



Assessing the Impact of Artificial Intelligence in Indian Securities Market: A Regulatory and Ethical Perspective

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ABSTRACT:

The advent of Artificial Intelligence (AI) into the capital market of India is revolutionizing the way investments are being made. What was once the sole preserve of human minds previously is now being regulated more and more by AI-based software that is capable of scanning humongous data, identifying patterns, and executing trades on its own. From robo-advisory and algorithmic trading websites to predictive analytics and sentiment-based investment products, AI is powering both retail and institutional investors such as mutual funds, hedge funds, pension funds. This rapidly progressing revolution is a structural transformation of financial decision making and market navigation. This paper critically evaluates the pervasive impact of AI across major market players, investment sentiments, and India's evolving regulatory landscape. Although AI has certainly enhanced the market efficiency, speed and convenience, it has presented a sequence of pressing concerns. These include algorithmic obscurity, data model-baked-in structural bias, over-reliance on AI-recommended recommendations, and system-level threats stemming from coordinated algorithmic responses (herding), and ethical concerns regarding undue threats to black-box machine learning models' interpretability and responsibility. For institutions, AI has strategic benefits like real-time portfolio optimization and risk modelling. Simultaneously, there are compliance and fiduciary concerns that goes with it. Retail platforms like ET Money, Small case, Kuvera and Sensibull have democratized investing but can be havoc-inducing through cognitive overloading, predictive analytics misinterpretation, and illusion of security through automated recommendations. This necessitates standardized investor education and responsibility at an emergent level. The role of the regulator is increasingly central to this ecosystem. SEBI has taken forward-looking initiatives like the Regulatory Sandbox and issued consultative papers on retail investor algorithmic trading. Nevertheless, there is a regulatory gap when it comes to AI-related risks like non-auditability, fairness, data protection, and systemic risk management. To fill this gap, the paper also examines cross-border trends such as the European Union's Artificial Intelligence Act, the UK Financial Conduct Authority (FCA) rules and U.S. Securities and Exchange Commission (SEC) policy notices, using the latter as a basis of planning India's regulation.

In brief, the research advocates practical policy suggestions to enhance the ethical, transparent, and responsible use of AI in India's securities market. These are mandatory disclosure of use of AI by investment managers, AI literacy certification of retail clients, XAI compliance requirements, indigenous AI innovation for Indian market conditions, and a whistle-blower protection system with AI abuse emphasis. The article argues that sustained success in capital markets for AI would be a case of achieving harmony between innovation and integrity, such that it protects investor confidence, market stability, and global competitiveness for India's financial sector.

Keywords: Artificial Intelligence, Indian Securities Market, Algorithmic Trading, Retail and Institutional Investors, Regulatory Framework

INTRODUCTION

Indian securities market is seeing a dramatic shift as Artificial Intelligence (AI) is being picked up at a fast pace, where AI earlier was utilized for automating back-end operations. It is today a strategic investment asset to employ in investment decision – making, risk management, and regulatory surveillance. AI is not only the facilitator of speed and organizational effectiveness in terms of stock trading and portfolio management, it is revolutionizing how data is understood, decisions are taken and strategies are implemented. Machine learning, natural language processing, and big data analytics are all AI technologies which are revolutionizing financial ecosystems across the world. And now these technologies are poised to boom in India, considering that AI is being inserted into trading algorithms, robo-advisors, sentiment analysis technology as well as real time fraud detection

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systems. AI has the ability to process huge structured and unstructured data, and with that, investors are able to detect market trends, predict trends, and asset allocation optimize with precision.

Regulators such as Securities and Exchange Board of India (SEBI) and Reserve Bank of India (RBI) have begun integrating AI in their supervisory systems. It can be utilized for processes like identifying insider trading and market manipulation and promoting innovation within the fintech sector, signaling a move towards more data driven, and intelligent one³. But the quick expansion of AI poses its own complications, including concerns over data quality, algorithm transparency, regulatory instability and ethics. One of the largest issues is that numerous AI models possess a 'black box' feature by which they frequently are opaque and incomprehensible. This fundamental opacity causes problems of accountability most notably when it is used by automated systems for making high stake investment choices. Also, financial markets are dynamic and, by nature, uncertain, and their action is subject to global – socio – political and disturbance, constraining the precision of AI systems based on prediction when acting on past data. Above all this, the growing use of third – party AI service providers entails a high security and regulatory compliance threat, especially in situations where the financial data are processed in the cloud.⁴To further add to the problem, India still does not have a specific set of regulations for AI use in securities trading and investment management. Although there are limitations to what artificial intelligence can do in the Indian securities market, the effect is enormous.

While the tools become increasingly sophisticated, AI is also facilitating easier access by retail investors to more sophisticated investment tools, improving institutional strategies in the process with the proper mix of innovation and regulation. Use of AI technology facilitates market efficiency and risk control while promoting financial inclusion and protection of investment goals. This study attempts to critically assess the impact of AI on investment policy in the Indian capital market. It delves into the potential of AI for decision-making, portfolio management, and understanding market dynamics to reflect both the inbuilt opportunities, and challenges of this technological change. It aims to provide a holistic insight into the impacts of AI and also provide worthwhile insights to the policy makers, innovators and the proper integration of AI in the Indian financial sector.

II. LITERATURE REVIEW

In order to understand the impact of AI and pre-existing mechanisms to combat the challenges and the prevalent loopholes, we have conducted Literature Review of 4 papers in this arena.

A Study on Impact of Artificial Intelligence and Machine Learning on Indian Financial Market

AUTHOR: R. Suresh and A. Vignesh

YEAR OF PUBLICATION: 2024

OBJECTIVE: This paper attempts to explore how technologies of AI & ML are reshaping some of the most important areas of the Indian financial system, namely, stock trading, risk assessment, fraud detection, and customer service.

FINDINGS AND CONCLUSION: AI and ML have enhanced speed and efficiency of trading through algorithmic means, improved loan risk detection and eliminated frauds by predictive analytics. They've also improved customer service responsiveness. The research, however, finds concerns such as rising market volatility, ethical concerns regarding algorithmic decision-making, and outdated regulations. It summed up that AI/ML are of high potential but must be used responsibly along with current laws and substantial effect in the long term.⁵

Impact of Artificial Intelligence on Stock Price Prediction in India

AUTHOR: Amalendu Bhunia

YEAR OF PUBLICATION: 2025

FINDINGS AND CONCLUSION: The study concludes that machine learning -based models, i.e., LSTM and hybrid models of ARIMA and deep learning, show greater robustness and accuracy than the traditional forecasting methods.

These artificial intelligence-based models work well to extract nonlinear interactions and dynamic market patterns and thus are more appropriate in the context of the volatility of the Indian stock market. There are challenges including data quality issues, potential for over fitting, uninterpretability of the model and regulatory ambiguity. Also, the absence of cost – benefit analysis and restricted real – time use in Indian financial institutions posed hurdles

³ Securities and Exchange Board of India. (n.d.). [Www.sebi.gov.in](https://www.sebi.gov.in). <https://www.sebi.gov.in>

⁴ World Economic Forum. (2023). *The World Economic Forum*. World Economic Forum. <https://www.weforum.org>

⁵ Suresh, R., & Vignesh, A. (2024). A study on impact of artificial intelligence and machine learning on Indian financial markets. *International Journal of Financial Management and Economics*, 7(2), 648–652. <https://doi.org/10.33545/26179210.2024.v7.i2.432>

for large – scale applicability. Though the study supports the inclusion of AI in financial forecasting, it emphasizes ethical use, model explainability, and regulatory synchronization for effective and ethical implementation.⁶

Prediction of Stocks and Stock Price using Artificial Intelligence

AUTHOR: S. P. Padalkar, S. S. Bere, D. B. Hanchate

YEAR OF PUBLICATION: 2024

OBJECTIVES: The research would aim to promote the precision in predicting stock price through the power of advanced machine learning and artificial intelligence algorithms including LSTM, RNN, ARIMA, and combined models. It would hope to incorporate dynamic model testing with sentiment analysis so that forecasting is made more accurate.

FINDINGS AND CONCLUSION: The study concludes that AI based models, namely ensemble methodologies using neural networks and sentiment analysis, are decidedly superior to classical methodologies. Despite the better predictive accuracy and agility offered by AI, ethical imperatives and matters of practical usability – transaction expenses and compliance costs – are big issues. The study calls for continuous model optimization, human action, and tight validation protocols so that faithful and ethical utilization of AI in finance prediction is adopted.⁷

Algorithmic Trading and AI: A Review of Strategies and Market Impact

AUTHOR: Wilhelmina Afua Addy, Adeola Olusola Ajayi – Nifise, and et al.

YEAR OF PUBLICATION: 2024

OBJECTIVE: The objective here is to explore how algorithmic trading and AI interact to revolutionize financial market. The emphasis is on development, strategic applications, market impacts, ethical concerns and regulatory reactions.

FINDINGS AND CONCLUSION: Trend following, statistical arbitrage, and high – frequency trading are greatly enhanced by machine learning.

There are still major problems, though, such as algorithmic bias, system risk associated with homogeneous trading patterns, and ethical concerns about transparency and fairness.

The paper highlights several case studies that show successful applications as well as catastrophic failures (such as the Knight Capital glitch, and Flash Crashes). The paper concludes that although the future of AI In trading is bright with trends such as quantum computing and decentralized finance there is a need for strong regulations and ethical standards to ensure innovation keeps pace with market stability.⁸

III. OBJECTIVE AND METHODOLOGY

The objectives of this research are as follows:

- To analyze the practical and regulatory impact of AI on investment behavior, market structure, and investor outcomes in the Indian securities market.
- To examine the implications of AI tools on institutional and retail investor decision-making, portfolio construction, and risk management.
- To evaluate the readiness and adequacy of current Indian regulatory frameworks (especially SEBI's guidelines) in managing AI-led financial innovation.
- To suggest a policy road map grounded on comparative jurisprudence and global best practices for moral adoption of AI in capital markets.

To enable this, the following Methodology has been adapted by us:

- Doctrinal Legal Analysis: Review of SEBI guidelines, consultation papers, circulars, and sandbox regulations has been conducted in order to analyze India's regulatory framework towards AI-based investment and trading.

⁶ Bhunia, A. (2025). Impact of Artificial Intelligence on Stock Price Prediction in India. *Journal of Finance and Accounting*, 13(1), 1–6. <https://doi.org/10.12691/jfa-13-1-1>

⁷ Padalkar, S. P., Bere, S., & Dinesh Bhagwan Hanchate. (2024). Prediction of Stocks and Stock Price using Artificial Intelligence. *ResearchGate*, 12(5), 2320–2882. https://www.researchgate.net/publication/383239912_Prediction_of_Stocks_and_Stock_Price_using_Artificial_Intelligence

⁸ Afua, W., None Adeola Olusola Ajayi-Nifise, Bello, G., Tubokirifuruar, S., None Olubusola Odeyemi, & None Titilola Falaiye. (2024). Algorithmic Trading and AI: A Review of Strategies and Market Impact. *World Journal of Advanced Engineering Technology and Sciences*, 11(1), 258–267. <https://doi.org/10.30574/wjaets.2024.11.1.0054>

- Comparative Study: Analysis of foreign AI regulatory policies such as that of EU Artificial Intelligence Act, U.S. SEC bulletins, and UK FCA guidelines on algorithmic decision-making has been conducted.
- Empirical Review: Comparative review of PwC, Deloitte, McKinsey, and World Economic Forum institutional reports to measure market trends and applied purposes of investment of AI.
- Case-based Observation: Theoretical formulation of thematic failures (e.g., over fitting leading to portfolio underperformance) and flash crashes for problem identification of algorithmic convergence introduction.
- Policy-Oriented Recommendations: Ranked policy suggestions according to order of ethical priority, technical applicability, and investor protection needs.

IV. EVOLUTION OF FINANCIAL MARKET BY MEANS OF ARTIFICIAL INTELLIGENCE

India's financial market has been transformed immensely with the implementation of AI, substituting the conventional method of trading with data-driven methods. Then came the age of algorithmic trading that was when institutional investors used automation to enhance trading efficiency.

With technology getting advanced, and using real – time data analysis, natural language processing, machine learning algorithms, and even auto-trading, the combination of cloud computing and big data has truly made finance accept AI as a powerful player in it. ⁹AI models able to instantly trade, draw information from financial news reports, and detect errors through the power of super-computer processing have revolutionized trading itself and made it even more efficient and precise.

With time, more technology in the form of real time data analytics development, natural language processing, and advancements in machine learning algorithms enabled AI models to carry out forecasting, risk evaluation and even self-trading. The combination of cloud computing and big data turns into a double – edged sword, assisting AI in solidifying its position in the financial sector. AI models with the ability of high –frequency trading signal, extract information from financial news, and immediately identify discrepancies, utilizing industrial – scale computing resources, has totally transformed the dynamics of market trading, making it efficient and accurate. Robo-advisors and intelligent investment platforms are fostering the emergence of AI powered fintech companies, termed as platforms that enabled financial inclusion for small and medium investors not just in India but globally. Customized investment approaches unseen for individual investors before are now available to them leading to the emergence of a fresh category of retail investors. Regulatory authorities such as SEBI are acquiring control of this movement towards AI in finance. They have compiled rules for algorithmic trading to increase fairness and transparency in the market. Through this transition from statistics to AI, market players can comprehend intricate data. Sophisticated AI risk assessment systems are assisting businesses to identify and control risks by processing large and diverse financial data sets.

Regulatory authorities such as SEBI have taken up the mantle of spearheading the move towards AI based analytics in the market. SEBI has established a framework for algorithmic trading with adequate guidelines to further promote fairness, transparency and risk management. The transition from method of utilizing statistics to AI, now market participants are able to decipher complex data. While that is happening, sophisticated AI powered risk assessment and surveillance systems are also assisting companies to identify and deal with systematic risks in advance by processing enormous and varied financial data sets. AI is transforming the stock market by making it more precise, accelerating processes, and giving greater access. ¹⁰

With technology continuing to evolve. AI will be contributing more to the financial industry to get more investors into the market. AI has been seen to changing stock market rapidly with more precision, quicker processes and simpler accessibility. The rate of technology advancements is increasing and in future, AI is likely to serve the financial world extensively while expanding the investors of the country as well.

V. IMPACT OF ARTIFICIAL INTELLIGENCE ON INVESTORS

With the rapid growth of Artificial Intelligence (AI) in the world of investment is revolutionizing how markets function and how investors make decisions. If you are an old hand at investing or only just starting out, it's worth understanding this evolving world. AI increasingly is being utilized to drive trading and investing decisions. On the upside, it can settle by using reasoning based on information rather than emotive intuition and thus potentially reduce

⁹ Padalkar, S. P., Bere, S., & Dinesh Bhagwan Hanchate. (2024). Prediction of Stocks and Stock Price using Artificial Intelligence. *ResearchGate*, 12(5), 2320–2882. https://www.researchgate.net/publication/383239912_Prediction_of_Stocks_and_Stock_Price_using_Artificial_Intelligence

¹⁰ SEBI / Consultation Paper on Algorithmic Trading by Retail Investors. (2021). Sebi.gov.in. https://www.sebi.gov.in/reports-and-statistics/reports/dec-2021/consultation-paper-on-algorithmic-trading-by-retail-investors_54515.html?trk=public_post_comment-text

abrupt market oscillations. On the downside, there is the reverse effect—if numerous AI programs respond to common market cues similarly, it would amplify market moves, inadvertently increasing volatility instead of dampening it.

AI is changing the game for investors in numerous ways. It provides investors with easy access to financial advice, expanding investment and financial information access. Automated systems also minimize the cost of portfolio management, offering professional-quality investment tools to individual investors. Perhaps the most potent asset of AI is its speed and capacity to process vast amounts of data, enabling investors to make faster and better-informed decisions. However, there's a potential pitfall that is, since the AI systems learn from historical data, and if those data contain human biases, those biases can be embedded in the AI's logic—leading to results that may not be entirely fair or balanced. Though, one of the biggest issues with relying on AI is that its decision-making process can be opaque. Most AI use cases are "black boxes," i.e., their decision-making is not transparent or explainable. This increases the challenge for end users to fully understand or question the basis of recommendations or decisions.

Impact on Institutional Investors

The use of Artificial Intelligence (AI) within the investment process of institutional investors—pension funds, mutual funds, sovereign wealth funds, and hedge funds—has transformed the capital allocation, risk management, and analysis of market dynamics. Institutional investors differ from individual investors in that they manage massive sums of money on behalf of others and thus need precision, speed, and scalability. Here, AI has not only become a tool for automation but as a strategy tool for outperformance and risk management. One of the most valuable applications of AI in institutional investing is data management at scale. Old-school analysis was controlled by quarterly reports, macroeconomic facts, and market indicators. Now, with AI technology, one can take in and digest huge volumes of structured and unstructured data ranging from weather and satellite data to alternative data such as credit card transactions, ESG scores, and news sentiment. Alpha generation by processing of such alternative data enables funds to spot weak trends and create alpha even before they realize those trends in mainstream data.¹¹

Additionally, institutional investors significantly use forecasting modeling in portfolio rebalancing, risk hedging, and exposure management. Algorithmic AI is also able to observe asset volatility and liquidity movement in real time and continuously adjust the portfolios accordingly. Hedge funds like Renaissance Technologies and Two Sigma have notoriously used machine learning and natural language processing to improve trade timing and execution, minimize slippage, and maximize return per unit of risk. These uses demonstrate that AI is not only being used as a predictor but also as an execution-based tactics engine. AI enables institutional investors to manage risk by allowing them to model out-of-the-world market scenarios – even black swans – and stress their portfolios accordingly. More sophisticated methods like reinforcement learning enable AI platforms to learn and adjust from market feedback, increasingly well prepared to manage volatility and downturns. Institutional investors like pension funds and insurance companies requiring steady returns will benefit from this ability an urgently needed hedge against long-term liabilities and macroeconomic shocks.¹²

In addition to this, AI provides regulatory reporting and compliance functionality in the form of RegTech solutions. Institutional asset managers are increasingly being required to fulfill transparency requirements by institutions such as SEBI, the SEC, and the FCA. Machine learning solutions are now utilized to monitor trading activity, identify suspicious trades, and generate automated compliance reports, thus mitigating operational risk and streamlining regulatory compliance requirements¹³. But all these benefits are at the risk of systemic and ethical danger. Institutional investors stand the chance of fitting AI models to the past data sets guaranteed to fail in the apprehension of future uncertainty. There is also the risk of algorithmic herding—where multiple funds employing similar AI models making similar types of positions in the market create systemic risk. This was observed during the COVID-19 market shock, when some portfolios based on AI experienced sudden drawdowns based on correlated risk exposures. The unintelligibility of certain institutional AI models adds yet another layer of complexity. If the fund managers themselves do not have the ability to explain the reasoning behind why an algorithm had made an allocation or a trade, then the accountability becomes confused. This is an issue in terms of fiduciary responsibilities owed to customers and causes potential issues under existing financial regulations requiring transparency and explainability of investment processes. In the coming years, institutional investing with AI will grow more interdisciplinary as regulators, ethicists, data scientists, and financial engineers

¹¹ *Uncovering the ground truth AI in Indian financial services*. (2022). <https://www.pwc.in/assets/pdfs/research-insights/2022/ai-adoption-in-indian-financial-services-and-related-challenges.pdf>

¹² McKinsey & Company. (2023). *McKinsey & Company*. McKinsey & Company. <https://www.mckinsey.com>

collaborate to ensure that such systems of unprecedented power are used responsibly. Institutional investors will be required to implement governance frameworks that prioritize explainability, fairness, and flexibility when leveraging AI for competitive advantage.

Impact on Retail Investors

Market Democratization Artificial Intelligence (AI) has significantly brought about financial markets into Indian retail investors. Retail investors are now able to construct and manage portfolios without or with minimum utilization of traditional financial advisers through products like robo-advisors and theme-based investing platforms—like ET Money, Kuvera, and Small case. These platforms offer features such as goal-based planning, rebalancing on the go, and risk analysis by individual profiles so that customers can tap into capital markets in a more convenient and informed manner. Thus, AI has emerged as a key driver of financial inclusion, particularly for first-time investors or investors in Tier II and Tier III cities who have previously not had access to formal investment advisory services.¹⁴

The Rise of Predictive and Sentiment-Based Solutions most importantly, the use of AI to analyze in real-time data from various sources including macroeconomic information, financial news announcements, and social sentiment may be the most important retail investing trend. Platforms such as Zerodha's Streak and Sensibull employ machine learning algorithm and pattern recognition models to support investors in taking well-informed decisions through reference to the market trend and human behavior. Although these platforms boost investor engagement and decision-making based on data, they also motivate some level of dependency on algorithmic predictions, which may not always remain wholly transparent or contextually aware. The actual risk is that of exaggerating the objectivity of these devices, which, though sophisticated, may be based on historical trends that could overlook live aberrations or market craziness.

The Problem of Cognitive Overload and Overdependence Decision-making software, as useful as they are programmed to simplify decision-making, can put their users at ease when in reality, they could not be. Most investors, especially new market investors, will be able to accept AI-suggested recommendations as absolute or fact, mixing model-produced predictions and guarantees. The "black box" character of much AI in fact makes such acceptance possible, giving none or little insight into how a conclusion is being formulated. All such lack of openness undermines accountability and can sow ruinous financial results, particularly where models misread information or interpret biased inputs. The Need for AI Literacy with increasing pervasiveness of retail investing, financial literacy classes must adapt to these developments. Financial education one-size-fits-all won't suffice anymore. Investors will need to learn how AI models operate, where they fall short, and risk of algorithmic bias. Even institutions like the Securities and Exchange Board of India (SEBI) might consider making it compulsory to be certified in AI literacy in the form of National Institute of Securities Markets (NISM) modules for those using advanced investment platforms. This would not only strengthen consumers but also reverse the evils of uninformed reliance on black box technology.

VI. APPLICATION OF ARTIFICIAL INTELLIGENCE IN INVESTMENT STRATEGIES

Integration of artificial intelligence in financial activities has profound implications on investment decisions. AI improves data processing, automates, and refines predictions. AI plays a crucial role in algorithmic trading, portfolio management, and trend forecasting. It revolutionizes how people make financial choices. AI in financial markets has a profound impact on algorithmic trading. AI systems use machine learning and execute trades in real-time as they process data. These systems typically used in high frequency trading, recognize and capitalize on market inefficiencies in microseconds. Global companies like Citadel Securities and XTX Markets illustrate how powerful these techniques are. In India more and more institutional investors and proprietary trading companies are using AI to enhance trading performance. Although, market volatility and transparency issues highlighted by incidents such as the 2010 Flash Crash have led Indian regulators such as SEBI to propose more stringent guidelines prioritizing explain ability and algorithmic accountability. AI is the driving force behind the emergence of robo-advisors, which are creating investment decisions based on investor profiles, risk tolerance, and market data as well as optimizing investment by automating.

Through continuous evolving to the user behavior and market condition, these new platforms are providing customized and low-cost investment solutions. Portfolio tools based on AI are making investment planning and making it more streamlined and possible and also possessing some advanced capabilities such as tax optimization and dynamic rebalancing¹⁵. The fact that they have the possibility to eliminate human factor also make them more dependable during volatile periods. AI functionality in predictive analytics gives investors an advantage, not only does AI rely on historical data, new sentiment,

¹⁴ Deloitte. (2023). *Deloitte UK / Audit, Consulting, Financial Advisory and Tax services*. Deloitte United Kingdom. <https://www2.deloitte.com>

¹⁵ U.S. Commodity Futures Trading Commission / U.S. Securities and Exchange Commission. (2010). *FINDINGS REGARDING THE MARKET EVENTS OF MAY 6, 2010 REPORT OF THE STAFFS OF THE CFTC AND SEC TO THE JOINT ADVISORY COMMITTEE ON EMERGING REGULATORY ISSUES Market Event Findings*. <https://www.sec.gov/news/studies/2010/marketevents-report.pdf>

economic data but also qualitative data such as news sentiment to build its models. AI techniques like LSTMs (Long Short Term Memory networks), and ensemble methods are used to make predictions and aid the investor in decision-making, so financial institutions use them to a great extent to forecast future trends, manage risk, and optimize the performance of the respective fund.

AI is crucial in preventing fraud and customers using behavior analysis and AI to offer more secure and use – friendly services. Rapid expansion in Fintech has been facilitated as a result of strong infrastructure and future – oriented regulatory policies. AI in investment strategies, as reinforcement to human insight and building a future where responsive algorithms and expert outlook coexist to make improved, effective financial decisions.

VII. REGULATORY AND ETHICAL CONSIDERATIONS

The quick advancement of Artificial Intelligence (AI) in India's capital market presents challenges as much as opportunities to regulators. AI has the capability to enhance efficiency, lower human errors, and ease access to investment products, but simultaneously, it carries with it existential issues of transparency, equity, data privacy, and algorithmic accountability. These require a sophisticated regulatory response—one that will be in a position to balance innovation and protection of the investor. The Existing Regulatory Framework in India India's main market regulator, the Securities and Exchange Board of India (SEBI), has moved early to deal with algorithmic and artificial intelligence-driven trading. A step in the right direction was taken with the establishment of a Regulatory Sandbox Framework in 2019, allowing fintech startups and regulated institutions to test new technology under the surveillance of SEBI. A Consultation Paper on Algorithmic Trading by Retail Investors was also released by SEBI in 2021, indicating its awareness of the heightened retail use of AI-based techniques. These efforts are to be welcomed but go only part of the way towards addressing AI-specific concerns such as model bias, explainability, or abuse of data. Ethical Hazards: Partiality, Lack of Clarity, and Abuse of Data AI algorithms are only as good as the data they are trained on. This is a material risk: if there are historic biases in data used to train the model—e.g., biased results by firm size, industry, or geography—those biases can make their way into the model's results. For securities markets, this can result in systemic discrimination, where smaller or newer companies will be underpriced solely because they don't have comparable historic data. Moreover, investors can be directed towards particular products or industries, not due to quality, but because of very highly embedded algorithmic prejudices.¹⁶

Another issue is the "black box" nature of AI models. Most people and even professionals cannot interpret or question the result produced by black box algorithms. In addition to producing ethical concerns regarding fairness, this lack of interpretability also creates compliance-related problems, particularly when companies cannot explain the basis for a certain investment recommendation provided by AI. Furthermore, investor information processing poses serious privacy issues. Since AI systems are built out of large-scale personal and behavioral data, they raise issues of consent, storage, disclosure, and third-party access. India, lacking an overarching data protection law (while the Digital Personal Data Protection Act, 2023 has been proposed) contributes to regulatory risk. International Regulatory Models: Dealing with Globally Comparable Regulators Across the globe, regulators have now set about building stronger AI systems. The Artificial Intelligence Act of the European Union views AI in finance as "high-risk" and demands transparency, human oversight, and accountability of algorithms. Likewise, the UK Financial Conduct Authority (FCA) demands that firms utilizing AI for decision-making outcomes make it audit-proof and understandable, particularly for use by consumers. In the US, the Securities and Exchange Commission (SEC) released a series of bulletins addressing the threat of algorithmic investment advice, particularly in being presented with so little disclosure or client awareness. India can take note of all this global experience and develop its own framework. Where it is a plus to have a sandbox being leveraged for innovation, it should be augmented by mandatory governance guidelines, regular AI tool audits, and enforceable investor rights to explanation and redress. The Way Forward: Ethical Guardrails for AI in Securities Markets Where finance meets AI, ethics has to move beyond the technical adequate.

The rulebook to pre-empt itself needs to:

- Facilitate explainability of AI-driven investment proposals, particularly to retail investors.
- Offer fairness audits to guarantee outputs are unbiased.
- Ensure they have strong data governance arrangements in place ahead of impending national privacy legislation.
- Offer opt-outs for investors from AI-driven personalization or targeting.
- Establish culture of algorithmic accountability with officers whose names appear behind AI model outputs. It is important to frame these frameworks not only for market fairness but also for safeguarding investor confidence in a very quickly digitized world.

¹⁶ Publications. (2016, April 21). FCA. <https://www.fca.org.uk/publications>

VIII. CHALLENGES FACED WHILE USING ARTIFICIAL INTELLIGENCE IN MAKING INVESTMENTS

Platform Failures and the Over fitting Trap Possibly the most critical issues that have emerged from the deployment of AI across financial markets is the failure of algorithmic models in real, turbulent market conditions. In India, several AI-based platforms have introduced thematic investment portfolios, chosen by machine learning models that have been trained on historical sectoral performance. But during periods of macroeconomic upheaval—such as the COVID-19 pandemic or the 2022 war in Ukraine—most of these AI-generated portfolios failed. Investors who invested based on advice based on historical growth in some industries (pharma or tech, say) exposed themselves to sector-specific declines, which the models were unable to predict due to overfitting and inability to dynamically re-tune. Over fitting effects are instances where models get too specialized to the past and fail when used to predict future patterns. This deficiency was glaringly revealed during periods of turbulence and liquidity loss in small-cap and mid-cap markets, where funds had been disproportionately invested by AI according to previous bull phases. These periods underscore the danger of solely depending on fixed AI outcomes with no scope for flexibility in the face of the constantly changing nature of market trends. Retail Investor Misinterpretation and Automation Risk Democratization of AI tools—expanding access—also caused mass misinterpretation and misuse by retail investors. With sites having "AI-powered" trading robots or prediction dashboards, most novice or amateur investors perceive these sites as providing certainty or assurances of guaranteed returns. Lack of disclaimers, explainability elements, or user training has led to cases where investors imitated algorithmic recommendations without cross-validation or even perceiving risk exposure. One such high-profile example of this type of case is the example of retail traders using auto-execution functionalities during the March 2020 market crash, causing enormous sell-offs at enormous losses, purely on the basis of panic signals generated by simplistic sentiment analysis bots. This example would be reflective of the necessity of platform accountability, risk labeling, and alert to users, particularly in the case of high-volatility markets. Regulatory Oversight and the Limits of Enforcement. The Securities and Exchange Board of India (SEBI) has identified the potential systemic risks of algorithmic and AI-based trading, particularly to the retail customer.

SEBI, through its 2021 consultation paper, proposed retail algorithmic strategies be regulated through subjecting them to broker authorization, system openness, and requirements of disclosure. Though it is hard to enforce the above due to the decentralized platforms of AI, the limited compliance capabilities in-house, and missing audit trails of black-box models. Regulators across the globe are also confronted with the challenge of striking a balance between innovation and investor protection. The Financial Conduct Authority (UK) and U.S. Securities and Exchange Commission have issued guidance on robo-advisory and AI governance, but implementation across platforms remains uneven. SEBI's sandbox initiative offers a test-bed for evaluating AI tools, yet participation remains voluntary, and its scope limited. Market Volatility and AI-Feedback Loops AI trading platforms, particularly in high-frequency settings, have the capability of magnifying market activity instead of dampening it. When various trading software react in the same way to identical price signals—e.g., RSI down or quarterly revenues—the outcome may be feedback loops, which yield snowballing sell-offs or sudden spikes.

US 2010 Flash Crash, although not prompted by AI independently, exposed vulnerabilities in algorithmic networks operating collectively with no human touch. The very same mini-crashes have since emerged in India on illiquid shares where AI-dominated algorithmic trades distorted prices and induced fake perceptions of liquidity. Those episodes make calls for firmer circuit-breaker regulations, diversity of models, and vigilant scrutiny of volatilities built up by AI on a real-time basis. Over-Reliance and the Black Swan Blind Spot AI models naturally fall behind—they learn from history to predict the future. Black swan occurrences like the COVID-19 pandemic, worldwide wars, or surprise policy moves (e.g., demonetization) lie beyond the model's sight space. Over-reliance on AI without any human intervention in such black swan occurrences normally results in mis-judgments in investment, abrupt drawdowns, and systemic risk. To this, banks will have to adopt hybrid strategies that blend AI intelligence with human reasoning, scenario simulation, and strong risk management systems. Regulators can also enforce model resilience tests and mandate human-in-the-loop controls on specific kinds of transactions. Future Trends and Future Prospects The direction of AI for the securities markets, especially of India, is moving towards a capital markets operation and investor sentiments, institutional and retail, transformation with financial products. As it grows in importance and depth in its creation, AI will be able to dig deeper in some of its key areas. Generative AI and Natural Language Interfaces. The advent of Generative AI (GenAI) has brought functionality beyond the dominant quantitative models. These technologies are being taught to generate investment research reports, translate lengthy regulatory releases, and engage with customers in natural language queries, replicating the role of a financial advisor. For the retail investor, this lessens reliance on human intermediaries but also brings into question the accuracy and supervisory liability of AI advice. Explainable and Ethical AI (XAI). The most talked-about trend around the world is the transition towards explainable AI in making financial decisions. Regulators in North America and Europe have begun requiring investment firms to provide explainability of autonomous decisions, particularly those that affect retail customers.

SEBI in India may need to impose similar requirements to make algorithmic outputs—be it portfolio suggestions or risk scores—auditable and understandable by compliance officers and clients. Human-AI Hybrid Advisory Models As much as the automation has improved, there is now an expanding literature base affirming hybrid advisory systems, wherein AI augments—but never replaces—human discernment. Experimentation has established that algorithm-centric methods will never work in scenarios of high-stress market scenarios or black swan occurrences. Models for the future

will integrate AI pattern recognition capabilities and human advisors' contextual awareness to produce more diversified and adaptive methodologies. Regulatory Technology (RegTech) and Oversight Mechanisms AI is also transforming the working of regulators. With RegTech solutions, SEBI and exchanges such as NSE can start using AI to track insider trading, front-running, and market manipulation in real-time. Under this active surveillance system, the regulators would shift from their present passive system and be able to react more contemporaneously and evidentially. ESG and AI Integration Finally, with environmental, social, and governance (ESG) investing more in the mainstream, AI systems more and more are used to track and score firms on sustainability scores. But until there is standard ESG data, there is the risk that firms will employ AI to overstate or make false declarations—a new green washing. Thus, incorporating ESG transparency into AI models remains a vital policy goal.¹⁷

IX. RECOMMENDATIONS

The arrival of Artificial Intelligence (AI) to India's securities market is of huge potential but with corresponding regulatory and ethics challenges that cannot be avoided. For maximizing the opportunity while protecting the integrity of markets and investor interests, the following pragmatic suggestions are offered:

Mandatory AI Disclosure Framework The regulators should compel mutual funds, portfolio managers, and listed firms to disclose unequivocally the extent and character of AI deployment in investment activity. Transparency regarding how algorithms are being used for portfolio construction, risk evaluation, and trade execution will improve investor trust and informed participation. In a market where retail investors increasingly rely on online platforms, this is particularly relevant. **AI Literacy Certification for Retail Investors** With an increasing number of AI-based investment products available, mainly in the form of mobile apps, there is the potential for abuse.

SEBI can promulgate a compulsory AI-literacy certification—similar to the National Institute of Securities Markets (NISM) modules—to investors using algorithmic platforms. This would expose the users to the principles of AI models, the drawbacks of algorithmic forecasts, and the dangers of over-dependence. **AI Audit Sandbox** SEBI's sandbox guidelines can also be extended in particular to provide a vertical for AI audits. Under this framework, financial institutions and fintech can voluntarily be audited through anonymized or synthetic data for detecting bias, back testing performance, and scrutinizing decision logic. It would improve AI tool trust with the assurance of data security and confidentiality of competitors. **Standardization of Explainable AI (XAI) Protocols** to reduce the obscurity typically associated with AI-based systems, SEBI can consider the standardization of Explainable AI (XAI) protocols making them compulsory. These protocols would mandate investment firms to have understandable reasons for automatic decisions—specifically high-risk or high-volume trades. Post hoc explanations would not only guarantee compliance but also facilitate the resolution of disputes and communication with clients. **Innovation Incentives for Ethical AI** to promote responsible innovation, SEBI, in collaboration with the Ministry of Electronics and Information Technology (MeitY), may initiate a fintech challenge or grants scheme aimed at creating AI tools with fairness, non-discrimination, and investor protection as prerequisites. This would incentivize startups to integrate ethical considerations into technical design. **Real-Time AI Monitoring for Systemic Risk** Instead of seeing AI as a risk in itself, it is equally possible for regulators to leverage it as a defense. There may be an AI monitoring system targeted in SEBI or NSE that detects anomalous spiking in trades, manipulation signals, or flash crash activity, enabling faster intervention in risky events. These systems have already been tested on pilot scales in markets such as the U.S. and EU.

Human-in-the-Loop (HITL) Safeguards For high frequency or quantity algorithmic trades, there must be a mandatory human-in-the-loop functionality. This would entail mandating supervisors or compliance officers in control to approve or review trades above a certain risk trigger. Human oversight can counter cascading errors born out of echo-chambers of algorithms. **Encouraging Localized AI Development** India's market dynamics differ drastically across regions and investor segments. SEBI and industry associations should encourage the development of AI models trained on local sentiment in local languages, regional economic cycles, and Indian retail investor behavioral trends. This will result in more contextual, precise, and hard predictions. **Academia-Regulator Cooperation** Regulation of AI should be as fast as the technology is developing. SEBI should form systematic alliances with institutions like IIMs, IITs, and NLU to co-create research, regulatory sandboxes, and whitepapers. This would increase policymaker vision and align with best international practices. **AI Whistleblower Protection Mechanism** SEBI's current whistleblower mechanism should be strengthened to address AI abuse—model manipulation, biased results, and algorithmic market rigging. Front-line developers or traders who notice inappropriate behavior should be able to report through protected, incentivized channels. It would create a culture of responsibility and ethical practice in tech-enabled financial services.

¹⁷ European Commission. (2021, April 21). *Proposal for a Regulation laying down harmonised rules on artificial intelligence | Shaping Europe's digital future*. European Commission. <https://digital-strategy.ec.europa.eu/en/library/proposal-regulation-laying-down-harmonised-rules-artificial-intelligence>

X. CONCLUSION

AI is entering the Indian security market as part and parcel, and swinging open doors to efficiency, access, and innovation. AI has reduced thresholds of access for retail investors, ramped up speed and accuracy in the speed and precision at which trades are being executed, and made possible portfolio building using data. All of this is subject to a very harsh caveat, however. Concerns such as algorithmic clarity, bias amplification, dependence, and raised volatility during a period of pressures in the system are still such challenging issues. The future of the market shall be a result of India's regulatory framework—SEBI, RBI, and interested exchanges—"keeping up with the Joneses.". There should be a policy of vision in which innovation is promoted and ethical guarding, model answerability, and education of investors are embedded. Most importantly, technology should augment judgment, not supplant it. The future paradigm cannot be one where machines dominate but one in which machines and human beings collaborate toward an optimal, efficient, transparent, and fair market system.