# Grocery Recommendation System

Data from InstaCart

#### Data

21 Departments

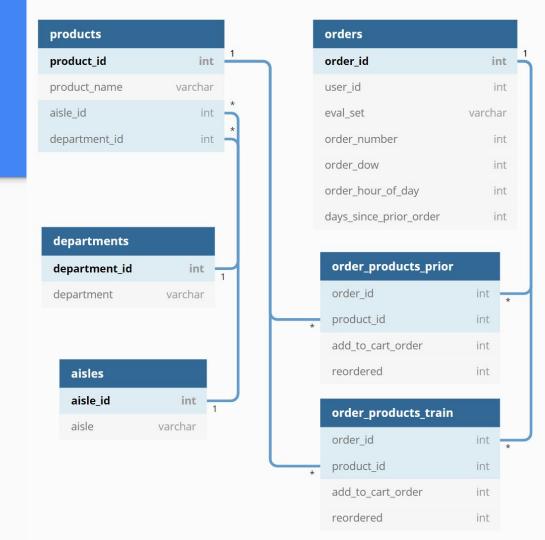
134 Aisles

49,688 Products

206,209 Users

3,421,083 Orders...

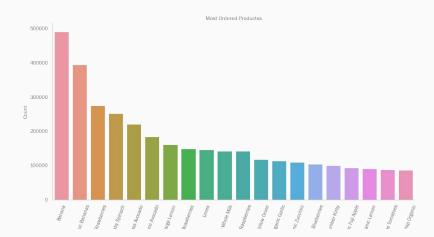
33,819,106 Ordered Products



#### Fun Food EDA

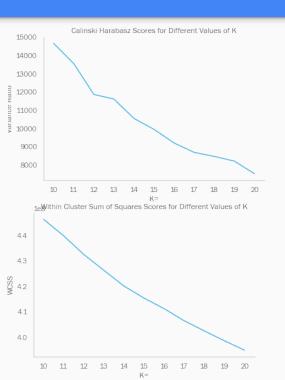
Most Ordered Product - Bananas

2nd Most Ordered Product - Organic Bag of Bananas



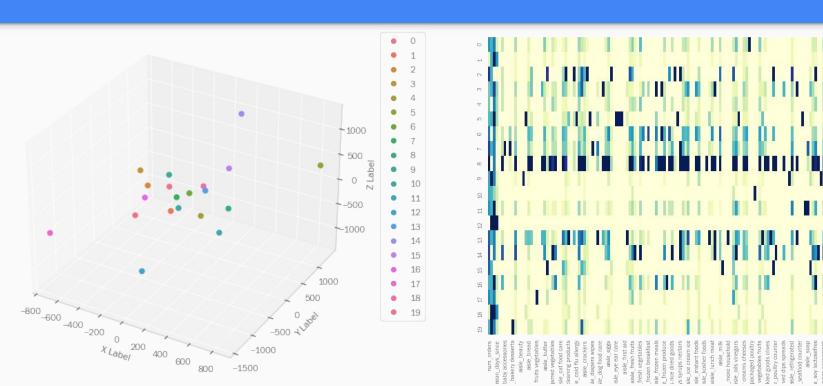


# **KMeans Clustering**



print(cluster\_metrics[5]) # Lots of personal care / pharmacy type products print(cluster\_metrics[7]) # Baby products print(cluster\_metrics[8]) # Lots of orders, shortest days between orders, big buyers print(cluster\_metrics[9]) # Alcohol purchasers print(cluster\_metrics[11]) # Soap and skin care print(cluster\_metrics[12]) # Very large cluster, with fewest number of orders and highest days between orders print(cluster\_metrics[13]) # Household, laundry, cleaning products print(cluster\_metrics[15]) # Chocolate, gum and soft drinks, least veggies print(cluster\_metrics[16]) # Vegan products and tofu

## Cluster visualizations



#### Metadata Recommendation

```
vectorize_products_based_on_metadata('Oreos')
   Out[276]: 22014
                                  Thin Mint Crisp Oreos
              23995
                       Halloween Oreos Sandwich Cookies
              Name: product name, dtype: object
           vectorize products based on metadata('Premium Almonds')
   Out[249]: 49178
                                                         Premium Almonds
              24511
                       Condoms, Premium Latex, Ultra Thin, Premium Lu...
              44962
                                                  Roasted Tamari Almonds
              7272
                                                  Yogurt Covered Almonds
              5597
                                                  Organic Tamari Almonds
              23466
                                              Platinum Premium Lubricant
              20405
                                                Roasted Unsalted Almonds
              21698
                       Pistachios, Premium Blend, Pomegranate, with A...
              18035
                                                     Premium Horseradish
              25923
                                           Premium Lubricant Condoms Enz
              Name: product_name, dtype: object
In [253]:
           ▶ vectorize products based on metadata('Red Potatoes')
   Out[253]: 13732
                                    Red Potatoes
              5651
                            Organic Red Potatoes
              3492
                               Baby Red Potatoes
              44892
                                      B Side Red
              10469
              47794
                                        Red Wine
              13706
                                   Essential Red
              6877
                                       Red Blend
              14259
                       Organic Red Potatoes, Bag
              4739
              Name: product name, dtype: object
           vectorize products based on metadata('randomword')
              No similar products found. Please refine your search terms and try again
```

**Natural Language Processing** 

Uses product names, aisle name, and department name with a Count Vectorizer and calculates cosine similarity.

Works like search engine optimization

Stemmed rather than lemmatized

Like to put stronger weight on nouns rather than adjectives

## Recommendation System

Too large to perform memory/neighborhood based models

Surprise SVD Model with grid search

User Id, Item Id, Number of times purchased (1-100)

RMSE 3.25, but high values seems far off

Currently scaling number of times purchased to a 1-5 rating scale

## Market Basket Analysis

Predict what a given user will order next. **Association rules** are normally written like this: **{Diapers} -> {Beer}** which means that there is a strong relationship between customers that purchased diapers and also purchased beer in the same transaction.

<u>Support</u>: Probability of buying X and Y products together: Support(X, Y) = Freq(X,Y)/N

<u>Confidence</u>: This says how likely item Y is purchased when item X is purchased.

Confidence(X, Y) = Freq(X,Y) / Freq(X)

<u>Lift</u>: Shows how likely item Y is purchased when item X is purchased, while controlling for how popular item Y is. **Lift = Support (X, Y) / (Support(X) \* Support(Y) )** 

#### More Work

Front end application?

**Jupyter Dash?** 

**SQL** tables