Open Access Research Journal of Science and Technology

Journals home page: https://oarjst.com/ ISSN: 2782-9960 (Online) OARJ OPEN ACCESS RESEARCH JOURNALS

(REVIEW ARTICLE)

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Enhanced business intelligence through the convergence of big data analytics, AI, Machine Learning, IoT and Blockchain

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Open Access Research Journal of Science and Technology, 2021, 02(02), 023-030

Publication history: Received on 25 July 2021; revised on 14 September 2021; accepted on 17 September 2021

Article DOI: https://doi.org/10.53022/oarjst.2021.2.2.0042

Abstract

In this era of the information economy, the integration of advanced technologies has become essential in making improvements to business intelligence. The paper provides a conceptual framework in the synergistic integration of Big Data Analytics, Artificial Intelligence, Machine Learning, the Internet of Things, and Blockchain for overcoming such limitations found in traditional BI solutions. The framework leverages the advantages of these technologies: big data analytics and IoT are trusted sources of comprehensive and real-time data, while AI and ML perform advanced predictive and prescriptive analytics. Blockchain technology comes in when it is a question of trusting issues in the management of data, providing security, transparency, and integrity to the data. It outlines a number of use cases for this integration in supply chain, retail, healthcare, and finance; predictive logistics and end-to-end visibility; customized customer experiences and inventory optimization; predictive diagnostics and real-time patient monitoring; fraud detection; and risk management. This paradigm proposed will drive businesses to be more creative, operationally more effective, and make better decisions. Technical difficulty, ethical concerns with AI and data security, and financial barriers are all carefully considered. Despite these challenges, the study has shown how integrating many technologies can result in a BI system that is suited to the needs of modern businesses. Examining possible future developments shows that more research, ingenuity, and practical implementation are required for this framework to reach its full potential. Businesses would also be able to steer toward safe, intelligent, and dynamic decision-making processes for effective continuity if this were put into practice.

Keywords: Artificial Intelligence (AI); Big Data Analytics; Business Intelligence; Internet of Things (IoT); Machine Learning (ML)

1. Introduction

In the contemporary business environment, the exponential growth of data has necessitated the evolution of Business Intelligence (BI) systems. Traditional BI frameworks, while instrumental in data collection and reporting, often fall short in providing real-time insights and predictive analytics. The integration of advanced technologies—Big Data Analytics, Artificial Intelligence (AI), Machine Learning (ML), the Internet of Things (IoT), and Blockchain—offers a promising avenue to enhance BI capabilities (Aazam et al., 2018). This paper explores a synergistic framework that amalgamates these technologies to overcome existing limitations and drive informed decision-making. Big Data Analytics serves as the cornerstone of modern BI, enabling organizations to process and analyze vast datasets to uncover hidden patterns and trends. The incorporation of AI and ML further augments this capability by facilitating advanced predictive algorithms and automation, thereby enhancing the accuracy and efficiency of data analysis. IoT contributes by providing real-time data from interconnected devices, enriching the depth and timeliness of insights (Al-Fuqaha et al., 2015). Blockchain technology ensures data integrity and security through its decentralized and immutable ledger system,

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fostering trust in data handling processes. The convergence of these technologies has been the subject of recent scholarly attention. The combination of Big Data, AI, and IoT in BI improves data analytics and decision-making processes. The research highlights that Big Data Analytics reveals concealed patterns, AI offers advanced predictive algorithms, and IoT supplies immediate data from interconnected devices, enhancing the detail and timeliness of insights (Banerjee et al., 2018). Similarly, Bottaccioli et al. (2017) discuss the integration of machine learning, blockchain technologies, and Big Data analytics for IoTs, emphasizing the potential to address open issues and critical challenges in the field.

However, the implementation of such an integrated framework is not without challenges. Technical complexities arise from the need to ensure interoperability among diverse systems and platforms. Ethical and legal concerns, particularly regarding data privacy and compliance with regulations, must be meticulously addressed. Financial barriers, including the high initial investment and the cost of maintaining advanced technological infrastructure, can also impede adoption. Despite these challenges, the potential benefits of integrating Big Data Analytics, AI, ML, IoT, and Blockchain into a cohesive BI framework are substantial, offering enhanced decision-making capabilities, operational efficiency, and opportunities for innovation (Cao et al., 2019). The practical applications of this integrated framework are vast and varied. In healthcare, real-time patient monitoring and predictive diagnostics can be achieved by combining IoT devices with AI analytics, leading to improved patient outcomes. In the retail sector, personalized customer experiences and efficient inventory management can be realized through the analysis of consumer data and supply chain information. The financial industry stands to benefit from enhanced fraud detection and risk management by leveraging predictive analytics and secure data transactions facilitated by blockchain (Dorri et al., 2017). In supply chain management, end-to-end visibility and predictive logistics can be achieved through the integration of IoT, Big Data, and AI, leading to increased operational efficiency.

This paper aims to propose a comprehensive framework for the integration of these technologies to enhance BI. By examining current literature and analyzing practical applications, we seek to provide insights into the benefits, challenges, and future directions of this integration. The goal is to offer a roadmap for organizations aiming to leverage this synergistic framework to achieve a competitive advantage in the data-driven business environment.

2. Literature Review and Research Gap

The convergence of Big Data Analytics, AI, ML, IoT, and Blockchain has been attracting the imagination of not a few scholars in the recent literature, considering their cumulative implications for Business Intelligence. A rather comprehensive work by Duan et al. (2019) has engaged the combination of Big Data, AI, and IoT in a BI perspective. They propose a structured approach: data from IoT devices, processed through Big Data technologies and analyzed with AI algorithms. This would combine to provide better depth and speed of insight to inform the decision-making process (Ed-daoudy et al., 2019).

On this note, Bottaccioli et al. (2017) described the integration of machine learning with blockchain technologies and Big Data analytics in the IoT context. The work further hints that such an integration may cure most of the prevailing issues of data security, scalability, and interoperability of BI systems. In line with these results, there is an urgent need for an integrated framework wherein strengths are borrowed from these three technologies to bring optimized data analytics along with operational efficiency (Ejaz et al., 2017). Against the background of such developments, there appears to be a serious lacuna as far as research on how all these technologies can come together under an integrated BI framework is concerned. Most of the earlier studies have considered the question of the bilateral integration of a few selected technologies, such as Big Data and AI or IoT and Blockchain, rather than taking up all the available tools. There is a literature gap with regard to research that will systematically investigate the synergistic potential of integrating Big Data Analytics, AI, ML, IoT, and Blockchain into an integrated framework for BI enhancement (Ejaz et al., 2017; Ed-daoudy et al., 2019).

Addressing this gap means the exploration of how these technologies can be put together in order to transcend the limitations of traditional BI systems. Future research should, therefore, focus on the development and validation of frameworks that not only combine these technologies but also ensure seamless interaction among them. Such studies will have to consider issues related to data interoperability, system scalability, real-time processing capabilities, and robust security measures (Flouris et al., 2017; Zheng et al., 2018). Further, empirical studies on the practical applications of such integrated frameworks in diverse industries would also be of great help toward their effectiveness and adaptability. Conclusion The literature to date provides only the basic understanding of how advanced technologies can be integrated into BI; what is still needed and lacking is research that integrates these elements into one optimized framework. Such efforts would be a great contribution to the development of BI systems, making them more responsive, intelligent, and secure in the emerging data-driven world.

3. Applications in Business Intelligence

Big Data Analytics, AI, ML, IoT, and Blockchain-like advanced technologies have redesigned Business Intelligence in industry after industry. Such technologies help an organization reap the most benefit from data in actionable insights to further improve efficiency, decision-making, and innovation (Figure 1). Some of the commanding applications of BI in the healthcare industry relate to areas like patient monitoring in real-time and predictive diagnosis. For instance, IoT devices such as wearables generate health data constantly that is observed by AI and ML algorithms in detecting anomalies or the prediction of a health condition (Hossain et al., 2019). For example, continuous monitoring of a patient who suffers from some kind of chronic ailment leads to early interventions that minimize the rate of readmission and therefore result in improvement. According to Sibanda et al. (2021), BI enables the personalization of customers' experiences and effective inventory management in the retail sector. Big Data Analytics and AI for pattern recognition in customers' purchasing habits and preferences will send personalized recommendations to customers. IoT-enabled smart shelves and predictive analytics further improve the inventory accuracy to reduce waste and guarantee the availability of a product. According to Chaudhary & Goel, 2021, BI is in wide application in the financial industry within fraud detection and risk management. ML models detect unusual patterns with real-time, potentially fraudulent activities when trained on transactional data. Similarly, in supply chain management, AI-powered risk assessment enables financial institutions to evaluate creditworthiness and control market risks according to Zhang et al., 2020. Business intelligence enhances the whole-chain visibility and predictive logistics.



Figure 1 Integration of artificial intelligence in applications of business intelligence

IoT sensors track shipments, providing real-time updates on location and condition. Predictive analytics help optimize delivery routes, reduce delays, and manage inventory levels, ensuring smooth operations (Jha et al., 2021). BI systems powered by advanced technologies enable predictive and prescriptive analytics, helping organizations anticipate future trends and take proactive measures. Predictive models use historical data to forecast outcomes, while prescriptive analytics suggest optimal actions based on these predictions. Real-time reporting and interactive dashboards provide decision-makers with up-to-date insights, ensuring informed strategies (Gandomi & Haider, 2021). Despite the benefits, implementing BI systems involves addressing several challenges. Tackling data silos is critical, as data spread across different systems hampers seamless analysis. Integrating disparate data sources through interoperable platforms ensures a unified view. Additionally, ensuring data security and privacy is paramount, especially in industries like healthcare and finance. In terms of solutions, blockchain technology creates secure and immutable records, hence protecting sensitive information (Khan et al., 2021).

4. Benefits and Opportunities

Big Data analytics amalgamated with AI, ML, and IoT on a blockchain-backed Business Intelligence system opens a whole spate of unprecedented benefits and opportunities that each organization is looking forward to harness. Coming back to that, these up-and-coming technologies address hosts of traditional limits of BI systems and open a whole new world toward new enabling avenues of growth and innovation by data (Koch et al., 2018; Yu et al., 2015). The major benefits that accrue with the integration of these technologies include improved decision-making due to insights gained from the data. Big Data Analytics processes vast amounts of structured and unstructured data to show hidden patterns and trends (Kolozali et al., 2014). Insights derived further enhance AI and ML to offer predictive and prescriptive analytics for enabling organizations to anticipate future trends and undertake proactive actions accordingly. IoT-driven real-time analytics ensures that the latest data is available with the decision-makers for more accurate and relevant strategies.

All these technologies go a long way in enhancing operational efficiency through automation of repetitive tasks and smoothing of workflows. For instance, ML algorithms can optimize processes down the value chain, such as demand forecasting and inventory management, while IoT devices monitor machinery and send warnings to managers before any machines break down (Makridakis et al., 2018; Jagielski et al., 2018). Further, blockchain ensures secure and transparent transactions, hence reducing audits, which are considered time-consuming. It helps a company carve out resources more appropriately to optimize its operating costs for maximum productivity. As customer satisfaction has become the driver in today's competitive marketplace, personalization stands first. It helps monitor preferences and behavior so businesses can deliver just what each customer is looking for (Mehdiyev et al., 2015; Rantos et al., 2018.). AI-enabled chatbots and recommendation engines will amaze customers with instant, relevant responses. IoT devices extend this idea of up-leverage, making customers more comfortable, with seamless users. This will improve customer loyalty by a big margin and will have long-term, positive impacts on brand reputation and loyalty. When these technologies combine in various ways, whole new innovation and business model opportunities arise. For instance, monetization of IoT data can come directly or via indirect services including predictive maintenance based on the very data, and health monitoring (Ramkumar et al., 2019). Blockchain applications open up a vista for developing different models in decentralized ways where new ecosystems can rise by way of inserting trust. AI and ML will lead the way to a new era of innovative smart product development and service development encompassing autonomous vehicles to personal healthcare solutions. It is innovations such as these that put businesses at the top of their industries, drive growth, and ensure competitive advantage. With all this, integrating advanced technologies into BI systems is going to conquer not only operational pain points but also open new avenues to growth and innovation. In such a way, organizations can be more agile, customer-oriented, and competitive in the rising tide of data-driven business.

5. Challenges and Limitations

Integration of advanced technologies such as Big Data Analytics, AI, ML, IoT, and Blockchain into BI systems has a lot of challenges and limitations. These can be further divided into technical challenges, ethical and legal issues, and financial constraints. The convergence of different technologies requires seamless integration, which is often complicated by the differences in architecture, data formats, and communication protocols. Interoperability amongst these systems requires sophisticated middleware solutions and standardized interfaces. Sgantzos et al. (2019) believe that the integration of Big Data, AI, and IoT becomes a formidable process to harmonize data collection, processing, and analysis, which may be technically demanding. As data coming from IoT devices and from other sources grows, scalability and performance of the system are at stake. The large-scale ingestion, storage, and real-time processing of data are required without loss of speed and efficiency. Kim et al. (2018) shed light on integrated system challenges with respect to scaling up, emphasizing the need for strong architectures to handle an increasing load of data.

The more the use of personal and sensitive data, the greater the privacy concerns. Organizations have to navigate through complex regulatory landscapes such as the General Data Protection Regulation to ensure compliance. Breach of data privacy will result in a number of legal consequences and loss of trust among consumers. Suliman et al. (2018) emphasize the right to data protection when embedding these technologies into BI systems. AI and ML algorithms can inadvertently perpetuate biases present in training data, leading to unfair or discriminatory outcomes. Some of the key considerations that need to be ensured in AI-driven decisions are fairness, accountability, and transparency. The ethical challenges associated with AI applications, highlighting the need for ongoing efforts to mitigate bias and ensure ethical use (Kang et al., 2016). In general, such advanced technologies require huge capital investments in infrastructure, software, and skilled manpower. Financial problems may be the constraint to organizational adaptation, especially for SMEs. According to Yin et al. (2019), many organizations face high costs associated with integrating technologies into their systems. An integrated BI system requires a workforce with highly specialized skills in the fields of data science,

AI, IoT, and blockchain technologies, which is not readily available. This shortage increases labor costs. Moreover, continuous maintenance and updates to keep the systems secure and efficient further increase the cost. According to Sgantzos et al. (2019), organizations face difficulties in finding and retaining skilled professionals, and there is also a continuous cost related to system upkeep.

The various steps to surmount these challenges would involve developing standardized integration frameworks, investment in scalable infrastructure, implementation of strong ethical guidelines, and investment in training and development. Proactive management of such limitations by an organization enables it to exploit the benefits of integrating advanced technologies into its BI system.

6. Future Directions

Business Intelligence is one of the most emergent technologies-driven fields. During 2020-2021, a variety of trends and innovations were spotted which would be decisive in building up the future of BI. One of the main trends going ahead is augmented analytics: embedding AI and ML in the BI tools themselves to automate data preparation and insight generation. This would amplify data analysis in speed and accessibility by users of all cadres of skills. For instance, in the year 2021 alone, CIO reported that the integration of AI and machine learning signals a new era of augmented analytics in BI (CIO, 2021). The other recent development in this area is an integration of AI, IoT, and Blockchain technologies. In fact, it targets making the BI system more secure, efficient, and intelligent. A 2021 study from Frontiers in Blockchain discusses a viewpoint where, owing to the convergence of these technologies, the next wave of digital transformation has already started to take place regarding improvements in data security and operation efficiency. On the AI side, improvement of natural language processing makes interaction with BI tools easier. NLP enables the perusal of data in conversational language to have more interactive querying and insight from it. According to an article published by Amar Infotech in 2021, NLP is fast turning the mainstay in making BI systems accessible to non-technical users. The integration of Blockchain with AI and IoT opens up newer possibilities of applications (Amar Infotech, 2021).



Figure 2 Showcasing core components including machine learning, pattern recognition, neural networks, automation, data mining, and problem solving for the future business solution

For instance, it might be that Blockchain provides a decentralized and secure structure for IoT devices, while their operation is optimized by AI. According to an article released by 4ire Labs in 2021, with the help of Blockchain integration, it is possible to achieve even faster and highly secure transactions within an IoT ecosystem-driving AI algorithms basically (4ire Labs, 2021). It calls for academia and industry to develop interoperable systems that exploit the emerging trends integrated by AI, IoT, and Blockchain technologies. In addition, with the volumes emanating from

devices, these shall be prime enablers for scalable architecture designs and securely designed. With all of these come ethical concerns: questions over data privacy, bias in algorithms, which therefore call for responsible deployment (Figure 2). Collaboration and continuous interaction of academia and industry will be fundamental in driving innovations and overcoming various challenges associated with the integration of such technologies into BI systems. Being in the know and involved in active, interdisciplinary research helps stakeholders really harness the power of these technologies in the transformation of Business Intelligence.

7. Conclusion

Big Data Analytics integrated with AI, ML, IoT, and Blockchain provides an integrated framework to update the prevailing paradigm for BI. These address some major gaps in conventional BI systems related to the provision of delayed insight, fragmented data, and various other security vulnerabilities with real-time analytics, seamless integration of data, and secure transaction of data. With such information at the podium, enterprises will be further empowered to make more accurate decisions, operational efficiencies, and innovations. The framework's transformative potential therefore lies in the predictive and prescriptive analytics it offers, automating workflows and creating value-added services that are needed in each industry. For example, AI and ML algorithms-inclusive insight to foresee the trends and behavior, while IoT devices enrich the data with real-time information. Such processes are made more secure through encryption by blockchain, and therefore, trust is instilled in the handling of data. Together, these technologies will continue to reshape organizational interaction with data, customers, and processes in ways that better position them for sustained competitiveness in a rapidly changing digital environment. The full benefit of this framework will only be realized with further research and implementation in practice. Research will be done in scalable and interoperable systems with issues concerning data privacy, regulatory compliance, and ethical challenges. Such success with the vision rests on collaboration at many levels through academia, industry practitioners, and policymakers. In the final analysis, integration of advanced technologies into BI is a shift rather than advancement in paradigm. A business entity could harness unparalleled opportunities for growth and innovation through embracing this framework. Further explorations and adoptions of such technologies mark the road to the future of data-driven decision-making, creation of intelligent, secure, responsive BI systems, which would meet the different demands of today's modern business environment.

Compliance with ethical standards

Acknowledgments

We would like to express our gratitude to all the co-authors for their contribution and critical reviews from the anonymous reviewers.

Funding

This research received no external funding.

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