



Al: the good, the bad, and the possibly amazing

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Should librarians be scared of (Generative) AI?





The Bad: Al can be biased:



Bias in data [Scheuermann et al]:

A particularly interesting example was PUBFIG, which had two gendered annotations: "male" and "attractive woman," of which there was no associated "female" [87]. The absence of annotations for "female" or "attractive man", however, highlights the culturally-situated values around gender that can emerge within an annotation schema (c.f., [150]).

Bias in algorithms [Amnesty International Report]:

Xenophobic machines: Discrimination through unregulated use of algorithms in the Dutch childcare benefits scandal

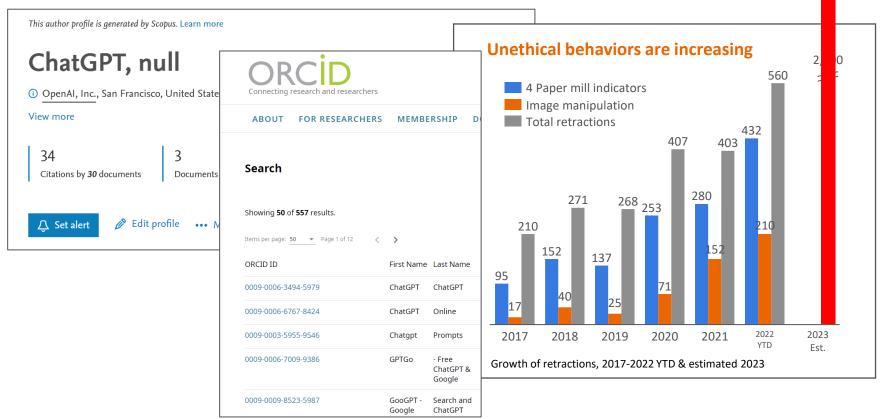
October 25, 2021, Index Number: EUR 35/4686/2021

EUROPE AND CENTRAL ASIA TECHNOLOGY AND HUMAN RIGHTS

Social security enforcement agencies worldwide are increasingly automating their processes in the hope of detecting fraud. The Netherlands is at the forefront of this development. The Dutch tax authorities adopted an algorithmic decision-making system to create risk profiles of individuals applying for childcare benefits in order to detect inaccurate and potentially fraudulent applications at an early stage. Nationality was one of the risk factors

The Bad: Al can be weaponized:





https://www.the-geyser.com/chatgpt-says-its-not-an-author/ -- https://www.elsevier.com/about/policies/publishing-ethics/the-use-of-ai-and-ai-assisted-writing-technologies-in-scientific-writing

Internal data but covered by Sarah Jenkins at https://www.stm-assoc.org/events/stm-research-integrity-master-class-2/

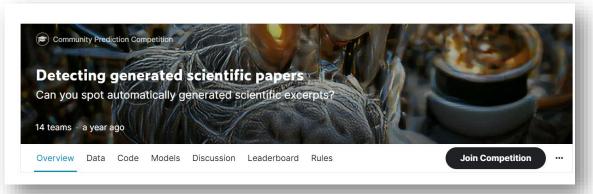
The Good: But AI can also do great things:



Al can help protect patients and train nurses:

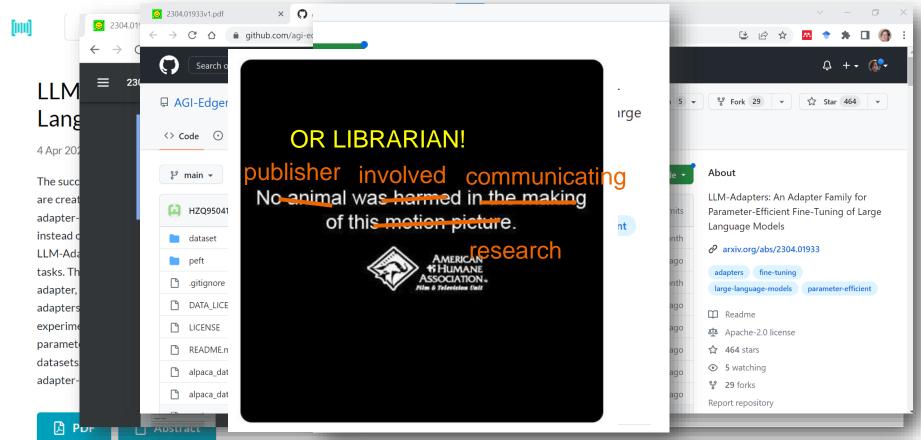


Al can help detect Al-generated Content!



How is AI research published? shared? used!





Paperswithcode: https://paperswithcode.com/paper/llm-adapters-an-adapter-family-for-parameter

Where does Elsevier stand, on all this?



Dubliching ethics

Juties of the Dublisher

Juties of Editor

Duties of Poviower

Duties of Author

Deferenc

<u>Elsevier's Al author policy</u> states that *authors* are allowed to use generative Al and Al-assisted technologies in the writing process before submission, but only to improve the language and readability of their paper and with the appropriate disclosure, as per our instructions in Elsevier's Guide for Authors.

THIS DOLLEY HAS DEED THEREIGN ON THE LISE OF SCHEDALING WE WIND WE ASSISTED TECHNOLOGICS. AND

Generative AI or AI-assisted technologies should not be used by editors to assist in the evaluation or decision-making process of a manuscript as the critical thinking and original assessment needed for this work is outside of the scope of this technology and there is a risk that the technology will generate incorrect, incomplete or biased conclusions about the manuscript.

<u>Al Principles</u>, such as those used during the screening process to conduct completeness and plagiarism checks and identify suitable reviewers.

about the manuscript and/or the authors. For this reason, editors should not upload their letters into an AI tool, even if it is just for the purpose of improving language and readability.

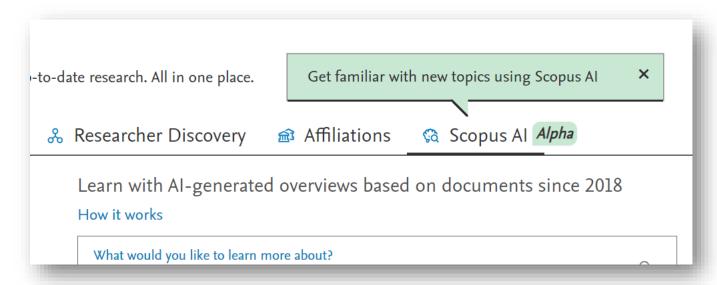
Peer review is at the heart of the scientific ecosystem and Elsevier abides by the highest

https://beta.elsevier.com/about/policies-and-standards/publishing-ethics?trial=true

For instance: ScopusAl







Vector Search

- Generate vectors quickly and efficiently, in under 300ms
- Results are of high quality and match user intent

Prompt Engineering

- Ensure outputs are not just accurate, but also meaningful
- Outputs are produced in usable formats (JSON)

Transparency

- Ensure every claim and statement is grounded in academic research
- All research can be traced back to source

Bringing the Responsible Al Principles in Practice



AS PART OF OUR RESPONSIBLE AI APPROACH...



1. We consider the real-world impact of our solutions on people



2. We take action to prevent the creation or reinforcement of unfair bias



3. We can explain how our solutions work



4. We create accountability through human oversight



5. We respect privacy and champion robust data governance

Generative AI output is **evaluated against sample data** before
release

Ensure generative AI is **grounded** in **academic information** and **traceable**

Observe Elsevier guidelines around personal data usage – GDPR

Behind all this: the pyramid of trust



Can be asynchronous, by different annotators, for different reasons

Relations/ Comparisons

Comparison to other work:

New, supporting/contradicting
findings; has impact on x/y/z, is cited/citing,
used in policy/patents, etc

Machine readable provenance of annotations

We know this citations, tweets a

ssertions, mentions,

Key entities/Assertions

Here is what this article claims and covers: Genes, stars, studies, etc We know this because we ran **SciBite** to find genes and relations, **Telescope** to find equipment, **Reaxys** to find compounds, and the key assertions through **NLP**...

Solid science/Reproducibility

This paper describes a {clinical trial; biology experiment} that complies with community standards in {RCTs, cell biology}, and links to data and software; it has been peer-reviewed and revised

We know this because we checked that the data is linked/ reviewed, we checked it with CONSORT, SAGER & STAR guidelines, we ran double-blind peer-review on date X and received Y revisions...

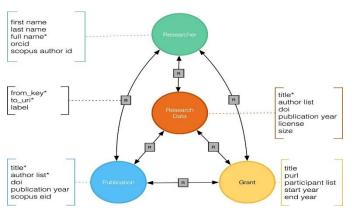
Research integrity/Ethics

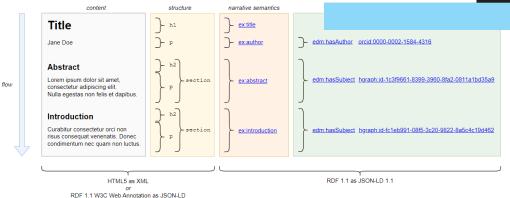
This is verified as real research, written by an identified (and human) researcher, it is conflict-free, and novel scholarly output

We know this because we ran iThenicate and checked ORCID, conflict of interest, etc

Behind that: Persistent Identifiers and Open Knowledge Gralovemy



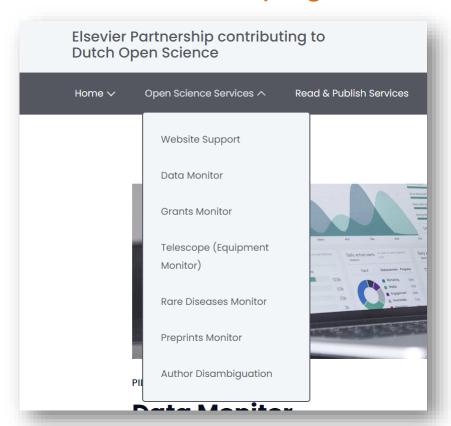


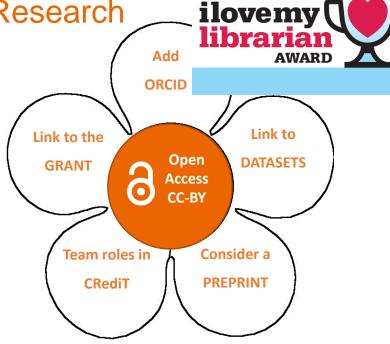


- 2010: ORCID: PIDs for authors
- 2011: Force11 Data Citation Principles: PIDs for data
- 2014: COPDESS Enabling FAIR Data: shared author's instructions
- 2014 FAIR Data Principles more than open, also interpretable data
- 2015: CrediT Taxonomy: roles of researchers
- 2017: Scholix leading to (2023) Open Science/Research Graphs for Fair Data
- 2018: Research Object Authoring tool: creating a linked data graph for an output
- 2022: Linked Document Standard: adding metadata as we go
- 2023: Peer review terminology: structured levels that can be reported
- 2023: NIST Research Data Framework: stakeholders and roles



This leads to helping Enable Open Research





- 1. Use ORCIDs to find all outputs for each researcher
- 2. Find **DATASETS** wherever they are
- 3. Aggregate outputs per funder based on **GRANT IDs**
- 4. Acknowledge team members using **CRediT**
- 5. Find PREPRINTS wherever they are



More info about our trusted collaboration & principles in the Netherlands: EPDOS.nl

In summary:

- Al can be good, bad, or amazing
- It is important to be responsible when using AI
- To move forward we need to embed these technologies into a robust community infrastructure with:
 - 1. **Provenance** to enable verification
 - Persistent identifiers for all components
 - **3. Knowledge graphs** to connect them
- The way forward is all together: institutions, funders, publishers and librarians
- In other words:





Maybe AI should be scared of librarians?



Tak for jeres opmaerksomhed



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