

The Influence of Artificial Intelligence on Financial Decision Making and Risk Management

Sanjay Kumar Jha

Asst. Professor, S, N, College, Bhayander (E)

Sanjaykha2368

Abstract: *Artificial Intelligence (AI) is rapidly reshaping the way financial decisions are made and risks are managed in modern organisations. In today's dynamic financial environment, marked by digital transactions, market volatility, and massive data generation, traditional analytical tools often struggle to provide timely and accurate insights. AI offers a powerful alternative by enabling organisations to process complex datasets, identify hidden patterns, and generate predictive outcomes with greater speed and precision.*

This study explores how AI influences financial decision-making efficiency and strengthens risk management practices, particularly within the context of emerging economies like India. It examines applications such as machine learning-based forecasting, automated credit scoring, fraud detection systems, and intelligent financial reporting. Using a descriptive and analytical research design, data were collected from finance professionals across multiple sectors to assess the relationship between AI adoption and financial performance indicators.

The findings indicate that organisations integrating AI into financial processes experience improved forecasting accuracy, faster budgeting cycles, enhanced credit risk assessment, and stronger internal controls. However, concerns related to data privacy, cybersecurity risks, regulatory uncertainty, and algorithmic bias remain significant challenges.

Overall, the study concludes that AI does not replace managerial judgment but enhances it. When combined with human expertise and ethical governance, AI serves as a strategic support system that improves financial resilience, efficiency, and long-term sustainability..

Keywords: Artificial Intelligence, Financial Decision-Making, Risk Management, Machine Learning, Digital Finance, Financial Performance, India

1. Introduction

Financial decision-making plays a crucial role in determining organisational growth, stability, and long-term sustainability. Managers regularly make decisions related to capital investment, financing options, liquidity management, and risk mitigation. Traditionally, these decisions relied on historical financial statements, ratio analysis, statistical forecasting models, and managerial experience.

However, globalisation and digitalisation have significantly increased financial complexity. The rise of online transactions, algorithmic trading, fintech innovations, and cross-border capital flows has generated massive datasets that require advanced analytical capabilities. In such an environment, Artificial Intelligence (AI) has emerged as a transformative technological tool.

AI enables financial systems to analyse structured and unstructured data, identify hidden patterns, and generate predictive insights in real time. In India, the rapid expansion of digital payments, fintech startups, and AI-enabled banking platforms has demonstrated the practical relevance of intelligent systems in financial management.

This study aims to examine how AI influences financial decision-making efficiency and risk management effectiveness, with particular attention to emerging markets such as India.



2. Review of Literature

Several scholars have examined the growing influence of Artificial Intelligence on financial systems. Agrawal, Gans, and Goldfarb (2018) argued that AI reduced the cost of prediction, thereby improving the quality of managerial decision-making. Their work emphasised that AI primarily enhanced forecasting capabilities rather than replacing human decision-makers.

Brynjolfsson and McAfee (2017) observed that digital technologies, including AI, had improved productivity by automating analytical and operational processes. They noted that organisations adopting AI experienced enhanced efficiency and competitive advantage.

Davenport and Ronanki (2018) found that AI initiatives were most successful when integrated into core business functions. Their study suggested that companies using AI in financial analytics achieved better operational outcomes compared to firms using traditional systems.

In the context of financial forecasting, Lee and Shin (2020) demonstrated that machine learning algorithms outperformed traditional regression models in predicting market trends. Their research showed that AI-based systems provided greater predictive accuracy, particularly in volatile markets.

Indian scholars also contributed to this discourse. Bansal (2022) examined AI adoption in Indian banking and found that machine learning-based credit scoring systems significantly improved loan default prediction. Sharma and Gupta (2023) analyzed AI integration in Indian financial institutions and concluded that AI enhanced decision speed and reduced credit risk exposure.

Kumar and Singh (2022) studied AI applications in non-banking financial companies (NBFCs) and reported that alternative data analytics improved credit evaluation processes. Similarly, Rao (2023) found that AI-driven automation strengthened internal control mechanisms and minimised compliance errors in Indian corporations.

NITI Aayog (2021) emphasised that AI had the potential to enhance financial inclusion and risk assessment frameworks in India. However, Shrestha, Ben-Menahem, and von Krogh (2019) cautioned that AI-supported decision structures required strong governance mechanisms to prevent ethical and operational risks.

Although prior research confirmed the benefits of AI in finance, comprehensive empirical studies linking AI adoption with measurable financial performance indicators in emerging markets remained limited.

3. Research Gap

The review of literature indicated that most studies focused on developed economies or specific sub-sectors such as banking and fintech. Limited research had examined the combined impact of AI on financial decision-making efficiency and risk management effectiveness across diverse industries in the Indian context. Furthermore, empirical analysis connecting AI adoption levels with financial performance indicators remained insufficient. This study attempted to address these gaps by analysing the broader organisational impact of AI integration.

4. Objectives of the Study

The study aimed to:

Examine the impact of AI adoption on financial decision-making efficiency.

Analyse the role of AI in enhancing risk management practices.

Evaluate the relationship between AI integration and financial performance.

Identify challenges associated with AI implementation in financial systems.

5. Research Methodology

Research Design

The study adopted a descriptive and analytical research design. It aimed to evaluate the relationship between AI adoption and financial management effectiveness.



Data Collection

Primary data were collected through structured questionnaires distributed to finance professionals, including accountants, financial analysts, risk managers, and banking officers. Secondary data were gathered from academic journals, policy reports, industry publications, and financial statements.

Sample Size and Sampling Technique

A sample of 135 finance professionals from banking, manufacturing, fintech, and service sectors was selected using purposive sampling. Respondents were chosen based on their experience in financial analysis and decision-making.

Variables

Independent Variable:

Level of AI adoption in financial processes

Dependent Variables:

Financial decision-making efficiency

Risk management effectiveness

Financial performance indicators

Hypotheses

H1: AI adoption has a significant positive impact on financial decision-making efficiency.

H2: AI implementation significantly enhances risk management effectiveness.

H3: Organisations integrating AI show improved financial performance.

H4: AI adoption strengthens fraud detection and internal control systems.

5. Data Analysis Tools

The collected data were analysed using statistical tools such as correlation analysis, regression analysis, and ANOVA. Reliability of the questionnaire was tested using, which indicated acceptable internal consistency.

6. Data Analysis & Interpretation

The statistical analysis revealed a strong positive correlation between AI adoption and financial decision-making efficiency. Organisations using AI-based forecasting systems reported faster budgeting processes and improved predictive accuracy.

Regression analysis showed that AI adoption significantly predicted improvements in risk management effectiveness. Firms using AI-driven fraud detection systems experienced fewer instances of financial irregularities.

ANOVA results showed significant differences in financial performance between organisations with high AI integration and those relying on traditional systems. Higher profitability ratios and improved cost control were observed among AI adopters.

These results suggested that AI enhanced analytical depth and reduced uncertainty in financial decision-making processes.

7. Findings

The study found that AI significantly improved forecasting accuracy and budgeting efficiency.

AI-based credit scoring models reduced default rates and enhanced credit risk assessment.

Automated fraud detection systems strengthened internal controls and minimised financial losses.

Organisations adopting AI reported improved operational efficiency and financial performance.

Challenges such as data privacy concerns, regulatory uncertainty, and skill gaps remained significant barriers to effective AI implementation.



8. Suggestions & Conclusion

Suggestions

Organisations should adopt a phased and strategic approach toward AI implementation.

Continuous training programs should be conducted to enhance digital competencies among financial professionals.

Robust cybersecurity measures must be implemented to safeguard financial data.

Regulatory authorities should develop comprehensive frameworks to ensure ethical AI usage.

A hybrid decision-making model combining AI insights with human ability should be encouraged.

Conclusion

Artificial Intelligence has emerged as a powerful tool transforming financial decision-making and risk management. By enhancing predictive accuracy, strengthening internal controls, and reducing operational inefficiencies, AI contributed significantly to improved organisational performance. However, its successful integration depended on effective governance, ethical oversight, and managerial competence. In the Indian context, rapid digital transformation provided immense opportunities for AI-driven financial innovation. Sustainable financial management in the future will require a balanced integration of intelligent technologies and human judgment.

REFERENCES

- [1]. Agrawal, A., Gans, J., & Goldfarb, A. (2018). Prediction machines. Harvard Business Review Press.
- [2]. Bansal, R. (2022). Artificial intelligence adoption in Indian banking. *Indian Journal of Finance*, 16(4), 45–58
- [3]. Brynjolfsson, E., & McAfee, A. (2017). Machine, platform, crowd. W. W. Norton & Company.
- [4]. Davenport, T. H., & Ronanki, R. (2018). Artificial intelligence for the real world. *Harvard Business Review*, 96(1), 108–116.
- [5]. Kumar, S., & Singh, P. (2022). AI-driven credit risk assessment in Indian NBFCs. *Journal of Emerging Market Finance*, 21(3), 212–229.
- [6]. Lee, I., & Shin, Y. J. (2020). Machine learning for financial forecasting. *Sustainability*, 12(10), 1–15.
- [7]. NITI Aayog. (2021). National strategy for artificial intelligence. Government of India.
- [8]. Rao, M. (2023). Automation and internal controls in Indian corporations. *Indian Accounting Review*, 27(2), 89–104.
- [9]. Sharma, V., & Gupta, N. (2023). Artificial intelligence and financial performance in India. *Global Business Review*, 24(5), 745–760.

