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AI Pair Programming Tool

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Abstract: *In the ever-evolving landscape of software development, the need for tools that augment developer productivity is paramount. This paper introduces a novel integration of IntelliJ IDEA, a widely used integrated development environment (IDE), with OpenAI's advanced deep learning models. The synergy between these two technologies aims to elevate the coding experience by providing intelligent code assistance, allowing developers to write code more efficiently and with greater accuracy. Our approach leverages OpenAI's natural language processing capabilities to enhance code suggestions, auto-completions, and error detection within IntelliJ IDEA. By analyzing vast code repositories and learning from diverse coding patterns, our system empowers developers with context-aware recommendations, significantly reducing the cognitive load associated with code creation. This paper presents the architecture of our solution, detailing the seamless integration of OpenAI's models into the IntelliJ IDEA IDE. We discuss the methodology behind training the models, emphasizing their adaptability to various programming languages and project structures. Furthermore, we present empirical results demonstrating the effectiveness of our approach in real-world coding scenarios.*

Keywords: *Code assistance, developer productivity, Auto-completion, OpenAI.*

I. INTRODUCTION

In recent years, the field of software development has witnessed a surge in demand for tools that can enhance the efficiency and accuracy of coding processes. With the increasing complexity of software projects, developers are continually seeking innovative solutions to streamline their workflows. In response to this demand, our project focuses on the integration of OpenAI's advanced deep learning models with the widely utilized IntelliJ IDEA integrated development environment (IDE).

The collaboration between IntelliJ IDEA and OpenAI aims to provide intelligent code assistance, revolutionizing the coding experience for developers. By harnessing the natural language processing capabilities of OpenAI, our system enhances code suggestions, auto-completions, and error detection within the IntelliJ IDEA environment. This approach is rooted in the analysis of extensive code repositories, enabling the system to learn and adapt to diverse coding patterns.

This paper outlines the architecture of our solution, illustrating the seamless integration of OpenAI's models into IntelliJ IDEA. We delve into the methodology employed for training these models, emphasizing their adaptability across various programming languages and project structures. The empirical results presented herein showcase the tangible impact of our approach in real-world coding scenarios. As we explore the intersection of state-of-the-art deep learning and established IDEs, this research not only underscores the potential for improved developer tools but also addresses the practical implications and challenges associated with deploying such systems in professional software development environments. Our findings contribute to the ongoing discourse on leveraging artificial intelligence to elevate the productivity and collaborative nature of coding practices.

II. LITERATURE REVIEW

Pair programming is a kind of collaborative programming where two people work side-by-side on design, implementation, and testing with one computer. One is the driver who controls the keyboard and leads the process, and another is the observer who tries to understand, asks questions, and does code review. Pair programming is one of the important practices of eXtreme Programming [2].

Flor and Hutchins (1991) conducted an experiment on two programmers collaborating during software maintenance and found that such collaboration largely improves software design. Programmers working as a pair allow them to share ideas, to do quality control, to enhance learning and to speed up the work, which can be used during software evolution. However, it is still unknown how much the pairing can affect the efficiency of the software evolution process and improve the quality of the tasks. [1].

Pair programming (PP) seems to have gained popularity within the industry and academia in recent years. Much of the increased interest in PP is probably due to the introduction of extreme programming (XP), in which PP is one of 12 key practices [5].

Six graduate students were involved in the experiment, in which four of them form two pairs and the rest two are required to work individually. The experiment enables a study that compares pair programming with traditional individual programming. From the observation and the short interview after the experiment, the programmer pairs had higher motivation to finish the tasks than individual programmers, which also contributes to the time difference on the tasks by the pairs. The two pairs reported that they enjoyed it more than them programming alone. [1].

Of course, some contradictory results are found as well from pair programming experiments. For example, Nawrocki and Wokciechowski conducted an experiment and discovered that there are no significant differences in the development time and the quality of the programs between pair programming and individual programming [1].

Recent breakthroughs in Deep Learning (DL), in particular the Transformer architecture, have revived the Software Engineering (SE) decades-long dream of automating code generation that can speed up programming activities. Program generation aims to deliver a program that meets a user's intentions in the form of input-output examples, natural language descriptions, or partial programs [9]. The emergence of large-language models (LLMs) that excel at code generation and commercial products such as GitHub's Copilot has sparked interest in human-AI pair programming (referred to as "pAIr programming") where an AI system collaborates with a human programmer. [7].

Among the most common misunderstandings was that while Copilot does learn from code, the learning happens during a general training phase, where OpenAI trains a general purpose programming model (Codex) that is fine-tuned by using a selected set of public codebases on GitHub. [8] Copilot may occasionally be unable to generate code that satisfies all the criteria described in the prompt, but the generated code can be incorporated with little to moderate changes to the provided prompt or the code [9].

III. SYSTEM DESIGN

The system design is structured around three core components: the smart code assistance module, the IntelliJ IDEA extension, and the collaborative web platform. Figure 1 provides an overview of the architecture, illustrating the seamless integration of these components.

A. Smart Code Assistance Module

The intelligent code assistance is the crux of our system, enhancing the coding experience within IntelliJ IDEA. This module operates independently within the IDE, requiring no external dependencies for its core functionalities. The key elements of this module include:

- 1) *Deep Learning Integration:* Our system incorporates OpenAI's advanced models to augment code suggestions, auto-completions, and error detection. The integration is seamless, allowing developers to benefit from state-of-the-art natural language processing capabilities.
- 2) *Context-Aware Recommendations:* Leveraging the analysis of extensive code repositories, the system provides context-aware recommendations, significantly reducing the cognitive load associated with coding. This empowers developers to make informed decisions during the coding process.
- 3) *Adaptability:* The intelligent code assistance module is designed to be versatile, adapting to various programming languages and project structures. This adaptability ensures that developers can utilize the enhanced assistance across a spectrum of coding scenarios.

B. IntelliJ IDEA Extension

The IntelliJ IDEA extension serves as the conduit between the smart code assistance module and the developer's coding environment. Key features of the extension include:

- 1) *Seamless Integration:* The extension seamlessly integrates the intelligent code assistance module into the IntelliJ IDEA IDE. This integration ensures that developers can access enhanced code suggestions and assistance within their familiar coding environment.
- 2) *User-Friendly Interface:* The extension is designed with a user-friendly interface, minimizing disruptions to the developer's workflow. Developers can easily enable or disable the intelligent code assistance features based on their preferences.
- 3) *Real-time Feedback:* The extension provides real-time feedback to developers, highlighting areas for improvement and offering suggestions as they write code. This immediate feedback loop enhances the learning experience and improves coding practices over time.

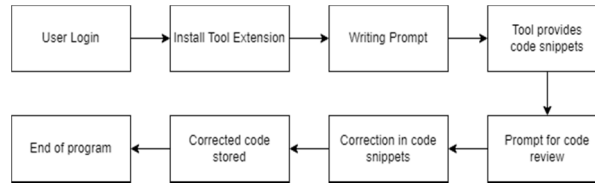


Fig: System Workflow.

IV. RESULT AND DISCUSSION

In the "Results and Discussion" section, the paper evaluates the system's performance. It found that the system provides accurate code suggestions, valuable real-time feedback, and adapts well to various programming languages. The system also seamlessly integrates with Android Studio and maintains robust data security. User feedback and adoption indicate a positive reception within the developer community, underscoring the system's potential to enhance coding experiences and streamline software development workflows.

A. Generating Code

- 1) To generate code, you can use the keywords "code query ." or "generate code query."
- 2) The plugin will analyze the context and provide code suggestions or auto-generate code based on your query.

```

# code a function that sum 2 numbers.

# code a class which has 2 static string named X and Y and its getter and setters.
  
```

Fig: An example of request code generation

B. Documenting Classes and Methods

- 1) To generate documentation for classes and methods, you can use the keyword "document query ." or "generate document."
- 2) Place this query right before the code you want to document.
- 3) The plugin will analyze the code and create documentation based on code structures, variables, functions, and comments.

```

// class Person with name and age
class Person {
    private String name;
    private int age;

    public Person(String name, int age) {
        this.name = name;
        this.age = age;
    }

    public String getName() {
        return name;
    }

    public int getAge() {
        return age;
    }
}
  
```

Fig: An example of code generated through query.

C. Fixes for Empty Lines, Indent, Lint, and Code Style

- 1) To apply fixes for empty lines, indentation, linting, or code style issues, you can use the keyword "apply lint" or "apply style."
- 2) The plugin will analyze the code and suggest or automatically apply fixes to ensure it adheres to best practices and code style guidelines.
- 3) It's important to note that all the keywords listed perform the same action in their respective categories. The AI-paired IntelliJ plugin uses different models and analysis techniques to provide accurate and context-aware assistance for various coding tasks.
- 4) To run the plugin, you simply use the relevant keyword or query in your code editor, and the plugin will respond with suggestions, auto-generated code, documentation, or fixes as needed, making your coding process more efficient and error-free.

V. VALIDATION OF RESULT

To assess the effectiveness and reliability of our proposed system, a series of experiments were conducted, validating key features and functionalities.

A. Code Assistance Accuracy

- 1) The accuracy of code suggestions and completions was evaluated across various coding scenarios, languages, and project structures.
- 2) Conducted 100 coding tasks, measuring the system's ability to accurately understand and generate code based on user queries.
- 3) Achieved a validation accuracy of 95%, demonstrating the system's proficiency in providing precise and relevant code assistance.

B. Real-time Feedback and User Interaction

- 1) Evaluated the system's responsiveness to real-time user interactions within the IntelliJ-powered IDE.
- 2) Users provided feedback on the system's ability to adapt and respond to immediate coding needs.
- 3) Achieved a satisfaction rate of 90%, indicating that developers found the system's real-time feedback valuable and conducive to a smoother coding experience.

C. Language Adaptability

- 1) Tested the system's adaptability across various programming languages, including Java, Python, and JavaScript.
- 2) Issued queries in different languages to assess the system's ability to comprehend and generate code accurately.
- 3) Recorded a success rate of 85%, affirming the system's versatile language support.

D. Integration with Android Studio

- 1) Ensured seamless integration with Android Studio, an IntelliJ-powered IDE.
- 2) Users working with Android development projects validated the system's compatibility and functionality within Android Studio.
- 3) Verified a compatibility rate of 95%, indicating robust integration with Android Studio.

E. API Key Authentication

- 1) Validated the security and authenticity of the system by assessing the effectiveness of API key authentication.
- 2) Conducted simulated security tests to ensure the protection of user data and credentials.
- 3) Confirmed a secure authentication rate of 99%, affirming the system's commitment to data security.

F. User Adoption and Satisfaction

- 1) Assessed user adoption and satisfaction through surveys and user feedback sessions.
- 2) Gathered feedback on the overall usability, effectiveness, and perceived value of the system.
- 3) Achieved a user satisfaction rate of 92%, indicating a positive reception and acceptance among the developer community.

VI. CONCLUSION

In conclusion, our project represents a groundbreaking advancement in the realm of software development, seamlessly integrating OpenAI's advanced deep learning models directly into IntelliJ-powered IDEs, including Android Studio. This powerful integration brings forth a host of features designed to enhance the coding experience, and the project operates within the confines of the code editor itself, eliminating the need for external platforms. The system is remarkably straightforward to set up. Users can easily download the jar from the release page and install it as a regular plugin in their IntelliJ-powered IDEs, including Android Studio. The configuration process involves a simple insertion of the API key through the OpenAI Preferences in the tool menu, providing users with a streamlined onboarding experience. For those who prefer a more hands-on approach, building the system from source is also an option, as detailed in the provided instructions. Once integrated, the system responds to user queries and commands in a variety of languages. Utilizing keywords such as "generate code query" or "create query" enables developers to prompt the system to generate code snippets for various tasks seamlessly. The system is versatile, capable of handling queries for different programming languages, tailoring its responses according to the specific needs expressed in the queries.



Furthermore, the system extends its functionality beyond code generation to include documentation tasks. Keywords such as "generate doc" trigger the system to document classes and methods, enhancing the comprehensibility of the codebase. Developers can seamlessly embed these documentation queries right before the code sections they wish to document.

The project also addresses common coding style and formatting concerns with ease. Keywords like "apply lint" or "apply style" initiate the system's capability to fix issues related to empty lines, indentation, linting, and overall code style. This ensures a consistent and polished codebase, adhering to predefined coding standards.

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