

# The Role of AI and Machine Learning in Optimising Cloud Migration Processes

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## Abstract

The systematic literature review explores the role of Artificial Intelligence (AI) and Machine Learning (ML) in enhancing cloud migration processes. It highlights that these technologies improve efficiency by automating workload assessments, optimising resource allocation, and enabling predictive analytics for risk management. Key findings indicate that tools like AWS Migration Hub and Azure Migrate significantly streamline migrations, reducing costs and time. However, challenges like a shortage of qualified personnel and data security concerns persist. The review concludes that adopting AI and ML can lead to more effective cloud migration strategies, with future directions focusing on integrating emerging technologies and exploring user-centric approaches.

**Keywords:** AI, Machine Learning, Cloud Migration, Efficiency, Predictive Analytics

on-premise systems to the cloud is a delicate process that must be planned, implemented, and optimised to minimise operational impacts. Fig. 1 explains the role of AI in Cloud Computing.

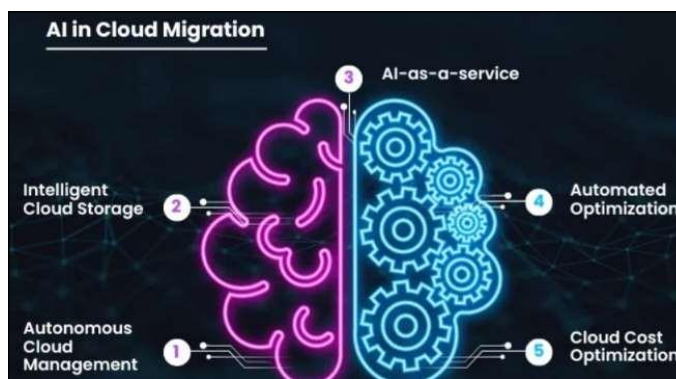


Fig. 1: AI in Cloud Computing (Flash, 2024)

## Introduction

### Overview

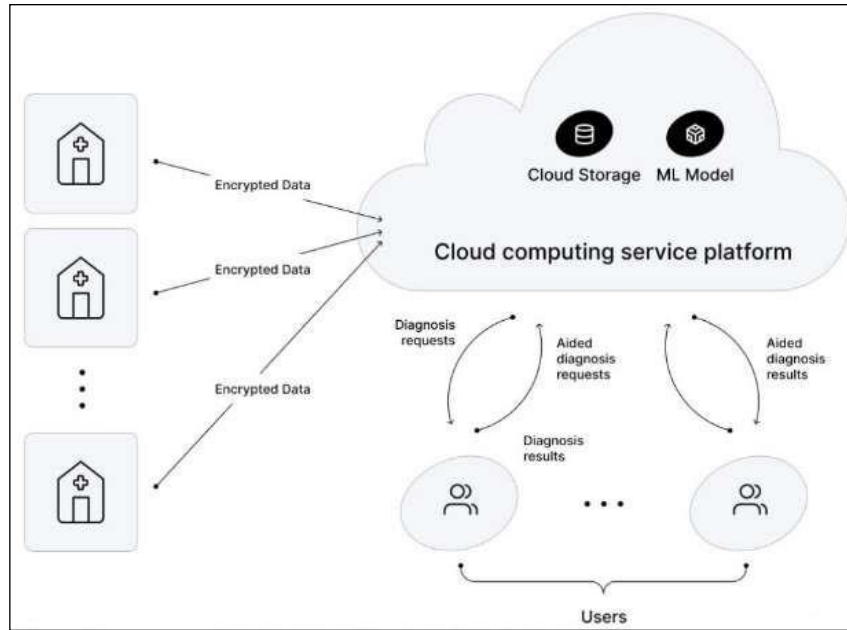
Cloud Migration is the process of transitioning applications, data, or any other business-critical resources from a local infrastructure to a cloud environment (Baharuddin et al., 2021). With the rising interest of organisations in organisational transformation, the cloud today is a valuable enabler to scale, optimize, and innovate. Cloud Computing secures on-demand delivery of computing resources, automated server provisioning, and better cost performance than infrastructure (Baharuddin et al., 2022). Nevertheless, moving from

## Role of AI and Machine Learning in Cloud Migration

Machine Learning and Artificial Intelligence are the most important technologies that help companies achieve better cloud migration outcomes. These technologies can mitigate traditional migration problems involving workload analysis, application dependencies, and the best resource configuration by automating migration processes (Lozano-Blasco et al., 2023). AI and ML also help maintain data security by analysing threats as soon as they are detected and how they should be resolved. In other words, while predictive analytics in AI will enable one to predict and detect possible migration risks, ML algorithms will improve migration over time (Pratsri & Nilsook, 2020). Fig. 2 explains the benefits of Machine Learning with Cloud Computing.

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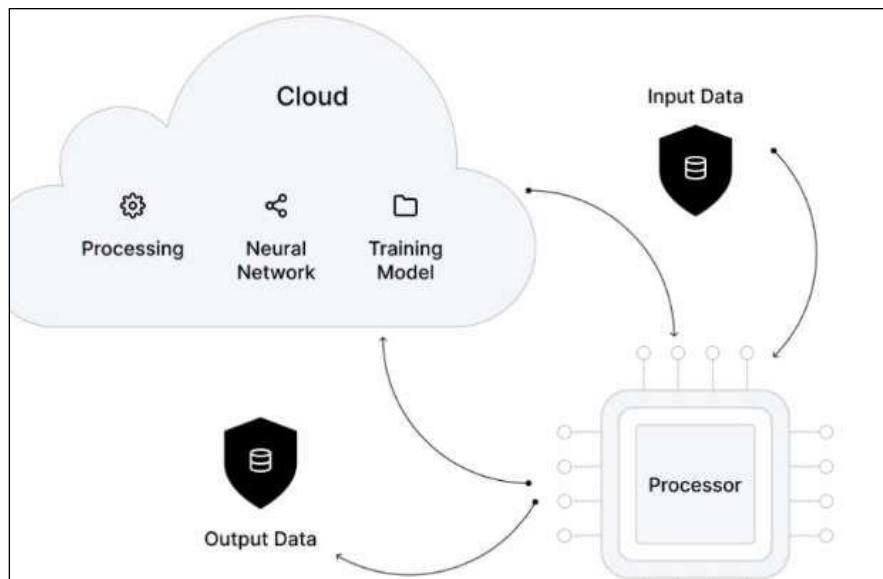


**Fig. 2: Benefits of Machine Learning with Cloud Computing** (<https://symphony-solutions.com/insights/machine-learning-in-the-cloud-what-are-the-benefits>)

### Challenges in Cloud Migration

However, organisations conducting cloud migration experience many challenges even as cloud computing accrues several benefits. Some of these include the feasibility of undertaking workloads without major disruption to service, data protection, and the overall

cost and complexity of workload transfer (Ullah, Nawi & Ouham, 2021). Nonetheless, existing systems built without the cloud in mind present their challenges since these systems need to be redesigned for proper deployment to the cloud (Akter et al., 2020). Further, there could be issues of poor migration performance and resource commitment, poor understanding of migration



**Fig. 3: Role of Cloud Computing in Machine Learning** (<https://symphony-solutions.com/insights/machine-learning-in-the-cloud-what-are-the-benefits>)

commitments by the applications that depend on it, poor visibility of applications and resource allocation, and poor identification of performance bottlenecks (Mogadem, Li & Meheretie, 2021). This makes cloud migration a risky business that needs sophisticated ways of handling operations and financial risks. Fig. 3 explains the Role of Cloud Computing in Machine Learning.

### Aim of the Review

Therefore, this systematic literature review (SLR) aims to present a review of how AI and ML enhance cloud migration processes, especially in terms of risk management, cost reduction, and efficiency. This review will discuss recent related studies and highlight the most preferred AI—and ML-based tools and approaches to cloud migration. Further, the review will discuss the existing use of AI-based cloud migration techniques and research the prospects of AI and ML in progress.

- *Evaluate Tools:* Compare AI and ML tools to effectively migrate to the cloud.
- *Analyse Costs:* Consider the potential relative cost advantage of AI and ML regarding migration.
- *Improve Decision-Making:* Review how AI and ML affect the decision to migrate.
- *Identify Best Practices:* Identify the proper guidelines for adopting AI and ML.
- *Address Challenges:* Discuss issues related to adopting AI and ML in migration.

### Methodology

Information gathering for this review involves conducting an SLR review to identify previous studies, conference

proceedings, and technical papers related to the role of AI and Machine Learning in optimising cloud migration processes. The primary aim is to explore how these technologies enhance cloud migration’s efficiency and performance and identify the tools and techniques currently in use.

This will entail a comprehensive search across several academic databases, including SpringerLink, ScienceDirect, and ERIC, using relevant keywords such as “Cloud Migration,” “Artificial Intelligence,” “Machine Learning,” “Efficiency,” “Tools,” and “Techniques.” A systematic search strategy will be employed, utilising Boolean operators to refine the search results. The inclusion criteria will focus on peer-reviewed articles published between 2020 and 2024 that are openly accessible and relevant to the study’s objectives.

The methodology will involve extracting key data regarding the AI and ML tools applied in cloud migration, the metrics used to assess efficiency, and any challenges identified in the implementation process. This study aims to fill knowledge gaps regarding the practical applications of AI and ML in cloud migration by synthesising information from diverse sources, ultimately contributing to a more comprehensive understanding of their impact on cloud migration strategies.

This systematic approach is expected to yield valuable insights into the implications of AI and Machine Learning for optimising cloud migration processes, thereby informing best practices and guiding future research in this critical area of cloud computing (Table 1).

**Table 1: PICOC**

Keyword	Description
<b>Cloud Migration</b>	The process of transferring data, applications, and services to a cloud computing environment.
<b>Artificial Intelligence</b>	Technology enabling machines to simulate human intelligence, enhancing decision-making and automation.
<b>Machine Learning</b>	A subset of AI that focuses on algorithms that learn from and make predictions based on data.
<b>Efficiency</b>	The ability to achieve maximum productivity with minimum wasted effort or expense in cloud migration.
<b>Tools</b>	Software and applications that facilitate cloud migration processes and enhance operational efficiency.
<b>Techniques</b>	Methods and strategies employed to optimize the cloud migration process using AI and ML.

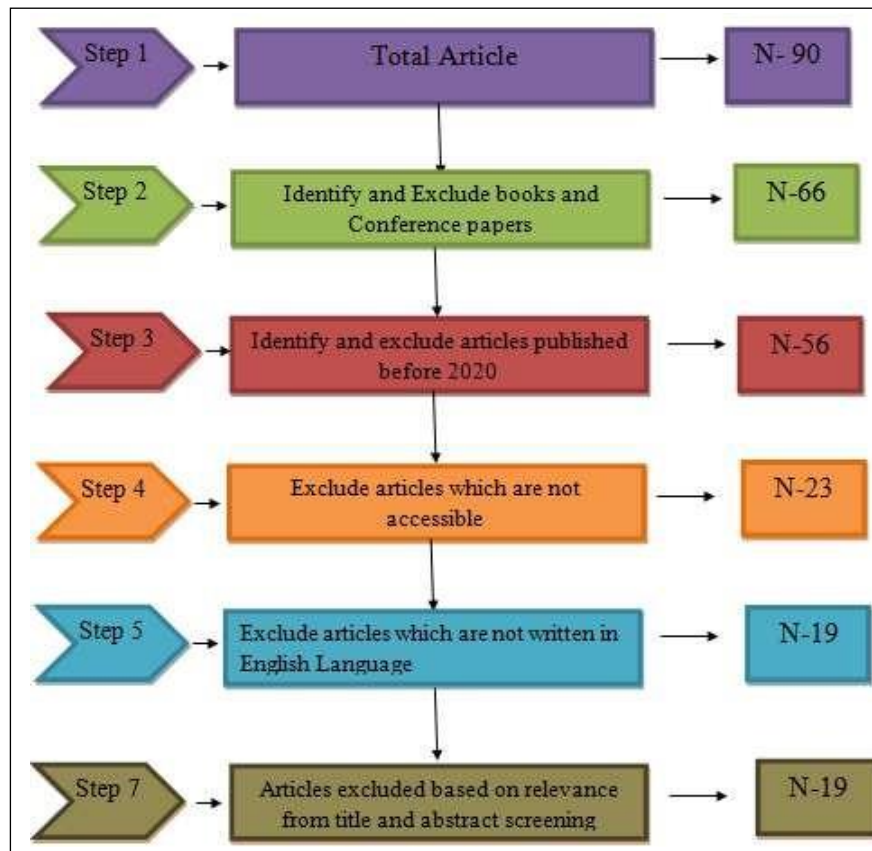
## Research Question

How do AI and Machine Learning improve the efficiency and performance of cloud migration processes?

## Search Strategy

In the process of article selection, the keywords relevant across the databases and scholarly journals will be used,

given the current trends related to the study. In this case, Grey's literature will also be searched to get as many findings about the topic as possible, including books and articles not included in the list of peer-reviewed ones. Due to the desire to improve the specificity and depth of the results, a systematic search applying keywords and Boolean operators will be employed in the selected databases (Fig. 4).



**Fig. 4: PRISMA Framework**

The following databases will be searched:

- SpringerLink
- ScienceDirect
- ERIC

- Efficiency
- Tools
- Techniques

Keywords:

- Cloud Migration
- Artificial Intelligence
- Machine Learning

## Search String

“Cloud Migration” AND (“Artificial Intelligence” OR “AI”)(ML OR “Machine Learning”) AND (“Efficiency” OR “Optimisation”) AND (“Tools” OR “Techniques”)

## Inclusion and Exclusion Criteria

Inclusion Criteria:

- Articles published in peer-reviewed journals or conference proceedings.
- Publication year between 2020 to 2024.
- Open-access articles only.
- Focus on Interoperability and Cross-Chain Security.
- Relevance to the topic is evident in the title and abstract.
- Article Published in English Language.

Exclusion Criteria:

- Publications outside the specified publication timeframe.
- Articles without open access availability.
- Irrelevant to title and abstract screening.
- Article Published in English Language.

## Data Extraction

Information for realising this SLR will be obtained through selective gathering of information from the identified studies, conference proceedings, and technical papers related to using AI and Machine Learning to enhance cloud migration strategies. Specific AI and ML tools and techniques identified and used will also be part of the key data points along with the measures of efficiency and effectiveness, as well as the performance of the diverse cases. These articles shall be reviewed to fit the research questions; otherwise, pieces of information will be classified regarding the impact of these technologies on cloud migration performance. Data extraction will also involve the methods employed in the studies, sample size, datasets used, and results realised. This is a well-coordinated approach to performing the review that will facilitate an organised collection of relevant information to achieve the review's objectives (Table 2).

**Table 2: Search Result**

Database	Total	Criteria 1	Criteria 2	Criteria 3	Final
ScienceDirect	56	50	46	15	11
SpringerLink	8	7	5	4	4
ERIC	26	9	5	4	4

## Data Synthesis

Therefore, this data synthesis method will entail identifying trends, patterns, and gaps in the current formally written literature on the role and application of AI and Machine Learning in cloud migration. This synthesis will provide interconnection between these findings and show that together with tools and techniques, there are shared and effective methods of enhancing operational fidelity and performance during migration. The synthesis will reveal the best practices and challenges people face when applying AI and ML to cloud migration by comparing outcomes across multiple scenarios and approaches. Last, more synergy would help acquire a richer and more comprehensive understanding of how these technologies could support cloud migration plans in the supporting field's theoretical perusal and functional real-location.

## Findings and Discussion

### Enhancement of Efficiency and Performance Through AI and Machine Learning

This systematic literature review finds that both AI and ML contribute to improving the effectiveness and accuracy of cloud migration initiatives. Other research shows that these technologies are useful tools for analysing the workload, allocating resources, and automating competitive activities, which minimises the time and energy needed to migrate. Technological solutions such as predictive analytics ensure organisations can recognise any disruptions during migration and adapt afloat to reduce the disruption to an absolute minimum. In addition, the ML algorithms used in migrations evolve; hence,

they adapt from previous migrations to deliver better recommendations and procedures.

## AI and ML Tools and Techniques in Cloud Migration

The literature review produces multiple tools and techniques that address the utilisation of AI and ML in cloud migration. Applications such as AWS Migration Hub, Google Cloud Migrate for Compute Engine, and Microsoft Azure Migrate use artificial intelligence to analyse the workload and advise where it should be migrated from and where it should be migrated to. These tools use features such as automated workload classification, dependency charting, and resource allocation. Research shows that companies using these AI and ML tools pay less and receive their migrations faster while optimising resources than conventional techniques.

## Challenges Faced in AI and ML Implementation

Nevertheless, organisations encounter limitations when utilising AI and ML for cloud migration. Important challenges are the shortage of qualified staff, problems with further integration of the system with other systems, and issues with data security and sanctions. Several studies suggest that organisations' inability to properly interpret insights generated by AI solutions about decision-making is a chief drawback of using AI. In addition, the initial investment required to employ sophisticated forms of AI and ML may pose a challenge, especially to relatively young firms.

## Best Practices and Guidelines for Adoption

This paper has highlighted several of the best practices for AI and ML adoption as part of cloud migration in the above review. Before adopting AI systems, organisations are advised to undertake a needs assessment of their current structures and migration plans. Furthermore, creating a culture that encourages learning/ upskilling across teams will improve the application of these technologies. Collaborating with experienced vendors and consultants can facilitate smoother integration and adoption processes.

## Answering the Research Question

*How do AI and Machine Learning improve the efficiency and performance of cloud migration processes?*

Paper (Baharuddin et al., 2021) is based on the deep learning methods for detecting malicious activities in IoT networks. It suggests using the neural network to help detect and counter cyber threats. Paper (Baharuddin et al., 2022) explores the combination of blockchain with artificial intelligence to increase the security of cloud computing by proposing a decentralised means of granting access and verifying identity. Paper (Lozano-Blasco et al., 2023) discusses the use of predictive models based on Machine Learning to identify phenotypes in rice, as well as the difficulty of constructing good predictors for different rice types worldwide. Paper (Pratsri & Nilsook, 2020) introduces adapting the electronic supply chain management framework with big data, AI, and IoT for delivering logistics services in the digital economy. Paper (Ullah, Nawi & Ouham, 2021) proposes a blockchain implementation for an autonomous energy trading system in microgrids, where AI and IoT technology are incorporated for effective marginal trading. Paper (Akteer et al., 2020) discusses a privacy-preserving AI architecture for the healthcare sector and the safe transfer of data and threat identification utilising federated learning and blockchain. Paper (Mogadem, Li & Meheretie, 2021) proposed an innovative decision support system based on AI in smart city energy management systems that discusses how AI and IoT can enhance energy efficiency and secure the systems simultaneously. Paper (Farahzadi, Farahsary & Rezazadeh, 2021) introduces artificial intelligence for traffic prediction in smart cities and presents an ensemble method to enhance prediction accuracy about the ethical aspects of AI. Paper (Kumar et al., 2024) resembles solid cybersecurity architecture for Zero Touch Networks (ZTN) and utilises AI with explainable models, smart contracts, and digital twins for intrusion detection and trust management. Paper (Bangui, Buhnova & Ge, 2023) analyses the ethical treatment of trust management technologies in SIoT and the effectiveness of AI maturity for trust enhancement in vehicular and underwater applications. Paper (Dominic, Cenggoro & Pardamean, 2023) reviewing other rice phenotypes on rice predicting models, which concluded

in complex global rice variety and interferences, requires complex deep learning models. Paper (Mohsen, 2023) covers the application of Supply Chain Operations through AI, IoT, and big data analytics that advance decision-making, logistics, and process management. Paper (Chang, Hahm, Xu, Vijayakumar & Liu, 2024) discusses Machine Learning techniques in banking contexts, focusing on B2B banking and green finance, identifying the impact of Machine Learning on risk management, profitability, and operational performance. Paper (Cao & Iansiti, 2022) examines the relationships between data architecture coherence and adaptive machine learning, especially within recently migrated server companies. Paper (Chang et al., 2023) discusses the services to be delivered through IoT and wearable devices across the healthcare ecosystem, suggesting a data analytic framework for supply chain management in the healthcare sector, considering the ethical issues. Paper (Xu, Jayne & Chang, 2024) introduces a multi-view deep learning model that incorporates emojis to address the problems of Twitter sentiment classification. Paper (Bukhari, Agarwal, Koundal & Zafar, 2023) describes anomaly detection in smart cities, which applies AI-resilient techniques in identifying and mitigating cyber threats in the IoT system. Paper (Pradeep Kumar et al., 2023) uses encryption techniques; the author suggests protecting patients' EHRs residing in the cloud, emphasising their privacy and security profile. Paper (Oliveri, Chiacchio, D'Urso, Munni & Russo, 2023) critiques Italian enterprises' current digital transformation strategies and introduces a fifth key training strategy for optimal digital transformation.

## Conclusion

The SLR analysis regarding the place and use of AI and Machine Learning in cloud migration has shown how the two tools improve the effectiveness and productivity of migration. This research shows that while AI and ML automate several classical migration tasks, they additionally offer tools for predictive risk assessment, resource optimisation, and real-time application performance monitoring. Integrating those advanced technologies addresses some of the impacts of migration to the cloud, such as workload evaluation and the management of application dependencies, to help foster the seamless and efficient movement of workloads to

cloud environments.

## Key Findings and Insights

- *Enhanced Efficiency:* AI and ML perform well because they ease cloud migration by automatically assessing and allocating resources.
- *Predictive Analytics:* These technologies offer a great deal of useful information that enables effective organisational risk management.
- *Improved Tools and Techniques:* AWS Migration Hub and Azure Migrate have been designed to help with migration by using artificial intelligence and machine learning.

## Future Directions

*Integration of Emerging Technologies:* Analysing how quantum computing and advanced AI models might enhance cloud migration and optimisation approaches.

*Longitudinal Studies:* Reviewing time series data that will enable researchers to determine more about the ongoing effects of adopting both AI and ML on general cloud migration performance in the long run.

*Broader Industry Applications:* Exploring the implementation of AI and ML in cloud migration at an organisational level with a special focus on regulatory environments relevant to various industries.

*User-Centric Approaches:* The creation of frameworks that would address journeys, feedback, and user experiences to improve user interfaces, AI, and ML to enable and support cloud migration.

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## Acronyms

- AI - Artificial Intelligence
- ML - Machine Learning
- SLR - Systematic Literature Review
- AWS - Amazon Web Services
- PRISMA - Preferred Reporting Items for Systematic Reviews and Meta-Analyses
- ERIC - Education Resources Information Center