

The Role of Artificial Intelligence and Digital Technologies in Enhancing Customer Experience in Service Organizations

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Abstract: *The integration of artificial intelligence (AI) and digital technologies is fundamentally reshaping customer experience (CX) in service organizations. This paper examines how AI-powered systems—including chatbots, service robots, and intelligent assistants—transform service delivery through enhanced personalization, operational efficiency, and predictive analytics. Drawing on service-dominant logic and socio-technical systems theory, the analysis explores AI's dual role as both a resource and a non-human agent in digital service ecosystems. The paper identifies four archetypes of AI deployment: discriminative experience enhancers, protective ecosystem orchestrators, ecosystem innovation companions, and personalized service composers. While AI offers significant opportunities for improving customer satisfaction and loyalty, the analysis also addresses critical challenges including diminished perceived warmth, privacy concerns, unintended consequences, and the importance of preserving human-centric service values. The paper concludes by proposing a balanced, socio-technical approach to AI implementation that aligns technological innovation with customer expectations and organizational capabilities.*

Keywords: Artificial Intelligence, Customer Experience, Service Organizations, Digital Technologies, Service-Dominant Logic, Personalization, Service Robots, Chatbots

I. INTRODUCTION

Modern service encounters have evolved from traditional dyadic customer-provider interactions to complex polyadic interactions involving multiple customers, providers, and third-party stakeholders within digital service ecosystems (Lipkin & Heinonen, 2022; Vargo et al., 2017). Artificial intelligence (AI) has emerged as a transformative force within these ecosystems, introducing non-human participants—such as AI-based chatbots, service robots, and intelligent assistants—that fundamentally alter the nature of value co-creation (Bartelheimer et al., 2025). AI is distinct from general information technology as it can learn, connect, adapt, and display aspects of human intelligence (Huang & Rust, 2018). Service robots represent a specific application of customer interface AI-enabled technology, defined as "system-based autonomous and adaptable interfaces that interact, communicate and deliver service to an organization's customers" (Wirtz et al., 2018, p. 909).

The rapid proliferation of AI in service contexts is evidenced by adoption trends: reports indicate that 78% of organizations have implemented AI across at least one business function, particularly in IT, marketing, sales, and service operations. AI enables companies to reduce costs by up to 30% while improving efficiency, customer satisfaction, and return on investment. In the hospitality sector specifically, the International Federation of Robotics reported a 37% growth in global sales of service robots in 2022, projecting a market value of US \$62.35 billion by 2030.

Despite these advancements, the relationship between AI deployment and customer experience remains complex and multifaceted. While AI offers unprecedented opportunities for personalization and efficiency, concerns persist regarding diminished service warmth, privacy risks, ethical dilemmas, and unintended consequences. This paper

examines the role of AI and digital technologies in enhancing customer experience in service organizations, addressing both opportunities and challenges. The analysis is structured around three key questions: (1) How do AI and digital technologies transform service delivery and customer experience? (2) What are the mechanisms through which AI enhances or diminishes customer experience? (3) How can service organizations balance technological innovation with human-centric service values?

II. THEORETICAL FOUNDATIONS

2.1 Service-Dominant Logic and Digital Service Ecosystems

Service-dominant logic (SDL) provides a foundational framework for understanding value co-creation in service ecosystems (Vargo & Lusch, 2004, 2008, 2016). From an SDL perspective, actors integrate resources to co-create value within service ecosystems—defined as "relatively self-contained, self-adjusting systems" of actors co-creating value (Lusch et al., 2016, p. 11). Digital service ecosystems now encompass a growing number of non-human entities powered by AI that increasingly engage in value co-creation processes (Abedin et al., 2022; Larivière et al., 2017).

However, AI's role in SDL presents conceptual challenges. Traditionally, information technology is classified purely as a resource (Lusch & Nambisan, 2015; Mansour & Lindenfalk, 2019). Yet AI increasingly exceeds resource boundaries by performing tasks autonomously, adapting to contextual input, and engaging in interactive service processes (Zimmermann et al., 2021). Simultaneously, AI does not fit neatly into SDL's notion of actors, as actors are presumed to exhibit intentionality, to be guided by norms and beliefs, and to experience purpose and value-in-use (Lusch & Nambisan, 2015; Tronvoll, 2017). To address this conceptual ambiguity, recent scholarship proposes a tripartite view of AI as either a resource, a non-human agent, or an actor. A non-human agent is defined as "an entity capable of acting, adapting, and learning within defined constraints, yet without possessing the inherent agency a human actor has in SDL" (Pakkala & Spohrer, 2019). Its agency is contextually enacted and functionally delegated by human principals (Beck et al., 2022; Wilga et al., 2024).

2.2 Socio-Technical Systems Theory

Socio-technical systems (STS) theory provides a complementary framework for analyzing AI integration in customer service. STS theory conceptualizes organizations as interdependent combinations of technological and social subsystems. The technological subsystem encompasses AI capabilities that enable service processes—including natural language processing (NLP), machine learning (ML) algorithms, and computer vision technologies. The social subsystem comprises human actors, organizational practices, and behavioral dynamics—including customer expectations, trust perceptions, employee skills, and adaptation to hybrid human-AI environments.

Crucially, the interaction between these subsystems is co-evolutionary and interdependent. Technology and social elements continuously shape and reconfigure one another. AI deployment transforms service delivery mechanisms while being contingent upon human factors such as user trust, ethical concerns, and organizational readiness. This perspective highlights the importance of aligning technological innovation with human-centered design principles to achieve sustainable service transformation.

III. AI APPLICATIONS IN CUSTOMER SERVICE

3.1 Core Technologies

AI applications in customer service draw on several core technologies. Natural Language Processing (NLP) enables conversational interfaces that understand and respond to customer queries. Machine Learning (ML) algorithms support predictive analytics and personalization. Computer vision technologies enable automated and sensor-based service environments. The International Telecommunication Union (ITU) defines advanced intelligent customer service systems (AICSS) as systems integrating "large language model, multimodal large model, NLP, ASR and TTS technology, omnichannel access, omni media interaction and multimodal interaction capabilities" to provide "reliable, efficient, unified customer service experiences".

3.2 Archetypes of AI in Service Ecosystems

Recent research identifies four archetypes of AI in digital service ecosystems based on systematic taxonomy development and multiple case study analysis. The *Discriminative Experience Enhancer* focuses on optimizing individual service encounters through discriminative AI capabilities. The *Protective Ecosystem Orchestrator* coordinates across ecosystem participants to ensure security and reliability. The *Ecosystem Innovation Companion* drives innovation through generative AI and ecosystem-level learning. The *Personalized Service Composer* tailors services to individual customer needs through data integration and personalization algorithms. These archetypes vary along dimensions of ecosystem data usage, perceived ecosystem embeddedness, perceived behavior, and perceived value, providing a framework for understanding AI's diverse manifestations in service contexts.

3.3 Personalization and Customer Engagement

AI-driven personalization represents a key mechanism for enhancing customer experience. Recommendation engines and personalized content delivery contribute to heightened customer satisfaction and loyalty. Digital technologies enable organizations to collect and analyze vast amounts of data, providing valuable insights into customer behavior, market trends, and operational performance, thereby helping businesses make informed decisions and personalize their offerings. Self-service portals and online customer support systems can enhance the overall customer experience.

Evidence from a randomized field experiment at a meal delivery company demonstrated that AI-assisted agents responded faster, engaged customers more deeply, and achieved greater improvements in customer sentiment. Text analysis revealed that AI-assisted agents exhibited higher levels of key response characteristics: empathy, information, and solution orientation. The benefits were most pronounced for less-experienced agents, suggesting AI can serve as a capability-enhancing tool for frontline staff.

3.4 Service Recovery and Remediation

AI plays an increasingly important role in service recovery following failures. Service failures are a driving factor for customer switching behavior and are highly likely to result in loss of consumer satisfaction or loyalty, negative word-of-mouth, customer churn, and brand image damage. AI service remediation tools include intelligent customer service robots, automated refund/compensation systems, predictive maintenance AI that identifies service risks in advance, and dynamic interface adjustment AI that adapts interactions based on user emotions.

For example, predictive compensation mechanisms can identify anomalies through real-time transaction monitoring, automatically initiate refunds, and push compensation coupons. Multimodal emotion recognition can detect user anxiety through voice analysis and respond with appropriate service interventions. However, the effectiveness of AI service recovery depends on appropriate design. Research reveals that the effect of AI anthropomorphism on customer participation in service recovery is mediated by shifting emotion and aversion, with time pressure serving as a boundary condition.

IV. CHALLENGES AND UNINTENDED CONSEQUENCES

4.1 Diminished Perceived Warmth

A significant challenge in AI deployment is the potential reduction in perceived service warmth. Perceived warmth—defined as "a positive, mild, volatile emotion involving physiological arousal and precipitated by experiencing directly or vicariously a love, family, or friendship relationship" (Aaker et al., 1986, p. 366)—is a crucial predictor of impression formation and long-term relationships. Research demonstrates that AI (vs. human) service significantly reduces customers' perceived service warmth due to diminished mind perceptions (i.e., experience and agency) of AI staff as service providers. This warmth loss effect further leads to lower customers' continuous usage intention of AI service. This effect is particularly pronounced in high-touch industries such as hospitality, where perceived service warmth derived from customer-employee interactions is crucial for customer satisfaction and retention.

However, the effect can be attenuated when AI staff interacts with customers using an informal (vs. formal) language style or for customers with a utilitarian (vs. hedonic) motivation. These findings suggest that communication style and customer segmentation can mitigate negative warmth perceptions.

4.2 Unintended Consequences of Service Robots

Systematic review of service robot research identifies six key themes of unintended consequences: (1) customers' emotional responses, (2) customer misbehavior, (3) employee technostress, (4) privacy, ethics, and fairness concerns, (5) post-purchase behaviors and brand evaluations, and (6) negative aspects in business-to-business contexts. Unintended consequences occur when "the outcomes of a purposive act differ from what was sought (whether desirable or undesirable)" (Polonsky et al., 2024). These consequences often result from formally purposeful choices when alternatives exist, such as using a chatbot to replace customer service agents versus using AI to empower customer service agents. They can arise from trade-offs between desirable outcomes (e.g., efficiency, 24/7 customer service) and potential undesirable outcomes (e.g., dehumanization of interactions, deterioration of customer well-being).

4.3 Privacy and Ethical Concerns

AI implementation in customer service raises significant privacy and ethical concerns. Organizations face increasing risks such as inaccurate predictions, security violations, and intellectual property violations. Consumer dissatisfaction with AI customer service often focuses on AI's inability to accurately understand user needs, inefficient communication where problems remain unsolved, and cumbersome processes for transferring to human customer service. Research indicates that 51.4% of consumers indicated that apart from fixed scripts, AI could not solve personalized problems; 47.9% indicated AI could not accurately understand questions and answered off point; and 43% indicated responses were stiff and mechanical, lacking emotion.

4.4 Spillover Effects from Negative AI Experiences

Research reveals important spillover effects from negative AI experiences. Customers who have experienced chatbot comprehension failures and are then connected to AI-assisted human agents may experience negative sentiment because unusually rapid responses lead them to believe they are still communicating with a chatbot. This suggests that prior negative experiences with AI can contaminate subsequent interactions, even when human agents are involved.

V. INTEGRATION FRAMEWORK AND MANAGERIAL IMPLICATIONS

5.1 Reconceptualizing Service Quality in the Digital Era

Digital technologies dissolve the provider-consumer boundary and transform quality into a real-time, algorithmically mediated construct shaped by social validation rather than internal compliance. Prior quality management frameworks (e.g., TQM, SERVQUAL) examined quality or digital technologies in isolation. A new integrative framework embeds digital technologies as constitutive elements—not peripheral tools—that actively reshape how service quality is defined, delivered, and measured. This reconceptualization challenges the notion of quality as a static and provider-defined construct, proposing it instead as a dynamic, co-created phenomenon shaped in real-time by digital interactions. By mapping quality practices onto digital touchpoints, managers can deploy AI-powered personalization and sentiment analytics to monitor and enhance guest experiences in real-time. The tourism sector's improved service quality leads to better customer experiences, job satisfaction for employees, and sustainable business practices.

5.2 The Servicescape Paradigm Extension

The traditional servicescape model (Bitner, 1992) has been enhanced to incorporate AI-driven ambient conditions, spatial functionality, and interactive artifacts while introducing AI-augmented social interactions and a digital/virtual servicescape. This revised framework highlights how AI enhances personalization and efficiency, moderated by

factors such as customer technology readiness and service context, while emphasizing the need to preserve human-centric service values .

5.3 Implementation Recommendations

For service organizations implementing AI, several key recommendations emerge from the literature:

First, organizations should adopt a socio-technical perspective that recognizes the interdependence between technological innovation and social acceptance . Misalignment between high technological sophistication and low user trust may lead to suboptimal outcomes, highlighting the importance of integrated design and implementation strategies.

Second, AI deployment should consider conversation contexts, such as customer intent and prior chatbot interactions . Companies should understand how AI integration affects customer sentiment based on the nature of the service request and the customer's prior technology experiences.

Third, anthropomorphic design requires careful calibration. While higher levels of anthropomorphism can enhance customer willingness to interact, technical limitations exposed through highly anthropomorphic interaction can significantly amplify negative user experiences . The design of AI agent form and behavior must consider potential service failure scenarios and customer emotional responses.

Fourth, organizations should balance AI efficiency benefits with the preservation of perceived service warmth. Informal communication styles and targeting customers with utilitarian motivations can attenuate negative warmth perceptions .

VI. CONCLUSION

Artificial intelligence and digital technologies are fundamentally reshaping customer experience in service organizations. Through enhanced personalization, operational efficiency, predictive analytics, and service recovery capabilities, AI offers significant opportunities for improving customer satisfaction and loyalty. However, the relationship between AI and customer experience is complex and context-dependent. Challenges including diminished perceived warmth, privacy concerns, unintended consequences, and the risk of dehumanizing interactions require careful attention.

Service organizations should adopt a balanced, socio-technical approach that aligns technological innovation with human-centric service values. By understanding AI's diverse roles—from discriminative experience enhancer to ecosystem orchestrator—organizations can strategically deploy AI while preserving the warmth and trust that remain central to meaningful customer relationships. The future of customer service lies not in replacing human interaction but in augmenting it through intelligent, empathetic, and ethically-designed AI systems.

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