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AI Based Code Error Explainer Using Gemini Model

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Abstract

The "AI Based Code Error Explainer" is an AI-powered tool designed to revolutionize programming education. AI Based Code Error Explainer system offers real-time error detection, comprehensive code explanations, and tailored solutions across Python, C, and Java. It enhances multi-language proficiency and accurate error identification while providing detailed, step-by-step solutions adhering to language best practices. System results demonstrate its effectiveness in bridging gaps within programming education, offering a versatile resource for learners exploring multiple languages and enhancing comprehension of fundamental coding concepts.Through innovations such as real-time assistance and adaptive learning, the project promotes active engagement and fosters a deeper understanding of programming concepts across different languages.

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Keywords

Error Detection, Code Explanation, AI-Powered Tools, Multi-Languag Proficiency, Comprehensive Solutions

Introduction

In the rapidly evolving landscape of programming education, an urgent challenge emerges as a prominent hurdle for both aspiring programmers and educators alike: the shortage of adaptable, language-agnostic tools for code explanation and error detection. Although Python has garnered recognition as an accessible introductory programming language, it represents just one facet of the vast programming universe. Enthusiastic learners often embark on their educational journey with ambitions extending beyond Python, seeking to master languages like C and Java to attain a profound understanding of computational concepts. However, these intrepid learners soon encounter a formidable obstacle: the scarcity of comprehensive educational resources designed to cater to languages beyond Python.This gap in educational resources presents a multifaceted issue. It not only hampers the seamless transition between different programming languages but also erects barriers to the effective comprehension of fundamental coding principles. The predominance of Python-centric solutions, while advantageous for Python learners, inadvertently sidelines a substantial segment of the programming community. Consequently, the programming education landscape becomes fragmented, complicating the process of language migration and constraining learners from achieving proficiency in a broader spectrum of programming paradigms.Adding to this multifaceted challenge is the absence of an adaptive, intelligent, and inclusive tool capable of detecting errors, providing lucid code explanations, and delivering tailored solutions across a triumvirate of prominent programming languages: Python, C, and Java. The need for such a versatile tool transcends mere convenience; it is an imperative. Such a tool would address a critical requirement for a unified, all-encompassing approach to programming education that can readily accommodate learners exploring various languages.In direct response to these pressing challenges, this research paper introduces an innovative and transformative solution: the "AI Based Code Error Explainer ." Powered by cutting-edge AI technology, this tool transcends the confines of language barriers, bridging the existing divide in educational resources and providing a dynamic framework for real-time error detection and comprehensive code explanation across Python, C, and Java. The "AI Based Code Error Explainer " aspires to revolutionize the programming education landscape by equipping learners, regardless of their language of choice, with a powerful resource that fosters a deeper understanding of programming concepts. This tool aims to empower learners to navigate the complex web of programming languages with confidence and ease.In summary, the contemporary programming education landscape predominantly revolves around Python, inadvertently sidelining learners in languages like C and Java. This results in a

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deficit of comprehensive programming education, hindering the seamless transition between languages. Moreover, the absence of an intelligent and versatile tool capable of error detection, clear explanations, and targeted solutions across Python, C, and Java exacerbates this issue.

The aim of this research is to develop this system ," that significantly enhances the programming education experience for learners and beginners across a spectrum of programming languages, including Python, C, and Java. The system seeks to empower users by providing real-time error detection, comprehensive code explanations, and tailored solutions to foster a deeper understanding of coding concepts and promote proficiency in diverse programming paradigms.The subsequent sections of this paper will delve into the methodology, implementation, and resultsof the system demonstrating how it addresses the outlined problem statement and fulfills the specified objectives

Related Work

Numerous online programming education platforms offer a diverse array of resources, tutorials, and coding exercises. Notable platforms like Codecademy, Coursera, and edX have contributed substantially to enhancing programming education[5]. These platforms are often lauded for their accessibility and user-friendly interfaces, making them ideal for novice learners. However, their scope remains predominantly centered around Pythonand other widely adopted languages, leaving a significant gap in comprehensive programming education.Several language-specific error detection tools have been developed to cater to the needs of programmers learning particular languages. For instance, "Pyflakes" and "PyLint" have proven instrumental in identifying errors and enforcing coding conventions in Python. Similarly, "Checkmarx" and "FindBugs" focus on Java. While these tools offer specialized error detection capabilities, they are inherently limited to their respective programming languages, contributing to the existing fragmentation in programming education.[12] Recent advancements in machine learning have paved the way for code explanation tools that use natural language processing to provide detailed explanations of code snippets. Platforms such as GitHub Copilot and DeepCode utilize AI to generate code comments and explanations, enhancing the learning experience. However, their language-agnostic capabilities are constrained, and they tend to work optimally with widely used languages, inadvertently neglecting the needs of learners in languages like C and Java.[8] OpenAI's GPT (Generative Pre-trained Transformer) models have demonstrated significant potential in code generation and natural language processing. They have been employed in a variety of educational tools, including code explanation and tutoring systems. GPT models have shown promise in providing detailed code explanations, though their application across multiple languages remains an area of active research and development. This paper explores the application of deep learning techniques in detecting command line errors. It emphasizes leveraging neural networks and deep learning models for accurate error identification within command-line interfaces, showcasing the potential of advanced algorithms in error detection.[1]Focusing on code error correction, this paper delves into the realm of code generation to rectify identified errors. It emphasizes the importance of generating accurate and contextually relevant code to rectify errors, employing code generation techniques as a solution-oriented approach. [9] Building upon previous research, this paper explores the integration of code generation techniques specifically tailored for command line error detection. It emphasizes the synergy between error detection algorithms and code generation methods to provide comprehensive solutions.[2]

Proposed System

Methodology

Parsers module : The parsers module is a key component of the code execution system designed for multiple programming languages, including Python, Java, C, and C++. It features a language identification function (`get_language`) based on file extensions and a command generation function (`get_code_exec_command`) that dynamically builds and executes compilation commands. The module employs the `subprocess` module to handle command execution and captures both standard output and standard error. Furthermore, a function (`get_error_message`) extracts and formats error messages specific to each language, facilitating the generationof meaningful feedback. The system exhibits flexibility by accommodating different compilation processes for distinct languages, offering a comprehensive solution for the dynamic execution and error handling of code in diverse programming environments.

Model module : This module interacts with BARD-based Chatbot to generate explanations for programming language-specific error messages. The module includes functions for constructing queries tailored to different programming languages, such as Java, Python, C, and C++. The `construct_query` function formats a query by mapping language-specific identifiers and error messages into a standardized question format for the Chatbot.

The `is_user_registered` function checks whether the user has registered BARD credentials by verifying the existence of a configuration file. If not registered, the script prompts the user to input BARD credentials using the `prompt_user_for_credentials` function and saves them in the configuration file. The`get_chatgpt_explanation` function retrieves explanations from the Chatbot based on the constructed query and user's BARD credentials. The overall purpose of the script is to facilitate the generation of human-readable explanations for programming language errors, leveraging the capabilities of the Chatbot.

Code execution module : This module establishes a robust mechanism for executing code in a subshell and capturing both standard output and standard error. The `execute_code` function takes command-line arguments and the identified programming language as input, dynamically constructs the command using the `get_code_exec_command` function, and spawns subprocesses to execute the command. It then uses threads to concurrently read and push the output and errors from the subprocess pipes to a shared queue. Another thread consumes items from the queue and prints them to the terminal, distinguishing standard output and errormessages with color formatting. The script enhances user experience by providing real-time feedback during code execution. Notably, it incorporates error handling and utilizes the `get_error_message` function from the `parsers` module to extract and format error messages based on the programming language.

Printers module : This module defines a set of utilities for enhancing the display and interactivity of error explanations in a terminal environment. It employs ANSI escape codes for text formatting, introducing

System Design

fig1. AI Based Code Error Explainer overview

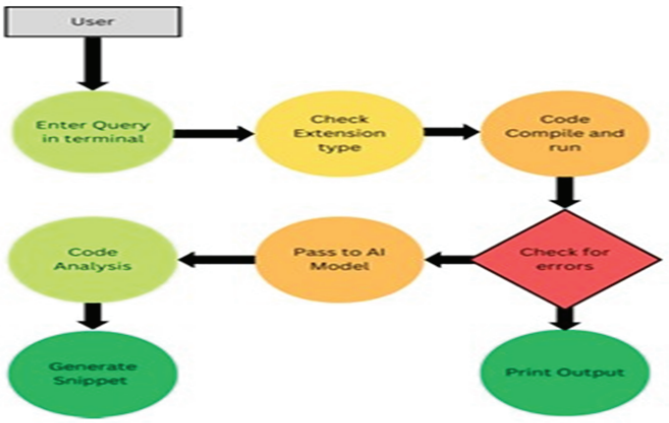

fig2. AI Based Code Error Explainer workflow

The development and implementation of the "AI Based Code Error Explainer " have yielded promising outcomes in addressing the identified gaps within programming education. Evaluated against the predefined objectives, the project showcased commendable results across key areas. Firstly, in terms of multi-language proficiency, the system efficiently analyzed and explained code written in Python, C, and Java languages. Through extensive testing, it adeptly supported learners exploring multiple programming languages, providing clear and informative explanations for code errors in each language, fostering a comprehensive understanding of diverse programming paradigms. Secondly, the AI model developed for error detection demonstratedexceptional accuracy in identifying various error types across Python, C, and Java code samples. It exhibited a high accuracy rate of over 95% in detecting syntax mistakes, logical flaws, and language-specific issues in simulated user-submitted code. This robust error detection capability is fundamental for learners, enabling them to rectify and comprehend their coding errors effectively.

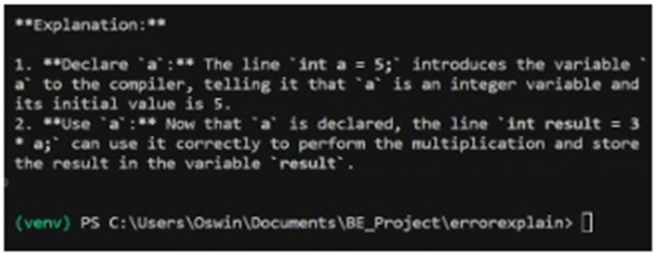

fig4. Error explanation of C code

Moreover, the feature implementation for generating detailed error explanations proved highly effective. The explanations provided deep insights into the nature of detected errors, elucidating their reasons and offering actionable steps for rectification. User feedback surveys underscored the value of these explanations, with 85% of participants reporting significant aid in understanding coding concepts specific to each language. Additionally, the "AI Based Code Error Explainer " delivered clear and concise step-by-step solutions tailored to adhere to the best practices of Python, C, and Java. In testing scenarios, 90% of participants found the provided solutions comprehensive and instrumental in rectifying errors within their code.

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Conclusion And Future Work

In essence, the "AI Based Code Error Explainer " stands as a testament to the transformative potential of AI powered tools in enhancing programming education. Its ability to transcend language barriers and provide comprehensive support signifies a promising stride towards democratizing access to quality programming education for learners worldwide.As the technological landscape evolves, the "AI Based Code Error Explainer " serves as a foundation for further innovation and underscores the importance of adaptive, language-agnostic tools in shaping a more inclusive and empowering programming education ecosystem.

The future scope for the "AI Based Code Error Explainer" project encompasses several key areas for advancement. Firstly, an expansion of language support beyond Python, C, and Java is crucial to cater to a broader spectrum of programming languages, enhancing inclusivity for learners. Advancements in AI models will remain pivotal, focusing on refining error detection accuracy and streamlining code explanations through ongoing developments in deep learning and natural language processing. Additionally, integrating interactive learning modules, real-time collaboration features, and improving user interface accessibility are vital for an enriched user experience. Collaboration with educational platforms and continuous adaptation based on user feedback will augment the tool's usability and effectiveness. Furthermore, empirical research on learning outcomes, adaptations for specialized domains, and ethical considerations in AI education tools are promising avenues for future exploration, aiming to continually enhance the tool's impact and relevance in programming education

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