

Open Access

Navigating the integration of large language models in healthcare: challenges, opportunities, and implications under the EU AI Act

Elena Bignami^{1*}, Michele Russo¹, Roberto Lanza¹ and Valentina Bellini¹

Introduction

In the rapidly evolving field of artificial intelligence (AI), large language models (LLMs) have emerged as powerful tools with transformative potential across healthcare sectors. These technologies promise enhanced diagnostic capabilities, patient education, and operational efficiencies. However, their integration into clinical practice is not without challenges, particularly in the context of stringent regulatory frameworks like the European Union's AI Act [1, 2]. This editorial explores the juxtaposition of innovation and regulation, offering insights into how healthcare professionals can navigate these dynamics responsibly.

Discussion

Transformative potential of LLMs in healthcare

LLMs are increasingly recognized for their ability to process vast datasets and generate human-like text, with applications spanning medical diagnostics, administrative tasks, and patient engagement [3]. From simplifying radiology reports to drafting discharge summaries, these tools streamline workflows while improving patient comprehension [4, 4–7]. Furthermore, they hold promise in drug discovery and personalized medicine, fostering innovation at an unprecedented scale [8].

*Correspondence:

elenagiovanna.bignami@unipr.it

LLMs can also democratize access to healthcare knowledge. By generating plain-language explanations of complex medical concepts, these models empower patients and support clinicians in underserved areas. In resourcelimited settings, LLMs can act as an accessible adjunct to healthcare professionals, mitigating the challenges of staffing shortages and expertise gaps [9].

However, these advancements must be balanced against inherent limitations. Their reliance on training datasets raises concerns about the quality, representativeness, and biases of the data, which can significantly impact outcomes. For instance, incorrect or oversimplified outputs may erode trust in healthcare systems or lead to adverse clinical decisions, underscoring the need for rigorous validation [10, 11].

The EU AI Act: a regulatory milestone

The European Union's AI Act, poised for adoption in 2024, establishes a comprehensive regulatory framework for AI systems, categorizing them by risk level—unacceptable, high, and limited. For healthcare, this risk-based approach translates into heightened scrutiny of AI applications in critical areas like diagnostics and treatment planning [2].

Key provisions and their implications

Risk categorization and transparency The Act mandates transparency for limited-risk systems and stringent requirements for high-risk applications. For healthcare providers, this ensures that AI tools are used with



© The Author(s) 2024. **Open Access** This article is licensed under a Creative Commons Attribution 4.0 International License, which permits use, sharing, adaptation, distribution and reproduction in any medium or format, as long as you give appropriate credit to the original author(s) and the source, provide a link to the Creative Commons licence, and indicate if changes were made. The images or other third party material in this article are included in the article's Creative Commons licence, unless indicated otherwise in a credit line to the material. If material is not included in the article's Creative Commons licence and your intended use is not permitted by statutory regulation or exceeds the permitted use, you will need to obtain permission directly from the copyright holder. To view a copy of this licence, visit http://creativecommons.org/licenses/by/4.0/.

Elena Bignami

¹ Anesthesiology, Critical Care and Pain Medicine Division, Department of Medicine and Surgery, University of Parma, Parma, Italy

informed oversight, fostering trust among clinicians and patients.

Prohibition of high-risk systems AI systems deemed unacceptable, such as those involving biometric categorization or workplace emotion recognition, are explicitly banned. This safeguards fundamental rights and aligns with ethical principles integral to medical practice.

Governance and accountability The establishment of regulatory sandboxes facilitates innovation while ensuring compliance. These controlled environments allow for the testing of AI tools in real-world scenarios, providing valuable insights without compromising ethical standards.

Ethical and legal challenges

While the AI Act offers a structured approach, its implementation poses significant challenges. Key issues include the following:

- i. Data privacy: LLMs process large datasets that may inadvertently include sensitive patient information, raising questions about data security and consent. Compliance with General Data Protection Regulation (GDPR) and related legislation are critical.
- Bias and equity: Models trained on skewed datasets risk perpetuating healthcare disparities. Proactive measures are required to identify and mitigate such biases, particularly when deploying LLMs in multicultural settings.
- iii. Intellectual property: The ownership of outputs generated by LLMs remains a contentious issue, particularly in collaborative medical research where authorship and credit must be carefully managed.

OpenAl's privacy policies and healthcare considerations

The EU AI Act necessitates a revaluation of data handling practices by entities like OpenAI. While its privacy policies address general principles, greater specificity is required for high-risk sectors such as healthcare [12]. For instance as follows:

- i. Handling sensitive data: Clear guidelines on managing healthcare data are essential to ensure compliance with GDPR and other local regulations.
- ii. Transparency and user awareness: OpenAI must enhance disclosures regarding AI-generated content, particularly in clinical contexts where decisions may significantly impact patient outcomes.

iii. Mitigation of risk: OpenAI should consider developing healthcare-specific safeguards, including limitations on the use of LLMs in critical care decisions without clinician oversight.

Fostering trust and collaboration

For LLMs to be effectively integrated into healthcare, fostering trust among all stakeholders is paramount. Patients, clinicians, and policymakers must be confident that AI tools are both safe and beneficial. This necessitates the following:

- i. Patient-centric AI: Models must prioritize patient welfare, including clear communication of their limitations and a robust mechanism for addressing errors.
- ii. Interdisciplinary collaboration: A collaborative approach involving engineers, ethicists, clinicians, and legal experts will ensure that LLMs are developed and deployed with comprehensive oversight.
- iii. Global standards: The fragmented nature of AI regulation highlights the need for international harmonization. As the EU leads with the AI Act, other nations must align their frameworks to ensure consistency and interoperability.

Balancing innovation and responsibility

The integration of LLMs in healthcare exemplifies the tension between technological progress and ethical responsibility. To achieve a sustainable balance as follows:

- i. Education and training: Healthcare professionals must be equipped with the skills to evaluate AI tools critically. Curricula should incorporate AI literacy, focusing on its applications, limitations, and ethical implications [13].
- ii. Continuous evaluation: AI tools should undergo ongoing assessments to ensure they meet evolving regulatory standards and clinical needs.
- Encouraging innovation: Regulatory sandboxes and similar initiatives allow innovation to flourish while maintaining ethical oversight, creating a fertile ground for AI-driven advancements.

The role of healthcare leaders

Leadership in healthcare will play a pivotal role in determining the trajectory of AI integration. By advocating for responsible innovation, healthcare leaders can influence policy development, foster interdisciplinary collaboration, and guide the ethical deployment of LLMs. Their

Table 1 Summary table: integration of large language models in healthcare

Category	Key points
Transformative potential	Streamlines clinical workflows with tasks like medical document summarization and discharge summaries
	Enhances medical education and personalized care through rapid synthesis of literature and tailored health advice
	Supports multimodal LLMs (M-LLMs) for integrating text, images, and sensor data to improve diagnostic accuracy
Challenges	Concerns over data privacy, particularly handling sensitive patient information
	Risks of biased outputs from nonrepresentative datasets perpetuating healthcare inequities
	Issues with model reliability and interpretability in critical clinical decisions
Ethical concerns	Ensuring fairness and transparency in Al-generated content
	Addressing biases to prevent exacerbating existing disparities in healthcare delivery
	Safeguarding patient trust through robust ethical and regulatory oversight
Role of EU AI Act	Establishes a risk-based framework categorizing AI systems by risk (limited, high, and unacceptable)
	Mandates transparency and prohibits high-risk systems like biometric categorization
	Encourages innovation via regulatory sandboxes, balancing progress with safety
Future directions	Calls for interdisciplinary collaboration among clinicians, technologists, and ethicists
	Promotes continuous evaluation and refinement of AI models to align with evolving healthcare needs
	Advocates for global standardization of regulations for consistent Al governance

voices are crucial in shaping a future where AI enhances, rather than replaces, the human touch in medicine [14] (Table 1).

Conclusion

LLMs present a dual-edged sword: immense potential to enhance healthcare delivery paired with challenges that demand meticulous oversight. The EU AI Act provides a robust framework, but its success hinges on collaborative efforts among stakeholders. As healthcare professionals, embracing these tools responsibly will ensure they augment, rather than undermine, clinical excellence.

Abbreviations

Al	Artificial intelligence
LLMs	Large language models
EU	European Union
M-LLMs	Multimodal large language models
GDPR	General Data Protection Regulation

Acknowledgements

Not applicable

Authors' contributions

E.B., M.R., R.L., V.B. conception, writing and proofreading of the paper.

Funding

No funding was obtained for the present study.

Data availability

No datasets were generated or analysed during the current study.

Declarations

Ethics approval and consent to participate Not applicable.

Consent for publication

Not applicable.

Competing interests

The authors declare no competing interests.

Received: 22 November 2024 Accepted: 28 November 2024 Published online: 02 December 2024

References

- Wang D, Zhang S (2024) Large language models in medical and healthcare fields: applications, advances, and challenges. Artif Intell Rev 57:299. https://doi.org/10.1007/s10462-024-10921-0
- European Parliament. EU AI Act: first regulation on artificial intelligence. 2. https://www.europarl.europa.eu/news/en/headlines/society/20230601ST O93804/eu-ai-act-first-regulation-on-artificial-intelligence. Accessed 17 November 2024.
- 3. Cascella M. Montomoli J. Bellini V. Bignami E (2023) Evaluating the feasibility of ChatGPT in healthcare: an analysis of multiple clinical and research scenarios. J Med Syst 47(1):33. https://doi.org/10.1007/ \$10916-023-01925-4
- 4. Jeblick K, Schachtner B, Dexl J, et al (2022) ChatGPT makes medicine easy to swallow: an exploratory case study on simplified radiology reports. arXiv. 2212.14882. https://doi.org/10.48550/arXiv.2212.14882
- 5. Lyu Q, Tan J, Zapadka ME et al (2023) Translating radiology reports into plain language using ChatGPT and GPT-4 with prompt learning: results, limitations, and potential. Vis Comput Ind Biomed Art 6(1):9. https://doi. ora/10.1186/s42492-023-00136-5.
- Reddy S (2023) Evaluating large language models for use in healthcare: 6 a framework for translational value assessment. Informatics in Medicine Unlocked. 41:101304. ISSN 2352-9148. https://doi.org/10.1016/j.imu. 2023.101304.
- 7. Patel SB, Lam K (2023) ChatGPT: the future of discharge summaries? Lancet Digit Health 5(3):e107-e108. https://doi.org/10.1016/S2589-7500(23) 00021-3
- 8. Wang L, Wan Z, Ni C, Song Q, Li Y, Clayton EW, Malin BA, Yin Z (2024) A systematic review of ChatGPT and other conversational large language models in healthcare. medRxiv [Preprint]. 2024.04.26.24306390. https:// doi.org/10.1101/2024.04.26.24306390.
- 9 Nassiri K. Akhloufi MA (2024) Recent advances in large language models for healthcare. BioMedInformatics 4(2):1097-1143. https://doi.org/10. 3390/biomedinformatics4020062.

- 10. Jui TD, Rivas P (2024) Fairness issues, current approaches, and challenges in machine learning models. Int J Mach Learn & Cyber 15:3095–3125. https://doi.org/10.1007/s13042-023-02083-2.
- Cascella M, Bellini V, Montomoli J, Bignami E (2023) The power of evolution cannot be contained, so let it be. Minerva Anestesiol. https://doi.org/ 10.23736/S0375-9393.23.17484-0.
- 12. Open Al privacy policy. (https://openai.com/policies/eu-privacy-policy Accessed: 17 November 2024).
- Moldt JA, Festl-Wietek T, Fuhl W, et al (2024) Assessing AI awareness and identifying essential competencies: insights from key stakeholders in integrating AI into medical education. JMIR Med Educ. 10:e58355. Published 2024 Jun 12. https://doi.org/10.2196/58355.
- 14. Sriharan A, Sekercioglu N, Mitchell C, et al (2024) Leadership for Al transformation in health care organization: scoping review. J Med Internet Res. 26:e54556. Published 2024 Aug 14. https://doi.org/10.2196/54556.

Publisher's Note

Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.