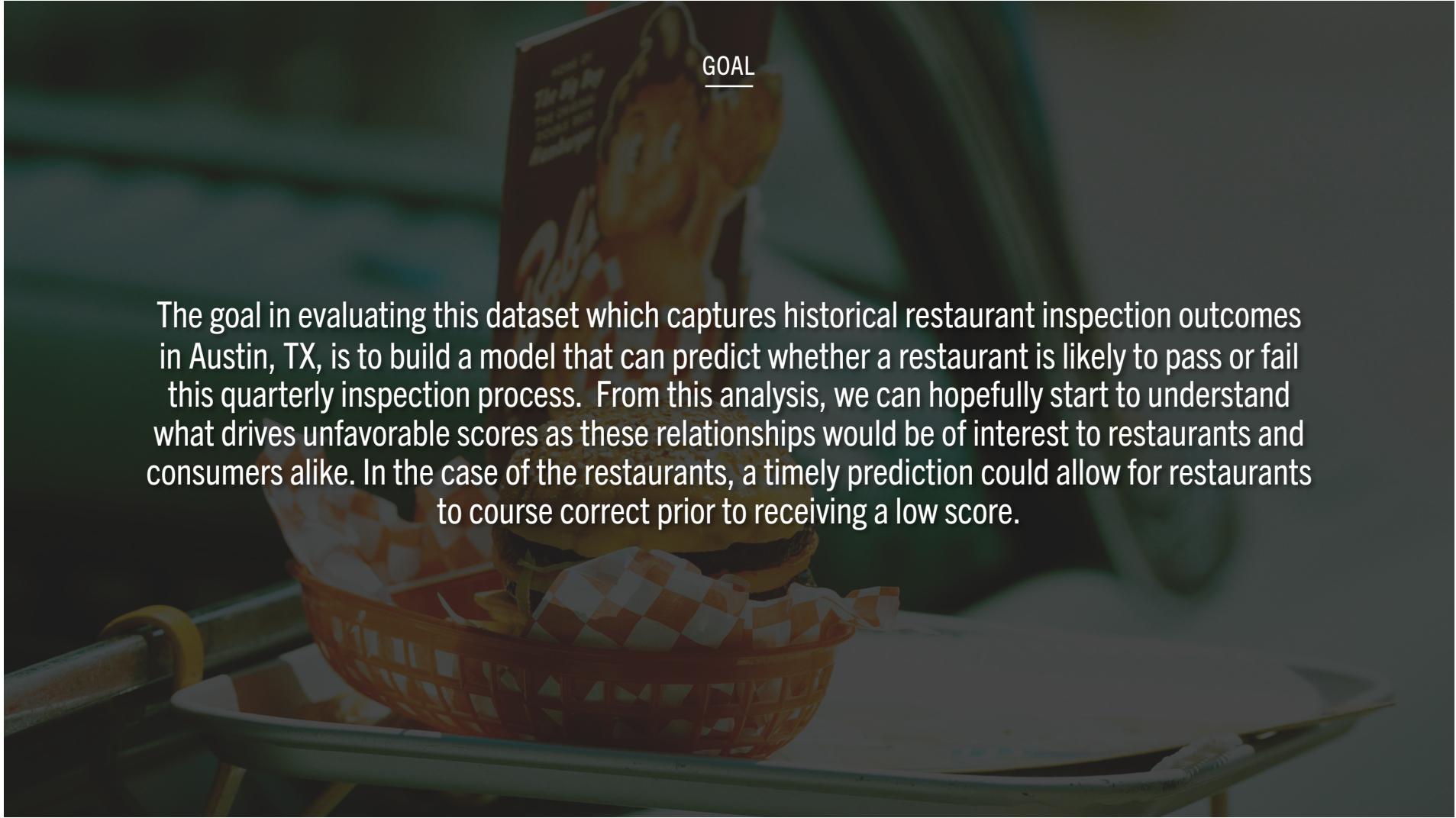




PREDICTING RESTAURANT INSPECTION OUTCOMES

CAPSTONE 2 - PRESENTATION



GOAL

The goal in evaluating this dataset which captures historical restaurant inspection outcomes in Austin, TX, is to build a model that can predict whether a restaurant is likely to pass or fail this quarterly inspection process. From this analysis, we can hopefully start to understand what drives unfavorable scores as these relationships would be of interest to restaurants and consumers alike. In the case of the restaurants, a timely prediction could allow for restaurants to course correct prior to receiving a low score.

MOTIVATION AND CHALLENGES

MOTIVATION:

- GREAT OPPORTUNITY TO DO SOME ANOMALY DETECTION
 - THE DATA IS HIGHLY CLASS IMBALANCED

CHALLENGES:

- THE DATA IS HIGHLY CLASS IMBALANCED 😊
(WITH ONLY XX% OF INSPECTIONS RESULTING IN A FAILING SCORE)
- MANY RESTAURANTS HAVE FEW OR NO REVIEWS



THE DATA

SOURCES:

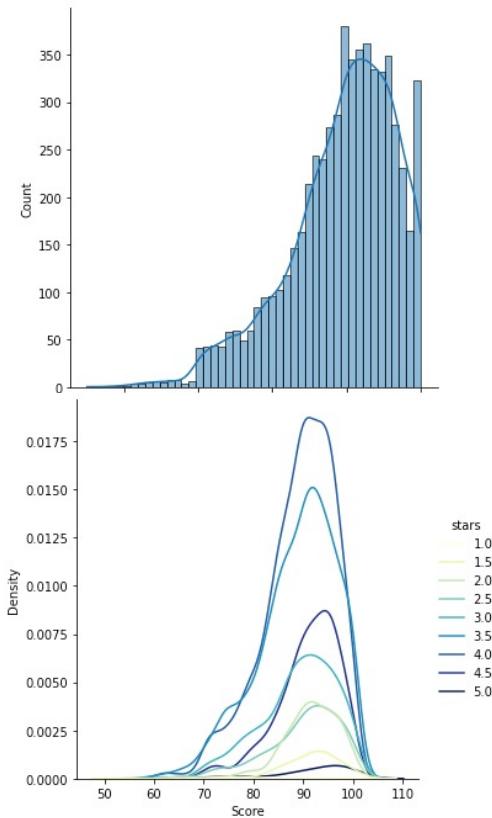
- [AUSTIN RESTAURANT INSPECTIONS SCORES](#) | AUSTIN PUBLIC HEALTH
- RESTAURANT REVIEWS | YELP

HIGH-LEVEL STATS:

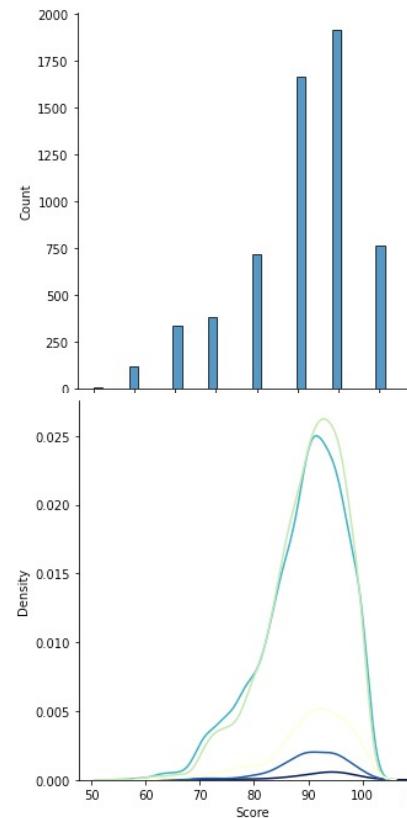
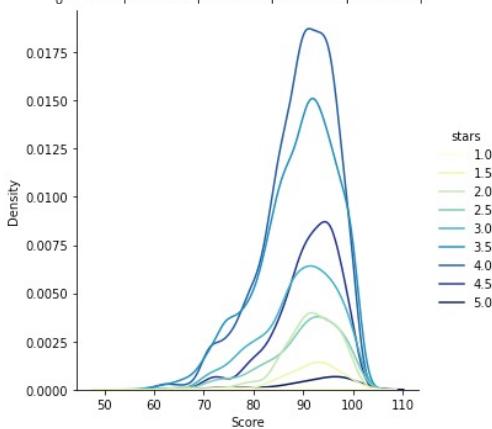
- 3 YEARS OF INSPECTIONS HISTORY
- INSPECTIONS OCCUR QUARTERLY
- CONTAINS SCORES FOR **XX** UNIQUE RESTAURANTS
- **XX** TOTAL REVIEWS
- ALLOWING A 60-DAY LOOKBACK FOR REVIEWS-BASED FEATURES
- ASSUMES THAT NON REVIEWS-BASED FEATURES ARE STATIC OVER THE 3 YEARS

HOW IS THE DATA DISTRIBUTED?

- INSPECTION SCORES AND RATINGS ALIKE ARE LEFT-SKewed AND TEND TO BE HIGH



- INTERESLINGLY AND ENCOURAGINGLY INSPECTIONS SCORES FOLLOW A SIMILAR DISTRIBUTION ACROSS RATINGS AND PRICE RANGE



THE SCORING IS HIGHLY IMBALANCED

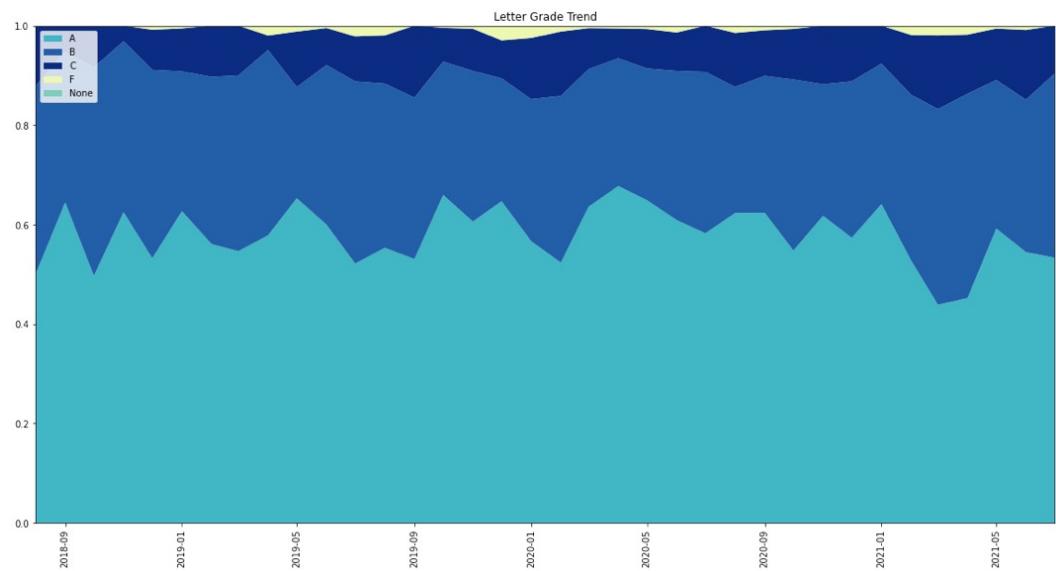
A: 58%

B: 32%

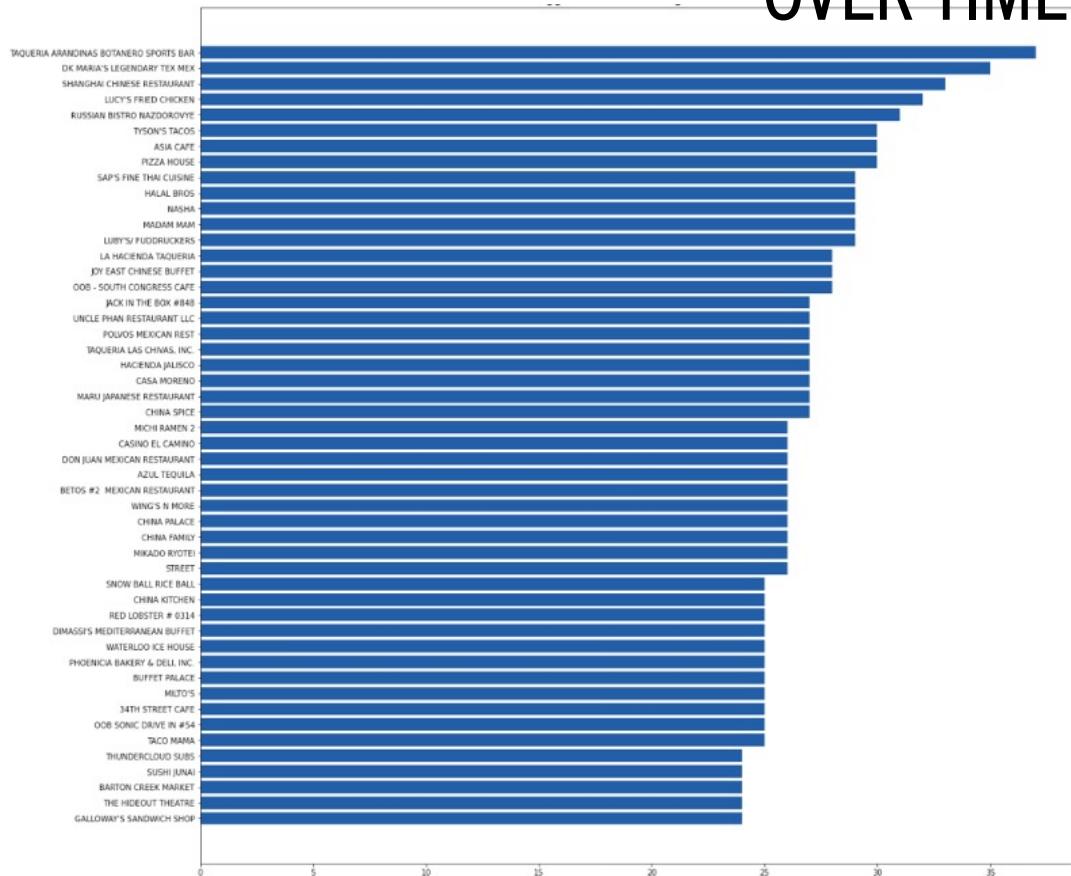
C: 10%

F*: 0.9%

* 70 OR BELOW



BUT AT THE RESTAURANT LEVEL WE DO SEE VARIATION OVER TIME

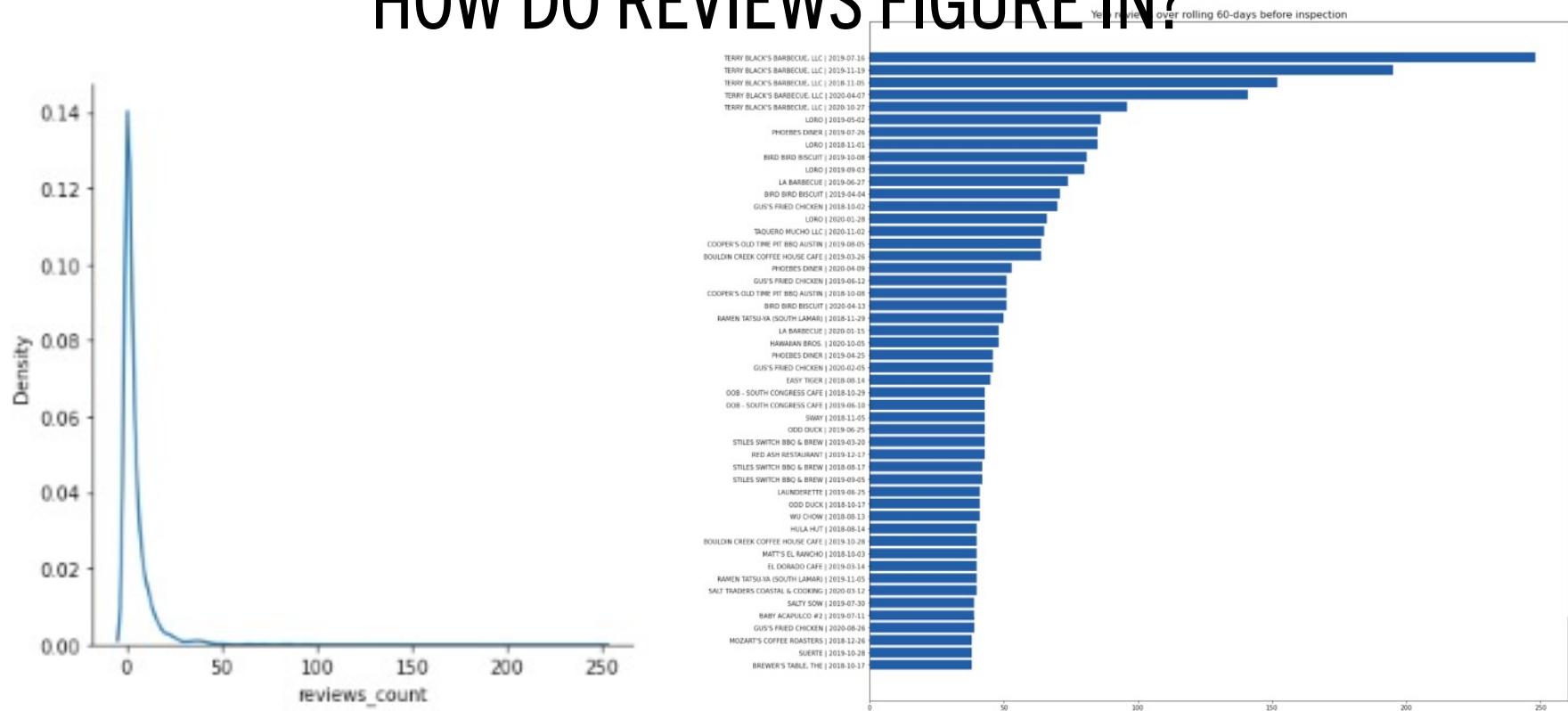


- IT'S A MIXED BAG IN TERMS OF RESTAURANTS SCORE OVER TIME
- THE LARGEST VARIATION (IN ABSOLUTE TERMS) OVER THE 3 YR PERIOD WAS 35 POINTS



WHAT CAN WE LEARN FROM THE YELP
DATASET??

HOW DO REVIEWS FIGURE IN?



MOST RESTAURANTS DO NOT HAVE VERY MANY IF ANY REVIEWS DURING THE 60-DAY LOOKBACK WINDOW

THE COVERAGE IS NOT GREAT, BUT IT IS A RICH DATASET SO LET'S DIG A BIT DEEPER



SENTIMENT ANALYSIS



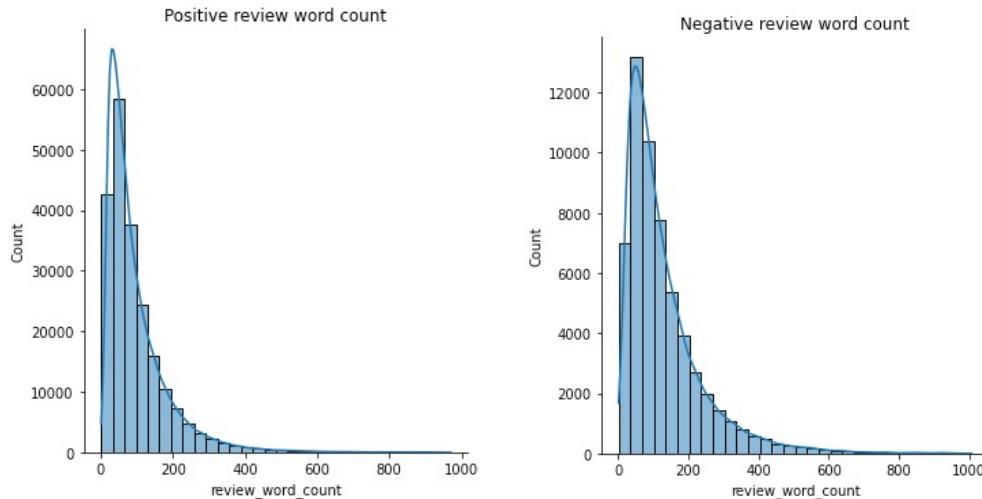
THIS ANALYSIS PRODUCED SOME INTEREST LEARNINGS BUT IT WOULD NEED TO REVISITED A BIT MORE DEEPLY FOR BETTER FEATURE EXTRACTION AS WELL AS TO GAUGE THE POTENTIAL FOR INFORMATION GAIN IN A SUBSEQUENT MODEL

```
positive['text'][1:2].values
```

array(['Definitely one of my favorite places for vietnamese grub! Very authentic and although they h, the people here are really nice. Will always recommend others to try it.']),

```
negative['text'][3:4].values
```

array(["I have been to this place a couple of times and each time, I never ceased to be disappointed. par at best and the place is dirty. One time a friend of mine went there and found a lizard (yes, that ard) in his vermicelli bowl. The service staff is not very helpful either. They are hard to understand em to be in a bad mood. So if you enjoy paying too much for a can (yes, CAN) of soda with a cup of ice ards in your food, I would recommend Pho Van."],



OTHER FEATURES

THIS ANALYSIS PRODUCED SOME
INTEREST LEARNINGS BUT IT WOULD
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FOR BETTER FEATURE EXTRACTION AS
WELL AS TO GAUGE THE POTENTIAL FOR
INFORMATION GAIN IN A SUBSEQUENT
MODEL



BURGERS

SHAKES

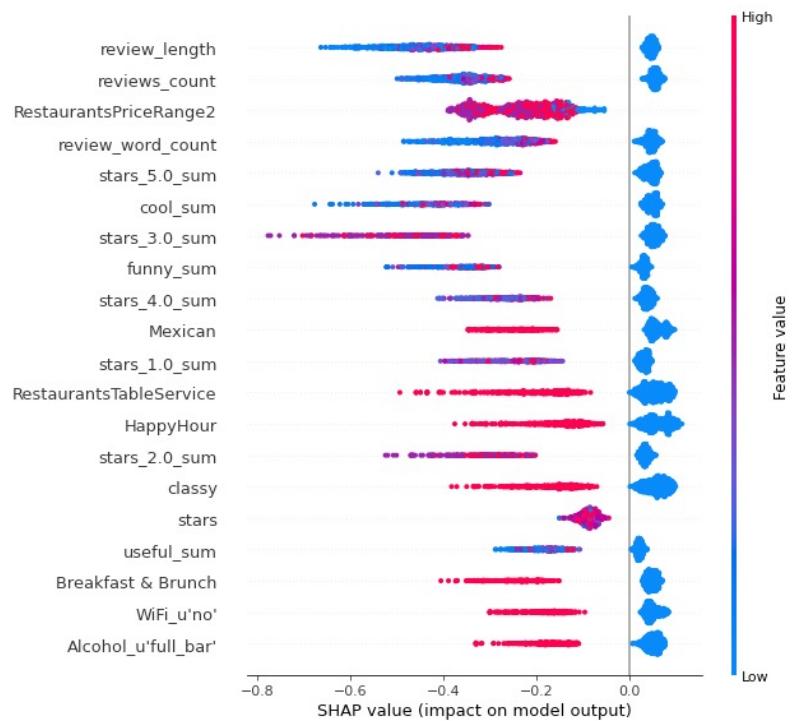
BUILDING THE MODEL

Pal's

OTHER FEATURES

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MODEL

ISOLATION FOREST RESULTS



AUTOENCODER

