

# Sentiment Analysis API with LLM Integration

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## 1. Approach to Solving the Problem

The goal is to implement an API that allows users to upload a CSV or XLSX file containing customer reviews, and then return sentiment analysis results (positive, negative, neutral) for each review using the Groq API for enhanced LLM-based analysis.

Key Considerations:

- - File Handling: The API should support common file types like CSV and XLSX, handling potential errors like missing columns or incorrect formats.
- - Sentiment Analysis: Using a Large Language Model (LLM) through the Groq API to classify reviews as positive, negative, or neutral.
- - Structured Response: The API must return results in a clear and structured JSON format, allowing for easy integration into other systems.
- - Error Handling: Robust error management is crucial to handle potential issues with file upload or processing.

## 2. How the Structured Response Was Implemented

The structured response is implemented using Flask's jsonify function, which returns a JSON object after analyzing the sentiment of uploaded reviews. The response is organized to include a count of positive, negative, and neutral reviews, ensuring clarity and ease of interpretation.

Implementation:

```
return jsonify({
    "positive": sentiment_counts["positive"],
    "negative": sentiment_counts["negative"],
    "neutral": sentiment_counts["neutral"]
})
```

Example Output:

```
{
  "positive": 5,
```

```
"negative": 3,  
"neutral": 2  
}
```

### 3. Examples of API Usage with Sample Inputs/Outputs

#### Example 1: Positive Sentiment Review

Input Review: "This product exceeded my expectations!"

Sentiment Analysis: Positive

Output:

```
{  
  "positive": 1,  
  "negative": 0,  
  "neutral": 0  
}
```

#### Example 2: Negative Sentiment Review

Input Review: "The product broke within a week, very disappointed."

Sentiment Analysis: Negative

Output:

```
{  
  "positive": 0,  
  "negative": 1,  
  "neutral": 0  
}
```

#### Example 3: Neutral Sentiment Review

Input Review: "It does what it's supposed to do, nothing more, nothing less."

Sentiment Analysis: Neutral

Output:

```
{
```

```
"positive": 0,  
"negative": 0,  
"neutral": 1  
}
```

## 4. Analysis of Results, Including Limitations and Potential Improvements

### Strengths:

- - Ease of Use: The API simplifies sentiment analysis by allowing users to upload a file and quickly obtain results.
- - Structured Output: JSON responses are clear, making it easy to integrate into other applications.
- - Error Handling: The API checks for file types and column presence, ensuring robustness.

### Limitations:

- - Basic Sentiment Categories: The sentiment classification is limited to positive, negative, and neutral. More nuanced sentiments, like "mixed" or "sarcastic," aren't captured.
- - Keyword Bias: In cases where simple keyword-based analysis is used, it might misclassify sentiment, as it does not account for context.
- - LLM Limitation: The Groq API integration provides better context analysis, but performance may vary depending on the model used.

### Potential Improvements:

- - Advanced Sentiment Categories: Incorporating more sentiment categories (e.g., very positive, very negative) can enhance the granularity of analysis.
- - Multilingual Support: Expanding support to handle reviews in multiple languages would make the API more versatile.
- - Batch Processing Optimization: For larger datasets, the API could benefit from optimizations in the LLM processing to reduce latency.
- - Customization: Allowing users to set certain parameters, such as sensitivity to negative or positive sentiment, could make the tool more flexible.

## 5. Additional Insights and Observations

### Observations:

- - Handling Emojis and Special Characters: The Groq API manages emojis and symbols well, which often add context to sentiment (e.g., "😄" indicating positivity). It's important to ensure these are processed correctly.

- - Flexibility: By utilizing LLMs, this solution moves beyond keyword-based analysis, providing more accurate and context-sensitive classification.

### Output:

```
1 {  
2   "negative": 23,  
3   "neutral": 7,  
4   "positive": 21  
5 }
```

**Github Link:** <https://github.com/AravindReddy16/Sentiment-Analysis-API-with-LLM-Integration>