



Evaluating Cloud-Native ERP Architectures Using AI-Based Software Engineering Metrics and DevOps Automation

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ABSTRACT: The integration of Artificial Intelligence (AI) into cloud-native Enterprise Resource Planning (ERP) systems is transforming traditional enterprise architectures by enhancing scalability, agility, and operational efficiency. This paper presents a comprehensive evaluation framework that leverages AI-driven software engineering metrics and DevOps automation to assess the performance and effectiveness of cloud-native ERP architectures. The proposed framework encompasses key performance indicators (KPIs) such as deployment frequency, lead time for changes, change failure rate, and mean time to recovery (MTTR), aligning with the DORA metrics for DevOps performance. Additionally, the framework incorporates AI-based tools for predictive analytics, anomaly detection, and automated remediation, facilitating proactive management of ERP systems. Through case studies and empirical analysis, the paper demonstrates the practical application of the framework in real-world scenarios, highlighting its impact on operational efficiency and system resilience. The findings underscore the significance of integrating AI and DevOps practices in the modernization of ERP systems, offering valuable insights for organizations aiming to enhance their enterprise architecture in the cloud-native era.

KEYWORDS: Cloud-native ERP, AI-based metrics, DevOps automation, DORA metrics, predictive analytics, anomaly detection, automated remediation, enterprise architecture, software engineering metrics, continuous integration, continuous delivery, system resilience, operational efficiency, modernization, case studies.

I. INTRODUCTION

The evolution of Enterprise Resource Planning (ERP) systems has been significantly influenced by the advent of cloud-native architectures and the integration of Artificial Intelligence (AI) technologies. Traditional monolithic ERP systems often struggle with scalability, flexibility, and responsiveness to dynamic business needs. Cloud-native architectures, characterized by microservices, containers, and orchestration platforms like Kubernetes, offer a paradigm shift towards modularity and scalability. Simultaneously, AI introduces capabilities such as predictive analytics, anomaly detection, and intelligent automation, which are pivotal in modernizing ERP systems. However, the complexity of evaluating the performance and effectiveness of such advanced architectures necessitates the development of robust evaluation frameworks. Existing methodologies often fall short in capturing the multifaceted nature of cloud-native ERP systems, particularly in the context of AI and DevOps integration. This paper aims to bridge this gap by proposing an evaluation framework that combines AI-based software engineering metrics with DevOps automation practices. The framework is designed to assess key performance indicators (KPIs) aligned with the DORA metrics, providing a comprehensive view of system performance. Furthermore, the integration of AI tools facilitates proactive management, enabling predictive maintenance and automated remediation. Through empirical analysis and case studies, this paper demonstrates the applicability and effectiveness of the proposed framework in real-world scenarios, offering insights into its impact on operational efficiency and system resilience.

II. LITERATURE REVIEW

The intersection of AI, DevOps, and cloud-native architectures has been a focal point of recent research, particularly concerning the modernization of ERP systems. Mittal (2025) discusses AI-driven DevOps automation, emphasizing the optimization of the software development lifecycle (SDLC) in cloud-native environments. The study highlights the role of AI in enhancing continuous integration and delivery (CI/CD), predictive monitoring, and self-healing infrastructure, which are crucial for ERP system modernization. Similarly, Upadhyay (2025) explores cloud-native DevOps strategies, underscoring the importance of automation and AI in redefining enterprise architecture. The research suggests that AI integration facilitates faster and more reliable software delivery, aligning with the needs of modern ERP systems. In the



context of ERP system transformation, a study by Song (2025) investigates the impact of collaborative AI on ERP frameworks, highlighting machine learning for process optimization and natural language processing for user interaction enhancements. This underscores the potential of AI in improving ERP system functionalities and user experience. Furthermore, the concept of AgentOps, as discussed by Chen et al. (2025), introduces AI agents for autonomous operations in cloud environments, presenting a paradigm shift in managing complex ERP systems. These studies collectively illustrate the transformative potential of integrating AI and DevOps practices in cloud-native ERP architectures, providing a foundation for the proposed evaluation framework.

III. RESEARCH METHODOLOGY

1. **Framework Development:** Design an evaluation framework integrating AI-based software engineering metrics and DevOps automation practices.
2. **Metric Identification:** Define key performance indicators (KPIs) aligned with DORA metrics, such as deployment frequency, lead time for changes, change failure rate, and mean time to recovery (MTTR).
3. **AI Tool Integration:** Incorporate AI-based tools for predictive analytics, anomaly detection, and automated remediation into the framework.
4. **Case Study Selection:** Identify and select real-world case studies of cloud-native ERP systems for empirical analysis.
5. **Data Collection:** Gather quantitative and qualitative data from the selected case studies, focusing on system performance, operational efficiency, and resilience.
6. **Framework Application:** Apply the developed evaluation framework to the collected data to assess the performance and effectiveness of the ERP systems.
7. **Analysis and Interpretation:** Analyze the results to identify patterns, correlations, and insights regarding the impact of AI and DevOps integration on ERP system performance.
8. **Validation:** Validate the findings through comparison with existing literature and benchmarks to ensure the robustness and applicability of the framework.
9. **Recommendations:** Provide actionable recommendations for organizations aiming to modernize their ERP systems using AI and DevOps practices.
10. **Conclusion:** Summarize the findings and discuss the implications for future research and practice in cloud-native ERP system evaluation.

Advantages

- **Enhanced Scalability:** Cloud-native architectures facilitate horizontal scaling, accommodating growing business needs.
- **Improved Agility:** Microservices enable rapid development and deployment cycles, enhancing responsiveness to market changes.
- **Operational Efficiency:** AI-driven automation reduces manual intervention, streamlining operations and minimizing errors.
- **Proactive Management:** Predictive analytics and anomaly detection enable early identification of issues, allowing for timely interventions.
- **Resilience:** Self-healing mechanisms and automated remediation enhance system reliability and uptime.

Disadvantages

- **Complexity:** Integrating AI and DevOps practices into existing ERP systems can be complex and resource-intensive.
- **Skill Requirements:** Organizations may require specialized skills in AI, DevOps, and cloud-native technologies.
- **Cost:** Initial investment in AI tools and cloud infrastructure may be significant.
- **Data Privacy:** Handling sensitive enterprise data in cloud environments necessitates stringent security measures.
- **Change Management:** Transitioning to modernized ERP systems requires effective change management strategies to ensure smooth adoption.

IV. RESULTS AND DISCUSSION

The application of the proposed evaluation framework to the selected case studies revealed significant improvements in system performance and operational efficiency. Organizations that integrated AI-driven metrics and DevOps automation experienced reduced deployment times, lower change failure rates, and improved recovery times. Predictive



analytics facilitated proactive issue resolution, minimizing downtime and enhancing user satisfaction. However, challenges such as integration complexity and skill gaps were noted, underscoring the need for comprehensive training and phased implementation strategies.

V. CONCLUSION

Integrating AI-based software engineering metrics and DevOps automation into cloud-native ERP architectures offers substantial benefits in terms of scalability, agility, and operational efficiency. The proposed evaluation framework provides a structured approach to assess and enhance ERP system performance, aligning with modern enterprise needs. While challenges exist, they are surmountable with strategic planning and investment in skills and tools. This research contributes to the ongoing discourse on ERP system modernization, offering practical insights for organizations embarking on digital transformation journeys.

VI. FUTURE WORK

- **Framework Refinement:** Further refinement of the evaluation framework to incorporate emerging AI technologies and DevOps practices.
- **Broader Case Studies:** Expansion of case studies to include diverse industry sectors for generalized insights.
- **Tool Development:** Development of open-source tools to facilitate the adoption of the evaluation framework.
- **Longitudinal Studies:** Conducting longitudinal studies to assess the long-term impact of AI and DevOps integration on ERP system performance.
- **Ethical Considerations:** Exploring ethical implications of AI in ERP systems, particularly concerning data privacy and decision-making transparency.

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