countries

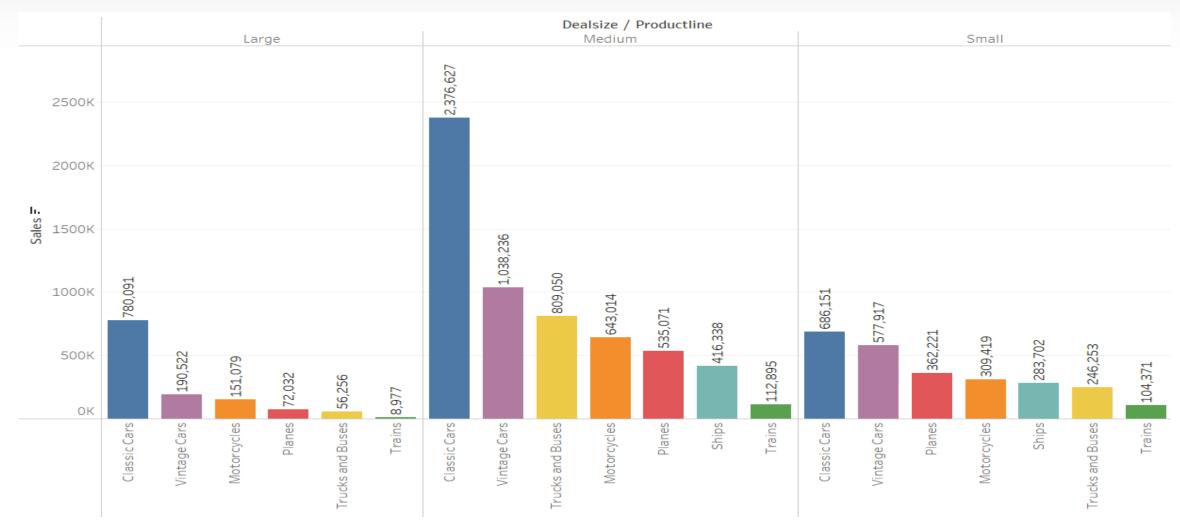


- USA contributed the highest in terms of sales with total of 33,55,576 currency units.
- Ireland
 contributed to the
 least amount of
 sales with 57,756
 currency units.

						, è	
	USA 3,355,576	France 1,110,917		Australia 630,623		UK 478,88	80
9		Italy 374,674	Singapore 288,488		Sweden 210,014		ustria 02,063
3	Spain	Finland 329,582	Denmark 245,637 Canada		Japan 188,168		
	1,215,687	Norway 307,464	224,079 Germany 220,472	Belgium 108,413		Phil 94,0	ippines 016
						Irela	and

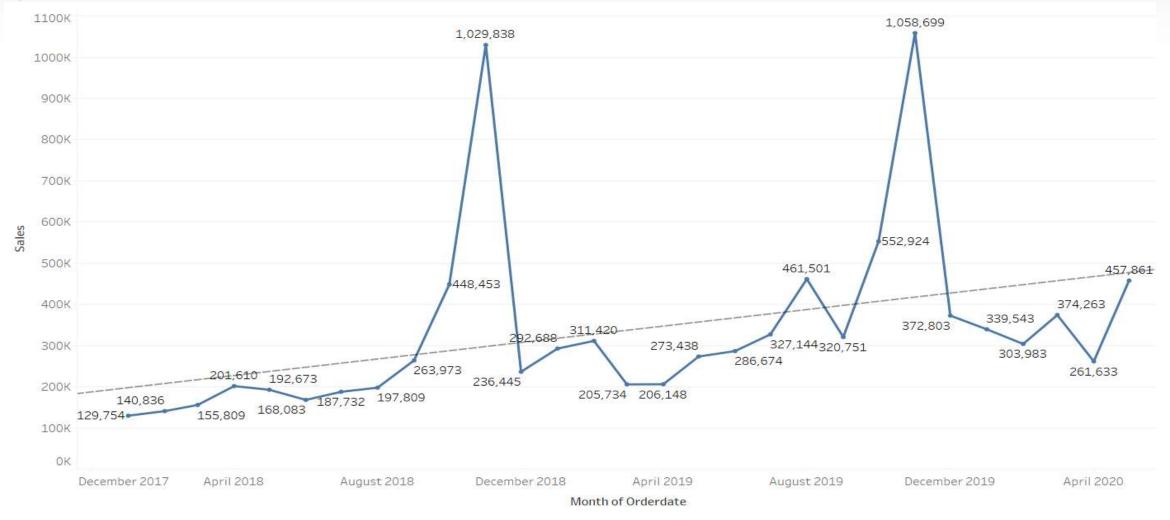
Sales vs Productlines across each deal size.

 Across all the deal sizes, Classic Cars are the highest sold and Trains are the least sold.



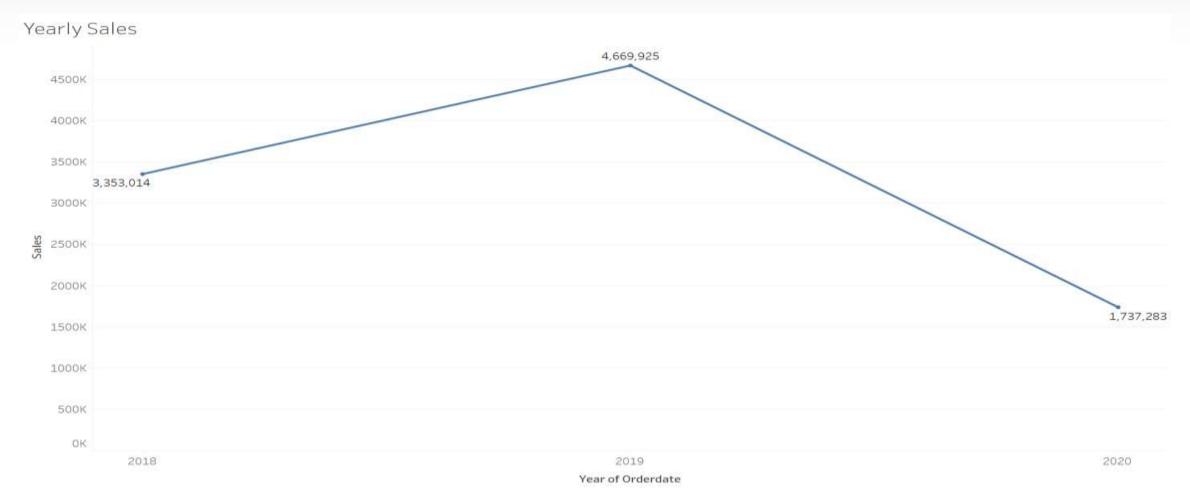
Sales Trend/Seasonality

There is a clear seasonality across years with peak sales in November, trending slighthly upwards.



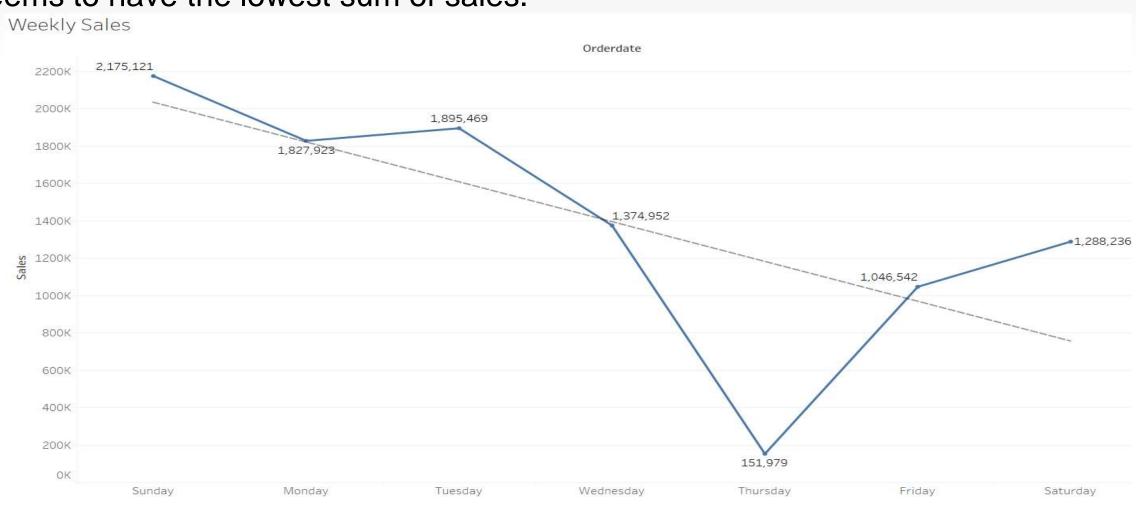
Yearly Sales

There is considerable dip in sales in year 2020, which should be given attention to. 2019 is the year with more sales out of all the three years.



Weekly Sales

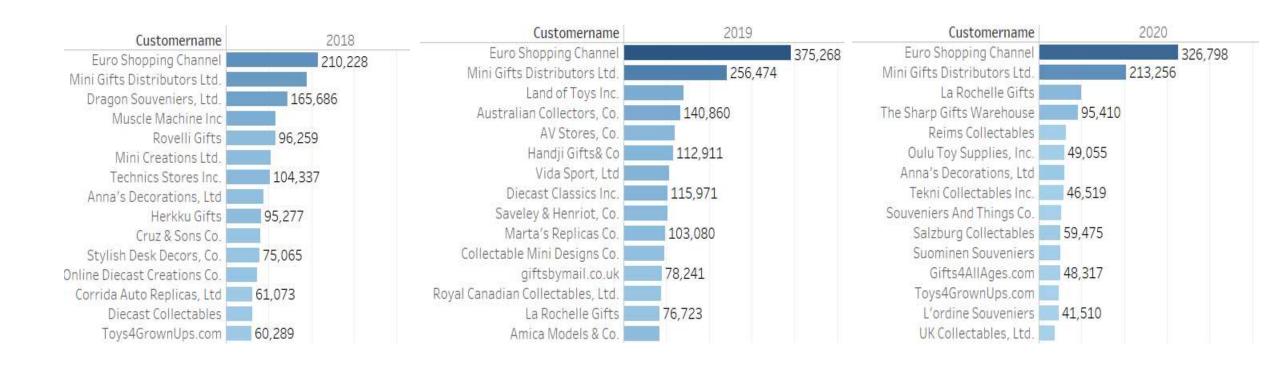
On an average, Sunday seems to have highest number of sales and Thursday seems to have the lowest sum of sales.



Top 15 Customers across years



Euro Shipping and Mini Gifts are the most contributing customer across all three years. Other customer among top 15 across years are as follows.



Highest selling product code

Below is the view of highest selling product code across each product line.

Classic Cars S18_3232	Classic Cars S10_1949	Classi S12_1	c Cars L108	Classic Ca S18_2238	Car	ssic 's 2_3891	Cars C	/intage cars i18_2795	Vintage Cars S18_31	Ca	rs	Vintage Cars S18_3856	Motorcycles S10_4698	Motorcycles S12_2823		ies _1662	Plan	nes
Classic Cars	Classic Clas		Classic	Classic		lassic	Vintage Cars S18_1342	Vintage S18_33		Vintage Cars	Vintage Cars				Plan	ies Pl	anes I	Plane
S24_3856	Cars Cars S10_4962 S18	1984	Cars S18_34	Cars 32 S18_4		ars	Vintage Cars S18_2949	Vintage S18_45	22	Vintage Cars	Vintage Cars	e			Plan	nes Pla	nes P	Plane
Classic Cars S18_4027							Vintage Cars S24_4258	Vintage S18_31	Cars 36						5 1511		.,	
-1	Classic Cars S18_1589	Class		assic Clars Ca		Classic Cars	-			Vintage Cars								
Classic Cars S18_1129	Classic Cars						Vintage Cars S24_3151	Vintage S18_46	68	Vintage Cars			Motorcycles S24_2000		Plar S24	es _3949		
Classic Cars 512_1099	S10_4757		sic Cars 1889				Trucks and Buse S12_1666		ks and Bi _2319	i	Frucks and	Trucks and	Motorcycles S24_2360	Motorcycles	Plan S18	es _2581		
	Classic Cars S700_2824		sic Cars				-				Buses 518_1097	Buses	Ships S24_2011	Ships	Ships	Ships	Trains S18_32	
Classic Cars S18_3685	Classic Cars	100000000000000000000000000000000000000	4620	Classic	arc	Classic	Trucks and Buse S24_2300		ks and Bu _4473	uses			2					
	S24_4048		sic Cars 3278	S24_33	71	Cars					Frucks and Buses		Ships S700_3505	China		China	Trains	
Classic Cars S12_3148	Classic Cars			Classic (S24_10	16	Classic	Trucks and Buse S18 4600	200	ks and Bu _1392	uses	32_3522			Ships S18_3029		Ships		
16_3140	S12_4675		sic Cars 3191	Classic	ars	Cars	520_4000	550	_1392		Frucks and Buses		Ships S700_3962	Ships	Ships			

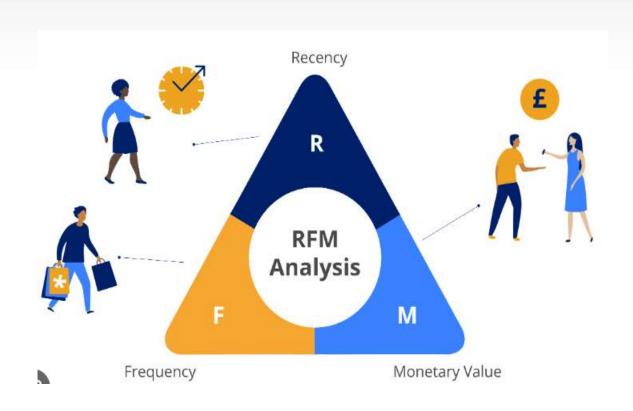
Customer Segmentation using RFM Model



- RFM stands for
 - Recency(days since last ordered/bought)
 - Frequency(number of times same customer is contributing towards sales.)
 - Monetary(Quantity of each item*Price of each item)
- RFM analysis is a great technique to identify set of customers for distinctive treatments.
- RFM segmentation allows marketers to target specific groups of customers.
- With promotions that are more significant for their actual behaviour, we can better help our individual customer.

Customer Segmentation using RFM Model





- Customer segmentation has been done into four categories(High Value, Medium Value, Low Value and Very Low Value)
- Below aggregations were used while segmenting customers.
 - ✓ Recency Days since last ordered.
 - ✓ Frequency Unique count of OrderID.
 - ✓ Monetary Total Sum of sales for each customer.

Inferences from RFM(Golden Customers)

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Below are our best customers with high monetary values, highly frequent buying and more active(recent) buying style.

CUSTOMERNAME	RFM_Score
The Sharp Gifts Warehouse	444
Technics Stores Inc.	444
Souveniers And Things Co.	444
Salzburg Collectables	444
Mini Gifts Distributors Ltd.	444
Land of Toys Inc.	444
La Rochelle Gifts	444
Euro Shopping Channel	444
Diecast Classics Inc.	444
Australian Collectors, Co.	444
Anna's Decorations, Ltd	444

Inferences from RFM(Medium Value Customers)

Below are our customers with average recency, average frequent and medium monetary value. We can entice these customers to convert into golden customers.

CUSTOMERNAME	RFM_Score
Baane Mini Imports	443
Online Diecast Creations Co.	424
UK Collectables, Ltd.	423
Oulu Toy Supplies, Inc.	423
Gift Depot Inc.	423
FunGiftIdeas.com	423
Auto Canal Petit	423
Quebec Home Shopping Network	422
Lyon Souveniers	422
Gifts4AllAges.com	422

Inferences from RFM(Low Value Customers)

Below are our customers that are at the verge of churning.

CUSTOMERNAME	RFM_Score
Amica Models & Co.	313
Atelier graphique	321
Tekni Collectables Inc.	322
Classic Legends Inc.	322
Toys of Finland, Co.	323
Suominen Souveniers	323
La Corne D'abondance, Co.	323
Scandinavian Gift Ideas	324
Corrida Auto Replicas, Ltd	324
Volvo Model Replicas, Co	342
Tokyo Collectables, Ltd	343
Reims Collectables	344

Inferences from RFM(Very low value Customers)

These are customers that can be considered lost as they have very less frequent and less monetary values and not very active customers.

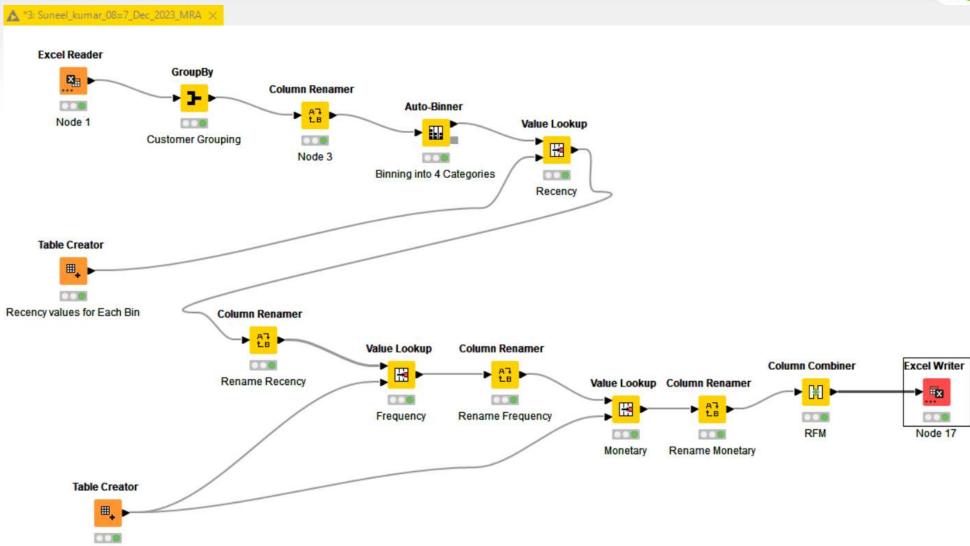
CUSTOMERNAME	RFM_Score
Signal Collectibles Ltd.	111
Iberia Gift Imports, Corp.	111
Double Decker Gift Stores, Ltd	111
Clover Collections, Co.	111
Bavarian Collectables Imports, Co.	111
Norway Gifts By Mail, Co.	112
Diecast Collectables	112
Mini Auto Werke	121
Gift Ideas Corp.	121
Auto-Moto Classics Inc.	121

Note: Customers mentioned in each category are just top 10 each category. Please refer attached excel for total list of customers.

Knime Workflow

Frequency & Monetary Values for Each Bin







Recommendations:

- Golden customers are the ones because of whom, the company is surviving. We should retain them at any cost.
- Medium customers are the ones who can be enticed with offers and services and make them our golden customers.
- Low value customers are very hard to retain but we can reduce the convertion to Lost customers by giving basic offers only after planned expenditure is made on golden and Medium value customers.
- Very Low value customers are lost customers and we dont need to do anything to retain them.



Thank You