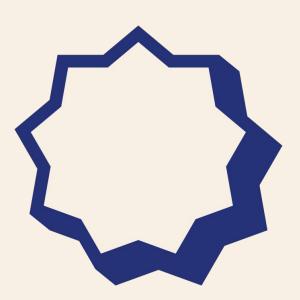


Dongkyu Kim | Marukha Hussaini | Natalie Lira | Sachin Pandya

# Project 4

# Objectives



# Develop an Unsupervised Model

We began by developing an unsupervised model to analyze restaurant data and to separate the values into two different clusters

# Standardize the data to allow for comparison

All data was scaled to the same power for ease of comparison and conclusion

#### **Future Goals**

Our Future Goals are to then tweak this model so that it can integrate into a restaurant review service like Yelp and produce groups of restaurants as "Go" and "Don't Go"

## Original Data

- The original dataset consists of 17 columns and 8368 rows.
- Includes both categorical and numerical values
- Requires standardization and cleaning of values for comparison

<cla< th=""><th>ss 'pandas.core.frame.Da</th><th>taFrame'&gt;</th><th></th></cla<>	ss 'pandas.core.frame.Da	taFrame'>	
Rang	eIndex: 8368 entries, 0	to 8367	
Data	columns (total 17 colum	ns):	
#	Column	Non-Null Count	Dtype
0	Name	8368 non-null	object
1	Location	8368 non-null	object
2	Cuisine	8368 non-null	object
3	Rating	8368 non-null	float64
4	Seating Capacity	8368 non-null	int64
5	Average Meal Price	8368 non-null	float64
6	Marketing Budget	8368 non-null	int64
7	Social Media Followers	8368 non-null	int64
8	Chef Experience Years	8368 non-null	int64
9	Number of Reviews	8368 non-null	int64
10	Avg Review Length	8368 non-null	float64
11	Ambience Score	8368 non-null	float64
12	Service Quality Score	8368 non-null	float64
13	Parking Availability	8368 non-null	object
14	Weekend Reservations	8368 non-null	int64
15	Weekday Reservations	8368 non-null	int64
16	Revenue	8368 non-null	float64
	es: float64(6), int64(7) ry usage: 1.1+ MB	, object(4)	

## Standardize the Data and

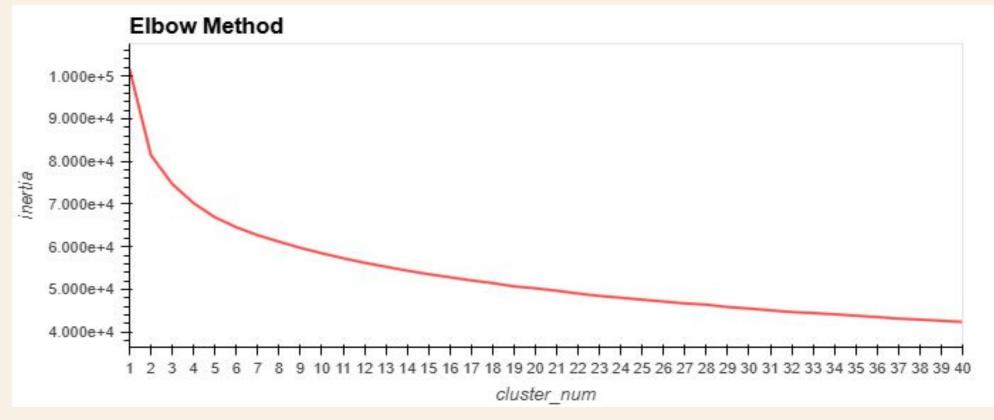
PCA

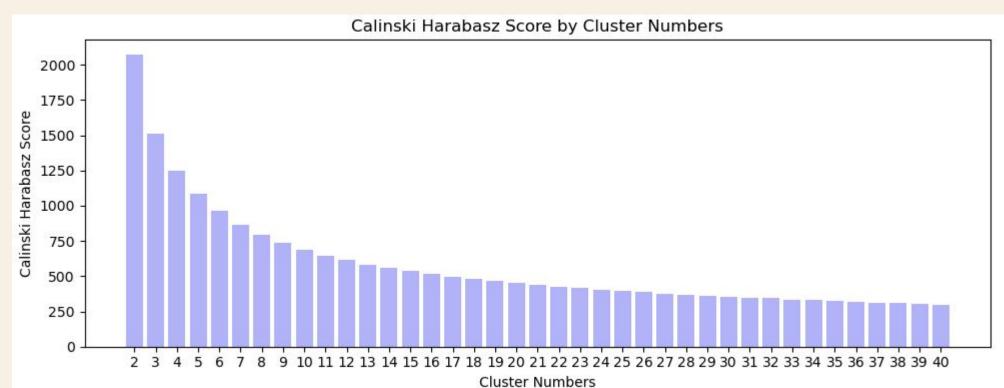
	Rating	Seating Capacity	Average Meal Price	Marketing Budget	Social Media Followers	Chef Experience Years	Number of Reviews	Avg Review Length	Ambience Score	Service Quality Score	Weekend Reservations	Weekday Reservations	Revenue
0	-0.014202	-1.276714	1.819441	-0.544861	-0.686274	0.534421	-1.219380	-0.178419	-1.639150	0.576566	-0.823590	-1.261571	-0.064043
1	-1.390099	0.907389	-1.380216	0.656375	0.351622	-0.371987	0.036038	-0.361285	-1.134352	-0.815332	0.924293	-1.161586	-0.620285
2	1.189708	-0.701950	0.027437	-0,231400	0.058746	1.440830	1.190445	-1.637931	-0,085926	0.460574	-0.124437	-0.761648	-0.428956
3	0.673746	-1.506619	0.254838	-1.124106	-1.126017	0.534421	-1.590955	0.425916	-0.357740	-1.047314	-1.023349	-0.611671	-0.940598
4	1.533682	1.597106	1.958951	0.230572	0.213665	-0.190705	-1.605385	0.929408	1.195484	-1.317961	0.374958	-0.161740	3.122598
		411		m		on .	""			***			
8363	-1.046125	-0.357091	-0.910068	-1.159727	-1.336227	0.171858	-0.515913	1.099394	1.544960	-0.196710	0.374958	-1.461540	-0.828044
8364	-0.530163	-0.644473	-0.768466	-0.674191	-0.845917	-0.190705	0.685392	0.011393	-1.095521	-1.124642	0.374958	-0.411702	-0.901626
8365	1.189708	1.597106	-0.071614	1,496473	1.489845	-0.734551	-0.313892	0.669276	-0.280079	-1.472616	2.672176	-0.411702	1.025907
8366	-1.562086	-1.679048	-0.234841	-1.376191	-1.557816	-1.640959	0.743112	0.051572	0.224719	-1.317961	-1.173167	-0.411702	-1.288631
8367	-0.014202	-1.564095	1.616454	-0,665971	-0.640002	-0.371987	-1.176090	-0.318526	0.147058	0.769885	-1.223107	-0.861632	-0.455978

Location_Rural	Location_Suburban	Cuisine_American	Cuisine_French	Cuisine_Indian	Cuisine_Italian	Cuisine_Japanese	Cuisine_Mexican	Parking Availability_No	Parking Availability_Yes
True	False	False	False	False	False	True	False	False	True
False	False	False	False	False	False	False	True	False	True
True	False	False	False	False	True	False	False	True	False
True	False	False	False	False	True	False	False	False	True
False	False	False	False	False	False	True	False	True	False
False	True	False	False	True	False	False	False	False	True
True	False	False	False	True	False	False	False	True	False
False	False	False	False	False	True	False	False	False	True
True	False	True	False	False	False	False	False	True	False
True	False	False	False	False	False	True	False	False	True

	PCA_1	PCA_2	PCA_3	PCA_4	PCA_5	PCA_6	PCA_7	PCA_8	PCA_9
0	-1.602329	-1.717388	1.636562	-1.393867	-1.705817	-0.210651	0.097664	0.326914	0.408310
1	0.441447	1.358972	-1.409816	-0.710771	-0.172635	0.511681	0.809583	-0.487749	1.704060
2	-0.779930	0.468890	1.343955	-0.353624	0.538145	0.535154	0.998055	2.139401	0.279931
3	-2.535516	-0.174292	1.015148	-1.686116	-0.077267	-0.640989	-0.547675	-0.862953	-0.334777
4	2.860161	-2.526592	1.211125	-0.760200	0.943812	-0.518266	-1.510619	-1.638053	0.214561
8363	-2.168573	0.060189	-1.092439	0.477991	1.108680	-1.067013	-1.265438	-0.399140	1.338251
8364	-1.677503	0.512871	-0.794318	-0.501850	0.174091	0.291397	1.426922	-0.668104	0.547127
8365	3.380806	0.677995	0.008023	-0.402776	0.372431	0.014191	0.324511	-1.757077	2.050849
8366	-3,407766	-0.334320	-0.686713	0.519816	0.756618	1.255804	0.432478	-1.461543	-0.400274
8367	-1.962954	-1.389461	1.659146	-0.247970	-0.900507	0.429666	-1.181607	0.155208	-0.164609

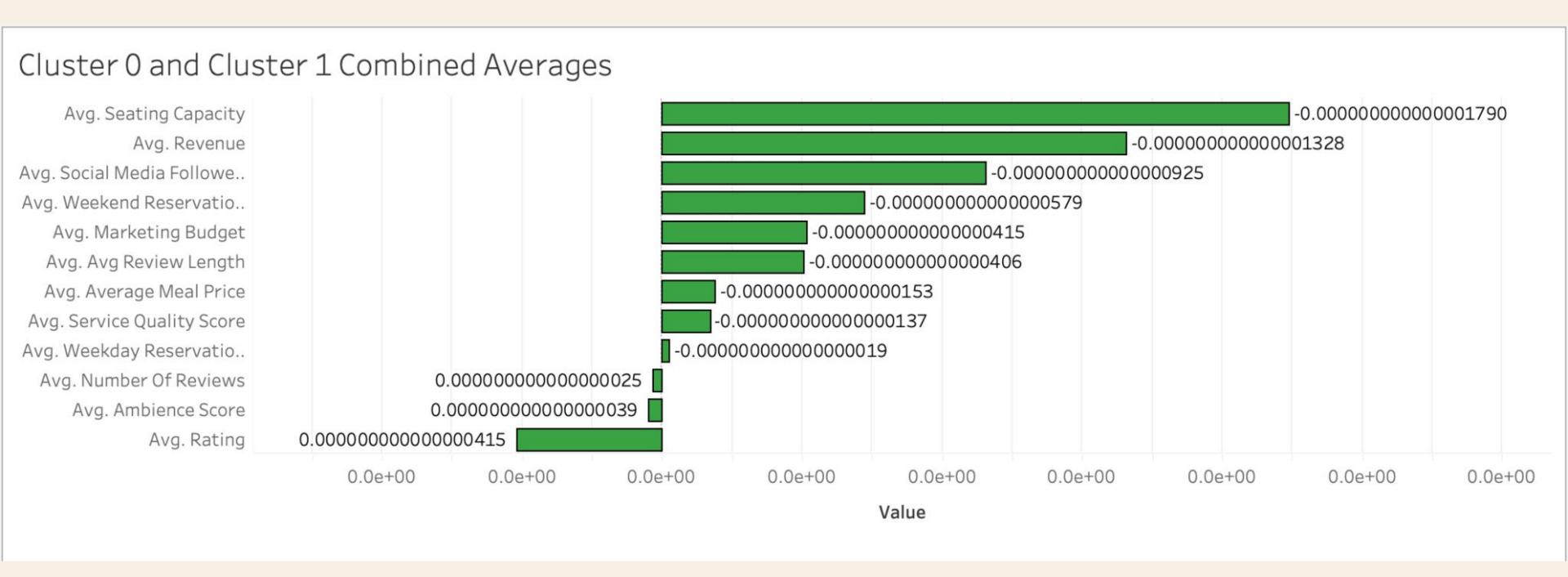
# Determining the Number of Clusters & K-means Clustering



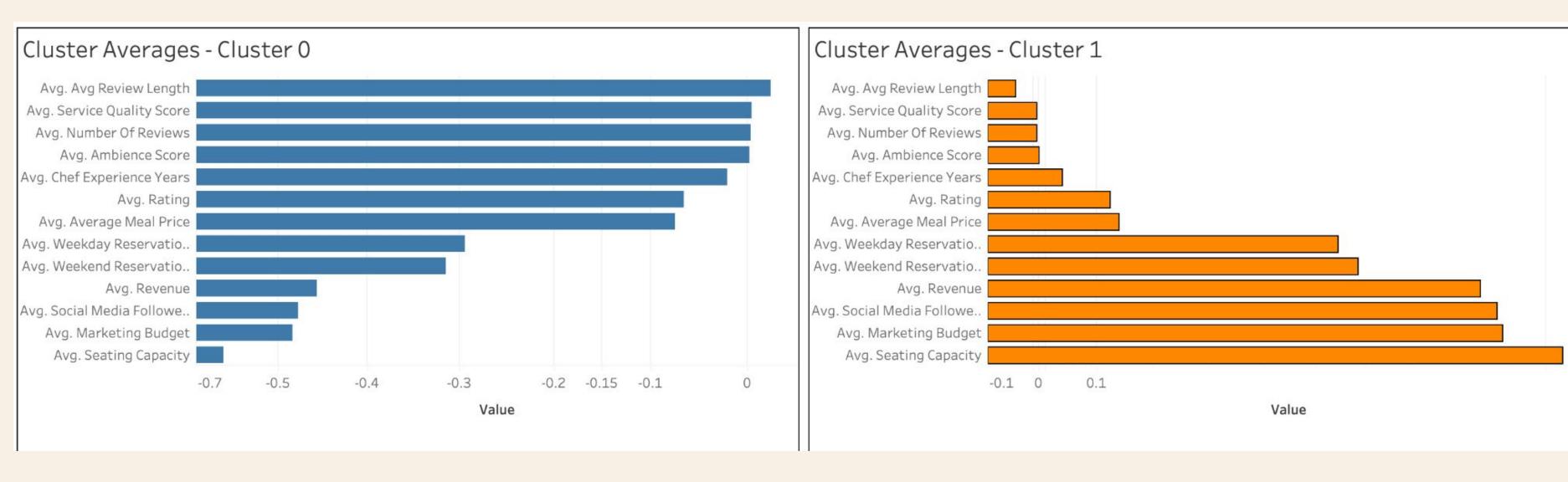


	PCA_1	PCA_2	PCA_3	PCA_4	PCA_5	PCA_6	PCA_7	PCA_8	PCA_9	Cluster
0	-1.602329	-1.717388	1.636562	-1.393867	-1.705817	-0.210651	0.097664	0.326914	0.408310	0
1	0.441447	1.358972	-1.409816	-0.710771	-0.172635	0.511681	0.809583	-0.487749	1.704060	1
2	-0.779930	0.468890	1.343955	-0.353624	0.538145	0.535154	0.998055	2.139401	0.279931	0
3	-2.535516	-0.174292	1.015148	-1.686116	-0.077267	-0.640989	-0.547675	-0.862953	-0.334777	0
4	2.860161	-2.526592	1.211125	-0.760200	0.943812	-0.518266	-1.510619	-1.638053	0.214561	1
			***	***	***	***			***	
8363	-2.168573	0.060189	-1.092439	0.477991	1.108680	-1.067013	-1.265438	-0.399140	1.338251	0
8364	-1.677503	0.512871	-0.794318	-0.501850	0.174091	0.291397	1.426922	-0.668104	0.547127	0
8365	3.380806	0.677995	0.008023	-0.402776	0.372431	0.014191	0.324511	-1.757077	2.050849	1
8366	-3.407766	-0.334320	-0.686713	0.519816	0.756618	1.255804	0.432478	-1.461543	-0.400274	0
8367	-1.962954	-1,389461	1.659146	-0.247970	-0.900507	0.429666	-1.181607	0.155208	-0.164609	0

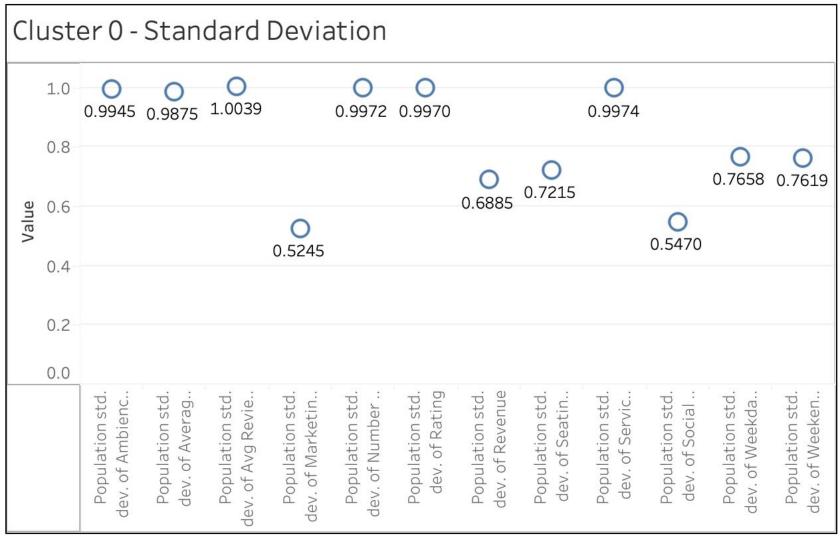
### Cluster Averages - Combined



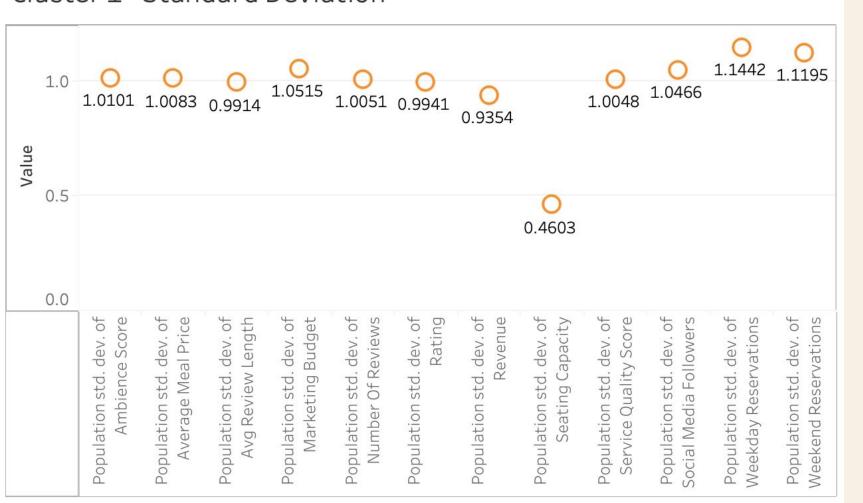
### Cluster Averages - Separated



 With the standardized data, the average values for each of the reported items are complementary to each other and help create a visual for how the model selected between the two clusters



#### Cluster 1 - Standard Deviation



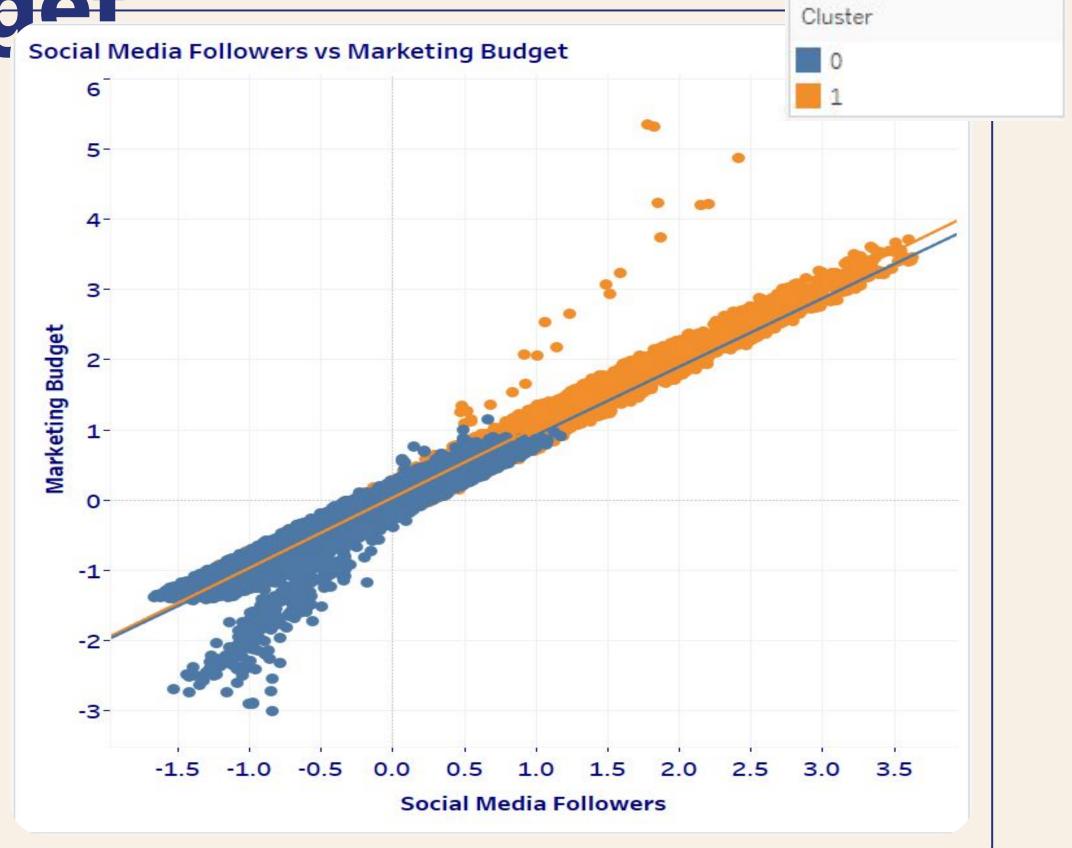
### **Standard Deviations**

- Between the two clusters we see more variability in the SD of the measures for Cluster 0 versus Cluster 1
- Cluster 1 overall has a more consistent, but larger, standard deviation for each measure
- In large, it appears that Cluster 1 has more "normal" distribution of the data and may help to indicate that these are the restaurants that should be included in the "Go" group.

VS

# Social Media Marketing Budget

- Both clusters show a positive relationship between Social Media Followers and Marketing Budget.
- There are indicators that businesses investing more in marketing efforts, tend to gain more followers on social media.
- While both clusters show similar results, *Cluster 1* may reflect more substantial investments in marketing, contributing to an even greater increase in social media followers compared to *Cluster 0*.



Followers

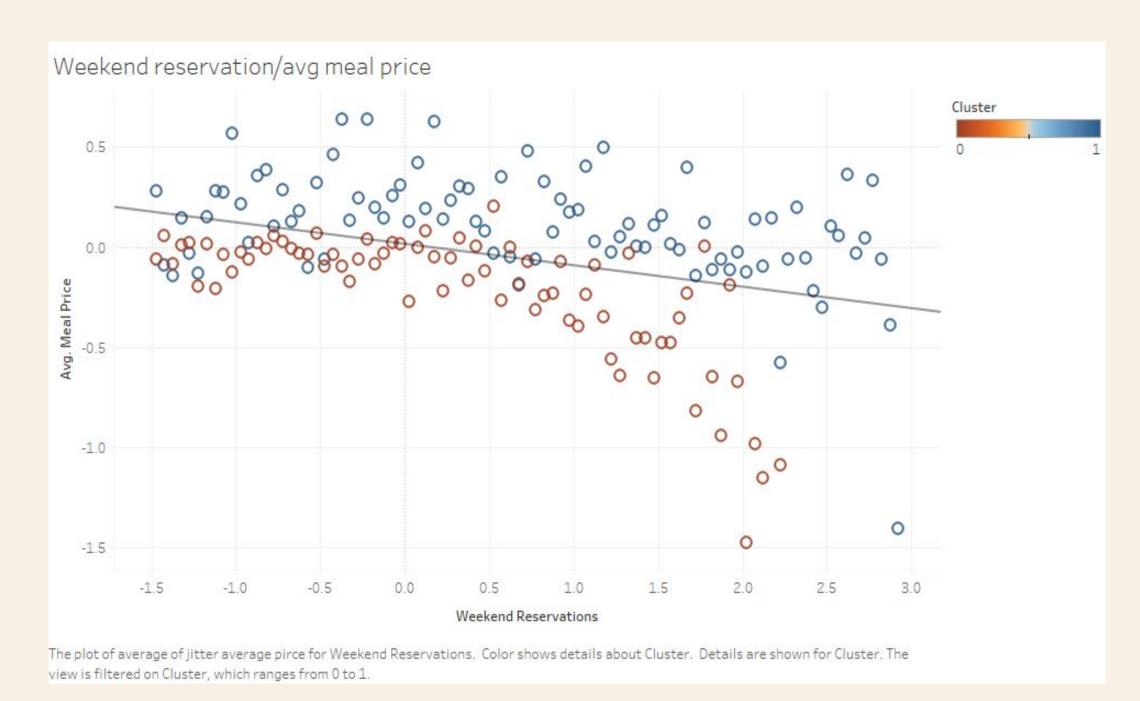
### Marketing Budget vs Revenue

- Both clusters show a negative relationship between Marketing Budget and Revenue.
- As Marketing Budget increases, Revenue tends to decrease for both clusters, but the rate of decrease varies between them.
- Businesses in Cluster 0 may need to reassess their marketing strategies, as their higher marketing budget is not yielding revenue gains to match.
- Cluster 1 shows a more moderate
  decline in Revenue with increasing
  Marketing Budget, suggesting that
  marketing efforts might be more
  effective or have diminishing returns
  at a slower rate.



# Weekend Reservations vs Avg.Meal Price

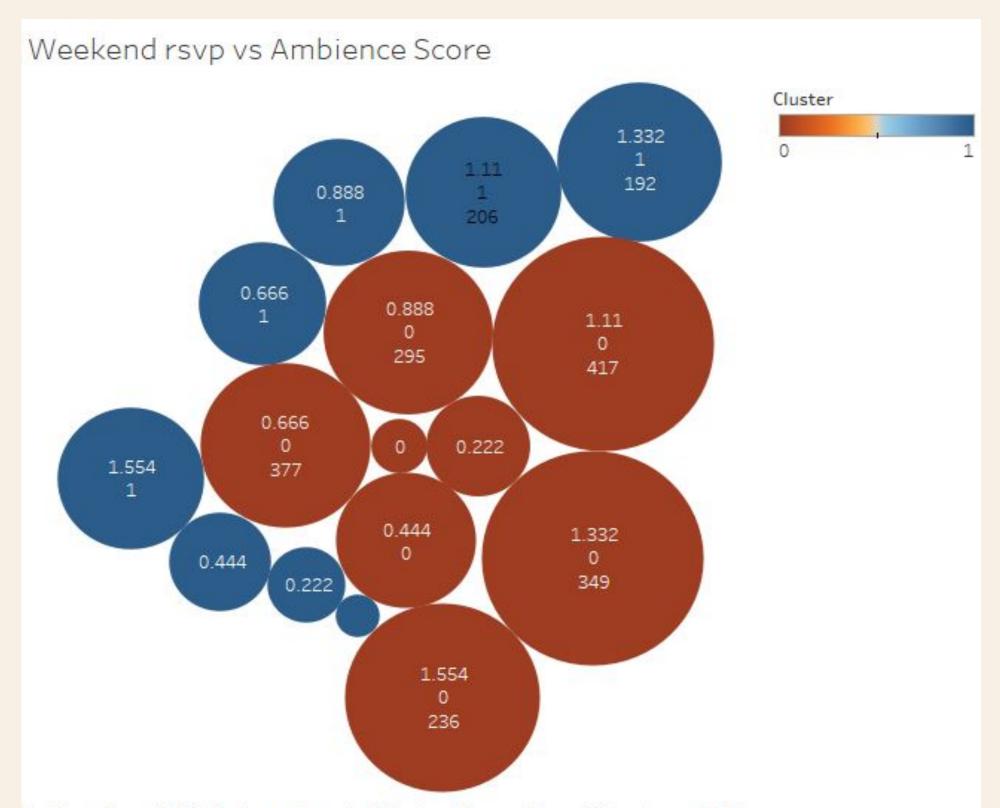
- Clusters shows negative relationships between Weekend Reservations and avg.Meal Price
- Avg Meal Price decrease for both clusters, Weekend RSVP increase
- Cluster 0 experiences a sharper decline in Average Meal Price as Weekend Reservations increase, indicating that businesses in this cluster may need to reassess their pricing strategies, as more reservations are correlating with lower prices rather than higher ones.
- Cluster 1 gradual decrease in Average Meal Price as Weekend Reservations rise, suggesting these businesses have a more stable pricing strategy that maintains meal prices despite more reservations.



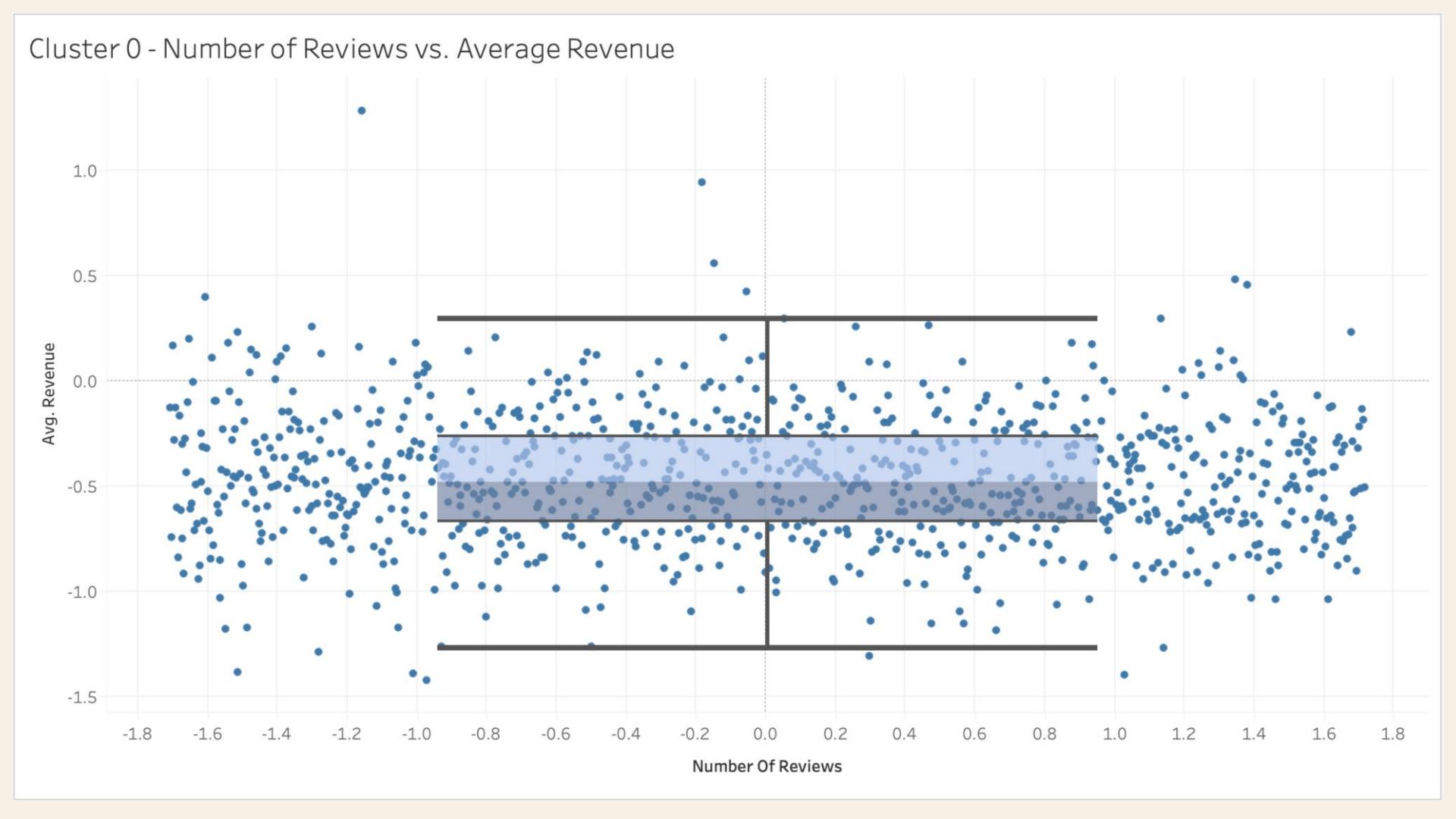
### Weekend Reservations vs Ambience

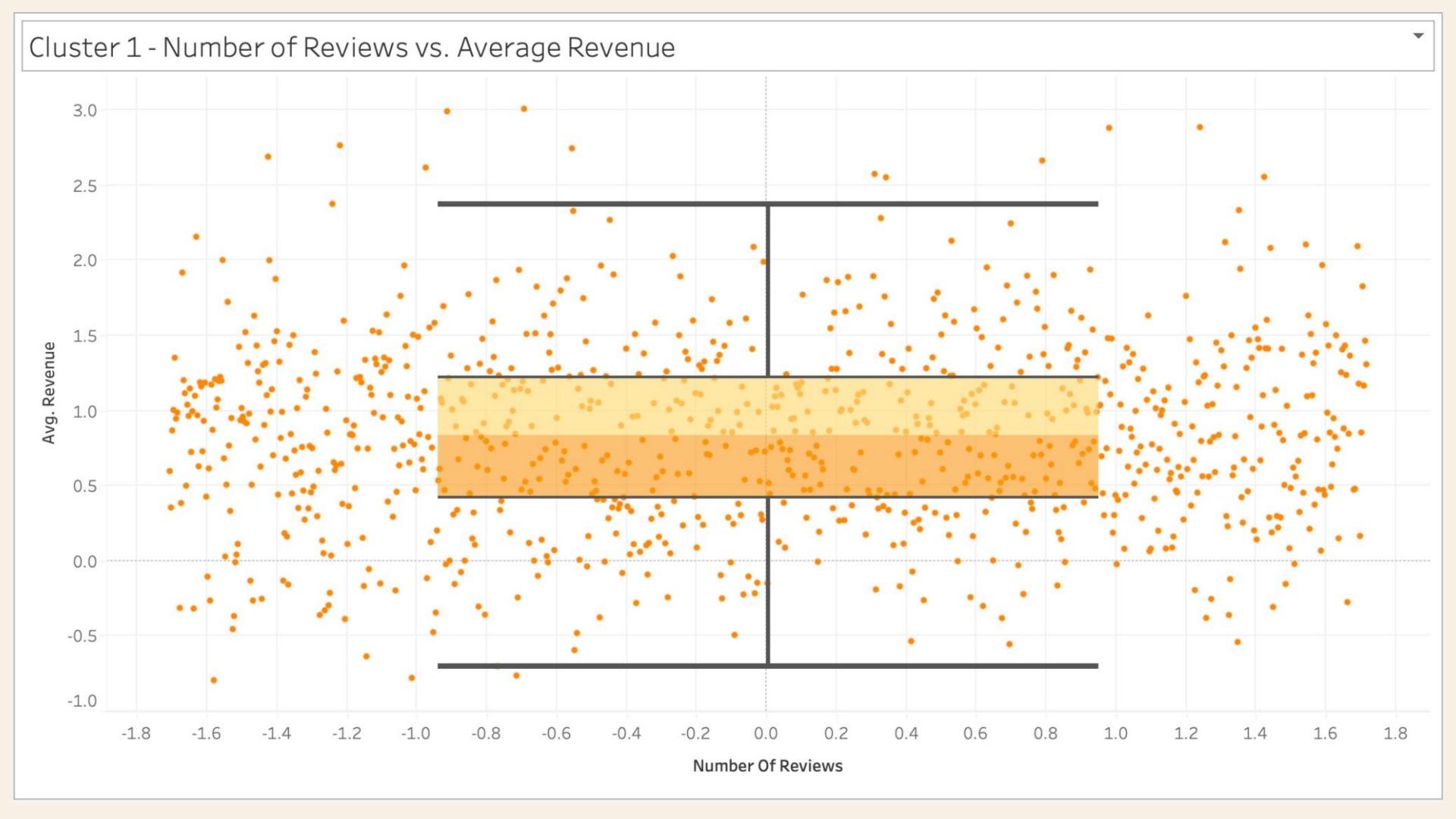
#### Score

- The chart shows the relationship between Weekend Reservations and Ambience Score for two clusters
- Larger bubbles indicated more Weekend RSVP
- Cluster 0 (Red) tends to have higher
   Ambience Scores and more
   Weekend Reservations
- Cluster 1 (blue) has lower Ambience scores and fewer Weekend Reservations.
- The bubble chart suggests that business with higher Ambience scores tend to have more Weekend reservations.
   However, the relationship is not necessarily linear, and there are variations within each cluster.



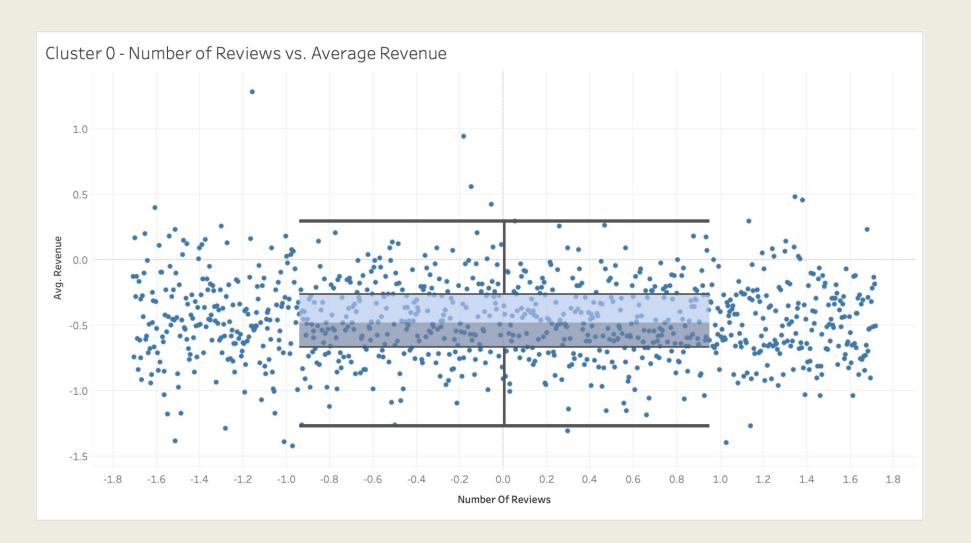
Ambience Score (bin), Cluster and count of Weekend Reservations. Color shows details about Cluster. Size shows sum of Ambience Score. The marks are labeled by Ambience Score (bin), Cluster and count of Weekend Reservations. The view is filtered on Cluster, which ranges from 0 to 1.





### Number of Reviews vs. Average

- Reache late average revenue.
  - The major difference between the clusters is the total revenue as a whole with Cluster 0 having a lower median value.
  - Cluster 1 shows a larger standard deviation and more outliers than Cluster 0 at a higher median value





## Questions?

### Resources

#### **Dataset:**

https://www.kaggle.com/datasets/anthonytherrien/restaurant-revenue-prediction-dataset/data







