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# And Plato met ChatGPT: an ethical reflection on the use of chatbots in scientific research writing, with a particular focus on the social sciences

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This interdisciplinary paper analyzes the use of Large Language Models based chatbots (LLM-chatbots), with ChatGPT the most known exponent, in scientific research writing. By interacting with LLM-chatbots, researchers could reduce efforts and costs as well as improve efficiency, but taking important risks, limitations, and weaknesses, which could highly-order erosion scientific thought. While many scientific journals, as well as major publishers such as Springer-Nature or Taylor & Francis, are restricting its use, others advocate for its normalization. Debate focuses on two main questions: the possible authorship of LLM-chatbots, which is majority denied because their inability to meet the required standards; and the acceptance of hybrid articles (using LLM-chatbots). Very recently, focusing on the education area, literature has found analogical similarities between some issues involved in Chatbots and that of Plato criticisms of writing, contained in the Phaedrus. However, the research area has been neglected. Combining philosophical and technological analysis, we explore Plato's myth of Theuth and Thamus, questioning if chatbots can improve science. From an interdisciplinary perspective, and according with Plato, we conclude LLM-chatbots cannot be considered as authors in a scientific context. Moreover, we offer some arguments and requirements to accept hybrid articles. We draw attention to the need for social science publishers, an area where conceptual hypotheses can take a long time to confirm, rather than solely on experimental observations. Finally, we advocate that publishers, communities, technical experts, and regulatory authorities collaborate to establish recommendations and best practices for chatbot use.

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## Introduction

Artificial intelligence (AI) is rapidly transforming several aspects of our lives. We have joined the age of AI. Large Language models (LLMs) based chatbots (hereafter, LLM-chatbots), like ChatGPT, are the great exponent of the hype generated in this decade. LLM-chatbots are increasingly used in the domain of investigation and scientific publication (Giray et al., 2024). The fact that communication between scientists involves both researchers and chatbots (which are capable of producing new content, without the immediate supervision of humans) is something unique, unknown, and with very profound implications (Dwivedi et al., 2023b). In fact, this tool is changing how science is done (Van Noorden, 2022).

By interacting with LLM-chatbots, researchers could reduce efforts and costs as well as improve efficiency (Cf. Table 2, Appendix), but taking important risks (Kendall, da Silva (2024)). ChatGPT's rapid adoption presents a wide range of concerns, limitations, and weaknesses (Cf. Table 1, Appendix), which could highly erode scientific thought (Stokel-Walker, 2023). For instance, despite GPT-4 being technically a significant improvement over GPT-3.5, 18% of its citations are still fabricated and 24% contain significant errors.

Faced with this ambivalence, the scientific community is divided. While many journals and major publishers (Cf. Table 3, Appendix) are restricting its use, other publishers and journals consider that the prohibition is unfeasible and simply ask that its use be reported (Sallam, 2023). Debate focuses on two main questions (Thorp and Vinson, 2023; Lund and Naheem, 2024; Nazarovets and Teixeira da Silva, 2024): the possible authorship of chatbots; and the acceptance of hybrid articles (using LLM-chatbots).

Very recently, some authors have found the fears we see with LLM-chatbot echo, in some ways, the fears Plato captured in the Phaedrus (Bedington et al., 2024), establishing analogical similarities between Plato criticisms of writing and some issues involved in chatbots such as authenticity (Deptula et al., 2024), moral hazard (Loos and Radicke, 2024) or plagiarism (Misra and Ravindran, 2021). While most articles focus on the education area (Aylsworth and Castro, 2024; Bingham, 2024; Kitinger, 2024), employing an instrumental analysis of pros and cons such as literature synthesis, citations, data analysis, etc., (Rahman et al., 2023), the research area has been neglected. Combining philosophical and technological analysis, we will explore Plato's myth of Theuth and Thamus as a means of providing insight and approaches to the use of ChatGPT on research, critically discussing LLM-chatbots authorship and hybrid articles in academic journals. We will conclude LLM-chatbots cannot be considered as authors in a scientific context. Moreover, we offer two practical and one theoretical argument to accept hybrid articles produced with the LLMs-chatbot's help, but with some preliminary requirements with clear specification of where and for what purpose an LLM-chatbot must be used. Finally, we advocate that publishers, communities, technical experts, and regulatory authorities collaborate to establish recommendations and best practices for chatbot use.

Today, LLM-chatbots could generate incorrect ("hallucinations") and/or inappropriate or unacceptable results (Alkaissi and McFarlane, 2023; Xu et al., 2024; Waldo and Boussard, 2024), but, after their evolution in our hands, some of these problems could be solved (Tonmoy et al., 2024). The real challenge lies not in unresolved issues or cumbersome functionalities, but in determining how and to what extent researchers should employ chatbots to create value for science and society. This requires critical thinking and understanding the potential hazards for science, beyond merely calculating costs and benefits.

We emphasize the need for social science publishers to be aware of the risks associated with data, algorithms, and hypotheses. This area requires special attention, as hypotheses are often based on conceptual frameworks that take a long time to confirm, rather than solely on experimental observations.

To meet this objective, this paper is divided into eight sections. The second section briefly presents the methodological approach. The third section realizes a comprehensive and multidisciplinary review of the literature to investigate the nature and context of advanced LLM-chatbots in scientific research, describing both benefits and concerns. We make an aside to consider the case of social sciences, where literature detects a certain naivety when judging algorithms and data. The fourth section explores Plato's myth of Theuth and Thamus, obtaining critical resources for the formulation of a critical analysis focusing on scientific authorship and hybrid publications. Combining both philosophical and technological views, the fifth section critically analyzes what is an author? The sixth section analyzes the hybrid articles, closing with a formulation of requirements that align ChatGPT with the values of science. The seventh section shortly analyzes the specificity of the social sciences. Finally, we present our conclusions and describe the limitations of this study. An additional appendix includes three tables outlining a wide range of concerns: the limitations and weaknesses, the advantages of using LLM-chatbots for research as discussed in recent literature, and the analysis of the positions of leading journals and major publishers.

## Methods: methodological framework for a critical analysis

Given the research question, in formulating this article, we have incorporated several key scientific methods as a methodological framework. It is focused on two analyses. On the one hand, we employ a comprehensive and multidisciplinary literature review, analyzing opinions from various authors, updated with the latest advances. This review does not merely aim to effectively present relevant information on the research topic, which would allow researchers to become familiar with the relevant concepts and ideas. Our extensive review attempts to help scientists build background knowledge so that they can bring an opinion. To achieve a holistic understanding of the topic, the article conducts a multidisciplinary exploration.

On the other hand, we tackle an analogical critical thinking based on the analogical discussion on Plato's criticisms of writing. Literature confirms its relevance in science and technology contexts (Cf. Kosar, 2024, p. 8), due to its capacity for (a) making sense of technological systems for lay people, (b) facilitating epistemological reflection on chatbots from a novel perspective, and (c) understanding and assessing emerging technologies in dialogue settings such as the scientific communities (Cf. Bucchi, M. and Trench, 2008). In contrast to scenarios of science fiction and permitting to know how society responded to previous technologies and the consequences these responses entailed, analogical thinking can enhance anticipatory capacities about governance processes.

We use analogous discussion on Plato's criticisms of writing, as a means of providing insight and approaches to the use of ChatGPT on research. With Clements (2022), we observe analogy between critics in the chatbots age and those that Plato observed, including the misuse or abuse of published information, the lack of context and, in the end, the role of memory. The Platonic distinction writing/logos; appearance/reality; verisimilitude/truth (Cf. Derrida, 1981, p. 181) permit us, arguing that, as a pharmakon, LLM-Chatbots could be a poison but also a medicine, which can be associated with better knowledge.

This methodological approach allows us to propose our hypothesis: An LLM-chatbot should not be considered an author of a paper, but it can serve as a tool that assists in the writing process, fitting into the hybrid article model. This hypothesis is explored in the following four sections, guided by the methodological framework and the critical analysis conducted by the authors.

### LLM-chatbots and science: a comprehensive analysis and literature review

Technically, LLM-chatbots are designed to understand and generate human-like text based on the input they receive. They integrate multiple technologies such as deep learning, unsupervised learning, instruction fine-tuning, multi-task learning, in-context learning and reinforcement learning (Cf. Wolfram, 2023). Its first pillar was the Transformer architecture, presented in the paper “Attention is all you need” by Google researchers (Vaswani et al., 2017), in which they claimed to have reached a new state of the art in generative AI. The experts agreed. In a macro survey conducted that year (Grace et al., 2018), they dated the years in which machine performance would surpass that of humans: translation (2024), high school essay writing (2026), truck driving (2027), writing best-sellers (2049) or general surgery (2053).

Based on a Transformer framework via unsupervised learning, in 2018, OpenAI, a non-profit organization with open-source software and publicly available advanced tools, presented the second pillar: the Generative Pre-trained Transformer model (GPT), a massive deep learning language model. The tool was trained on Microsoft Azure’s AI supercomputer using 45 terabytes of text from Common Crawl, WebText2, books and Wikipedia (Brown et al., 2020). In the initial stage of generative pretraining, the algorithm assigns numerical values to words. Words that frequently appear together are associated within an “embedding,” a multidimensional representation of their relationships. Chatbots utilize these embeddings to generalize concepts by leveraging the relationships between words in the training data. During this phase, weights are adjusted to minimize loss, which is the difference between the desired and actual results. The outcome is a tool capable of generating functional “knowledge” across various contexts, which is then refined through reinforcement learning with human feedback, involving supervised fine-tuning on specific tasks relevant to the user.

ChatGPT has been iteratively updated from GPT-1 to GPT-4 (2018–2024) building through an easy-to-use web interface and capable of generating conversational-mode responses similar to those of an “expert” human to queries formulated in natural language, based on the input provided. GPT-4, with accepting image and text inputs and emitting text outputs, exhibits human-level performance on various professional and academic benchmarks (Wu et al., 2023). If the GPT parameters were increased tenfold in GPT-2 (one hundred times in GPT-3), the GPT-4 version has 500 times more parameters than GPT-3, approaching the number of neuronal connections in the human brain and proving to be functional for a wide range of applications (cf. Rudolph Tan (2023)). OpenAI recently introduced a new series, designed to spend more time reasoning through problems before responding, which helps them solve more complex tasks in areas such as science, coding, and mathematics (Cf. Zhong et al., 2024). For instance, the GTP o3 LLM model represents a significant advancement in AI capabilities, particularly in complex tasks such as coding, mathematics, and science. It demonstrates improved performance and reasoning skills, making it a valuable tool for various applications. The model achieved a breakthrough high score of 87.5% on the ARC-AGI benchmark, which is designed to

test genuine intelligence (human performance is benchmarked at 85 percent).

Along the way, OpenAI became a for-profit company, giving Microsoft exclusive access to its GPT-3 language model. The open became partially closed (the architectures for GPT-3 and GPT-4 have not been published), and the democratization of AI became doubtful. The issue is not trivial because, with huge economic and environmental costs in terms of energy and water (Hao, 2020), innovation is being developed by only a small number of Big Tech players (United Nations. AI Advisory Body (2024)).

While some authors emphasize an LLM-chatbot does not think (Dehouche, 2021), others emphasize it can pass public exams in medicine (Kung et al., 2023) or law (Katz et al., 2024), create poetry indistinguishable from human poetry (Köbis and Mossink, 2021) or display social intelligence at the level of licensed psychologists and doctors (Sufyan et al., 2024). Table 2 (appendix) summarizes main advantages for researchers, classified into efficiency improvement, quality improvement and cost saving in terms of time, effort, and money. However, risks and concerns described in Table 1 (appendix) are also present in scientific research, including incorrect and fictitious answers; fabricated or biased literature; inaccuracies in the code; limited logical reasoning skills; plagiarism; overconfidence; copyright, security or privacy violations, etc. To provide more clarity, we should have holistic performance metrics, considering both pros and cons. Currently, we lack reliable metrics.

However, a positive balance does not guarantee that its use will improve the quality of our scientific activity and our publications. We must remember chatbots are nondeterministic by nature. This is because they predict the probability of a word given the context, represented by a sample of words, which produces randomness (Ouyang et al., 2024). As a result, identical instructions/prompts can yield different responses to separate requests, and to check for hallucinations is not possible (Alkaissi and McFarlane, 2023; Xu et al., 2024), which affects the reliability and reproducibility of research. Therefore, the potential of LLM-chatbots to improve (or depreciate) research must be carefully analyzed and constantly monitored.

The main examen is the evaluation of the hard changes in individual and collective memory. Let’s not fool ourselves. As an algorithmic crystallization of memory, ChatGPT can generate human-like texts from a simple message without further human intervention. But, far from being an AI agent that (re)distributes and (re)activates memories, ChatGPT is a kind of active guardian of memory, effectively deciding which sources of information and which interpretations of the collective past gain more visibility and are accessible to the platform users, and therefore shaping how this past is remembered or a decline in memory retention for humans (Bai et al., 2023).

As an entity responsible for the organization and retrieval of information in response to human queries, and given its non-deterministic nature, ChatGPT can change our scientific practices. In fact, our truth discovery becomes a challenge, which integrates multi-source noisy information by estimating the reliability of each source. For instance, the algorithmic initialization of the truths can be obtained using voting/averaging approaches, using weighted vote for categorical data and weighted median for continuous data to update truths (Cf. Li et al., 2016). Statistic could supplant truth, as, in the past, writing attempted to supplant logos. To expose this factor, we will turn to Plato.

### Plato and ChatGPT: philosophical roots of the criticism

In ‘The Phaedrus’ (274e7–275b2), Plato relates the myth of the Egyptian King Thamus and the god Theuth, inventor of writing. Animated by the desire of popularizing his invention, Theuth

presents it to Thamus, as “an elixir of memory and wisdom” (274 e), the best cure (pharmakon) for oblivion and ignorance. But the king argues that it is not an elixir of memory (mneme), but of reminding (hypomnesis) and can be a poison. While writing allows for a vast accumulation of knowledge, it can affect both (a) our ability to remember (it makes us forgetful), and (b) the integrity and power of the individual to interact with knowledge and reality.

The Thamus’ critique is developed in three steps.

The first one is the criticism of technical manuals, which, according to Thamus, is needed but not sufficient condition for producing “results that are clear or certain”. Despite a set of true and useful theorems would be contained in a manual of mathematics, the owner of this manual is not a mathematician. A mathematician must be in possession of a set of abilities and skills, which permit to correctly understand propositions, to articulate the relevant true propositions to some concrete problem; to explain why they are relevant and true; to know how to justify decisions to others, as well as be able to teach others to become accomplished practitioners like himself. Those capabilities cannot be obtained in technical manuals (Cf. Rabbås, 2010, p. 32-33). Finally, the accumulation of data, writings or technical manuals, which can be deposited in external libraries or technological artifacts, by itself do not generate knowledge, and their memorization does not form a living memory. It is a “pharmakon-poison,” which connects with hypomnesic memory and makes us forgetful, while providing us with the illusion of wisdom.

The second step is developed across the analogy with painting. (275d4– e5). “Writing, Phaedrus... is very like painting; for the creatures of painting stand like living beings, but if one asks them a question, they preserve a solemn silence. And so it is with written words; you might think they spoke as if they had intelligence, but if you question them... they always say only one and the same thing” (275 de). According to Plato, the written word is dead because it cannot enter in a dialogue, adjusting to each person through a dialectical exchange, neither to specific audiences and situations. Writing is a “pharmakon-poison” (275e) for the memory because (a) it cannot answer the questions it raises; (b) it can reach people to whom it is not addressed, and it can be misinterpreted or misinterpreted; (c) when questioned, it cannot defend itself (Wieland, 1976; Barceló-Aspeitia and González-Varela, 2023).

In the third step, Thamus focuses on knowledge “in the soul of the student” (276a1-b1). Only the text capable of touching the soul of the student achieves the goal, which is to provide the resources that the student needs to support his practice toward truth. We are talking about a certain tacit knowledge (Polanyi, 1958), which is difficult to write or visualize, but which can make the writing alive for the student. In this way, the crucial difference between speaking in person and writing for others is not linguistic structure but efficacy as residing in truthfulness (Hyland, 1968, p. 39; Cf. Heidegger, 1931/32) and in readers able to be cojoined with true knowledge. If wisdom is not a manual but includes a set of abilities and skills to do things and search truth, and these abilities and skills are properties of the soul, not merely part of the content of his or her mind, then, when the discourse is successful planted, and, with time rooted and grow, the results change the learner (276e4–277a4). The scientist needs to remember occasions when she/he exercised his knowledge, which can be used as objects of comparison to instruct her/himself (or others) about what was right or best in that situation. On the opposite, the text without soul is as petrified, static and passive writing, dead words, being easy to remember, is only susceptible to identical repetition, without criticism or reinterpretation, dogmatic. However, since

their appearance does not sufficiently evidence their instrumental character, they can give us the illusion of embodying knowledge.

After criticizing writing, Plato wrote dialogues. Why? According to literature (Cf. Clements, 2022; Rabbås, 2010; Staehler, 2013; Wieland, 1982), Plato’s criticism of writing is not absolute, and, therefore, to interpreting writing as a pharmakon-remedy is possible.

The most obvious benefit of writing is to have a certain generative power of its own, which overcomes the limitation imposed by time and space. This power could preserve and defend, for example, Socrates’s memory as well as the true concepts to those who are unable to participate in dialectic (Allen, 2011, p.61). Another benefit is that writing text is always “on record” and accessible to scrutiny. As Derrida (1981, p. 113) shows, this element is present in Plato’s perception of the need for the law to be written. In summary, the reader could use writing as remedy, recognizing certain conditions.

These conditions are described at the end of the Phaedrus. Socrates argues that good discourse has two features: their seeds (a) must have been planted in minds able to philosophy, and (b) must be cojoined with true knowledge (Cf. Allen, 2011 p. 60). When Phaedrus, to the detriment of philosophy, shows its preference for rhetoric which is cleverer, persuasive and requires less effort, Socrates argues that he never will be a good rhetorician if he does not become a philosopher. The aspiration of permanence and vividness is not only related to time and space, but also to the common (republic) good, which requires being based on truth (Zuckert, 2009, p. 323). Thus, the first crucial difference between speaking in person and writing for others is not linguistic structure but efficacy as residing in truthfulness (Hyland, 1968, p. 39; Cf. Wrathall, 2004). The second condition is in the abilities and soul of the reader, who must be intimately fused with genuine knowledge. In the writing law, this condition is satisfied because “if the writer is the legislator, the judge is its reader” (Derrida, 1981, p. 113).

A similar situation appears in science. Scientists are not only knowledge-gatherers or candidates to authorships, but also as societal stakeholders and as members of a scientific community, which protects a legacy (Mills and Sætra, 2024). As Polanyi (1962, p. 7) underlines, the activities of scientists take place around the published results of other scientists into the current professional standards of science, which must impose a framework of discipline and at the same time encourage rebellion against it. The authority of science over the lay public is held by a multitude of individuals, who controlling the access of candidates until they reach a degree that allows them to assume the responsibility of joining that *republic*. In the end, writing as “Pharmakon-remedy” is embedded in the legacy, which can be collected, interpreted, criticized or discarded, a living memory correlated with the anamnestic memory, with which one converses, renews or adds new meanings, and is therefore non-dogmatic. While the “Pharmakon-poison” can equate appearance and truth causing the degradation of science (Hendrycks et al., 2023; Loos and Radicke, 2024), the “Pharmakon-remedy”, closely connected with traditions and hermeneutic filters, helps to lead the person to the truth in perpetual tension between the temporal and the eternal, the real and the ideal, the finite and the infinite.

Therefore, the question here is if texts produced by LLM-chatbot must be considered dead letters, or a legacy with a soul. If, finally, a platonic dialogue between humans and chatbots is possible, chatbot should be considered an author; in other case, no. This question is important in practice because, despite humans, who are highly social beings, tend to respond to complex machines as if they were as “social” as we are (Sætra, 2020; cf. Foss and Saebi, 2017). This will allow us to steer the discussion



towards our hypothesis of the use of LLM-Chatbots for hybrid articles.

### Authorship. What is an author?

In “What is an author?” (1969), Foucault wonders whether a scientific article is not simply something written by a scientist and recalls that in the Middle Ages texts only had truth value if they were signed by their author. Two decades later, the same question is focused on a computer (Samuelson, 1990), and fifty years later, Ginsburg and Budiardjo (2019) reformulate the question focusing the most technologically advanced machines. They suggest machines are little more than faithful agents of the humans who design or use them, supporting the “right” question is how to evaluate the authorial claims of the humans involved in either preparing or using the machines that “create”.

Nowadays, authorship is described since scientific standards (cf. ICMJE (2020)) which require substantial contributions to: a) the conception and design of the article; b) the collection, analysis and interpretation of data; c) drafting or revising the article critically; and d) final approval of the version to be published.

Although some argue that the idea of “substantial contribution” is too subjective and that there are important non-intellectual contributions that should not be unfairly neglected, chatbots cannot approve the final version, nor be responsible for all aspects of the work to ensure accuracy and completeness (Van Woudenberg et al., 2024); nor understand a conflict-of-interest statement, hold copyright or have independent affiliation (Flanagan et al., 2023; Teixeira da Silva and Tsigaris, 2023).

According with Plato, there are other aspects which must be considered as author:

Firstly, “*the creation itself*”. Beside intellectual property and the entitled to own, to grant permission to reproduce the final version, to appropriate the revenues and the academic and social prestige that publications generate (Efthymoulou, 2008), authorship implies moral identity, whereby the researcher must recognize him/herself as a creative author (Epstein et al., 2023) and be competent to make judgments in a scientific community and take responsibility for them (Nannini, 2023; Scerbo, 2023). Authors are called to defend their work if it is challenged by scientific community, editing and critically review their contribution to science. Moreover, only an author can challenge legacy and the rules of the scientific method. Even if, in the future, chatbot could be conceptualized as a legal person to whom we attribute taxes, criminal liability, etc., moral identity is needed: an author must be able to a dialectic challenge with community, to claim misinterpreted or misunderstood; and, when challenged, to defend.

Not everyone agrees. While countries such as Korea define a “work” as “a creation that expresses the thoughts or feelings of a human being” and an “author” as “a person who creates a work”, preventing chatbots authorship, India or Canada have already accepted applications for copyright registration with computer applications (Lee, 2023). Moreover, some authors reclaim to describing new standards and guidelines. For example, Polonsky and Rotman (2023) point to (a) articles written by 100 or more authors, with each author’s contribution blurred; (b) articles signed by authors who died before publication; or (c) organizational authors with no identifiable individuals, with each author’s scope of formal approval and responsibility like that of a chatbot. They also indicate that protocols such as Vancouver recognize that some authors have specialized roles and journals provide a list of author roles in which ChatGPT could be included. On the other hand, they stress that chatbots are becoming so sophisticated so quickly that they will be hard to ignore.

Secondly, “*the responsibility*”. As chatbots lie beyond the boundaries where moral values apply, publishers can only

demand from named individuals the basic requirements of honesty, ethics and integrity that are essential for journals to advance science. Without proper attribution of authorship, it is unclear who is responsible for the content of the manuscript, the review, the veracity of the sources, the unduplicated publication, or the inclusion of all relevant points of view. (Ginsburg & Budiardjo, 2019). Breaches of codes of conduct and ethical behavior in scientific publishing (certainly more common in the social sciences according to (Xie et al., 2021)) undermine the integrity of the whole system, damaging both the reputation of journals and the value of science. Whether through misconduct (falsification, fabrication, plagiarism) or questionable research practices (cf. Xie et al., 2021), only individuals can be held accountable. Editors must therefore certify that authors are who they claim to be and that they have participated sufficiently in the study and critical analysis to be able to publicly guarantee its content and adherence to the rules of the scientific method and to guarantee the intellectual property of the ideas disclosed.

Third, “*the membership*”. The scientist is always, first of all, a reader/student, who receives the written past, appropriates it and unravels for himself the meaning it suggests to him. Before this appropriation, he must ask himself where the information comes from, and how much truth it contains. Because the context in which it was written differs from that of researcher, author requires clarifications to illuminate the truth of the message (Staehler, 2013, p. 87). Some of these come from knowing the author of the quotes because, through his career, we can get closer to his thoughts and, as Foucault pointed out, evaluate their reliability. As for the choice of one method or another, this “truth” is even more diffuse. Placing a text within a particular topic legacy is a quality of an expert specialist (Gupta, 2024). An expert is someone who possesses extensive and authoritative knowledge in a particular area that is not possessed by most individuals in a community, being candidate to trust (cf. Croce, 2019). Because a chatbot cannot distinguish between “accurate and false information” (Bhattacharyya et al., 2023, p. 6), the success criterion of an LLM-chatbot is verisimilitude, not veracity, being not candidate to trust. Only those who have a thorough knowledge of the scientific literature and the various theories on a topic can place hypotheses in a scientific context, distinguish meaningful connections from seemingly irrelevant information, make accurate deductions, and apply this knowledge to specific situations. A chatbot does not have the capacity to perform this task, which inevitably requires a degree of human judgment.

The three conditions are resolved in scientific publication through the dialectic between authors and referees, and among authors into the scientific community. This dialectic would be the remedy, or the “antidote” to poison (Cf. Horbach and Halffman, 2020).

A special emphasis requires the role of editors and reviewers. If a journal is the dynamic outcome of a collective and interdisciplinary effort among scientists to create an ethical, open and shared framework for conducting scientific research. The role of the editor is to ensure that the framework remains open, thereby facilitating critical analysis and the search for truth inherent in any scientific endeavor. By validating any contemporary research in a competent journal, editors possess a valuable and human cooperative tool to disagree and move forward. An editor who is responsible for aligning algorithmic thinking that can benefit certain stages of the research program with ethical values in scientific production must have a clear understanding of this issue.

Some recent works support the use of LLMs as expert reviewers (Lu et al., 2024), permitting fast and lower-cost evaluations and decreasing the number of needed expert reviewers (Cf. Baek et al., 2024; Lu et al., 2024). Analyzing the use of LLM-chatbots for peer

review, Donker (2023) finds good summaries, description of the main aim and conclusions, comments on style and some constructive criticism, but also hallucinations, incorrect or uncorrelated answers to the text or not appropriate to the data, recommendation of non-existent articles and intellectual property problems. He concludes that reviewers should refrain from using chatbots tools and be trained to detect when and how a chatbot has been used correctly.

According to the Aristotelian tradition, the crucial function of a journal is not epistemic (i.e., facilitating “understanding of the underlying principles of the world,” where a chatbot, despite its numerous shortcomings, can be of great help), but to assist in “scientific performances in the world,” which is a matter of phronesis, or practical wisdom (Cf. Lee, 2005). This is a dialogical relationship between thought and action (Atkin, 2007, p. 68), where a chatbot is not effective. Considering only the epistemic function of AI, it may partially displace human (Cohen, 2013, p. 1926). However, when it comes to practical wisdom, an artificial science without critical thinking (Cohen, 2013, p. 2921) and without room for altruism, empathy, and other essential elements of human flourishing does not be accepted. We understand that neither the accumulation of data nor speed can override academic analysis and the critical search for truth.

But if the editor reviewing the text wants to be sure and asks why a particular work by a particular author is cited or not, or why a particular method is used, the editor will not get an answer: a chatbot is not able to explain or justify its choices (cf. Chavanayarn, 2023; Thorp and Vinson, 2023; Stokel-Walker, 2023).

In summary, while ChatGPT passes the Turing test and convinces people that its answers are human (Guo et al., 2023), produces academic texts that are difficult to distinguish from those written by humans (Campbell et al., 2022), generates high-quality, evidence-based research questions (Lahat et al., 2023), and writes a convincing manuscript in less than an hour (Elbadawi et al., 2024), it does not possess the ability to distinguish between accurate and false information. The responsibility for the creation and verification of information inevitably requires human judgment. Therefore, its outputs are not considered scientific documents, and it cannot be regarded as a scientific author of a text or review.

### Hybrid articles: the path to help with LLM-Chatbots

A scientist working with a chatbot produces a polyphonic response (Kjeldsen, 2024), which contains the voice of the scientist in a choral scientific music, and another (or several) voice of the machine. Last one, clearly distinct from the first, transmits (at best) no-contextualized messages attributable to other human beings who are not subject to current scientific authority (Sætra, 2020) and mission.

Socrates distinguishes between the writer who is merely “a maker of speeches”, whose mission is to persuade the audience (277e5–278b4), and the writer who “has composed his writings with knowledge of the truth” (276d–78e). Criticizing the first, Socrates signals that, without truth, students will become wise in their own opinion (doxa) instead of wise (sophos) (275a–b), opposing the probable (eikos) (that “happens to spring up in the many, through likeness versus the truth”) to the truthfulness, and finally the “public” opinion (doxa) to knowledge (episteme) of scientists (246A–49D). While scientists contribute episteme to the polyphonic sound, chatbot contributes doxa based on eikos.

For poorly prepared scientists or those outside of science, the outputs of a chatbot will create a kind of illusory knowledge. If it were false or erroneous, they will not identify it and will report it as true. Learning without teachers (Phaedrus 275a) is equivalent

to doing science without scientific contrast: it can produce the illusion of wisdom in the ignorant, but it cannot expand knowledge.

However, a good scientist is able to produce a harmonic song with human and machine songs. We have two practical arguments and one theoretical one for accepting hybrid articles produced with the help of the LLM-chatbot.

The first one is the conviction that AI deployment is unstoppable and, therefore, a policy of non-co-production unfeasible in the long term. Recent studies reveal that two-thirds of authors are utilizing chatbots without disclosing this fact (cf. Khalifa and Ibrahim (2024); Lund and Naheem, 2024; Alkaisi and McFarlane, 2023), creating fairness and legality problems for publishers. Assuming that authors are being honest when stating that they do not use them or when not stating that they use them is an unviable option.

The second one is that currently it was extraordinarily challenging to detect what articles have been produced with an LLM-chatbots (Gao et al., 2022). Despite to essays (cf. Van Dis et al., 2023; Köbis and Mossink, 2021), there is not motivation to invest in this function, appearing as unprovable to have soon have effective detectors. In their own self-interest and given the irreversibility of the publication decision and its implications (Stahl et al., 2022), publishers have strong incentives to reject (Gordijn and Have, 2023) or to reclaim transparency.

Finally, on the theoretical argument, there is a more compelling reason. As human agents, researchers intend both the end and the means, praxis and poesis. On the one hand, they intend the improvement of the science itself, with centrality on persons and truth. On the other, they intend its practical development, measured, for example, in investigation’s results (i.e., papers, patents, etc.), which are influenced by something external (regulations, ethical guides, journal requirements, etc.) elements and are executed only to the extent necessary to achieve. This last has embedded some poetic activity. Despite, by nature, chatbots are unable to praxis, which implies logos and the creation of a shared reality, chatbots can emulate many of the poetic elements embedded in scientific activity. This could suppose a major but risky advancement in specific scientific tasks, which must be clearly described. Here, it is crucial to develop a protocol, which should clearly describe the whole analysis process, which at the same time, can be useful to the realm of creativity and efficiency.

In summary, we advocate avoiding restrictions, but we call for a unifying comprehensive ethical framework for their use, in terms of advertising transparency, accountability or responsibility on the use of chatbots, and authorship, as well as for publishers/editors.

*Advertising transparency* recommendations range from those asking for basic information to be attached in the acknowledgments section, to those asking for strong details of (a) its specifications (name, version, model and source), (b) how it has been used and (c) why, and categorizing it in the “Methodology” sections. Based on the literature and our own judgment, we agree with the stronger version. With (a) because different chatbot models exhibit varying types and degrees of biases, employ diverse workarounds, and possess distinct characteristics (cf. Sufyan et al., 2024), with (b) because a chatbot can be used in many ways and for many purposes, with different effects (cf. Zhuo et al., 2023a) and with (c) because it is necessary to know who is “playing that polyphonic tune” to maintain trust in science. We agree with the proposal of Jenkins and Lin (2023), who ask journals to provide procedures to give transparency about how and what content has been generated with AI and, mainly, for what purpose.

We also require authors to taking individual and/or collective *formal responsibility* for all contributions made using chatbots,

including the accuracy and proper attribution of all cited material, and to search for and cite omitted sources (cf. Taylor, 2024). This is in order to avoid liability gaps, which in the case of scientific articles is usually a forward-looking liability that is more focused on the actor than on the act itself (cf. Hedlund, 2022).

Third, *on the authorship*, we add an additional condition to “satisfy Plato”. Since editors are responsible for ensuring that authors answer the questions raised by their text and that it is correctly interpreted, we recommend authors defend their text and describe what their contribution to science has been.

### The particular focus and specificity of the social sciences

Some authors argue that chatbots could help especially social sciences, improving the objectivity of the scientific field to examine practicing scientists’ views of nature of science and explore possible relationships between these views and science social context. This offering a less arbitrary epistemology, purging the contamination of individual reasoning and improving the accuracy of results both at the initiation stage of research and in data analysis, reducing the potential for human error and achieving better reproducibility (Burger et al., 2023).

However, other authors warn against a certain naivety in social scientists who identify themselves as “end users”, discussing the many limitations of LLM-chatbots. Bails examines how bias in the data used to train these tools can negatively impact social sciences research—as well as a range of other challenges related to ethics, replication, environmental impact, and the proliferation of low-quality research (Bail, 2024).

A LLM-Chatbot is targeted at technologically literate users. To use it effectively, social researchers must be willing to fully embrace that technology, which often involves collaborating with experts, fostering interdisciplinary collaboration, and investing in training and development programs. Faced with a technology that causes radical changes, a scientist must not only know how to use it, but also gain the ability to understand and critique its results and use and to manage correctly the data. However, many social scientists are unaware of its risks. We want to draw attention to two gaps that a non-expert social scientist might face managing data.

**The first gap stems from its theoretical burden.** Data are not neutral “facts”, value-free observations, but imply goals, means, constraints and selection criteria (Sarasvathy, 2021, p. 249). As masterfully Coveney et al. (2016) put it, big data need big theory too. As S. Javed et al. (2021) pointed out, “letting nature speak for itself” is not the same as “letting the model speak for itself”, as the latter is based on a priori assumptions and theories.

**The second gap originates from the data selected for training.** Fossil fuels may be polluting, but they produce the same result (heat) as clean energy. This is not the case here: the performance of LLM-Chatbots depends directly on the quality and relevance of the data on which it is trained (cf. Dwivedi et al., 2023b). There is evidence that tools trained on problematic data reproduce problematic associations (Grassini, 2023) in the form of stereotypes (Font and Costa-Jussa, 2019), abusive language and hate speech (Kennedy et al., 2018), and all kinds of biases of gender (Tschopp, Salam, 2023), race (Buolamwini and Gebru, 2018), sexual orientation or religion (Kurita et al., 2019), etc.

To ensure that technology is harnessed for the benefit of humanity and scientific progress without impeding valuable research, social science editors must possess skills in data analysis, algorithm and code design, and training in the latest technological advancements. Given the scarcity of individuals who are both technologically proficient and ethically informed, it is crucial

for journals to form interdisciplinary teams and seek advice from independent experts. This approach will help navigate the complexities of the new technological landscape effectively.

The first gap stems from its theoretical burden, which requires a deep understanding of the underlying principles and frameworks guiding technological advancements. The second gap originates from the data selected for training, highlighting the importance of using diverse and representative datasets to ensure fairness and accuracy in technological applications. By addressing these gaps, social science editors can better manage the integration of technology in research and publication, ultimately fostering a more ethical and progressive scientific community.

Therefore, in social science, the application of ChatGPT must be accompanied by a meticulous analysis of the data and its inherent characteristics. Equally important is the human contribution in crafting the conceptual framework of the paper, developing hypotheses, and establishing expectations for practical scenarios. As we have mentioned, the conceptual frameworks and hypotheses may take long time to confirm, that is, several years to validate. Therefore, it is imprudent to rely solely on the probabilistic nature of ChatGPT, which is grounded in text and learned knowledge, to generate hypotheses or conceptualize theories in the social sciences. This is in addition to everything indicated for hybrid articles, which is generic in nature in any scientific setting.

This circumstance makes us pay special attention to the particular case of the social sciences. In any case and scenario, any hypothesis for the future, taking a long time to confirm, in any field, must be proposed and validated by a human being. Judea Pearl’s theory on imagination highlights a current challenge that remains beyond the capabilities of LLM-Chatbots. Pearl’s “Ladder of Causation” (Pearl and Mackenzie, 2018) emphasizes that while AI can observe and perform tasks, the ability to imagine and hypothesize about future scenarios is uniquely human. This imaginative capability is crucial for developing and validating hypotheses, particularly in the social sciences, where conceptual frameworks often require extensive time for confirmation. Thus, the human role in proposing and validating hypotheses remains indispensable, underscoring the limitations of AI in this domain.

### Concluding remarks, limitations, and future directions

LLM-chatbot technologies are increasingly used in the domain of investigation and scientific publication (Giray et al., 2024). The fact that communication between scientists involves both researchers and chatbot (which can produce new content, without the immediate supervision of humans) is something unique, unknown and with very profound implications (Dwivedi et al., 2023b). The huge benefits in terms of reducing efforts and costs as well as improve efficiency are measured against concerns, limitations and weaknesses, an evaluation still pending (Grassini, 2023; Stokel-Walker, 2023; Zhou et al., 2023).

LLM-chatbot passes the Turing test and makes people believe that its answers are human (Guo et al., 2023); produces academic texts that are difficult to distinguish from those produced by humans (Campbell et al., 2022); generates high-quality, evidence-based research questions (Lahat et al., 2023); or writes a convincing manuscript in less than an hour (Elbadawi et al., 2024), but it presents a wide range of concerns, limitations, and weaknesses (Stokel-Walker, 2023).

However, in our opinion, the main question is not about the balance between pros and cons, but if the integration of LLM-based chatbots in scientific writing improves or erodes scientific thought the quality and integrity of published research (Nazarovets and Teixeira da Silva, 2024). The scientific community is divided. While many journals and major publishers (such as



Springer-Nature, Elsevier, Lancet or Taylor and Francis) have restricted its use, other publishers and journals consider that the prohibition is unfeasible and simply ask that its use be reported (Sallam, 2023). Debate focuses on two main questions highlighted in the introduction: the possible authorship of LLM-chatbots, and the acceptance of hybrid articles (using LLM-chatbots) (Thorp and Vinson, 2023; Lund and Naheem, 2024; Nazarovets and Teixeira da Silva, 2024).

To contribute to the debate, we turn to Plato’s criticisms of writing, contained in the Phaedrus, an analogy which present many connections with chatbots. Very recently, it has been employed in the analysis of some peculiar aspects (Deptula et al., 2024; Loos and Radicke, 2024) mainly focusing on education area and from instrumental analysis (Aylsworth & Castro, 2024; Bingham, 2024; Kitzinger, 2024; Rahman et al., 2023). However, research area and the analysis of science itself have been neglected. We explore Plato’s myth of Theuth & Thamus as a means of providing insight and approaches to qualify the use of LLM-chatbots on research.

From an interdisciplinary perspective, our conclusions are as follows:

Firstly, it is necessary and urgent to establish a unified and comprehensive ethical framework for the use of LLM-chatbots in research. Editors, most of whom could suffer from a certain technical illiteracy, need that the publishers, communities, technical experts and regulatory authorities sit together and lay down recommendations and good practices about its use, an uncompleted mission jet.

Secondly, an LLM-chatbot cannot be considered as authors/co-authors or be cited as authors/co-author in a scientific context. Not only because they do not comply with the present editorial standards, but because they are unable to escape Platonic criticism of writing.

Third, hybrid articles may be acceptable under certain strong conditions. We offer some preliminary recommendations to help guideline: *Advertising transparency*, *formal responsibility* for all contributions made using chatbots, and *on the authorship*, we add an additional condition to “satisfy Plato”, *the creation itself and the membership as author*, requiring clarifications to illuminate

the truth of the message. A chatbot does not have the capacity to perform this task, which inevitably requires a degree of human judgment.

This paper provides a platform for debate, analysis, and reflection on the use of chatbots in the final writing of scientific papers. It contributes to the ongoing discussion within the scientific community, as cited in this paper. We hope it adds to the scientific legacy. As with any study, we acknowledge certain limitations, which also present opportunities for new avenues of exploration. We discuss some of these limitations.

Firstly, and due to the recentness, dynamic and novelty of the technology, the findings of this research seem to raise more questions than they provide answers. Advances in the reasoning processes of the new LLM-Chatbots models are ongoing, although the limitation posed by Judea Pearl remains on the horizon, imagine.

Second, very recently, OpenAI has launched the mentioned OpenAI o3, known as Strawberry, the third model in a new series designed to perform complex reasoning tasks, allowing it to generate more complex answers. precise and thoughtful. Despite we think our conclusions are also valid, this paper does not analyze this innovation. It would be interesting to analyze the use of the LLM-chatbots in the generalization of theories and hypotheses based on past data they have learned and on the reasoning capacity of the new models (Strawberry, Gemini, Llama family, Claude, etc.) (Movva et al., 2024). Advances in the complex reasoning processes of the new LLM-Chatbots models are ongoing.

Thirdly, an analysis of the impact of chatbots on the social sciences, and the responsibility question would require much more space than these pages allow, and a space-temporal analysis.

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Appendix  
Table 1

Table 1 Key concerns associated with generative AI.	
Author rigths and plagiarism	Cotton et al., 2023; Dehouche, 2021; Flanagan et al., 2023; Floridi, 2019; Gao et al., 2022; Hagendorff, 2020; Hutson, 2021; Xie et al., 2021; Susnjak, 2022
Biases, discrimination, prejudice, racism, sexism, stereotypes, injustice, inequity, ideological biases, marginalization of minorities	Basta et al., 2019; Benthall & Haynes, 2019; Binns, 2018; Buolamwini & Gebru, 2018; Dwivedi et al., 2023b; Font & Costa-Jussa, 2019; Fraser et al., 2023; Hutson, 2021; Huang & Chen, 2019; Kurita et al., 2019; Maarten et al., 2020; Tschopp & Salam, 2023; Zhao et al., 2019; Wang et al., 2023.
Dehumanization, loss of collective human identity	Chen, 2023; Leung et al., 2021; Menderlsohn et al., 2020; Solaiman et al., 2023; Anderljung et al., 2023; Hendrycks et al., 2023.
Digital gap and technical Illiteracy	Bozkurt & Sharma, 2023; Hagendorff, 2024; Leung et al., 2021.
Hallucinations or similar phenomena	Azamfirei et al., 2023; Borji, 2023; Li, 2023; Dwivedi et al., 2023b; Ji et al., 2023; Susarla et al., 2023; Shelby et al., 2023; Ray, 2023; Weidinger et al., 2023.
Harmful or inappropriate content: (offensive, pornographic, toxic or violent)	Basta et al., 2019; Carlini et al., 2020; Grassini, 2023; Gehman et al., 2020; Hendrycks et al., 2023; Illia et al., 2023; Kendall & da Silva, 2024; Kennedy et al., 2018; Mozes et al., 2023; Shelby et al., 2023; Zhuo et al., 2023a.
Incomplete, inaccurate, incorrect or false information, confident justifications, fabricated references, disinformation, advertisements, deepfakes	Azaria et al., 2023; Aydin & Karaarslan, 2023; Basta et al., 2019; Bender et al., 2021; Donker, 2023; Khan et al., 2023; Megahed et al., 2024; Wang et al., 2023; Zhan et al., 2023; Zhao et al., 2019.
Maleficence	Brundage et al., 2018; D'Alessandro et al., 2024; Hagendorff, 2023; Hagendorff, 2024; Hendrycks et al., 2023; Mozes et al., 2023; Turchin & Denkenberger, 2020; Wang, et al., 2023; Weidinger et al., 2023; Zhan et al., 2023.
Training data quality and codification’s errors	Azaria, 2023; Azaria et al., 2023; Dwivedi et al., 2023b; Su & Yang, 2023.



**Table 1 (continued)**

Mass manipulations, hate speech, social punctuation techniques, political manipulation, democratic risk Lost of work	Brown, 2020; Hutson, 2021; Hartmann et al., 2023; Kennedy et al., 2018; McConnell-Ginet, 2020; Mozes et al., 2023. Abdullah et al., 2022; Dwivedi et al., 2023b; Qadir et al., 2022; Lazar & Nelson, 2023.
Security, privacy, data protection, unintentional extraction, leakage of confidential or private information.	Alshater, 2022; Cohen, 2013; Chen, 2023; Grassini, 2023; Huang et al., 2022; Kobis & Mossink, 2021; Ray, 2023; Vassilev, 2023; Wang et al., 2023; Wu et al., 2023.
Sustainability	George & George, 2023; Holzapfel et al., 2022; Holzapfel et al., 2022; Gill & Kaur, 2023; Holzapfel et al., 2022; Barnett, 2023; Mannuru et al., 2023; Sastry et al., 2024; Shelby et al., 2023.
Transparency, explainability, Evaluation & Auditing, controllability, opacity and black box issues Tendency to oligopoly and excessive power	Anderljung et al., 2023; Castelvechi, 2016; Hendrycks et al., 2023; Ji et al., 2023; Mökander & Floridi, 2023; Wang et al., 2023. Hao, 2020; Hagendorff, 2024; Ray, 2023; Mannuru et al., 2023; Dwivedi et al., 2023b; Weidinger et al., 2023.

Table 2

**Table 2 Main benefits associated with generative AI for researchers.****Improving efficiency: saving time and effort**

Literature review and various bibliographic sources, recommendation of articles and summaries of the state of the art Writing of study proposals. Structured presentation of reports and research results Automatic formatting, elaboration of tables, graphs and aesthetic figures Generating code to solve analytical or computational tasks Guidance on ethics, compliance and professional regulation Erasing the language barrier for non-English speakers Discovery of new knowledge. Peer-review process response; citation count	Adam, 2021; Aithal & Aithal, 2023a; Alshater, 2022; Anson & Straume, 2022; Baidoo-Anu & Owusu, 2023; Bail, 2024; Banerjee et al., 2023; Burger et al., 2023; Cheng & Liu, 2023; Dergaa et al., 2023; Deroy et al., 2023; Dowling & Lucey, 2023; Else, 2023; Hill-Yardin, et al., 2023; Khalifa & Ibrahim, 2024; Kalla & Smith, 2023; Liebreiz et al., 2023; Lund et al., 2023; Ray, 2023; Salvagno et al., 2023; Sallam, 2023; Thorp & Vinson, 2023.
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**Quality' improvement**

Obtaining expert advice Generation of new questions based on patterns and trends and gaps in the research Methodological orientation, selection of methods for experimental design Statistical analysis, hypothesis generation, hypothesis testing, sample size, randomization, and control group allocation Data visualization and exploration Detection and correction of errors and biases	Aithal & Aithal, 2023b; Bail, 2024; Banerjee et al., 2023; Dasborough, 2023; Dwivedi et al., 2023a; Gao et al., 2022; Gill & Kaur, 2023; Guo et al., 2023; Huang & Chen, 2019; Kung et al., 2022; Lahat et al., 2023; Lund et al., 2023; Mahama et al., 2023; Pavlik, 2023; Rice et al., 2024; Ziems, 2023.
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**Cost savings****Data collection: surveys, questionnaires or interviews**

Data processing, cleaning, coding. Interpretation of findings: conclusions and inferences Experience identification Replication of the research findings Collaborative platform Overcoming language barriers	Anson & Straume, 2022; Bahrini et al., 2023; Bail, 2024; Banerjee et al., 2023; Carlini et al., 2020; Coveney et al., 2016; Elbadawi et al., 2024; Gefen & Arinze, 2023; Grace et al., 2018; Helberger & Diakopoulos, 2023; King, 2023; McDermott et al., 2019; Rice et al., 2024; Sallam, 2023; Salvagno et al., 2023.
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Table 3

**Table 3 Some needed questions in relation with Hibrid Paper's acceptance.****Editors must ensure that authors**

Do not include chatbots as coauthor Declare the use of AI generative tools in the paper, and specifications of the chatbot used: name, version, model, source. Take formal responsibility for all contributions made by chatbots, including accuracy and proper attribution of all cited material, and for searching for and citing omitted sources. Detail how the chatbot has been employed in the paper, including method of application (query structure, syntax). Detail specifically the contribution to the science	Bhatia & Kulkarni, 2023; Contractor et al., 2022; Flanagan et al., 2023 Gao et al., 2022; Graf & Bernardi, 2023; Gupta, 2024; Hosseini, et al., 2023; Hu, 2023; Lee, 2023; Leung et al., 2021; Liebreiz et al., 2023; Long & Magerko, 2020. Samuelson, 2023; Scerbo, 2023; Sengupta & Honavar, 2017; Stokel-Walker, 2023; Teixeira da Silva & Tsigaris, 2023; Uchendu et al., 2023; Yeo-Teh & Tang, 2023; Zhuo et al., 2023b; Zielinski et al., 2023.
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Table 3 (continued)

Editors must ensure that authors	
Editors must ensure that referees	
Is a human expert reviewer	Baek et al., 2024; Lu et al., 2024; Si et al., 2024; Leung et al., 2021.
Have a skilled and/or interdisciplinary team able to evaluate content efficiently and accurately.	
Have appropriate tools to help them detect content generated or altered by AI, with or without it being declared.	
Have a skilled and/or interdisciplinary team able to evaluate content efficiently and accurately.	

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## Author contributions

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## Informed consent

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## Additional information

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