
Personalized Recommendation System For instacart

Presented by Wisdom Chen, Boya Zeng, Jessy Hu, Yuqing Wu



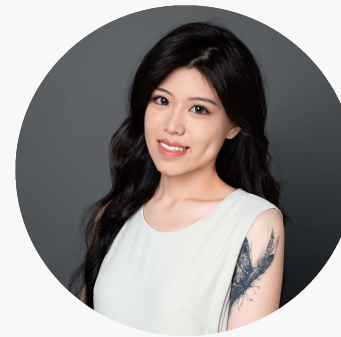
Wisdom Chen

Data Scientist



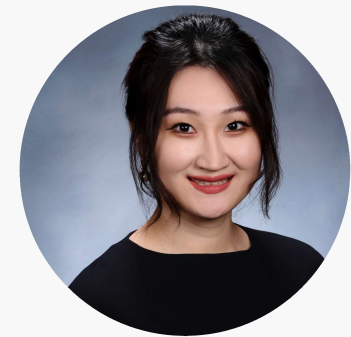
Boya Zeng

Machine Learning Scientist



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Big Data Analyst

Our Mission

Translate complex data into actionable strategies, driving success and competitive advantage for our clients

OUR GOAL: LEVERAGING DATA ANALYTICS TO DRIVE PERSONALIZED MARKETING STRATEGIES IN ONLINE GROCERY SHOPPING

INTRODUCTION



Situation

In the highly competitive online grocery delivery sector, businesses like Instacart strive to **understand customer behaviors, preferences, and segmentation** to enhance service personalization and operational efficiency.

We aimed to dissect the extensive Instacart dataset available on Kaggle to **uncover actionable insights on customer purchasing behaviors**.

Task



Action

We will analyze **purchasing behaviors**, **segment customers**, perform **market basket analysis**, develop a **recommendation system**, and create a **recommendation dashboard**

The recommendation system and dashboard facilitated data-driven decisions, setting the stage for a personalized marketing strategy to enhance customer engagement and boost sales.

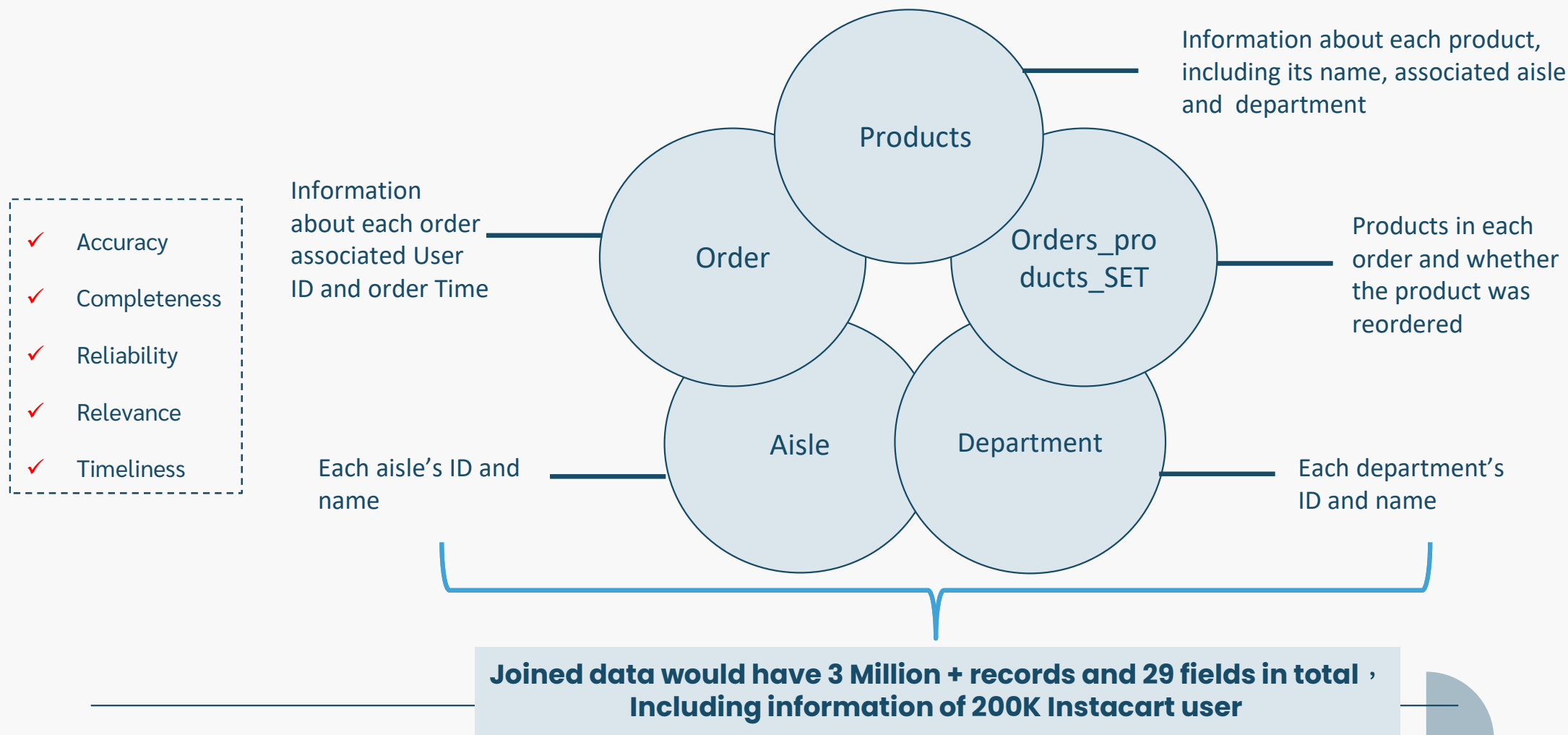
Result



**DATA OVERVIEW: JOINED DATASET ENSURES QUALITY WITH OVER 3 MILLION RECORDS
ACROSS 29 FIELDS CAPTURING COMPREHENSIVE INSTACART USER ACTIVITY**

INTRODUCTION

DATA
INSIGHTS

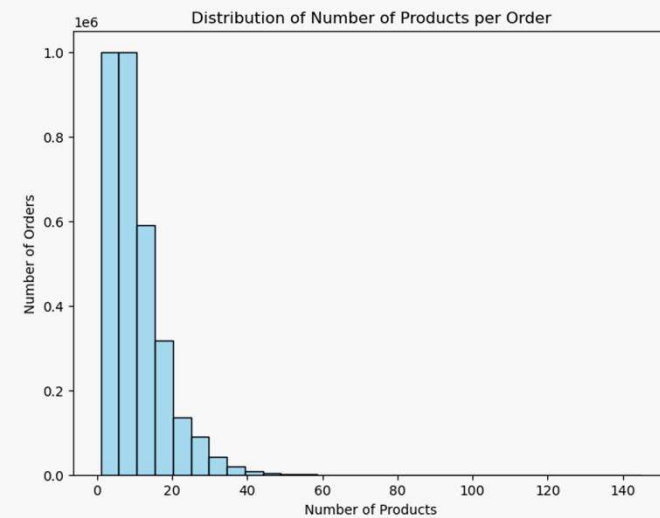


DATA EXPLORATION REVEALS 59% REORDER RATE; CUSTOMERS COMMONLY ORDER

20 ITEMS, PREFERRING FOOD AND PERSONAL CARE PRODUCTS

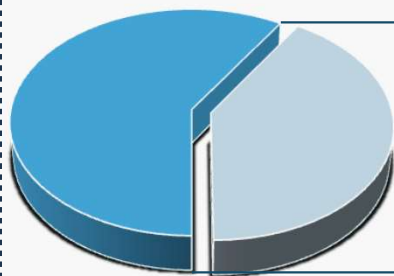
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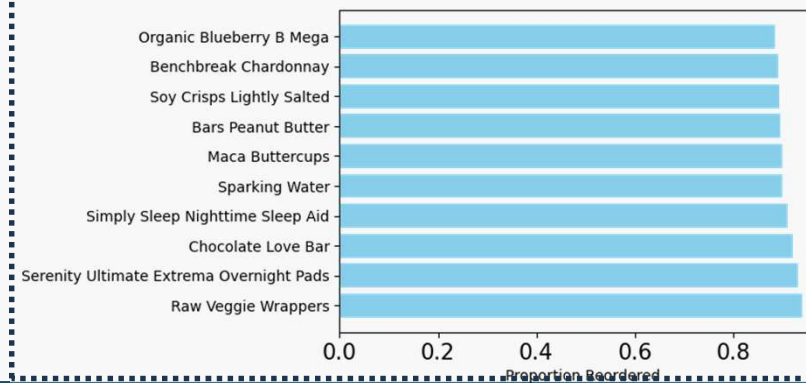
Customer typically purchase **20 products** within a single order

Reorder Distribution



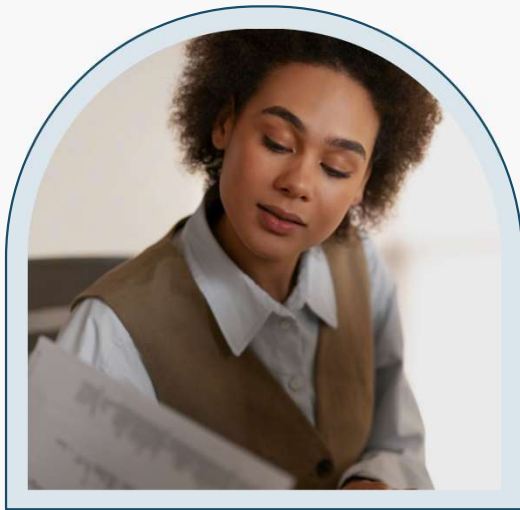
41% of the customer did not reorder since the initial purchase

■ Reordered ■ Did Not Reordered



Food and Personal Care are most frequently reordered by user

DATA EXPLORATION OF CUSTOMER'S PURCHASE BEHAVIOR



Name

Anna Wilson

Age

22 years old

Profile

Loyal Customer

Trait 1

peak activity around midday

Trait 2

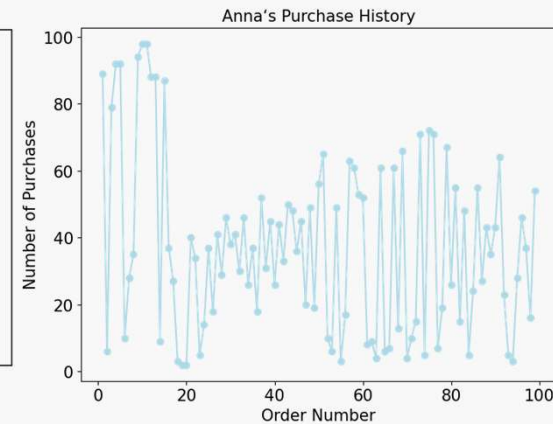
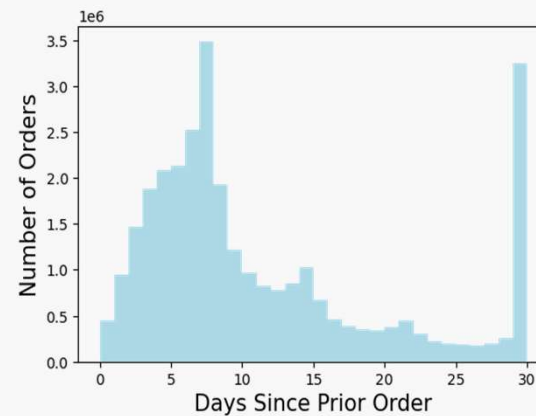
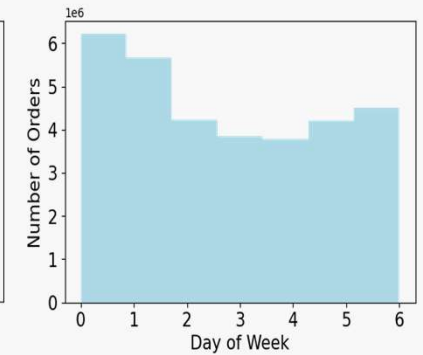
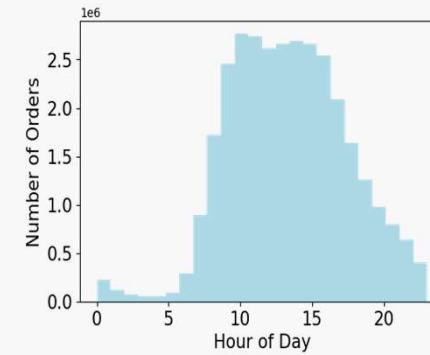
stocking up early in the week

Trait 3

daily or monthly replenishment.

Trait 4

routine and bulk purchasing behaviors



Customers ordering peaks midday and early in the week, with frequent same-day reorders or monthly patterns, while individual habits, like Anna's, vary from regular small purchases to intermittent bulk orders.

IDENTIFY DISTINCT CUSTOMER SEGMENTS USING K-MEANS CLUSTERING BASED ON THEIR SHOPPING PATTERNS TO TAILOR FURTHER ANALYSIS

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MODELING

Data Preparation

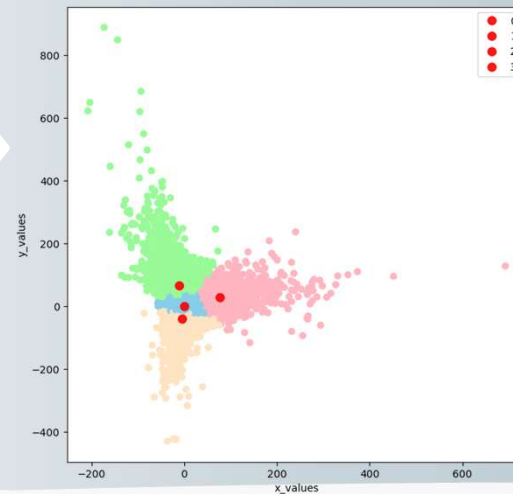
Aggregated purchase data by customer and aisle using cross tab

Aisle User ID	Baby food formula	Yogurt	Tea	...
1	0	3	0	...
2	1	0	0	...
...
134	12	0	0	...

E.g., User #134 purchased 12 times of baby food formula

Segment Customer with K-Means

- Performed **PCA** for dimension reduction
- Utilized **k-means clustering** to segment customers into groups with similar aisle preferences.
- Determined optimal cluster number through **Elbow Method** and **Silhouette Score analysis**

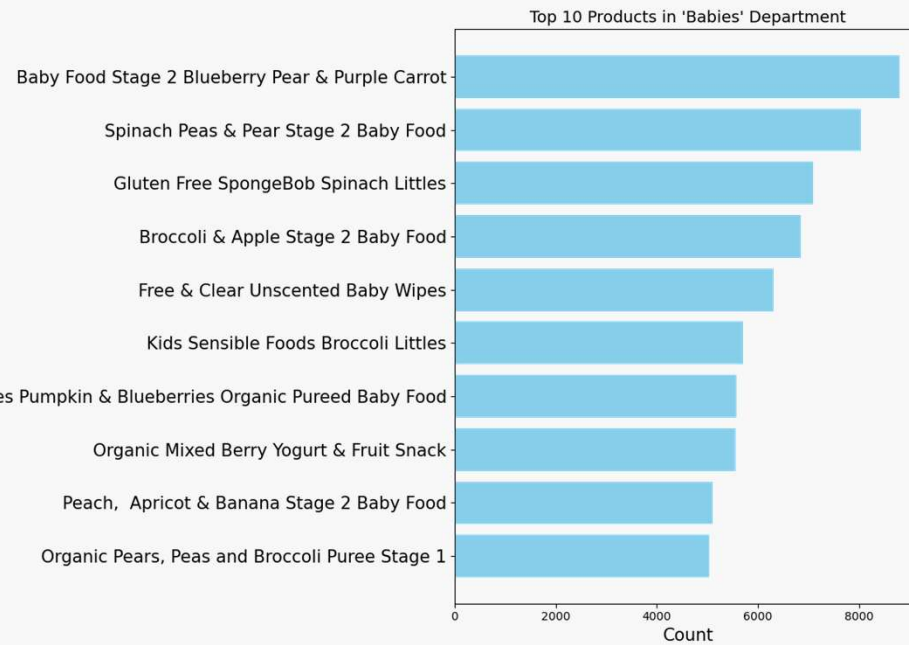


Distinguish Segment of Interest

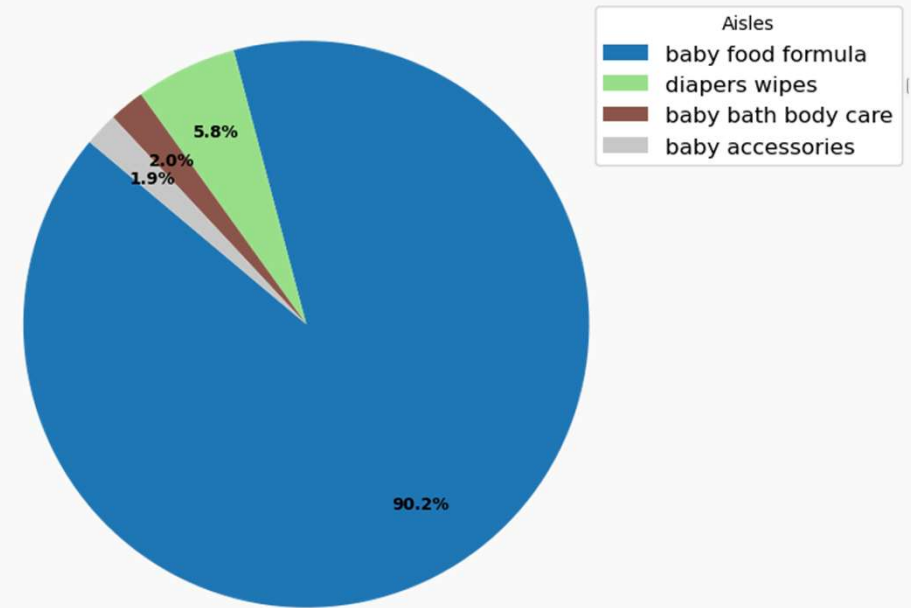
- 3 out of 4 clusters primarily purchase fresh fruits/vegetables
- while 1 distinct cluster mainly buys baby food formula.**

Cluster 1	Cluster 2	Cluster 3	Cluster 4
Fresh veggie	Fresh fruit	Fresh fruit	Baby food formula
Fresh fruit	Fresh veggie	Yogurt	Fresh fruit
...

TAKE A CLOSER LOOK INTO BABIES DEPARTMENT



Top10 Products in Department of Babies



Distribution of Products by Aisle

Next Step : Identify patterns of product purchases and discover associations between different products.

PERFORM MARKET BASKET ANALYSIS USING APRIORI ALGORITHM TO IDENTIFY ASSOCIATION BETWEEN ITEMS

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Association Rule Mining with Apriori

- Transform order data of cluster 4 to list of transactions with one-hot encoding

	Coffee Filters	Organic Milk	Greek Yogurt	...
0	True	True	False	...
1	False	True	False	...
2	False	False	False	...
...

- Set minimum threshold: 0.01 for support, 1 for lift

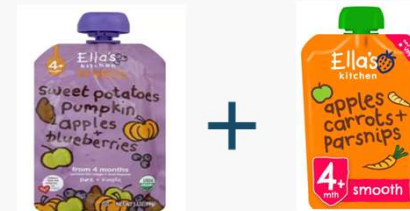
	Antecedents	Consequents	Support	Confidence	Lift
0	Whole Wheat Bread	Organic Bananas	0.01	0.30	1.51
1	Organic Bananas	Apple Honeycrisp Organic	0.02	0.08	1.62
...
193	Organic Hass Avocado	Organic Bananas	0.05	0.39	1.97

Item Sets Example



Support	0.01
Confidence	0.50
Lift	2.52

- High confidence
- Indicates these two items are often bought together

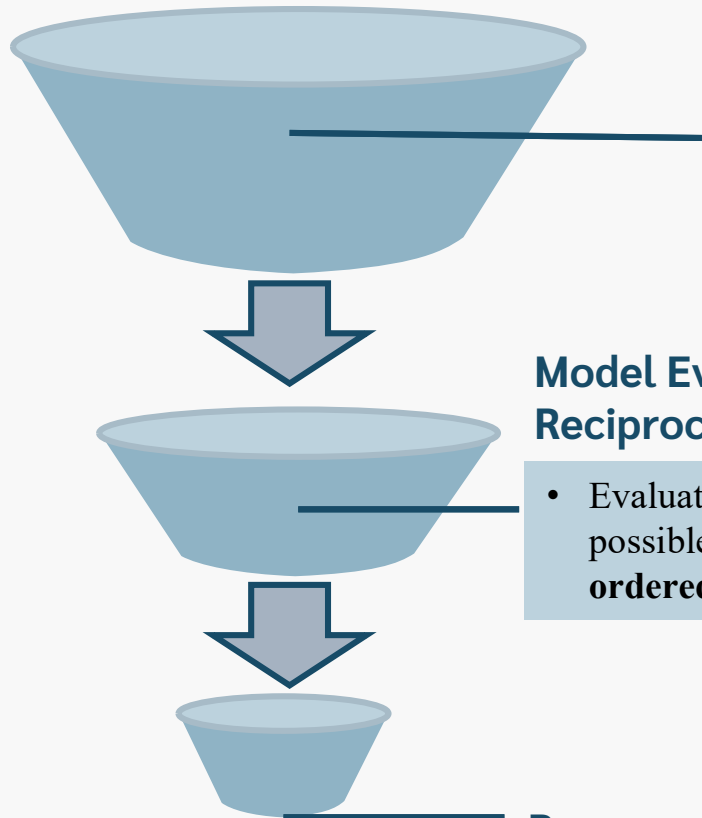


Support	0.01
Confidence	0.29
Lift	15.89

- High lift
- Indicates if A is sold the ratio of sale of B will increase more than 15 times.

Hybrid Recommendation Systems

- **Collaborative Filtering:** Historical of reordered product
- **Content-Based:** Recommend items that are similar in content which are the users' choice



Model Evaluation -- Mean Reciprocal Rank (MRR)

- Evaluating any process that produces a list of possible responses to a sample of queries, **ordered by probability of correctness**

Recommendation Dashboard

MRR (Apple)	
Items	Scores
Pink Lady Apple	0.78
Apple Cinnamon Fig Bar	0.62
Organics Vitamin C Apple Juice	0.51
Apple Apple Applesauce On The Go Pouches	0.49
Braeburn Apple	0.47

TAILORED **MARKETING STRATEGY** BASED ON OUR MODELING ENABLING POTENTIAL **BUSINESS VALUE**

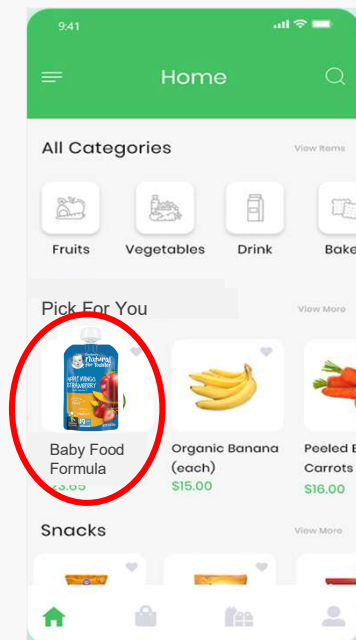
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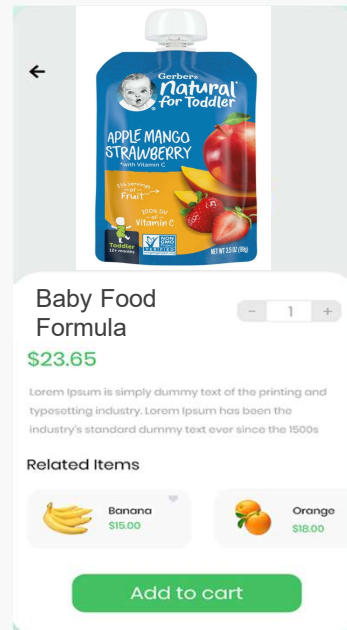
BUSINESS
VALUE

Homepage Promotion



- ✓ Engage visitors from the moment they land on the homepage by using Recommendation System

Add-to-Cart Recommendation



- ✓ Maximize the opportunity for additional sales using Market Basket Analysis



Business Value

- Enhanced Customer Segmentation
- Personalized Shopping Experience
- Efficient Marketing and Promotions

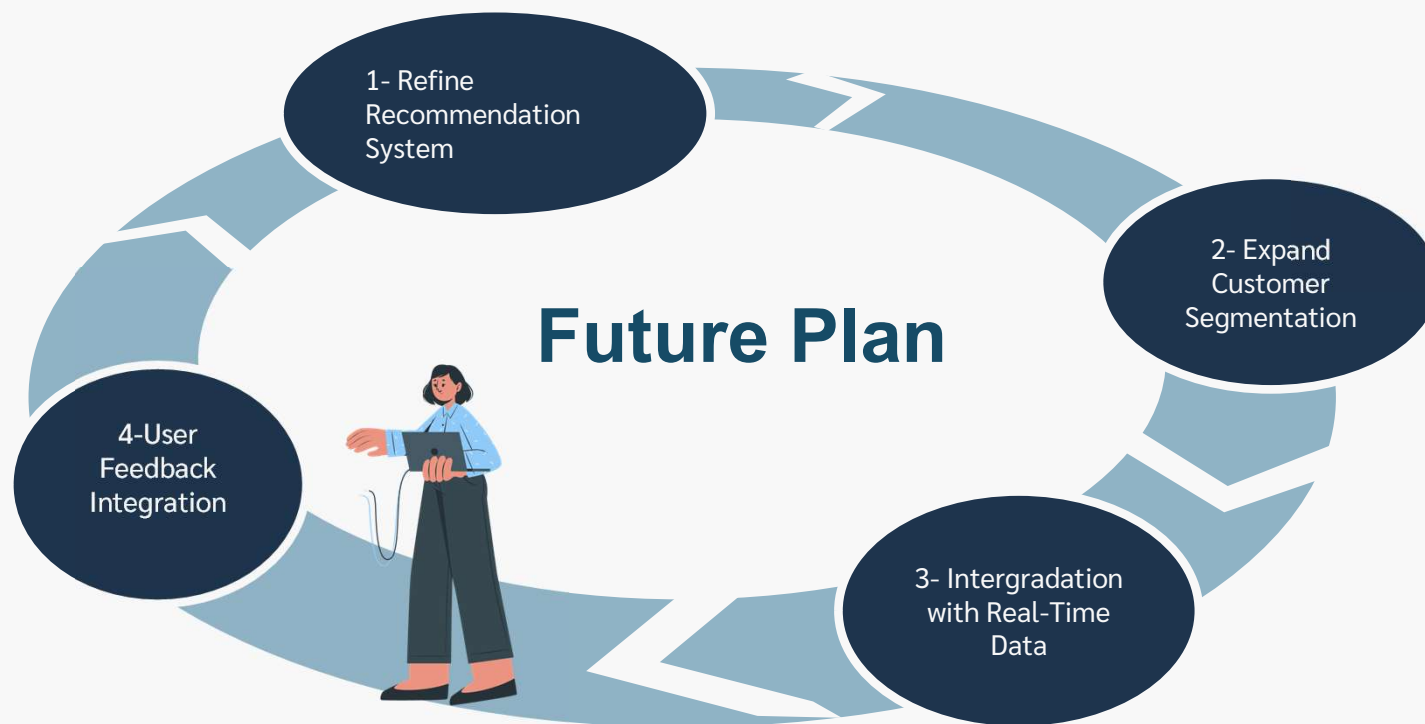
OUR **FUTURE PLAN** STAYS TRUE TO FLEXIBLE AGILE PPRINCIPLE WITH 4 STEPS

ALLOWS CONTINUOUS FEEDBACK AND ITERATIVE DEVELOPMENT

INTRODUCTION

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MODELING

BUSINESS
VALUEFUTURE
PLAN

Slide 12

YCO

Business values + Marketing Strategy = 1 Slide

Wisdom Chen, 2024-03-02T20:19:56.119

Thank You

Any questions?

Team Contribution

EDA : Wisdom Chen, Boya Zeng, Jessy Hu

Customer Segment: Jessy Hu, Yuqing Wu

Recommendation System: Wisdom Chen

Market Basket Analysis: Yuqing Wu, Boya Zeng

Recommendation System Video Demo: Boya Zeng

Slides: All





Appendix



Dashboard Demo Video

Recommendation Dashboard

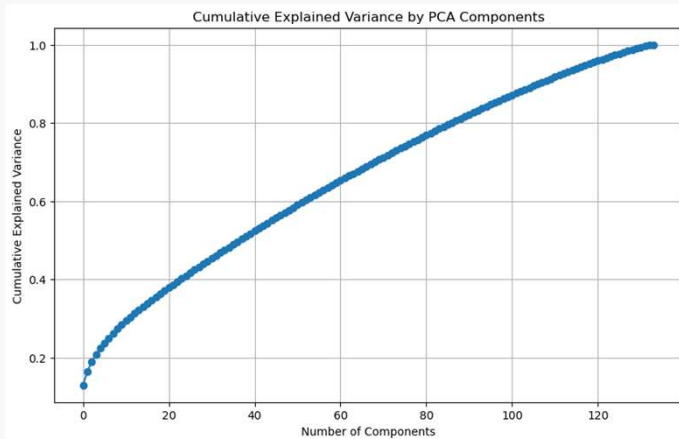


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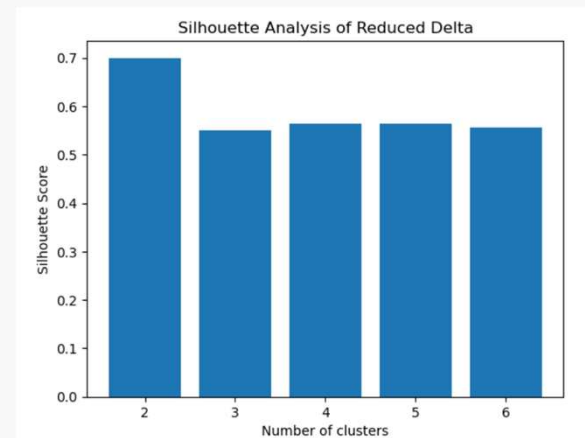
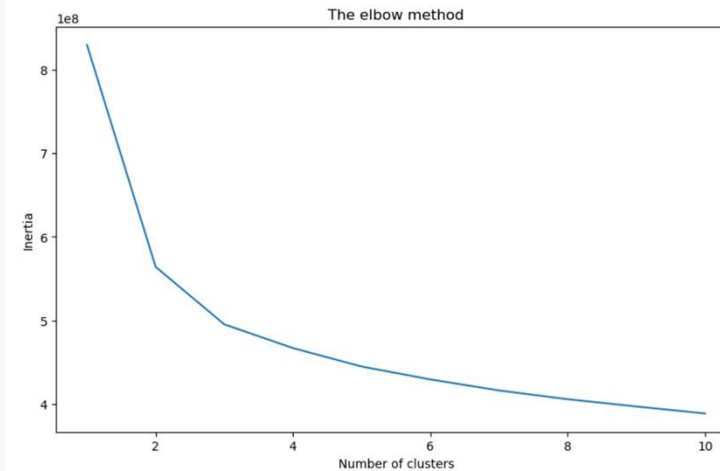


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PCA & Elbow Method & Silhouette Score Analysis



Selecting the number of components=78 to explain at least 75% of the variance



We will choose $k=4$ here. although $k=2$ seems to be the optimal, but there will be less information. $k=5$ will have the second largest silhouette score, but after we plot it out, we see one of the cluster is too small to compare. therefore, we ultimately chose $k=4$