Homonyms Problem in the text Use Case

1. Introduction

1.1.Problem Statement

In Natural Language Processing (NLP), homonyms have the same spelling but different meanings depending on the context in which they are used. This creates significant challenges for sentiment analysis tasks, where understanding a given sentence's sentiment (positive or negative) is essential. A common issue is when words like "hate" appear in sentences with opposite sentiments due to different contextual meanings. For example:

- Sentence 1: "I hate the selfishness in you" The sentiment here is negative because "hate" refers to a strong dislike.
- Sentence 2: "I hate anyone who can hurt you" This sentence also includes the word "hate," but its meaning in context is closer to a protective or caring sentiment, leading to a positive label.

This discrepancy arises from the homonym "hate," which has different emotional implications depending on its context. The challenge lies in accurately capturing the context to resolve such ambiguities and determine the correct sentiment.

1.2. Objectives

- To develop a solution that addresses the issue of homonyms and contextual ambiguity in sentiment analysis.
- To implement a machine learning pipeline capable of distinguishing between different meanings of words like "hate" in varying contexts.
- To improve sentiment classification performance by enhancing the model's ability to disambiguate homonyms based on the surrounding context.

2. Data Processing

The goal is to preprocess the text data efficiently, it involves serval steps:

- Text Cleaning: Remove any unwanted characters, punctuation, HTML tags, or special symbols, lowercasing.
- Text Tokenization: this step in handling textual data is breaking down the sentences into individual tokens (words). This is important because homonyms often rely on their surrounding words to derive the correct meaning.
- Contextual Representation: Use pre-trained language models like BERT or RoBERTa to generate contextual embeddings for each word in a sentence. These models are capable of capturing the nuanced meanings of words based on their surrounding context, helping to resolve homonym ambiguities.

3. Sentiment Analysis

The goal of sentiment analysis is to classify sentences into categories such as positive, negative, or neutral. However, for sentences containing homonyms like "hate," the sentiment classification model must:

- Disambiguate the word "hate" based on context.
- Leverage contextualized word embeddings to differentiate between instances of the word "hate" used in positive or negative contexts.
- Predict the sentiment by considering the overall meaning of the sentence rather than focusing solely on individual words.

4. Tools Used

- Pre-trained Language Models (e.g., BERT, RoBERTa)
- NLP Libraries (Hugging Face Transformers)
- Text Processing Tools

5. External Resources

- Pre-trained Models
- NLP Research Papers

6. Project Learnings

6.1. Challenges

• Contextual Ambiguity: Homonyms often have multiple meanings based on context, and resolving this ambiguity is difficult for traditional sentiment analysis models. For example,

"I hate the selfishness in you" and "I hate anyone who can hurt you" may both use the word "hate," but with opposing sentiment meanings.

7. Conclusion

The key to solving the homonym problem in sentiment analysis lies in leveraging advanced contextual models like BERT, Roberta, or GPT, which capture subtle variations in meaning depending on surrounding words. By carefully processing data, fine-tuning models, and addressing the challenges of ambiguity and context, we can improve sentiment classification, even when homonyms are involved.