# Customer Segmentation Report

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## 1. Collecting and Preparing Data

- Loading Data: We started by loading four CSV files: orders, products, order\_products, and aisle.
  - Orders Data: Contains information about each order.
  - Products Data: Lists details about each product.
  - Order-Products Data: Shows which products were in each order.
  - Aisle Data: Groups products by aisle (category).

## 2. Merging the Data Together

- Combining Datasets: Merged these files to create one complete dataset.
  - Merging Orders and Order-Products: Linked them by order\_id so each order shows the products in it.
  - Adding Products Data: Merged with the products list on product\_id to include product details.

### 3. Sampling Data for Easier Analysis

- Stratified Sampling: Selected a sample that represents all customer types, using 99% of the data to keep it accurate.
  - Sample Size: We kept the sample large to cover most customer behaviors.

## 4. Adding Aisle Names

- Mapping Aisle IDs to Names: Converted the aisle data into a dictionary (list of pairs) to make aisle names easier to read.
  - Aisle Mapping: Replaced aisle\_id with the actual aisle name for better understanding.

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#### 5. Cross-Tabulating Users and Aisles

- Creating a User-Aisle Table: Created a table to see how often each user buys from each aisle.
  - Row Normalization: Adjusted each row to compare user data more easily.
  - User-Aisle Matrix: Shows each user's shopping habits across different aisles.

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### 6. Reducing Dimensions with PCA (Principal Component Analysis)

- Simplifying Data: Used PCA to reduce the number of variables and keep the most important information.
- Normalization: Standardized values so they are all on a similar scale.
- Choosing Key Components: Selected components that capture the most important data patterns.

## 7. Finding Clusters with K-Means

- K-Means Clustering: Used K-means to group customers into segments based on similar buying patterns.
  - Elbow Method: Chose the best number of clusters by looking at a graph of within-cluster variation.
  - Optimal Cluster Number: Found the right number of clusters that balanced detail with simplicity.

## 8. Visualizing Clusters

- Plotting Clusters: Made a scatter plot of clusters using the first two principal components to show group differences.
  - Understanding Cluster Differences: Noted how different clusters stand out, helping us understand each group's buying habits.

### 9. Finding Popular Products in Each Cluster

- Top Products per Cluster: Listed the most popular products for each customer group.
  - Popular Product Analysis: Showed common products in each cluster to identify unique preferences.
  - Bundle Suggestions: Suggested product bundles based on popular items within each group.

### 10. Final Insights and Recommendations

- Customer Insights: Summarized key buying behaviors and patterns for each group.
- Purchase Patterns: Highlighted trends like when customers shop most and which products they often buy together.
- Product Bundling Ideas: Recommended bundles tailored to each group to boost sales and satisfaction