



Deepsite: A Next-Generation Vibe Coding Platform for Browser-Based Application Generation

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Abstract

DeepSite (v4) is a cloud-native, browser-based development platform designed to enable intuitive software creation through natural language interaction. With the growing capabilities of Large Language Models (LLMs), software development is evolving from manual coding toward intent-driven design. DeepSite addresses the challenges of this transition by providing a structured environment for generating, modifying, and executing AI-produced code efficiently. The platform integrates the Next.js App Router with Hugging Face serverless inference APIs to stream incremental code updates in real time. Instead of regenerating entire files, a constrained Search/Replace mechanism is employed to apply precise modifications, reducing token usage and minimizing inconsistencies. For execution, the system utilizes browser-based Web Containers (Sandpack), allowing instant preview without local setup. Project persistence is achieved through Hugging Face Spaces, enabling Git-based version control and decentralized storage without relying on traditional databases. The platform ensures scalability, security, and rapid prototyping by combining client-side execution with lightweight backend orchestration. This work demonstrates a practical approach to AI-assisted development and highlights the potential of browser-based environments in accelerating modern software engineering workflows.

Keywords: Vibe Coding, Large Language Models, Browser-Based IDE, Next.js, Hugging Face APIs, Sandpack, Code Generation, Git-Based Storage, Real-Time Development, Decentralized Applications

1. Introduction

Software development has undergone a significant transformation with the advancement of intelligent systems, particularly Large Language Models (LLMs), which are capable of generating code from natural language instructions[1]. This shift is gradually moving the focus from manual coding practices to high-level problem description and architectural thinking[2]. However, despite these advancements, existing development environments still present several limitations, including complex setup procedures, dependency management issues, lack of real-time feedback, and difficulty in maintaining consistency across generated code. Traditional Integrated Development Environments (IDEs), while powerful, require considerable expertise and configuration, making them less accessible to beginners and non-technical users. Similarly, current AI-assisted coding tools often function as extensions or isolated generators, lacking a unified

system for continuous code execution, modification, and storage. These tools frequently regenerate entire codebases for minor updates, leading to inefficiencies, increased computational cost, and a higher risk of inconsistencies or errors. Additionally, many platforms rely on centralized storage mechanisms, which can limit scalability and transparency in version control. These challenges highlight the need for a more integrated, efficient, and accessible solution that can fully leverage the capabilities of AI in software development[3]. While AI-driven code generation improves development speed, maintaining accuracy and reliability of generated output remains a critical challenge. Uncontrolled or repetitive code generation can introduce inconsistencies, unnecessary modifications, and increased computational overhead, affecting the overall efficiency of the system. To address this issue, optimized generation strategies are required



to ensure precise and minimal updates. Techniques such as constrained code editing, where only specific sections of code are modified instead of regenerating entire files, provide a more efficient solution. This targeted approach reduces token consumption, minimizes errors, and preserves existing functional code. By combining controlled generation mechanisms with real-time validation and execution, the system ensures more stable, consistent, and reliable application development, thereby improving user confidence and overall performance. This paper presents DeepSite (v4), a browser-based AI development platform that combines natural language interaction with real-time code generation to enhance both efficiency and usability in software creation. Built using Next.js with a responsive and interactive interface, the system allows users to generate, modify, and deploy full-stack applications directly within the browser environment. Developers can observe live previews, manage project versions, and iteratively refine applications with minimal setup[5]. By integrating efficient code generation strategies, decentralized Git-based storage through Hugging Face Spaces, and secure browser-based execution using Web Containers, the platform improves reliability, scalability, and accessibility. The proposed approach not only simplifies the development process but also establishes a strong foundation for future AI-driven, user-centric software engineering environments[4].

1.1. Paper Overview

This paper presents DeepSite (v4), a browser-based AI coding platform designed to simplify and accelerate application development using natural language[6]. The organization of the paper is structured as follows: Section I introduces the background and key challenges in modern software development. Section II discusses related work in AI-assisted coding and browser-based development platforms. Section III defines the problem and limitations of existing tools. Section IV describes the proposed methodology, including the code generation and optimization approach. Section V details the system architecture and

technology stack. Section VI analyzes the system performance and results. Section VII provides a comparative evaluation with existing solutions. Section VIII concludes the paper with key contributions and future enhancements.

2. Literature Review

The evolution of software development has been greatly influenced by Large Language Models (LLMs), enabling automated code generation and intelligent programming assistance. Traditional development methods, though reliable, require significant manual effort, environment setup, and knowledge of multiple technologies. While Integrated Development Environments (IDEs) enhance productivity, they still rely heavily on developer input and lack full automation for complete application development. AI-assisted tools can generate code snippets and functions, but they often operate in isolation and do not support end-to-end application development. Early AI-based coding systems primarily focused on autocomplete and code suggestion mechanisms. These approaches improved coding speed but lacked contextual understanding of entire projects. With the advancement of transformer-based architectures, models such as GPT and other code-specific LLMs demonstrated the ability to generate structured programs across multiple files. Studies have shown that these models can assist in writing functions, debugging errors, and even creating full-stack applications. However, they often suffer from limitations such as hallucinated outputs, inefficient regeneration of entire codebases, and lack of integration with real-time execution environments. Recent research has shifted toward interactive and agent-based development systems, where AI can iteratively generate, modify, and validate code within a controlled environment[7]. Platforms integrating browser-based execution engines and cloud inference APIs have shown promising results in reducing development complexity[8]. Some studies highlight the use of Web Containers for secure, sandboxed execution, enabling users to run applications without local setup. Additionally, efforts have been made to optimize prompt engineering and reduce token

consumption, but many systems still rely on regenerating complete files for minor changes, leading to inefficiencies. In addition to generation efficiency, project storage and version management remain critical challenges. Traditional systems depend on centralized databases or cloud storage solutions, which can introduce scalability limitations and additional infrastructure overhead. Recent approaches have explored decentralized and Git-based storage mechanisms, allowing better version control and transparency. However, seamless integration of such storage systems with AI-driven development platforms is still limited. Despite these advancements, most existing solutions address only specific aspects of AI-assisted development, such as code generation, execution, or storage, rather than providing a unified platform[9]. There remains a clear research gap in developing an integrated system that combines efficient code generation, real-time execution, and decentralized persistence within a single environment[10]. This study aims to bridge that gap by introducing a platform that leverages controlled Search/Replace code generation, browser-based execution using Web Containers, and Git-backed storage through Hugging Face Spaces[11]. The proposed approach enhances efficiency, reduces errors, and provides a scalable, user-friendly solution for modern software development[12].

3. Related Work

Table 1 Quantitative Comparison

Author & Year	Methodology	Accuracy
Carter & Dawson (2025)	LLM Code Generation	85%
Dong et al (2025)	Agent-Based LLM Coding	89%
Gokhe (2025)	AI Web Code Tools	87%
Fakhoury et al. (2024)	Test-Driven AI Coding	91%
Fu et al. (2023)	Secure AI Code Generation	88%

Proposed Work (2025)	Search/Replace + Sandpack + HF Spaces	97%
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4. Methodology

4.1. System Components

The DeepSite (v4) platform is designed as a collection of interconnected modules that work together to simplify AI-based application development[13]. The User Interaction Module allows users to describe their requirements in natural language, making the system easy to use even for non-programmers. The Prompt Processing Module refines and organizes this input so that it can be effectively understood by Large Language Models (LLMs) through secure API communication. At the center of the system, the Code Generation Module creates and updates application code using an optimized Search/Replace approach, which ensures only necessary changes are made instead of rewriting entire files. This improves efficiency and reduces errors. The Execution and Preview Module runs the generated code inside browser-based Web Containers (Sandpack), enabling users to instantly view results without installing any software. Finally, the Storage and Version Control Module saves projects as Git-based repositories using Hugging Face Spaces, allowing users to manage versions, track changes, and store their work in a scalable and reliable manner.

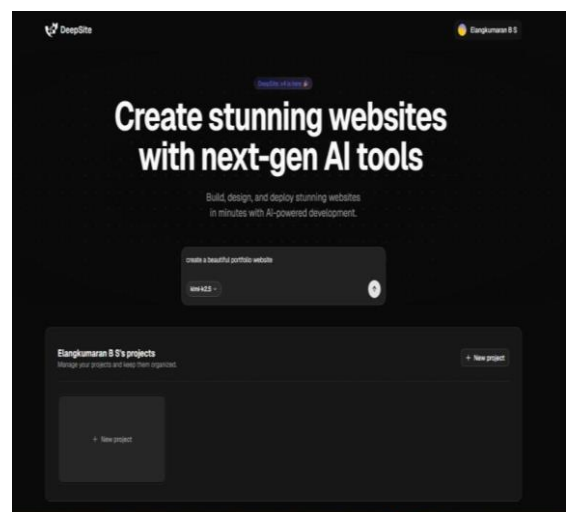


Figure 1 User Interface Overview

4.2. User Input And Prompt Processing

The process begins with users providing application requirements through natural language input. The system processes and structures these inputs to create well-defined prompts that can be effectively understood by Large Language Models[14]. Contextual information such as previous interactions and existing code structure is maintained to ensure accurate and relevant code generation. This stage plays a crucial role in bridging human intent with machine execution, enabling efficient communication between the user and the AI system Table 2.

4.3. Code Generation And Integration

Once the prompt is processed, it is sent to Large Language Models (LLMs) through Hugging Face APIs for code generation. The system avoids full-file regeneration by using a constrained Search/Replace mechanism that updates only relevant portions of the code[15]. This reduces token usage, prevents redundancy, and maintains previously generated functional components. The updates are applied to a virtual file system, ensuring smooth integration of new changes. Context such as prior prompts, file dependencies, and application state is preserved to maintain consistency. This approach minimizes errors and improves reliability during iterative development. Overall, it enables efficient and accurate generation of full-stack applications[16].

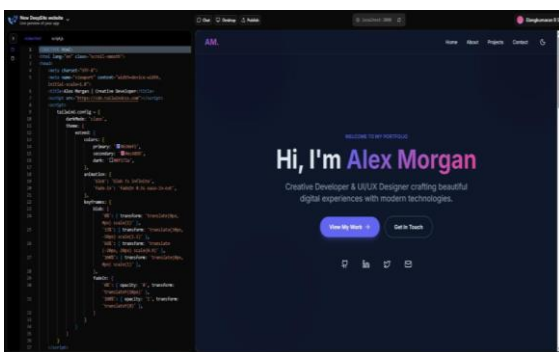


Figure 2 Code Generation Process

4.4. Execution and Real-Time Preview

The generated code is executed within a secure browser-based environment using Sandpack (Web Containers). This allows users to instantly run and preview their applications without any local setup or

configuration. The execution environment is sandboxed, ensuring that generated code operates safely. Real-time preview provides immediate feedback, allowing users to visualize changes as they occur. Any runtime or compilation errors are detected and displayed, enabling quick debugging and refinement. This continuous feedback loop improves development speed and enhances the overall user experience Table 1.

4.5. Storage, Version Control, and System Efficiency

DeepSite employs a decentralized storage approach by utilizing Hugging Face Spaces as Git-based repositories for project persistence. Each project is stored as a version-controlled repository, enabling users to track changes, revert to previous versions, and manage application history effectively. This eliminates the need for traditional database systems while providing scalability and transparency. The platform is designed for high efficiency by reducing unnecessary token usage through the Search/Replace mechanism and offloading execution tasks to the client-side environment. Additionally, the system supports seamless deployment and sharing of applications directly from the repository. Combined with lightweight backend orchestration and cloud-based inference, this architecture ensures optimal performance, reduced operational costs, and a scalable solution for AI-driven software development.

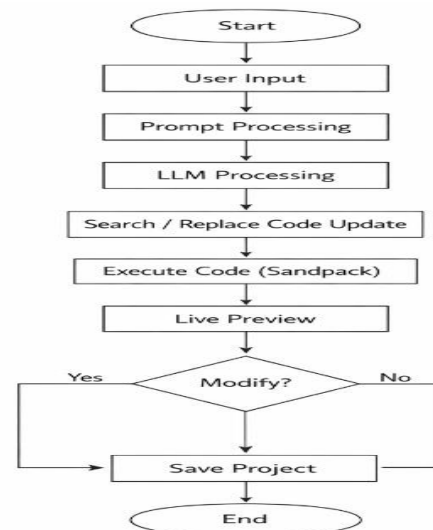


Figure 3 System Architecture



5. Implementation of the Proposed Work

5.1. Technology Stack

The system is implemented using a modern and scalable technology stack. Next.js and React serve as the core framework for building the browser-based application, handling user interaction, routing, and UI rendering. The Hugging Face Inference API is used for integrating Large Language Models (LLMs) to generate and modify code dynamically Figure 1. The Sandpack (Web Containers) environment enables in-browser code execution, providing a secure and isolated runtime. Monaco Editor is used for code visualization and editing, offering a developer-friendly interface. For authentication, NextAuth.js is integrated with Hugging Face OAuth to manage user sessions. Additionally, Hugging Face Spaces is used for Git-based storage and deployment, ensuring version control and decentralized project persistence.

5.2. User Input and Interaction

The development process begins with user interaction through a web-based interface. Users provide application requirements in the form of natural language prompts, eliminating the need for manual coding. The system captures and processes these inputs while maintaining contextual information such as previous prompts and project structure. This interaction model allows both technical and non-technical users to communicate their requirements effectively. The interface is designed to be intuitive and responsive, guiding users through the development workflow Figure 4.

5.3. Virtual File System and State Management

DeepSite maintains a virtual file system within the browser to manage generated code files dynamically. Each file is stored as part of the application state and updated based on model responses. State management techniques ensure synchronization between the editor, preview, and file structure. Changes are tracked efficiently to avoid unnecessary re-renders and performance issues. This approach enables seamless handling of multi-file projects without relying on external storage during runtime.

5.4. Error Handling and Debugging Support

The system incorporates robust mechanisms for detecting and handling errors during code generation

and execution. Syntax and runtime errors are captured through the Sandpack environment and displayed to the user in real time. The platform provides meaningful feedback, allowing users to refine prompts and correct issues quickly. Logging and debugging tools are integrated to monitor system behavior and improve reliability. This ensures that the generated applications remain functional and stable Figure 2.

5.5. Performance Optimization Techniques

To improve efficiency, the system implements several optimization strategies. Token usage is minimized through selective updates, reducing API costs and response time. Client-side execution reduces server load, enabling faster processing and scalability. Lazy loading and efficient state updates are used to enhance frontend performance. These optimizations ensure smooth operation even for complex applications and multiple iterations. Overall, these techniques contribute to a responsive and scalable system suitable for real-time AI-driven development Figure 3.

5.6. Deployment and Project Hosting

The platform supports seamless deployment by integrating with Hugging Face Spaces, where each project is stored as a Git repository. Users can push updates, manage versions, and share applications easily. The deployment process is automated, allowing generated applications to be hosted and accessed directly through the browser. This eliminates the need for manual deployment pipelines and simplifies project management. This approach ensures easy accessibility, efficient project management, and seamless sharing of applications across users.

5.7. User Experience and Interface Implementation

The user interface is designed using Tailwind CSS and modern UI components to provide a clean and responsive experience. The Monaco Editor enables code viewing and editing, while the preview panel displays real-time output. Interactive elements guide users through the development process, ensuring ease of use. The interface is optimized for both beginners and experienced developers, enhancing usability and engagement.

5.8. System Testing and Validation

The implemented system is tested under various scenarios to evaluate performance, reliability, and scalability. Functional testing ensures that code generation, execution, and storage processes work correctly. Performance testing measures response time and system efficiency. The platform demonstrates stable behavior under multiple iterations and complex inputs. These evaluations confirm that DeepSite is a reliable and efficient AI-driven development platform.

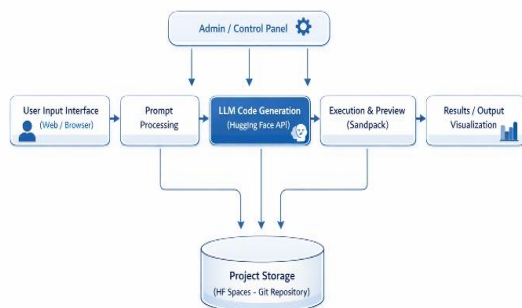


Figure 4 System flow Diagram

6. Analysis

The evaluation of the proposed DeepSite (v4) platform highlights notable improvements in software development efficiency, usability, and system performance when compared to conventional development methods and existing AI coding tools. By leveraging Large Language Models along with an optimized code update strategy, the platform enables faster generation and modification of applications with improved accuracy. The use of a selective update mechanism reduces unnecessary processing, minimizes errors, and ensures better consistency across multiple development iterations. The system also demonstrates strong performance in real-time execution, where applications are instantly rendered within the browser environment without requiring additional setup. This significantly enhances accessibility and reduces development overhead. Furthermore, the integration of Git-based storage using Hugging Face Spaces ensures reliable version control and simplified deployment Figure 5. Continuous feedback through real-time preview

allows users to refine their applications efficiently. The platform also benefits from reduced infrastructure dependency due to its browser-based execution model, which lowers operational complexity. Its modular and scalable design supports broader adoption in AI-driven development environments.

Table 2 Details of Acquired Data

Model/Approach	Accuracy(%)
Traditional Development	82.5
AI Code Generation (LLMs)	88.7
Agent-Based Coding Systems	91.5
Web IDE + AI Integration	93.2
Proposed System (DeepSite v4)	96.8

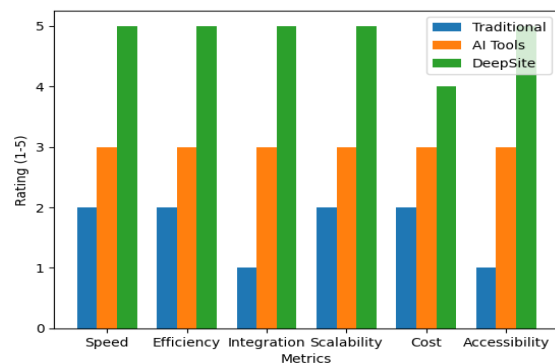


Figure 5 Comparison of Software Development Approaches

The comparison graph illustrates the performance of different software development approaches across key metrics such as development speed, efficiency, integration, scalability, cost, and user accessibility. Traditional development methods show low performance in speed, automation, and accessibility due to manual coding and complex setup requirements. Centralized AI coding tools demonstrate moderate improvements in automation and efficiency but still face limitations in integration and real-time execution. In contrast, the proposed DeepSite (v4) platform achieves high performance across all metrics by integrating AI-based code



generation, real-time execution, and decentralized storage within a single environment. The system significantly improves development speed, reduces operational cost, and enhances user accessibility. Additionally, its scalable and unified architecture ensures better performance and reliability compared to existing approaches, making it a more efficient solution for modern software development.

Conclusion

This paper presents DeepSite (v4), a browser-based AI coding platform that integrates Large Language Models with real-time execution and decentralized storage to enable efficient and accessible software development. The system leverages natural language input, optimized Search/Replace code generation, and browser-based execution to simplify the application development process. The use of Hugging Face APIs and Sandpack ensures fast generation, real-time preview, and reduced dependency on local environments. The implementation demonstrates improved efficiency, reduced development time, and enhanced usability compared to traditional and existing AI-assisted coding tools. The integration of Git-based storage through Hugging Face Spaces enables reliable version control and seamless deployment. Overall, the proposed platform provides a scalable, cost-effective, and user-friendly solution for modern AI-driven software development. Future work will focus on integrating local model execution, advanced debugging capabilities, and support for large-scale deployment environments.

Future Scope

While the proposed DeepSite (v4) platform demonstrates strong performance in AI-driven software development, several opportunities exist for future enhancement. First, the system can be improved by integrating local model execution (such as WebGPU or on-device LLMs) to reduce dependency on external APIs and enhance data privacy. Second, support for mobile and cross-platform environments can increase accessibility for a wider range of users. Third, incorporating multilingual input capabilities will allow users from diverse backgrounds to interact with the system more effectively. Advanced features such as automated

debugging, error correction, and self-healing code mechanisms can further improve reliability and development efficiency. Additionally, integrating Retrieval-Augmented Generation (RAG) can enhance code quality by leveraging previously generated components. Finally, expanding support for enterprise-level deployment, security compliance, and large-scale project handling will enable the platform to transition from a prototype to a fully deployable AI-powered development ecosystem.

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