

# Online Retail Customer Segmentation



#### Content

PROBLEM STATEMENT

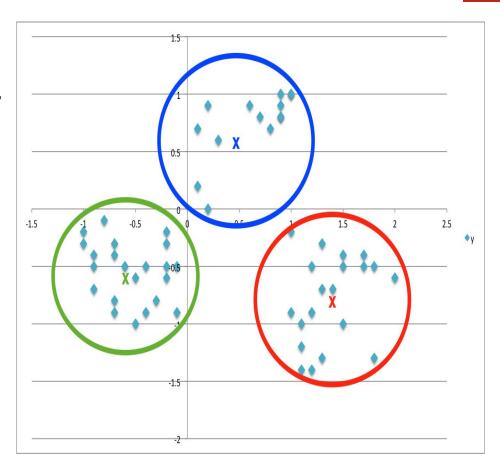
DATA SUMMARY

**ANALYSIS** 

CHALLENGES

CONCLUSION

Q & A





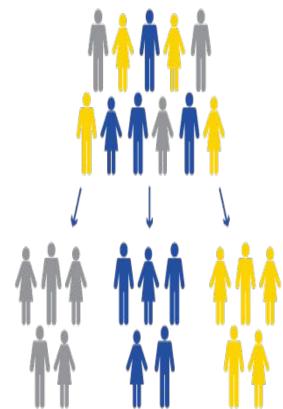
#### **Problem Statement**

Given a dataset related to a online retailer based out of the UK, we need to analyse and identify major customer segments using K Means algorithm and also using different verification method to confirm the result.



### What is Customer Segmentation?

- Practice of dividing a customer base into groups of individuals that are similar in specific ways relevant to marketing, such as age, gender, interests and spending habits.
- Allows us to better understand our customers helping us target these customers in a more efficient manner and improve the customer experience.





### **Data Summary**

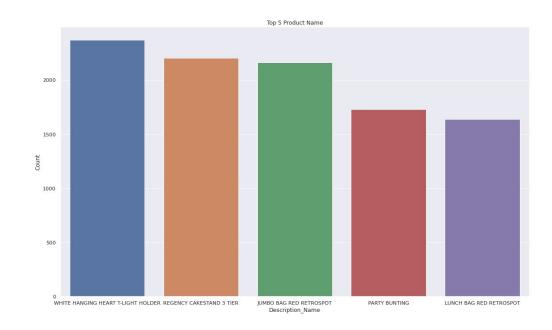
- A transnational data set with transactions occurring between 1st December 2010 and 9th
   December 2011 for a UK-based online retailer.
- The company mainly sells unique all-occasion gifts.
- Many customers of the company are wholesalers.

InvoiceNo	StockCode	Description	Quantity	InvoiceDate	UnitPrice	CustomerID	Country
536365	85123A	WHITE HANGING HEART T-LIGHT HOLDER	6	2010-12-01 08:26:00	2.55	17850.0	United Kingdom
536365	71053	WHITE METAL LANTERN	6	2010-12-01 08:26:00	3.39	17850.0	United Kingdom
536365	84406B	CREAM CUPID HEARTS COAT HANGER	8	2010-12-01 08:26:00	2.75	17850.0	United Kingdom
536365	84029G	KNITTED UNION FLAG HOT WATER BOTTLE	6	2010-12-01 08:26:00	3.39	17850.0	United Kingdom
536365	84029E	RED WOOLLY HOTTIE WHITE HEART.	6	2010-12-01 08:26:00	3.39	17850.0	United Kingdom



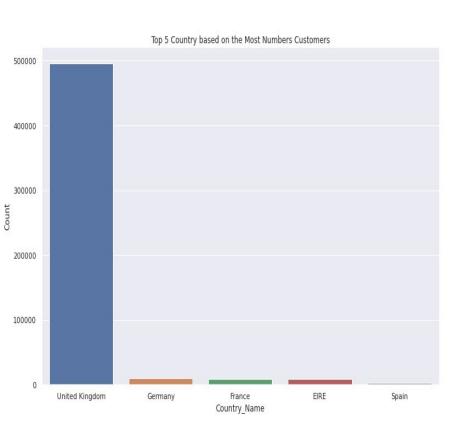
## Cleaning Data Finding the most Purchased Products

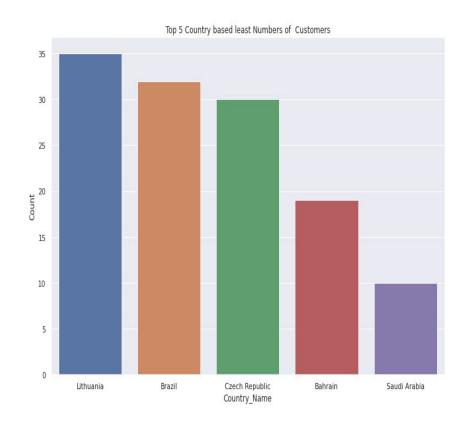
Description_Name	Count
WHITE HANGING HEART T-LIGHT HOLDER	2369
REGENCY CAKESTAND 3 TIER	2200
JUMBO BAG RED RETROSPOT	2159
PARTY BUNTING	1727
LUNCH BAG RED RETROSPOT	1638





### **Top 5 vs Bottom 5 countries**

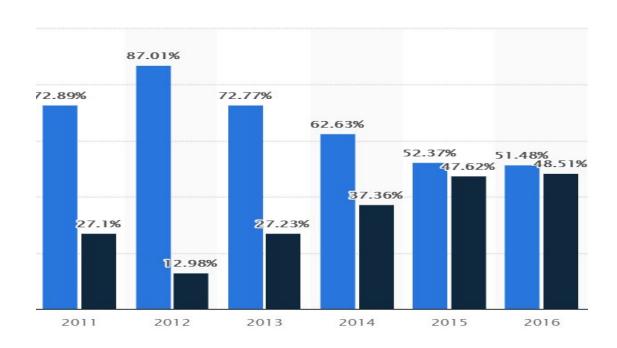






### **Analysis**

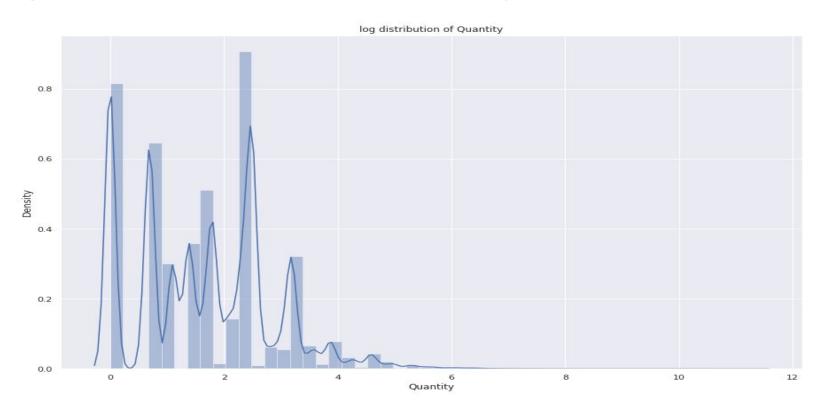
#### UK Saudi



Source obtained from Statista comparing online purchases from 2011 to 2016

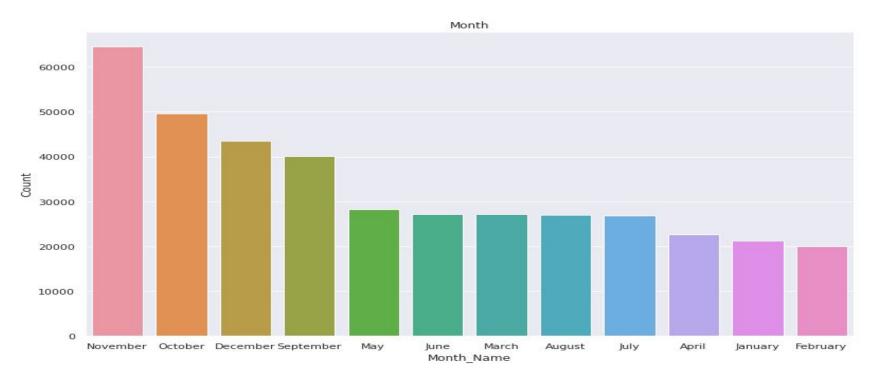


### **Log Distribution of Quantity**





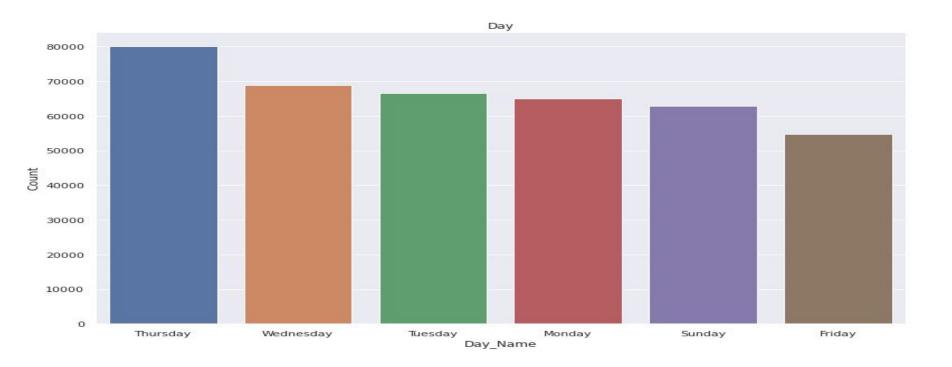
### **Month-wise analysis**



November and December could be the months with highest sales in anticipation of Christmas

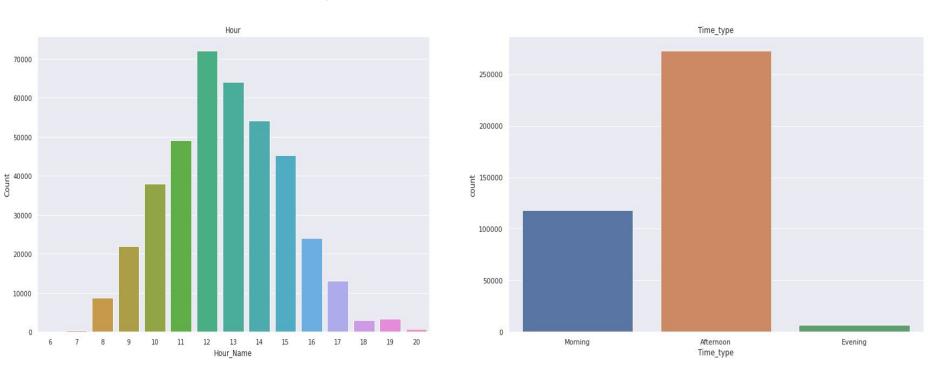


### **Daywise analysis**





### **Hourwise analysis**



Working hours witnessing the highest sales could be attributed to the fact that a large part of the dataset is Wholesalers' data



### Recency, Frequency, Monetary values

#### **RFM Metrics**



The freshness of the customer activity, be it purchases or visits

RECENCY

E.g. Time since last order or last engaged with the product



#### **FREQUENCY**

The frequency of the customer transactions or visits

E.g. Total number of transactions or average time between transactions/ engaged visits



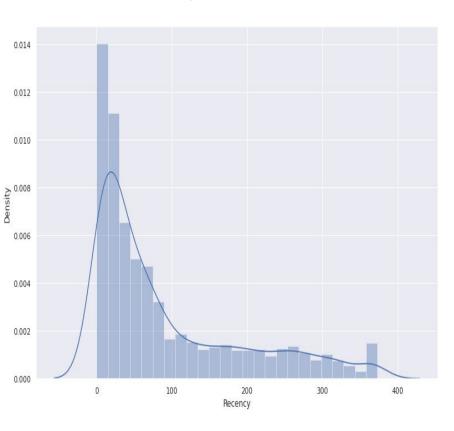
#### **MONETARY**

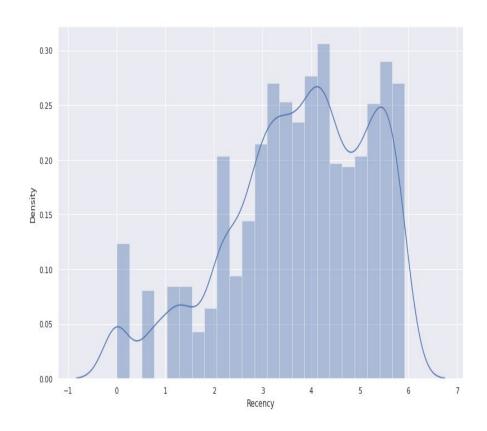
The intention of customer to spend or purchasing power of customer

E.g. Total or average transactions value



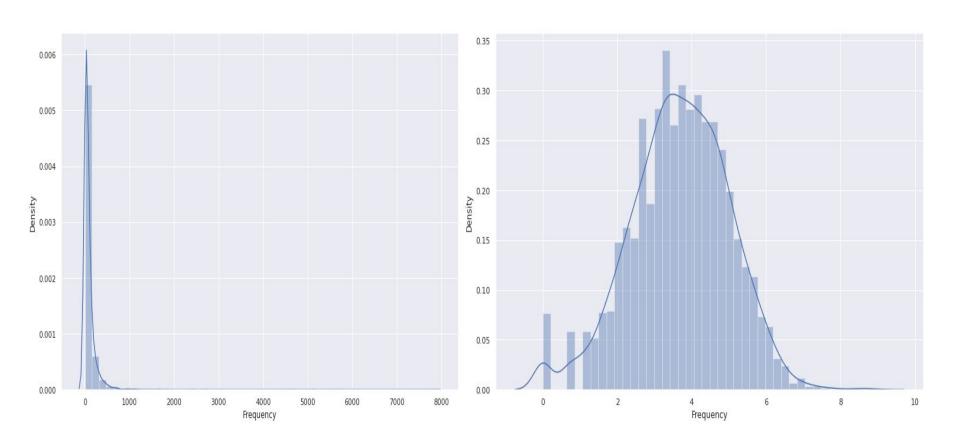
### **Recency**





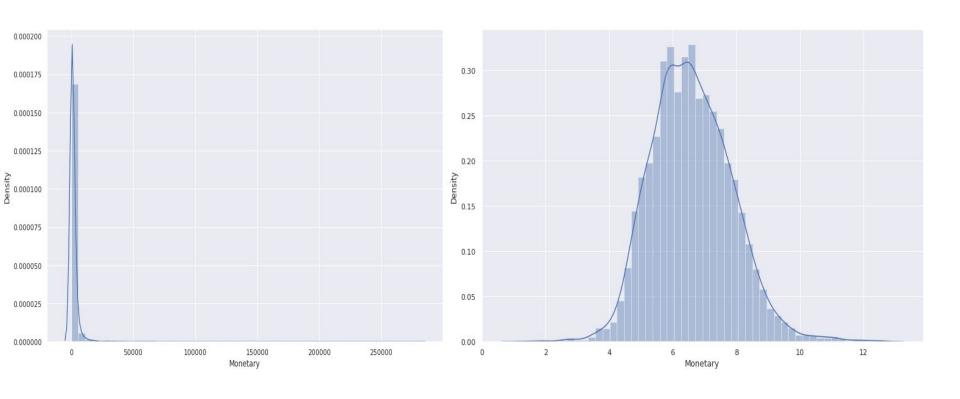


### **Frequency**





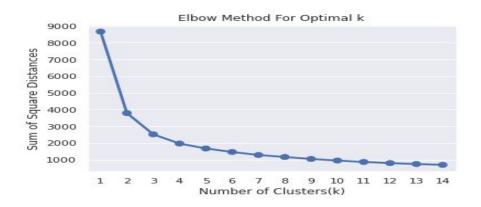
### **Monetary**

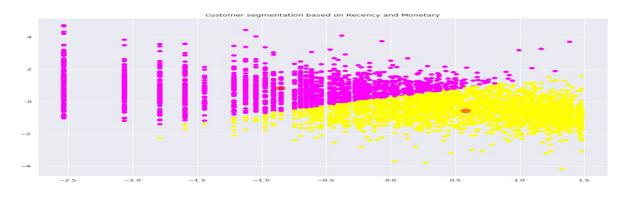




#### Silhouette score and Elbow method on R&M

```
For n_clusters = 2, silhouette score is 0.4216081125935063
For n_clusters = 3, silhouette score is 0.3432957775914936
For n_clusters = 4, silhouette score is 0.36494104664274657
For n_clusters = 5, silhouette score is 0.33668503688485785
For n_clusters = 6, silhouette score is 0.34397809419193187
For n_clusters = 7, silhouette score is 0.345867202377316
For n_clusters = 8, silhouette score is 0.33919727934627264
For n_clusters = 9, silhouette score is 0.3458423886312394
For n_clusters = 10, silhouette score is 0.34850666375861195
For n_clusters = 11, silhouette score is 0.3385166366909024
For n_clusters = 12, silhouette score is 0.3427649471441594
For n_clusters = 13, silhouette score is 0.34083950250492523
For n_clusters = 14, silhouette score is 0.3406096956008792
For n_clusters = 15, silhouette score is 0.34223526314989594
```

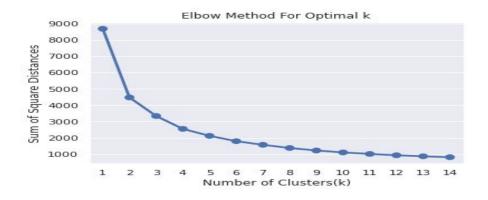


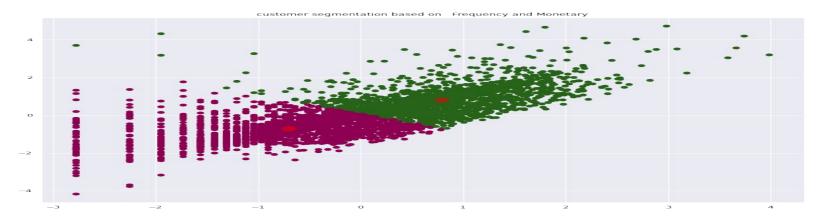




### Silhouette score and Elbow method on F&M

For n\_clusters = 2, silhouette score is 0.478535709506603
For n\_clusters = 3, silhouette score is 0.40764120562174455
For n\_clusters = 4, silhouette score is 0.3713782596510203
For n\_clusters = 5, silhouette score is 0.34479733808079405
For n\_clusters = 6, silhouette score is 0.35974563779013946
For n\_clusters = 7, silhouette score is 0.33835032540639154
For n\_clusters = 8, silhouette score is 0.3519892091800133
For n\_clusters = 9, silhouette score is 0.3460160650521864
For n\_clusters = 10, silhouette score is 0.3619887930235607
For n\_clusters = 11, silhouette score is 0.36822618560766546
For n\_clusters = 12, silhouette score is 0.3640489785135785
For n\_clusters = 13, silhouette score is 0.3624674157300161
For n\_clusters = 14, silhouette score is 0.36520616987776316
For n\_clusters = 15, silhouette score is 0.36101570873847355





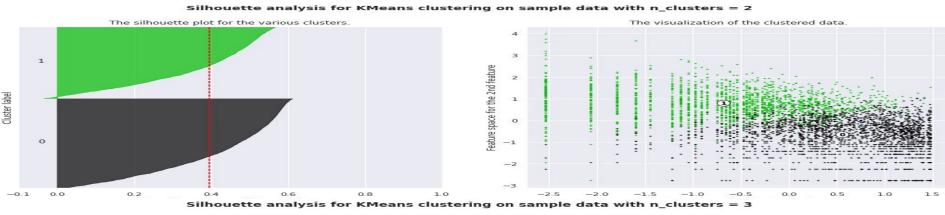


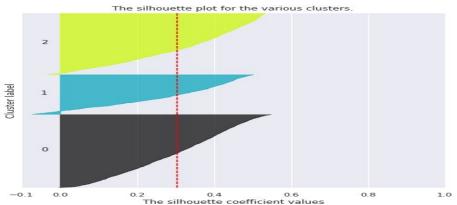
### Silhouette analysis on R, F and M

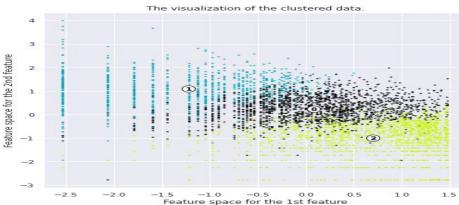
```
For n clusters = 2 The average silhouette score is : 0.395423791756615
For n clusters = 3 The average silhouette score is : 0.3031065868149085
For n clusters = 4 The average silhouette score is : 0.30272551749681986
For n clusters = 5 The average silhouette score is : 0.2788034616608947
For n clusters = 6 The average silhouette score is : 0.27854318607070516
For n clusters = 7 The average silhouette score is : 0.2623613650755882
For n clusters = 8 The average silhouette score is : 0.2638608672365028
For n clusters = 9 The average silhouette score is : 0.25878886517568883
For n_clusters = 10 The average silhouette_score is : 0.25947712786853405
For n clusters = 11 The average silhouette score is : 0.2594602425001122
For n clusters = 12 The average silhouette score is : 0.26359981003963245
For n clusters = 13 The average silhouette score is : 0.26216905448550776
For n clusters = 14 The average silhouette score is : 0.2610200890360579
For n clusters = 15 The average silhouette score is : 0.2549657732066674
```



### Silhouette analysis on RFM

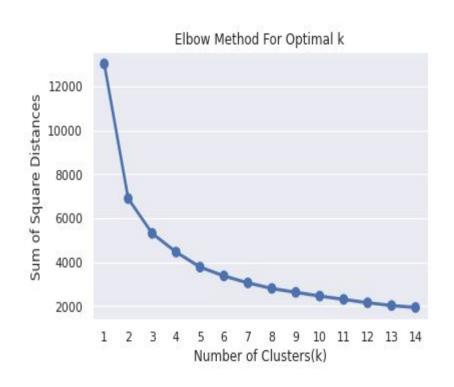


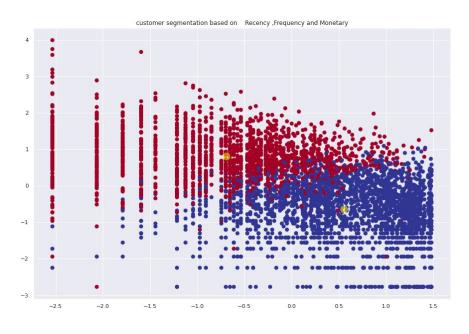






#### **Elbow method and Cluster chart on RFM**





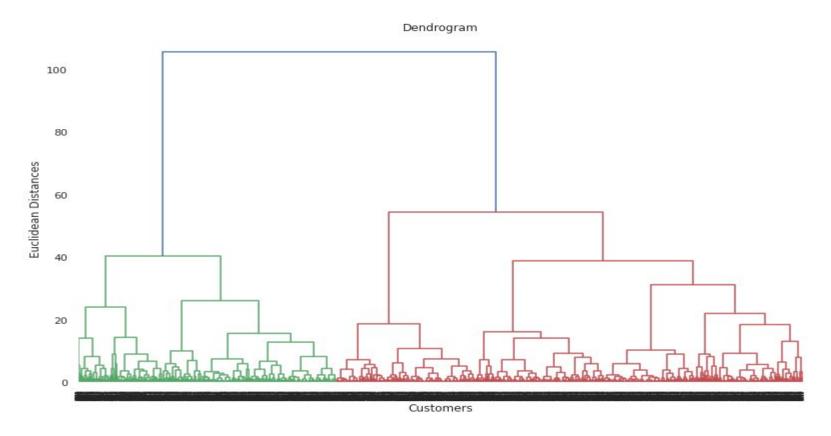


### **RFM Analysis**

	Recency	Frequency	Monetary	R	F	M	RFMGroup	RFMScore	Recency_log	Frequency_log	Monetary_log	Cluster
CustomerID												
12346.0	325	1	77183.60	4	4	1	441	9	5.783825	0.000000	11.253942	1
12347.0	2	182	4310.00	1	1	1	111	3	0.693147	5.204007	8.368693	0
12348.0	75	31	1797.24	3	3	1	331	7	4.317488	3.433987	7.494007	1
12349.0	18	73	1757.55	2	2	1	221	5	2.890372	4.290459	7.471676	0
12350.0	310	17	334.40	4	4	3	443	11	5.736572	2.833213	5.812338	1
12352.0	36	85	2506.04	2	2	1	221	5	3.583519	4.442651	7.826459	0
12353.0	204	4	89.00	4	4	4	444	12	5.318120	1.386294	4.488636	1
12354.0	232	58	1079.40	4	2	2	422	8	5.446737	4.060443	6.984161	1
12355.0	214	13	459.40	4	4	3	443	11	5.365976	2.564949	6.129921	1
12356.0	22	59	2811.43	2	2	1	221	5	3.091042	4.077537	7.941449	0



### Dendrogram





### **DBSCAN**





### Challenges

- Tackling refunds
- Right number of 'k' for clusters



### Conclusion

Model Name	Data	Optimal Number of Clusters			
K-Means with Silhouette Score	RM	2			
K-Means with Elbow method	RM	2			
DBSCAN	RM	2			
K-Means with Silhouette Score	FM	2			
K-Means with Elbow method	FM	2			
DBSCAN	FM	2			
K-Means with Silhouette Score	RFM	2			
K-Means with Elbow method	RFM	2			
Hierarchical Clustering	RFM	2			
DBSCAN	RFM	3			