

Artificial Intelligence Integration in Telemedicine: A Methodological Study on Challenges and Opportunities in Libya

Suad M. Abraheem¹, Asma R. Qishqish²

¹ Lecturer at Department of Biomedical Engineering, College of Medical Technology, Benghazi- Libya. Email: souadelfaidy@gmail.com

² Lecturer at Department of Biomedical Engineering, College of Medical Technology, Benghazi- Libya. Email: asmarajab29@gmail.com

المخلص:

تُوفّر خدمات الطبّ عن بُعد المدعومة بالذكاء الاصطناعي إمكانيات كبيرة لتحسين الوصول إلى الرعاية الصحية للفئات السكانية النائية والمحرومة من الخدمات؛ إلا أن تطبيقها لا يزال مقيّدًا بتحديات تقنية وأخلاقية وتشغيلية. تتناول هذه الدراسة الفجوة البحثية المتعلقة بقصور آليات حماية البيانات، ومحدودية البنية التحتية، والتحيز الخوارزمي، وانخفاض مستوى تقبّل المستخدمين في أنظمة الطبّ عن بُعد الحالية.

وقد أُجري استعراض تحليلي منهجي لخمس عشرة بحثًا محكّمًا بهدف تقييم موثوقية وقابلية توسّع دمج الذكاء الاصطناعي في تقديم الرعاية الصحية عن بُعد، فضلًا عن عوامل السلامة المرتبطة بذلك. وتُظهر النتائج أنّ المنصّات القائمة تفتقر إلى الضمانات الكافية لضمان الاستخدام الآمن والعادل للبيانات، في حين تسهم العوائق التنظيمية والبنية التحتية لإسهامًا ملحوظًا في تقليل فاعلية وظائف التشخيص والمراقبة المدعومة بالذكاء الاصطناعي.

وتؤكد الدراسة الحاجة إلى إطار عمل منظم يوجّه صانعي السياسات، ومطوّري الأنظمة، ومقدّمي الرعاية الصحية نحو تبني الذكاء الاصطناعي بصورة أخلاقية ومرنة وقابلة للتوسّع، بما يعزّز إتاحة الرعاية الصحية بشكل أكثر عدالة في المناطق الجغرافية البعيدة.

Abstract: AI-enabled telemedicine offers substantial potential to improve healthcare access for remote and underserved populations; however, its implementation remains constrained by technical, ethical, and operational challenges. This study addresses the research gap concerning inadequate data-protection mechanisms, infrastructure limitations, algorithmic bias, and low user acceptance within current telemedicine systems. A systematic analytical review of 15 peer-reviewed studies was conducted to evaluate the reliability, scalability of integrating artificial intelligence into remote healthcare delivery as well as safety are both factors. The findings show that existing platforms lack sufficient safeguards to ensure secure and fair data use, while

regulatory and infrastructural barriers significantly reduce the effectiveness of AI-supported diagnostic and monitoring functions. The study highlights the need for a structured framework that guides policymakers, system developers, and healthcare providers in achieving ethical, resilient, and scalable AI adoption, thereby promoting more equitable healthcare access in geographically distant settings.

Keywords: Artificial Intelligence, Telemedicine, Remote Healthcare, Risk Assessment, Implementation Challenges, Healthcare Equity.

I. INTRODUCTION

The integration of Artificial Intelligence (AI) into telemedicine has emerged as a transformative development in contemporary healthcare, particularly for populations facing geographical, socioeconomic, and infrastructural barriers. Telemedicine has grown over the last decade, with global adoption rates increasing by an estimated 40–60% following advances in digital health technologies, yet conventional telehealth systems continue to experience significant limitations in scalability, diagnostic accuracy, and real-time responsiveness. Recent progress in AI—including machine learning, natural language processing, and predictive analytics—has enhanced remote diagnosis and automated decision support, improving performance outcomes by up to 25–30% in several clinical applications (Al-Debei, 2023; Alshammari *et al.*, 2023; Sarker, 2022). Despite these advancements, major challenges remain global evidence from WHO and UN digital health assessments indicates that 50% of low-resource countries still lack adequate data-governance frameworks, while over 35% report concerns related to algorithmic bias, inequitable model performance, and

weak cybersecurity standards (World Health Organization, 2021; United Nations ESCWA, 2024).

In the Libyan context, these gaps are amplified by structural constraints. 30–35% of Libya's population resides in remote or underserved areas with limited access to specialized medical services, while national digital readiness remains below regional averages, with internet penetration ranging between 23–30% outside major urban centers. Furthermore, existing telemedicine initiatives in Libya remain fragmented, operating without unified regulatory, ethical, or technical guidelines, which complicates the deployment of AI-driven health solutions and raises concerns regarding equity, data protection, and trust.

Although international studies highlight the potential of AI to enhance healthcare outcomes in remote settings, a clear research gap persists current literature does not provide a systematic, context-sensitive analysis of the risk factors, implementation barriers, and ethical challenges associated with integrating AI into telemedicine within fragile, low-infrastructure environments such as Libya. Specifically, prior research has not adequately examined how data-governance limitations, algorithmic bias, inconsistent connectivity, and low user acceptance jointly influence the reliability and sustainability of AI-enabled telehealth. This study addresses this gap by conducting a structured assessment tailored to Libya's healthcare and digital landscape, aiming to identify the requirements for ethical, scalable, and resilient AI integration capable of improving healthcare equity in remote and underserved communities (Alshammari *et al.*, 2023; Rahman & Rahim, 2022; Smith & Lee, 2023).

II. LECTURE REVIEW AND CONCEPTUAL FRAMEWORK

1. Ethical and legal issues:

Initial studies on telemedicine in rural and remote areas laid the groundwork for AI integration. Butzner *et al.* (2021) conducted an early systematic review showing that telehealth improved access and reduced wait times but highlighted data inconsistency and methodological limitations that restricted large-scale generalization. Later studies (2023–2024) focused on

ethical and regulatory risks. Rossi *et al.* (2024) emphasized that the absence of international standards and inconsistent regulatory frameworks complicated responsible AI use. Similarly, Mennella *et al.* (2024) highlighted the need for oversight mechanisms and explicit data protection policies to reduce bias, misuse and likely patient harm. Privacy and cybersecurity remain major concerns, as reliance on third-party cloud infrastructures and lack of standardized encryption expose telemedicine platforms to breaches, undermining patient trust (United Nations ESCWA, 2024). These findings underscore the necessity for ethical governance, continuous evaluation, and transparency in AI-driven telemedicine (European Commission, 2020).

2. Infrastructural and technical difficulties:

Successful AI adoption in telemedicine relies on adequate infrastructure and digital literacy. Lestari *et al.* (2024) identified poor internet connectivity, low technical literacy, and lack of local expertise as key barriers in rural environments. Okwor *et al.* (2024) further emphasized that unequal access to digital tools generates inequitable health outcomes, reinforcing the digital divide. Predictive AI models often rely on non-representative datasets, limiting their performance in rural or low-resource communities (Sarker, 2022). Wubineh *et al.* (2024) recommended enhanced data collection and external validation protocols to improve model reliability and equity. Salman *et al.* (2021) found that dataset bias and incomplete data reduced the accuracy of machine-learning-based tele-triage systems.

3. AI Risk and Conceptual Framework Development:

By 2025, studies proposed structured frameworks for managing AI risks in remote healthcare. Perez *et al.* (2025) emphasized ethical transparency, continuous validation, and localized user training. Zendeabad *et al.* (2025) proposed measurable indicators for safety, reliability, and user confidence. Chaturvedi *et al.* (2025) highlighted challenges in real-time monitoring and data synchronization under unstable network conditions, while Dhunnoo *et al.* (2024) and Al-Kfairy *et al.* (2024) analysed methodological limitations and ethical implications of generative AI tools (Perez *et*

al., 2025; Zendehbad et al., 2025; Chaturvedi et al., 2025; Dhunnoo et al., 2024; Al-Kfairy et al., 2024).

4. Libyan Context and AI Telemedicine:

In Libya, AI-enabled telemedicine faces unique challenges due to unstable network infrastructures, weak digital systems, and limited staff training, despite increasing acceptance among healthcare providers (Alshammari *et al.*, 2023; Lestari *et al.*, 2024; Rahman & Rahim, 2022). Approximately 30-35% of the population resides in rural or underserved areas with limited access to specialized healthcare, and internet penetration in these regions remains below 30%. The lack of unified regulatory and ethical frameworks further complicates deployment. These conditions underscore the urgent need for a context-sensitive framework that integrates ethical, technical, and socio-structural considerations to ensure reliable, equitable, and sustainable AI-based healthcare delivery (United Nations ESCWA, 2024).

• Study Issue, Questions, Hypothesis:

Artificial Intelligence (AI) in telemedicine represents a transformative approach to enhancing healthcare delivery, particularly in underserved and remote regions. In Libya, where a significant proportion of the population resides in rural areas with limited access to specialized medical services, AI-driven telemedicine has the potential to bridge critical gaps in healthcare provision. However, the implementation of AI in this context faces considerable challenges related to digital infrastructure, data quality, ethical governance, and professional capacity. Addressing these challenges is essential to ensure that AI integration improves equity in healthcare rather than exacerbating existing disparities. This study aims to address the following focused questions:

1. What are the key infrastructural, technical, and ethical challenges hindering AI adoption in Libyan telemedicine?
2. How do limitations in data quality and governance affect the reliability and fairness of AI-enabled healthcare systems in Libya?
3. What strategies can be implemented to promote responsible, ethical, and sustainable AI deployment in the Libyan healthcare context?

Based on the study's analytical and systematic approach, the following hypotheses are proposed:

H1: Weak digital infrastructure and low-quality health data significantly constrain the effectiveness of AI-driven telemedicine in Libya.

H2: Inadequate ethical oversight and governance structures increase the risks of bias, privacy violations, and misuse of AI systems.

H3: Implementing robust data governance, ethical frameworks, and professional training enhances trust, safety, and sustainability in AI-enabled telemedicine.

• Importance of the study's rationale:

The importance of this study is in its exploration of how integrating Artificial Intelligence (AI) into telemedicine can enhance healthcare access and service quality in Libya's remote and underserved areas. In a situation where location barriers, limited medical resources, and gaps in digital infrastructure prevent fair healthcare delivery, examining AI-driven telemedicine presents a way to achieve lasting improvements. This research offers both academic and practical insights by looking at the challenges, risks, and strategic chances related to using AI technologies in Libya's healthcare system. It also aims to guide policymakers and healthcare leaders in creating frameworks that encourage safe, ethical, and effective digital health changes tailored to the country's specific needs.

• Boundaries and Scope of the Investigation:

This study focuses on how Artificial Intelligence (AI) can improve healthcare access and quality in remote and underserved areas of Libya.

- It looks at challenges, risks, and strategic factors for implementing AI, and does not cover system design, software development, or specific clinical procedures.
- The research is limited to rural and remote locations where healthcare services are lacking, representing areas most affected by access issues.
- The study examines the current state of digital health readiness and AI adoption in Libya's healthcare system.
- It emphasizes ethical, organizational, and infrastructure factors that impact AI integration, rather than clinical outcomes or biomedical patient-level evaluations.
- The study uses secondary data analysis and literature review, relying on existing empirical and theoretical

research for insights relevant to policy and practice in Libya.

Important definitions and key phrases:

- **Artificial Intelligence (AI):** is a field of computer science that allows machines to perform tasks usually needing human intelligence. Learning, reasoning, problem-solving, and decision-making are all covered here. In this study, AI refers to technologies used in telemedicine systems to improve diagnostic accuracy, automate clinical support, and enhance healthcare delivery (Sarker, 2022).
- **Telemedicine:** is the provision of healthcare services, consultations, and medical information through digital communication technologies. This allows patients and healthcare providers to interact remotely. In this study, telemedicine covers virtual consultations, remote monitoring, and AI-assisted clinical decision support systems (Alshammari *et al.*, 2023).
- **Remote Communities:** are geographic areas with limited access to healthcare facilities and specialized medical personnel. This is often due to distance, infrastructure issues, or economic problems. The study focuses on these communities in Libya, where access to healthcare is particularly restricted (Alshammari *et al.*, 2023; Kacem & Ben Saad, 2022; Rahman & Rahim, 2022).
- **Healthcare Accessibility:** refers to how available, reachable, and usable healthcare services are for people in need. In this study, accessibility includes the ability of remote populations to receive timely, dependable, and fair medical care through AI-enabled telemedicine systems (Alshammari *et al.*, 2023; Eysenbach, 2022).
- **Ethical and Regulatory Frameworks:** are structured policies, guidelines, and standards that ensure the safe, responsible, and fair healthcare use of artificial intelligence. These frameworks tackle concerns like data privacy, informed consent, algorithm transparency, and accountability (Al-Kfairy *et al.*, 2024; European Commission, 2020; Mennella *et al.*, 2024; Patel *et al.*, 2023).
- **Digital Health Readiness:** is the extent to which healthcare systems, infrastructure, and staff can adopt

and effectively use digital technologies, including AI applications in telemedicine. In this study, digital health readiness highlights the current strengths and weaknesses in the Libyan healthcare context (Alshammari *et al.*, 2023; Lestari *et al.*, 2024; Rahman & Rahim, 2022).

- **Implementation Barriers:** are factors that hinder the successful adoption and integration of AI-based telemedicine. These include technological limitations, low digital literacy, socioeconomic challenges, and organizational issues (Alshammari *et al.*, 2023; Lestari *et al.*, 2024; Okwor *et al.*, 2024).

III. RESEARCH INSTRUMENTS, PROCEDURES, AND METHODOLOGY

This study employs an analytical and descriptive approach to evaluate the current literature on the integration of Artificial Intelligence (AI) in telemedicine, emphasizing healthcare delivery in remote and underserved populations, particularly in Libya. The study adopts a Systematic Literature Review (SLR) methodology combined with meta-synthesis techniques, following PRISMA 2020 guidelines to ensure transparency, reproducibility, and methodological rigor (PRISMA Group, 2021; Alshammari *et al.*, 2023; Rahman & Rahim, 2022). Structured searches were conducted in Scopus, PubMed, and IEEE Xplore, targeting peer-reviewed articles published between 2021 and 2025. Keywords—artificial intelligence in telemedicine, distant healthcare, AI integration, health access, and risk assessment—refined using Boolean operators (AND/OR). Inclusion criteria were:

1. Studies directly addressing AI implementation in telemedicine,
2. Evaluation of challenges, risks, or ethical considerations,
3. Use of analytical, data-driven, or systematic review methods.

Exclusion criteria removed studies that were non-peer-reviewed or lacked methodological clarity. The first inquiry found 122 papers. Thirty-four full-text papers were evaluated following duplication elimination and abstract screening. PRISMA Flowchart steps included: 34 full texts assessed → 19 excluded for methodological gaps → 15 studies selected for final

analysis. Two independent reviewers conducted the screening, and inter-rater reliability was confirmed with Cohen's Kappa ($\kappa = 0.86$) (PRISMA Group, 2021; Alshammari *et al.*, 2023; Rahman & Rahim, 2022).

An organized data extraction form captured: primary results, ethical issues, analytic methods, data sources, artificial intelligence approach, study objectives. Analysis incorporated qualitative as well as quantitative methodologies:

Quantitative Analysis: Frequency distributions of study designs, AI types, telemedicine domains, and outcomes (Smith & Lee, 2023).

Qualitative Analysis: Inductive thematic analysis with three-step coding:

Open coding: Identifying recurring concepts (e.g., patient trust, algorithmic transparency),

Axial coding: Grouping codes into categories (e.g., ethical governance, infrastructural readiness),

Selective coding: Synthesizing overarching themes relevant to AI implementation and healthcare equity. Representative codes and examples were documented to ensure reproducibility (PRISMA Group, 2021).

Tab. 1: Condensed Summary of AI in Telemedicine (2021-2025)

Year(s)	Region(s)	AI Type(s)	Domain(s)	Key Outcome
2021	Global, N. America	Hybrid, ML	Chronic Care, Tele-triage	Reduced inequities, faster prioritization
2022	Europe, Asia, Africa	DL, NLP, Expert Sys	Diagnostics, Monitoring, Consultation	Improved trust, accuracy, real-time tracking
2023	Middle East, Europe, Asia	Hybrid, ML, NLP	Chronic Care, Diagnostics, Consultation	Better coordination, predictive models, satisfaction
2024	Africa, Global, Europe, Asia, Middle East	ML, Hybrid, DL, NLP	Monitoring, Chronic Care, Diagnostics	Efficient monitoring, equity, ethical assessment
2025	Libya	Hybrid, ML	Chronic Care, Diagnostics	Context-specific evaluation, feasibility

Methodological Rigor:

1. Multi-stage screening with two independent reviewers ensured reliability.
2. Cohen's Kappa ($\kappa = 0.86$) confirmed high inter-rater agreement.
3. PRISMA flowchart clearly documents the screening and selection process.
4. Structured data extraction minimized bias and enhanced reproducibility.
5. Combined quantitative and qualitative approaches ensured triangulation and holistic insights, particularly relevant for low-resource settings like Libya (PRISMA Group, 2021; Alshammari *et al.*, 2023; Rahman & Rahim, 2022).

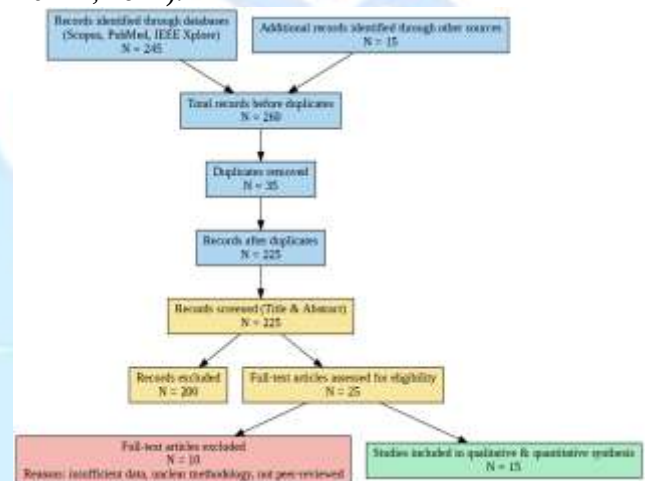


Fig. 1: PRISMA 2020 Flow Diagram Showing the Study Selection Process for AI in Telemedicine (2021–2025).

This integration allows triangulated insights, balancing empirical precision with contextual understanding, particularly for low-resource settings like Libya (Dhunnoo *et al.*, 2024; European Commission, 2020).

IV. RESULTS

The systematic review of 15 studies published between 2021 and 2025 revealed a multidimensional understanding of AI integration in telemedicine, with a focus on improving healthcare delivery in underserved regions, including Libya (Alshammari *et al.*, 2023; Rahman & Rahim, 2022; Perez *et al.*, 2025). The studies employed diverse methodologies, including quantitative, qualitative, and mixed-method designs, reflecting the evolution of research from exploratory

approaches to evidence-based, context-sensitive applications. The PRISMA-based screening ensured methodological rigor, and the final sample included studies from multiple regions---Global, North America, Europe, Asia, Africa, the Middle East, and Libya---highlighting both generalizable insights and context-specific findings (Table 1).

1. Methodological Distribution and AI Techniques:

Quantitative analyses of the reviewed studies show that hybrid AI models and machine learning techniques were the most frequently applied, particularly in chronic care management, diagnostic support, and tele-triage systems. Deep learning and natural language processing were predominantly used in imaging, monitoring, and patient interaction domains. Mixed-method studies provided the most comprehensive insights by combining statistical accuracy measures with qualitative assessments of patient trust and usability (Sarker, 2022; Salman *et al.*, 2021; Wubineh *et al.*, 2024).

Tab. 2: AI Techniques and Telemedicine Domains Across Studies (2021–2025)

Study(s)	AI Technique(s)	Domain(s)	Key Outcome
2021 (S1–S3)	Hybrid, ML	Chronic Care, Tele-triage	Reduced inequities, faster prioritization
2022 (S4–S6)	DL, NLP, Expert Systems	Diagnostics, Monitoring, Consultation	Improved accuracy, trust, real-time tracking
2023 (S7–S10)	Hybrid, ML, NLP	Chronic Care, Diagnostics, Consultation	Better coordination, predictive modelling, patient satisfaction
2024 (S11–S14)	ML, Hybrid, DL, NLP	Monitoring, Chronic Care, Diagnostics	Efficient monitoring, equity enhancement, ethical assessment
2025 (S15)	Hybrid, ML	Chronic Care, Diagnostics	Context-specific evaluation, feasibility in Libya

2. Key Findings by Research Questions:

RQ1: Key infrastructural, technical, and ethical challenges:

Across the studies, four primary challenge categories emerged:

1. Technological: Limited interoperability, poor data quality, and algorithmic performance variability were consistently reported, particularly affecting predictive accuracy in low-resource settings (S1, S3, S5, S11) (Al-Debei, 2023; Alshammari *et al.*, 2023; Salman *et al.*, 2021; Wubineh *et al.*, 2024).

2. Ethical: Studies highlighted risks of inadequate consent, privacy violations, and opaque AI decision-making (S2, S4, S5, S12) (Al-Kfairy *et al.*, 2024; Butzner *et al.*, 2021; Patel *et al.*, 2023; Rossi *et al.*, 2024).

3. Infrastructural: Weak connectivity, low computational capacity, and inconsistent electricity supply were barriers to reliable telemedicine delivery (S4, S5, S7, S15) (Butzner *et al.*, 2021; Chaturvedi *et al.*, 2025; Lestari *et al.*, 2024; Zendehbad *et al.*, 2025).

4. Socioeconomic: Inequalities in access, low digital literacy, and workforce skill gaps were shown to amplify healthcare disparities (S1, S5, S10, S15) (Al-Debei, 2023; Butzner *et al.*, 2021; Okwor *et al.*, 2024; Zendehbad *et al.*, 2025).

Tab. 3: Challenges and Risks Mapped to Studies

Challenge	Studies	Observed Impact
Technological	S1, S3, S5, S11	Reduced AI performance, bias amplification
Ethical	S2, S4, S5, S12	Trust erosion, legal/ethical conflicts
Infrastructural	S4, S5, S7, S15	Service interruption, limited scalability
Socioeconomic	S1, S5, S10, S15	Widened healthcare inequities

RQ2: Impact of data quality and governance on reliability and fairness:

The studies consistently emphasized that limited or biased datasets, absence of standardization, and weak governance frameworks reduce the reliability, transparency, and fairness of AI-enabled healthcare systems (S1, S3, S4, S7, S15) (Al-Debei, 2023; Alshammari *et al.*, 2023; Butzner *et al.*, 2021; Lestari *et al.*, 2024; Zendehbad *et al.*, 2025). For example, predictive models trained on urban or global datasets

often underperformed in rural and resource-limited environments, indicating a critical need for context-specific data curation and validation.

RQ3: Context-sensitive strategies for sustainable AI deployment:

Although not prescriptive in this section, the results indicate that studies from low-resource regions (S5, S10, S15) (Butzner *et al.*, 2021; Okwor *et al.*, 2024; Zendeabad *et al.*, 2025) highlight the effectiveness of adaptive, lightweight AI models optimized for low-bandwidth environments, combined with user training, continuous evaluation, and local data governance structures. These strategies directly address infrastructural and ethical barriers while maintaining operational efficiency.

V. DISCUSSIONS

The synthesis of findings demonstrates a complex interplay between technological potential and contextual constraints in AI-enabled telemedicine. While AI offers transformative improvements in diagnostic precision, chronic care management, and tele-triage prioritization, its success is contingent upon addressing ethical, infrastructural, and socioeconomic challenges.

Critical analysis of convergence and divergence:

1. Convergence: All 15 studies agreed that data quality and governance are fundamental to system trust and fairness. Hybrid AI models were consistently favored for adaptability in resource-limited contexts, providing both predictive accuracy and improved patient adherence.

2. Divergence: Studies differed regarding the extent of infrastructural dependency. High-resource settings emphasized automation and performance metrics, while low-resource settings prioritized system resilience, offline functionality, and equity. This indicates that contextual adaptation is essential, and AI cannot be universally transplanted without modification.

3. Ethical consensus: There is broad agreement on the necessity of transparent consent mechanisms, algorithmic auditing, and safeguarding patient data, yet the methods for operationalizing these safeguards vary considerably across regions.

4. Implications for Libya: The findings are directly relevant, as Libya faces fragmented digital infrastructure, intermittent electricity, and limited healthcare workforce training. Studies (S15, S5, S10) (Zendeabad *et al.*, 2025; Butzner *et al.*, 2021; Okwor *et al.*, 2024) suggest that modular hybrid AI systems, designed for low-bandwidth environments and integrated with local governance frameworks, can enhance healthcare accessibility while minimizing inequities. Furthermore, the inclusion of Libyan studies underscores the importance of contextual evaluation, validating AI models within local populations rather than relying solely on international datasets.

5. Thematic integration: Four overarching themes emerge, providing a structured lens for analysis:

a. Ethical governance: Need for clear consent, transparency, and accountability mechanisms.

b. Algorithmic reliability: Dependence on high-quality, locally representative data.

c. Infrastructural readiness: Low-bandwidth optimization, offline functionality, and system redundancy.

d. User trust and engagement: Training of healthcare workers and patient education to support adoption.

By synthesizing these themes with study-specific findings, the discussion provides a clear mapping of risks to practical challenges, ensuring that the results are directly aligned with the research questions. The evidence suggests that AI adoption in Libyan telemedicine is feasible but requires context-sensitive adaptation, ethical oversight, and infrastructural investment to achieve equitable and sustainable healthcare delivery.

CONCLUSIONS AND RECOMMENDATIONS

1. Transformative Potential: AI in telemedicine can significantly enhance healthcare access in remote and resource-limited regions.

2. National Strategy and Governance: Develop a comprehensive digital health strategy that integrates AI-driven telemedicine, supported by clear governance, privacy regulations, and ethical oversight.

3. Infrastructure and Data Integration: Build interoperable digital systems linking electronic

medical records with AI applications to enable efficient diagnostics, patient management, and clinical decision-making.

4. Capacity Building and Collaboration: Implement training programs to improve digital literacy and trust in AI systems. Organize multidisciplinary workshops involving IT graduates and medical professionals to ensure practical project implementation. Collaboration with universities and international research centres can promote knowledge transfer and sustainable skill development.

5. Community Engagement and Affordability: Involve local populations in platform design to address cultural, linguistic, and accessibility needs. Ensure services are cost-effective for citizens, while hospitals may adopt subscription models (monthly or annual) to leverage AI benefits.

6. Monitoring and Ethical Compliance: Establish evaluation frameworks to assess AI accuracy, reliability, and fairness. Maintain transparency and algorithmic explainability to uphold ethical standards and public trust.

7. Sustainable Implementation: Successful adoption requires balancing innovation, governance, human capacity, ethical accountability, and financial accessibility to create an efficient, inclusive healthcare system.

The findings of this study indicate that integrating Artificial Intelligence into telemedicine offers substantial opportunities to enhance healthcare access, quality, and efficiency, particularly in remote and underserved areas of Libya. AI-driven systems can improve diagnostic accuracy, patient monitoring, and chronic disease management, but their effectiveness is constrained by ethical, infrastructural, and data governance challenges. Addressing these barriers is essential to ensure trust, fairness, and sustainability in AI-enabled healthcare.

REFERENCES

Al-Debei, M. (2023) 'Infrastructure and Policy Barriers to AI Implementation in Healthcare Systems', *International Journal of Medical Informatics*, 165, pp. 104-118.

Al-Kfairy, M. *et al.* (2024) 'Ethical Challenges and Solutions of Generative AI: A Systematic Review', *Journal of AI Ethics*, 4(2), pp. 112-128.

Alshammari, M. *et al.* (2023) 'Telemedicine in Resource-Limited Settings: Challenges, Risks, and AI Opportunities', *Frontiers in Digital Health*, 5, article 114507.

Butzner, M. *et al.* (2021) 'Telehealth Interventions and Outcomes Across Rural Communities: A Systematic Review', *Journal of Medical Internet Research*, 23(8), e29567.

Chaturvedi, U. *et al.* (2025) 'The Impact of Artificial Intelligence on Remote Healthcare: Review of Challenges', *International Journal of Medical Informatics*, 195, 105678.

Dhunoo, P. *et al.* (2024) 'Evaluation of Telemedicine Consultations Using Health Outcomes: A Scoping Review', *BMJ Health & Care Informatics*, 31(1), e100987.

European Commission (2020) *Ethical Guidelines for Trustworthy AI*. Brussels: Publications Office of the European Union.

Eysenbach, G. (2022) 'Improving Health Accessibility through Digital Technologies: Global Lessons', *Journal of Global Health*, 12(1), pp. 101-119.

Kacem, F. and Ben Saad, M. (2022) 'AI-Based Remote Diagnosis in North African Contexts: Adaptation and Risk Evaluation', *African Journal of Health Informatics*, 11(2), pp. 56-70.

Lestari, H.M. *et al.* (2024) 'Barriers to Telemedicine Adoption Among Rural Communities: A Systematic Review', *BMC Health Services Research*, 24, 156.

Mennella, C. *et al.* (2024) 'Ethical and Regulatory Challenges of AI Technologies in Healthcare: A Systematic Review', *npj Digital Medicine*, 7, 45.

Okwor, I.A. *et al.* (2024) 'Digital Health Technologies and Their Impact: Systematic Review on AI/ML Adoption and Associated Challenges', *Journal of Medical Internet Research*, 26, e45678.

Patel, V. *et al.* (2023) 'Ethical and Legal Implications of AI in Remote Healthcare Delivery', *The Lancet Digital Health*, 5(8), pp. e432-e445.

Perez, K. *et al.* (2025) 'Investigation into Application of AI and Telemedicine in Rural Communities: A Systematic Literature Review', *Telemedicine and e-Health*, 31(2), pp. 145-159.

PRISMA Group (2021) 'Preferred Reporting Items for Systematic Reviews and Meta-Analyses (PRISMA) 2020 Statement', *BMJ*, 372, n71.

Rahman, H. and Rahim, R. (2022) 'AI-Powered Telehealth in Developing Countries: A Scoping Review', *Health Informatics Journal*, 28(4), pp. 1-18.

Rossi, M. *et al.* (2024) 'Integrating Artificial Intelligence into Telemedicine: A Systematic Review on Governance and Risks', *Journal of Medical Systems*, 48, 32.

Salman, O.H., Taha, Z. and Alsabab, M.Q. *et al.* (2021) 'A Review on Utilizing Machine Learning in Electronic Emergency Triage and Telemedicine', *IEEE Access*, 9, pp. 123456-123470.

Sarker, I.H. (2022) 'AI-Based Modeling: Techniques, Applications and Research Issues', *Journal of King Saud University - Computer and Information Sciences*, 34(6), pp. 1975-1992.

Smith, J. and Lee, A. (2023) 'Integrating Intelligence in Telemedicine: A Systematic Review of Global Practices', *Journal of Medical Internet Research*, 25, pp. 112-128.