# Predicting Sucess on yelp:

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### Objective

Our aim is to develop a success prediction for new restaurants

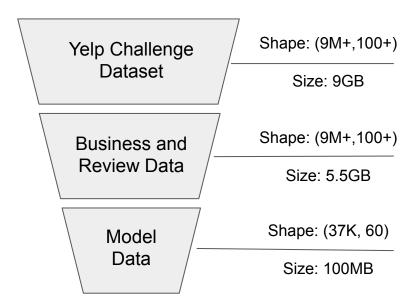
Success is defined as 4+ stars

The predictor can help new restaurants evaluate whether they are 'on track' to becoming successful restaurants



#### Data overview

#### The dataset comes from Yelp and was honed down in Cloud SQL



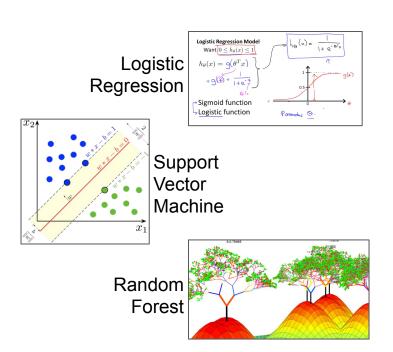
#### Data Overview:

- Data provided through <u>Yelp 2019</u>
  <u>Challenge</u>
- Used Google Cloud SQL to store data and run initial cuts
- Types of data:
  - Business information
  - Reviews and tips
  - User information
- Data ultimately used in model:
  - Business information
  - Aggregate characteristics of reviews for each restaurant

### Approach

#### We will attempt 3 models: Logistic Regression, SVM and Random Forest

#### Models



#### Success Metrics:

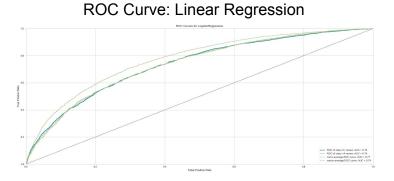
- Accuracy we aim to maximize accuracy.
  False negatives and false positives have relatively similar value in this context.
  Target classes are balanced.
- Efficiency models should run with our available computing power

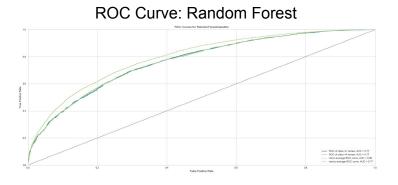
### Model and Hyperparameter Tuning

#### Models, after hyperparameter tuning, resulted in similar results

- Models capped out at ~0.71 accuracy

- We believe that improvement hyperparameter tuning have been exhausted
- Further improvement may be had by adding features from other data in the dataset





#### **Model Selection**

Accuracy is close between Random Forest and SVM. We choose Random Forest as our model based on performance and resource requirements

Classifier	Accuracy*	Resource Requirement	Other Consideration
Random Forest	71.7585%		Feature Importance
SVM Nystroem	71.3564%		
SVM Monte Carlo	70.4559%		
Logistic Regression	70.9743%		

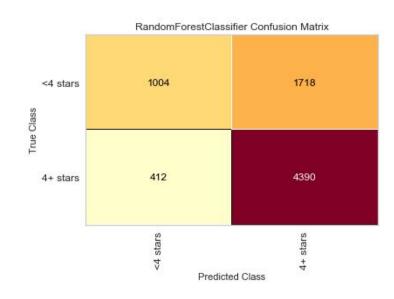
\*Mean accuracy in k-fold cross validation

### Model Performance

#### Key features include location as well as aggregate review characteristics

Importance		Feature
4	0.063159	meanuseful
1	0.062690	latitude
14	0.060382	lowerquartilewordcount
2	0.055452	longitude
12	0.052334	medianwordcount
5	0.048457	avgwordcount
0	0.046001	review_count
13	0.044722	upperquartilewordcount
3	0.041821	meanfunny
10	0.035700	avgusefulwordcount

Key features include location as well as aggregate characteristics from business reviews



# Recommendations & Limitations

# Further Research and Next Steps

#### Potential improvement areas:

- Feature addition from reviews
- Reviews
- Changing model to look at initial reviews as a predictor of later success

## Thank You!



# Slide Template Option

## **Template Option**

**Key Point:** lorum ipsum etc

