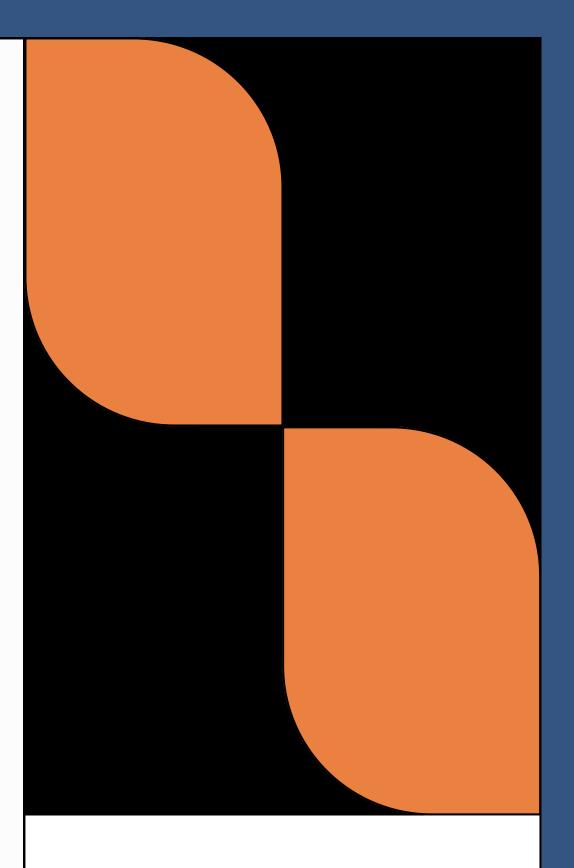


# YelpSense

Unleashing the Power of Customer Feedback

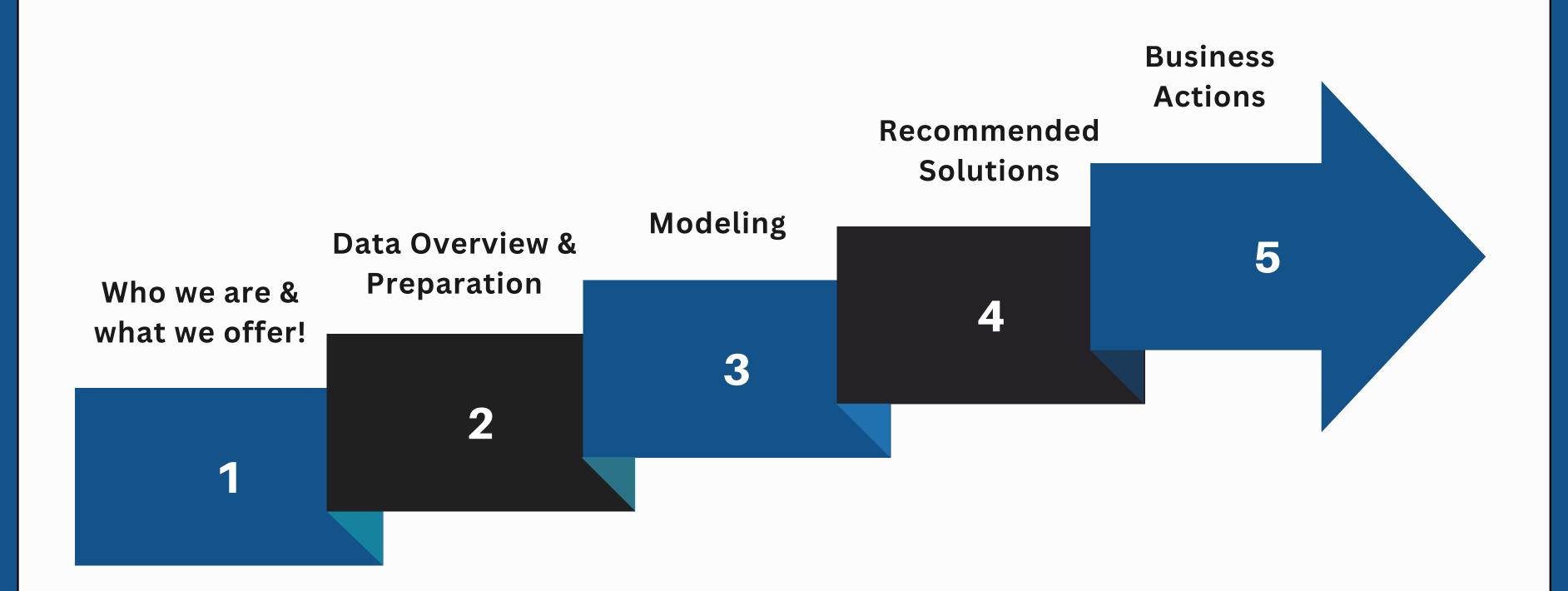
### Presented by:

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# Agenda





# Business Problem Statement

60% of people read reviews before visiting restaurant. With the rise of online reviews, businesses face the challenge of analyzing vast amounts of customer feedback to understand their sentiment and make datadriven decisions. Traditional manual methods are time-consuming, subjective, and not scalable to handle big data





## Who we are?

Unleashing the Power of Customer Feedback

YelpSense is a leading provider of advanced sentiment analysis solutions for product reviews. We aim to provide an automated, scalable, and accurate sentiment analysis solution to help businesses extract valuable insights from Yelp reviews.



# Machine Learning Canvas

The Machine Learning Canvas (v0.4) Designed for: Group 2 Designed by: Iteration: Decisions Value ML task Data Sources Collecting Data Propositions How do we get new data to How are predictions used to Input, output to predict. Which raw data sources can make decisions that provide type of problem. we use (internal and learn from (inputs and What are we trying to do for the the proposed value to the end-user? external)? outputs)? end-user(s) of the predictive system? Predictions are used to What objectives are we serving? Type: Classification make decisions on various Improve Customer Continuously collect new Yelp business actions, such as: Data sources: Yelp reviews Satisfaction: review data to learn from, Identifying areas of dataset containing text Identify pain points including both inputs (review reviews, ratings, timestamps, improvement for Input: Yelp reviews user and areas of text) and outputs (sentiment and metadata. enhancing customer dissatisfaction. data and restaurant satisfaction Implement targeted data Refining product offerings measures for based on customer improvement. sentiment towards Output: Sentiment Regularly monitor competitors prediction (positive, feedback and Promptly addressing measure impact. negative) negative sentiment to Competitive manage reputation Analysis: effectively Analyze sentiment trends for Making Offline Features Building Models competitors' Evaluation Predictions offerings. Input representations When do we create/update Identify strengths extracted from raw data models with new training and weaknesses. When do we make predictions on new Methods and metrics to evaluate the sources. data? How long do we have to inputs? How long do we have to system before deployment. Refine own featurize training inputs and create a featurize a new input and make a offerings and prediction? capitalize on Methods: Split the labeled opportunities. Extract relevant features from dataset into training and Reputation the raw data sources, such as Models are created or updated Predictions are made on validation sets to assess Management: bag-of-words representation, with new training data new inputs (Yelp model performance. Proactively detect TF-IDF scores, or word periodically to adapt to reviews) in real-time as Metrics: Evaluate and address embeddings. changing sentiment patterns. they are received. classification accuracy. negative sentiment. The frequency of model The time available for precision, recall, F1 score, Respond updates depends on the rate making predictions and possibly AUC-ROC or professionally and of data availability and the depends on the desired AUC-PR to measure the offer solutions. need for model refinement. speed of response to quality of predictions prior to Build a positive customer feedback and deployment. brand image and the urgency of customer trust. addressing any negative sentiment. Live Evaluation and Business Metrics: Merchant Defect Rate, Merchant Failure Rate, Customer Satisfaction Rate, NPS Score Monitoring Model Metrics: Evaluate classification accuracy. Methods and metrics to evaluate the precision, recall, F1 score, and possibly AUC-ROC or system after deployment, and to AUC-PR to measure the quality of predictions prior to quantify value creation. deployment.

## Data Overview

Source - Yelp

### **User Reviews Dataset**

- Consists of customer ratings and text reviews
- ~ 7M reviews
- 9 features

### **Restaurant Database**

- Consists restaurant demographics
- ~ 150k restaurant records
- 14 features

# Data Preparation

#### **Review Dataset - Selected 3 features**

- Business ID
- Review Stars
- Review Text

### **Restaurant Dataset - Selected 2 features**

- Business ID
- Category of Restaurant

Inner join on Business ID

- Analyzed "English Reviews **ONLY**"
- Removed 3-star reviews
- Labelling Sentiment
  - Stars < 3 Classified as 0</li>
  - Stars > 3 Classified as 1

#### **FINAL DATASET**

- Balanced 0 and 1 reviews
  - (21.3% vs 78.6%)
- Removed stop words
- Final columns:
  - Cleaned reviews &
  - Sentiment

# Model Building

1	Data Randomly Split : 70:30	
2	Feature Engineering and Text Vectorization  Models  Bag-of-Words Representation  TF-IDF  nGrams	
3	<ul> <li>Classification Models</li> <li>Logistic Regression</li> <li>Random Forest</li> <li>SVM</li> </ul>	
4	Model Evaluation and Integration	

# Model Evaluation

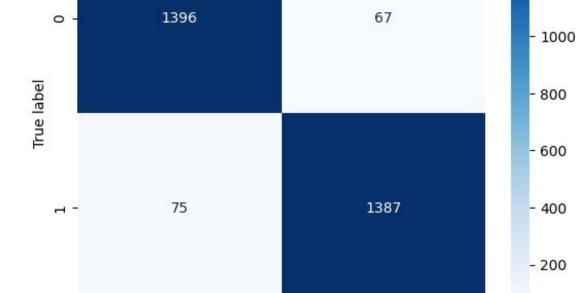
## **Using TD-IF**

<u>Models</u> Metrics	Logistic	Random Forest	SVM
Accuracy	0.95	0.94	0.97
Precision	0.95	0.97	0.97
Recall	0.94	0.97	0.98
F1 Score	0.95	0.97	0.97

# Confusion Matrix

### **Logistics Regression**

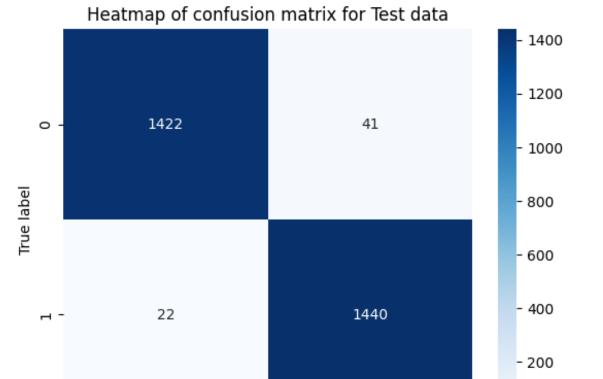




Predicted label

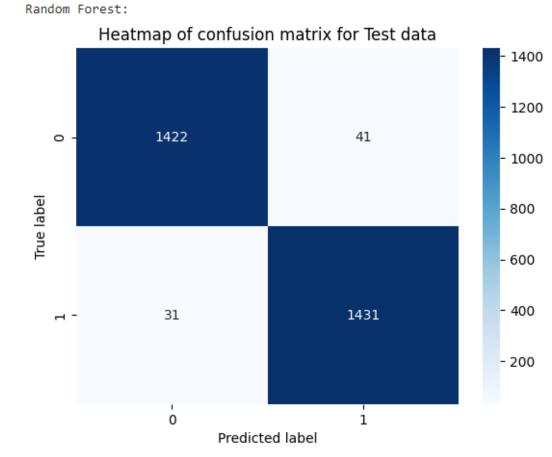
### **SVM**





Predicted label

### **Random Forest**



# Integration and Deployment



### **Seamless Integration**

Our sentiment analysis solution can be seamlessly integrated into existing systems and workflows.



### Scalable Deployment

Our solution is designed to handle large volumes of Yelp reviews, making it scalable to accommodate growing data needs as the business expands



### **Customization Options**

We offer customization options to align with specific business requirements and objectives ensuring a tailored approach.



### **User-Friendly Interface**

Our solution comes with a user-friendly interface that allows businesses to easily access and interpret sentiment analysis results, empowering decision-makers with actionable insights.

# Recommended Business Solution

### **ADVANCED NLP**

- State-of-the-art algorithms for sentiment analysis
- Contextual understanding and nuanced language analysis

## MACHINE LEARNING MODELS

- Custom sentiment analysis models
- Accurate classification of reviews (positive, negative, neutral)

## REAL-TIME MONITORING AND ALERT SYSTEMS

- Continuous scanning of Yelp reviews
- Prompt identification of negative sentiment

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# How can we help?

## IMPROVE CUSTOMER SATISFACTION

- Identify pain points and areas of dissatisfaction.
- Implement targeted measures for improvement.
- Regularly monitor feedback and measure impact.

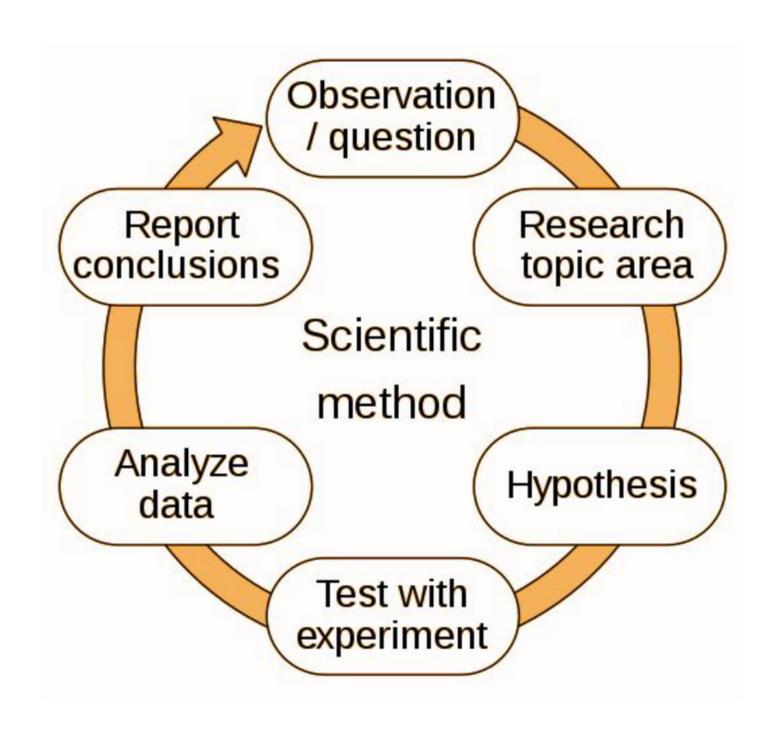
### **COMPETITIVE EDGE**

- Analyze sentiment trends for competitors' offerings.
- Identify strengths and weaknesses.
- Refine own offerings and capitalize on opportunities.

#### **BRAND IMAGE**

- Proactively detect and address negative sentiment.
- Respond professionally and offer solutions.
- Build a positive brand image and customer trust.

# Conclusion



We offer businesses a powerful solution by leveraging advanced NLP techniques, ML models, and real-time monitoring to extract valuable insights from customer feedback at scale, enabling data-driven decision-making.

Our solution empowers businesses to unlock the full potential of Yelp reviews and drive business outcomes through actionable sentiment analysis.

# Thank You

Invest in our vision and join us on this exciting journey

