# Final Report: VeritasVigil — The Truth Watchman

## **Custom Tokenizer Design**

## Goals:

- Handle informal text commonly found in online news content.
- Normalize elongated/repeated characters (e.g., sooo → so + REPEAT:3).
- Process emoticons, contractions, and punctuation.
- Lowercase and split into meaningful tokens

A manual dictionary is used to expand common contractions:

```
"don't" \rightarrow "do not", "it's" \rightarrow "it is", "i'm" \rightarrow "i am"
```

# Rule-Based POS Tagger

POS	Rule
VERB	Words ending in -ing, -ed
NOUN	Words ending in -tion, -ity, -ment, -ness
ADJ	Words ending in -ous, -ful, -able, -ive

## Eg: running -> stemmed to run

.

## **Custom Lemmatizer**

#### Goals:

- Reduce words to base form considering POS context
- Avoid stemming errors like "better" → "bett"

## Methodology:

- VERB:
  - $\circ$  Remove -ing, -ed endings (e.g., running  $\rightarrow$  run)
- NOUN:
  - Remove -ness

## **Feature Extraction**

## **Methods Used:**

- 1. Bag-of-Words (BoW):
  - o Frequency count of tokens
  - o Captures term occurrence
- 2. TF-IDF (Term Frequency-Inverse Document Frequency):
  - Weighs terms that are frequent in a document but rare in corpus
  - o Highlights unique keywords

## Classification

#### **Models Trained:**

- 1. Multinomial Naive Bayes (NB):
  - o Fast, probabilistic model for text data

## 2. Linear Support Vector Machine (SVM):

o Margin-based classifier for high-dimensional spaces

## **Impact Analysis**

## **Repeated-Character Normalization**

- Helps model understand exaggeration and sentiment often present in fake news headlines or user-like stories.
- Improves vocabulary normalization by collapsing redundant variants (e.g., "gooood", "goood", etc.)

## **POS-Guided Lemmatization**

- Prevents over-truncation compared to stemmers.
- Enhances semantic clarity by converting words only when contextually appropriate.