Impact of AI Chatbot-Enhanced Customer Satisfaction on Customer Loyalty: The Mediating Role of Customer Trust

Hui Liang¹ Sutthikarn Khong-khai^{2*} Winitra Leelapattana³ and Chin-Fa Tsai⁴

Received: 11/04/2024, Revised: 07/06/2024, 20/06/2024, Accepted: 20/06/2024

Abstract

This study examines the impact of AI chatbot-enhanced customer satisfaction on trust and loyalty, with a particular focus on the mediating role of customer trust. A multistage sampling technique was employed to select a diverse group of 436 participants from the target population of Chinese online retail customers with experience using AI chatbots. Analysis using Structural Equation Modeling in SPSS and AMOS revealed a significant positive relationship between customer satisfaction derived from AI chatbot interactions and both trust and loyalty. Moreover, customer trust was found to significantly mediate the relationship between satisfaction and loyalty. These findings underscore the critical role of AI chatbots in not only improving customer satisfaction but also in fostering trust and loyalty within the online retail sector. The study provides valuable insights into how digital customer service tools like AI chatbots can be strategically leveraged to enhance customer relationships and loyalty, offering important implications for online retailers looking to optimize their engagement strategies in the digital marketplace.

Keywords: AI Chatbot; Customer Satisfaction; Customer Loyalty; Customer Trust

¹ International College, Maejo University, Chiang Mai, Thailand

² International College, Maejo University, Chiang Mai, Thailand

³ International College, Maejo University, Chiang Mai, Thailand

⁴ International College, Maejo University, Chiang Mai, Thailand

^{*} Corresponding Author E-mail: mjuonn@gmail.com

Introduction

In the swiftly evolving digital landscape, the integration of Artificial Intelligence (AI) chatbots into customer service strategies has gained significant traction (Chatterjee et al., 2019). This integration has reshaped conventional customer service paradigms by facilitating seamless interactions between AI chatbots and customers, providing immediate support and personalized assistance (Mero, 2018). As businesses strive for enduring customer relationships, understanding how AI chatbot-enhanced customer satisfaction influences consumer loyalty has emerged as a critical focus area (Ho, 2021).

Central to this dynamic is customer trust, serving as a pivotal mediator between satisfaction and loyalty. Trust, bolstered by the reliability and integrity of AI chatbots, profoundly impacts their efficacy in enhancing the customer experience (Følstad et al., 2018). It is the foundation of customer loyalty initiatives, with effective AI chatbot use—marked by accurate information, consistent service, and empathetic interaction—raising customer satisfaction and, subsequently, loyalty (McGrath et al., 2024; Babatunde et al., 2024).

Despite the recognized importance of AI chatbots in customer service, a discernible research gap exists in comprehensively understanding the mediating role of trust between AI chatbotenhanced customer satisfaction and loyalty. While the impact of AI chatbots on satisfaction and the general relationship between satisfaction and loyalty have been explored, less is known about how trust, specifically fostered through AI chatbot interactions, shapes the transition from satisfaction to loyalty. This study aims to bridge this gap by delving into the nuanced relationship between these elements, providing insights for businesses to optimize AI chatbot strategies in building trust and fostering loyalty (Chen et al., 2023).

The practical implications of addressing this gap are multifaceted. In an era of digital innovation, businesses continuously seek ways to leverage technology to bolster customer satisfaction and loyalty. A deeper understanding of the mediating effect of trust can offer valuable insights for tailoring AI chatbot services to meet customer expectations and preferences, enhancing overall satisfaction and loyalty (Liang et al., 2018; Bahri-Ammari & Bilgihan, 2019). Moreover, as remote interactions become increasingly common, establishing trust through AI chatbot is paramount. This research aims to shed light on the challenges and opportunities presented by AI chatbot implementation, helping businesses refine their strategies to enhance trust, satisfaction, and loyalty among their customer base (Viberg et al., 2023).

Guided by the objectives to explore the relationships between AI chatbot-enhanced customer satisfaction, trust, and loyalty, and to understand the mediating role of trust, this research seeks to provide actionable insights. It aims to empower businesses to leverage AI chatbots more effectively, fostering a deeper sense of trust and loyalty among customers, thereby enhancing their competitive edge in the digital marketplace.

Literature Review and Hypotheses Developments

Relationship Between Customer Satisfaction and Customer Loyalty

The relationship between customer satisfaction and loyalty remains a central focus in marketing, as highlighted by Sani et al. (2024) in their study on the impact of service quality on customer loyalty through satisfaction. Customer satisfaction, defined by the degree to which consumer expectations are met or surpassed by a product or service, is often hailed as the bedrock upon which customer loyalty is built (Lu et al., 2020). Customer loyalty, demonstrated by repeat patronage and the willingness to recommend a service, reflects customer satisfaction and is a critical factor in an organization's financial success, as discussed by Khan (2024).

The transition from satisfaction to loyalty is a dynamic process that encompasses positive customer attitudes and a deep-seated commitment to the brand, thus underscoring the importance of satisfaction in customer retention strategies (Kotler et al., 2014). Recent studies highlight customer satisfaction as a crucial driver of customer loyalty. Yum and Yoo (2023) showed that in mobile social media, customer satisfaction significantly impacts loyalty. Similarly, Huang et al. (2020) demonstrated that enhanced customer satisfaction through improved service quality directly boosts customer loyalty.

The advent of AI chatbots in customer service introduces a new dimension to this relationship. AI chatbots can offer a seamless, personalized experience, significantly enhancing customer satisfaction by providing instant responses and tailored recommendations. Studies have shown that the use of AI chatbots in online retail can lead to higher customer satisfaction due to their efficiency and ability to handle multiple inquiries simultaneously (Følstad & Brandtzæg, 2017). Consequently, this enhanced satisfaction is likely to foster greater customer loyalty.

H1: Customer satisfaction through the use of AI chatbots has a positive impact on customer loyalty.

Relationship Between Customer Satisfaction and Customer Trust

The integral role of customer trust in the nexus between satisfaction and loyalty is increasingly recognized within the realms of marketing and consumer behavior (Vijay et al., 2019). Recent studies emphasize that customer satisfaction is crucial for building trust in retail, especially with high-quality service (Yum & Yoo, 2023). Recent studies reinforce the significant correlation between customer satisfaction and trust, showing that high levels of satisfaction, particularly through AI chatbot interactions, are instrumental in engendering trust among consumers (Huang et al., 2024).

Trust, in this context, not only fosters loyalty but also acts as a buffer, mitigating the impact of negative experiences and ensuring that customer satisfaction remains intact even amidst service failures or challenges (Shpëtim, 2012). The transition from mere satisfaction to loyalty is significantly influenced by the degree of trust a customer places in the brand, emphasizing the importance of trust in transforming satisfied customers into steadfast brand advocates (Dam & Dam, 2021).

In the digital age, where AI chatbots are becoming ubiquitous in customer service, understanding the dynamics of trust in the context of AI interactions becomes paramount. AI chatbots can enhance trust by consistently providing accurate information, respecting user privacy, and learning from interactions to improve future customer experiences. Research indicates that customers tend to trust AI chatbots that exhibit human-like conversational abilities and ethical behavior (Araujo, 2018).

H2: Customer satisfaction through the use of AI chatbots has a positive impact on customer trust.

Relationship Among Customer Satisfaction, Customer Trust, and Customer Loyalty

The multifaceted relationship between customer satisfaction, trust, and loyalty underscores the nuanced role of trust, serving both as a critical outcome of satisfaction and as an essential antecedent to loyalty (Hride et al., 2020). The pathway to loyalty involves more than just satisfaction; it is a synergetic effect of both satisfaction and the trust that customers place in a brand or service (Hart & Johnson, 1999). The literature consistently demonstrates that customer trust exerts a profound, positive influence on loyalty, affirming its significance in customer relationship dynamics (Kantsperger & Kunz, 2010; Chung & Shin, 2010).

In the context of AI chatbot interactions, customers' perceptions of trust extend beyond the technology's functionality to encompass the reliability and consistency of the service experience (Hassanein & Head, 2007). This trust is pivotal, acting as a bridge that connects

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satisfaction with loyalty, and plays a mediating role in this relationship, as evidenced by recent research (Troshani et al., 2021). Understanding how trust in AI chatbots influences loyalty is crucial, especially as these technologies become more integrated into customer service frameworks.

H3: Customer trust through the use of AI chatbots has a positive impact on customer loyalty.

H4: Customer satisfaction influences customer loyalty through the mediating of customer trust.



The study's conceptual framework is visually represented in Figure 1.

Figure 1 Conceptual Framework

Research Methodology

To examine the relationship among customer satisfaction, loyalty, and trust by utilizing AI Chatbot, as well as the mediating role of customer trust between satisfaction and loyalty, this study employed a quantitative research approach through closed-ended questionnaire surveys. The study targeted customers who have shopped at online retail stores in China, totaling 845 million individuals. A multistage sampling technique was utilized to select participants from the target population of Chinese online retail customers who have used AI chatbots. A multistage sampling technique was utilized to select participants from the target population of Chinese online retail customers from the target population of Chinese used AI chatbots. The steps involved in the multistage sampling technique were as follows:

1. Selection of Major Cities: Four major cities in China-Beijing, Shanghai, Guangzhou, and Shenzhen-were selected as primary sampling units due to their significant economic activity and diverse consumer behaviors.

2. Allocation of Sample Size: A minimum of 100 participants were targeted from each city to ensure balanced representation and robust data.

3. Participant Selection within Cities: Within each city, a combination of convenience and systematic sampling methods was employed. Convenience sampling facilitated accessibility to participants, while systematic sampling ensured randomness and representativeness within the selected sample.

Research Instrument and Indicators

The study adapted and modified items from existing related research to fit the current study's context. Customer satisfaction was measured using indicators of competence, and warmth (Chen et al., 2021; Fang et al., 2014). Loyalty was assessed through indicators of recommendation intention, and repurchase intention (Zaato et al., 2023; Anggraeni & Saidani, 2022). Lastly, customer trust was gauged using indicators of innovativeness, and optimism (Coulter & Coulter, 2003; Soderlund & Julander, 2003). A 5-point Likert scale was employed, with 1 indicating "strongly disagree" and 5 indicating "strongly agree".

Validity and Reliability

Before commencing data collection, the questionnaire underwent testing by five experts to ensure the validity of item-objective consistency (IOC). Experts rated each item's relevance on a 1-5 scale, and their ratings were averaged, resulting in an IOC of 0.98. In a pilot study involving a sample of 30 participants, the reliability of the items was examined, resulting in a Cronbach's alpha of 0.98, indicating satisfactory quality of the research instrument. A total of 436 valid questionnaires were collected.

Data Analysis

Data analysis was conducted using Analysis of Moment Structures (AMOS) and the Statistical Package for the Social Sciences (SPSS). Descriptive statistics were utilized, encompassing frequency, percentage, mean, standard deviation, skewness, and kurtosis. Additionally, structural equation modeling (SEM) and confirmatory factor analysis (CFA) were employed, considering good-fit indices as well as convergent and discriminant validity indicated by factor loading, composite reliability (CR), average variance extracted (AVE), correlation matrix, and square root of AVE. The anticipated model was expected to yield satisfactory good-fit indices, though adjustments were permissible based on modification indices if the model was deemed unfit (Knekta et al., 2019).

Results

Respondents' Profiles and Studied Variables

The sample exhibited broad urban representation across four major Chinese cities: Beijing (24.5%), Shanghai (24.3%), Guangzhou (24.3%), and Shenzhen (26.8%). Predominantly youthful, most respondents fell within the 21-30 (27.5%) and 31-40 (30.5%) age brackets, reflecting China's tech-savvy younger consumer base. Gender distribution was nearly balanced, with a slight female majority (51.8%). Educationally, a significant portion held undergraduate degrees (34.4%) or junior college diplomas (28.2%). Professionally, full-time employees (30.7%) and students (20.6%) constituted the largest groups. Shopping preferences leaned towards both online and offline options (45.6%), with a notable daily online shopping frequency (33.3%). Moderate familiarity with AI chatbots was observed among a considerable segment (37.4%), with weekly interaction times primarily under 1 hour (40.4%). Income levels varied, with 27.3% earning between 6001-8000 RMB, reflecting a middle-income demographic. The majority resided in urban areas (47.2%), mirroring the urban-centric deployment of AI chatbots in retail.

Confirmatory Factor Analysis (CFA) and Construct Validity Assessment

The study begins with Confirmatory Factor Analysis (CFA) and Construct Validity Assessment, using Composite Reliability (CR) as the primary metric. This process ensures the accuracy of each construct's representation and evaluates the fit of the data to the proposed model. It establishes a foundation for the theoretical framework and operationalization of constructs within the context of AI chatbot-enhanced customer satisfaction and loyalty in the Chinese retail sector.



Figure 2 Overall Confirmatory Factor Analysis

To ensure clarity in the standards for model-empirical data harmonization, it is essential to highlight the key fit indices used to validate the model's adequacy, as shown in Table 4. These indices collectively provide robust evidence of the model's structural validity and its suitability for examining the proposed relationships within the customer satisfaction domain. The Chi-square to degrees of freedom ratio (CMIN/DF) of 3.356 suggests a moderate fit, falling within the acceptable range of 1 to 5. The Goodness-of-Fit Index (GFI) achieved a value of 0.985, indicating an excellent fit and surpassing the threshold of 0.95. The Adjusted Goodness-of-Fit Index (AGFI) value of 0.946 also shows a good fit, exceeding the 0.90 benchmark. The Root Mean Square Error of Approximation (RMSEA) value of 0.074 suggests a moderate fit, being below the 0.08 threshold. The Root Mean Square Residual (RMR) value of 0.011 confirms a good fit, as it is well below the 0.05 cut-off. Furthermore, the Comparative Fit Index (CFI), Tucker-Lewis Index (TLI), and Normed Fit Index (NFI) values of 0.993, 0.983, and 0.991 respectively, all surpass the 0.95 criterion for a good fit.

The displayed confirmatory factor analysis (CFA) model suggests a positive relationship between customer satisfaction and customer trust, as well as between customer satisfaction and loyalty, with customer trust also positively influencing loyalty. High factor loadings for the observed variables on their respective latent constructs indicate a reliable measurement model. The structural paths imply that trust may play a mediating role between satisfaction and loyalty, although further analysis would be required to confirm this mediation effect.

Variables	CR	AVE	MSV	MaxR(H)
Customer Satisfaction	0.941	0.888	0.135	0.942
Customer Trust	0.941	0.889	0.131	0.942
Customer Loyalty	0.935	0.877	0.135	0.949

Table 1 Reliability s & Convergent Validity

The measurement model's constructs-Customer Satisfaction, Customer Trust, and Customer Loyalty-display exceptionally robust psychometric properties, well-aligned with the stringent criteria for confirmatory factor analysis within the structural equation modeling framework. The Composite Reliability (CR) scores are outstanding, with Customer Satisfaction and Customer Trust

each at 0.941 and Customer Loyalty at 0.935, which is significantly higher than the recommended benchmark of 0.7, denoting high internal consistency and reliability of the constructs (Hair et al., 2010). Convergent validity is strongly established, as evidenced by the Average Variance Extracted (AVE) values (Customer Satisfaction at 0.888, Customer Trust at 0.889, and Customer Loyalty at 0.877), all considerably exceeding the threshold of 0.5 (Fornell & Larcker, 1981). This indicates that the majority of the variance in the observed variables is attributable to their respective latent variables, suggesting that the constructs are well-specified.

Moreover, the Maximum Shared Variance (MSV) figures fall below the AVE values for their respective constructs, confirming that each construct is distinctly captured and thus supports the model's discriminant validity (Henseler et al., 2014). The closeness of the Maximum Reliability (MaxR(H)) to the CR scores corroborates the high reliability of the constructs, suggesting that the measurement model is highly stable (Hu & Bentler, 1999).

	Customer Satisfaction	Customer Trust	Customer Loyalty
Customer Satisfaction	0.942		
Customer Trust	0.305***	0.943	
Customer Loyalty	0.368***	0.362***	0.937

Table 2 Pearson Correlation and AVE Square Root

Note: The diagonal bold number is the square root of AVE; ***p<0.001

For assessing discriminant validity, the foundational criteria postulate that the square root of the Average Variance Extracted (AVE) for each construct should surpass the correlation coefficients among the constructs (Henseler et al., 2014). This principle underpins the establishment of distinctiveness among the constructs within the research model. The table 2 is a part of a statistical analysis to examine the relationships between three constructs: Customer Satisfaction, Customer Trust, and Customer Loyalty. Diagonal elements, presumably representing the square root of the Average Variance Extracted (AVE), confirm the discriminant validity of the constructs as they are higher than the inter-construct correlations, aligning with the criteria set forth by Fornell and Larcker (1981).

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Customer Satisfaction shows a moderate and positive correlation with Customer Trust (r = 0.305, p < 0.001), indicating that as customer satisfaction increases, there is a corresponding increase in customer trust. The correlation between Customer Satisfaction and Customer Loyalty is also positive and moderate (r = 0.368, p < 0.001), suggesting that higher customer satisfaction is associated with increased customer loyalty. Furthermore, Customer Trust and Customer Loyalty are positively correlated (r = 0.362, p < 0.001), which implies that trust is a potential precursor to loyalty. The significance level (p < 0.001) marked by asterisks indicates a high level of statistical confidence in these relationships.

Structural Equation Model

In the assessment of normality, reference values for skewness and kurtosis play a pivotal role. According to Kline (2023), absolute skewness values below 2.0 and kurtosis values below 7.0 are deemed moderate thresholds for normality, applicable in evaluating multivariate normality within the contexts of factor analysis and MANOVA.

	Skewness		Kurtosis		Distribution
	Statistic	Std. Error	Statistic	Std. Error	
Warmth	0.342	0.117	-1.041	0.233	Normal
Competent	0.322	0.117	-1.085	0.233	Normal
Innovativeness	0.409	0.117	-1.240	0.233	Normal
Optimism	0.326	0.117	-1.109	0.233	Normal
Recommendation Intention	0.397	0.117	-1.218	0.233	Normal
Repurchase Intention	0.236	0.117	-1.246	0.233	Normal

Table 3 Normality Test

As can be inferred from Table 3, the skewness and kurtosis statistics for the variables Warmth, Competent, Innovativeness, Optimism, Recommendation Intention, and Repurchase Intention indicate slight asymmetry and flatter peak distributions, respectively. However, all values fall within the acceptable range for normality (skewness within ± 2 and kurtosis within ± 2), suggesting that the distribution of these variables does not substantially deviate from a normal distribution (Hair et al., 2012), which validates the use of parametric tests in the analysis.



Figure 3 Final Model SEM Results

The structural equation model (SEM) framework developed for this study is designed to explore the relationships among the various constructs defined in the research model. This framework serves as a visual representation of the hypothesized relationships and provides a foundation for the SEM analysis. In figure 3 describes the framework diagram from this study, illustrating the constructs and the hypothesized paths.

Index	Value	Criterion	Result
CMIN/DF	3.356	<3.00, good fit; 3.0 - 4.0 moderate fit	Moderate fit
GFI	0.985	>0.95 good fit; 0.9 – 0.95 moderate fit	Good fit
AGFI	0.946	>0.9 good fit; 0.85 – 0.9 moderate fit	Good fit
RMSEA	0.074	<0.05, good fit; 0.05-0.08 moderate fit	Moderate fit
RMR	0.011	<0.05, good fit; 0.05-0.08 moderate fit	Good fit
CFI	0.993	>0.95 good fit; 0.9 – 0.95 moderate fit	Good fit
TLI	0.983	>0.95 good fit; 0.9 – 0.95 moderate fit	Good fit
NFI	0.991	>0.95 good fit; 0.9 – 0.95 moderate fit	Good fit

Table 4 Summary of Model Fitting Outcomes

The confirmatory factor analysis (CFA) for the model assessing relationships among Customer Satisfaction, Customer Trust, and Customer Loyalty demonstrates a mix of good to moderate fit based on various fit indices. The Chi-square to degrees of freedom ratio (CMIN/DF) of 3.356 falls within the moderate fit range, slightly exceeding the preferred criterion of less than

3.00, which suggests a reasonably good fit to the data. The Goodness-of-Fit Index (GFI) at 0.985 and the Adjusted Goodness-of-Fit Index (AGFI) at 0.946 both exceed the threshold of 0.9, indicating a good fit. The Root Mean Square Error of Approximation (RMSEA) value is 0.074, which is within the moderate fit range, suggesting that the model could be improved. The Root Mean Square Residual (RMR) at 0.011 is well below the 0.05 cut-off, demonstrating a good fit. Relative fit indices, including the Comparative Fit Index (CFI) at 0.993, the Tucker-Lewis Index (TLI) at 0.983, and the Normed Fit Index (NFI) at 0.991, all surpass the 0.95 benchmark for a good fit, reinforcing the model's overall adequacy. These indices collectively provide confidence in the model's structural validity for examining the proposed relationships within the customer satisfaction domain.

Hypothesis Investigation

No.	Hypothesis	Standardized Regression	Ρ
H1	Customer satisfaction through the use of AI chatbots has a positive impact on customer loyalty.	0.284	***
H2	Customer satisfaction through the use of AI chatbots has a positive impact on customer trust.	0.305	***
H3	Customer trust through the use of AI chatbots has a positive impact on customer loyalty.	0.275	***
H4	Customer satisfaction influences customer loyalty through the mediating of customer trust.	0.084	***

 Table 5 Regression Weight

Note: *** p<0.001

The investigation of hypotheses within this research aimed to elucidate the impacts of AI chatbot-enhanced satisfaction on customer loyalty in the retail business sector in China. A series of hypotheses were formulated and tested through standardized regression analyses to understand the interrelations between customer satisfaction, trust, loyalty as influenced by the use of AI chatbots (Table 5).

Hypothesis H1, proposing that customer satisfaction through the use of AI chatbots positively impacts customer loyalty, is supported with a standardized regression weight of 0.284 (p < 0.001). This suggests a moderate positive effect of satisfaction on loyalty within the AI chatbot context. Hypothesis H2, which posits a positive effect of customer satisfaction via AI chatbots on customer trust, is also supported, evidenced by a standardized regression weight of 0.305 (p < 0.001). This indicates that as customers' satisfaction with AI chatbots increases, their trust in the service or brand similarly increases. Further, Hypothesis H3 is corroborated, with a standardized regression weight of 0.275 (p < 0.001), denoting a positive impact of customer trust, developed through AI chatbot interactions, on customer loyalty. This finding underlines the role of trust as a significant determinant of loyalty in the realm of AI-enabled customer service. Lastly, Hypothesis H4, examining the mediating role of customer trust between satisfaction and loyalty, is substantiated by a statistically significant indirect effect of 0.084 (p < 0.001). This mediation effect, though smaller in magnitude compared to the direct effects, is nontrivial and indicates that customer trust partially mediates the relationship between satisfaction and loyalty. The presence of this indirect effect signifies that the impact of customer satisfaction on loyalty is not only direct but also channeled through the development of customer trust.

Conclusion and Discussion

This study successfully achieved its dual objectives. Firstly, it established that AI chatbotenhanced customer satisfaction significantly influences customer trust and loyalty, highlighting the importance of sophisticated AI chatbot interactions in modern customer relationship management. The AI chatbots' ability to provide instant responses, personalized experiences, and reliable service was found to enhance customer satisfaction, which in turn fostered greater trust and loyalty among customers.

Consistency with Hypotheses and Objectives

The findings confirmed the hypothesized relationships, showing that customer trust mediates the effect of satisfaction on loyalty. This consistency aligns with our theoretical expectations and research objectives, validating the role of AI chatbots in enhancing customer experiences.

Customer Satisfaction and AI Chatbots

The findings revealed that AI chatbots significantly enhance customer satisfaction by providing efficient, personalized, and immediate responses to customer inquiries. This aligns with

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the work of Følstad & Brandtzaeg (2017), who noted that the efficiency and multitasking capabilities of AI chatbots contribute to higher customer satisfaction levels. The positive association between customer satisfaction and trust, as well as between satisfaction and loyalty, extends the work of Følstad et al. (2018) and Glikson & Woolley (2020), who highlighted the pivotal role of credibility and ease of use in fostering cognitive trust.

Customer Trust and AI Chatbots

The study also illuminated the mediating role of customer trust, confirming that trust is not merely a byproduct of satisfaction but an essential conduit through which satisfaction is converted into loyalty. AI chatbots build trust by consistently providing accurate information and demonstrating reliability, which is crucial for nurturing a loyal customer base. This echoes the findings of Nordheim et al. (2019) and Chen et al. (2021), who underscored the chatbot-related factors that contribute to building consumer trust. Our investigation reaffirms their conclusions within the retail context, signifying that the competence demonstrated by AI chatbots is instrumental in nurturing a loyal customer base.

Customer Loyalty and AI Chatbots

Furthermore, the mediation analysis confirms the pivotal role of trust in bridging the satisfaction-loyalty nexus, resonating with Mozafari et al. (2022) and Chi et al. (2021). The study showed that customer loyalty is significantly influenced by the trust built through AI chatbot interactions. This suggests that businesses aiming to cultivate durable customer relationships and loyalty should prioritize the deployment of sophisticated AI chatbots. The observed significant influence of trust on loyalty amplifies the discourse by showcasing how AI's tangibility and immediacy can foster trust and loyalty, drawing a parallel with earlier assertions by Skogland & Siguaw (2004) and Suh & Youjae (2006) regarding the direct correlation between satisfaction and loyalty.

Implications for Theory and Practice

This research contributes to the literature by providing empirical evidence on the pivotal role of AI chatbots in shaping customer trust and loyalty, and by elucidating the mechanism through which customer satisfaction influences loyalty via trust. The conclusions drawn here point toward a new paradigm in customer engagement, where AI chatbots are central to fostering lasting customer relationships.

Theoretical Recommendations

The findings of this study offer several theoretical recommendations for advancing the understanding of AI chatbot integration in customer service. Scholars should consider exploring the nuanced mechanisms through which AI chatbots influence customer trust and satisfaction, potentially incorporating theories related to technology acceptance and relationship marketing. Further, examining the role of perceived empathy and personalization in AI chatbot interactions could enrich the theoretical frameworks that explain customer loyalty formation.

Practical Recommendations

Based on the study's findings, practical recommendations include businesses prioritizing the development of AI chatbots that not only provide efficient and accurate service but also embody elements of personalization and empathy to enhance customer satisfaction and trust. Companies should invest in continuous improvement and testing of AI chatbot functionalities to ensure reliability and relevance to customer needs. Additionally, incorporating customer feedback loops into the AI chatbot development process can help tailor interactions to meet and exceed customer expectations, thereby fostering loyalty. These strategies are crucial for businesses looking to leverage AI chatbots as a tool for building stronger, trust-based relationships with their customers in the digital age.

Suggestions for Future Research

Future research should extend beyond the retail sector to diverse industries and adopt longitudinal approaches to capture dynamic customer-chatbot interactions over time. Investigating the impact of evolving AI chatbot capabilities on customer perceptions and exploring additional mediators like perceived value and service quality could deepen understanding. Integrating objective data alongside self-reported measures could also enhance the robustness of findings, offering a more comprehensive view of AI chatbots' role in shaping customer trust and loyalty.

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