

The current and potential future roles of artificial intelligence in ophthalmology

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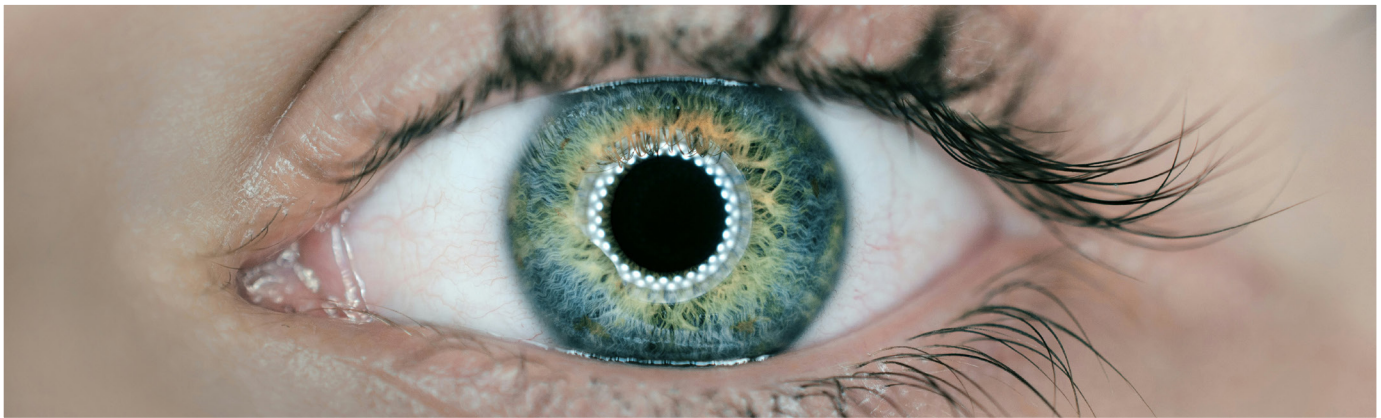
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Abstract

Artificial Intelligence (AI), initially conceptualised through Alan Turing's Turing Test in 1950, has since evolved into a transformative technology. AI encompasses a broad range of computational methods that mimic human intelligence, with Machine Learning (ML) as a subfield focused on pattern detection in structured datasets. Deep Learning (DL), a further subset of ML, utilises neural networks modelled after the human brain to solve complex data-driven problems. As computer processing power advances, AI's role in fields such as ophthalmology has expanded. Current applications include AI-powered diagnostic systems, such as IDx-DR for diabetic retinopathy, and DeepSeeNet for age-related macular degeneration (AMD), which have demonstrated high accuracy, sensitivity and specificity. These advancements promise reduced diagnostic costs and improved disease management. While challenges such as algorithmic bias and ethical concerns persist, the potential of AI to revolutionise patient care and medical education remains substantial as AI tools continue to evolve.

Abbreviations

AI – artificial intelligence

AMD – age-related macular degeneration

ML – machine learning

DL – deep learning

DR – diabetic retinopathy

VR – vitreoretinal

Introduction

When Alan Turing introduced the Turing Test in 1950, he could not have anticipated how his idea would lay the groundwork for artificial

intelligence (AI), which is described by some as the 'fourth revolution of mankind'.¹ AI integrates datasets with computer science to mimic human intelligence in solving problems. Within AI, machine learning (ML) identifies patterns and models results from structured datasets, while deep learning (DL), a subset of ML, utilises neural networks to process data and solve complex problems² (**Figure 1**). As a result of its potential, there had been an increasing amount of excitement in the research in the application of AI in the medical field. This article explores the current applications of AI in ophthalmology and its untapped potential.

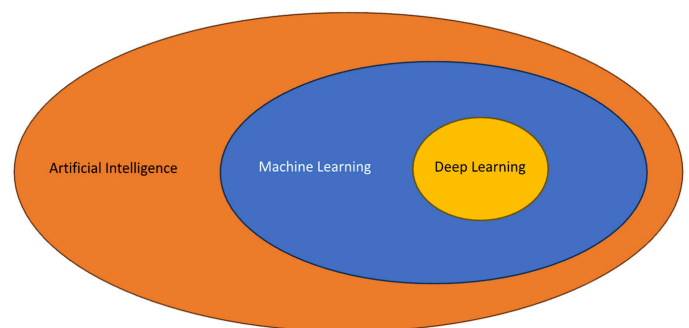


Figure 1. Diagram explaining the terms of Artificial Intelligence (AI), Machine Learning (ML), and Deep Learning (DL)

AI as a diagnostic tool

A good diagnostic tool is characterised by high sensitivity and specificity, cost effectiveness and ease of deployment. Such tools enable ophthalmologists to detect and manage diseases promptly, reducing their progression and avoiding complications.

Diabetic retinopathy (DR), a complication of diabetes, affects

one-third of diabetics and can lead to neovascularisation, retinal detachment and vision loss.³ Ophthalmology became the first medical specialty to adopt a US FDA-approved autonomous AI diagnostic system, IDx-DR.^{4,5} This system achieved sensitivities and specificities of 87.2% and 90.7% respectively in detecting DR in patients with diabetes with no previous history of the condition.⁴ NICE's Medtech Innovation Briefing highlights how AI technologies like RetinaLyze and Retmarker could reduce screening costs from £4.79 per patient to £0.35–£0.86 and decrease reliance on staff.⁶

Age-related macular degeneration (AMD) is the leading cause of irreversible vision loss in developed nations, accounting for over 9% of global blindness.⁷ DL models, such as DeepSeeNet, have demonstrated superior accuracy compared to retinal specialists in risk stratification for AMD.⁸ One study reported comparable sensitivity, specificity and precision in detecting late AMD progression.⁹ Furthermore, a separate system outperformed five out of six retinal experts in predicting disease progression, underscoring AI's potential in diagnostics.¹⁰

AI as a treatment tool

AI is particularly prominent in vitreoretinal (VR) surgery due to its complexity. AI was shown to be able to localise, classify and segment tissues and instruments during VR procedures, which can be used by surgeons to enhance precision and reduce surgical risks.¹¹

AI as an educational tool

AI can be used to track and analyse surgical movements in different stages of cataract surgery and compare the movements with expert surgeons to highlight the difference in techniques.¹² Moreover, AI can also be deployed to highlight key learning points of the surgery to facilitate independent learning.¹³ Cybersight, an online training and mentorship service for eye health professionals in developing countries created by Orbis, has integrated AI in the platform to enable users to utilise AI in learning, detecting and diagnosing eye conditions in fundus images.¹⁴

AI's limits and concerns

Despite its potential, AI has limitations. Algorithms require extensive datasets for training, making it challenging to develop accurate models for rare eye diseases due to limited data availability. Additionally, AI models are susceptible to algorithmic bias, as their effectiveness depends on the quality and diversity of their training datasets. For example, biases in datasets can lead to disparities in diagnostic outcomes across populations.¹⁵

Ethical concerns also pose challenges. Issues of accountability, data privacy and informed consent must be addressed to ensure AI's responsible implementation.¹⁵ The use of dataset in training AI models presents unique risks to data security, necessitating robust measures to protect sensitive information. Establishing clear guidelines for algorithm transparency and performance evaluation is critical for building trust and ensuring ethical compliance.

Future directions

The capacity of AI as an accurate diagnostic tool will strengthen in the future as more datasets in various ophthalmic conditions are produced and models are trained on them. With the maturation of knowledge in AI diagnostic tools, it is hopeful that deploying AI models will become more cost-effective, especially in developing countries.

The primary challenge for AI is to seamlessly integrate AI research into daily clinical practice while addressing ethical concerns. Ethical concerns can be addressed by adhering to AI-research guidelines such as Consort-AI and Spirit-AI, which aim to improve transparency

and completeness of clinical trial protocols involving AI as well as other future guidelines and frameworks.¹⁶

Conclusion

It is evident that the application of AI in ophthalmology will continue to expand as datasets become more robust and extensive, with more AI algorithms being developed and trained. Coupled with the increasing processing power of computer chips, the potential of AI in enhancing patient care will continue to grow. Equally important is the need to equip future healthcare professionals with AI-related knowledge, enabling them to utilise this multifaceted tool effectively and safely, thereby unleashing its full potential.

Contribution statement

Both authors had substantial contributions to the conception or design of the work; or the acquisition, analysis, or interpretation of data for the work and drafting the work or revising it critically for important intellectual content and the final approval of the version to be included in Inspire. Vincent Ng is responsible for the integrity of the work as a whole.

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Is a fifth-year medical student at Cardiff University with a keen interest in burns and reconstructive plastic surgery. Passionate about research, she has contributed to various clinical projects and societies, including INSPIRE and OSCEazy. She is also exploring the integration of AI in clinical practice and medical education.

Currently, Coco is on elective in New Zealand, gaining hands-on experience at the renowned National Burns Centre. Outside of medicine, Coco enjoys traveling and visual arts, using creativity as a means to explore new perspectives and express her passion for science and humanity.



Vincent Ng

Is an academic F2 doctor at University Hospitals Birmingham Trust. He has a strong passion for ophthalmology, technology and medical education. He is the co-founder of international medical education platform, OSCEazy and he is currently an honorary tutor for Cardiff University and an honorary research fellow for Moorfields

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