

International Journal of Artificial Intelligence in Healthcare

ISSN online: 3050-2470 - ISSN print: 3050-2462

https://www.inderscience.com/ijaih

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Esther Hwang, Yujong Hwang

DOI: 10.1504/IJAIH.2025.10072650

Article History:

Received: 27 November 2024
Last revised: 02 May 2025
Accepted: 03 May 2025
Published online: 20 October 2025

A literature review on artificial intelligence and healthcare management

Esther Hwang and Yujong Hwang*

DePaul University,

1 E. Jackson Blvd., Chicago, IL 60604, USA

Email: ehwang10@depaul.edu Email: yhwang1@depaul.edu *Corresponding author

Abstract: The purpose of artificial intelligence (AI) is to create an algorithm that functions autonomously to find the solutions to questions. However, the results that AI makes can lead to social biases and other selectivity issues. The social biases include negative statements to ethnic minority groups, gender biases, and cultural biases. Due to this reason, there is a research gap of AI and healthcare management such as AI biases and human-AI interaction. Thus, the goal of this literature review is to comprehensively examine the interaction of AI and users (patients who are in their mid or late-thirties, White, and live in the USA) specifically in the clinical healthcare environment to further enhance the usability of patients and AI.

Keywords: artificial intelligence; healthcare management; literature review; social bias; gender bias; cultural bias.

Reference to this paper should be made as follows: Hwang, E. and Hwang, Y. (2025) 'A literature review on artificial intelligence and healthcare management', *Int. J. Artificial Intelligence in Healthcare*, Vol. 1, No. 1, pp.48–59.

Biographical notes: Esther Hwang is an undergraduate student majoring data science in the Jarvis College of Computing and Digital Media at DePaul University. She is also in the University Honors program and received DePaul Presidential Scholarship Award and Arnold Mitchem Fellowship. She has published articles in *Human Technology, Journal of Economic Studies, Americas Conference in Information Systems*, and *Midwest Association for Information Systems Conference*. Her research interests include data science, human-computer interaction, artificial intelligence, healthcare management, and technology adoption.

Yujong Hwang is a Professor and Ezerski Endowed Fellow and Director of MIS program in the School of Accountancy & MIS at DePaul University, Chicago. He is also a Coordinator of Business Tech course at the Driehaus College of Business in DePaul. He received his PhD in business (MIS and accounting) from the University of South Carolina and was a Visiting Professor in the Kellogg School of Management at Northwestern University. He has published more than 90 refereed journal articles on human-AI interaction, e-commerce, and FinTech. He has received more than 10,000 Google Scholar citations with over 45 H-index.

1 Introduction

The integration of artificial intelligence (AI) into healthcare presents both unprecedented opportunities and complex challenges. As AI technologies become increasingly sophisticated, their potential to revolutionise disease prediction, diagnosis, and treatment is evident. However, the successful implementation of AI in clinical settings hinges on a crucial factor: the incorporation of patient perspectives. Understanding how patients perceive AI, what concerns they harbour, and what values they prioritise is essential for ensuring ethical, effective, and equitable AI integration. This literature review explores the multifaceted ways in which patient perspectives can inform the development and implementation of AI in healthcare, addressing key themes such as patient attitudes, ethical considerations, and potential limitations. By synthesising current research, this review aims to provide insights into how healthcare systems can harness AI's potential while maintaining patient-centred care.

2 Methodology

The university library's catalogue was the main search engine for retrieving sources. These sources were connected through databases such as DOAJ Directory, PubMed central, and much more. Table 1 represents the keywords that were used and the results.

Study	Keywords	Results
Themes (2024)	Artificial intelligence AND interaction	No results probably because applied the filter of IEEE Xplore in the database section
Adus et al. (2023)	Artificial intelligence AND interaction	Applied 'peer-reviewed articles' → still a bit broad of a search and should narrow
Amann et al. (2020)	Artificial intelligence AND interaction AND healthcare	Applied 'peer-reviewed articles'
Antes et al.	ChatGPT AND interaction AND	Applied 'peer-reviewed articles'
(2021)	healthcare	*Compared to number of results, ChatGPT is kind of narrow
Beets et al. (2023)	Artificial intelligence AND interaction OR experience AND healthcare	Applied 'peer-reviewed articles'
Eboigbe and Srinivasan (in press)	Artificial intelligence AND interaction AND cultural differences AND healthcare	Applied 'peer-reviewed articles'
Esmaeilzadeh et al. (2021)	Emotional artificial intelligence AND experience AND cultural	Applied 'peer-reviewed articles'

 Table 1
 A search log for this study (continued)

Study	Keywords	Results
Moy et al. (2024)	Artificial intelligence AND communication AND attitudes AND healthcare	Applied 'peer-reviewed articles' (a narrowed search result)
Richardson et al. (2022)	Artificial intelligence AND interaction OR usability AND perceptions OR attitudes AND healthcare	Applied 'peer-reviewed articles' but barely had any results related to my topic
Robertson et al. (2023)	Artificial intelligence in clinical applications AND interaction AND behaviours OR perceptions OR attitudes AND healthcare	Applied 'peer-reviewed articles' and results seem like they are from neuroscience?
Saurbrei et al. (2023)	Artificial intelligence in healthcare AND trust AND behaviours OR perceived attitudes	Applied 'peer-reviewed articles' and results are decent but not tailored to my needs
Witkowski et al. (2024)	Artificial intelligence in healthcare AND patients' experience OR user experience OR approaches AND behaviours OR beliefs	Applied 'peer-reviewed articles' and results are better than the previous search
Young et al. (2021)	Artificial intelligence in healthcare AND patients' experience OR approaches AND behaviours OR beliefs AND perceived attitudes	Applied 'peer-reviewed articles' and results are better than the previous search; to be honest, I am very satisfied with this search results

3 Important features to consider in using AI

Some of the important features to consider while using AI is to be informed about the potential benefits and risks of using them. While some may could say that benefits are the advantages and the risks are the disadvantages, it is not as simple as we think it is to define and assign benefits and risks in clinical healthcare. The common perceived benefits that most researchers claim about the use of AI in healthcare are predicting disease, making prognoses, and treating the patients (Beets et al., 2023; Esmaeilzadeh et al., 2021; Young et al., 2021). But, according to Beets et al. (2023) and Esmaeilzadeh et al. (2021), they provide more information regarding the formulation of benefits and risks by stating how the lack of systematic understanding the risks and benefits that AI technology in healthcare poses could pose a risk of providing the correct answers to the wrong questions in the healthcare community. Since both Beets et al. (2023) and Esmaeilzadeh et al. (2021) side with how patients' opinions are the driving force of formulating benefits and risks, this also justifies why Esmaeilzadeh et al. (2021) decided to focus on utilitarian aspects of the perceived benefits rather the motivational factors.

Even though Esmaeilzadeh et al. (2021) did not focus on the motivational factors, their research and much of the other researchers' findings in the literature showed a trend that emotion and psychology were the noticeable features of forming benefits and risks. Feelings, not pure facts, were shown to be common themes across the literature. For example, Antes et al. (2021) used a factor analysis to present of the two underlying response patterns reflecting a general extent of concern and perceived benefit. It appeared

that participants responded to the benefit/concern (i.e., positive/negative) framing of the items, not necessarily to the content of the topic discussed, but rather how the question was framed. In other words, the emotional trigger and not the pure content were the causes of formulating their attitudes of benefits and risks. In fact, Esmaeilzadeh et al. (2021) and Richardson et al. (2022) claim that rationale is not what patients use as their decision-making, but rather it is their anecdotal experiences. Only Richardson et al. (2022) did a focus group study of their design and is still justified how patients' overall opinions about using AI in healthcare are formed by experiences and backgrounds, which finalises to attitudes. The critical factors that influence the choice of using AI makes it very personal from person to person.

Lastly, if using AI is personal for each individual, the importance of trust and familiarity is another feature to consider. Trust and reliability seemed to be also a leading factor in Richardson et al.'s study. In addition, Saurbrei et al. (2023)'s study also highlighted that one way to know whether or not patients demonstrate their trust in AI is by indicating at their reliability. Saurbrei et al. (2023)'s study is an important study to be considered here because they focus on the doctor-patient relationship. Young et al. (2021) even presents how comfort levels also have an influence and impact on the use of AI. Overall, much of the researchers claim how increased familiarity and demonstrating reliability can foster greater trust.

Based off of these findings of what the important features of using AI in healthcare, more researchers are now finding what and how the viewpoints of using AI in healthcare tell us. Even if there is a lack of consensus based on findings due to formulating opinions of diverse groups of individuals, those opinions and viewpoints should not be disregarded in this field.

4 Patient viewpoints of using AI in healthcare

As the shape of attitudes is formed uniquely for each individual, the first concern that is present according to researchers' study is the lack of transparency and explainability. Another word for explainability is interpretability of AI decision-making processes. Young et al. (2021) proves that the findings of the literature show how the patients are concerned about not having their values considered. In addition, as Saurbrei et al. (2023) focuses on the doctor-patient relationship, they stress how if not used for this purpose, then it can reduce the accountability which relates to explainability. Within accountability, Saurbrei et al. also states how empathy is a crucial viewpoint and component for patients. Their findings conveyed how patients value empathy in the doctor-patient relationship that using AI can free up doctors' loads and have more time to connect with patients. However, if doctors have more time to spend talking to patients, but if they are unable to provide the necessary explanations about certain treatment decisions/prognoses and/or diagnoses suggested by the AI, the benefits of extra time may be limited. The lack of clear explanations and AI decisions can lead to ethical and legal dilemmas too. Explainability is also important because it determines whether the perceived benefits of using AI tools will be beneficial to patients. In fact, Esmaeilzadeh et al.'s study, their findings present how the perceived risks - feelings of uncertainty about healthcare outcomes, rather than AI-specific technical flaws - can cause the concerns of lack of transparency. Beets et al. (2023) also convey how Americans express a desire for transparency in the management of AI disease-diagnosis technology. This highlights the crucial factor of being open and transparent within the field of healthcare AI. But whether explainability should be reduced or maximised is still a question among researchers according to Saurbrei et al. (2023) as there is a lack of consensus.

Another dimension of trust is patient autonomy and control over their healthcare decisions. Research suggests that patients are more comfortable with AI when they feel it serves as a decision-support tool rather than a replacement for human judgement (Sauerbrei et al., 2023). Trust also depends on the system's consistency and accuracy over time - patients need reassurance that AI will perform reliably across different medical scenarios and patient populations (Beets et al., 2023). In cases where AI fails to meet expectations or introduces biases that disproportionately affect certain demographics, patient skepticism increases, further hindering adoption (Young et al., 2021). Amann et al.'s (2020) study does align with Saurbrei et al.'s claim that if AI decisions are opaque, they can undermine patient autonomy, making shared decision-making difficult. This also proved in Beets et al.'s study and Richardson et al.'s findings also justified how the value of controlling patients' own data is a need for the patients. Even though Saurbrei et al. (2023) agrees with how patients want more autonomy, their findings suggest that the emerging literature is divided on whether AI will enhance the doctor-patient relationship by encouraging shared decision-making through increased patient autonomy or create a new form of paternalism by hindering value-plurality. To build trust, AI in healthcare must prioritise clear communication, transparency, and fairness in decision-making. AI models should be designed with explainability features that allow patients to understand the rationale behind their recommendations. Additionally, healthcare providers should play a crucial role in mediating AI interactions, ensuring that AI-generated insights complement rather than replace human expertise. By addressing these factors, AI developers can foster a patient-centred approach that improves confidence in AI's role in clinical healthcare.

5 Future implementation and addressing the barriers

To ensure successful and patient-centred AI integration, several practical considerations must be addressed, with a strong emphasis on incorporating diverse perspectives. AI in healthcare should not be designed solely from a technological or medical standpoint but should integrate patient values and preferences, recognising that these values vary across different demographic and experiential backgrounds. Regulations and development strategies must address patient concerns about AI's role, potential risks, and impact on healthcare equity, with a focus on how these concerns may differ among diverse patient groups.

A critical aspect of practical implementation is acknowledging and accounting for the diversity of patient perspectives. Patient perceptions of AI in healthcare are highly variable, influenced by factors such as age, gender, employment, personality, and cultural background. This diversity underscores the need for inclusive and representative approaches in gathering patient input to inform AI development. Understanding how different groups perceive the benefits and risks of AI, and tailoring implementation strategies accordingly, is essential for equitable adoption.

Limitations in gathering patient input, such as the risk of dominant voices overpowering minority opinions in focus groups, need to be carefully managed to ensure

diverse perspectives are accurately represented. Strategies to mitigate this include employing facilitation techniques that encourage participation from all members and using a variety of data collection methods to capture a broad range of viewpoints. Ultimately, the practical implementation of AI in healthcare depends on a holistic approach that prioritises patient perspectives, promotes transparency and autonomy, and addresses ethical concerns, while also ensuring that the unique needs and concerns of diverse patient populations are fully considered.

6 Discussion

To better understand the current landscape of patient perspectives on AI in healthcare, a thematic synthesis of recent literature was conducted. This approach allowed for the identification of recurring patterns, concerns, and values expressed across diverse studies. By organising these insights into key thematic categories, the review provides a structured overview of how patients perceive, interact with, and respond to AI technologies in clinical contexts. Table 2 summarises these findings across multiple sources.

The tabular synthesis of the literature reveals that patients' perceptions of AI in healthcare are shaped by a constellation of factors, including trust, familiarity, diversity, and autonomy. Trust emerged as a foundational component of AI acceptance, often outweighing traditional socio-demographic predictors. Emotional responses and personal experiences – not purely rational assessments – were shown to heavily influence attitudes toward AI, echoing findings from recent work on affective computing and patient-centred design. Additionally, the importance of including diverse voices, particularly from historically marginalised populations, aligns with broader critiques in the literature regarding algorithmic bias and health equity. While patients recognised potential benefits in diagnostics and treatment optimisation, concerns about explainability, legal accountability, and the erosion of autonomy indicate persistent structural and ethical barriers. These findings suggest that successful integration of AI requires not only technological refinement but also robust participatory frameworks that prioritise patient values and transparent communication. Future research should build on this groundwork by empirically testing AI applications in clinical contexts, with special attention to personalisation, inclusivity, and regulatory adaptation.

While it would enhance the scope of this study to include an examination of the practical application of AI innovations in healthcare, the main objective was to synthesise existing literature that explores how patients' perceptions shape and inform the implementation of AI within healthcare systems. This study deliberately adopted a conceptual and theoretical lens, aiming to provide foundational insights into the attitudinal and perceptual factors that may influence the successful integration of AI technologies in clinical contexts. A significant limitation of this approach, however, is the exclusion of detailed analysis regarding specific use cases or implementation strategies. Figure 1 that justifies this study's overall reason to represent the topic's case in a visual.

 Table 2
 A tabular representation of the themes gathered from each source

	Ì		ı
Young et al. (2021)		Notes how diverse and diverse and inclusive responses to AI are sponses to AI be 'accurate' be 'accurate'	
Witkowski et al. (2024)		Other scholars have noted AT's potential to imappropriately 'mudge', 'mudge', participants to a specific behaviour, even acknowledging potential iste uses of 'fliss technology of 'fliss technology towards minority populations	
Saurbrei et al. (2023)		Stresses the importance of importance of including diverse and diverse and backgrounds backgrounds	There seems to be a clear division of whether or not explainability should be fully clear to patients
Robertson et al. (2023)			To secure the chemistry of Al in clinical practice, finiture research on best methods of physician methods of physician and patient decision making is required because there seems to be a lack of consensus of what patients preference is to receive diagnosis from.
Richardson et al. (2022)		To know what and and who spatients attitudes and beliefs in healtheare Al, the ambient of a died feets group (studying from different sociodemographics)	Another barrier to overcome is that even though fecus groups did enable important cross-demographic dialogue, they could dominant voices or overpowering minority voices or pushing the group to artificial consensus.
Moy et al. (2024)	Education in Al according to patients considered as lesser value to them	Building a policy that provides provides standardised deficiencs, best practices, and boundaries on the use of AI in healthcare that could potentially help many people	The perspectives in government, healthcare organisations, and imformation technology found concerns to be centred around barriers within the legal or regulatory landscape
Esmaeilzadeh et al. (2021)	According to results, age, race, level of caduction, employment, and amutal mousehold income had no significant difference on personal innovativeness		
Eboig be and Srinivasan (in press)		Highlights the minorance of creating a diverse virtuanin with a proper creating a diverse suppercentation approach (related to diabetes and diabetes and diabetes and dispurities in health in high-risk groups.	
Beets et al. (2023)		Perceptions such as such as such as perceived applications, benefits, and risks differ from background to background	
Antes et al. (2021)	Results show that in recreation analyses and regression analyses, and regression analyses, the importance of trust and faith were associated with openness greatly and proved that other warmliess (socio-demographies, personality, etc.) did not have an influence to openness in openses.	The study highlights that perceptors of Al differ based on factors (like age, gender, elliple age, gender, personality traits.	Even though the appearability may be a predictor, authors claim that if personality trails are involved in myolved in myolved in cerebiologies, this fact might be somewhat might be somewhat might be somewhat with the somewhat which the before the might be somewhat with the somewhat with the somewhat with the somewhat might be somewhat with the before in myolith the foreign personality tends to be fixed in season as harrier to overcome a barrier to
Amann et al. (2020)			Unclear Al decision-making could lead to ethical and legal diformas, such as ilability concerns when Al makes increase recommendations
Adus et al. (2023)	Patients' literacy in using AI in healthcare should be taught in clinical settings	The inclusion of the voices of patents experiencing marginalisation may serve as a method to combat the well-known implications that AI may have in my have in healthcare, aspecifically with respect to a lack of represented in the data, and a lack of prioritisation of particisation of prioritisation of prioritisation of practices by AI researchers and elevelopers	
Themes (2024)	How Human factors contribute to AI adoption	Importance of diversity	Barriers/ limitations

 Table 2
 A tabular representation of the themes gathered from each source (continued)

 Table 2
 A tabular representation of the themes gathered from each source (continued)

Themes (2024)	Adus et al. (2023)	Amann et al. (2020)	Antes et al. (2021)	Beets et al. (2023)	Eboigbe and Srinivasan (in press)	Esmaeilzadeh et al. (2020)	Moy et al. (2024)	Richardson et al. (2022)	Robertson et al. (2023)	Saurbrei et al. (2023)	Witkowski et al. (2024))	Young et al. (2021)
Patient autonomy		If AI decisions are apparte, they can undermine patient autonomy, making shared decision- making difficult.		Patients hope to control and have autonomy with their data		Chronic patients silvo they are not willing to use by physicians by physicians	Restrict the Results show the role of Al as an the value of assistant rather controlling their than a decision-maker for the patients	Results show that the value of controlling their own data is a need for the patients		Results show want autonomy, want autonomy, want autonomy, want autonomy, also take about a shared decision-making where the AI makes the decisions but addresses potential challenges where the AI makes the decisions nather the period of promof		
Accounting	Participants stackowledged the importance of importance of importance of object of the process o	Regulations and development strategies must account for patient concerns about AI's role, potential a risks and impact equity.			The study undersoors the need for personalised for personalised profile health strategies that account for ethnic differences and baseline vitamin D levels	Different perceived factors are inclusions of the perceived factors of the potential benefits and risks	To address those differences, stressed the stressed the building a building a building a public policy with that acquairs with the use of Al in healthcare for equity	Even though the authors do not addrectly state the previved factors, the authors clearly corroborate that papterns: a paper of the pape		Trust levels, past experiences, and demographic factors have factors have and risks and risks	Parient comfort and comfort and consent and consent should be incorporated by allowing patients to have choices to have choices in how Al is incorporated is use.	
Familiarity/ reliability								Familiarity was another key component of how patients' perspectives can influence the implementation and development of Al	Found substantial resistance to the use of Al, what some have called algorithm aversion or 'robophobia'	One way to know whether or not patients demonstrate their trust in AI is by indicating at their reliability		Comfort levels with using AI have an influence

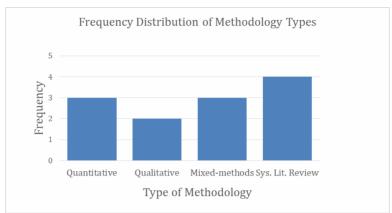


Figure 1 The frequency of the various methodology types based on literature (see online version for colours)

Most of the methodology other researchers chose for their study was done through a systematic literature review, as shown on the chart (Figure 1). Systematic literature reviews are the most used as a methodology to convey information about what the current research says about a topic. Though systematic literature reviews serve to organise the literature, there has not yet been many consensuses in what users' perceptions present about the development of AI in healthcare. The lack of consensus in the literature also represented in Figure 3 justifies the reason why this study took a theoretical approach rather than a practical approach. In fact, Figure 1 shows that there remains a critical need for empirical research that investigates the types of behaviours, expectations, and attitudes patients exhibit when interacting with AI-based systems in medical environments. Without a comprehensive understanding of these user perceptions, it may be difficult to advance discussions about the real-world deployment of AI in healthcare, including regulatory frameworks and operational integration.

Consequently, future research should expand beyond theoretical discourse and begin incorporating case studies of AI application done in healthcare such as diagnostic decision support, predictive analytics, patient monitoring, hospital workflow automation, and much more. These investigations should be accompanied by regulatory compliance, particularly with respect to established data protection laws such as Health Insurance Portability and Accountability Act (HIPAA) in the United States and the General Data Protection Regulation (GDPR) in the European Union. Additionally, forthcoming studies would benefit from identifying specific barriers to clinical adoption, such ethical concerns and proposing mechanisms to address algorithmic bias and underrepresentation of diverse populations.

In sum, even though this study does not provide practical insights of how to implement and incorporate users' perceptions of AI in clinical healthcare, it contributes to a growing body of literature that emphasises the importance of patient-centred perspectives in shaping future trajectory of AI in healthcare. As stated, subsequent research grounded in real-world application will be essential in bridging the gap between theoretical insight and practical implementation.

7 Conclusions

Overall, incorporating patient perspectives is paramount to the successful and ethical integration of AI in healthcare. By prioritising transparency, fostering trust, and addressing patient concerns, healthcare systems can ensure that AI serves as a tool to enhance, rather than undermine, patient-centred care. The themes discussed in this review highlight the importance of understanding patient attitudes, addressing ethical considerations, and implementing practical strategies that align with patient values. As AI technologies continue to evolve, ongoing research and dialogue with patients will be essential to navigate the complex landscape and realise the full potential of AI in transforming healthcare for the better. Ultimately, the future of AI in healthcare depends on a holistic approach that places the patient at the centre, ensuring that AI enhances empathy, autonomy, and the overall quality of care.

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