



Description

Generative AI has moved rapidly from a novelty to a routine tool in many scholarly workflows. Evidence from early adopters indicates substantial uptake: an empirical analysis of publications during late 2022 early 2023 found that language models had contributed to more than 10% of papers across a range of journals. This shift matters because AI can speed drafting, improve clarity, and help non-native speakers reduce language barrier to publication, yet it also raises questions about authorship, accuracy, and preservation of an individual researcher's scholarly voice. This article outlines what AI tools are, why they matter for researchers, common risks and misunderstandings, and practical, ethically grounded steps to use AI without losing intellectual ownership of one's work.

How Do LLMs Assist Writing

A *large language model (LLM)* is a type of machine-learning model trained on massive text corpora to predict and generate human-like language. LLMs (for example, GPT-family models, Gemini, or Claude) can summarize literature, suggest rewrites for clarity, generate outlines, and act as a dialogic partner for brainstorming. Because these models learn statistical patterns rather than verify facts, their output can be fluent yet inaccurate; authors therefore remain responsible for verifying content and sources. The technical limitations of LLMs including hallucinations, bias from training data, and sensitivity to prompt phrasing shape how they should be used in academic contexts.

Benefits of AI in Academia

Al tools can improve efficiency at multiple stages of the writing process. For non-native English speakers, empirical studies report measurable gains in fluency, grammatical accuracy, and clarity when Al is used as an editing or feedback partner; interventions using generative models have shown positive effects on writing quality in classroom and EFL settings. Al can also accelerate time-consuming tasks such as initial literature summarization, restructuring paragraphs to improve flow, and generating multiple phrasing options that preserve technical meaning. When used as a drafting aide or reviewer checklist, Al often enables authors to spend more time on conceptual framing and data interpretation rather than micro-editing.



Risks, Ethical Considerations, and Publisher Expectations

Al introduces specific ethical and quality risks that affect publishability and scholarly integrity. Major editorial bodies and publishers agree on two core points: Al tools cannot be credited as authors, and use of such tools must be transparently disclosed in submissions. The ICMJE and COPE guidance, reflected in publisher policies, emphasize that human authors remain fully accountable for accuracy, originality, and attribution; journals increasingly require a description of how Al was used in the methods, acknowledgements, or cover letter. Beyond attribution, Al can fabricate citations, produce subtly incorrect statements, and generate figures or images that mimic experimental output each of which can lead to serious ethical breaches if unchecked.

Detection Tools and Why They Cannot Be the Only Safeguard

Academic institutions and publishers have invested in Al-detection systems, yet the available evidence shows these detectors are imperfect. Evaluations of multiple detection tools found varying accuracy and nontrivial false-positive and false-negative rates; detectors can misclassify human writing (especially from non-native authors) as machine-generated and miss obfuscated or edited Al output. Consequently, relying solely on detectors to police Al use risks <u>unfair accusations or missed issues</u>. The responsible approach combines disclosure, human verification, and editorial policies rather than treating detectors as definitive proof.

Preserving Voice and Intellectual Ownership

Maintaining an authentic academic voice means using AI to *enhance* expression, not to replace original thought. Researchers should treat AI output as a draft or suggestion that requires rewrite and interrogation. When a model proposes phrasings or structural changes, authors should adapt the language to reflect their conceptual priorities, preferred terminology, and discipline-specific conventions. This practice keeps the manuscript's rhetorical choices tethered to the researcher's intent, and it ensures that interpretive claims remain attributable to human authors who can defend them during peer review.

A Practical Workflow for Responsible Al-Assisted Writing

Adopting a reproducible, transparent workflow reduces risk while harnessing Al's benefits. The following checklist provides actionable steps researchers can implement during drafting and submission:

- Before using AI: Decide what the tool's role will be (e.g., brainstorming, language editing, summarization) and whether the planned use requires disclosure under shortlisted journal policies.
- **During drafting**: Keep a changelog or brief notes indicating where AI was used (e.g., "AI suggested paragraph reorganization in Methods, 2025-06-10"), and do not accept factual statements without independent verification of primary sources.
- **Verifying content**: Cross-check any Al-provided statements or citations against original articles or databases; confirm experimental details, numerical values, and references personally.



- Attribution and disclosure: Follow the journal's or discipline's guidance disclose the tool name and version and describe the nature of its contribution in the manuscript and cover letter where required.
- **Final authorship check**: Ensure all listed authors meet authorship criteria (contribution, approval, accountability) and that AI has not been listed or treated as a contributor.

Using that workflow helps preserve voice and ethical accountability while keeping manuscripts aligned with current editorial standards.

Common Mistakes and How to Avoid Them

A frequent error is treating AI output as an authoritative source. Because models can generate plausible-sounding but incorrect information, authors should always validate references and avoid citing AI as a primary source. Another mistake is over-editing AI output to the point of losing critical nuance; instead, use AI suggestions as editable scaffolds and intentionally rephrase to match one's established terminology. Finally, failing to disclose AI use risks desk rejection or post-publication correction; check publisher policies early in the submission process.

How AI Use Differs Across Tasks

Al's suitability varies by task. For language polishing, summarization, or generating multiple phrasing options, LLMs are well suited and pose lower risk when outputs are verified. For conceptual design, interpretation of results, or literature synthesis where nuance and domain expertise matter, Al should be used sparingly and always reviewed by experts. For image generation or synthetic experimental figures, the risk of unintentional fabrication is high and many journals treat such outputs with particular scrutiny; image provenance must be transparent and justified. Understanding these task-dependent differences helps manage risk while leveraging strengths.

Tips and Tricks to Keep Your Voice While Using Al

When using AI to refine text, prompt deliberately: ask the model to preserve specified technical terms, sentence rhythm, or authorial stance. Use AI for constrained tasks (e.g., "Suggest three ways to make this methods paragraph clearer while retaining technical terms X, Y, Z"), then perform a manual rewrite to integrate favored suggestions. Maintain a personal style guide (common phrasing, preferred passive/active constructions, disciplinary conventions) and use it to edit AI drafts so the final manuscript reads consistently with the author's prior work. Keep edits iterative and small so that the rhetorical signature remains human.

When to Seek Professional Support

If language or formatting constraints are delaying submission, professional editorial support can complement AI use. Consider using manuscript-editing services that focus on refining paraphrase choices, ensuring citations are correctly integrated, and preparing responses to peer review all framed as assistance that preserves authorship and accountability. Such services can help implement disclosure language and improve clarity so that the researcher's voice and intellectual contributions remain central. Enago's manuscript-editing services, for example, can help refine language and citation



practices so disclosures and attributions meet journal expectations without obscuring authorship.

In Practice: Short Examples

A materials-science researcher uses an LLM to generate three possible opening paragraphs summarizing recent work on a technique. After selecting elements from each option, the researcher rewrites the chosen text to emphasize their laboratory's methodological nuance and references the original studies that the AI suggested only after verifying them independently. A psychology team uses AI to generate multiple phrasings of survey items; the team then evaluates each item for conceptual validity with domain experts before finalizing the instrument. These workflows show AI as a drafting partner never the final arbiter of content.

Closing Guidance

Al tools offer measurable benefits for clarity, accessibility, and drafting speed, especially when combined with discipline expertise and transparent reporting. At the same time, publishers and ethical bodies require disclosure and insist that human authors retain accountability for all manuscript content. By integrating a short, repeatable workflow plan use; verify outputs; document changes; disclose appropriately researchers can use Al productively while preserving voice, accountability, and scholarly integrity. Visit our Responsible Al Movement for a summary table of publisher policies, practical author workflow, and learning resources to help you use Al responsibly and productively!

Category

1. Reporting Research

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