### Product Recommendation for Online Groceries

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## **Current Solutions**

#### **Online Groceries**

- Data provides insights on consumption behavior and patterns
- Customers lose opportunity to 'chance' upon products compared to traditional retail store
- Depends largely on specific searches
- How to best utilise screen real estate for quality recommendations?



#### Past Purchases

See All >

♠ You might be running low on 1 product

5D

Add To Cart

♣ Bought 1 month ago ★★★★★ (4341)



\$12.25

- Frozen

500 g

**Tropical Maria** 

Raspberry Fruits

Add To Cart

\$12.90



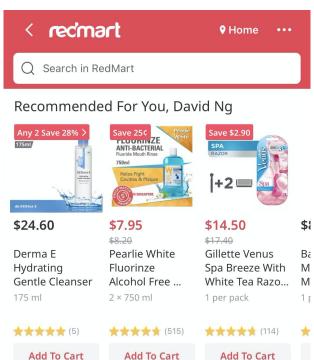




8D

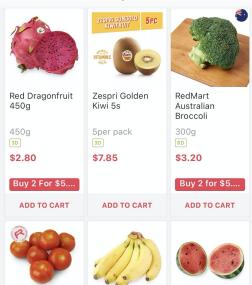
\*\*\*\* (2456)

Add To Cart





You may also like







1.8kg

Yay! You get free delivery.





**ADD TO CART** 

#### **Problems**

- No data on demographics on customers
- Without demographics, clustering becomes difficult
- Category managers unable to target specific groups within customer base
- Recommendations take on a very generic form
- Most recommendations come from the same category

### **Problem Statement**

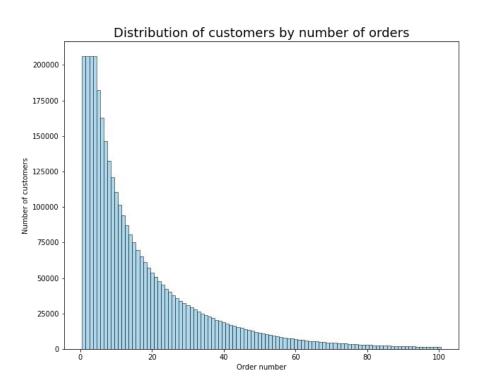
Without the flexibility of browsing physical store, e-grocers face difficulty in recommending **relevant** products to their customers, losing potential for **increased sales** and **customer retention**.

The objective is to develop a recommender system that leverages on consumption behavior to **group customers**. Based on their groups, relevant products based on **item association** will be recommended to the consumer.

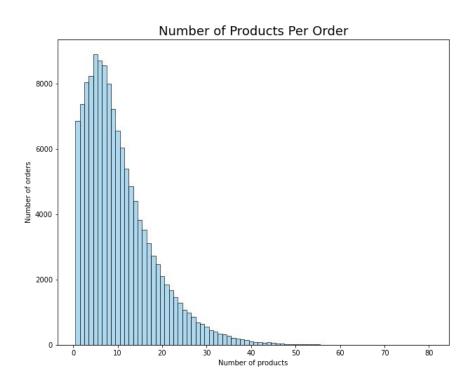
Success will be measured by the increase in **average order value**, and the **average number of new products** a customer checks out per order.

# **Data Highlights**

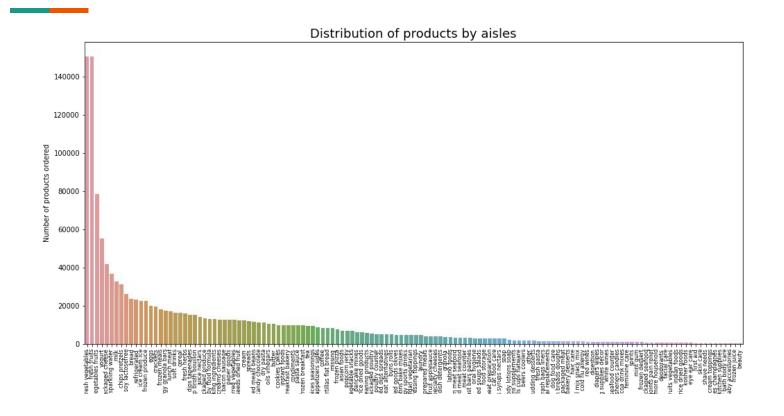
#### **Number of Orders**



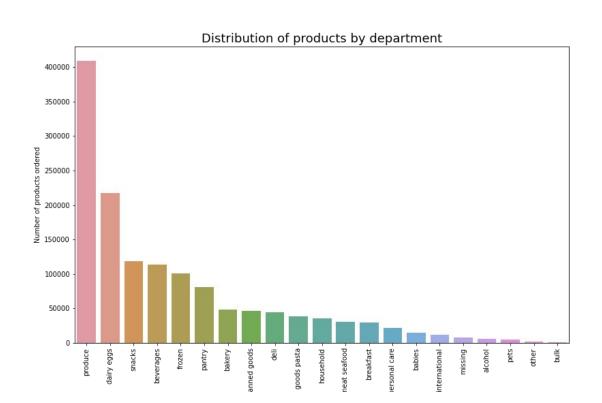
### **Items per Order**



#### **Aisle Distribution**



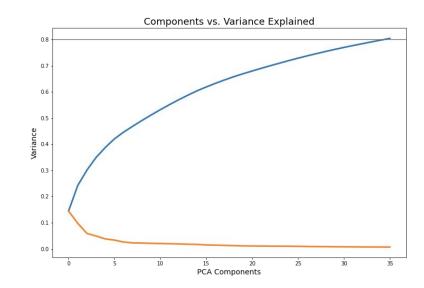
#### **Department Distribution**



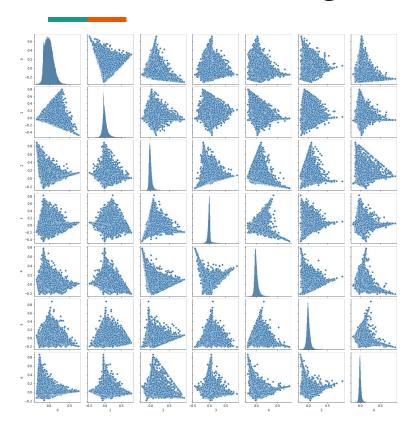
# Clustering

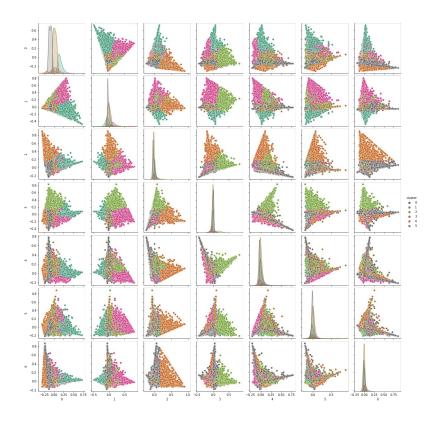
#### Principal Component Analysis (PCA)

- Data has total of 134 different aisles
- For each customer, the aisle weightage is computed
- Due to the high number of aisles, PCA was employed to help reduce the number of features
- PCA reduced features from 134 to 36, with
  ~80% of the variance explained



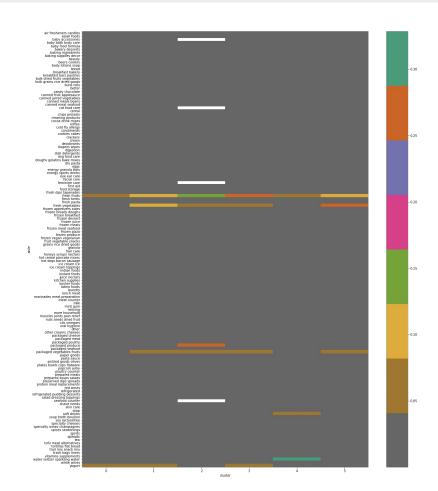
#### **K-Means Clustering**





#### Aisle Heatmap

- Cluster 0: Fresh fruits & Yogurt
- Cluster 1: Similar to 1 with additional preferences for vegetables
- Cluster 2: Baby accessories, cat food, feminine care, seafood
- Cluster 3: Particularly high for fresh fruits
- Cluster 4: Drinks and sparkling water
- Cluster 5: Fruits and vegetables



### **Recommender Basis**

#### **Association Rule Mining (ARM)**

- Dealing with implicit data (past purchases)
- ARM is more session based and caters for differing tastes in different periods
- Apriori algoritm
  - Suport: Probability of an item pair appearing
  - Confidence: Probability of a given item pair A & B given item A
  - Lift: Ratio of Confidence to Support

Recommendations are based on Lift. Lift values > 1 indicate positive correlation whereas < 1 indicates items being substitutes for each other.

#### **Association Rule Mining**

	item_A	item_B	product_name_A	product_name_B	freqAB	supportAB	freqA	supportA	freqB	supportB	confidenceAtoB	confidenceBtoA	lift
0	4985	36278	Homestyle Some Pulp Orange Juice	Chocolate Breakfast Biscuits Chocolate	27	0.000121	30	0.000134	32	0.000143	0.900000	0.843750	6281.437500
1	4985	48927	Homestyle Some Pulp Orange Juice	Fat Free Plain Yoghurt	24	0.000107	30	0.000134	62	0.000278	0.800000	0.387097	2881.806452
2	36278	48927	Chocolate Breakfast Biscuits Chocolate	Fat Free Plain Yoghurt	25	0.000112	32	0.000143	62	0.000278	0.781250	0.403226	2814.264113
3	24852	48927	Banana	Fat Free Plain Yoghurt	32	0.000143	50557	0.226368	62	0.000278	0.000633	0.516129	2.280045
4	38259	38576	Blueberry Greek Yogurt + Chia	The Epic Seed Greek Yogurt Strawberry & Chia	23	0.000103	33	0.000148	25	0.000112	0.696970	0.920000	6226.448485

## Recommender Results

Product added to cart: 9098 - Carrot Bunch. Customer is grouped in cluster: 0.

Product added to cart: 38548 - Gala Apple. Customer is grouped in cluster: 5.

	product_id	product_name
0	14168	Penne Pasta
1	49176	Cinnamon Roll Dough With Icing
2	29553	Super Soft Taco Flour Tortillas
3	48144	Tater Tots
	46041	Beef Franks

Product added to cart: 23087 - Organic Sweet Pea Sprouts. Customer is grouped in cluster: 0.

Product added to cart: 9874 - Organic Greek Yogurt Honey. Customer is grouped in cluster: 5.

	product_id	product_name
0	32734	Spaghetti
1	49176	Cinnamon Roll Dough With Icing
2	19508	Corn Tortillas
3	2452	Naturals Chicken Nuggets
4	46041	Beef Franks

product_name	oroduct_id	ŗ
Penne Rigate	12872	0
Organic Raspberries	27966	1
with Crispy Almonds Cerea	13042	2
Whole Wheat Tandoori Naar	7649	3
Cheese Pizza Snacks	31766	4

#### **Cluster Differentiation**

Product added to cart: 3235 - Sriracha Flavor Tortilla Chips. Customer is grouped in cluster: 1.

Product added to cart: 3235 - Sriracha Flavor Tortilla Chips. Customer is grouped in cluster: 2.

product_name	product_id	
Organic Good Seed Bread	2855	0
Trail Mix	37710	1
Organic Raisins	20448	2
Romaine Hearts	32689	3
Lowfat Kefir Smoothie Blueberry	20019	4

Product added to cart: 3235 - Sriracha Flavor Tortilla Chips. Customer is grouped in cluster: 3.

product_id	product_name
42701	Organic Sour Cream
19508	Corn Tortillas
33198	Sparkling Natural Mineral Water
12872	Penne Rigate
21267	Sourdough Bread
	42701 19508 33198 12872

	product_id	product_name
0	28199	Clementines, Bag
1	31651	Extra Fancy Unsalted Mixed Nuts
2	43889	Dark Chocolate Covered Banana
3	2855	Organic Good Seed Bread
4	19767	Old Fashioned Oatmeal

Product added to cart: 3235 - Sriracha Flavor Tortilla Chips. Customer is grouped in cluster: 5.

	product_id	product_name
0	21883	Irish Whiskey
1	19660	Spring Water
2	24489	Organic Whole Strawberries
3	21903	Organic Baby Spinach
4	46676	Total 0% Nonfat Greek Yogurt

# **Conclusion & Next Steps**

#### Conclusion

- Clustering seems to be effective in segregating customers according to their tastes, which is defined by the market share in the different aisles
- Recommender pushes products based on lift of item pairs, and goes further up to aisle lift if insufficient item lift pairs are available
- These products do not necessarily belong to the same category, allowing for pushing of cross category products and increasing exposure of products
- Clusters seem to be effective as customers from different aisles get recommended different products based on their cluster

#### **Next Steps**

- A/B testing required to evaluate effectiveness of recommender
- Metrics to be monitored:
  - Average Order Value
  - Average New Products Added
  - Product Recommendations Click Through Rate
  - Customer Churn

#### **Further Opportunities & Limitations**

- Currently, if insufficient items pairs with lift criteria are met, recommender resorts to pushing products from associated aisles
- These products can be prioritised by category managers (on the basis of product margin/promotions) in order to better influence click through rate of recommendations
- Additional information from customers such as order frequency and average order value can also be useful features in clustering
- Further breakdown dense aisles and combine loose aisles