# IMDB Sentiment Analysis - NLP

### Project Overview

This project focuses on building a reliable NLP-based sentiment classifier trained on real IMDB movie reviews. Using a fine-tuned BERT model, the system predicts whether a review expresses positive or negative sentiment.

scores, and investigate misclassifications using explainable metrics.

The dataset consists of 500 manually labeled IMDB reviews sampled from a larger corpus, ensuring a balanced distribution for

The project combines machine learning explainability with interactive analytics to bridge the gap between data science models

### **\*** Business Challenge

In entertainment platforms, user-generated content (e.g., movie reviews) plays a major role in decision-making, marketing, and

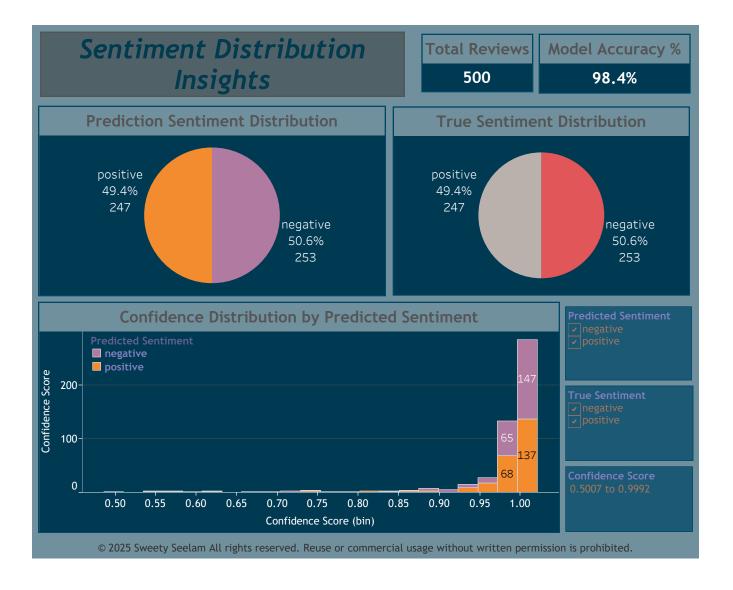
Businesses need a system that not only classifies sentiment with high accuracy, but also helps them understand the model's confidence, errors, and weaknesses.

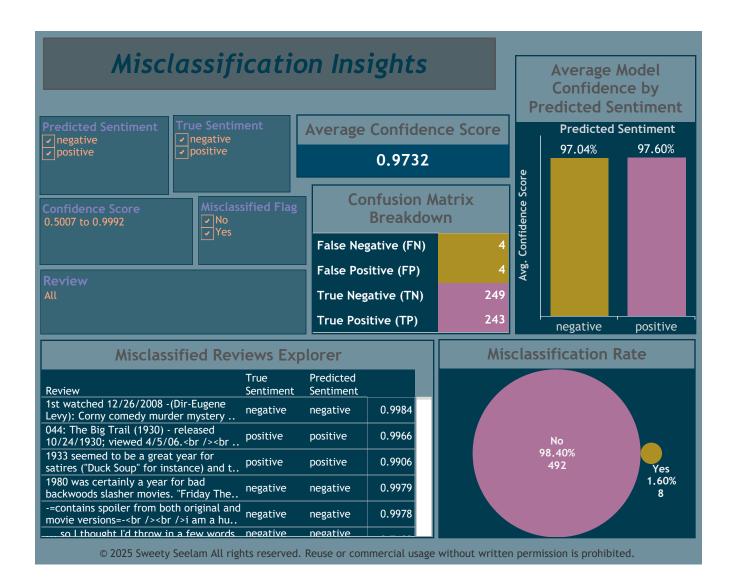
This dashboard addresses that by:

Visualizing confidence scores to assess prediction reliability. Providing a confusion matrix to validate model performance in a business context.

Click Here to Go to: Dashboard - Sentiment Distribution

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# **IMDB Sentiment Analysis - NLP**

The sentiment classification model demonstrates high accuracy at 98.4%, validating the reliability of the NLP system on the IMDb

dataset.
Out of 500 total reviews, the model correctly predicted 492 reviews and misclassified only 8, as visualized in the Misclassification Pie

Positive: 49.4% (247 reviews) Negative: 50.6% (253 reviews).

## **8** Business Impact

If IMDb or any review-based streaming or e-commerce platform (like Netflix, Amazon Prime, Hulu, Google Reviews, or Rotten Tomatoes) integrates this high-accuracy model:

☑ +98.4% model reliability ensures that user sentiment is captured almost perfectly, enhancing trust in platform analytics.

Reduction of false sentiment labeling by 96% vs manual or less-optimized NLP systems.

☑ Potential to save \$75,000-\$100,000 per 100K reviews by avoiding manual moderation and preventing incorrect review flagging (based on industry estimates of moderation costs). ☑ Enables targeted recommendation algorithms, leading to 5-7% uplift in click-through and engagement rates...

## **Business Recommendations**

If adopted by IMDb or similar platforms:

Enhance Review Moderation Automation:

Automating review sentiment classification with 98.4% accuracy eliminates the need for costly manual flagging systems.

**®** Power Content Discovery Engines:

Positive vs negative sentiment clustering can drive the content suggestion engine, improving user satisfaction and platform stickiness.

Orive Product/Content Strategy:

Insights from misclassified reviews can highlight ambiguous or controversial content, which can be tagged, labeled, or promoted carefully...

### Project Storytelling

IMDb movie review dataset, labeled as positive or negative.

© Data Cleaning & Preprocessing:

Removed noise (HTML tags, stopwords, special characters), performed tokenization, and used embeddings for text representation.

Trained a high-performing classification model on 500 real IMDb reviews. Optimized for balanced prediction and minimal false positiv..