



SENTIMENT ANALYSIS OF HOTEL REVIEWS

Enhancing Customer Satisfaction and Brand Loyalty

Team : Emotion Analysts

Overview



Problem Statement

Enhance guest experience by addressing negative feedback promptly.



MARKET ANALYSIS

Aggregation from review platforms and social media.



Data Preparation

Cleaning, preprocessing, and vectorization.

Modeling

Sentiment analysis using Logistic Regression, SVM, Random Forest.



Evaluation:

Metrics like precision, recall, ROC AUC.



Deployment:

Real-time integration and monitoring.



Business Problem Statement

1

Impact of Guest Reviews:

Negative feedback impacts bookings, revenue, and brand loyalty.

2

Feedback Challenges: Fragmented

Feedback sources delay issue recognition.

3

Importance of Timely Response:

Crucial to address negative sentiments promptly to mitigate dissatisfaction.

 just-shower-thoughts
The song "Hotel California" is basically a bad Yelp review and a 3 minute guitar solo.
 spinesongs
#the drinks were tasty but I got stabbed here
Source: just-shower-thoughts

 Jessie
@mnmajessie
Hotels be like, it's \$150 a night and you're staying 2 nights so that brings your total to \$947.43

 Alex Baze
@bazecrazz
Follow
I think hotels set their rates based on how hard the shower is to figure out.
7:30 AM · Jun 25, 2015
142 Retweets 4 Retweets

 Bonnie Gillespie (she/her)
@bonniegillespie
Follow
Not sure I'll ever understand a hotel that boasts 600 threadcount sheets and plush, cozy robes providing toilet paper and tissue made of sandpaper. 🙄🙄🙄
12:57 AM · Jul 25, 2018
39 Replies 4 Replies

Proposed Business Solution Approach

01

Real-time Sentiment Analysis Platform: Aggregates feedback from various channels, detects negative sentiments using NLP.

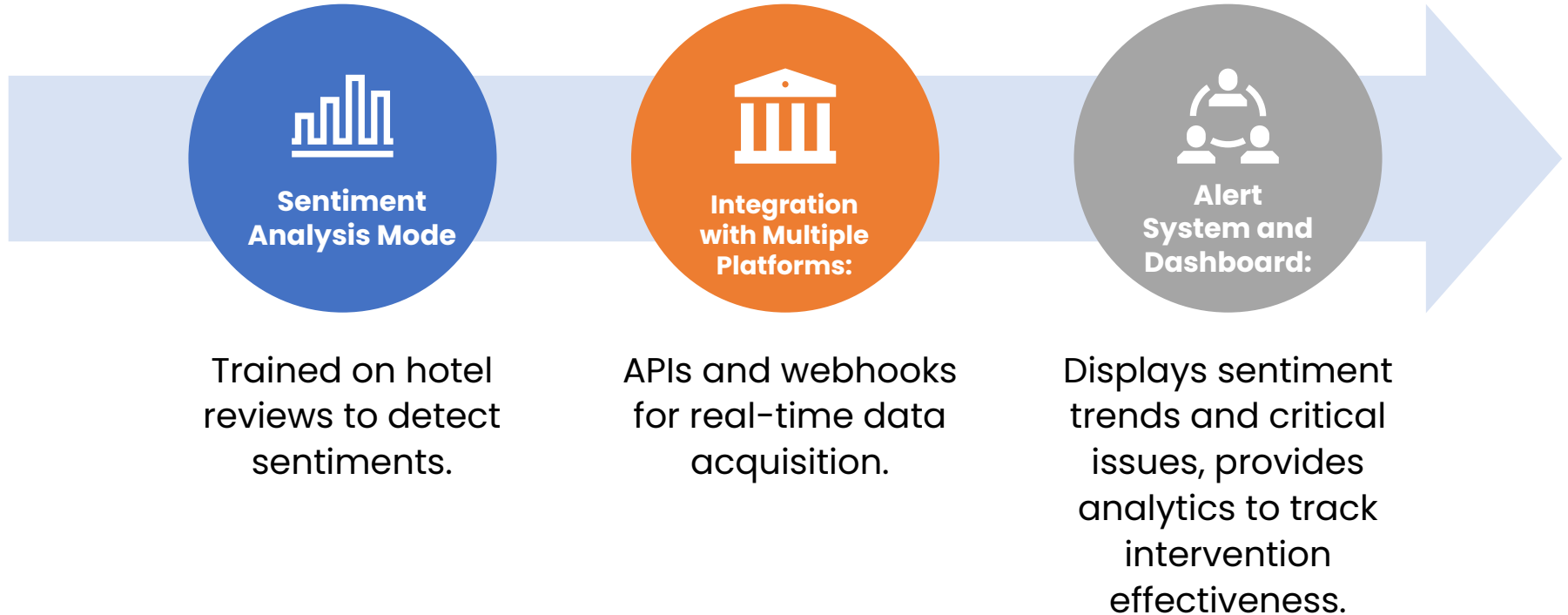
02

Task Force for Rapid Response: Specialized team addressing feedback immediately, enhancing guest satisfaction.

03

Feedback Loop and Continuous Improvement: Analyzes response effectiveness, implements targeted improvements.

Proposed Data Science Solution Approach



Dataset Utilization

Dataset: Hotel Reviews
Booking.com (Kaggle):
Comprehensive
dataset of hotel
reviews.













Detailed textual
feedback from
guests.



Reviews from
various hotels
worldwide.



Includes reviewer
nationality, stay
duration, trip type.

PREDICTION TASK  <p>The task is sentiment analysis aimed at detecting negative feedback from hotel reviews in real-time. The outcome observed is whether a review has a negative sentiment, which needs to be addressed promptly to avoid impact on bookings and brand loyalty.</p>	DECISIONS  <p>The predictions are used to identify negative sentiments, which trigger alerts for a rapid response team. This helps in taking immediate action to address guest dissatisfaction.</p>	VALUE PROPOSITION  <p>The end-users are hotel management teams who benefit from real-time insights into guest sentiments. The system helps in promptly addressing negative feedback, thereby enhancing guest satisfaction and maintaining brand loyalty.</p>	DATA COLLECTION  <p>Data collection involves real-time acquisition of detailed textual feedback from various hotels worldwide through APIs and webhooks.</p>	DATA SOURCES  <p>The data sources include a comprehensive dataset of hotel reviews from Booking.com available on Kaggle. This includes metadata like reviewer nationality, stay duration, and trip type.</p>
IMPACT SIMULATION  <p>The models can be deployed in a real-time environment, with performance assessed using test data. Evaluation metrics include precision, recall, and ROC-AUC to ensure the models are accurately identifying negative sentiments.</p>	MAKING PREDICTIONS  <p>Predictions are made in real-time as new reviews are posted. Time for feature extraction, post-processing, and computing the sentiment prediction is minimized to enable prompt responses.</p>	MONITORING  <p>The system includes dashboard that displays sentiment trends and critical issues, providing analytics to track the effectiveness of interventions.</p>		
			BUILDING MODELS  <p>The models used are Logistic Regression, SVM, and Random Forest. These are chosen for their capability to classify sentiments based on textual data.</p>	FEATURES  <p>Input features include text representations such as TF-IDF, Bag of Words, and nGrams. These features help in identifying important words and understanding the context within the feedback.</p>

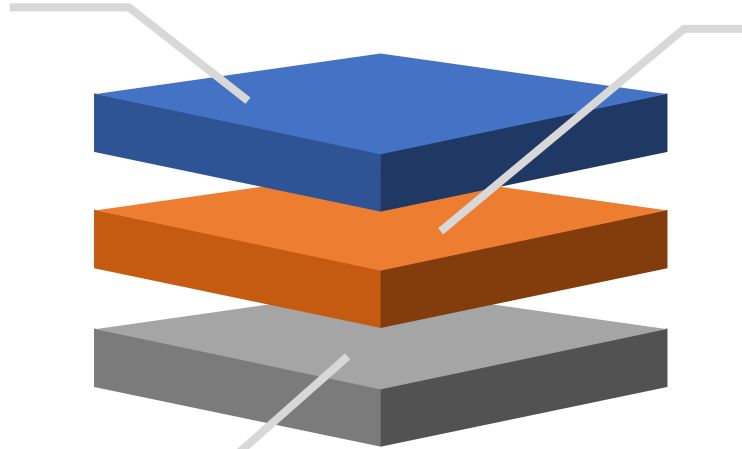
Text Representation Techniques

Bag of Words:

Simple model capturing word frequencies.

nGrams:

Captures sequences of words to understand context.



TFIDF

TFIDF (Term Frequency-Inverse Document Frequency): Highlights important words by reducing weight of common words.

THANK YOU
MOTEL

OFFICE

VACANCY

