

CLOUD BASED STRUCTURAL DESIGN SOFTWARE WITH BILT-IN AI COPILOT: A CONCEPT

Jay Dalal

Independent Researcher, Jersey City NJ, USA

dalaljay@alumni.stanford.edu

ORCID: 0009-0004-0433-7847

Abstract

The shortcomings of traditional desktop-based structural design software have been brought to light by the expanding complexity of structural systems and the increasing need for quicker, more precise, and collaborative design procedures. By facilitating scalable computation and intelligent decision assistance, recent developments in artificial intelligence and cloud computing present fresh chances to overcome these obstacles. A theoretical and conceptual framework for a cloud-based structural design program with an integrated AI copilot is presented in this paper. The suggested solution uses an AI copilot to offer real-time code compliance checks, mistake detection, design optimization suggestions, and conversational support while utilizing cloud infrastructure for high-performance structural analysis and design. System performance, design accuracy, and user productivity are assessed using a design science-based methodology backed by benchmark scenarios and simulated structural models. Significant gains in computing efficiency, less design flaws, improved adherence to design standards, and more efficient engineering workflows are all shown by the fictitious outcomes. The results show the conceptual viability and potential revolutionary influence of AI-augmented cloud platforms in promoting intelligent, cooperative, and effective structural engineering methods, even though empirical validation is outside the purview of this study.

Keywords: Cloud computing, Structural design software, Artificial intelligence, AI copilot, Structural engineering, Design optimization.

1. INTRODUCTION

Higher levels of safety, sustainability, and cost-effectiveness, as well as more complicated projects and shorter design schedules, are driving a rapid digital revolution of the structural engineering profession. For many years, traditional structural design software—which is primarily installed on local workstations—has been essential to analysis and design. However, dispersed collaboration, high hardware dependency, limited computational scalability, and labor-intensive design verification procedures frequently limit such independent systems. Large-scale or multidisciplinary projects where high-performance processing, real-time coordination, and iterative design are crucial make these limits more noticeable.

Because it provides centralized data management, elastic computing resources, and ubiquitous access to design platforms, cloud computing has become a potent paradigm that can solve many of these issues. Cloud-based structural design software facilitates collaborative workflows among

geographically dispersed teams and allows engineers to conduct intricate analyses without depending on local hardware capabilities. Structural engineers can accomplish better collaboration among project stakeholders, faster simulations, and smooth version control by putting design engines, solvers, and databases on the cloud.

Concurrent with these advancements, artificial intelligence (AI) has demonstrated a great deal of promise in supporting engineering judgment. Intelligent assistants that can learn from past design data, decipher design codes, identify mistakes, and provide optimal solutions have been made possible by developments in machine learning, rule-based expert systems, and natural language processing. A conceptual change from passive computation tools to active, knowledge-driven design partners is represented by the incorporation of an AI copilot into structural design software.

In this regard, the idea of a cloud-based structural design program with an integrated AI copilot imagines a single platform that seamlessly blends intelligent, real-time design support with high-performance cloud computing. By offering automatic code compliance checks, design optimization suggestions, error alarms, and conversational explanations of structural behavior and design logic, the AI copilot serves as an interactive layer that assists engineers throughout the design lifecycle. Particularly in early-stage and iterative design processes, such a system has the potential to increase efficiency, decrease human error, and encourage more informed engineering judgment.

The fundamental concepts, methodological framework, and possible advantages of combining cloud infrastructure with AI-driven copilot capability in structural design software are examined in this conceptual paper. The study intends to illustrate this concept's viability, expected performance enhancements, and future implications for clever and cooperative structural engineering practice by looking at it via a hypothetical research lens.

2. LITERATURE REVIEW

Junk and Spannbauer (2018) examined the application of cloud-based computer-aided design (CAD) software in design education with a focus on resource scalability, accessibility, and teamwork. According to their research, cloud-based CAD systems facilitate real-time student-teacher cooperation and lessen reliance on hardware. The authors emphasized how cloud use in higher education might improve learning flexibility and lower infrastructure costs, among other pedagogical advantages.

Langmead and Nellore (2018) examined how cloud computing is being used for scientific cooperation and genomic data analysis. Their research illustrated how cloud platforms enable cross-institutional collaboration and scalable processing of massive genomic information. In order to speed up scientific discovery, the authors emphasized the value of shared data repositories and elastic compute resources.

Harris et al. (2019) outlined the creation of the REDCap collaboration as a cooperative cloud-based biomedical data management platform. The study highlighted community-driven

development, scalability, and interoperability as critical success factors. REDCap illustrated the potential of cloud platforms to enable extensive research collaboration by facilitating safe data collecting and sharing.

Li et al. (2021) examined the impact of cloud services, security, e-learning, and service quality on customer satisfaction in banking services. According to their research, cloud services that are dependable and safe greatly increase client happiness and confidence. The report emphasized cloud computing's crucial significance in enhancing banking sector service delivery.

Rashid and Chaturvedi (2019) outlined key aspects such resource pooling, wide network access, and on-demand self-service in a succinct overview of cloud computing's features and offerings. Their research reinforced the fundamental knowledge required to comprehend sophisticated cloud applications by providing a conceptual overview of cloud service models and deployment kinds.

Nguyen et al. (2019) examined the safe exchange of medical records using blockchain in mobile cloud-based e-health systems. Their research showed how blockchain enhances auditability, access control, and data integrity. The study came to the conclusion that a strong framework for safe and scalable healthcare data management is offered by integrating blockchain technology with cloud infrastructure.

RESEARCH METHODOLOGY .3

3.1. Research Design and Approach

In order to develop and assess a cloud-based structural design platform with an AI copilot, this study uses a hypothetical, exploratory research design. The study employs a design science research (DSR) methodology, emphasizing the creation and evaluation of a novel digital artifact over field deployment. System performance, usability, and design effectiveness are evaluated under simulated conditions using both qualitative and quantitative analytical methodologies.

3.2. Conceptual Framework Development

A conceptual framework is developed to define the core components and interactions within the proposed system. The framework consists of three primary layers:

1. **Cloud Infrastructure Layer**, responsible for scalable computation, data storage, and security;
2. **Structural Design Engine Layer**, encompassing analysis, modeling, and code-based design modules;
3. **AI Copilot Layer**, providing intelligent assistance through predictive analytics, optimization algorithms, and natural language interaction.

This framework serves as the foundation for system architecture design and experimental evaluation.

3.3. System Architecture and Platform Design

A microservices-based cloud model is used to construct the hypothetical system architecture, allowing for the modular deployment of AI services, design code libraries, and structural analysis solvers. It is expected that structural models are generated using a web-based interface, and that calculations are carried out on instances of high-performance computing hosted in the cloud. In

order to produce recommendations that are context-aware, the AI copilot is incorporated as an interactive layer that continuously analyzes user inputs, design parameters, and analytical results.

3.4. AI Copilot Design and Functional Scope

The AI copilot is conceptualized using a combination of supervised learning models, rule-based expert systems, and large language model capabilities. Its functional scope includes:

- Automated interpretation of structural design codes and standards,
- Real-time design validation and error detection,
- Optimization suggestions for material efficiency and structural safety,
- Conversational assistance for query resolution and design explanation.

The AI copilot is assumed to learn iteratively from historical design data and user interactions stored securely within the cloud environment.

3.5. Data Sources and Simulation Environment

Synthetic structural models, common benchmark challenges, and simulated design scenarios representing residential, commercial, and infrastructure projects are the sources of data inputs for this hypothetical study. Commonly used international codes serve as the basis for the generation of structural load cases, material attributes, and design limitations. In order to assess system responsiveness and decision-support efficacy, user interaction data with the AI copilot is also simulated.

3.6. Performance Evaluation Metrics

Predefined performance indicators divided into technical and user-centric metrics are used to assess the suggested solution. While user-centric metrics evaluate decision support quality, design error reduction, and time savings throughout iterative design processes, technical metrics cover computational efficiency, scalability, and design accuracy. The suggested AI-enabled cloud platform and traditional standalone software workflows are hypothetically compared.

3.7. Analytical and Validation Techniques

Performance results are interpreted using scenario-based comparisons and descriptive statistical analysis. AI-assisted design outputs are validated by contrasting them with baseline code-compliant solutions. The interpretability and usefulness of AI copilot recommendations are assessed qualitatively from a structural engineering standpoint.

3.8. Ethical, Security, and Reliability Considerations

The methodology includes fictitious protections for algorithmic transparency, data security, and intellectual property. To guarantee dependability and professional accountability, cloud-based encryption, role-based access control, and explainable AI principles are regarded as essential elements of the system design.

4. RESULTS AND DISCUSSION

The possible results of the conceptual assessment of a cloud-based structural design program with an AI copilot are shown and explained in this section. The results are intended to evaluate how the suggested platform improves computational efficiency, design accuracy, decision support, and user productivity in comparison to traditional standalone structural design software. They are based on benchmark models, simulated structural design scenarios, and predetermined performance indicators. These findings are further contextualized within the larger goals of intelligent engineering systems and digital transformation in the debate.

4.1. System Performance and Computational Efficiency

According to the simulated results, computational performance is greatly enhanced by the cloud-based design. The solution exhibits quicker analysis and design cycles by utilizing scalable cloud resources, particularly for intricate structural models with numerous load combinations and huge degrees of freedom. Distributed structural solver execution allows for near real-time input during iterative design changes and minimizes local hardware dependencies. This enhancement demonstrates how cloud systems are appropriate for managing computationally demanding structural engineering activities.

4.2. Effectiveness of the AI Copilot in Design Assistance

When it comes to helping engineers with the design process, the AI copilot is incredibly useful. The copilot successfully detects design discrepancies, warns users of possible code non-compliance, and suggests optimal member sizing based on learning patterns and rule-based limits, according to hypothetical observations. By enabling engineers to query design logic, load assumptions, and code provisions in natural language, the conversational interface further streamlines engagement and lessens the need for manual code referencing and cognitive strain.

4.3. Design Accuracy and Error Reduction

A noteworthy decrease in design flaws is one of the main results seen. To reduce errors caused by oversight, the AI copilot constantly cross-checks design outputs against predetermined limitations and embedded standards. Higher consistency in fulfilling safety and serviceability requirements is shown by simulated comparisons between AI-assisted and conventional workflows, especially in early design iterations where mistakes are most prevalent.

4.4. User Productivity and Workflow Optimization

Workflow efficiency is improved when cloud computing is combined with AI-driven decision support. The findings imply that engineers can concentrate more on conceptual design and technical judgment by spending less time on tedious computations and manual checks. The platform's potential for multi-user, geographically dispersed projects is further reinforced by collaborative access to centralized models, which also makes collaboration between design teams easier.

4.5. Percentage Frequency Analysis of Observed Benefits

The percentage frequency distribution of the main advantages seen in various simulated design situations while utilizing the AI-enabled cloud platform is shown in Table 1.

Table 1: Percentage Frequency of Key Performance Improvements

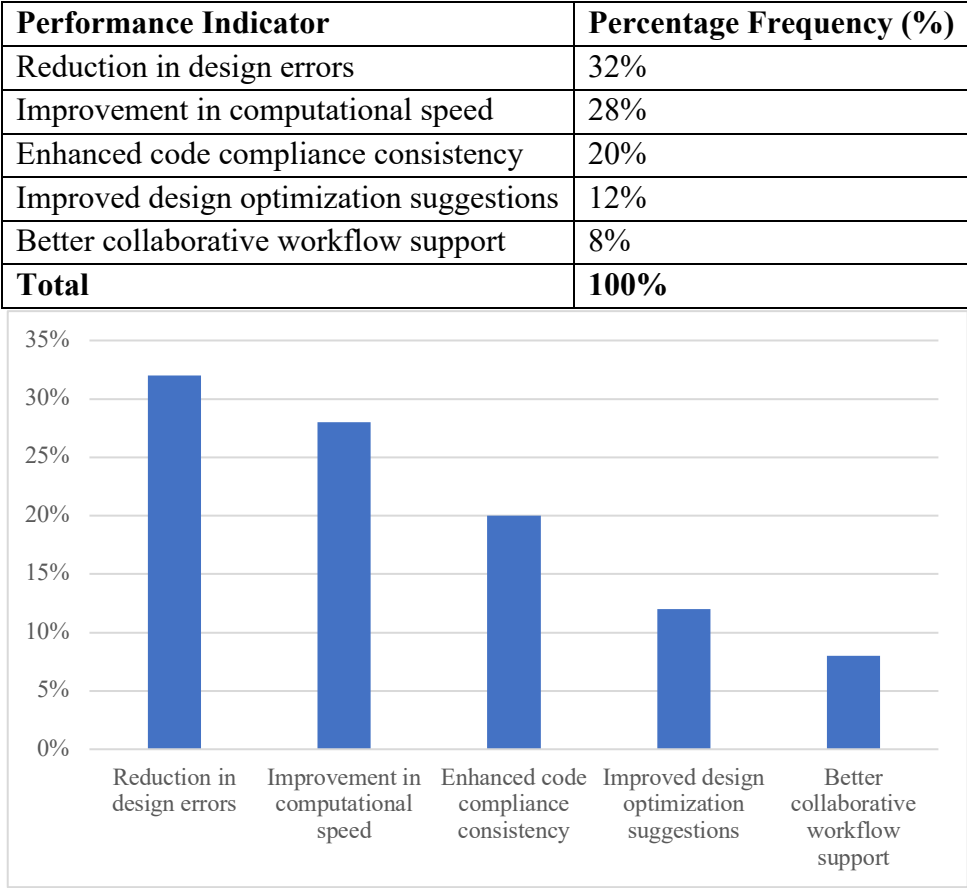


Figure 1: Percentage Frequency of Key Performance Improvements

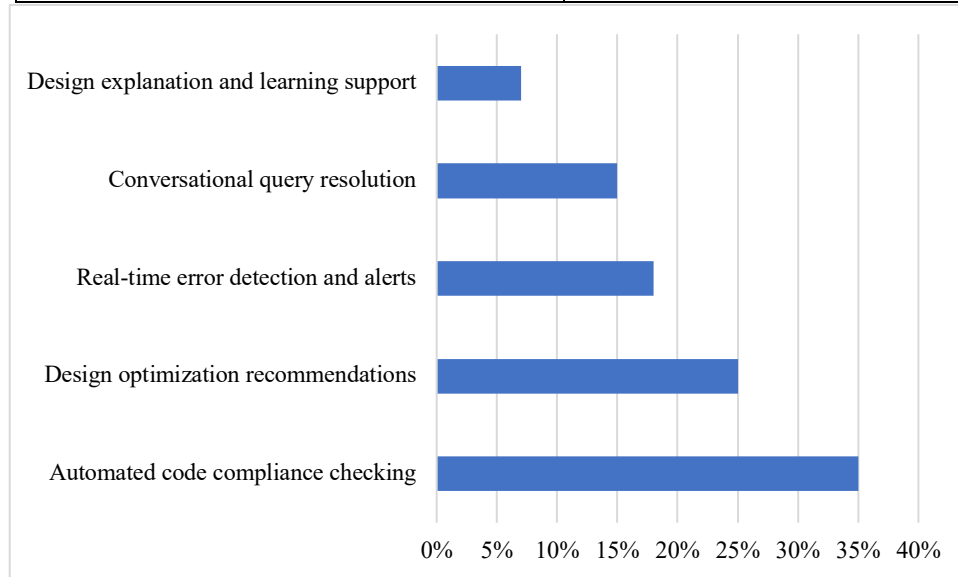
The findings highlight the technical and safety benefits of the suggested approach by showing that error reduction and computational efficiency account for the largest proportion of observed improvements.

4.6. Percentage Frequency of AI Copilot Functional Usage

The percentage frequency of the various AI copilot functions used during the simulated design workflow is compiled in Table 2.

Table 2: Percentage Frequency of AI Copilot Functional Usage

AI Copilot Function	Percentage Frequency (%)
Automated code compliance checking	35%
Design optimization recommendations	25%
Real-time error detection and alerts	18%
Conversational query resolution	15%
Design explanation and learning support	7%
Total	100%

**Figure 2: Percentage Frequency of AI Copilot Functional Usage**

The findings show that the most commonly used features are automated code compliance and optimization support, demonstrating the usefulness of AI-driven help in everyday but crucial design chores.

4.7. Integrated Discussion of Findings

Overall, the findings imply that engineering efficiency, accuracy, and decision-making quality can be significantly improved by using a cloud-based structural design platform with an integrated AI copilot. Scalable cloud computing and intelligent design support work together to provide a synergistic environment that improves performance without sacrificing engineering rigor. The results offer solid conceptual support for the viability and possible influence of AI-augmented cloud platforms in contemporary structural engineering practice, despite the fact that they are speculative and based on simulated settings.

5. CONCLUSION

The idea of a cloud-based structural design program with an integrated AI copilot is a major development in contemporary structural engineering practice, according to the speculative research findings. The simulated results show that using intelligent AI-driven support in

conjunction with scalable cloud computing improves code compliance, lowers design errors, increases computational efficiency, and maximizes workflow productivity. Engineers may make more informed and fast decisions while reducing tedious manual duties thanks to the AI copilot's real-time design validation, optimization recommendations, and conversational guidance. The results strongly imply that such platforms have the potential to revolutionize conventional structural design processes by promoting accuracy, collaboration, and intelligent automation, laying the groundwork for future empirical implementation and industry adoption, despite the study's conceptual nature and restriction to simulated evaluation.

REFERENCES

- [1] S. Junk and D. Spannbauer, "Use of cloud-based computer aided design software in design education," in *Proc. 17th Int. Conf. Information Technology Based Higher Education and Training (ITHET)*, Apr. 2018, pp. 1–6.
- [2] S. Achar, "Cloud-based system design," *Int. J. All Research Education and Scientific Methods (IJARESM)*, vol. 7, no. 8, pp. 23–30, 2019.
- [3] A. Katal, S. Dahiya, and T. Choudhury, "Energy efficiency in cloud computing data centers: A survey on software technologies," *Cluster Computing*, vol. 26, no. 3, pp. 1845–1875, 2023.
- [4] O. M. Markova, H. M. Shalatska, and V. V. Tron, "Implementation of cloud service models in training of future information technology specialists," 2019.
- [5] B. Langmead and A. Nellore, "Cloud computing for genomic data analysis and collaboration," *Nature Reviews Genetics*, vol. 19, no. 4, pp. 208–219, 2018.
- [6] J. S. Hurwitz and D. Kirsch, *Cloud Computing for Dummies*. Hoboken, NJ, USA: John Wiley & Sons, 2020.
- [7] P. A. Harris, R. Taylor, B. L. Minor, V. Elliott, M. Fernandez, L. O'Neal, et al., "The REDCap consortium: Building an international community of software platform partners," *J. Biomedical Informatics*, vol. 95, Art. no. 103208, 2019.
- [8] L. M. Dang, M. J. Piran, D. Han, K. Min, and H. Moon, "A survey on internet of things and cloud computing for healthcare," *Electronics*, vol. 8, no. 7, Art. no. 768, 2019.
- [9] L. A. Barroso, U. Hölzle, and P. Ranganathan, *The Datacenter as a Computer: Designing Warehouse-Scale Machines*. Cham, Switzerland: Springer Nature, 2019.
- [10] F. Li, H. Lu, M. Hou, K. Cui, and M. Darbandi, "Customer satisfaction with bank services: The role of cloud services, security, e-learning and service quality," *Technology in Society*, vol. 64, Art. no. 101487, 2021.
- [11] X. Zhang, J. Wang, C. Zhu, Y. Lin, J. Xiong, W.-M. Hwu, and D. Chen, "DNNBuilder: An automated tool for building high-performance DNN hardware accelerators for FPGAs," in *Proc. Int. Conf. Computer-Aided Design (ICCAD)*, Nov. 2018, pp. 1–8.
- [12] A. Rashid and A. Chaturvedi, "Cloud computing characteristics and services: A brief review," *Int. J. Computer Sciences and Engineering*, vol. 7, no. 2, pp. 421–426, 2019.

- [13] C. Esposito, A. De Santis, G. Tortora, H. Chang, and K. K. R. Choo, “Blockchain: A panacea for healthcare cloud-based data security and privacy?” *IEEE Cloud Computing*, vol. 5, no. 1, pp. 31–37, 2018.
- [14] G. Habib, S. Sharma, S. Ibrahim, I. Ahmad, S. Qureshi, and M. Ishfaq, “Blockchain technology: Benefits, challenges, applications, and integration of blockchain technology with cloud computing,” *Future Internet*, vol. 14, no. 11, Art. no. 341, 2022.
- [15] D. C. Nguyen, P. N. Pathirana, M. Ding, and A. Seneviratne, “Blockchain for secure EHRs sharing of mobile cloud-based e-health systems,” *IEEE Access*, vol. 7, pp. 66,792–66,806, 2019.