

Instacart:

Market Basket Analysis with Customer and Product Segmentation



What is Instacart?

An American company operating a grocery delivery service made abundantly necessary in the times of COVID-19.



“In a matter of a couple of weeks, we were already ahead of our end-of-year goal. A week later, we were ahead of our 2021 goals, and a few days after that, we were ahead of our 2022 goals.”

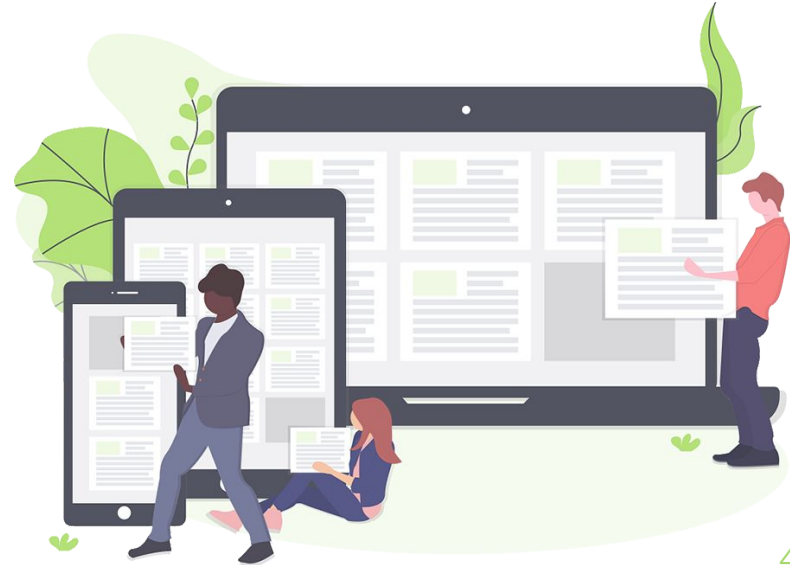
Apoorva Mehta, CEO, Instacart, Inc.
via Bloomberg Businessweek



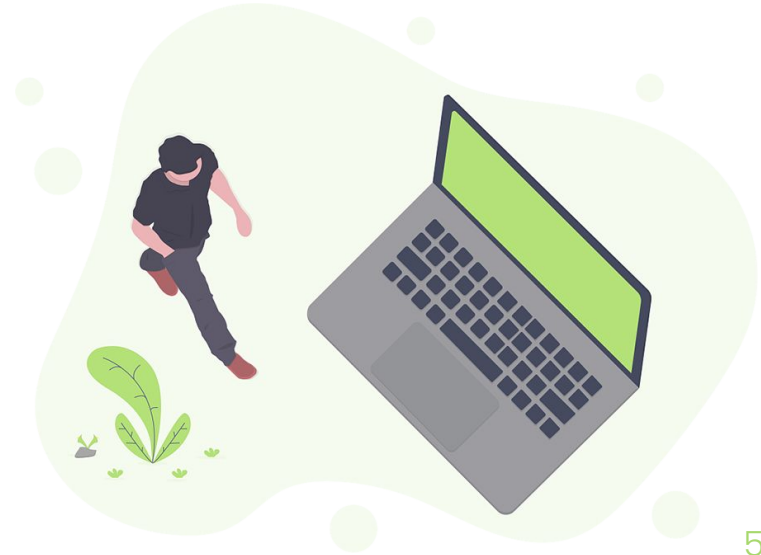
Our Selected Topic



- Market Basket Analysis
- Customer Segmentation
- Product Segmentation



The Exploratory Phase



1,315,590 orders

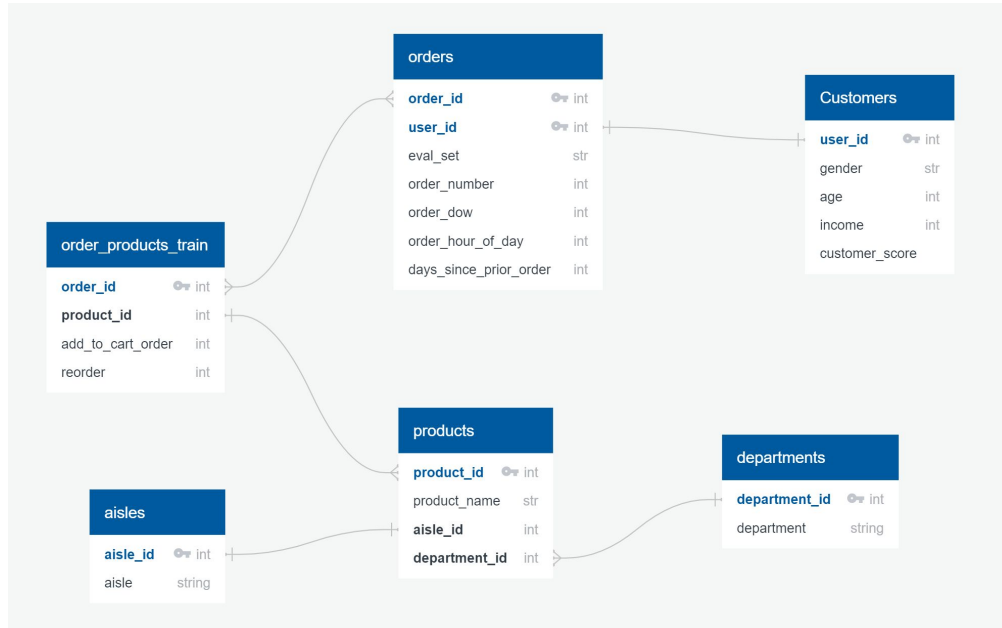
49,688 products

12,950 customers

*All data initially in six tables



The Exploratory Phase Continued..



Questions We Hope to Answer

Customer Segmentation:

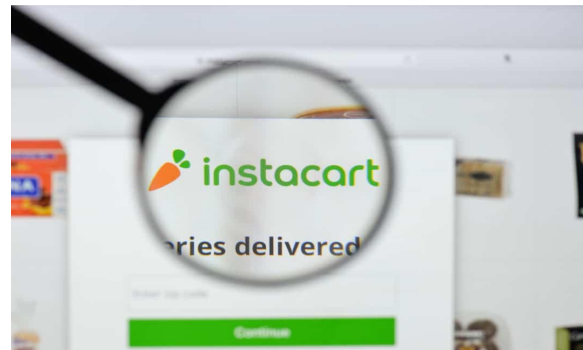
1. What is the distribution of annual Income by age?
2. What is the distribution of gender for Instacart users?
3. What is the age group of most frequent Instacart users?
4. What is the overall distribution of income level for instacart users?
5. How do male and female spending scores differ?
6. Who is our targeted customer and who is our least active customer?



Questions We Hope to Answer

Product Segmentation:

1. What are the Top 10 most ordered products?
2. What are the Top 10 most reordered products?
3. Which 2 products were ordered together the most?
4. Which day and time of the week was the busiest in terms of number of orders received?
5. Which products were not frequently reordered? This could help determine which products did not meet customers expectations.
6. What were the top aisles with the least amount of orders? This could be a recommendation that we make to our vendors in terms of restructuring their aisles.



The Analysis Phase

Tableau Analysis
and
Machine Learning

Customer Analysis | Product Analysis



Machine Learning

Customer Segmentation:

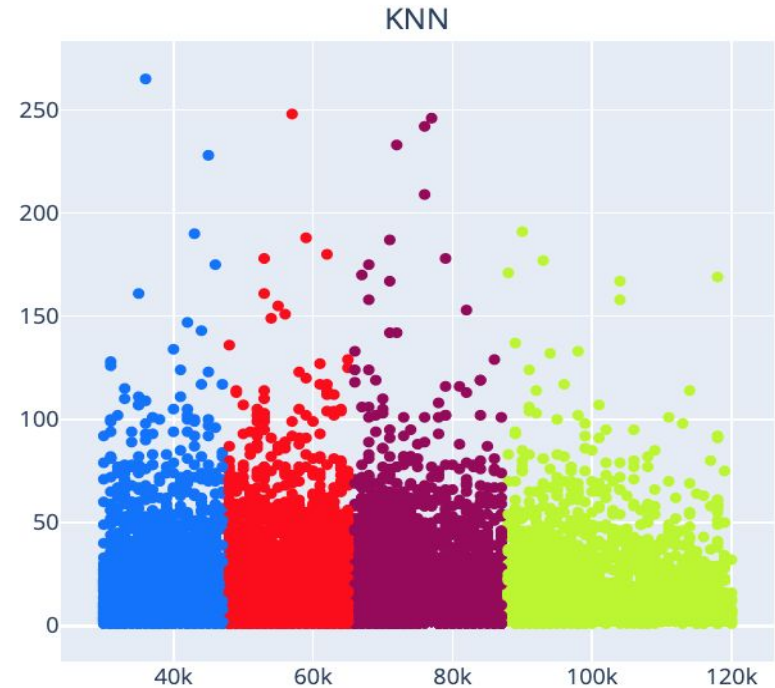
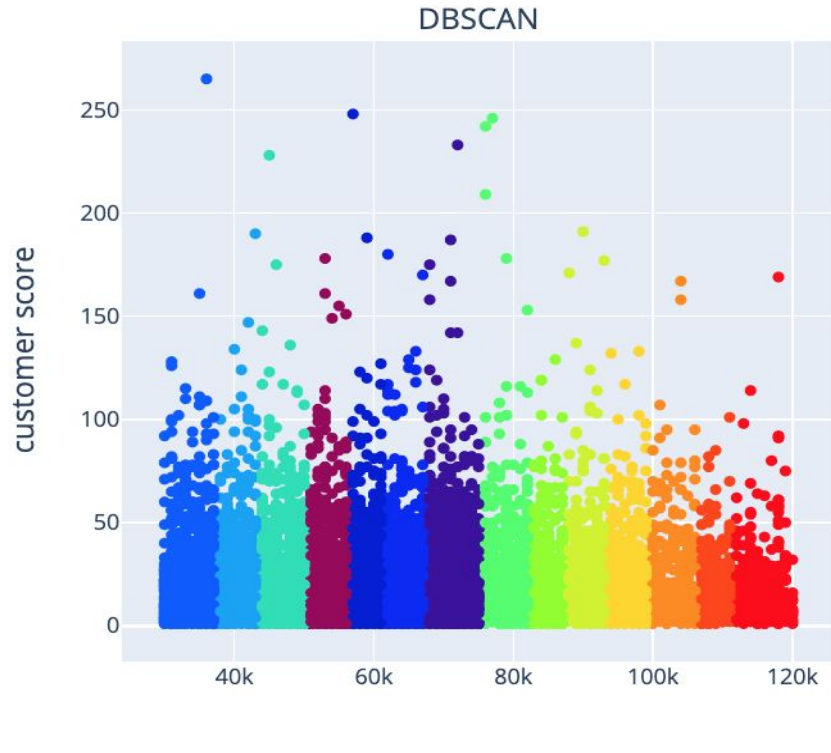
Utilizing KMeans, we hope to be successful in clustering and categorizing the Instacart customers into four main groups:

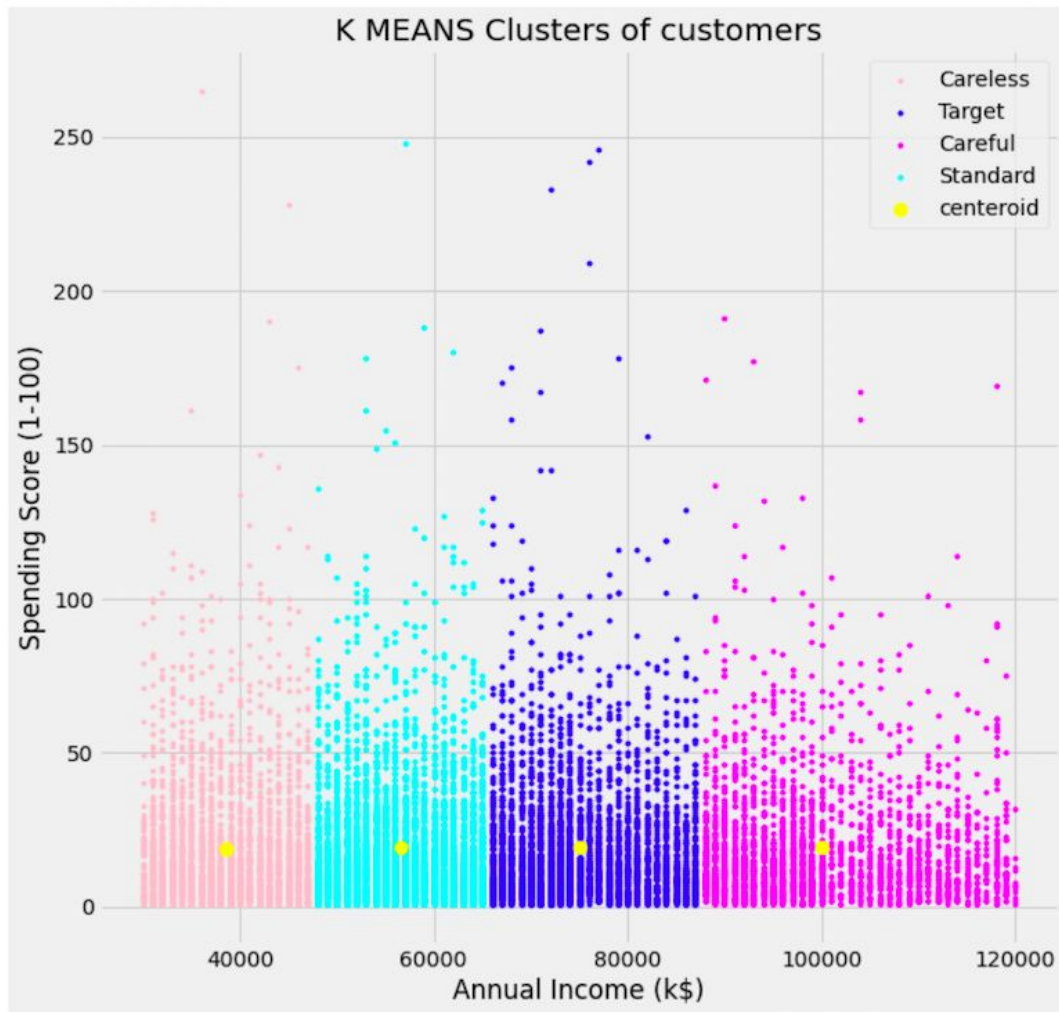
- Target
- Standard
- Careless
- Careful



Density Based Clustering

K-Means

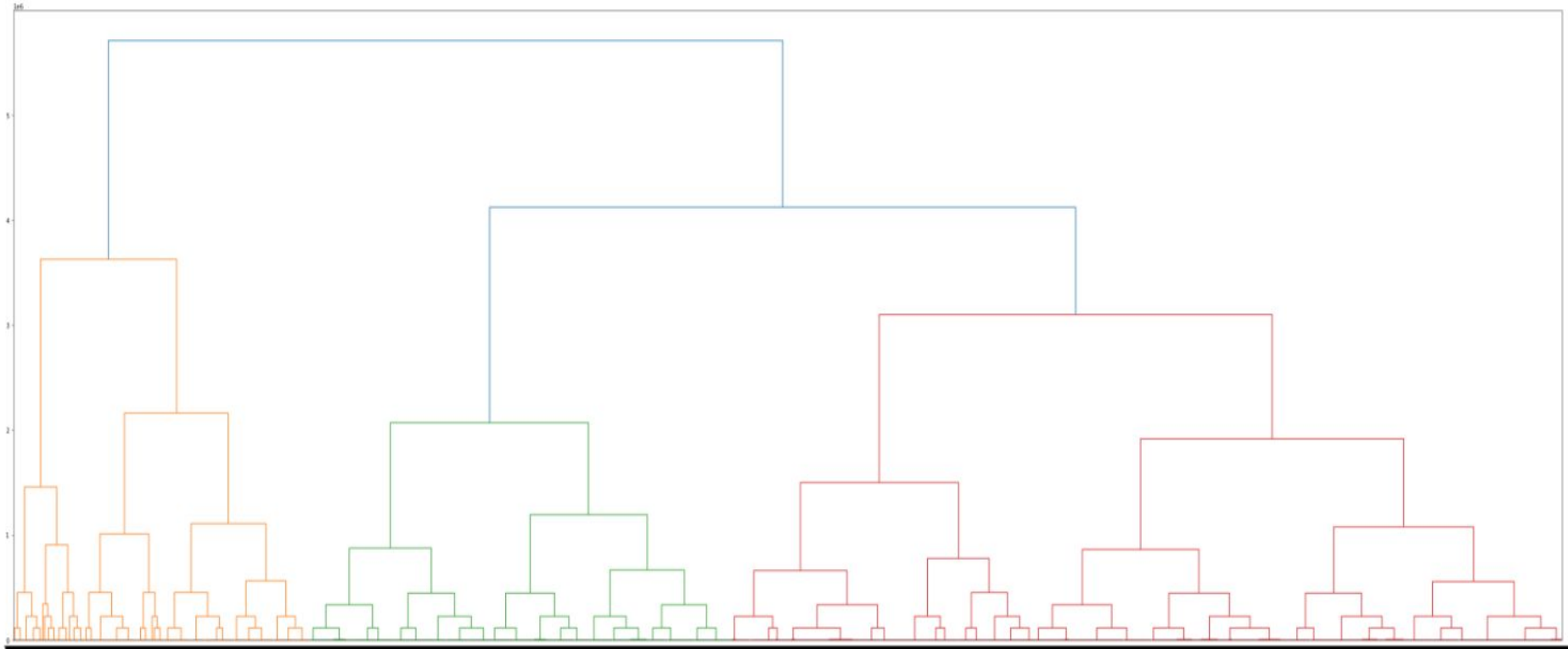




**Grouping the customers
based on their Annual
Income and Spending Score
using K MEANS**



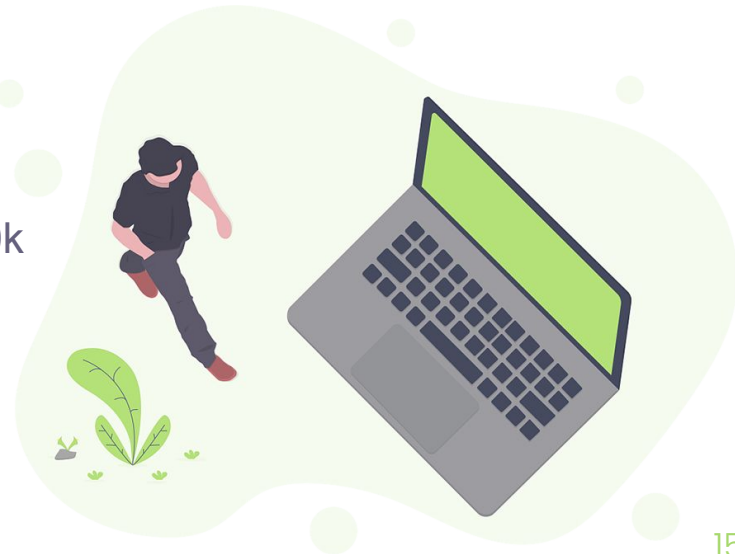
Dendrogram (Hierarchical Clustering)



Results of Analysis

Customer Segmentation

- The highest incomes came from middle aged and older customers
- Females were greater represented in the dataset (female was 58.63%, male was 41.37%)
- Ages 50-65 were our most frequent Instacart users
- Income distribution ranged from 30-120k with 50-70k being the most common income among our users.
- Males and females had incredibly similar spending scores and spending habits



Machine Learning

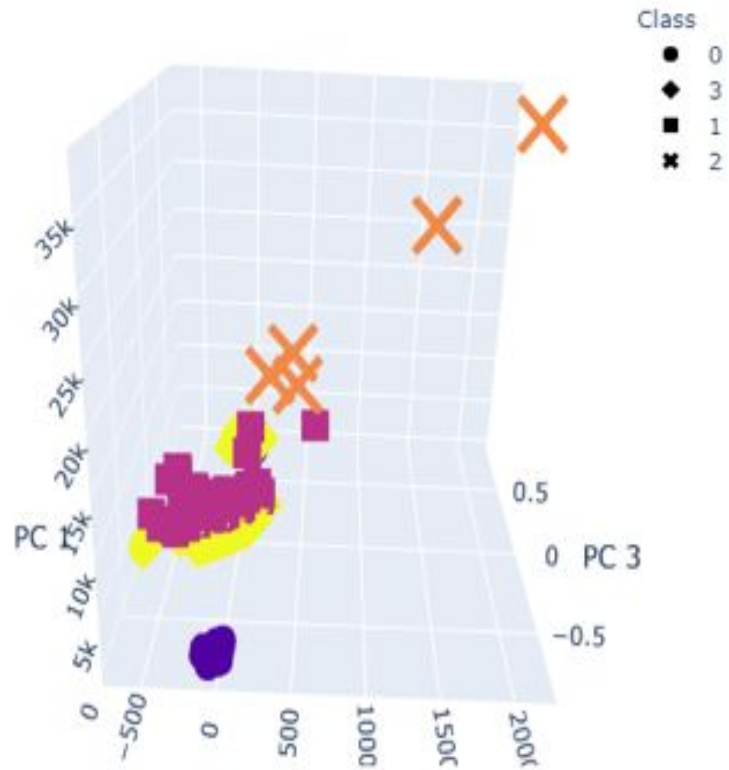
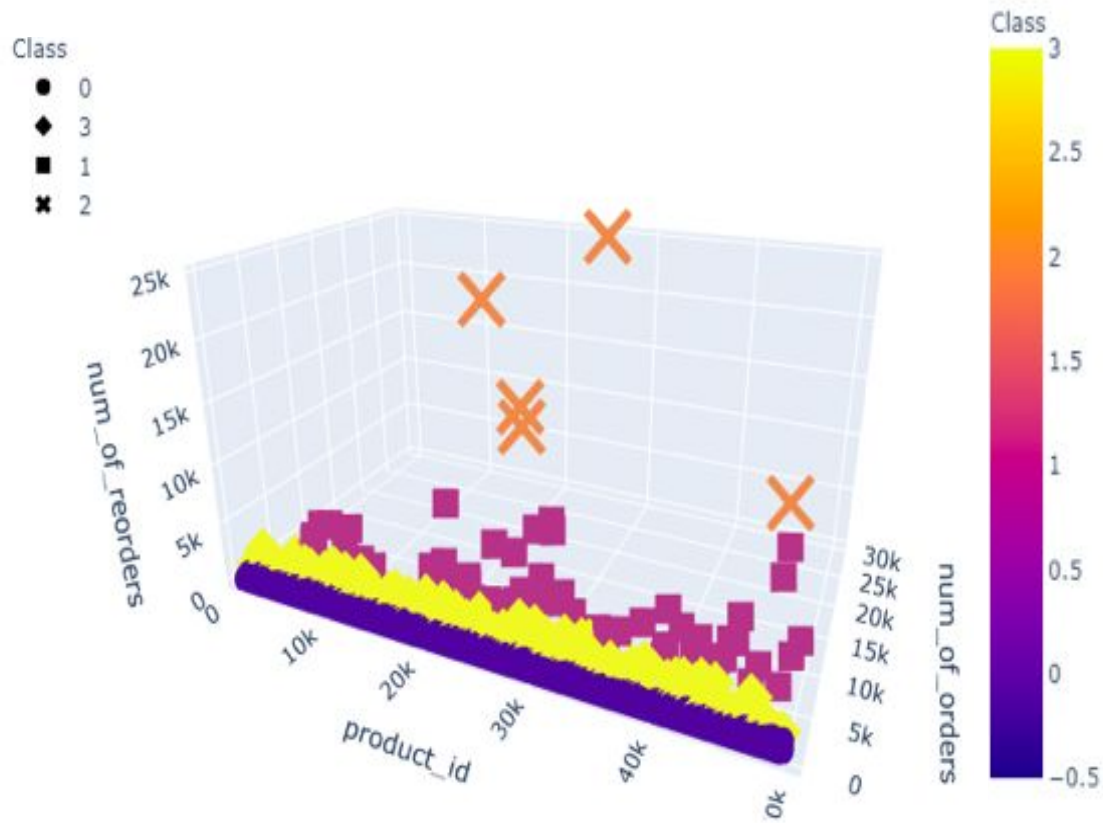
Product Segmentation:

Utilizing KMeans, we hope to be successful in clustering and categorizing the Instacart products into four main groups:

- Class 0 - Never Reordered
- Class 1 - Occasionally Reordered
- Class 2 - Often Reordered:
“Popular”
- Class 3 - Frequently Reordered:
“High Demand”



Product Clustering



Deep Learning Neural Networks to Predict Product Reordering

	order_dow	order_hour_of_day	days_since_prior_order	add_to_cart_order	reordered	num_of_reorders	num_of_orders	department	aisle
0	5	0	15.0	1	1	6180	8882	produce	fresh fruits
1	5	0	15.0	2	1	3831	6083	produce	packaged vegetables fruits
2	5	0	15.0	3	1	2461	3595	produce	packaged produce
3	5	0	15.0	4	0	211	387	meat seafood	hot dogs bacon sausage
4	5	0	15.0	5	0	208	317	produce	fresh fruits
...
1864657	2	9	7.0	16	1	146	275	snacks	nuts seeds dried fruit
1864658	2	9	7.0	17	1	699	1120	bakery	bread
1864659	2	9	7.0	18	1	224	393	snacks	chips pretzels
1864660	2	9	7.0	19	0	13	37	pantry	pickled goods olives
1864661	2	9	7.0	20	0	290	714	pantry	oils vinegars

1812546 rows × 9 columns

Model: "sequential_1"

Layer (type)	Output Shape	Param #
dense_6 (Dense)	(None, 570)	108870
dense_7 (Dense)	(None, 950)	542450
dense_8 (Dense)	(None, 950)	903450
dense_9 (Dense)	(None, 950)	903450
dense_10 (Dense)	(None, 950)	903450
dense_11 (Dense)	(None, 1)	951
Total params: 3,362,621		
Trainable params: 3,362,621		
Non-trainable params: 0		

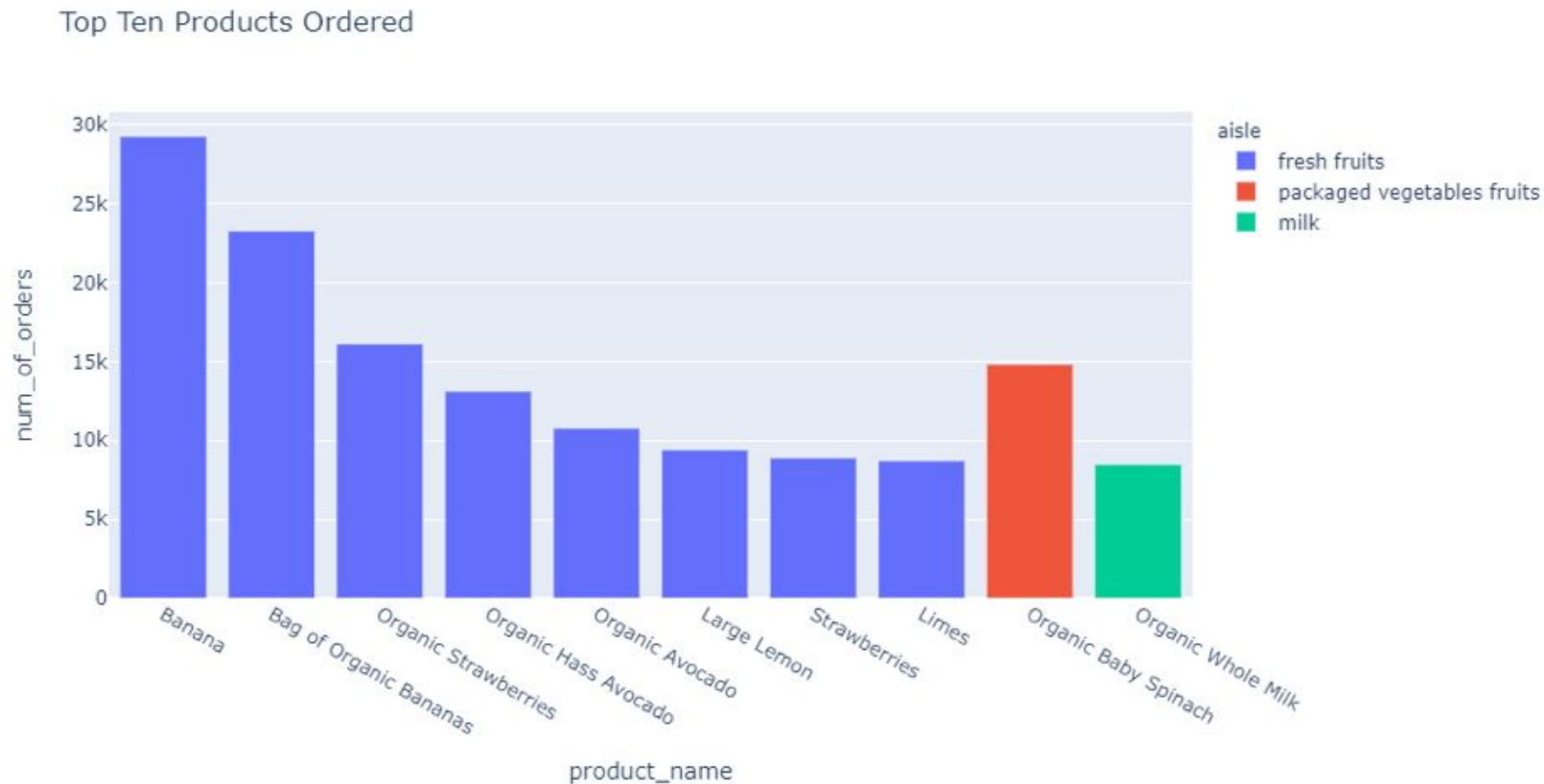
- Received an accuracy of 70.5%

8114/8114 - 88s - loss: 0.5889 - accuracy: 0.7007
Loss: 0.5889468789100647, Accuracy: 0.7007383704185486

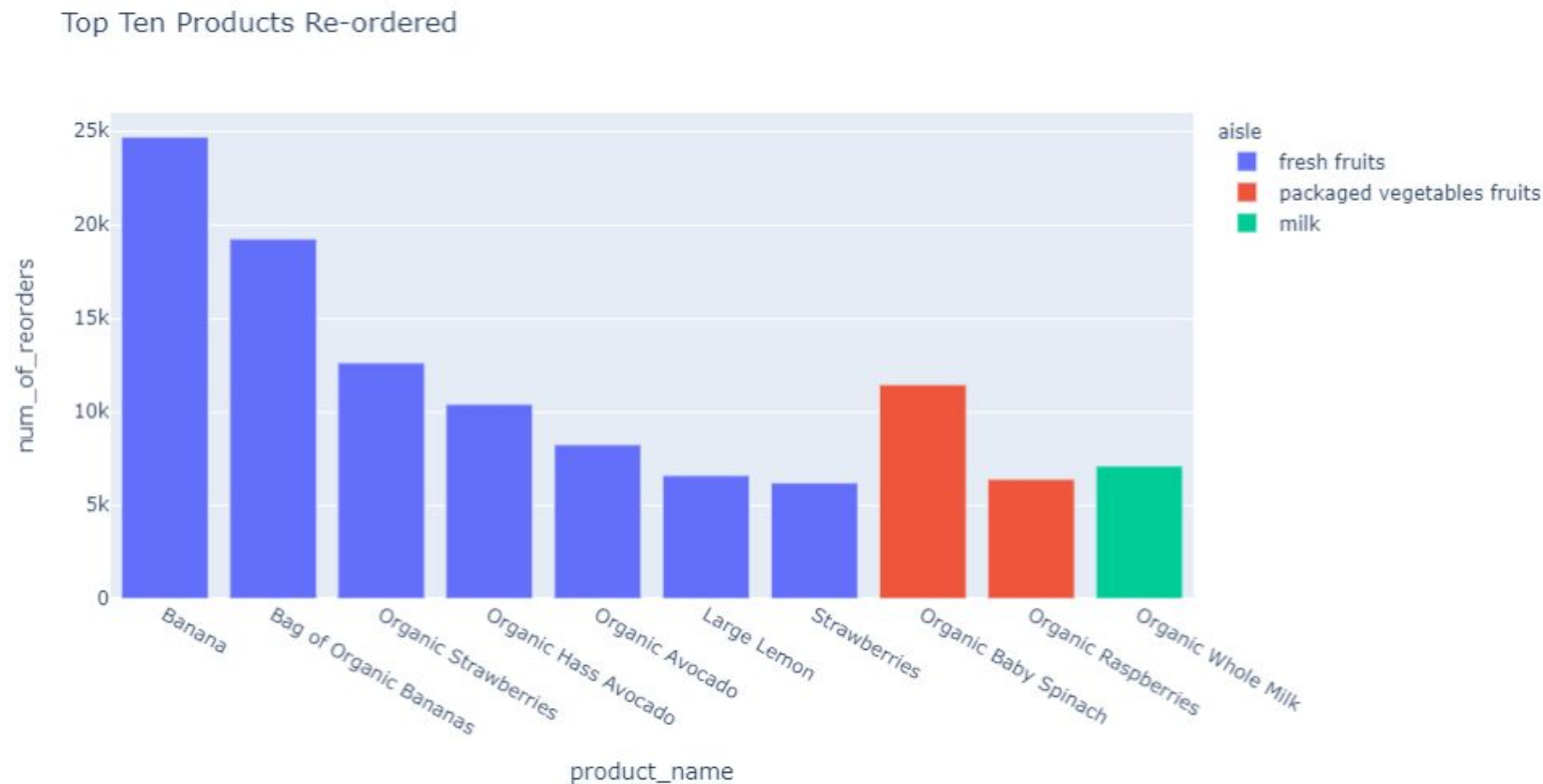
Training the Model



Top 10 Ordered Products



Top 10 Reordered Products



Results of Analysis

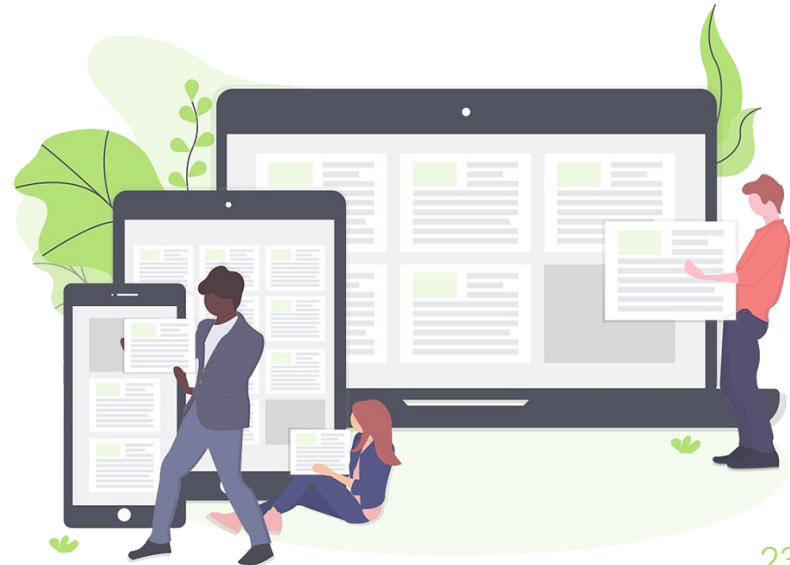
Product Segmentation

- Busiest times are on weekends:
 - Saturday(12-2 PM) and on Sundays(10-12 PM)
- Products that were not reordered frequently:
 - Corn Starch, Paprika, Bay leaves, Cayenne Pepper, Ground Ginger
- Aisles that had the least orders:
 - Beauty, Baby Accessories, Vitamin Supplements, First Aid, Kitchen Supplies
- Top products ordered together:
 - Bags of Organic Bananas + Organic Avocados
 - Bags of Organic Bananas + Organic Strawberries
 - Spinach + Bananas
- Top Departments with most orders:
 - Produce and Dairy Eggs and Beverages



Technologically Speaking

- Methodologies:
 - Jupyter Notebook, PGAdmin, Tableau, Google Slides, SQL, Python, Postgres, AWS, VS Code, GitHub
- Algorithms:
 - K-Means Clustering
 - Neural Networks
 - DB Scan
 - KNN
 - Dendrogram (Hierarchical Clustering)



What We May Have Done Differently

- Customer buying prediction model
- Build an API (time restrictions)
- Choosing a project topic based on available datasets instead of choosing a topic and having to scramble to find a suitable dataset



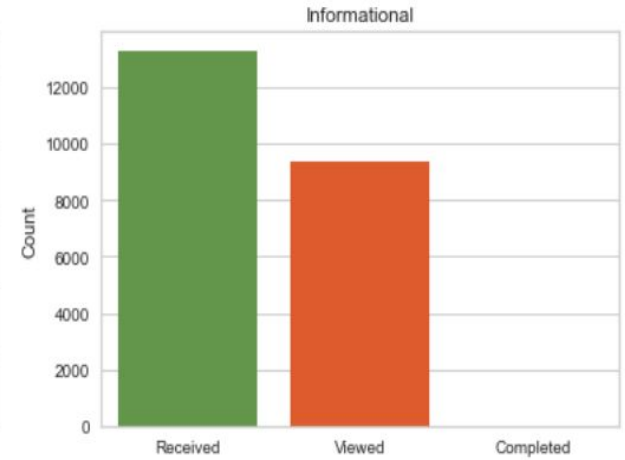
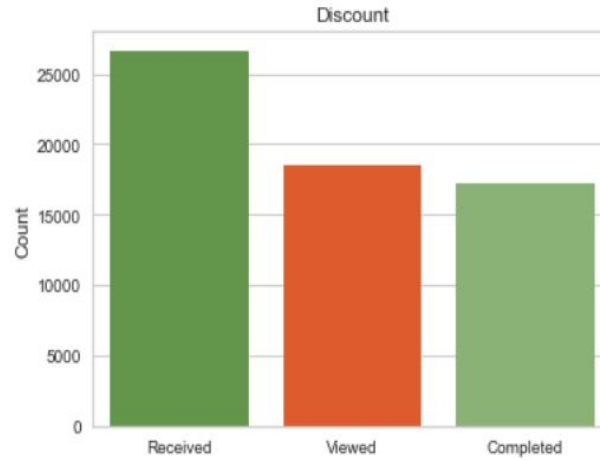
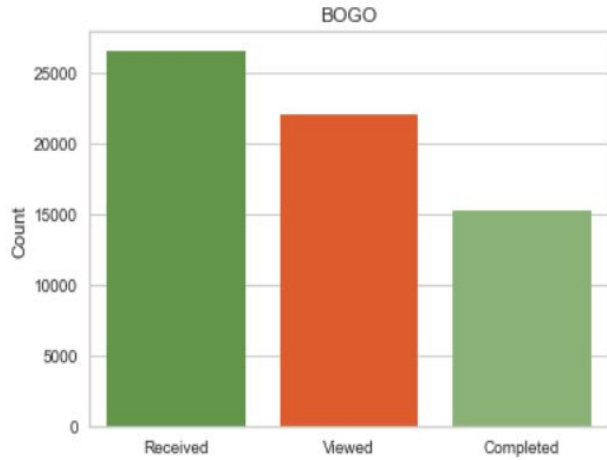
Recommendations for Future Analysis

- Combine more features from the customers dataset so that we can have a better prediction for specific customer groups
- Promos for under engaged demographics
- Promos for targeted customers
- Promos for targeting low selling products

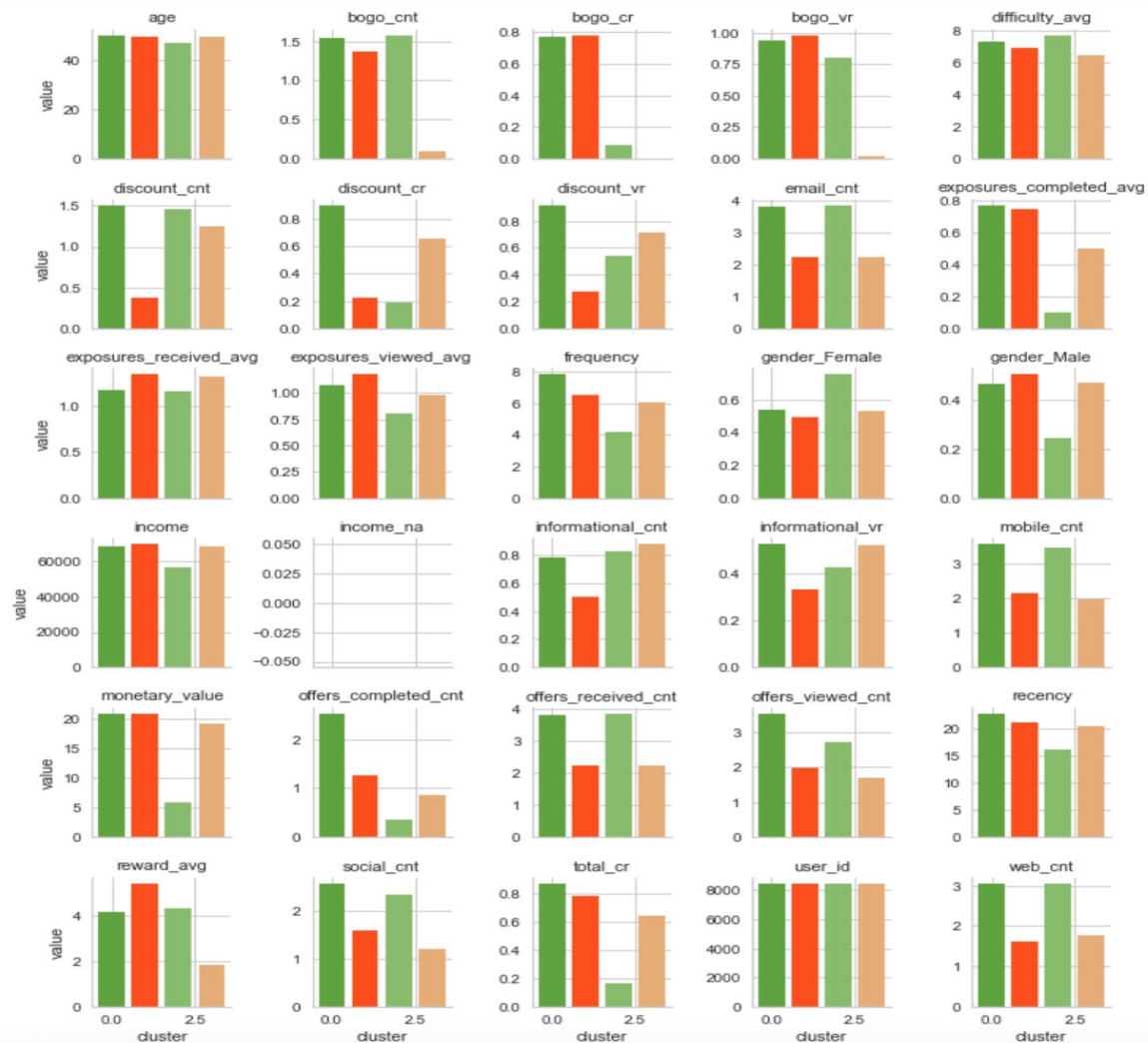


Promotions

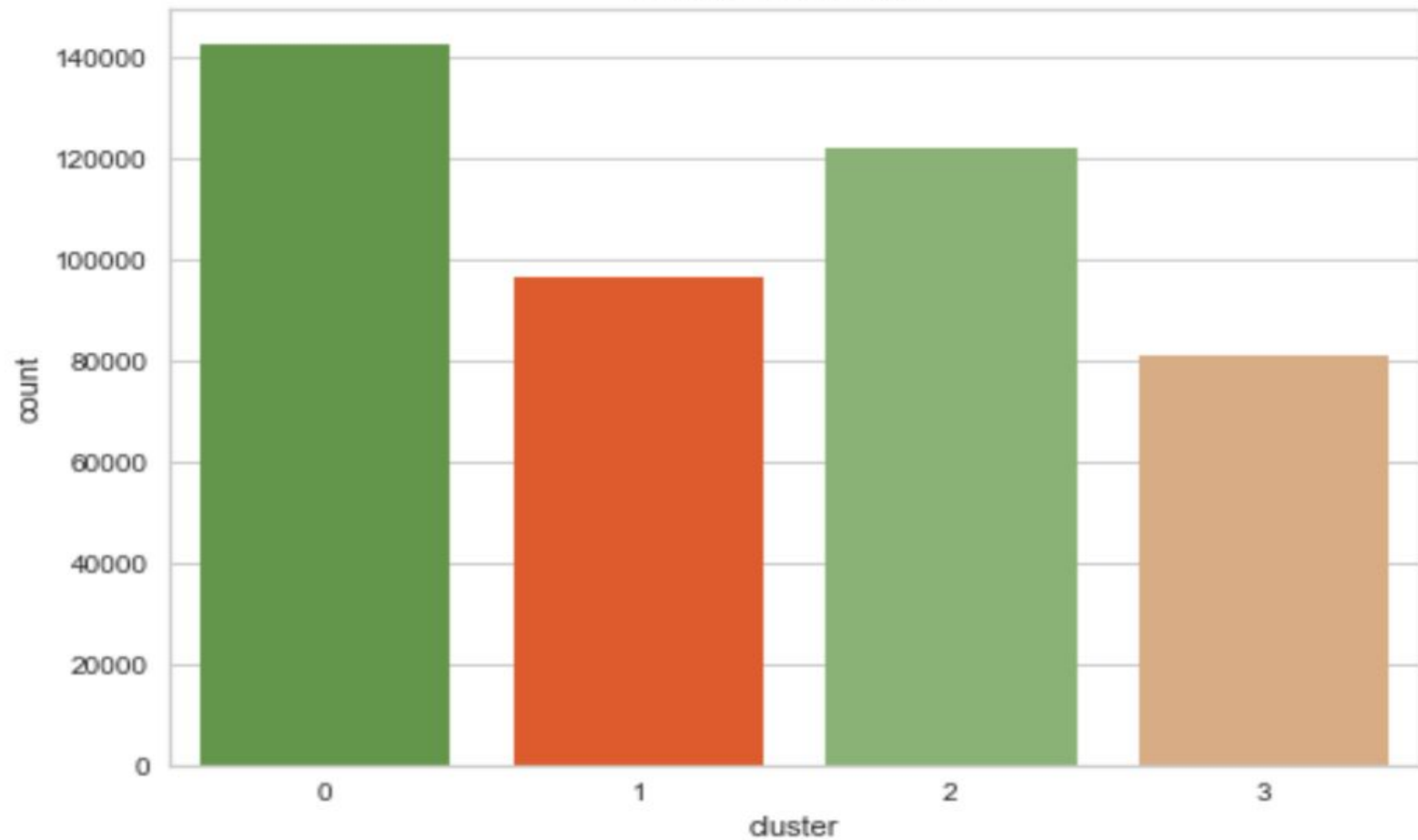
Offer conversion funnel by type



PROMOS



Customers by cluster



Team Presentation



Ketaki Pradhan
Team Member



Leora Talmor
Team Member



Surekha Chandramouli
Team Member



Edward Rivera
Team Member



Allison Corrales
Team Member

