MSR74

Data extraction from full-text PDFs using Large Language Models for systematic reviews

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Introduction

Rapid adoption of evidence-based decision-making in medicine, public health, and other fields has fueled a surge in evidence synthesis.

Traditional evidence synthesis methods are time-intensive and resource-demanding.

Advances in Artificial Intelligence (AI) and Large Language Models (LLMs) offer transformative potential to address these challenges by automating key tasks.

Problem and Question

Issues with LLM hallucinations, lack of model transparency, and inconsistent outputs hinder reliability and scalability.

Can a multi-layered and AI-assisted approach to data extraction increase efficiency while maintaining accuracy during systematic review?

Methods

We used an example systematic review with 193 included studies about digital health interventions.

Our workflow parsed each PDF to text, encoded the content, and applied task-specific prompt engineering with OpenAI GPT-4o. The model returned answers and highlighted supporting passages, allowing us to extract 40 data elements such as demographics/study characteristics, intervention details, and outcome data.

Human methodologists extracted the data assisted by the AI suggestions; we checked the accuracy of the AI by comparing the final human response to the AI recommended answer.

Results

Overall, the AI reached 72.3% accuracy in its suggestions, with individual questions ranging from 52.9% to 99.3%. No custom prompt engineering was applied.

Productivity increased from an average of 80 extracted elements per hour for human-only extraction to 195 extracted elements per hour for AI-assisted extraction, an estimated time savings of 59%. This approach also appeared to reduce fatigue, allowing methodologists to work for an additional 1-2 hours without breaks.

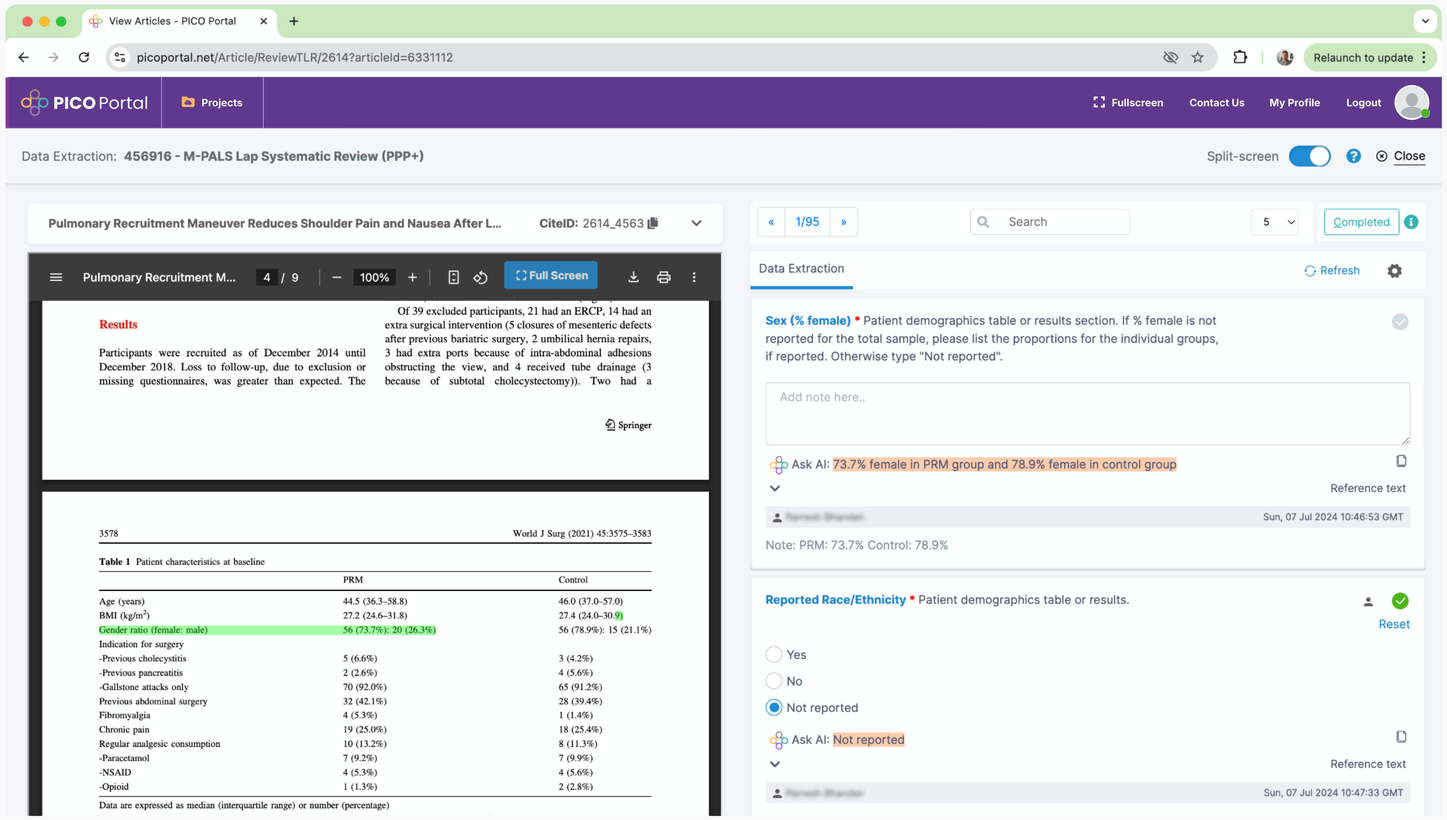

Figure: AI-assisted data extraction (right) with interactive PDF (left).

Conclusions

LLM technologies, combined with human oversight, have demonstrated the ability to reduce time, costs, and decision fatigue.

This approach represents a scalable solution to accelerate evidence syntheses across diverse fields including clinical research, public health policy making, and education.

Acknowlegments: We would like to acknowledge the EPC team at Minnesota University for allowing us to use their systematic review as an example.example.

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