In modern software development, code comments play a vital role in enhancing the readability and maintainability of codebases. They provide essential context and explanations that aid developers in understanding complex code, facilitating collaboration, and expediting the debugging process. However, despite their significance, many codebases are plagued with inconsistent or unclear comments, leading to misunderstandings, increased onboarding time for new team members, and challenges in code maintenance and scalability.

Traditional static analysis tools predominantly focus on analyzing code structure and identifying potential security vulnerabilities but often neglect the quality of code comments. This oversight creates a gap where the natural language aspects of code documentation are not adequately assessed, leaving room for significant improvement in code comprehension and developer productivity.

This project aims to address this gap by developing a tool that assesses and ensures the clarity, relevance, and consistency of code comments within a codebase. Leveraging the capabilities of Large Language Models (LLMs) like GPT-4, the tool evaluates comments against predefined quality metrics, assigning scores and providing actionable feedback to developers.

The proposed solution works by parsing code files to extract comments and their associated code blocks. Utilizing an LLM, it assesses the quality of these comments based on three key attributes: clarity, relevance, and consistency. The tool then offers suggestions for improvement, enabling developers to enhance their documentation practices effectively.

By improving comment quality, the tool aims to enhance code readability and maintainability, ultimately facilitating better collaboration among developers. Maintaining high documentation standards is crucial for efficient development and scaling of software projects, and this tool provides a practical approach to achieving that goal.

The Comment Quality and Consistency Checker is designed with a modular architecture that facilitates ease of maintenance and scalability. The tool comprises three primary components:

- 1. Code Parser and Comment Extractor
- 2. LLM Evaluation Module
- 3. Feedback Generator

The tool utilizes Python's built-in ast module to parse source code files. By traversing the abstract syntax tree, it identifies nodes corresponding to functions, classes, and modules. For each node, the tool extracts:

- Docstrings: Using ast.get\_docstring(node), capturing multi-line comments that describe functions, classes, or modules.
- Associated Code Blocks: Extracted via ast.get\_source\_segment(source, node), providing context for the evaluation.

Inline comments are also extracted using regular expressions that match patterns starting with #. These comments are associated with nearby code statements to provide context during evaluation.

The tool integrates with the OpenAI API to leverage the capabilities of advanced language models. The integration process involves:

- API Authentication: The API key is securely loaded from environment variables, ensuring that sensitive information is not hard-coded.
- Prompt Construction: For each comment and associated code block, a structured prompt is created to guide the LLM's evaluation.

```
def evaluate_comment_quality(comment, code_block):
Uses OpenAI's LLM to evaluate the quality of a comment.
Evaluate the following comment for clarity, relevance, and consistency with the associated code block.
Comment:
{comment}
Code Block:
{code_block}
Provide a score from 1 to 5 for each category:
- Clarity
Also, provide suggestions for improvement if necessary.
response = client.chat.completions.create(
    model='gpt-4o-mini',
    messages=[
        {'role': 'system', 'content': 'You are an assistant that evaluates code comments.'},
        {'role': 'user', 'content': prompt}
    max_tokens=200,
    temperature=0
```

- API Request: The tool sends a request to the ChatCompletion endpoint, specifying the model (e.g. gpt-4o-mini) and including the prompt in the required format.
- Response Handling: The LLM's response is parsed to extract the scores and suggestions, which are then processed for presentation.

The tool assesses comments based on three criteria:

- Clarity: The ease with which the comment can be understood.
- Relevance: How well the comment relates to and describes the associated code block.
- Consistency: The uniformity of commenting style throughout the codebase.

Each criterion is scored on a scale from 1 to 5, with 5 representing the highest quality.

After evaluation, the tool generates feedback that includes:

- Scores: Numerical values for clarity, relevance, and consistency.
- Suggestions: Specific advice on how to improve the comment, such as elaborating on details, correcting inaccuracies, or adopting a consistent style.

The feedback is formatted and presented in the console output for easy review by the developer.

The tool was developed using the following technologies:

- Python 3.8: The primary programming language for its readability and extensive standard library.
- OpenAl Python Library: Used to interact with the OpenAl API and LLMs.
- AST Module: For parsing and analyzing Python source code.
- Regular Expressions (re module): To extract inline comments via pattern matching.

