

PROJECT REPORT*

CUSTOMER SEGMENTATION

GOKAY BULUT



**Note: Since the company owns all intellectual property rights, this is not the original presentation provided to the company. Rather, it is a report explaining my work.*

AGENDA

- Introduction
- Data Wrangling & EDA
- Cohort Analysis
- RFMT Analysis
- K-Means Clustering
- Conclusion



Introduction

Problem:

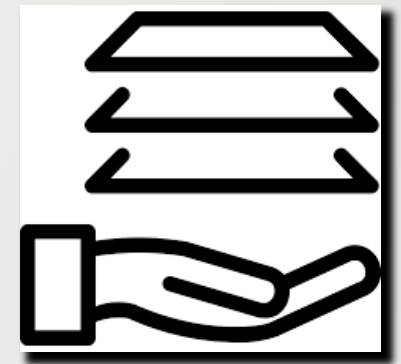
- group customers into segments
- to understand high level trends better
- by providing insights on metrics across product / service and customer lifecycle.



Introduction

Data set:

- 98,572 rows of customer transactions,
- provided by the company.



Introduction

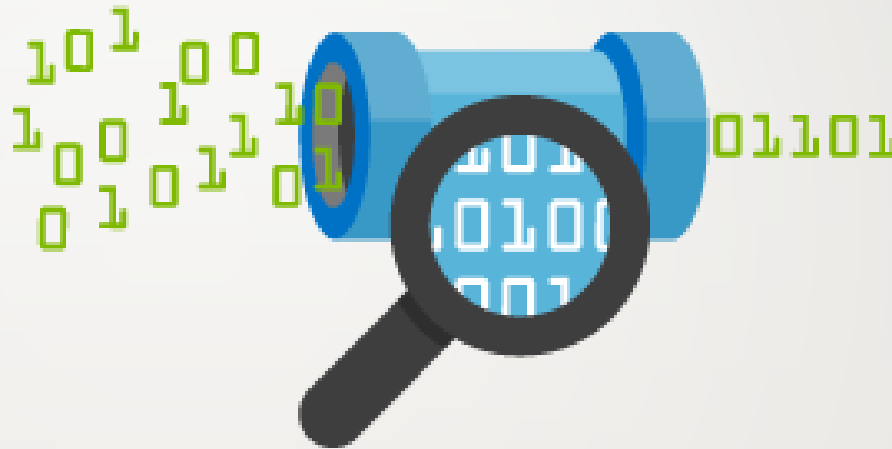


Features:

- 'Inv_No' - number for each transaction (integer).
- 'Inv_Date' - time of the transaction (string).
- 'name' - name of the customer (string).
- 'lastname' - lastname of the customer (string).
- 'Cust_ID' - unique number for identifying the customers (integer).
- 'Photo_Type' - types of photos taken (string).
- 'Amount' - amount paid by the customer for the transaction (float).
- 'Notes' - notes on the transaction (string).

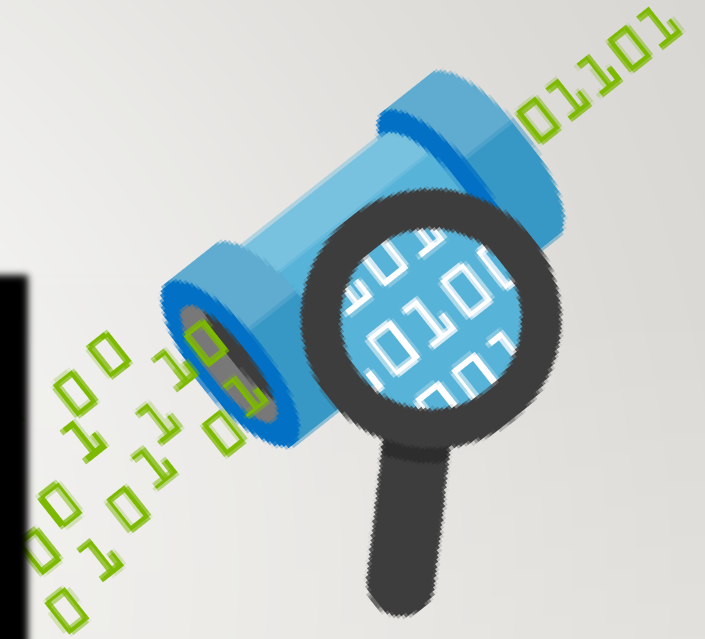
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Data Wrangling & EDA

| | Inv_No | Inv_Date | Cust_ID | Photo_Type | Amount | Notes |
|---|--------|------------|---------|------------|--------|------------------------------|
| 0 | 43891 | 01/01/2016 | 16106 | amatör | 29.78 | Co k acele |
| 1 | 43892 | 01/01/2016 | 10570 | pasaport | 32.94 | Acele, bir an once yapilamli |
| 2 | 43893 | 01/01/2016 | 13796 | vesikalik. | 29.45 | Bizim |
| 3 | 43894 | 01/01/2016 | 10246 | Okul | 23.94 | Liste -- oncelikli |
| 4 | 43895 | 01/01/2016 | 5158 | pasaport | 23.58 | Tanidik |

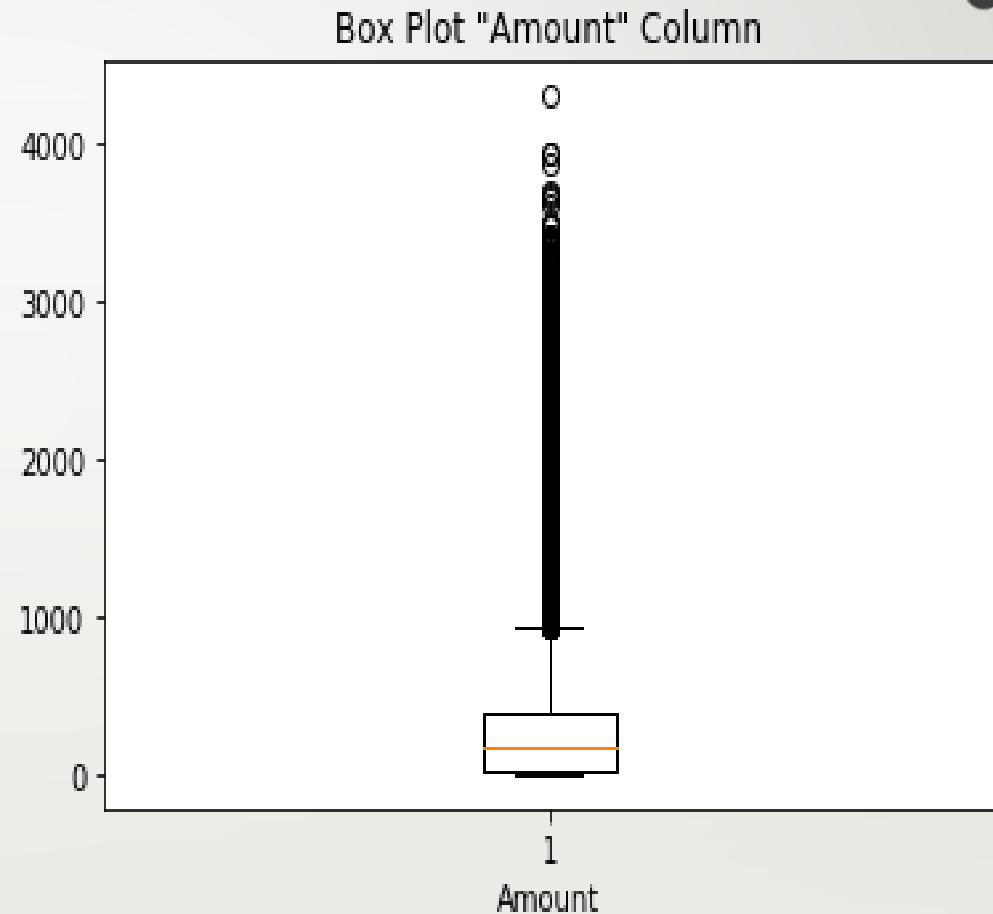


- Dropped redundant columns & rows,
- 'Photo_Type' column cleaned (lowercased & stripped dots),
- 'Inv_Date' column → datetime type for better analysis,

Data Wrangling & EDA



- Sales 'Amount' mean = 331.2.
- Returning customers.
- Invoice numbers are unique.
- Invoice dates cover 1 year.
- 7 unique photo types.



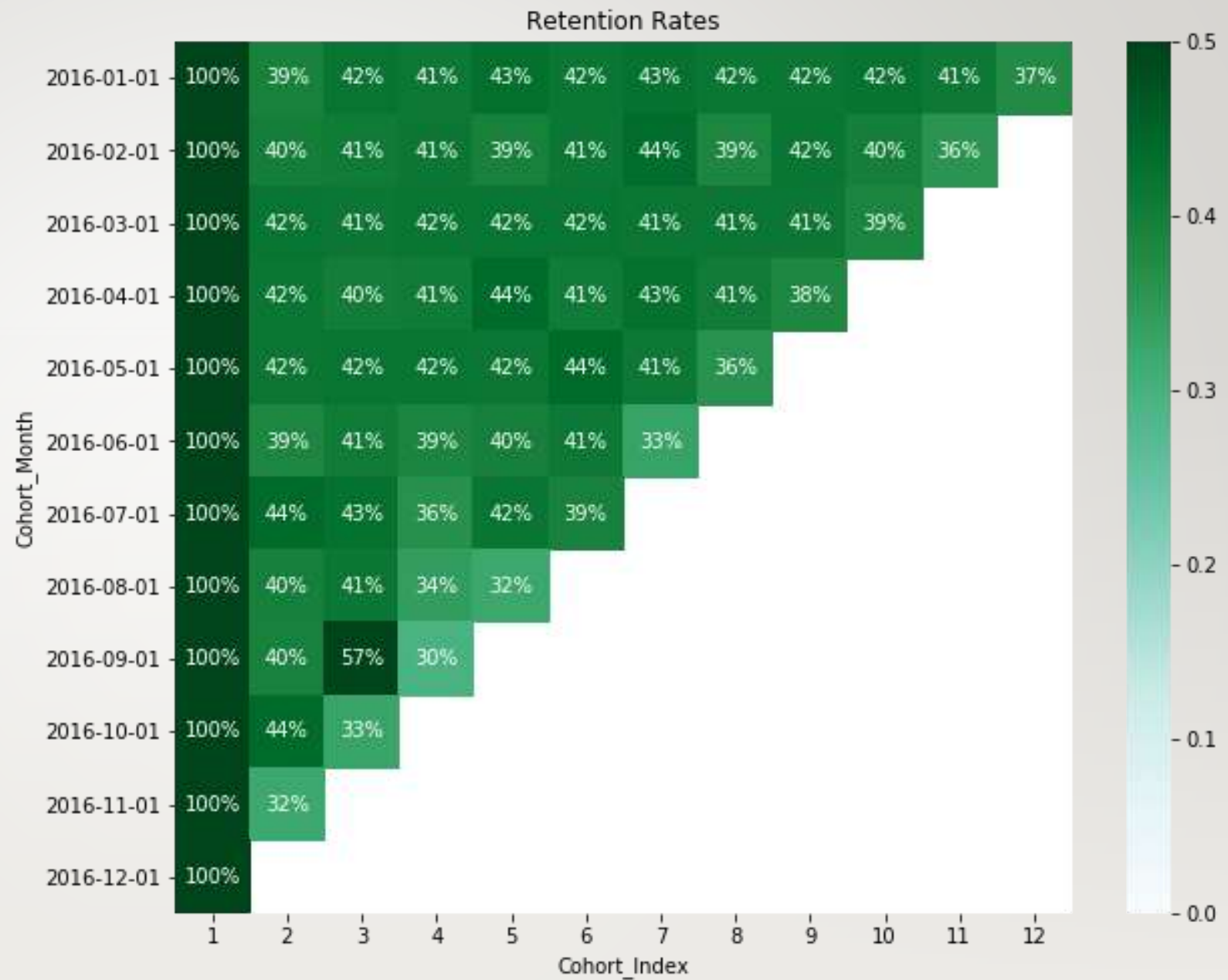
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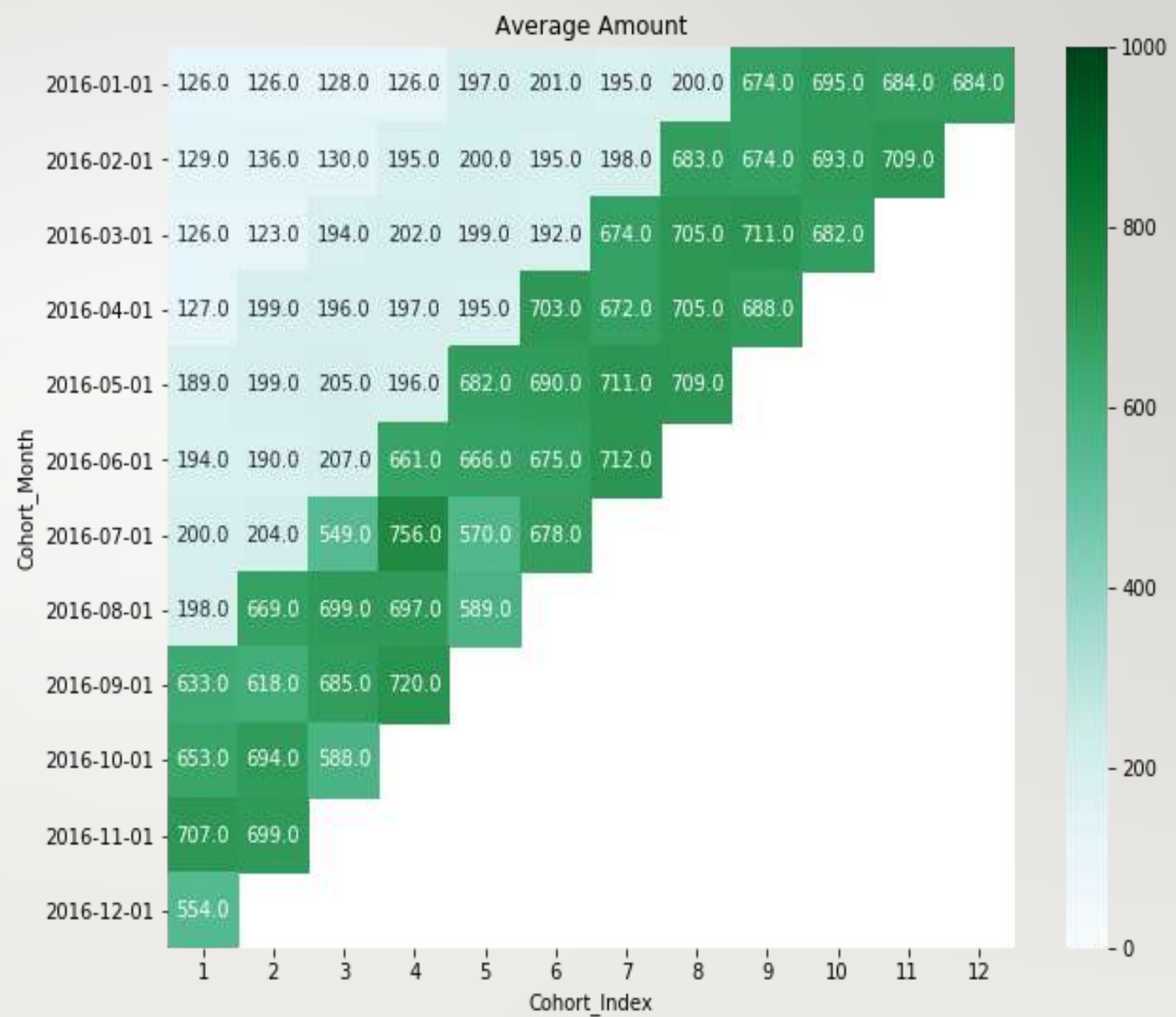
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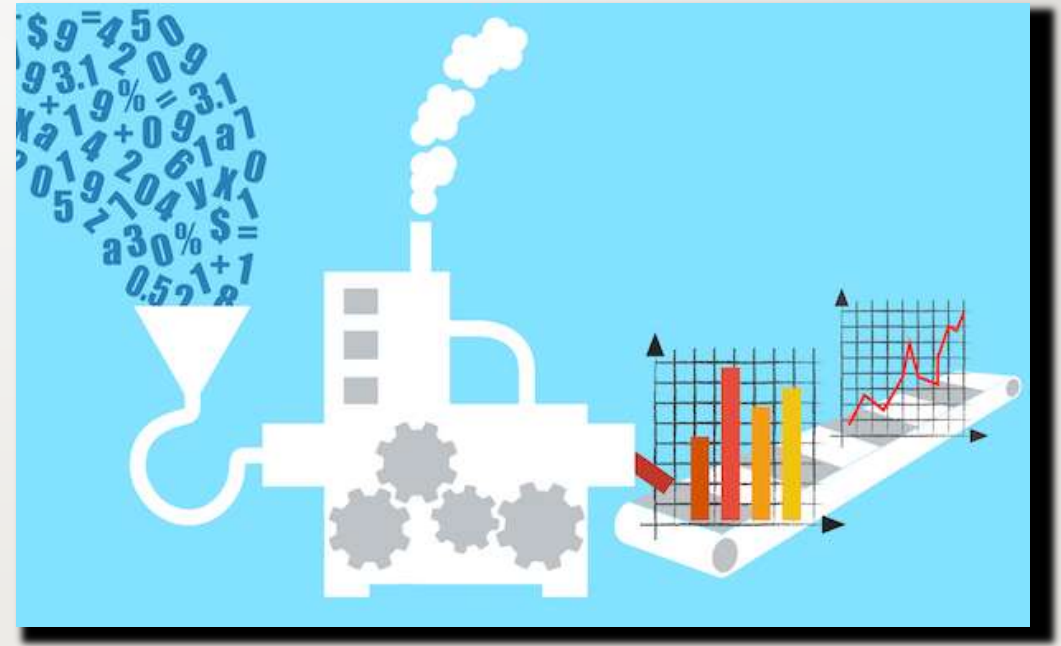
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RFMT Analysis

- Recency → how recent
- Frequency → how many times
- Monetary Value → how much
- Tenure → for how long

| _Segment | Recency | Frequency | Monetary_Value | | Tenure |
|----------|---------|-----------|----------------|-------|--------|
| | mean | mean | mean | count | mean |
| 1.Gold | 25.7 | 8.6 | 3177.3 | 5197 | 344.8 |
| 2.Silver | 74.3 | 5.0 | 1501.9 | 9207 | 302.3 |

| _Segment | Recency | Frequency | Monetary_Value | | Tenure |
|----------|---------|-----------|----------------|-------|--------|
| | mean | mean | mean | count | mean |
| 1.Gold | 25.7 | 8.6 | 3177.3 | 5197 | 344.8 |
| 2.Silver | 48.9 | 6.1 | 1985.1 | 4494 | 321.2 |
| 3.Bronze | 98.5 | 4.0 | 1041.2 | 4713 | 284.3 |

- segment 1 remained in both.
- segment 2 → divided in 2

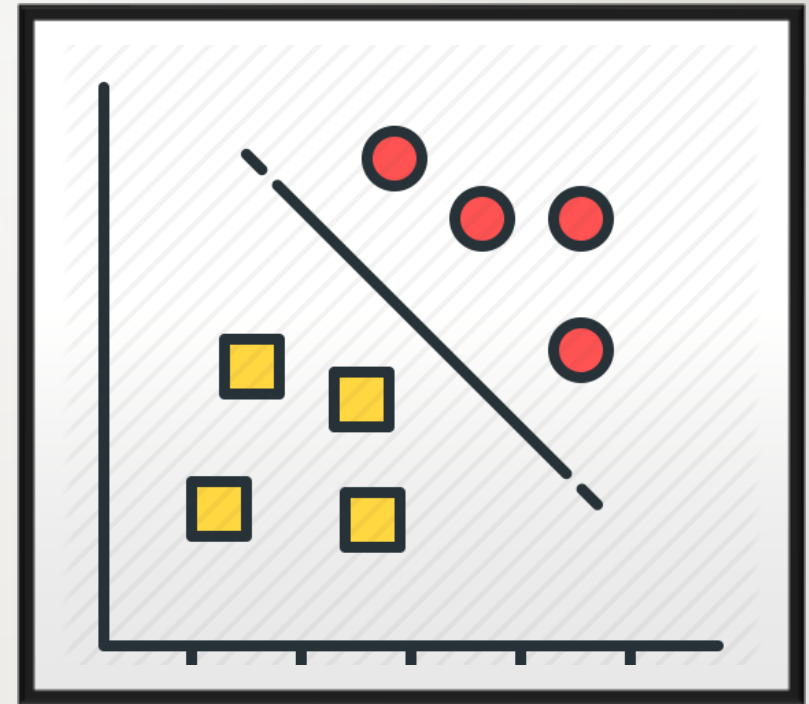
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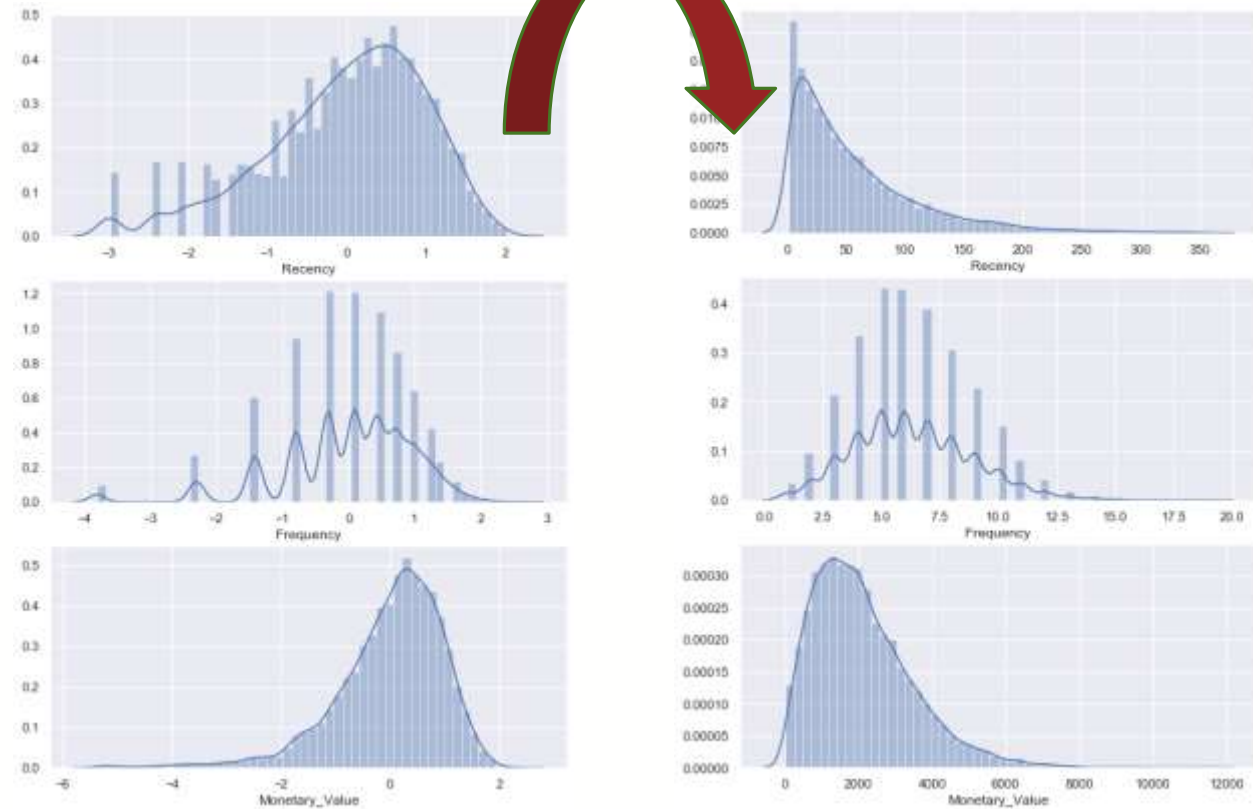
K-Means Clustering

- Partitions n observations into k clusters,
- Each observation belongs to cluster with nearest mean.
- Assumptions:
 - variables symmetrically distributed,
 - have the same mean and variance.



K-Means Clustering

- Log transformation to unskew the data
- Standardized to same mean
- Scaled to same std



K-Means Clustering

- k = Elbow point (where decrease in SSE slows down) & next point



K-Means Clustering

- k=2, clusters more distinct, cluster 1 = 2x cluster 0 in size.

| | Recency | Frequency | Monetary_Value | |
|---------|---------|-----------|----------------|-------|
| | mean | mean | mean | count |
| Cluster | | | | |
| 0 | 101.0 | 4.0 | 952.0 | 4994 |
| 1 | 34.0 | 8.0 | 2719.0 | 9410 |

- k=3, clusters 0 and 2 F & M close.

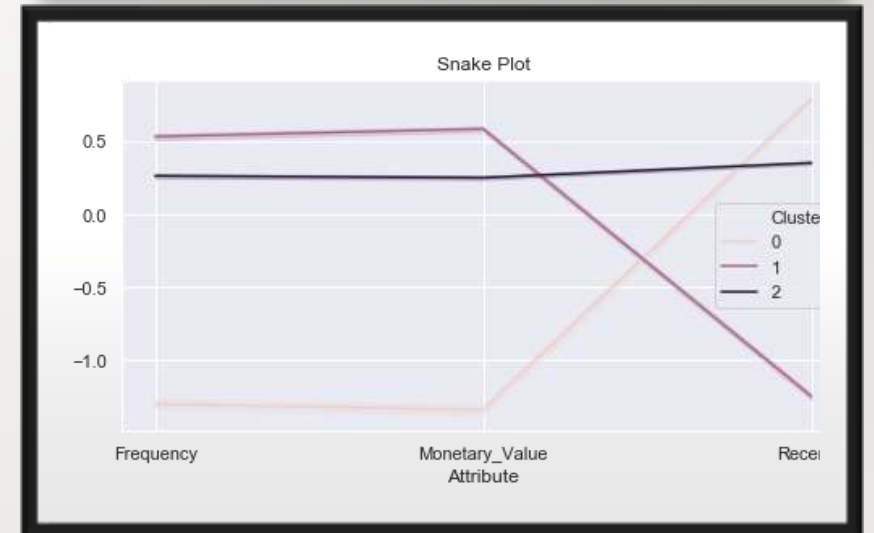
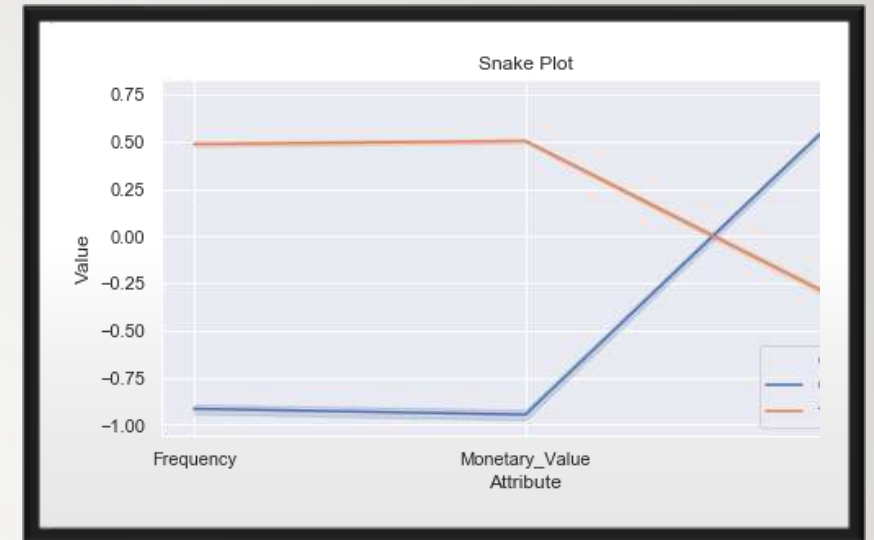
| | Recency | Frequency | Monetary_Value | |
|---------|---------|-----------|----------------|-------|
| | mean | mean | mean | count |
| Cluster | | | | |
| 0 | 112.0 | 3.0 | 699.0 | 3126 |
| 1 | 10.0 | 8.0 | 2972.0 | 4035 |
| 2 | 59.0 | 7.0 | 2232.0 | 7243 |

- k=4, clusters 0 and 3 F & M very close.

| | Recency | Frequency | Monetary_Value | |
|---------|---------|-----------|----------------|-------|
| | mean | mean | mean | count |
| Cluster | | | | |
| 0 | 7.0 | 7.0 | 2699.0 | 2966 |
| 1 | 143.0 | 3.0 | 394.0 | 1331 |
| 2 | 76.0 | 5.0 | 1335.0 | 4984 |
| 3 | 45.0 | 8.0 | 2959.0 | 5123 |

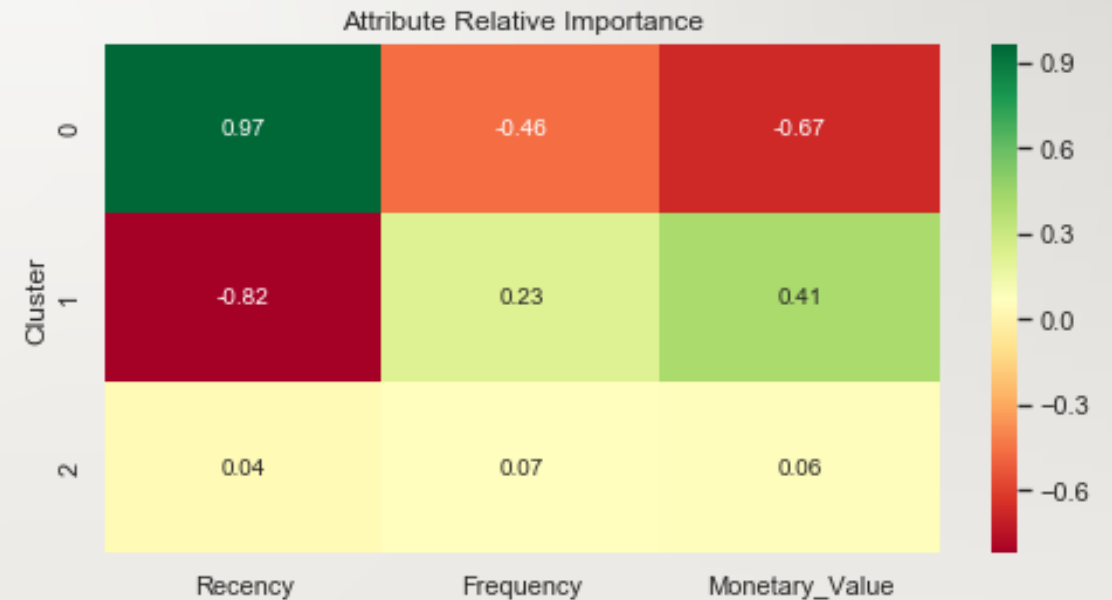
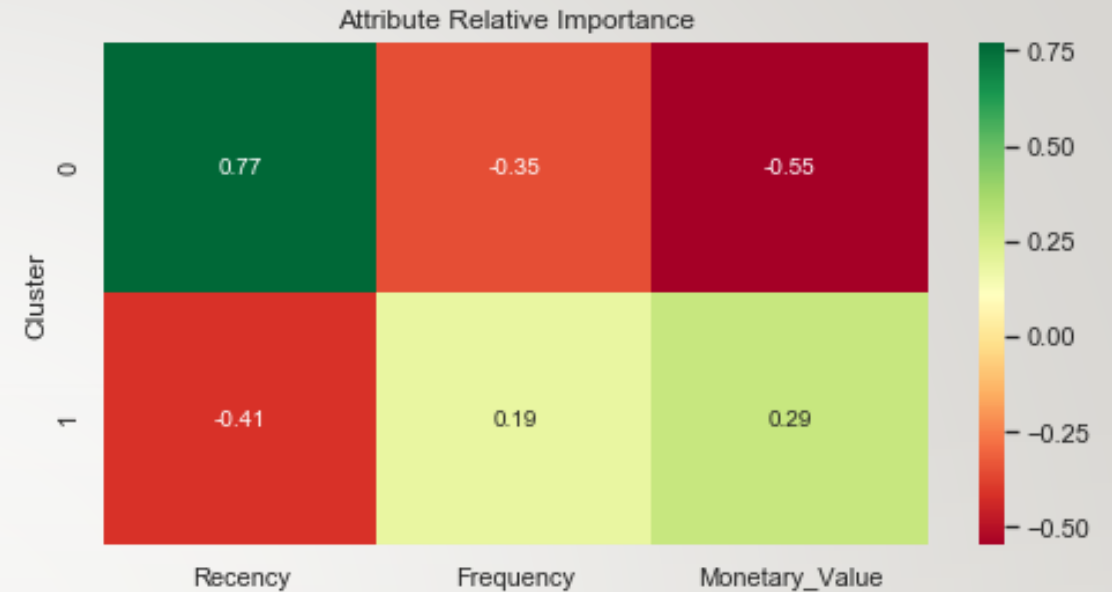
Snake Plots

- Tool for visualizing clusters
- Some overlap with 3 clusters.



Relative Importance of Segment Attributes

- proportion of cluster average to population average
- indicates none of the attributes are important for defining Cluster 2, compared to the population average.



Silhouette Analysis

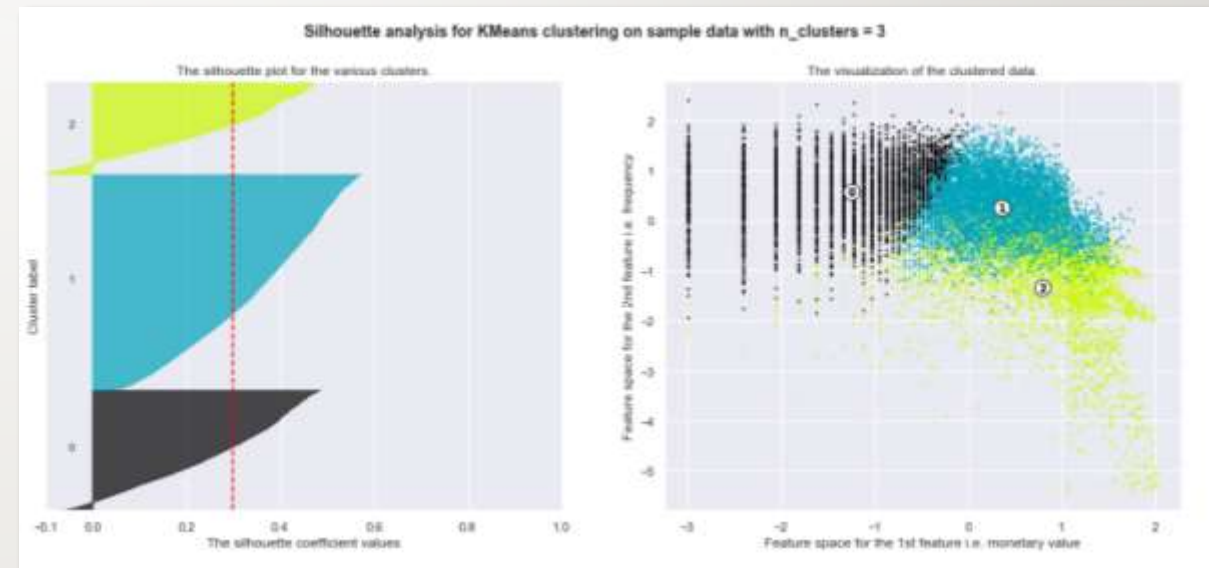
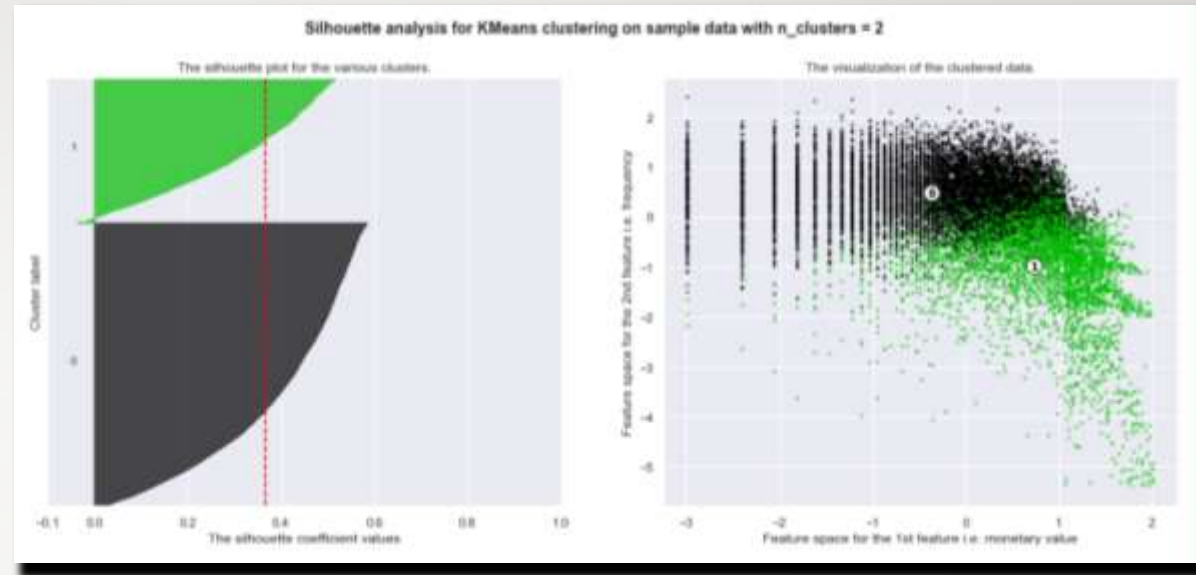
- measures how well each datapoint x_i "fits" its assigned cluster,
- and also how poorly it fits into other clusters.
- a_{xi} = avg distance from x_i to all other points within its own cluster k . The lower the value, the better.
- b_{xi} = min avg distance from x_i to points in a different cluster, minimized over clusters.

$$s(x_i) = \frac{b_{x_i} - a_{x_i}}{\max(a_{x_i}, b_{x_i})}$$

| Range | Interpretation |
|------------|--|
| 0.71 - 1.0 | A strong structure has been found. |
| 0.51 - 0.7 | A reasonable structure has been found. |
| 0.26 - 0.5 | The structure is weak and could be artificial. |
| < 0.25 | No substantial structure has been found. |

Silhouette Analysis

- $n_clusters = 2$,
Silhouette Score = 0.37.
- $n_clusters = 3$,
Silhouette Score = 0.3.
- Best score < 0.5 (with $n_clusters=2$),
- \rightarrow Structure is weak & could be artificial



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Conclusion

- Methods used:
 - RFMT Analysis,
 - K-Means clustering,
 - Snake Plots,
 - Relative Importance of Segment Attributes,
 - Silhouette scores.

Conclusion

- All methods identified two distinct clusters,
- While favoring 2 clusters, a 3 clusters option is also possible,
- RFMT Analysis → 3 almost equally distanced clusters feasible,



Conclusion

- For better capturing customer behavior,
- and more focused marketing to target diverse customers,
- suggest 3-clustered customer segmentation, pending managerial decision.