NLP HW1

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1. How do you select features for your model input, and what preprocessing did you perform to review text?

Ans:

In my model, I selected the following features for input:

- 1. Text features: I combined the "title" and "text" of the product review, then converted them to lowercase. I used CountVectorizer to transform the text into a bag-of-words representation with a maximum of 10,000 features.
- 2. Verified purchase: I encoded this categorical feature into numerical values using LabelEncoder.
- 3. Helpful votes: This numeric feature was included directly.

For preprocessing the review text, I applied the following steps:

- 1. Combined the title and review body.
- 2. Converted all text to lowercase.
- 3. Used CountVectorizer to convert the text into numerical features by counting word occurrences, limiting the feature set to the top 10,000 most frequent words.
- 2. Please describe how you tokenize your data, calculate the distribution of tokenized sequence length of the dataset and explain how you determine the padding size.

Ans:

In my code, I did not explicitly tokenize the text data. Instead, I used CountVectorizer from scikit-learn, which treats the text as a bag of words, splitting the text into individual words based on spaces. The vectorizer then

converts the text into a sparse matrix of word counts, with a maximum of 10,000 features representing the most frequent words in the dataset. Since I treat the text as bag-of-words features, there is no concept of sequence length or token padding.

3. Please compare the impact of using different methods to prepare data for different rating categories.

Ans:

Bag-of-Words: Simply counts word occurrences, treating words independently of each other.

Pros: Fast and simple, often effective for small datasets.

Cons: Does not capture word context or relationships between words, which can limit its ability to differentiate subtle sentiments in reviews.

Word Embeddings: Captures semantic relationships between words and context.

Pros: Often results in better performance by understanding the sentiment or meaning behind reviews, which is crucial for correctly predicting different ratings.

Cons: More computationally expensive and requires larger datasets.