# Obligatory Blank Slide

# Starbucks Customer Segmentation

Focused Marketing Tactics

# Using Customer Behavior to Update Offer Portfolio

### **Transaction Data**

### Mining App Usage Data

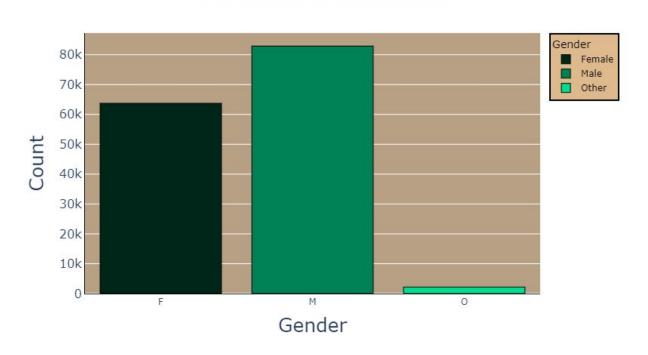
- Customer Profiles
- Offer Portfolio
- Transaction Details

# **Demographics**

**Distributions of Customer Data** 

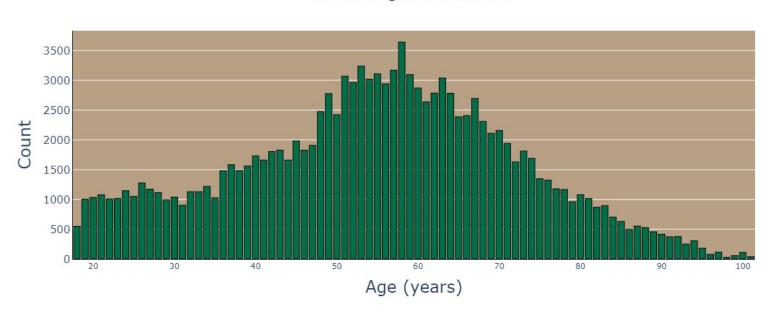
### Gender

#### Overall Gender Distribution



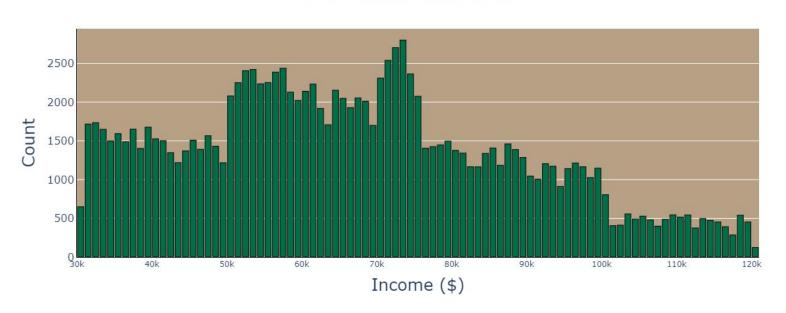


#### Overall Age Distribution



### Income

#### Overall Income Distribution



### **Spending**

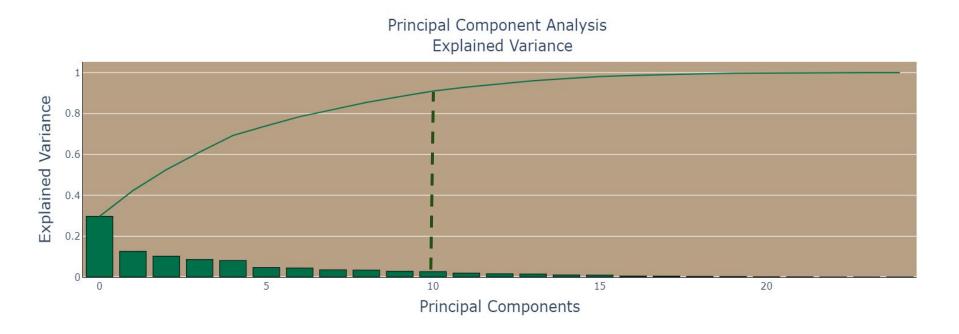




# Clustering

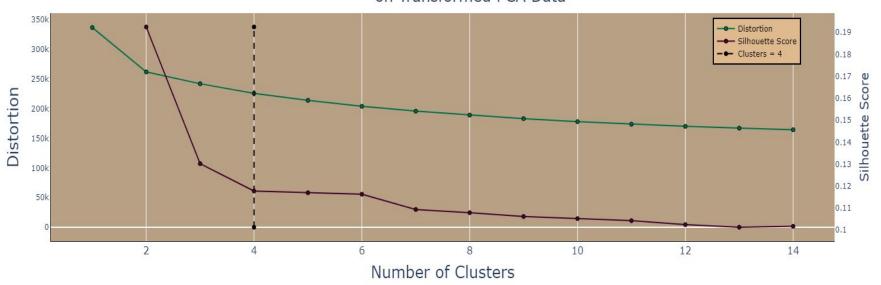
**Grouping Like Customers for Targeted Marketing** 

### **Dimensionality Reduction**



### **Clustering Heuristics**

Inertia and Silhouette Scores K-Means Fit on Transformed PCA Data

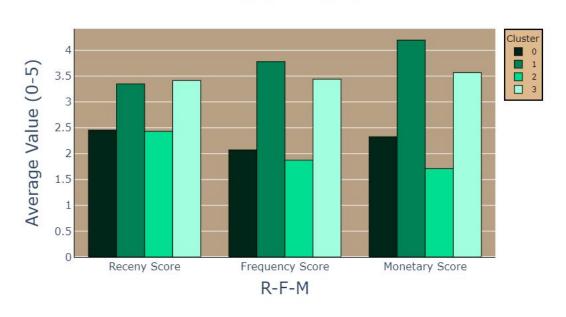


# Understanding Our Groups

What makes our clusters unique

### Recency - Frequency - Monetary

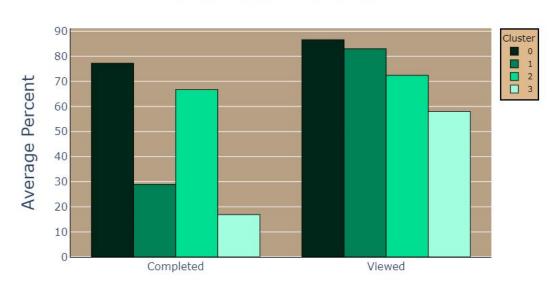
#### Average RFM Score



- Cluster 1 and 3 represent our best customers.
- Cluster 0 and 2 also show similarity



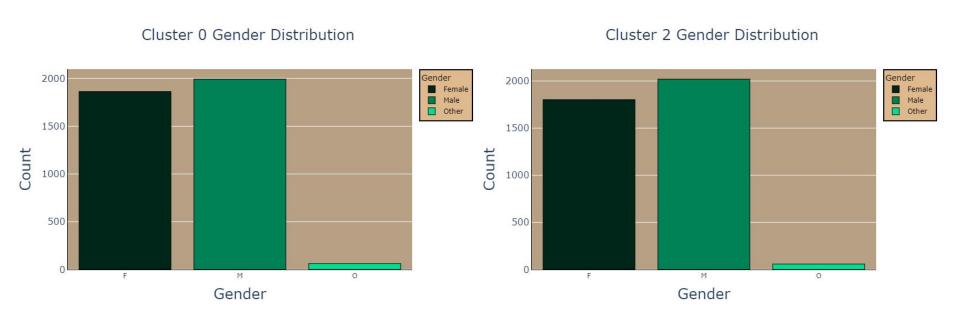
#### Overall Offer Engagement



- Cluster 0 and 1 represent our most engaged customers
- Cluster 1 views most offers but hardly complete any
- Cluster 3 views more than half of offers and completes only 18%



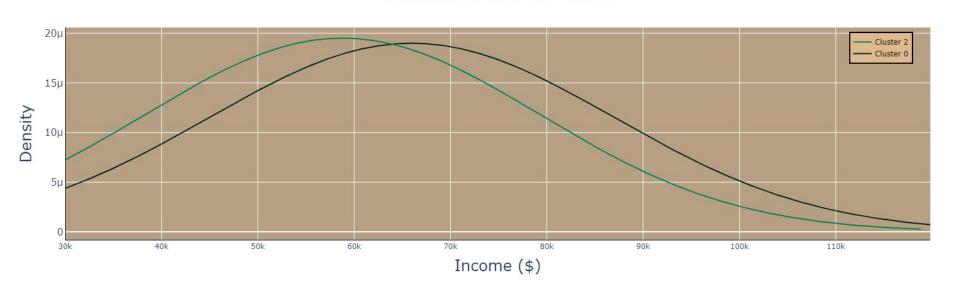
### Similarities of Clusters 0 and 2 (Gender)





### Differences of Clusters 0 and 2 (Income)

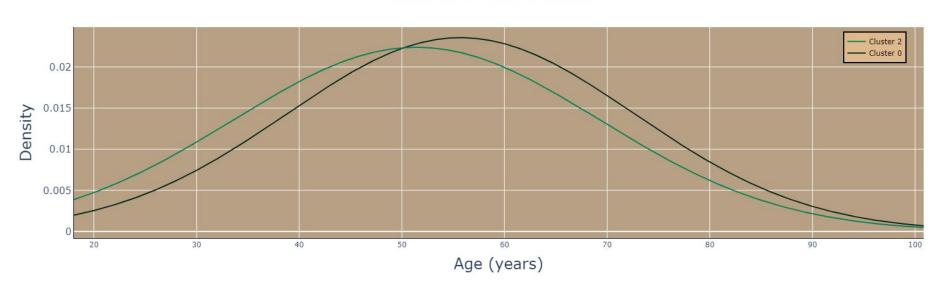
Distribution Income by Cluster





### Differences of Clusters 0 and 2 (Age)

#### Distribution Age by Cluster



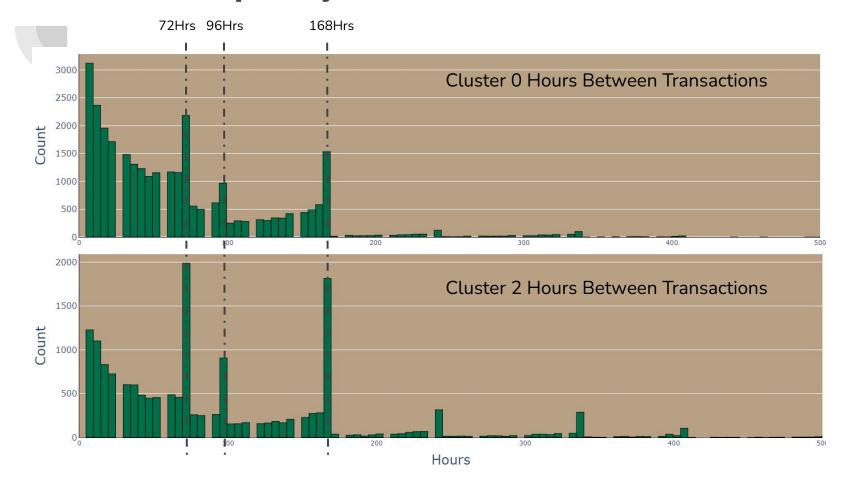


#### Differences of Clusters 0 and 2 (Revenue)





### **Transaction Frequency**



#### Recap: Who are clusters 0 and 2

#### Cluster 0

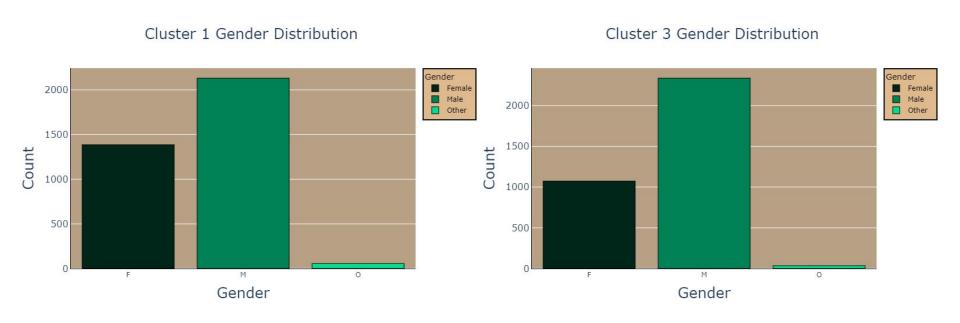
- Wealthy
- Older
- Low R.F.M.
- Low Spenders
- Females Over-Represented
- Engaged with App
- More than once per day buyers

#### Cluster 2

- Poorer
- Younger
- Low R.F.M.
- Lowest Spenders
- Females Over-Represented
- Engaged with App
- Every 3 or 7 day customers



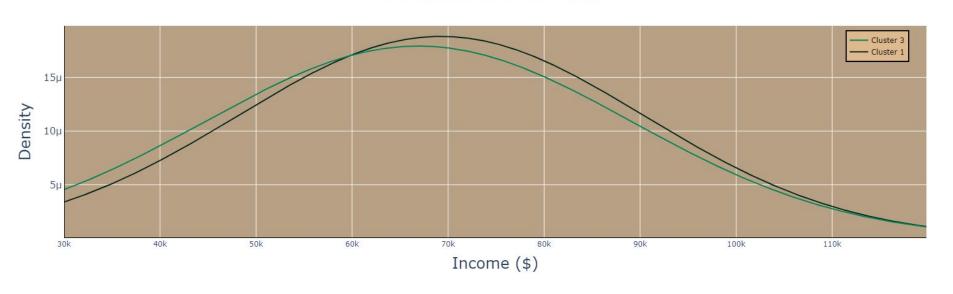
### Differences of Clusters 1 and 3 (Gender)





### Differences of Clusters 1 and 3 (Income)

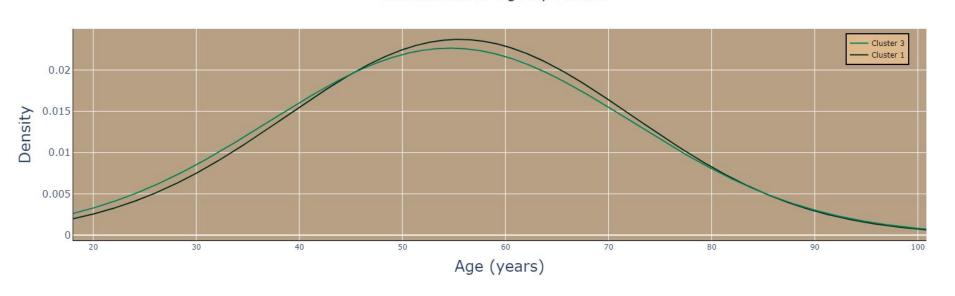
#### Distribution Income by Cluster





### Differences of Clusters 1 and 3 (Age)

#### Distribution of Age by Cluster



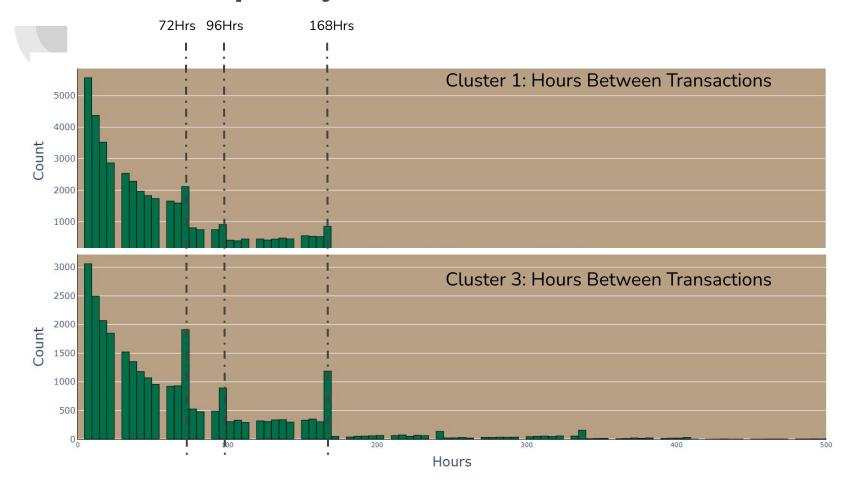


### Differences of Clusters 1 and 3 (Revenue)

#### Distribution Total Spent by Cluster



#### **Transaction Frequency**



#### Recap: Who are clusters 1 and 3

#### Cluster 1

- Moderate Income
- Median Age
- High R.F.M.
- Highest Spenders
- Males Over-Represented
- Not Engaged
- Highest Number of Purchases Per Day

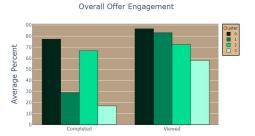
#### Cluster 3

- Moderate Income
- Median Age
- High R.F.M.
- High Spenders
- Males Over-Represented
- Not Engaged
- More than once per day buyers

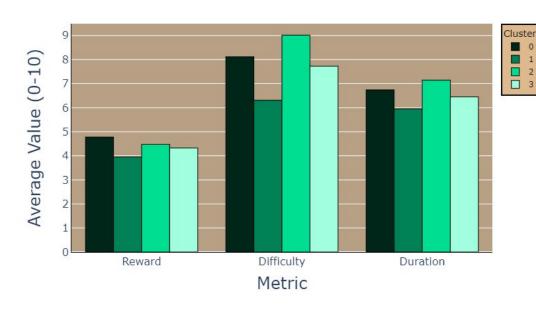
# How to Improve

**Targeted Marketing** 

### Reward - Difficulty - Duration



#### Overall Reward/Difficulty/Duration Distribution



- Reward and difficulty metrics scaled 0-10
- Duration is number of days the offer is valid
- Plot represents the average of all offers sent

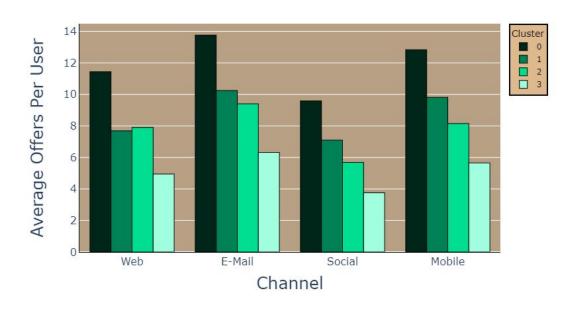
### Cluster 1 and 3 Offer Types







#### Overall Offer Channel Distribution



- Overall consistent trend with existing offers
- Cluster 2 received more web and email offers than the trend
- Social media offers are the least visible

## **Proposed New Offers**

| Cluster | Reward | Difficulty | Duration | Туре     | Web | Email | Social | Mobile |
|---------|--------|------------|----------|----------|-----|-------|--------|--------|
| 0       | 4      | 9          | 7        | BOGO     |     | Х     |        | Х      |
| 1       | 7      | 7          | 7        | Discount | X   | X     |        | X      |
| 2       | 5      | 9          | 3        | Discount | X   | X     |        | X      |
| 3       | 3      | 5          | 7        | Discount | X   |       | X      |        |

#### Proposed Follow-Ups

- Predict viewership and completion ratios for proposed offers and their impact on revenue
- Find more succinct ways of visualizing differences between clusters
- Use 1-4 instead of 0-3 for cluster naming, or create nicknames