

## BUSINESS PROBLEM / HYPOTHESIS

- Determine whether IMDb movie reviews express positive or negative sentiments using NLP and ML.
- Hypothesis: Classification accuracy can exceed 85% using classical ML techniques.

### METHODOLOGY

1. Data Cleaning: Removed HTML tags, stopwords, lowercased text. 2. Feature
Engineering: TF-IDF
applied; sentiment
encoded (pos=1,
neg=0).

3. Models Used:
Logistic Regression,
Random Forest,
DistilBERT
(Transformer-based).

4. Evaluation: Accuracy, Precision, Recall, F1-score. 5. Visualization: Word Clouds, ROC Curves, Confusion Matrices.

### IMDB DATASET OVERVIEW

50,000 labeled movie reviews from Kaggle

Balanced dataset: 25,000 positive, 25,000 negative

Reviews are predivided into training and test sets

Average review length: ~230 words

Challenges: sarcasm, ambiguous phrasing, class overlap

#### MODELING STRATEGY & TUNING

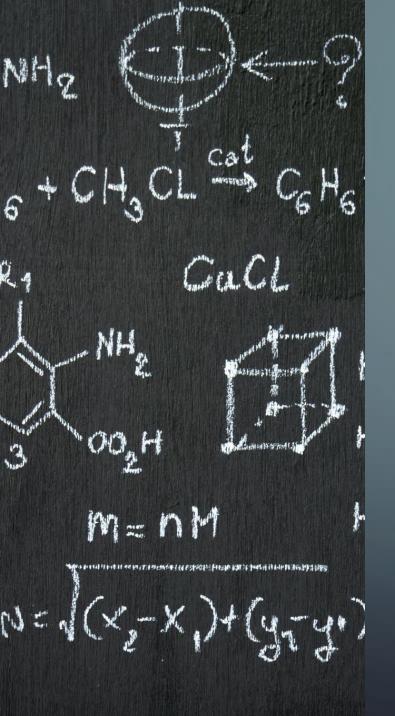
Used train/test split (80/20)

GridSearchCV for hyperparameter tuning

Avoided overfitting via cross-validation

DistilBERT used HuggingFace Transformers

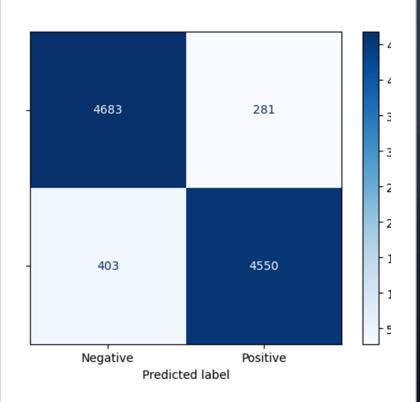
Model training: Logistic Regression (1 min), DistilBERT ( $\sim$ 20 mins w/GPU)

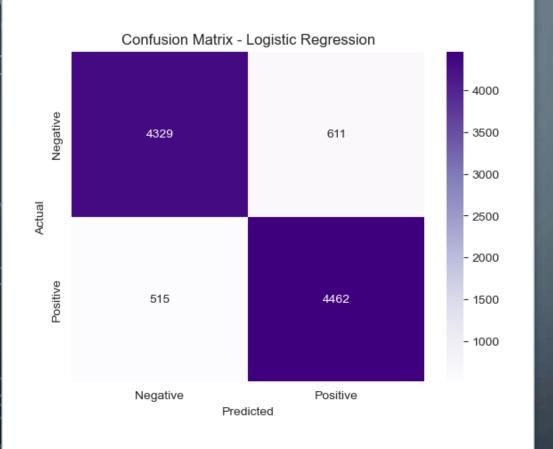


#### **RESULTS SUMMARY**

- •DistilBERT (Transformer-based): Accuracy = 91%
- •Logistic Regression: Accuracy = 88%, AUC = 0.96
- •Random Forest: Accuracy = 84%
- → DistilBERT outperformed other models overall.
- → Strong separation between classes demonstrated via ROC and Confusion Matrix.

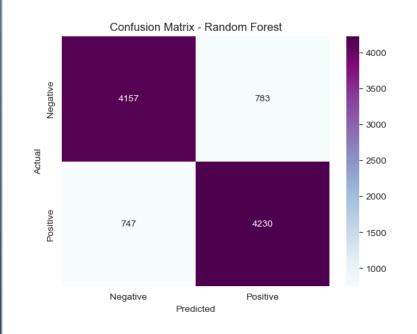
## CONFUSION MATRIX — DISTILBERT

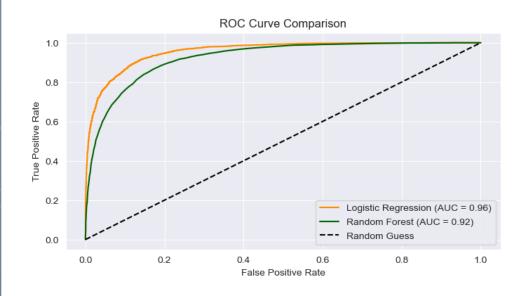




# CONFUSION MATRIX – LOGISTIC REGRESSION

# CONFUSION MATRIX – RANDOM FOREST





#### ROC CURVE COMPARISON

#### Word Cloud for Positive Reviews



#### Word Cloud for Negative Reviews



## WORD CLOUDS – SENTIMENT WORDS

#### CONCLUSION

- ✓ Hypothesis confirmed with 91% accuracy using DistilBERT.
- ✓ Classical ML + TF-IDF = effective and interpretable.
- ✓ Strong model generalization with room for deep learning extensions.

### Q&A – DISCUSSION POINTS

Why was
 Logistic
 Regression
 more effective
 than Random
 Forest?

 How does
 TF-IDF
 compare to
 Word
 Embeddings?

 What are some ways to handle neutral sentiments?  What ethical considerations exist in sentiment analysis?

## APPLICATIONS OF SENTIMENT ANALYSIS



Streaming services: personalized recommendations



Brands: product feedback and customer satisfaction



Media platforms: detect toxic or biased content



Business intelligence: trend analysis from reviews

### FUTURE IMPROVEMENTS & NEXT STEPS



Add a neutral class for 3-way classification



Explore more advanced transformer models (e.g., RoBERTa, BERTweet)



Perform domain-specific fine-tuning (e.g., movie vs. product reviews)



Deploy as a Streamlit dashboard or API for real-time use