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Review

Scientific discovery in the age of artificial intelligence

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Artificial intelligence (AI) is being increasingly integrated into scientific discovery to augment and accelerate research, helping scientists to generate hypotheses, design experiments, collect and interpret large datasets, and gain insights that might not have been possible using traditional scientific methods alone. Here we examine breakthroughs over the past decade that include self-supervised learning, which allows models to be trained on vast amounts of unlabelled data, and geometric deep learning, which leverages knowledge about the structure of scientific data to enhance model accuracy and efficiency. Generative AI methods can create designs, such as small-molecule drugs and proteins, by analysing diverse data modalities, including images and sequences. We discuss how these methods can help scientists throughout the scientific process and the central issues that remain despite such advances. Both developers and users of AI tools need a better understanding of when such approaches need improvement, and challenges posed by poor data quality and stewardship remain. These issues cut across scientific disciplines, so they require developing foundational algorithmic approaches that can enable scientists to find scientific understanding or acquire it autonomously, making them more effective for AI innovation.

The foundation for forming scientific insights and theories is built on the principles of scientific method, from physical instruments such as microscopes to mathematical models. The introduction of digitization in the early 1950s has transformed the way data are collected, transformed and understood. The development of machine learning in the early 2010s has significantly expanded the scope of scientific discovery processes. The general use of computing in scientific discovery since the 1950s has enabled AI to revolutionize scientific discovery by identifying scientifically relevant patterns in large datasets. The practices and procedures vary across stages of the scientific process, and the development of AI algorithms cuts across disciplines (Fig. 1). Such algorithms can enhance every stage of the execution of scientific studies. They are becoming

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