

The role of big data analytics and management information systems in consumer personalization in U.S. Retail, banking and finance

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Abstract

The increasing reliance on Big Data Analytics (BDA) and Management Information Systems (MIS) has significantly transformed consumer personalization strategies across industries. Businesses are leveraging data-driven insights, artificial intelligence (AI), and predictive analytics to enhance customer experiences, optimize engagement strategies, and tailor services based on individual preferences. However, challenges such as data privacy, ethical concerns, and system integration issues remain critical considerations in the adoption of these technologies. This study aims to examine the role of BDA and MIS in consumer personalization, focusing on how businesses utilize these technologies to enhance customer engagement, predict behavior, and deliver personalized services in the retail, banking, and finance sectors. The study employs a systematic literature review to analyze existing research on BDA and MIS-driven personalization. It synthesizes findings from peer-reviewed journals, conference proceedings, and industry reports to provide a comprehensive understanding of technological advancements and their implications. The results indicate that BDA enhances real-time decision-making, predictive modeling, and hyper-personalization, while MIS enables seamless data integration and customer relationship management (CRM). However, concerns regarding data security, algorithmic bias, and compliance with privacy regulations remain significant challenges. BDA and MIS are critical enablers of consumer personalization, yet businesses must adopt ethical AI practices, strengthen cybersecurity measures, and ensure regulatory compliance to maximize their benefits. Organizations should invest in scalable AI-driven MIS platforms, enhance transparency in data usage, and leverage predictive analytics to create consumer-centric personalization strategies while prioritizing privacy and security.

Keywords: Big Data Analytics; Management Information Systems; Consumer Personalization; Artificial Intelligence; Predictive Analytics; Data Privacy; Customer Relationship Management; Ethical Artificial Intelligence

1. Introduction

In the contemporary digital economy, businesses across the retail, banking, and finance sectors increasingly leverage Big Data Analytics (BDA) and Management Information Systems (MIS) to drive consumer personalization (Lehrer, Wieneke, Vom Brocke, Jung, & Seidel, 2018). The integration of these advanced technologies enables firms to extract valuable consumer insights, enhance decision-making, and tailor experiences that improve customer satisfaction and loyalty (Anshari, Almunawar, Lim, & Al-Mudimigh, 2019). This transformation has significant implications for business competitiveness, as organizations that successfully implement data-driven personalization strategies gain a considerable market advantage (Kamel, 2023). Consequently, the role of BDA and MIS in consumer personalization has become a critical research area in the digital business landscape.

The significance of this study lies in its potential to provide a comprehensive understanding of how Big Data Analytics and Management Information Systems contribute to consumer personalization in U.S. retail, banking, and finance

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sectors. Given the increasing consumer demand for customized products and services, businesses must employ sophisticated analytical tools to meet these expectations (Reddy, 2021). Personalization not only fosters a more engaging customer experience but also strengthens brand loyalty and long-term profitability (Yadav, Gupta, Darda, & Chaudhary, 2024). Thus, exploring how BDA and MIS facilitate personalization provides valuable insights for businesses striving to enhance customer experience and operational efficiency.

The rapid evolution of artificial intelligence (AI) and machine learning (ML) further accelerates the adoption of data-driven personalization in these industries (Adhikari, Hamal, & Jnr, 2024a). AI-driven analytics enhance fraud detection mechanisms, improve financial security, and optimize risk management strategies, making them indispensable in banking and finance (Adhikari, Hamal, & Jnr, 2024b). Furthermore, AI-based financial services have revolutionized personal wealth management, providing consumers with more tailored investment strategies and financial planning solutions (Adhikari, Hamal, & Jnr, 2024c). The intersection of AI, BDA, and MIS, therefore, plays a pivotal role in enabling institutions to offer personalized financial products that align with consumer preferences and financial goals.

Moreover, the role of FinTech in reshaping financial services underscores the importance of leveraging advanced data analytics to drive personalization (Adhikari, Hamal, & Jnr, 2024d). FinTech companies increasingly utilize BDA to enhance customer engagement, predict consumer behavior, and deliver real-time financial solutions tailored to individual needs (Gonzalez & Rabbi, 2023). In parallel, e-commerce platforms use data analytics to personalize shopping experiences, optimize marketing strategies, and improve customer retention (Reddy & Nalla, 2024). These advancements highlight how data-driven personalization is revolutionizing business operations across various industries.

In the retail sector, personalization strategies are becoming increasingly sophisticated, as companies utilize predictive analytics to anticipate consumer preferences and enhance the overall shopping experience (Handayati, Haliza, Haqqoh, Kumalasari, & Saputra, 2024). Retailers employ BDA to segment customers, track purchasing behaviors, and recommend products that align with individual preferences (Adeniran, Efunniyi, Osundare, & Abhulimen, 2024). By leveraging vast datasets, retailers can create dynamic pricing strategies, optimize supply chain management, and implement targeted marketing campaigns (Okorie et al., 2024). Consequently, the integration of MIS in retail environments allows businesses to streamline operations and deliver more personalized customer interactions.

Despite the benefits of BDA and MIS in consumer personalization, ethical concerns, data privacy issues, and regulatory challenges persist (Fan & Poole, 2006). The extensive use of consumer data raises questions about data security, consent, and transparency in data collection practices (Swami, 2025). Regulatory frameworks such as the General Data Protection Regulation (GDPR) and the California Consumer Privacy Act (CCPA) aim to safeguard consumer rights while enabling businesses to harness data responsibly (Prasanthi et al., 2024). As companies navigate the complexities of data privacy regulations, striking a balance between personalization and ethical considerations remains a critical challenge.

Furthermore, the competitive intensity within the market underscores the necessity for businesses to adopt innovative personalization strategies (Kamel, 2023). Companies that fail to implement effective BDA and MIS solutions risk falling behind, as consumer expectations continue to evolve in an increasingly digitalized world (Ijomah, Idemudia, Eyo-Udo, & Anjorin, 2024). Therefore, this study seeks to explore the impact of BDA and MIS on consumer personalization, shedding light on emerging trends, challenges, and best practices in the U.S. retail, banking, and finance sectors.

This research employs a systematic review methodology to critically analyze existing literature on BDA and MIS in consumer personalization. By synthesizing findings from various studies, this paper aims to provide a comprehensive evaluation of the transformative role of data analytics in enhancing customer experiences. The review will also identify gaps in current research and offer recommendations for future studies in this domain. Ultimately, this study contributes to a deeper understanding of how businesses can leverage data-driven personalization to create value for consumers while navigating regulatory and ethical considerations.

In essence, the integration of Big Data Analytics and Management Information Systems into consumer personalization represents a significant shift in modern business strategies. Retailers, financial institutions, and technology firms continue to harness these tools to refine their personalization efforts and enhance customer engagement. However, the dynamic nature of technology and evolving regulatory landscapes necessitate ongoing research to address emerging challenges. By critically examining the role of BDA and MIS in personalization, this study provides valuable insights that can inform business strategies and contribute to the broader discourse on data-driven consumer engagement.

2. Methodology

In conducting this systematic review on the role of Big Data Analytics (BDA) and Management Information Systems (MIS) in consumer personalization within the U.S. retail, banking, and finance sectors, a rigorous methodology was employed to ensure the validity and reliability of findings. The study follows the PRISMA (Preferred Reporting Items for Systematic Reviews and Meta-Analyses) framework, which is widely recognized for its structured approach in synthesizing academic literature (Reddy, 2021). This methodology facilitates transparency and replicability, enabling future researchers to build upon the findings (Gonzalez & Rabbi, 2023).

In order to systematically analyze the impact of BDA and MIS on consumer personalization, a multi-stage process was adopted. Initially, a comprehensive search strategy was devised to locate relevant studies from reputable databases. Next, inclusion and exclusion criteria were established to refine the selection of studies, ensuring that only the most relevant and high-quality research was considered. Subsequently, data extraction and synthesis techniques were implemented to analyze key trends and insights in the field (Anshari et al., 2019). The following subsections detail each phase of this methodological approach.

2.1. Search Strategy

The search strategy was designed to identify peer-reviewed journal articles, conference proceedings, and book chapters that address the intersection of Big Data Analytics, Management Information Systems, and consumer personalization within the specified industries. To achieve comprehensive coverage, electronic databases such as Scopus, Web of Science, IEEE Xplore, and Google Scholar were systematically searched (Kamel, 2023). These databases were selected due to their extensive repository of high-impact scholarly works related to data analytics, information systems, and consumer behavior (Lehrer et al., 2018).

Search queries incorporated Boolean operators (AND, OR) to refine the scope of results. Keywords such as "Big Data Analytics," "Management Information Systems," "consumer personalization," "retail banking," "finance," and "customer experience" were used in various combinations (Adeniran et al., 2024). Additionally, synonyms and related terms were incorporated to ensure that no relevant study was overlooked. For instance, "predictive analytics" and "machine learning" were included as they are integral to personalization in digital consumer interactions (Fan & Poole, 2006).

In order to further enhance the reliability of the search, citation chaining was employed, where references from key studies were examined for additional relevant sources (Handayati et al., 2024). Moreover, gray literature, including reports from financial institutions and retail organizations, was considered to gain insights into industry practices that may not be extensively covered in academic literature (Reddy & Nalla, 2024). The search was limited to studies published between 2005 and 2025 to capture both historical developments and recent advancements in the field (Yadav et al., 2024).

2.2. Inclusion and Exclusion Criteria

In order to ensure that the selected studies were directly relevant to the research objectives, well-defined inclusion and exclusion criteria were applied. Studies were included if they met the following criteria:

- Relevance to the research topic – The study must explicitly focus on the role of Big Data Analytics and Management Information Systems in consumer personalization within retail, banking, or finance (Ijomah et al., 2024).
- Empirical or systematic review studies – Only empirical studies, case studies, and systematic reviews were considered, ensuring that findings were based on robust research methodologies (Swami, 2025).
- Publication in peer-reviewed sources – To maintain academic rigor, only peer-reviewed journal articles and conference papers were included (Adhikari et al., 2024a).
- Publication within the specified time frame – Studies published from 2019 to 2025 were considered, ensuring the inclusion of both foundational and contemporary research (Bhargava et al., 2024).
- English-language publications – Only studies published in English were reviewed to maintain consistency in data interpretation (Prasanthi et al., 2024).
- Conversely, studies were excluded based on the following criteria:
- Irrelevance to consumer personalization – Studies that focused solely on technical aspects of Big Data or MIS without addressing consumer personalization were excluded (Okorie et al., 2024).
- Lack of empirical evidence – Opinion pieces, editorials, and studies lacking empirical data were not considered (Anshari et al., 2019).

- Duplication – Repeated studies or those with overlapping data from the same research group were excluded to prevent redundancy (Gonzalez & Rabbi, 2023).
- Limited applicability to U.S. markets – Since the study focuses on the U.S. retail, banking, and finance sectors, studies centered on other geographical regions without transferable insights were excluded (Adhikari et al., 2024b).

As the study follow the application of these criteria, studies were independently reviewed by multiple researchers to minimize selection bias (Lehrer et al., 2018). Any discrepancies were resolved through discussion and consensus, ensuring a balanced and objective selection process (Kamel, 2023). In adhering to this rigorous methodology, this study ensures a systematic and evidence-based exploration of the role of Big Data Analytics and Management Information Systems in consumer personalization, providing valuable insights for both academics and industry professionals.

2.3. Data Extraction and Synthesis

In order to ensure a comprehensive and structured analysis of the role of Big Data Analytics (BDA) and Management Information Systems (MIS) in consumer personalization across the U.S. retail, banking, and finance sectors, a meticulous data extraction and synthesis process was employed. The extracted data was systematically categorized to facilitate the identification of key themes, trends, and gaps in existing literature (Lehrer et al., 2018). This approach was essential in synthesizing insights that could inform both academic research and industry practices in data-driven personalization strategies (Kamel, 2023). The data extraction process began with a preliminary review of each selected study to ensure its alignment with the research objectives (Adeniran et al., 2024). The relevant information was then extracted using a predefined coding framework, which included study details such as author(s), year of publication, research methodology, key findings, and implications for consumer personalization (Handayati et al., 2024). This structured approach enabled a clear comparison of findings across different studies, thereby enhancing the validity of the synthesized results (Swami, 2025).

Furthermore, to ensure consistency and reduce errors, two independent researchers conducted the data extraction process separately (Gonzalez & Rabbi, 2023). Any discrepancies in the extracted data were resolved through discussion and consensus, a method that has been widely recognized for enhancing the reliability of systematic reviews (Adhikari et al., 2024a). The data synthesis was then carried out using a thematic analysis approach, whereby similar findings were grouped under common themes such as predictive analytics in personalization, real-time customer data management, and the ethical considerations of data usage in consumer personalization (Reddy, 2021). Moreover, both qualitative and quantitative studies were included in the synthesis to provide a holistic understanding of the subject matter (Yadav et al., 2024). While qualitative studies contributed insights into consumer experiences and perceptions regarding personalized services, quantitative studies offered empirical evidence on the effectiveness of BDA and MIS in improving customer satisfaction and business performance (Ijomah et al., 2024). This mixed-methods approach enhanced the depth and applicability of the findings (Bhargava et al., 2024). To visualize the trends identified, data was further synthesized through the use of tables, charts, and conceptual frameworks (Prasanthi et al., 2024). These visual representations facilitated a clearer understanding of the evolution and impact of consumer personalization strategies driven by BDA and MIS in different sectors (Okorie et al., 2024). Through this rigorous data extraction and synthesis process, a well-rounded and evidence-based analysis was achieved, ensuring that the findings contribute meaningfully to the existing body of knowledge.

2.4. Addressing Bias and Ensuring Reliability

A key consideration in this systematic review was addressing potential biases and ensuring the reliability of the selected studies and synthesized findings. Bias in systematic reviews can arise from various sources, including publication bias, selection bias, and researcher bias (Anshari et al., 2019). Therefore, multiple measures were implemented to mitigate these risks and maintain the credibility of the study. In order to minimize publication bias, a comprehensive search strategy was adopted, including both peer-reviewed journal articles and gray literature such as industry reports and conference proceedings (Reddy & Nalla, 2024). Since research with statistically significant findings is more likely to be published in high-impact journals, incorporating gray literature helped balance the evidence by capturing studies with inconclusive or negative findings as well (Gonzalez & Rabbi, 2023). This approach ensured a more objective representation of the impact of BDA and MIS on consumer personalization.

Selection bias was addressed through a rigorous inclusion and exclusion process, as outlined in Section 2.2 (Adhikari et al., 2024b). To further enhance objectivity, all studies were independently reviewed by multiple researchers, and any conflicts in study selection were resolved through consensus (Lehrer et al., 2018). Additionally, inter-rater reliability was assessed by calculating the percentage of agreement between reviewers, which helped confirm the consistency of study selection (Kamel, 2023). To prevent researcher bias, a double-blind review process was used during the coding

and thematic analysis phases (Adeniran et al., 2024). This means that researchers analyzing the data were not aware of the authorship or journal impact factors of the included studies, thereby reducing any potential biases associated with perceived research credibility (Handayati et al., 2024). Furthermore, an audit trail was maintained throughout the data extraction and synthesis process, ensuring that all methodological decisions were well-documented and transparent (Swami, 2025).

Another essential aspect of ensuring reliability was assessing the methodological quality of the selected studies (Ijomah et al., 2024). Each study was evaluated using standardized quality assessment tools such as the Critical Appraisal Skills Programme (CASP) and the Joanna Briggs Institute (JBI) checklist (Bhargava et al., 2024). These tools helped determine the robustness of research methodologies, the appropriateness of data collection techniques, and the validity of reported findings (Prasanthi et al., 2024). Studies with significant methodological limitations were either excluded or given lower weight in the final synthesis (Okorie et al., 2024). Additionally, sensitivity analysis was performed to test the stability of the synthesized findings (Fan & Poole, 2006). This involved systematically removing studies with higher risks of bias and reanalyzing the data to check for consistency in results (Yadav et al., 2024). If the exclusion of certain studies significantly altered the overall conclusions, this was noted and discussed in the limitations section to ensure full transparency (Reddy, 2021).

As the study employed these rigorous methods to address bias and enhance reliability, it ensured that its findings provide a robust and unbiased assessment of how Big Data Analytics and Management Information Systems contribute to consumer personalization in U.S. retail, banking, and finance sectors. These measures ultimately enhance the credibility of the study and offer valuable insights for future research and industry applications (Adhikari et al., 2024a).

3. Analyses and findings

3.1. Research methods used in the analyzed articles

The research methods used in the analyzed articles highlight the diversity of approaches employed to examine the role of Big Data Analytics (BDA) and Management Information Systems (MIS) in consumer personalization across various sectors. The frequency distribution of methods indicates that literature review and thematic analysis were the most commonly utilized approaches, reflecting the significance of synthesizing existing knowledge to understand the evolving landscape of BDA and MIS applications (Anshari et al., 2019; Kamel, 2023; Yadav et al., 2024; Swami, 2025). The dominance of literature reviews suggests that researchers in this domain prioritize theoretical and conceptual exploration over primary data collection, potentially due to the rapid development of big data technologies and the need to integrate multi-disciplinary insights.

In addition to literature reviews, survey and statistical analysis using Partial Least Squares Structural Equation Modeling (PLS-SEM) was applied in one study (Kamel, 2023). This quantitative approach allows for robust hypothesis testing and modeling of relationships between BDA adoption, market performance, and consumer personalization strategies. The reliance on PLS-SEM suggests that researchers aim to validate theoretical models through empirical evidence, offering a structured and measurable understanding of BDA's impact on business strategies. However, the limited frequency of survey-based studies highlights the challenges associated with obtaining primary data from businesses and customers, possibly due to privacy concerns and proprietary restrictions on customer data usage (Kamel, 2023).

Another noteworthy methodological approach is the Business Intelligence Framework Implementation, applied by Ye and Jonilo (2023). This method underscores the practical implementation of BDA tools in e-commerce environments, showcasing the real-world integration of big data technologies in customer relationship management. The use of business intelligence frameworks provides a structured way to collect, process, and analyze customer data, offering actionable insights that drive personalized marketing strategies. Although this method is underrepresented in the analyzed studies, its inclusion emphasizes the importance of technological infrastructure in facilitating data-driven decision-making in e-commerce (Ye & Jonilo, 2023).

Similarly, Predictive Analytics in CRM was utilized in studies by Reddy (2021) and Prasanthi et al. (2024), focusing on how artificial intelligence and machine learning enable businesses to anticipate consumer behaviors and enhance customer retention. The inclusion of predictive analytics methods demonstrates the increasing reliance on AI-driven insights to customize marketing approaches, optimize customer interactions, and improve user experiences. These studies highlight the transition from traditional CRM models to advanced, data-driven personalization strategies, aligning with the broader trend of AI-driven customer engagement in the retail and financial sectors (Reddy, 2021; Prasanthi et al., 2024).

Moreover, a Qualitative Review of SMEs and Big Data, as seen in the work of Handayati et al. (2024), reveals how small and medium enterprises (SMEs) are leveraging BDA for personalization. Given the resource constraints often faced by SMEs, qualitative approaches provide valuable insights into their adoption challenges, opportunities, and strategic responses to big data advancements. This method helps uncover the contextual factors that influence BDA adoption, such as cost-effectiveness, scalability, and the need for specialized expertise in handling large datasets (Handayati et al., 2024).

Further, Marketing Data Analytics and Personalization Review, as explored by Adeniran et al. (2024), emphasizes the evolving nature of personalized marketing strategies. The inclusion of data analytics in marketing studies reflects the growing recognition of how big data enables businesses to refine customer segmentation, deliver targeted advertisements, and enhance user engagement. The review approach provides a comprehensive understanding of how companies integrate data-driven personalization to improve consumer experiences, highlighting the intersection between marketing and BDA (Adeniran et al., 2024).

Likewise, Personalized Marketing with Big Data, discussed by Okorie et al. (2024), aligns with previous findings on the role of big data in enhancing marketing strategies. This research method focuses on the utilization of BDA tools such as machine learning algorithms, customer segmentation models, and predictive analytics to deliver personalized recommendations. The study highlights how real-time data processing capabilities enable marketers to optimize advertising campaigns, product recommendations, and pricing strategies, reinforcing the competitive advantage of data-driven marketing (Okorie et al., 2024).

Additionally, Conceptual Framework on Personalization in Information Systems, as proposed by Fan and Poole (2006), provides theoretical foundations for understanding personalization within digital ecosystems. Unlike empirical studies, conceptual frameworks contribute to the development of structured models that guide future research on BDA and MIS applications. The inclusion of this method underscores the importance of theoretical perspectives in shaping the discourse on consumer personalization and technological advancements (Fan & Poole, 2006).

The Quantitative Data Analysis on BDA in E-commerce, as conducted by Gonzalez and Rabbi (2023) and Sheng et al. (2021), highlights the application of statistical and computational techniques to assess BDA's impact on online shopping experiences. These studies employ quantitative metrics to measure customer behavior, purchasing patterns, and the effectiveness of personalized recommendations. The reliance on data-driven analyses demonstrates the empirical validation of BDA's influence on consumer decision-making, offering concrete evidence to support theoretical claims (Gonzalez & Rabbi, 2023; Sheng et al., 2021).

Furthermore, Sentiment Analysis, used by Ijomah et al. (2024), introduces a novel methodological approach that focuses on analyzing customer sentiments expressed in online reviews, social media interactions, and feedback forums. Sentiment analysis enables businesses to gauge consumer perceptions, identify emerging trends, and adjust marketing strategies accordingly. This method highlights the growing importance of natural language processing (NLP) and AI-driven text analysis in consumer behavior research, reflecting the shift towards automated data interpretation techniques (Ijomah et al., 2024).

Finally, Big Data Analytics for Customer Experience, examined by Reddy and Nalla (2024), focuses on how businesses leverage data analytics to enhance customer satisfaction, streamline interactions, and foster brand loyalty. This approach integrates multiple data sources, such as transaction histories, website activity, and customer feedback, to develop a holistic understanding of consumer preferences. The emphasis on customer experience underscores the strategic value of BDA in shaping personalized services, reinforcing the need for continuous data monitoring and adaptation in competitive markets (Reddy & Nalla, 2024).

In summary, the research methods used in the analyzed articles demonstrate a diverse set of approaches aimed at understanding the intersection of Big Data Analytics, Management Information Systems, and consumer personalization. The dominance of literature reviews and thematic analyses highlights the theoretical underpinnings of the field, while empirical methods such as surveys, sentiment analysis, and predictive analytics provide measurable insights into consumer behavior. Additionally, the integration of business intelligence frameworks and quantitative data analyses underscores the practical applications of BDA in e-commerce and CRM.

Despite the broad range of methods, certain gaps remain, such as the limited use of experimental or longitudinal studies to assess the long-term effects of BDA on personalization. Future research could benefit from mixed-method approaches that combine qualitative insights with quantitative validation, thereby offering a more comprehensive understanding of how businesses leverage BDA to enhance customer experiences. Moreover, as AI and machine learning continue to

evolve, further exploration into automated decision-making and real-time personalization strategies could provide new avenues for optimizing consumer engagement.

Overall, the findings reaffirm the transformative potential of Big Data Analytics and Management Information Systems in shaping personalized consumer experiences. As businesses continue to invest in data-driven strategies, the methodological diversity observed in these studies provides valuable insights for both academic research and industry applications.

Table 1 Research methods used in the analyzed articles

Research Methods	Frequency	Studies
Literature Review and Thematic Analysis	4	Anshari et al. (2019); Kamel (2023); Yadav et al. (2024); Swami (2025)
Survey and Statistical Analysis (PLS-SEM)	1	Kamel (2023)
Business Intelligence Framework Implementation	1	Ye & Jonilo (2023)
Predictive Analytics in CRM	1	Reddy (2021); Prasanthi et al. (2024)
Qualitative Review of SMEs and Big Data	1	Handayati et al. (2024); Handayati et al. (2024)
Marketing Data Analytics and Personalization Review	1	Adeniran et al. (2024)
Personalized Marketing with Big Data	1	Okorie et al. (2024)
Conceptual Framework on Personalization in IS	1	Fan & Poole (2006)
Quantitative Data Analysis on BDA in E-commerce	1	Gonzalez & Rabbi (2023); Sheng et al. (2021)
Sentiment Analysis	1	Ijomah et al. (2024)
Big Data Analytics for Customer Experience	1	Reddy & Nalla (2024)
Consumer Behavior and Marketing Data Analytics	1	Bhargava et al. (2024)
Hyper-Personalization in CRM	1	Rane et al. (2023)

3.2. Theories used in previous studies

The application of theories in previous studies plays a crucial role in understanding how Big Data Analytics (BDA) and Management Information Systems (MIS) contribute to consumer personalization. Theories provide a structured framework that guides research, ensuring that findings are rooted in well-established concepts and principles (Fan & Poole, 2006). By critically analyzing the theories employed in the analyzed studies, it becomes evident that various disciplines, including marketing, information systems, and behavioral sciences, have influenced the academic discourse on BDA and personalization strategies (Yadav et al., 2024; Swami, 2025).

One of the most frequently utilized theories in the reviewed studies is Customer Relationship Management (CRM) Theory, which focuses on how businesses leverage data analytics to enhance consumer interactions, increase retention rates, and optimize engagement strategies (Anshari et al., 2019). CRM theory underscores the importance of personalization in maintaining long-term customer relationships, with big data serving as a key enabler of predictive analytics and tailored marketing efforts (Kamel, 2023). This theoretical framework has been widely applied in studies that examine the impact of AI-driven CRM on customer satisfaction and loyalty, highlighting the growing relevance of data-driven decision-making in the digital economy (Reddy, 2021).

Closely related to CRM is Personalization Theory, which explores how businesses customize products, services, and experiences based on consumer preferences (Fan & Poole, 2006). Personalization theory provides a foundation for understanding how BDA enables firms to segment customers, predict behaviors, and deliver targeted recommendations in real-time (Gonzalez & Rabbi, 2023). The theory posits that personalization enhances consumer satisfaction by reducing decision-making complexity and improving the overall shopping experience (Okorie et al., 2024). Several studies have applied this theory to assess how big data-driven personalization impacts consumer trust, purchase intentions, and brand loyalty (Yadav et al., 2024).

In addition to marketing-centric theories, Organizational Information Processing Theory (OIPT) has been employed to examine how firms manage and utilize big data for strategic decision-making (Sheng et al., 2021). OIPT suggests that organizations must develop robust information-processing capabilities to handle the volume, velocity, and variety of big data effectively (Kamel, 2023). This theory has been particularly relevant in studies analyzing the adoption of business intelligence frameworks and data analytics systems in retail, banking, and finance sectors (Ye & Jonilo, 2023). By applying OIPT, researchers have demonstrated how firms with strong data analytics capabilities can gain a competitive edge through enhanced market intelligence and consumer insights (Reddy & Nalla, 2024).

Another significant theoretical framework observed in the analyzed studies is Technology Acceptance Model (TAM), which explains how users adopt and utilize new technologies, including big data analytics platforms (Prasanthi et al., 2024). TAM posits that perceived usefulness and ease of use influence user acceptance of technological innovations, making it particularly relevant in studies examining how businesses and consumers interact with data-driven personalization systems (Adeniran et al., 2024). Several studies have applied TAM to explore factors that drive the adoption of AI-powered recommendation engines, personalized customer service tools, and automated marketing platforms (Swami, 2025).

Moreover, Knowledge-Based View (KBV) Theory has been employed to assess how organizations leverage big data as a strategic asset (Kamel, 2023). KBV posits that firms with superior data analytics capabilities can create knowledge-driven competitive advantages by transforming raw data into actionable insights (Sheng et al., 2021). This theory has been particularly useful in studies analyzing the role of BDA in market intelligence, customer segmentation, and demand forecasting (Ye & Jonilo, 2023). By applying KBV, researchers highlight the value of data-driven decision-making in shaping business strategies and enhancing consumer engagement (Gonzalez & Rabbi, 2023).

Another theoretical framework that has gained traction in big data research is Consumer Decision-Making Theory, which examines how consumers process information and make purchase decisions (Okorie et al., 2024). This theory is particularly relevant in studies analyzing sentiment analysis, predictive analytics, and personalized marketing campaigns (Ijomah et al., 2024). It posits that consumers are more likely to engage with brands that provide personalized recommendations and streamlined purchasing experiences (Adeniran et al., 2024). By integrating big data into consumer decision-making processes, firms can enhance customer satisfaction and drive higher conversion rates (Yadav et al., 2024).

Additionally, Data-Driven Marketing Theory has been widely referenced in studies examining how big data influences personalized marketing strategies (Bhargava et al., 2024). This theory emphasizes the role of consumer data in shaping targeted advertisements, optimizing content delivery, and improving customer segmentation (Handayati et al., 2024). It aligns with findings from sentiment analysis studies, which reveal that businesses can refine their engagement strategies by analyzing customer feedback, social media interactions, and purchasing behaviors (Ijomah et al., 2024). The growing reliance on machine learning algorithms and real-time analytics further underscores the relevance of this theory in modern marketing research (Reddy & Nalla, 2024).

Lastly, Mass Customization Theory has been applied in studies that explore how firms tailor products and services to individual customer preferences using big data (Sheng et al., 2021). This theory argues that companies can simultaneously achieve efficiency and personalization by leveraging advanced data analytics tools, such as recommendation algorithms and predictive modeling (Swami, 2025). Several studies have used this framework to examine how e-commerce platforms, financial institutions, and retail businesses implement mass customization strategies to enhance customer experiences and increase brand loyalty (Ye & Jonilo, 2023).

In summary, the theoretical frameworks employed in previous studies provide a comprehensive understanding of how Big Data Analytics and Management Information Systems facilitate consumer personalization. While CRM Theory and Personalization Theory emphasize customer engagement and satisfaction, organizational theories such as OIPT and KBV highlight the strategic advantages of big data in decision-making. Additionally, TAM and Consumer Decision-Making Theory shed light on how consumers interact with and respond to data-driven personalization efforts. Meanwhile, Data-Driven Marketing Theory and Mass Customization Theory provide insights into how businesses optimize marketing strategies and product offerings using big data.

Despite the breadth of theoretical perspectives, there is still room for further exploration, particularly in the integration of psychological and behavioral theories to assess consumer responses to personalized experiences. Future research could also benefit from interdisciplinary approaches that combine marketing, information systems, and behavioral science theories to provide a more holistic understanding of big data-driven personalization. Ultimately, the application

of these theories underscores the transformative impact of big data on consumer experiences, reinforcing its role as a fundamental driver of business innovation and competitive advantage.

3.3. Role of Big Data Analytics in Consumer Personalization

Big Data Analytics (BDA) plays a pivotal role in shaping consumer personalization by enabling businesses to collect, process, and analyze vast amounts of consumer data to deliver tailored experiences. The increasing reliance on BDA in various industries, including retail, banking, and finance, underscores its transformative impact on consumer engagement and satisfaction (Reddy & Nalla, 2024). By leveraging data-driven insights, companies can understand customer preferences, predict future behaviors, and optimize marketing strategies to enhance consumer loyalty (Gonzalez & Rabbi, 2023). One of the most significant contributions of BDA to consumer personalization is its ability to facilitate real-time data processing and decision-making. Businesses can now analyze consumer interactions in real-time, allowing them to adjust marketing campaigns, recommend personalized products, and optimize customer service interactions instantaneously (Okorie et al., 2024). This real-time capability is particularly evident in e-commerce and digital banking, where companies utilize machine learning algorithms to provide instant recommendations based on browsing history, purchase behavior, and demographic data (Yadav et al., 2024).

Moreover, predictive analytics is a critical component of BDA that enables businesses to anticipate customer needs before they are explicitly expressed (Reddy, 2021). By analyzing historical transaction data, social media interactions, and online reviews, companies can identify emerging trends and proactively offer personalized products or services (Ijomah et al., 2024). For instance, in the financial sector, predictive analytics helps banks and credit institutions assess credit risk, detect fraudulent activities, and offer customized financial products based on consumer spending patterns (Adhikari et al., 2024). Another crucial aspect of BDA in consumer personalization is its role in sentiment analysis, which involves analyzing customer feedback to gauge satisfaction levels and sentiment trends (Handayati et al., 2024). Sentiment analysis allows businesses to understand how consumers perceive their products or services, providing valuable insights that can be used to refine marketing strategies and improve customer experiences (Adeniran et al., 2024). Retailers, for instance, leverage sentiment analysis to adjust product offerings based on customer reviews, thereby enhancing customer satisfaction and brand reputation (Gonzalez & Rabbi, 2023).

In addition to enhancing marketing efforts, BDA significantly improves customer segmentation and targeted advertising. Traditional marketing strategies often rely on broad demographic data, whereas big data-driven personalization enables businesses to segment customers based on behavioral patterns, preferences, and purchasing habits (Swami, 2025). By using machine learning algorithms and clustering techniques, businesses can group consumers into highly specific segments and tailor advertising messages accordingly (Kamel, 2023). This data-driven approach increases marketing efficiency, reduces advertising costs, and enhances the overall effectiveness of marketing campaigns (Okorie et al., 2024). Furthermore, BDA plays an essential role in enhancing customer relationship management (CRM) by providing organizations with deep insights into customer behaviors and expectations (Anshari et al., 2019). Companies that integrate BDA into their CRM strategies can offer personalized loyalty programs, customized discounts, and exclusive offers to retain high-value customers (Ye & Jonilo, 2023). This level of personalization fosters a stronger emotional connection between brands and consumers, leading to increased customer retention and higher lifetime value (Kamel, 2023).

Beyond marketing and CRM, BDA also contributes to hyper-personalization in e-commerce and financial services. Hyper-personalization takes traditional personalization a step further by leveraging artificial intelligence (AI) and real-time data analytics to deliver ultra-targeted experiences (Rane et al., 2023). For example, streaming services like Netflix and Spotify use AI-driven BDA to curate personalized content recommendations, while fintech companies use real-time analytics to offer dynamic pricing and financial advisory services tailored to individual user needs (Sheng et al., 2021). Moreover, the integration of BDA with Internet of Things (IoT) technologies has further expanded the scope of consumer personalization (Prasanthi et al., 2024). IoT devices, such as smart home assistants and wearable technology, continuously generate consumer data that businesses can analyze to personalize user experiences (Handayati et al., 2024). For example, smart refrigerators can suggest grocery lists based on past purchases, while fitness trackers can provide tailored health recommendations based on user activity levels (Gonzalez & Rabbi, 2023).

Despite its numerous benefits, the use of BDA in consumer personalization also raises several challenges, particularly in terms of data privacy and ethical concerns (Adeniran et al., 2024). With increasing regulatory scrutiny, businesses must ensure compliance with data protection laws, such as the General Data Protection Regulation (GDPR) and the California Consumer Privacy Act (CCPA), to safeguard consumer information (Swami, 2025). Ethical considerations, including data transparency, consent management, and algorithmic bias, must be addressed to build consumer trust and ensure fair and responsible use of personal data (Reddy & Nalla, 2024). Moreover, while BDA enhances

personalization efforts, there is a fine line between personalization and intrusion. Excessive personalization can lead to consumer discomfort, especially when businesses appear to know too much about their customers (Ijomah et al., 2024). Studies have shown that overly aggressive personalization strategies, such as highly targeted ads that follow consumers across multiple platforms, can lead to privacy concerns and brand distrust (Okorie et al., 2024). Therefore, businesses must strike a balance between leveraging data-driven personalization and respecting consumer boundaries (Yadav et al., 2024).

In conclusion, Big Data Analytics has revolutionized consumer personalization by enabling businesses to analyze vast amounts of customer data in real time. From predictive analytics and sentiment analysis to targeted marketing and hyper-personalization, BDA plays a crucial role in enhancing consumer experiences across various industries. Furthermore, its integration with AI and IoT technologies has expanded personalization opportunities, allowing businesses to deliver highly customized interactions that meet individual consumer preferences. However, despite its advantages, the use of BDA in personalization must be approached with caution, particularly regarding data privacy and ethical considerations. Ensuring transparency, compliance with regulations, and maintaining a balance between personalization and consumer autonomy are essential to fostering long-term customer trust. As technology continues to advance, businesses must continue refining their data analytics strategies to optimize personalization while prioritizing ethical responsibility. Ultimately, BDA remains a powerful tool for businesses seeking to enhance consumer engagement, build brand loyalty, and drive long-term growth in the digital economy.

3.4. Role of Management Information Systems in Consumer Personalization

Management Information Systems (MIS) play a fundamental role in enabling consumer personalization by providing businesses with the necessary tools to collect, store, analyze, and utilize vast amounts of consumer data efficiently. The increasing digitization of business operations has made MIS indispensable for organizations seeking to enhance customer experiences and optimize marketing strategies (Kamel, 2023). By integrating advanced information systems, businesses can develop more effective customer relationship management (CRM) models, streamline data processing, and improve decision-making processes, all of which contribute to more personalized consumer interactions (Anshari et al., 2019). A key contribution of MIS to consumer personalization lies in its ability to enhance data management and integration. Businesses today generate data from multiple sources, including customer transactions, social media interactions, and online browsing behaviors (Reddy & Nalla, 2024). MIS enables companies to consolidate these data points into centralized systems, ensuring seamless access and real-time analysis (Ye & Jonilo, 2023). For instance, Enterprise Resource Planning (ERP) systems and CRM platforms allow businesses to create unified consumer profiles that track individual preferences, past purchases, and engagement history, facilitating more targeted marketing efforts (Gonzalez & Rabbi, 2023).

Furthermore, MIS plays a crucial role in improving customer segmentation and predictive analytics, which are essential for personalization (Kamel, 2023). Traditional segmentation methods often rely on demographic data, but MIS integrates artificial intelligence (AI) and machine learning to refine customer groupings based on behavioral patterns and purchasing histories (Swami, 2025). For example, advanced CRM systems leverage MIS capabilities to identify high-value customers and recommend personalized promotions based on their past interactions (Adeniran et al., 2024). This predictive capability not only enhances customer satisfaction but also increases sales and customer retention rates (Okorie et al., 2024). Another critical aspect of MIS in consumer personalization is its ability to automate and optimize real-time interactions. Many organizations use Decision Support Systems (DSS) and AI-driven chatbots to engage with customers and provide instant recommendations (Yadav et al., 2024). E-commerce platforms, for instance, utilize MIS to dynamically adjust product recommendations based on browsing patterns, ensuring that each consumer receives a tailored shopping experience (Ijomah et al., 2024). Additionally, financial institutions leverage MIS-powered AI models to customize banking services, offering personalized financial products and advisory services to individual customers (Reddy, 2021).

Beyond automation, MIS plays a fundamental role in enhancing omnichannel personalization strategies. Today's consumers engage with businesses across multiple touchpoints, including websites, mobile applications, physical stores, and social media platforms (Handayati et al., 2024). MIS ensures a consistent and seamless customer experience by integrating data from all these channels, allowing businesses to personalize interactions regardless of the platform used (Prasanthi et al., 2024). For instance, retailers with robust MIS infrastructure can track consumer behavior across different touchpoints and provide unified recommendations, whether the customer is shopping online or in-store (Okorie et al., 2024). Moreover, MIS contributes to improving data-driven decision-making, which is crucial for personalization strategies (Sheng et al., 2021). Businesses that invest in robust MIS infrastructures gain the ability to analyze customer trends, measure the effectiveness of marketing campaigns, and adjust personalization efforts accordingly (Gonzalez & Rabbi, 2023). For example, financial institutions use MIS to assess customer spending patterns

and offer customized credit card promotions, thereby increasing customer engagement and satisfaction (Swami, 2025). Similarly, retail companies analyze consumer demand and adjust inventory levels to ensure that personalized product recommendations align with availability (Ye & Jonilo, 2023).

While MIS enhances personalization efforts, it also introduces significant challenges, particularly in data security and privacy (Adeniran et al., 2024). As businesses collect vast amounts of consumer data, ensuring compliance with regulations such as the General Data Protection Regulation (GDPR) and the California Consumer Privacy Act (CCPA) becomes increasingly important (Reddy & Nalla, 2024). MIS plays a crucial role in enforcing data protection measures by implementing encryption, access controls, and consent management systems (Yadav et al., 2024). Organizations must balance the benefits of personalization with ethical considerations to maintain consumer trust and prevent data misuse (Ijomah et al., 2024). Additionally, businesses must be cautious about algorithmic biases and ethical concerns when using MIS for personalization (Handayati et al., 2024). AI-driven recommendation systems, while powerful, can sometimes reinforce biases in data, leading to unfair or discriminatory targeting of consumers (Okorie et al., 2024). For example, financial institutions relying on predictive analytics for loan approvals must ensure that their models do not disproportionately disadvantage specific demographic groups (Swami, 2025). Addressing such biases requires businesses to implement transparent and accountable MIS frameworks that prioritize fairness in data-driven decision-making (Sheng et al., 2021).

Another limitation of MIS in consumer personalization is system integration challenges. Many organizations operate legacy systems that are not easily compatible with modern AI-driven MIS platforms (Gonzalez & Rabbi, 2023). The complexity of integrating various data sources and ensuring interoperability across departments can hinder personalization efforts (Ye & Jonilo, 2023). Businesses must invest in scalable and flexible MIS architectures that facilitate seamless integration and data flow across different systems (Reddy, 2021). Despite these challenges, the future of MIS in consumer personalization looks promising, especially with the integration of cloud computing and Internet of Things (IoT) technologies (Kamel, 2023). Cloud-based MIS solutions provide businesses with scalable storage and processing power, allowing them to analyze consumer data more efficiently (Adeniran et al., 2024). Similarly, IoT-enabled devices generate real-time consumer data that can be integrated into MIS platforms to offer hyper-personalized experiences (Reddy & Nalla, 2024). For example, smart home assistants like Amazon Alexa and Google Home use MIS to analyze user preferences and deliver personalized content recommendations (Swami, 2025).

In conclusion, Management Information Systems play an essential role in enabling businesses to personalize consumer interactions through advanced data management, predictive analytics, and real-time decision-making. MIS enhances customer segmentation, optimizes omnichannel experiences, and improves data-driven marketing strategies, ultimately increasing customer satisfaction and brand loyalty. However, businesses must address challenges related to data privacy, ethical concerns, and system integration to maximize the benefits of MIS in personalization. As technology continues to evolve, organizations that invest in advanced MIS solutions, such as AI-driven CRM systems and cloud-based analytics platforms, will have a competitive edge in delivering hyper-personalized experiences. Furthermore, businesses must remain vigilant about regulatory compliance and ethical considerations to maintain consumer trust and ensure fair personalization practices. Ultimately, MIS will continue to be a crucial driver of consumer personalization, shaping the future of data-driven business strategies and customer engagement in the digital era.

4. Discussions of Findings

The findings of this study underscore the transformative role of Big Data Analytics (BDA) and Management Information Systems (MIS) in consumer personalization across various industries, including retail, banking, and finance. The results indicate that businesses are increasingly leveraging these technologies to enhance customer experiences, improve decision-making, and drive personalized engagement strategies (Kamel, 2023; Ye & Jonilo, 2023). More importantly, the integration of artificial intelligence (AI), predictive analytics, and cloud-based systems has significantly improved the ability of firms to analyze consumer behavior, anticipate needs, and offer tailored solutions in real time (Reddy & Nalla, 2024).

However, while the benefits of BDA and MIS in consumer personalization are evident, the findings also reveal several challenges, including data privacy concerns, algorithmic bias, and ethical implications in using consumer data (Adeniran et al., 2024; Handayati et al., 2024). Furthermore, the study highlights the growing role of regulatory frameworks, such as the General Data Protection Regulation (GDPR) and the California Consumer Privacy Act (CCPA), in shaping how businesses collect, process, and utilize consumer data (Swami, 2025). This section critically discusses these findings, emphasizing the impact, challenges, and future implications of BDA and MIS in consumer personalization.

The results confirm that Big Data Analytics is at the core of modern consumer personalization strategies, providing businesses with real-time insights into consumer preferences, behaviors, and purchasing patterns (Okorie et al., 2024). Through machine learning algorithms, predictive analytics, and sentiment analysis, companies can dynamically adapt their marketing strategies to better meet consumer expectations (Gonzalez & Rabbi, 2023). For instance, e-commerce platforms such as Amazon and Netflix leverage BDA-driven recommendation systems to provide personalized product and content suggestions, enhancing user engagement and increasing conversion rates (Yadav et al., 2024).

Moreover, the study highlights the importance of predictive analytics in consumer personalization, allowing businesses to anticipate future trends and behaviors (Reddy, 2021). By analyzing historical transaction data, browsing habits, and social media interactions, organizations can predict consumer needs even before they are explicitly expressed (Ijomah et al., 2024). This proactive approach has been particularly beneficial in banking and finance, where institutions use BDA to offer personalized financial products, detect fraudulent activities, and optimize risk assessment models (Adhikari et al., 2024). Another significant finding is the role of sentiment analysis in improving customer engagement (Handayati et al., 2024). By extracting insights from online reviews, social media conversations, and feedback forums, companies can refine their marketing strategies and address consumer concerns in real time (Adeniran et al., 2024). This is particularly relevant in the retail sector, where businesses analyze sentiment data to tailor product offerings and promotional campaigns based on customer feedback (Gonzalez & Rabbi, 2023).

However, despite these advantages, the study also identifies several challenges associated with BDA-driven personalization. Chief among them is the issue of data privacy and security, as consumers are increasingly concerned about how their personal data is collected and used (Swami, 2025). Regulatory frameworks such as GDPR and CCPA require businesses to ensure transparency, obtain consumer consent, and implement strict data protection measures (Reddy & Nalla, 2024). Companies that fail to comply with these regulations risk reputational damage and financial penalties, making ethical data management a critical aspect of BDA adoption (Kamel, 2023). Furthermore, the study reveals that algorithmic bias poses a significant challenge in BDA-driven personalization (Okorie et al., 2024). Machine learning models trained on biased datasets can lead to unfair and discriminatory outcomes, particularly in areas such as credit scoring, job recommendations, and targeted advertising (Sheng et al., 2021). Addressing this issue requires businesses to implement fair and transparent AI models, conduct regular bias audits, and prioritize ethical AI development (Ye & Jonilo, 2023).

Looking ahead, the future of BDA in consumer personalization will likely be shaped by advancements in AI-driven automation, blockchain for data security, and federated learning for decentralized data analysis (Prasanthi et al., 2024). Businesses that invest in these technologies will be better positioned to deliver hyper-personalized experiences while maintaining compliance with privacy regulations (Adeniran et al., 2024).

The findings also emphasize the critical role of Management Information Systems (MIS) in supporting data-driven personalization strategies. MIS provides businesses with the infrastructure needed to manage, store, and analyze consumer data efficiently, enabling them to develop more responsive and adaptive personalization models (Anshari et al., 2019). The integration of MIS with Customer Relationship Management (CRM) systems, Enterprise Resource Planning (ERP) platforms, and Decision Support Systems (DSS) has enhanced the ability of firms to offer seamless and personalized consumer experiences across multiple touchpoints (Kamel, 2023). A key finding is that MIS significantly improves customer segmentation and targeted marketing efforts (Swami, 2025). Unlike traditional marketing approaches that rely on broad demographic data, modern MIS platforms leverage AI and machine learning algorithms to segment consumers based on behavioral patterns, preferences, and engagement levels (Reddy, 2021). This capability allows businesses to deliver highly targeted advertisements, personalized promotions, and customized product recommendations (Okorie et al., 2024).

Additionally, the study highlights that MIS enhances omnichannel personalization by integrating data from various customer interaction points, such as websites, mobile apps, social media, and physical stores (Handayati et al., 2024). For instance, retailers with robust MIS infrastructures can track consumer behavior across different channels and provide personalized experiences regardless of the platform used (Adeniran et al., 2024). This seamless integration ensures a consistent and unified consumer journey, ultimately improving customer satisfaction and brand loyalty (Yadav et al., 2024). Moreover, the findings suggest that MIS plays a crucial role in real-time decision-making and automation (Sheng et al., 2021). Many organizations use AI-powered MIS solutions, such as intelligent chatbots, automated customer service platforms, and real-time analytics dashboards, to enhance personalization (Reddy & Nalla, 2024). These technologies allow businesses to respond instantly to customer inquiries, optimize inventory management, and provide dynamic pricing strategies based on market demand (Gonzalez & Rabbi, 2023).

Despite these benefits, the study identifies several challenges in MIS-driven personalization, including system integration issues, data silos, and security vulnerabilities (Ye & Jonilo, 2023). Many businesses still rely on legacy IT systems that are not easily compatible with modern AI-driven MIS platforms, creating inefficiencies in data processing and personalization efforts (Kamel, 2023). Addressing this challenge requires firms to invest in scalable, cloud-based MIS solutions that enable seamless data integration and interoperability (Swami, 2025). Additionally, cybersecurity remains a major concern in MIS adoption, as businesses must protect consumer data from cyber threats and unauthorized access (Adeniran et al., 2024). Implementing advanced security protocols, such as encryption, multi-factor authentication, and blockchain-based data storage, will be crucial in ensuring consumer trust and compliance with data protection laws (Reddy & Nalla, 2024). Looking forward, the future of MIS in consumer personalization will likely involve greater reliance on AI-driven automation, edge computing for real-time data processing, and enhanced regulatory compliance measures (Okorie et al., 2024). Companies that successfully integrate these innovations into their MIS infrastructure will be well-positioned to deliver superior personalized experiences while maintaining ethical and secure data management practices (Gonzalez & Rabbi, 2023).

Overall, the findings demonstrate that both Big Data Analytics and Management Information Systems are essential in advancing consumer personalization strategies. While BDA enables real-time insights, predictive analytics, and sentiment analysis, MIS provides the technological backbone for data management, segmentation, and omnichannel integration. However, businesses must navigate key challenges such as data privacy, algorithmic bias, cybersecurity risks, and system integration issues to fully realize the potential of these technologies. Moving forward, companies that invest in ethical AI, robust security frameworks, and next-generation MIS solutions will gain a competitive edge in delivering personalized, data-driven consumer experiences in an increasingly digital economy.

5. Conclusion

The findings of this study reinforce the transformative impact of Big Data Analytics (BDA) and Management Information Systems (MIS) on consumer personalization across various industries, including retail, banking, and finance. The integration of BDA has enabled businesses to leverage machine learning, predictive analytics, and sentiment analysis to anticipate consumer needs, optimize engagement strategies, and offer tailored experiences in real-time (Reddy & Nalla, 2024; Okorie et al., 2024). Meanwhile, MIS has provided the essential infrastructure to collect, process, and integrate vast amounts of customer data, ensuring that businesses can make data-driven decisions and enhance customer relationship management (Kamel, 2023; Ye & Jonilo, 2023). One of the key takeaways from the study is that real-time data processing is a game-changer for consumer personalization. Companies that effectively utilize BDA can dynamically adjust their marketing campaigns, recommend products based on real-time behavior, and optimize customer service interactions instantaneously (Yadav et al., 2024). This level of agility has proven particularly beneficial in e-commerce, digital banking, and fintech, where hyper-personalization has become a competitive necessity (Handayati et al., 2024).

Additionally, the study highlights the importance of predictive analytics in personalization. Businesses that employ predictive models can not only tailor their marketing efforts but also forecast future trends, identify consumer preferences, and optimize supply chain operations based on data-driven insights (Swami, 2025). However, while BDA has revolutionized personalization, its effectiveness depends on the robustness of MIS infrastructure, which serves as the foundation for collecting, analyzing, and utilizing consumer data (Kamel, 2023). Despite these advantages, the study also identifies several challenges that businesses must address, particularly regarding data privacy, ethical considerations, and algorithmic bias (Adeniran et al., 2024). The increasing reliance on consumer data raises concerns about how businesses collect, store, and use personal information, necessitating compliance with regulations such as the General Data Protection Regulation (GDPR) and the California Consumer Privacy Act (CCPA) (Reddy & Nalla, 2024). Furthermore, bias in AI-driven personalization models poses risks of unfair or discriminatory outcomes, highlighting the need for ethical AI governance and transparency in data-driven decision-making (Sheng et al., 2021). Looking ahead, the future of BDA and MIS in consumer personalization will likely involve greater reliance on AI automation, federated learning for decentralized data processing, and blockchain for secure data management (Gonzalez & Rabbi, 2023). Businesses that successfully navigate data privacy concerns, invest in ethical AI frameworks, and integrate advanced MIS solutions will gain a significant competitive advantage in delivering personalized, data-driven consumer experiences (Okorie et al., 2024).

Recommendations

In order to fully capitalize on the potential of Big Data Analytics (BDA) and Management Information Systems (MIS) in consumer personalization, businesses must adopt a strategic and ethical approach to data management and decision-making. As the findings of this study reveal, the integration of advanced AI-driven analytics, predictive modeling, and

real-time data processing has significantly enhanced the ability of organizations to anticipate consumer needs and tailor personalized experiences (Kamel, 2023; Reddy & Nalla, 2024). However, despite these advantages, several challenges—ranging from data privacy concerns to algorithmic bias and system integration issues—must be addressed to ensure that BDA and MIS-driven personalization strategies remain effective, ethical, and sustainable. One of the most critical recommendations is for businesses to prioritize data privacy and regulatory compliance. As concerns surrounding consumer data protection and surveillance continue to grow, organizations must ensure strict adherence to data privacy laws such as the General Data Protection Regulation (GDPR) and the California Consumer Privacy Act (CCPA) (Swami, 2025). Companies should invest in robust data encryption techniques, secure storage solutions, and transparent data governance frameworks to protect consumer information from potential breaches and misuse (Adeniran et al., 2024). Additionally, businesses should implement user-friendly consent management systems that provide customers with greater control over their personal data, ensuring that personalization efforts remain ethical and legally compliant (Gonzalez & Rabbi, 2023).

In addition to data privacy, businesses must also mitigate algorithmic bias and promote fairness in AI-driven personalization. The study highlights that BDA models trained on biased datasets can result in discriminatory outcomes, particularly in areas such as loan approvals, job recommendations, and targeted advertising (Okorie et al., 2024). To address this issue, organizations should adopt bias detection tools, conduct regular AI audits, and incorporate fairness-aware machine learning techniques (Sheng et al., 2021). Furthermore, businesses should strive for greater transparency in AI decision-making, ensuring that consumers understand how their data is used to personalize experiences and that personalization models do not unintentionally reinforce societal inequalities (Reddy & Nalla, 2024). Another key recommendation is for organizations to invest in scalable MIS infrastructure to facilitate seamless data integration and real-time processing. Many companies still operate with fragmented legacy systems that create data silos, limiting their ability to derive actionable insights from customer data (Kamel, 2023). To overcome these challenges, businesses should transition to cloud-based MIS platforms and AI-driven CRM solutions that enable cross-departmental collaboration and streamlined consumer data analysis (Ye & Jonilo, 2023). Additionally, integrating edge computing technologies can enhance real-time data processing capabilities, allowing businesses to deliver instantaneous personalized experiences across multiple digital touchpoints (Handayati et al., 2024).

Moreover, predictive analytics should be leveraged to enhance proactive personalization. The findings indicate that predictive modeling can significantly improve customer engagement and retention by enabling businesses to anticipate consumer preferences, buying behaviors, and service needs in advance (Yadav et al., 2024). Companies should adopt advanced AI-driven predictive analytics tools to develop more precise consumer segmentation models, refine personalized marketing strategies, and optimize inventory management based on real-time demand forecasting (Adeniran et al., 2024). By harnessing predictive analytics effectively, businesses can move beyond reactive personalization and adopt a more proactive, consumer-centric approach (Gonzalez & Rabbi, 2023). Furthermore, strengthening cybersecurity measures is essential for ensuring the safety and reliability of BDA and MIS-driven personalization efforts. With the rising incidence of cyber threats and data breaches, businesses must prioritize the implementation of multi-factor authentication, blockchain-based data security, and AI-driven fraud detection systems (Swami, 2025). Secure data-sharing frameworks should also be established to protect consumer identities and prevent unauthorized access to sensitive information (Reddy & Nalla, 2024). By investing in stronger cybersecurity protocols, organizations can enhance consumer trust and confidence in personalized services, ultimately improving brand reputation and long-term customer relationships (Sheng et al., 2021).

Additionally, businesses should embrace transparency and consumer control over data usage. The study highlights that consumers are more likely to engage with personalized experiences when they have clear visibility into how their data is being used (Okorie et al., 2024). To address this, companies should implement interactive data dashboards and AI explainability tools that allow customers to review, adjust, or opt out of personalization features at any time (Gonzalez & Rabbi, 2023). Providing consumers with greater autonomy over their personal data not only enhances trust but also aligns with evolving regulatory expectations and ethical standards (Adeniran et al., 2024). Looking ahead, businesses should explore emerging technologies such as the Internet of Things (IoT) and edge computing to enhance real-time personalization. The integration of smart devices and connected systems enables businesses to gather real-time consumer insights and deliver context-aware recommendations (Swami, 2025). For example, wearable fitness devices can provide users with personalized health and wellness recommendations, while smart home assistants can optimize content delivery based on user preferences (Sheng et al., 2021). By leveraging IoT-powered personalization, companies can further enhance consumer engagement and satisfaction in an increasingly digital landscape (Reddy & Nalla, 2024).

Finally, cross-industry collaboration and ethical data-sharing partnerships should be encouraged. The study suggests that companies across retail, financial services, healthcare, and entertainment sectors can benefit from secure data-sharing frameworks that enhance personalization efforts while preserving consumer privacy (Adeniran et al., 2024).

Federated learning models, for instance, allow businesses to train AI models on decentralized datasets without exposing sensitive consumer information, ensuring both data security and effective personalization (Gonzalez & Rabbi, 2023). By fostering responsible data-sharing ecosystems, businesses can maximize the value of BDA while upholding ethical and regulatory standards (Okorie et al., 2024). In summary, to successfully implement BDA and MIS-driven personalization strategies, businesses must adopt a holistic approach that balances innovation, regulatory compliance, and consumer trust. By investing in privacy-focused AI, transparent personalization frameworks, and advanced predictive analytics, companies can enhance customer experiences while mitigating the risks associated with data-driven personalization (Swami, 2025). Furthermore, as technology continues to evolve, organizations must remain agile, adaptive, and proactive in refining their personalization strategies to align with changing consumer expectations and industry regulations (Reddy & Nalla, 2024). Ultimately, the future of consumer personalization will depend on businesses' ability to merge technological advancements with ethical responsibility, fostering a data-driven ecosystem that benefits both consumers and enterprises alike (Sheng et al., 2021).

Limitations of the Study

While this study provides valuable insights into the role of Big Data Analytics (BDA) and Management Information Systems (MIS) in consumer personalization, several limitations must be acknowledged. First, the scope of the study was primarily limited to the retail, banking, and finance sectors, which may restrict the generalizability of the findings to other industries such as healthcare, education, and entertainment (Okorie et al., 2024). Future research should explore how BDA and MIS influence personalization in a broader range of industries to enhance the applicability of the results (Swami, 2025). Second, the study relied heavily on secondary data and literature reviews, which, while valuable, may not fully capture real-time industry trends and emerging technologies (Reddy & Nalla, 2024). Conducting empirical

research, primary data collection, and case studies would provide a more comprehensive and practical perspective on how businesses are implementing BDA and MIS-driven personalization strategies (Handayati et al., 2024).

Additionally, while the study discussed data privacy and ethical concerns, it did not fully explore the regional variations in data protection laws and regulations, which could influence how businesses approach personalization across different markets (Adeniran et al., 2024). Future research should examine how organizations navigate global regulatory challenges and balance data-driven personalization with ethical considerations (Gonzalez & Rabbi, 2023). Finally, the study focused primarily on business applications of personalization but did not extensively examine consumer perspectives, acceptance, and potential resistance to personalized experiences (Yadav et al., 2024). Future studies should incorporate consumer surveys, behavioral experiments, and qualitative research to better understand user attitudes toward data-driven personalization (Sheng et al., 2021). Despite these limitations, this study offers a solid foundation for future research and industry applications of BDA and MIS in consumer personalization (Kamel, 2023).

Compliance with ethical standards

Disclosure of conflict of interest

No conflict of interest to be disclosed.

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