



Recommendations and guidance on responsible AI in evidence synthesis

And an introduction to the new AI Methods Group

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Trusted evidence.
Informed decisions.
Better health.



Overview

1. Why we need to embrace AI and the importance of infrastructure
2. A global challenge and introducing RAISE (Responsible AI use in evidence Synthesis)
3. Applying the RAISE recommendations in practice; what does it mean for authors?
4. Introduction to the new, joint AI Methods Group
5. Questions



Why we need to embrace AI



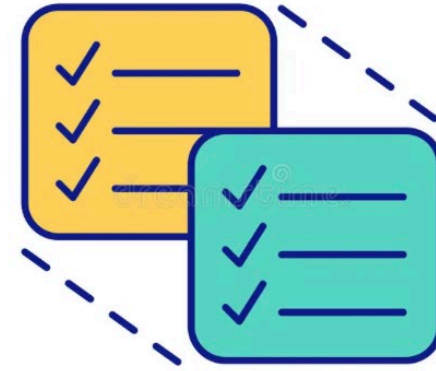
Context: core principles of evidence synthesis



Rigor



Transparency



Replicability

The stakes are high with important decisions based on evidence syntheses

The problem!

The vast and growing amount of research produced means that we are finding it increasingly difficult to keep up!



Pinpointing needles in giant haystacks (Shemilt 2017)

We have a huge **specificity problem** in information retrieval for evidence synthesis.

Questions are becoming more complex.

It's hard to find what we need without retrieving large quantities of noise.



The inefficiency of isolation

Siloed working and siloed data resulting in significant duplication of effort and research waste.

The manual process of evidence synthesis has become unsustainable



Artificial intelligence

An umbrella term which means different things to different people.

Ways to automate tasks and activities currently performed by humans.



Artificial Intelligence

1950s

Text
analytics

Rule based
systems

1960s

1980s

Natural
language
processing

Machine
learning

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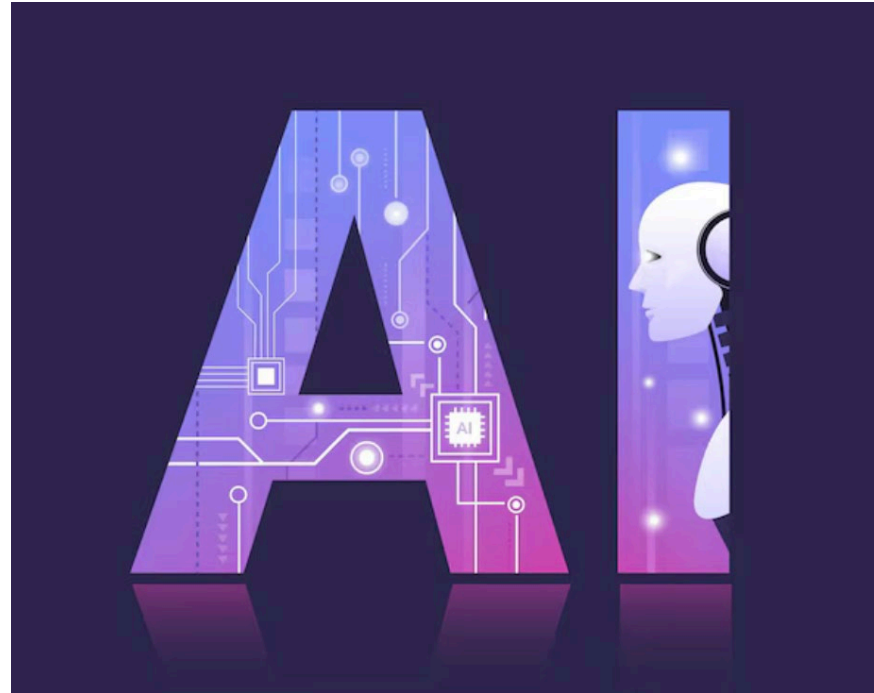
GenAI

Large Language Models offer new potential

What's so special about LLMs?

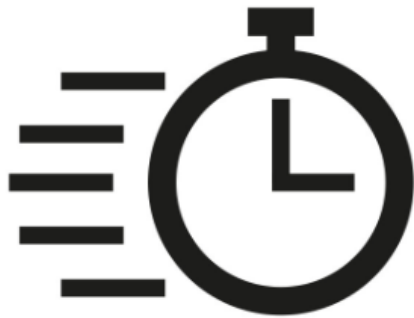
Generative AI uses Large Language Models (LLMs) which have been trained on vast datasets.

LLMs are able to recognize, summarize, translate, predict, and generate text without any specific task training or only a few instructions as a form of prompts

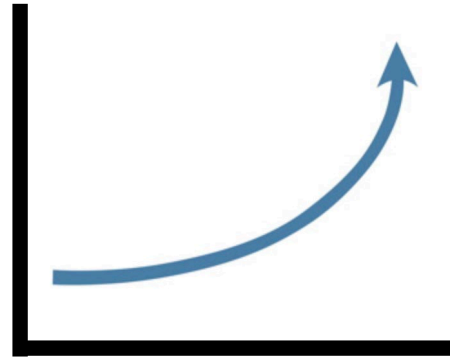


Plenty of excitement about AI

Increases capacity and brings new capability



Speed



Scale



Breadth

LLMs have dramatically opened-up the potential task space both in terms of capacity and capability

Plenty of excitement about AI

Increases capacity and brings new capability



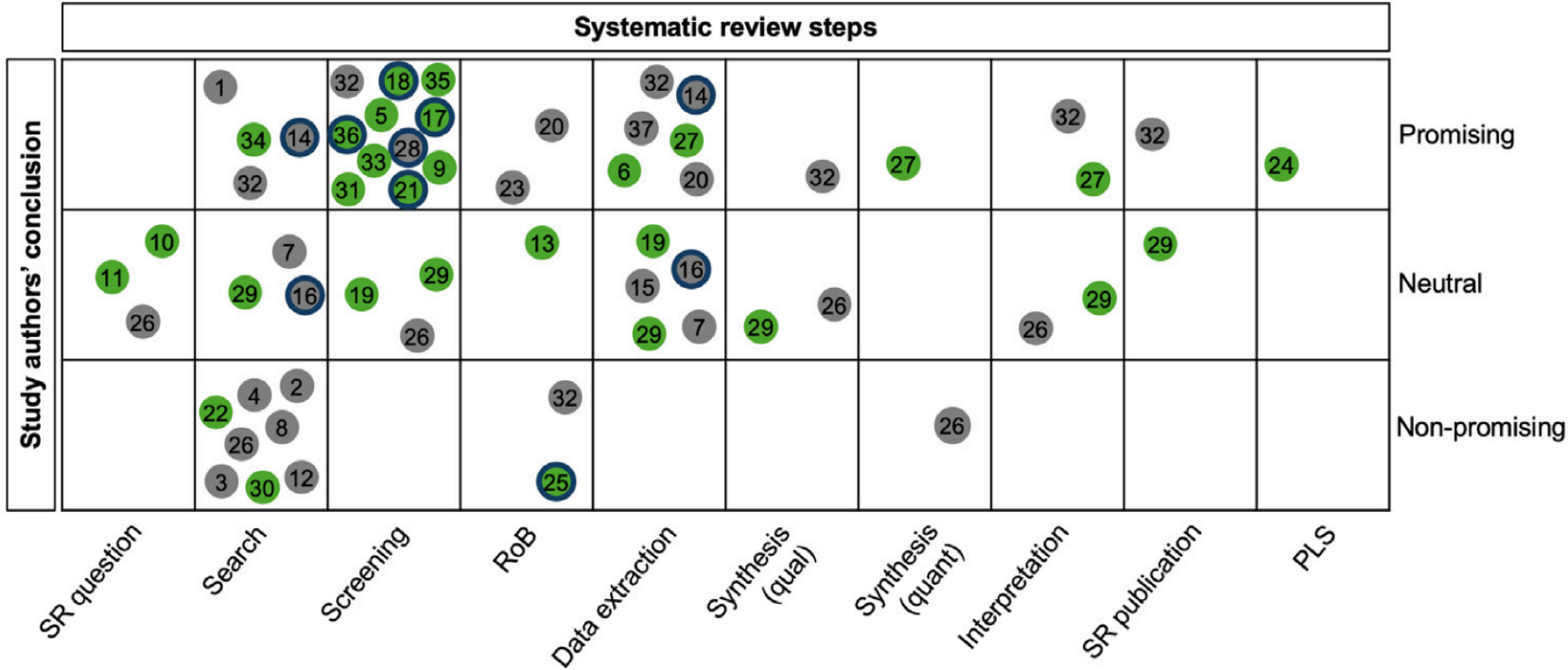
Accuracy

Gartlehner et al. performed a head-to-head comparison: two humans extracting data vs AI (Claude)+human.

“The AI-assisted approach had fewer incorrect extractions (9.0% vs. 11.0%) and similar risks of major errors (2.5% vs. 2.7%) compared to the traditional human-only method, with a median time saving of 41 minutes per study”

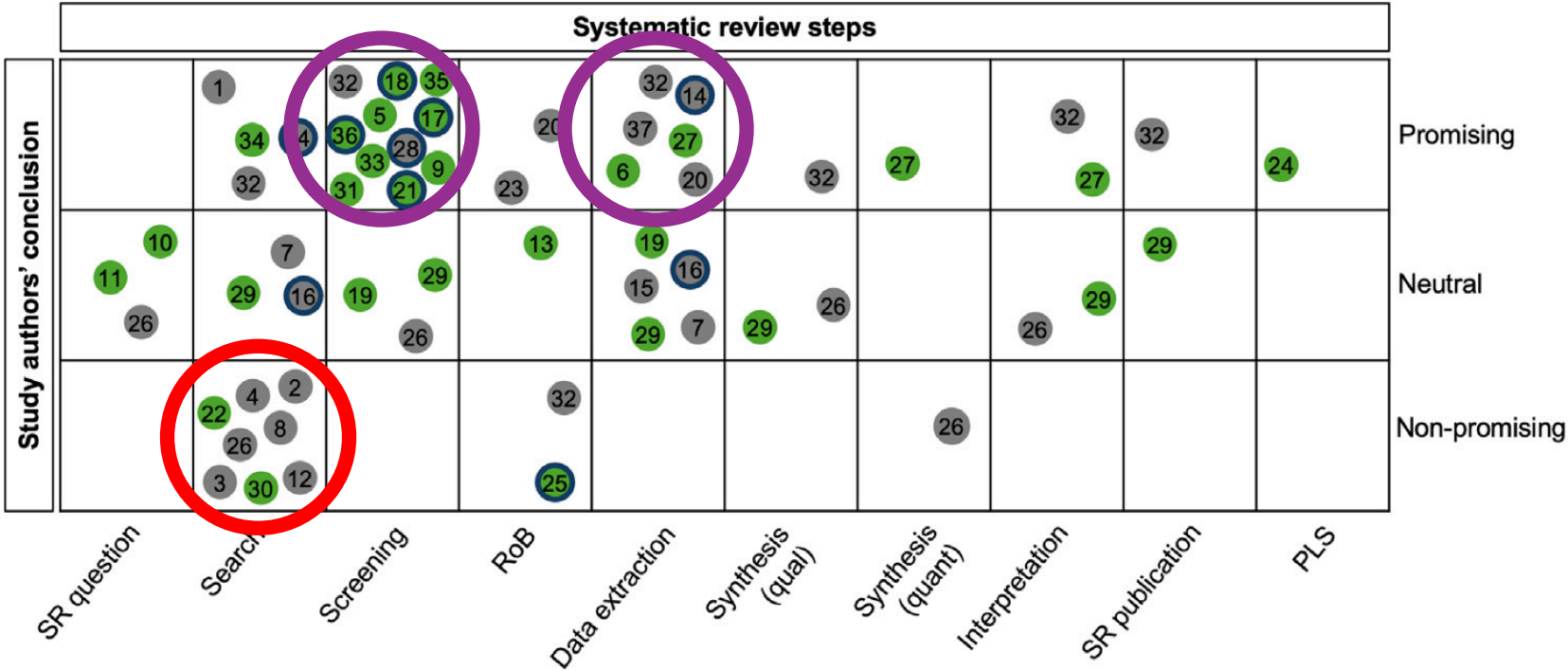
Large Language Models offer new potential

Methodological studies are emerging and have covered



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Methodological studies are emerging and have covered





But... (there's always a but!)

Challenges that might impact the utility of LLMs in evidence synthesis

Over-fitting

Algorithmic
bias

Black box
predictions

Hallucination



Barriers to effective implementation

Some significant non-technical challenges

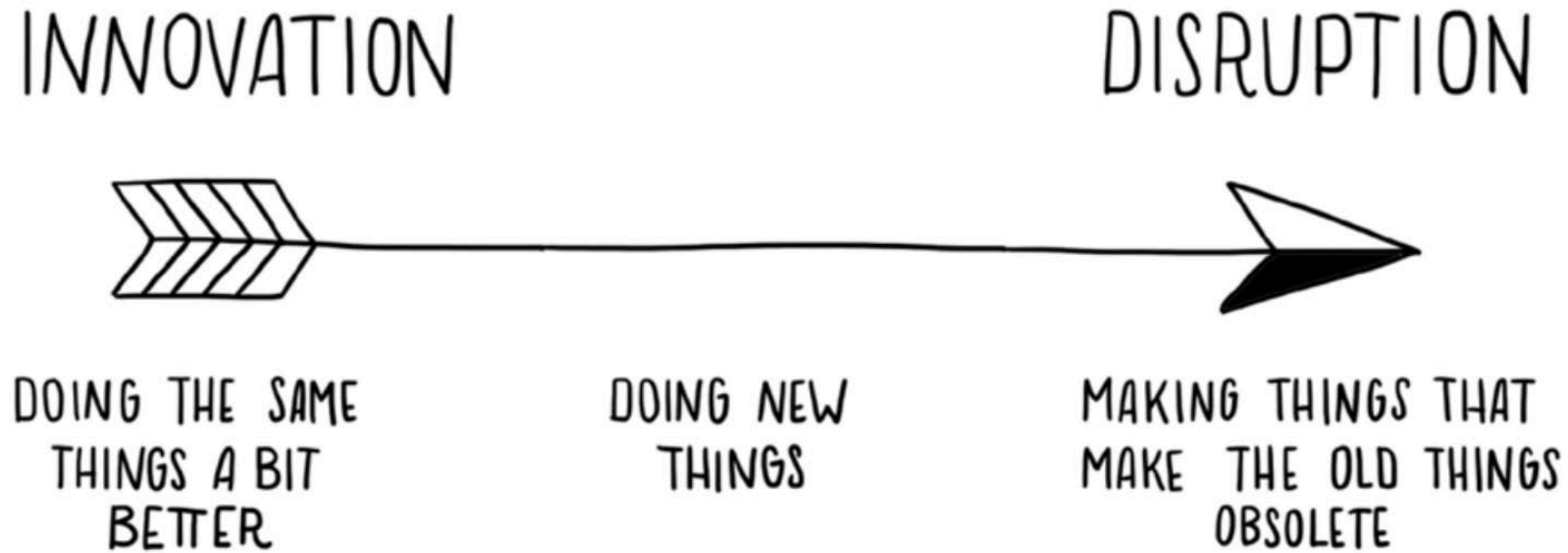
AI literacy

Access

Ethical

Fast pace of
change

From innovation to disruption



Cross-organization collaboration and infrastructure

We have a responsibility to use AI, but only if we can use it safely and responsibly.

We'll be more effective at understanding and overcoming the challenges together: both at a methodological level and at a data level.

To do this we need guidance, frameworks and infrastructure that enables us to learn and evolve as the technology itself learns and evolves.

