



# Will My Restaurant Close?

**An inside look at the power of prediction with Yelp reviews**

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## Problem Statement

The restaurant industry is extremely competitive with 80% of restaurants closing within their first 5 years of operation.

Looking solely at Yelp reviews, does the text have any predictive power into whether or not a restaurant's operating status is open or closed?



## Audience

A group of investors, with a semi-technical background, approached our team and asked for help in performing data analysis.

- Provided our team with a Yelp dataset
- Asked us to look into relationship between Yelp reviews and restaurant's operating status.



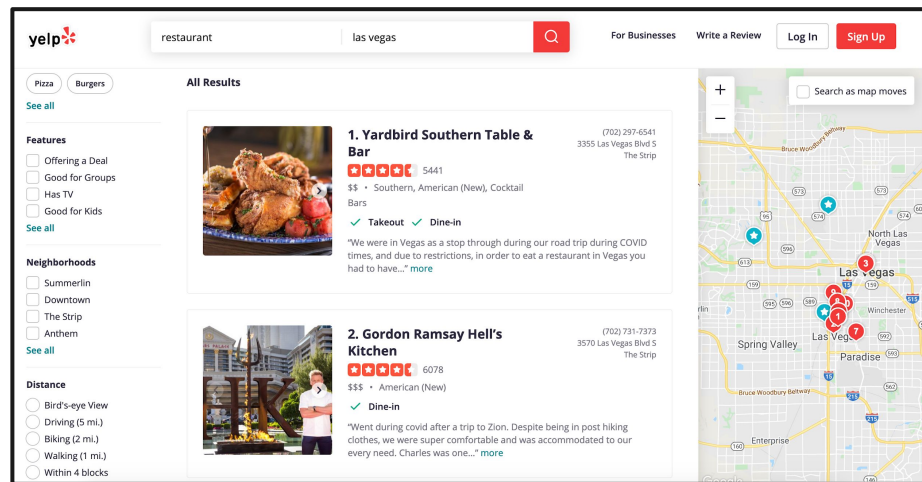


# Agenda

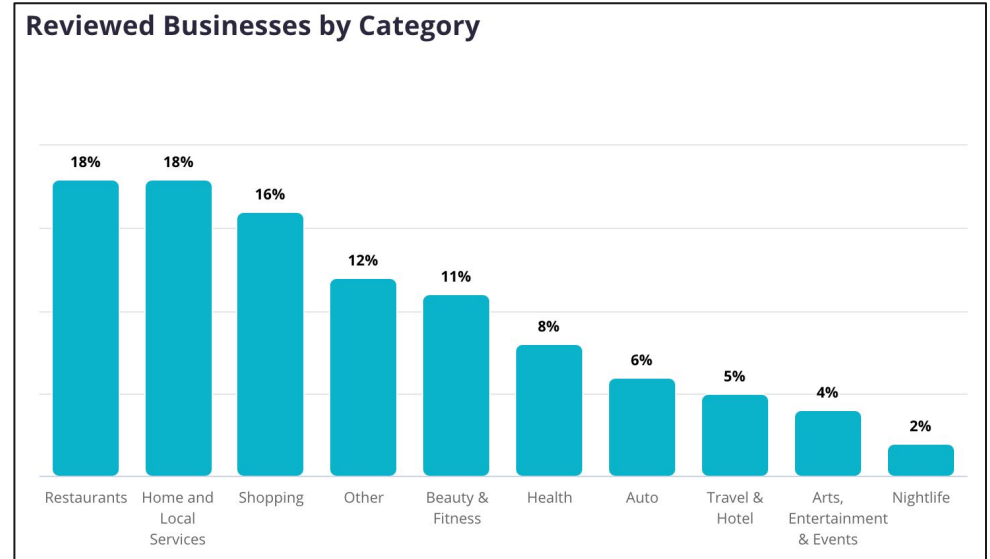
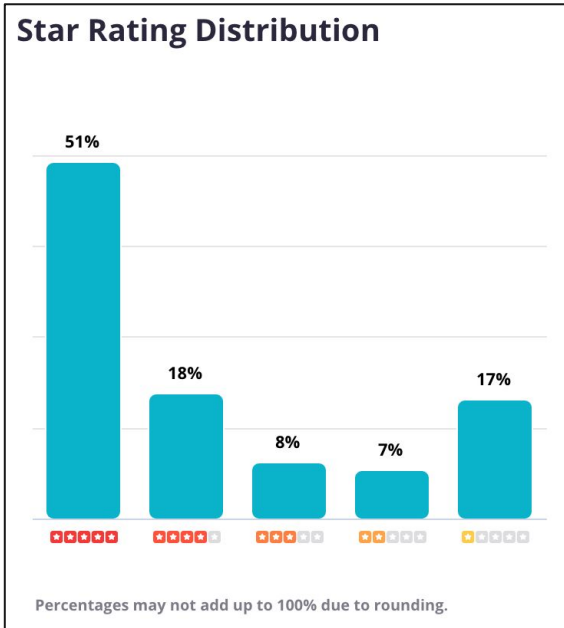
1. Introduction on Yelp
2. Data Collection
3. Exploratory Data Analysis (EDA)
4. Modeling
5. Conclusions / Recommendations
6. Future Areas of Focus
7. Questions

# Introduction on Yelp

“Yelp connects people with great local businesses. With unmatched local business information, photos and review content, Yelp provides a one-stop local platform for consumers to discover, connect and transact with local businesses of all sizes by making it easy to request a quote, join a waitlist, and make a reservation, appointment or purchase.”

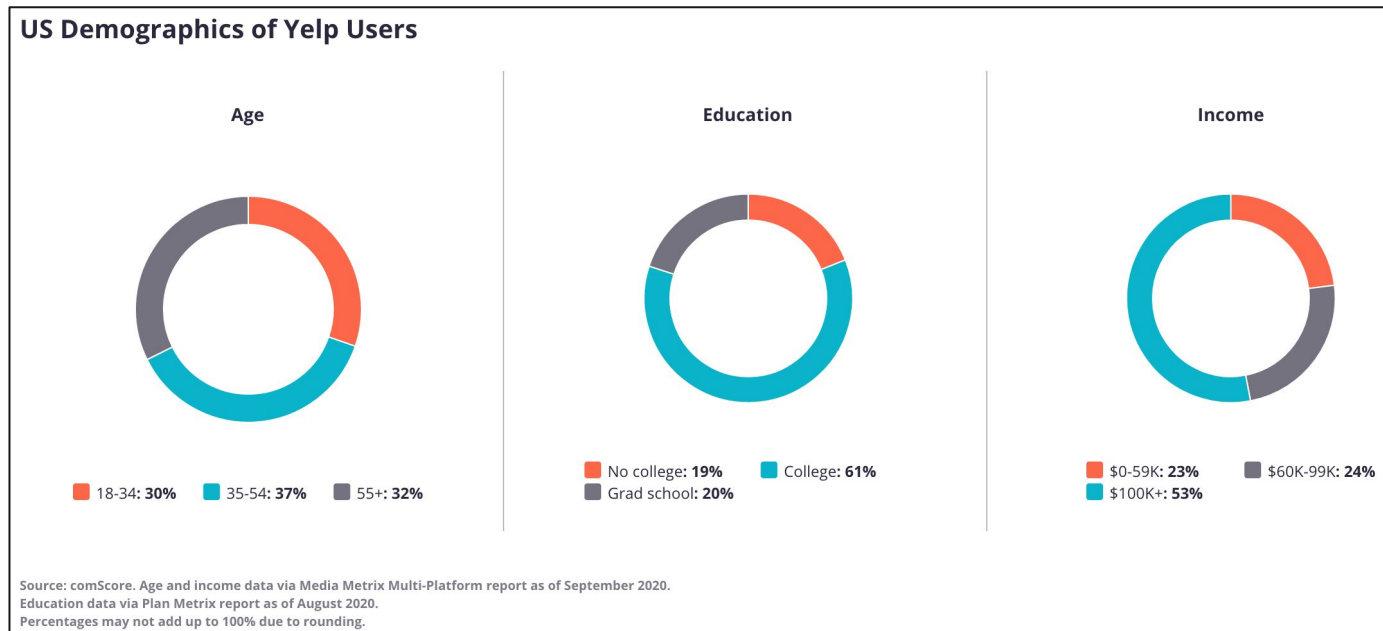


## Introduction on Yelp (cont.)



Source: <https://www.yelp-press.com/company/fast-facts/default.aspx>

## Introduction on Yelp (cont.)



Source: <https://www.yelp-press.com/company/fast-facts/default.aspx>

# Data Collection

## The Dataset



**8,021,122 reviews**



**209,393 businesses**



**200,000 pictures**



**10 metropolitan areas**

1,320,761 tips by 1,968,703 users

Over 1.4 million business attributes like hours, parking, availability, and ambience

Aggregated check-ins over time for each of the 209,393 businesses

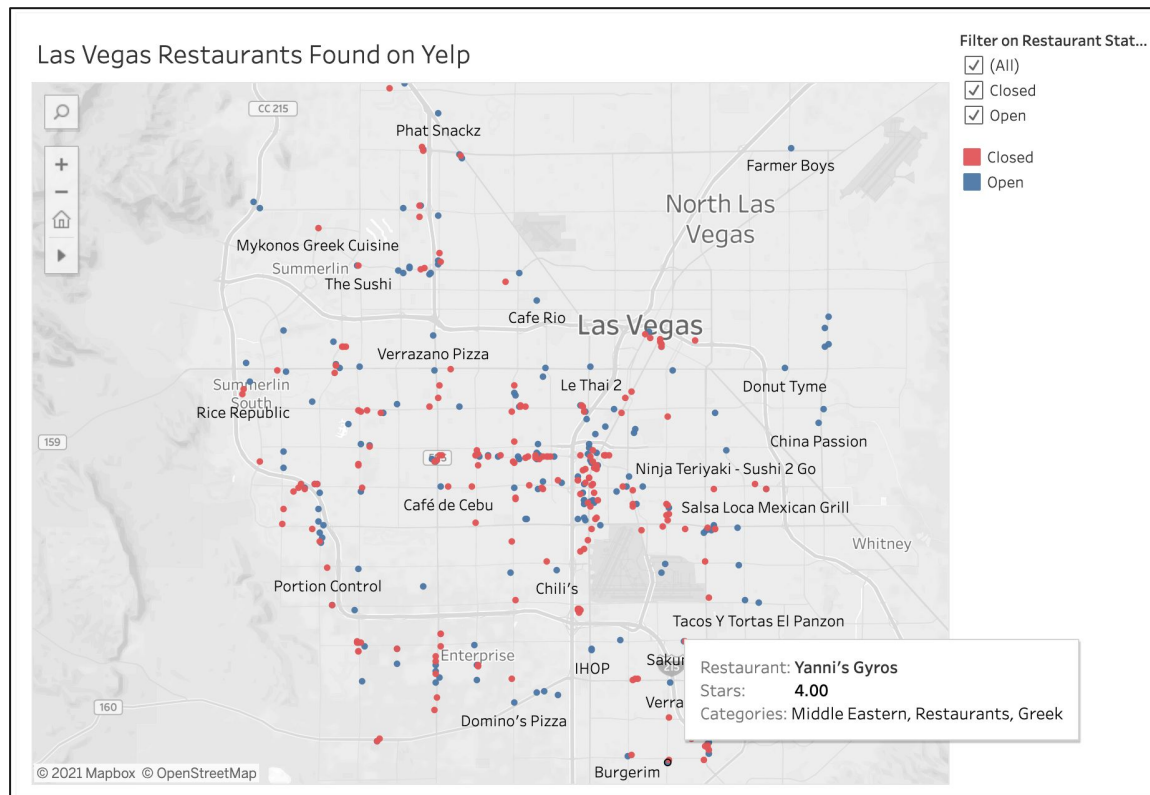


## Data Collection (cont.)

- Las Vegas
  - City with the largest amount of reviews in Yelp dataset
  - Dynamic city with high-level of tourism
- Data pull - Las Vegas restaurants that had between 100 and 300 reviews
- Result
  - EDA: 400 restaurants with 100-300 reviews
  - Modeling: all restaurants with 100-300 reviews



# Data Collection (cont.)





**EDA**

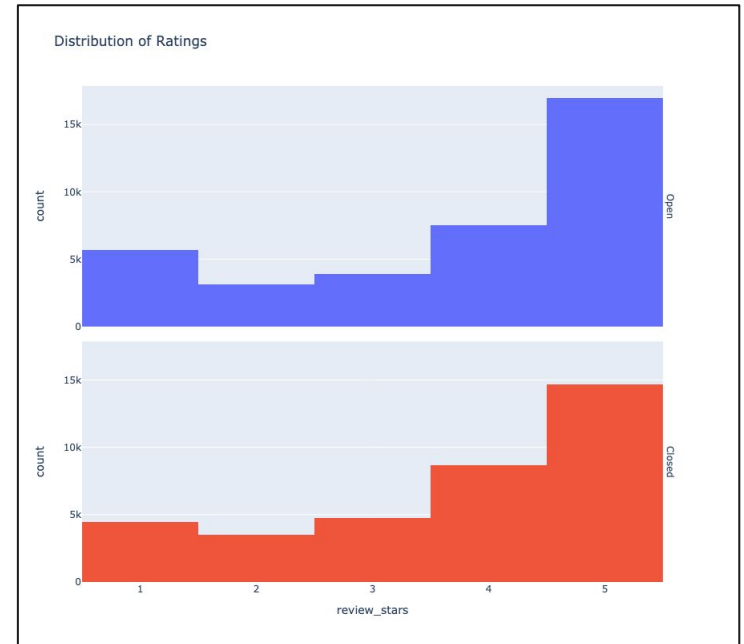
# Restaurants in Las Vegas

## Open Restaurants

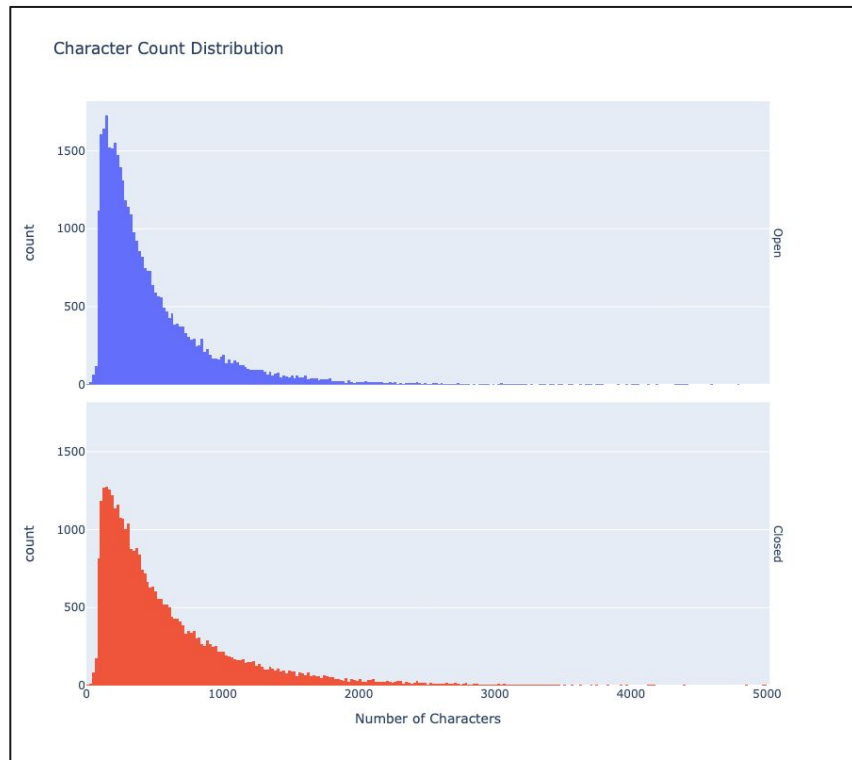
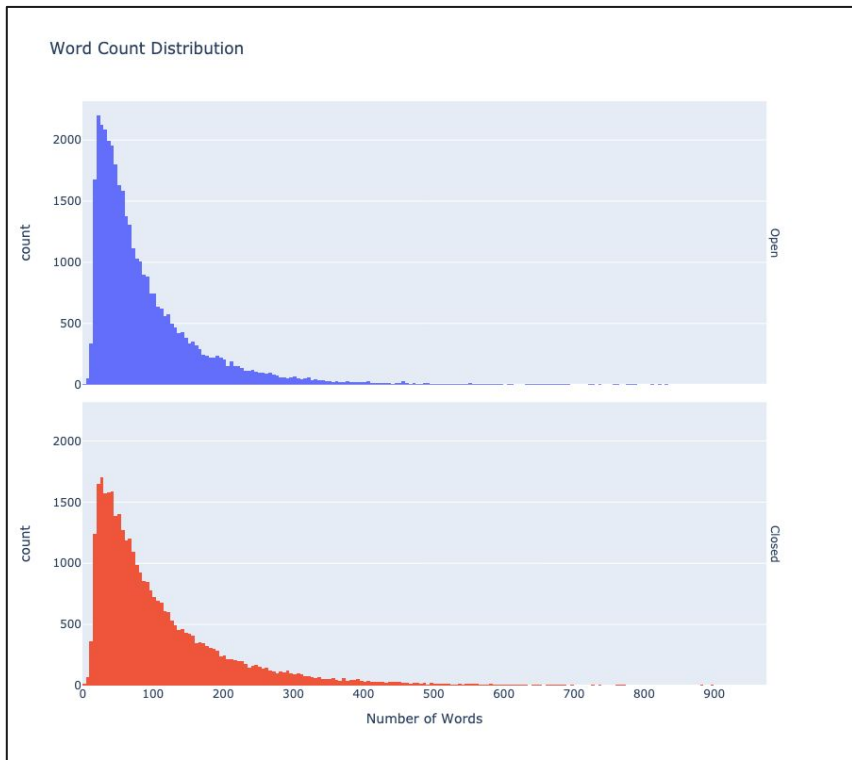
- 200 restaurants
- 37,211 reviews
- Average Rating: 3.7 stars

## Closed Restaurants

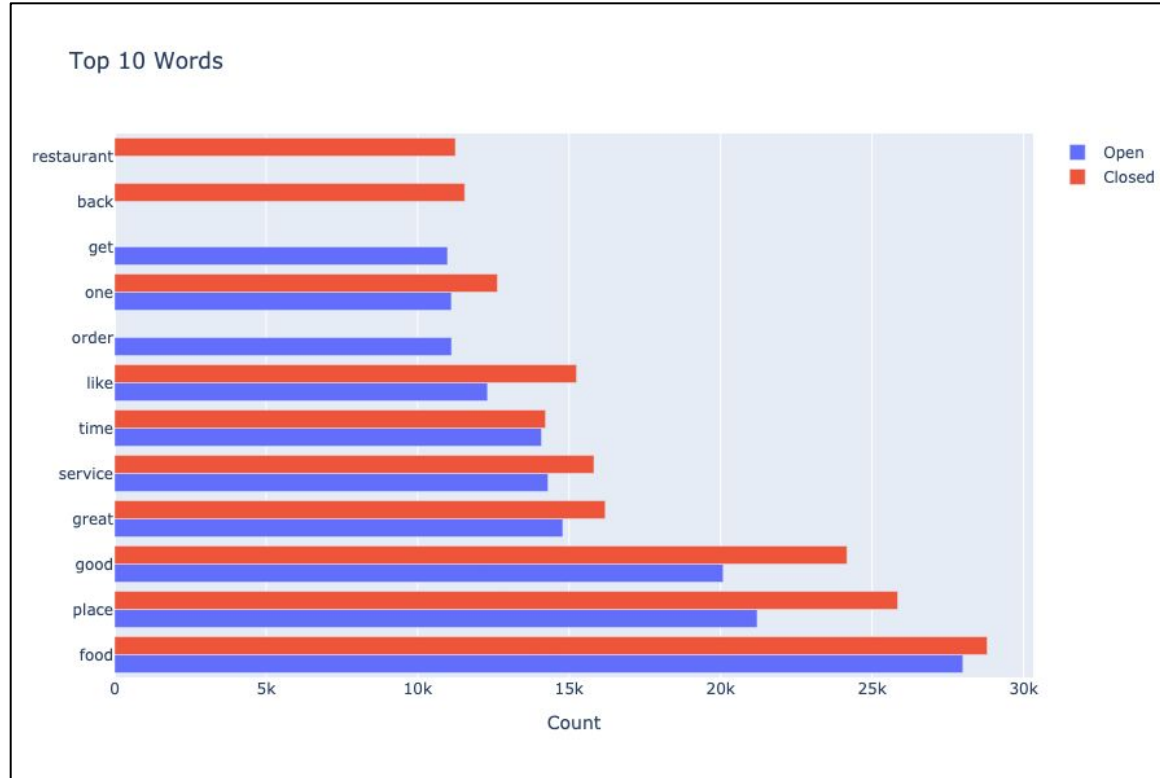
- 200 restaurants
- 36,067 reviews
- Average Rating: 3.7 stars



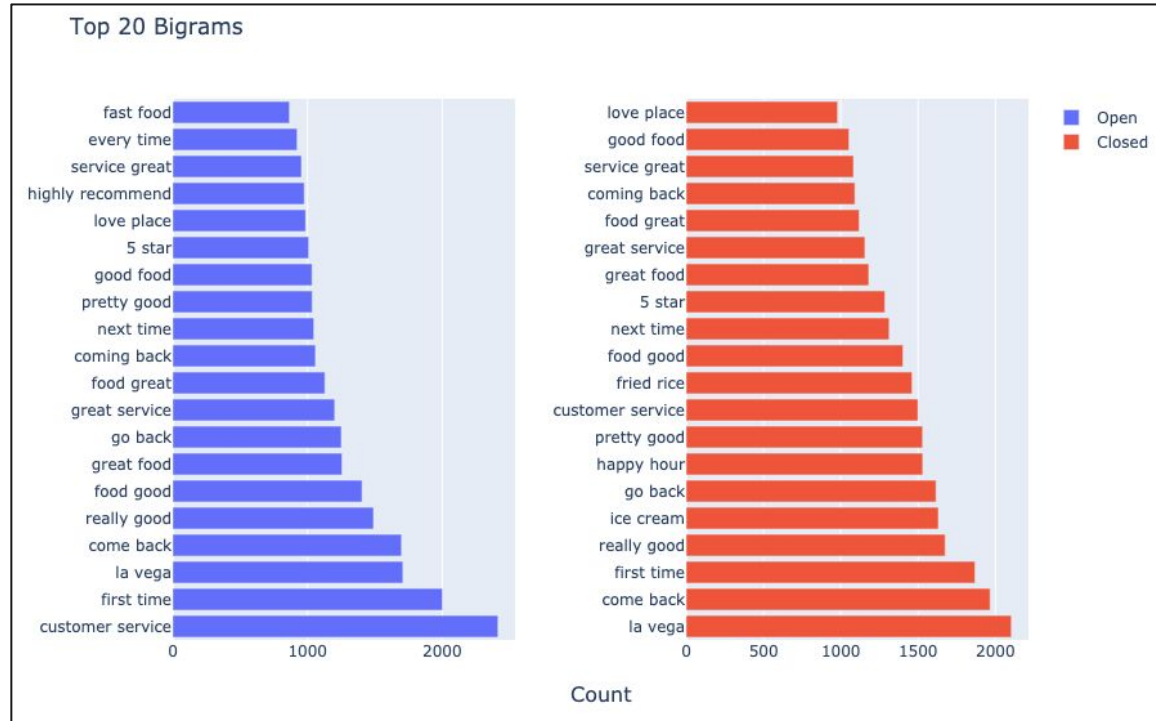
# Restaurants in Las Vegas (cont.)



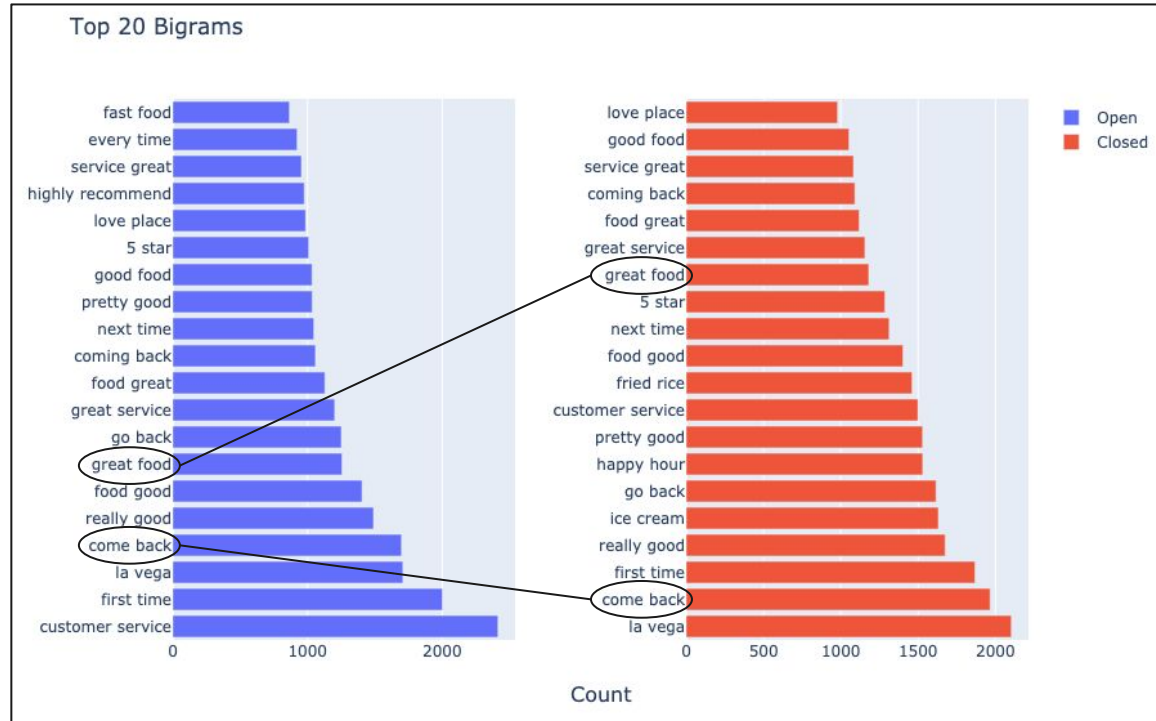
# Restaurants in Las Vegas (cont.)



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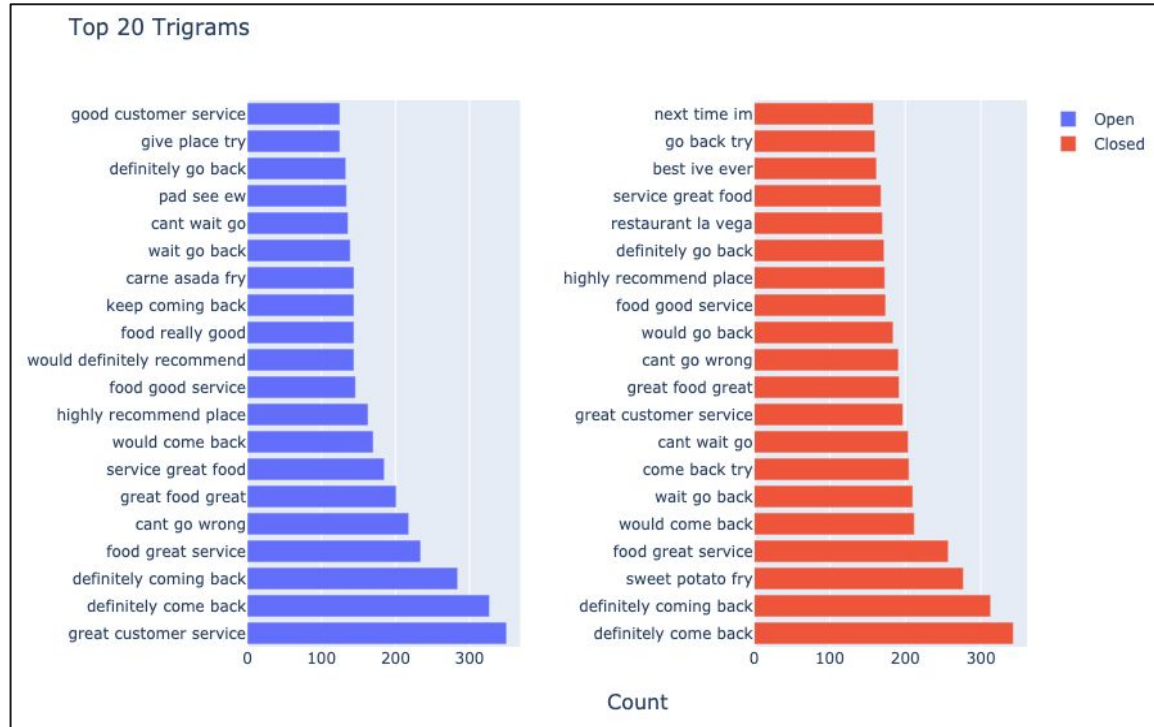


# Restaurants in Las Vegas (cont.)

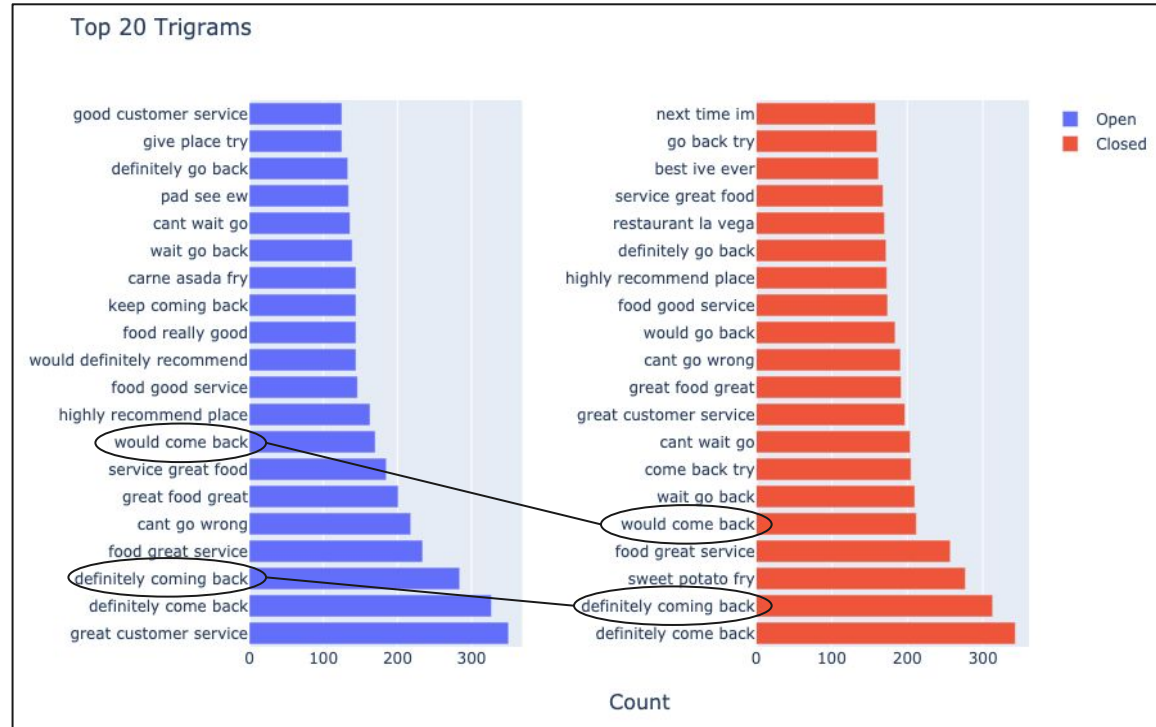




# Restaurants in Las Vegas (cont.)



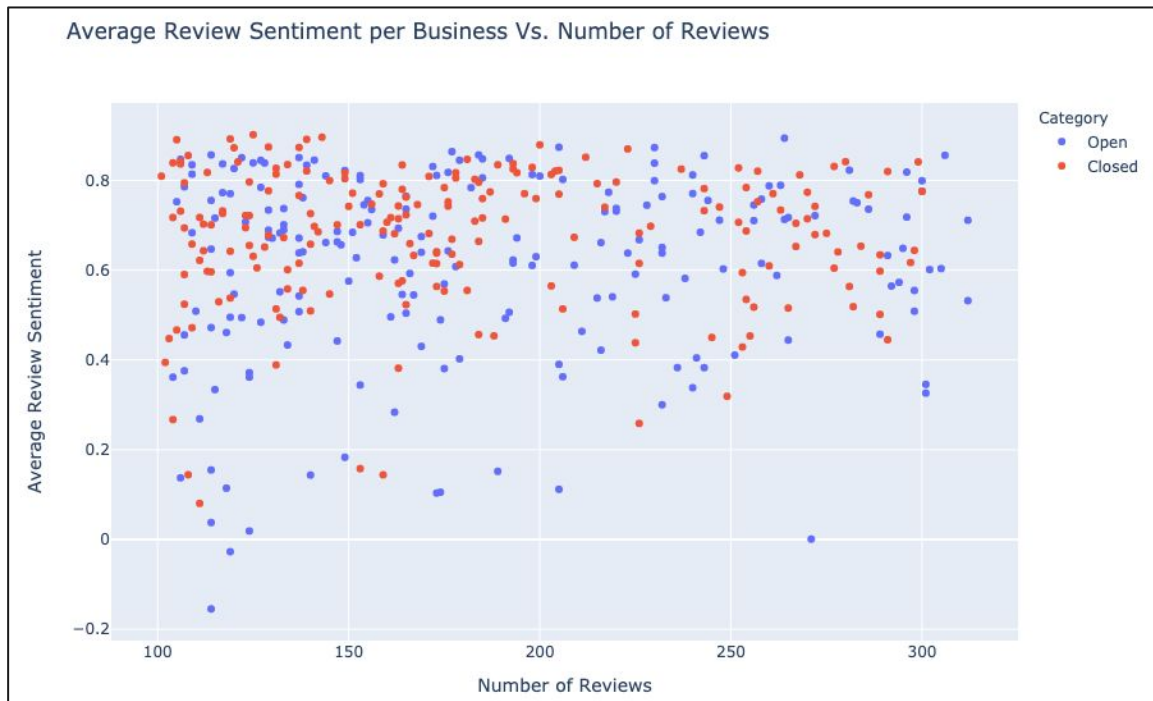
# Restaurants in Las Vegas (cont.)



# Restaurants in Las Vegas (cont.)

## Sentiment Analysis

- Evenly distributed between open and closed restaurants
- Wider spread of sentiment for restaurants with lower number of reviews





# Modeling



## Modeling - First Basic Models

The following algorithms were used:

- LogisticRegression
- MultinomialNB
- RandomForest
- ExtraTrees
- K-NearestNeighbors
- SVC
- AdaBoostClassifier
- GradientBoostingClassifier

**Baseline accuracy score - 0.72**

	Model	Preprocessing	Accuracy	Baseline improvement
1	GradientBoostingClassifier	TfidfVectorizer	0.830	0.105
2	AdaBoostClassifier	TfidfVectorizer	0.799	0.074
3	SVC	TfidfVectorizer	0.796	0.071
4	LogisticRegression	CountVectorizer	0.781	0.056
5	LogisticRegression	TfidfVectorizer	0.774	0.049
6	RandomForest	TfidfVectorizer	0.765	0.040
7	ExtraTrees	TfidfVectorizer	0.758	0.033
8	K-NearestNeighbors	TfidfVectorizer	0.742	0.017

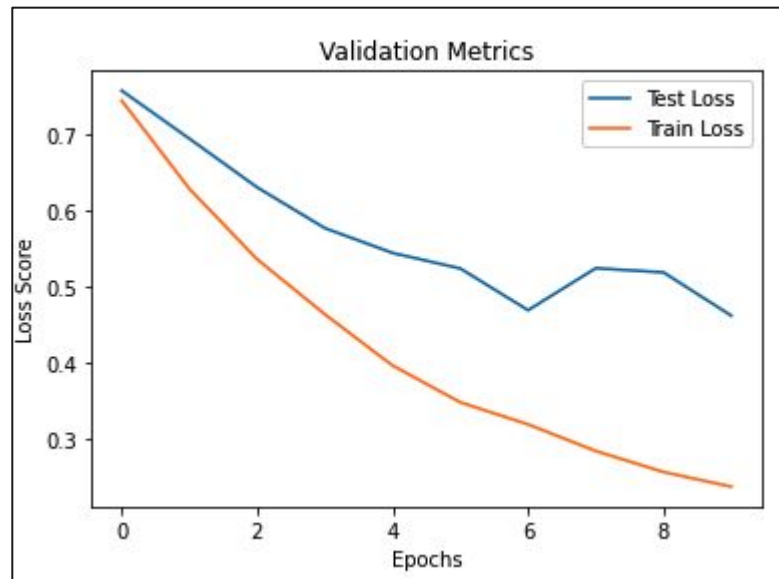


## Modeling (cont.)

	Model	Recall	Specificity	Balanced Accuracy	Accuracy	Model Improvement
1	GradientBoostingClassifier	0.959	0.549	0.754	0.846	0.016
2	SVC	0.919	0.566	0.742	0.821	0.025
3	AdaBoostClassifier	0.891	0.566	0.728	0.801	0.002
4	LogisticRegression	0.825	0.623	0.724	0.769	-0.011
5	RandomForest	0.984	0.205	0.595	0.769	0.005
6	KNN	0.950	0.164	0.557	0.733	-0.009

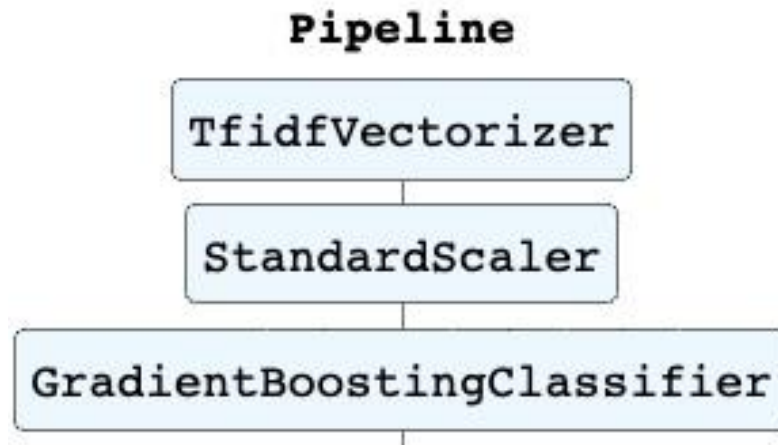
# Neural Network Model

- Grid Search - best parameters
  - Keras Classifier
    - Dropout: 0.5
    - Regularization: L2
    - Units: 128
- Scores
  - Accuracy: 0.832
  - Balanced Accuracy: 0.709



## Overall Best Model - GradientBoostingClassifier

- Model parameters:
  - Estimator parameters
    - Max depth = 3
    - Min samples leaf = 20
  - TfidfVectorizer parameters
    - Max Features: 1000
    - Ngram Range: (1, 1)
    - Stop Words: yes, english
- Scores
  - Accuracy: 0.846
  - Balanced Acc: 0.754





## Data Limitations / Constraints

- Dataset was large: over 10GB
- Only used restaurants with 100-300 reviews
- Data only current up to the end of 2019
- Reviews pulled from a Yelp provided dataset (unknown missing)





## **Conclusions / Recommendations**

## Conclusions / Recommendations

- While a lot can be learned from restaurant reviews, they are not very effective predictors of whether or not a restaurant will close
- Initial modeling efforts led us to believe that customer service played an important role, something to look into



## Future Areas of Focus

- More advanced NLP methods for better understanding of text
- Model on all restaurants in given city
- Model on other business types
- Expand our modeling efforts to new cities
- Gather recent reviews to observe effect of COVID on our models





**Questions?**



# BACKUP

1. Restaurant Statistics
  - a. <https://www.fsrmagazine.com/expert-takes/restaurant-profitability-and-failure-rates-what-you-need-know>
2. Yelp Information
  - a. <https://www.yelp-press.com/company/fast-facts/default.aspx>
3. Yelp Dataset
  - a. <https://www.yelp.com/dataset>
4. Tableau Dashboard
  - a. [https://public.tableau.com/profile/adam.pardo#!/vizhome/yelp\\_restaurants/Dashboard1?publish=yes](https://public.tableau.com/profile/adam.pardo#!/vizhome/yelp_restaurants/Dashboard1?publish=yes)



## BACKUP - Neural Network Model

Model: "sequential\_329"

Layer (type)	Output Shape	Param #
dense_987 (Dense)	(None, 12)	12012
batch_normalization_329 (Batch Normalization)	(None, 12)	48
dense_988 (Dense)	(None, 128)	1664
dropout_329 (Dropout)	(None, 128)	0
dense_989 (Dense)	(None, 1)	129

Total params: 13,853

Trainable params: 13,829

Non-trainable params: 24

# BACKUP Feature Importance - Open vs Closed Restaurants

