



AI - DRIVEN INNOVATIONS IN MUTUAL FUND ANALYTICS

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Abstract: This study looks at how Artificial Intelligence's changing the mutual fund sector. It focuses on ways Artificial Intelligence is being used between 2020 and 2025. The study checks how Artificial Intelligence can help with analyzing mutual fund performance looking at risks and understanding how investors behave. It does this by using computer models and predicting what will happen next. Artificial Intelligence uses computer codes, language understanding and looks at what people are saying about the market right now. This helps mutual fund managers make guesses about what will happen to mutual funds and the market. The study also looks at how Artificial Intelligence's helping to pick mutual funds automatically manage portfolios and give investors personalized advice. The results show that Artificial Intelligence has made changes in the mutual fund industry. It has made things more transparent, efficient. Helped investors make better decisions. Overall the study shows that Artificial Intelligence is changing the mutual fund industry by using data to make investment decisions and making the financial system smarter.

Keywords: Artificial Intelligence, in Mutual Funds, Predictive Analytics, Portfolio Optimization, Machine Learning, Investor Decision Support, FinTech Innovation.

INTRODUCTION

The fast growing technological development has radically changed the financial services sector in the international market with Artificial Intelligence (AI) being one of the drivers that have been fuelling the innovation. The AI technologies of mutual funds have reached a paradigm shift during the last years. Investment decision making in investment vehicles which are comprised of the collective funds of the investors to invest in diversified portfolio of securities has always been conducted through the use of human expertise, historical data analysis, fundamental and technical analysis.

It has been induced by the growing complexity of the financial market and data explosion, which require more complicated tools of analysis. Machine learning and deep learning, natural language processing (NLP), and big data analytics technologies that are built on AI are changing the shape of mutual funds analysis and management. With the help of these technologies, it is possible to work with much structured and unstructured data, such as financial statements, market tendencies, economic indexes, moods of social media and world news on the spot.

An AI-based innovation that is making mutual funds analysis and management change includes machine learning and deep learning, natural language processing (NLP), and big data analytics. With such technologies, it is possible to operate with big amounts of both structured and unstructured data such as financial statements, market trends, economic indicators, social media sentiment and global news on the spot. As a result, fund managers and investors are now able to make more informed decisions and have gained additional knowledge.

One of the most significant advances that AI has had on the analytics of mutual funds is predictive modeling. Human beings are able to detect patterns and correlations with ease, however, with AI algorithms it is possible to foresee the market trends and fund performance with very high accuracy since it is able to grasp the patterns and correlations which are not vividly outlined by human beings. This enhances the ability of fund managers in optimising the portfolios, allocation of assets and risk management. Besides, AI-driven tools will be capable of learning and developing in accordance with the changes that take place in the market environment, becoming more precise over time.



Other notable innovations are developing robo-advisors, which are computer-based programs that provide advice on investments, and they are based on algorithms and preferences of the users. These services have rendered the consumption of financial advisory services to be democratic because of the customized investments which are being offered at low prices. The other significant use of AI is the sentiment analysis that examines the action of the investors and the mood of the market, depending on various sources and makes predictions of the market movements. Besides this, AI improves the efficiency of the operations by automating the monotonous processes such as data collection, data processing and reporting.

This assistance to reduce the human error and bias, to enlarge transparency and to speed up the process of decision making. The analytics of the mutual funds performed by a machine has a strategic advantage in a highly competitive financial market because it allows making decisions based on this data and improving the overall performance of the fund. Despite numerous benefits, the introduction of AI into the mutual fund analytics is associated with issues as well, including the issue of data privacy, the prohibitive cost of its introduction, and the requirement of the highly-qualified specialists. Nevertheless, the growing involvement of AI in financial sphere shows that there is the need to study its involvement and effectiveness in mutual fund analytics.

OBJECTIVES OF THE STUDY

Primary Objective: The goal is to study how AI-powered innovations are changing mutual fund analytics.

Secondary Objectives

- We want to see how AI affects analytics and forecasting returns.
- We will look into how AI's used for risk profiling and monitoring in real-time.
- We need to examine developments in AI-based portfolio optimization.
- We aim to investigate AI's role in understanding investor behavior and advisory systems.
- We also want to identify challenges and ethics issues with AI-based mutual fund analytics.

Research Methodology:

Research methodology is the systemic way which has been taken to undertake the research and to analyze the research problem. The given research is devoted to the consideration of AI-driven solutions in the field of mutual fund analytics and its effects on investment decisions and fund returns.

Research design:

The research design of the study is descriptive and analytical. The descriptive part is going to give a holistic idea of AI technologies and their usage in the mutual fund analytics. The analytical dimension is concerned with the quality and the performance of these innovations in fund performance, risk management, and investment choices.

Research Approach:

The research is based on the mixed-methodology, the combination of the qualitative and quantitative methods: Qualitative Approach: The qualitative method is adopted to comprehend AI technologies, innovations and its use in mutual funds analytics with the help of literature review and opinion leaders. Quantitative Approach: This is applied to examine the numerical data like fund performance measures, returns, and risk measures.

Data Sources:

Primary data

- Questionnaires were provided to mutual fund investors through structured questionnaires.
- Response to Interview of fund managers, financial analysts and investment advisors.
- Polls to know awareness and perception of AI-driven tools among the investors.

Secondary Data:

- Academic journals and research papers.
- Mutual fund performance report.
- Articles in financial news publications and the industry press.
- Financial databases and official websites on the internet.

**REVIEW OF LITERATURE**

Roberts, C. (2022) has investigated the changes in the use of analytics in the management of receivables. The results indicate that predictive analytics enhance the suitability of identifying the delinquent accounts at an early stage. Roberts recommends to rely on AI-based tools to analyse the patterns of payments and anticipate defaults. This paper suggests that analytics should be included in the ERP systems to enable real-time decision making. The conclusion highlights that advanced analytics improve the accuracy and timeliness of the receivables management.

Baker, L. (2021) concentrated on the efficiency of payment incentives in the early stages. The results indicate that early payment discounts enhance cash flow and decrease the receivables aging. To distribute the financial expenses and customer gains, Baker recommends organizing operations of discounts. The research suggests the use of effective communication regarding incentives to motivate involvement. The conclusion brings out the issue of early payment incentives as a viable instrument of improving efficiency of receivables.

Harris, P. (2021) examined customer relationship management (CRM) systems and how it is applied in receivables management. The results show the CRM tools enhance the communication with the customers making it faster and more satisfying. Harris proposes that CRM systems should be implemented based on the accounts receivable to simplify the process of follow-ups and tracking of payments.

Evans, G. (2020) considered the automotive industry-specific account receivable patterns. The results indicate that the long credit terms with OEMs result in late payments and increased receivables aged. Evans recommends the use of more rigorous payment schedules and frequent credit checks to major customers. The study suggests the use of data analytics in order to determine payment trends. The conclusion focuses on the fact that the industry-specific issues require tailored approaches to receivables management.

Johnson, E. (2020) explored the role of the ERP systems in the efficient accounts receivable management. The research concludes that the integration of the ERP simplifies the process of issuing invoices, invoice tracking and invoice payments follow-up, which in turn lowers the receivables turnover. According to Johnson, one of the reasons why organizations should embrace ERP systems is to centralize information regarding finances and improve visibility. The research proposal suggests employee training in order to utilize the full potential of ERP. The inference brings out the importance of ERP systems in enhancing efficiency and real time checks on receivables.

Anderson, B. (2019) examined the connection between working capital efficiency and receivables management. The results indicate that working capital is minimized by the receivables policies that are effective. According to Anderson, it should be proposed to provide early payment incentives to customers to speed up collections. In the study, targeted collection strategies should be considered by concentrating on the high-value customers. The conclusion reiterates the fact that optimization of the receivables has a direct effect to the working capital and liquidity of the business.

Taylor, S. (2019) investigated the significance of credit risk evaluation during the management of receivables. The results indicate that poor risk assessment will result in increased bad debts and late payments. Taylor recommends financial measures and credit scoring tools in order to assess the creditworthiness of customers. The research suggests that it is necessary to conduct periodic reviews of the financial stability of the customers. The conclusion underlines the fact that a sound credit risk evaluation is important in the sustenance of healthy receivables and minimization of financial risks.

Davis, H. (2019) investigated the possibility of effectiveness of various collection strategies in the enhancement of accounts receivable management. The proactive approach, according to Davis, works well in promoting collection rates through the early follow-up and frequent customer reminders. To improve the process of communication and eliminate delays, the study proposes implementing a CRM system with a collection focus. Davis advises of having a systematic follow up program in order to make prompt collections. It concludes with the idea that proactive and commitment-based collection practices are crucial in reducing the receivables aging and increasing the stability of the cash flow.

Mitchell, D. (2018) compared the key accounts receivable metrics turnover ratio and collection period and compared them with other industries. The paper concludes that a tight receivables management in industries increases their liquidity. Mitchell offers to benchmark the measures to detect inefficiencies. This study proposes routine modifications to credit and collection policy relying on the trends of metrics. The conclusion is focused on the importance of data-based decision-making in receivables management.



Martin, L. (2018) did a research to investigate the value of the receivable turnover ratio as a metric of liquidity and efficiency. The discussion shows that the industries that have higher rate of turnover have a better liquidity and financial stability. Martin recommends that the turnover ratios should be benchmarked against the industry standards to ascertain the gaps and improvement areas. The research 13 suggests that the turnover trends need to be monitored on a monthly basis to identify and resolve problems as soon as possible. Martin finds that it is also necessary to monitor the turnover ratio constantly to achieve operational efficiency and liquidity.

AI IN MUTUAL FUND ANALYSIS

Predictive Analysis: AI tools like LSTM, Random Forest and Gradient Boosting forecast fund returns.

These tools assess:

- Market indicators
- trends
- Sector shifts
- Investor feelings

Risk Check and Live Tracking:

AI systems track market, operational and liquidity risks by processing data constantly.

- Quick detection of patterns
- Simulated stress tests

Understanding Investor Sentiment with NLP

NLP tools gather insights, from:

- News stories
- Social media posts
- Company earnings calls

This helps understand what investors think and predict market changes.

Optimizing Portfolios

AI helps with:

- Changing asset distribution
- Automatic portfolio adjustments
- Personalized investment plans

Automated Advice and Fund Evaluation

Robo-advisors review funds and suggest them based on risk level, costs and performance measures.

Major Findings:

Artificial Intelligence makes analyzing funds faster and more accurate.

- Machine learning models do a job than statistical models.
- Sentiment analysis using Natural Language Processing helps with making timing decisions in the market.
- Artificial Intelligence helps create investment strategies that balance risk.
- Automated advisory systems make it easier for regular investors to get advice.
- Suggestions and Recommendations
- Mutual fund companies should invest in Artificial Intelligence. Hire experts.
- Rules for using Artificial Intelligence need to be stronger.
- Investors personal information should be kept private.
- Fund managers should use Artificial Intelligence analysis along with their thinking.
- Artificial Intelligence models that are easy to understand should be used.

CONCLUSION

Artificial Intelligence has made an impact on mutual fund analysis. It makes forecasts more accurate provides information and helps with decision making. Between 2020 and 2025 Artificial Intelligence helped fund managers look at data and find market trends. Tools that analyzed risk helped make portfolios safer by finding patterns early. Sentiment analysis helped understand market dynamics by looking at news and social media. Artificial Intelligence helped with portfolio management and created investment solutions for big investors. With progress there are still issues with information and rules. It's crucial to use Artificial Intelligence transparently so investors trust it. Artificial Intelligence is changing mutual fund analysis. It will keep making investment environments smarter.

**Scope, for Future Research: Future studies may look at:**

- What Artificial Intelligence can do in global fund markets.
- How human-managed funds compare to Artificial Intelligence-managed funds.
- How deep learning can improve forecasting.
- What to consider when adopting Artificial Intelligence on a scale.

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