# Detecting Hate Speech in the Digital Age

- An ML-driven approach to identifying and mitigating harmful content.
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# Why This Matters Now

- Protect vulnerable groups from harassment and discrimination.
- Prevent online speech from escalating to real-world violence.
- Maintain platform trust, safety, and reputation.

# Kenya's Unique Challenge

- Twitter is a key platform for political discourse in Kenya.
- Risks: Hate speech, tribalism, and incitement during elections.
- Challenge: Manual moderation is slow, biased, and misses local nuances.
- Solution: Automated ML classification model for early detection.

### Our Mission & Goals

- Identify common inciting terms and phrases.
- Build a multi-class classification model (Hate Speech, Offensive, Neutral).
- Evaluate algorithms using F1 Score.
- Deploy best-performing model for real-time detection.

## Inside the Data

- Kenyan election-related hate speech dataset from Kaggle.
- Labels: Hate Speech, Offensive Language, Neutral.
- Thousands of annotated tweets for robust training.

#### How We Clean the Data

- Remove irrelevant characters, URLs, tags, and mentions.
- Tokenization into words/sub-words.
- Stop word removal and lemmatization.
- Standardizing text for better model understanding.

# Turning Text into Features

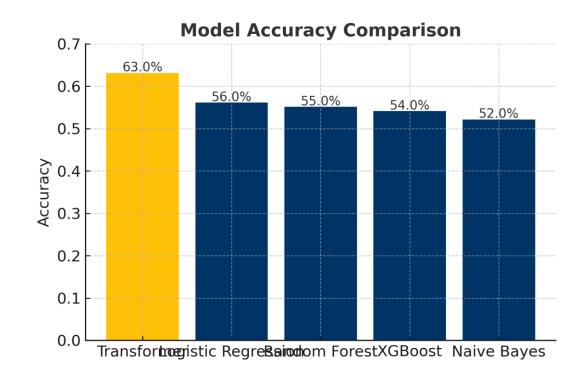
- TF-IDF scores: Highlight important terms.
- Sentiment polarity: Detect emotional tone.
- Word embeddings: Capture semantic meaning.
- Lexical features: Text length, emoji presence.

## Models We Tested

- Logistic Regression strong baseline, efficient.
- Naive Bayes probabilistic text classifier.
- Random Forest rule-based tree model.
- XGBoost high-performing ensemble method.
- RoBERTa Transformer contextual deep learning model.

# Which Model Wins?

- Logistic
  Regression: Best
  overall accuracy.
- RoBERTa: Strong on Hate Speech & Neutral detection.
- Both models show strong balance between precision and recall.



### From Model to Action

- Deploy best model via Streamlit web app.
- Real-time tweet classification.
- User-friendly dashboard for moderators.

### What We Learned

- ML models can detect harmful speech with high accuracy.
- TF-IDF and sentiment analysis are valuable features.
- Scalable system suitable for high-volume platforms.

#### Where We Go From Here

- Expand dataset for broader coverage.
- Improve model for detecting nuanced, local expressions.
- Integrate into social media monitoring systems.