

2023 GLOBAL ONLINE CONFERENCE | EMPOWERING LEARNERS FOR THE AGE OF AI

Insights from a meta review of AIEd research A call for increased ethics, collaboration, and rigour

Tuesday 24th October, 9am (UK time), 7pm (South Australia)

Dr. Melissa Bond

University College London, University of Stavanger, National Institute of Teaching

Dr. Hassan Khosravi
The University of Queensland





Margaret Newsonia Edition Deligis Reverse Science Scie

Living Communities



David R Horton (creator), © AIATSIS, 1996

Acknowledgement of Country

We would like to acknowledge the Traditional Custodians of the land on which this conference is being hosted, Kaurna country, and pay our respects to Elders past, present and emerging.

We respectfully acknowledge the Kaurna, Boandik and Barngarla First Nations Peoples and their Elders past and present, who are the First Nations' Traditional Owners of the lands that are now home to the University of South Australia's campuses in Adelaide, Mount Gambier and Whyalla. We are honoured to recognise our connection to the Kaurna, the Boandik and the Barngarla lands, and their history, culture and spirituality through these locations. We also acknowledge the other First Nations of lands across Australia, their Elders, ancestors, cultures and heritage.

Panel objectives

- What kind of research is being conducted in AIEd?
- What can it tell us about the affordances and challenges of using AI in various educational contexts?
- What opportunities for future research have been identified?
- What considerations are needed when undertaking robust studies in AIEd?

3

Review Team





Kunnskapssenter Dr. Melissa Bond



Prof. Maarten de Laat Ms. Phuong Pham Prof. George Siemens



Dr. Hassan Khosravi





Dr. Nina Bergdahl



Dr. Violeta Negrea Dr. Emily Oxley Prof. Sin Wang Chong

Presence of AI has increased in teaching & learning and in the public discourse

TECHNOLOGY

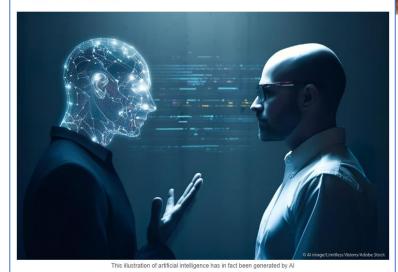
AI 'is clear and present danger to education'

School leaders announce joint response to tech

EU AI Act: first regulation on artificial intelligence

Society Updated: 14-06-2023 - 14:06 Created: 08-06-2023 - 11:40

The use of artificial intelligence in the EU will be regulated by the AI Act, the world's first comprehensive AI law. Find out how it will protect you.



Source: The Times

U.S. Department of Education Shares Insights and Recommendations for Artificial Intelligence

New policy report, part of Biden-Harris Administration's ongoing work to advance comprehensive approach to AI, summarizes the opportunities and risks for AI in teaching, learning, and assessment

MAY 24, 2023

Contact: Press Office, (202) 401-1576, press@ed.gov

Today, the U.S. Department of Education's Office of Educational Technology (OET) released a new report, "Artificial Intelligence (AI) and the Future of Teaching and Learning: Insights and Recommendations" that summarizes the opportunities and risks for AI in teaching, learning, research, and assessment based on public input. This report is part of the Biden-Harris Administration's ongoing effort to advance a cohesive and comprehensive approach to AI-related opportunities and risks.

The new report addresses the clear need for sharing knowledge, engaging educators and communities, and refining technology plans and policies for AI use in education. It recognizes AI as a rapidly advancing set of technologies that can enable new forms of interaction between educators and students, help educators address variability in learning, increase feedback loops, and support educators. It also outlines risks associated with AI—including algorithmic bias—and the importance of trust, safety, and appropriate guardrails to protect educators and students. The report recommends that the Department continue working with states, institutions of higher education, school districts and other partners to collaborate on the following steps:

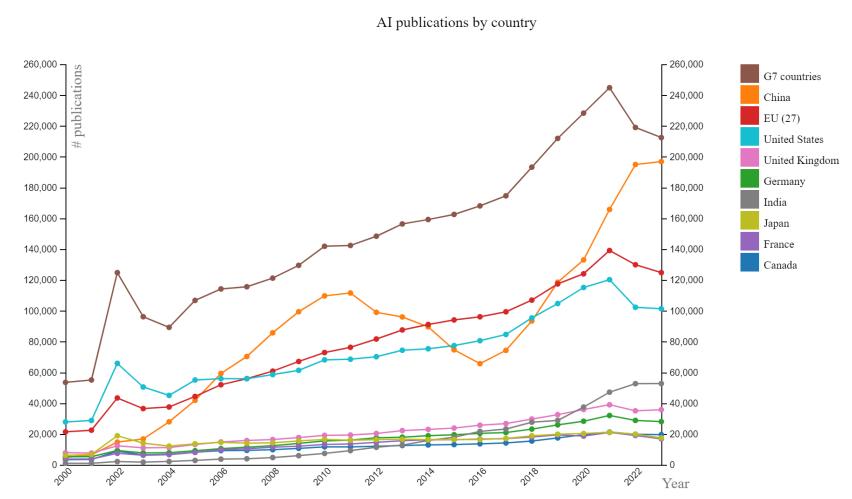
- Emphasize Humans-in-the-Loop
- 2. Align Al Models to a Shared Vision for Education
- 3. Design Al Using Modern Learning Principles
- 4. Prioritize Strengthening Trust
- 5. Inform and Involve Educators
- 6. Focus R&D on Addressing Context and Enhancing Trust and Safety
- 7. Develop Education-specific Guidelines and Guardrails

Source: U.S. Department of Education

Source: European Parliament



Interest in AI as a topic of research has also increased exponentially



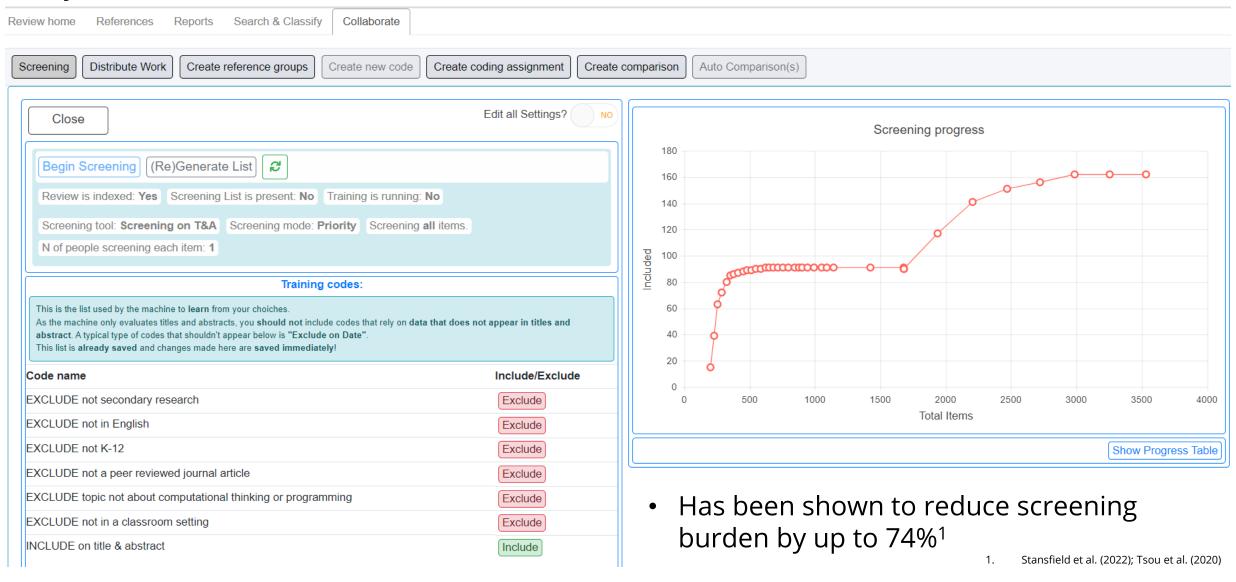
- Need for evidence syntheses to inform policy and practice
- Time & resource intensive

DEST (esp. automation) have helped make stages of the review process more efficient

Source: OECD.AI (2023), visualisations powered by JSI using data from OpenAlex., accessed on 17/9/2023, www.oecd.ai



Priority screening learns from items already coded on title & abstract and presents the most likely includes first



- Evidence of reticence and mistrust of using automated tools in research¹
- Not enough mistrust in Al by students (& educators)?²
- Are we producing quality evidence syntheses?
- How can we reduce 'research waste'?



Midjourney

" A biochemistry professor at a UK university"

Generative AI biases: A 'Midjourney' generated image, 30 June 2023, M. Compton

- . Jardim et al. (2022); Marshall et al. (2018)
- 2. Liu (2023); Smolansky et al. (2023)

Research Questions

What is the nature and scope of AIEd evidence synthesis?

- Review and publication types
- Authorship and geographical distribution
- International research collaboration
- Technology used
- Research quality
- General findings
- Benefits and challenges
- Research gaps



Methodology

Tertiary review¹ in order to "map the literature [...] and provide an opportunity to identify key concepts; gaps in the research; and types and sources of evidence to inform practice, policymaking, and research"².

Web of Science

Scopus

> ERIC

- > EBSCOHost
- > IEEE Xplore
- Science Direct
- > ACM Digital Library
- Google Scholar
- ResearchGate
- OpenAlex & snowballing.

AI	"artificial intelligence" OR "machine intelligence" OR "intelligent support" OR "intelligent virtual reality" OR "chat bot*" OR "machine learning" OR "automated tutor" OR "personal tutor*" OR "intelligent agent*" OR "expert system" OR "neural network" OR "natural language processing" OR "intelligent tutor*" OR "adaptive learning system*" OR "adaptive educational system*" OR "adaptive testing" OR "decision trees" OR "clustering" OR "logistic regression" OR "adaptive system*"
AND	
Education sector	"higher education" OR college* OR undergrad* OR graduate OR postgrad* OR "K-12" OR kindergarten* OR "corporate training*" OR "professional training*" OR "primary school*" OR "middle school*" OR "high school*" OR "elementary school*" OR "vocational education" OR "adult education" OR "workplace learning" OR "corporate academy"
AND	·
evidence synthesis	"systematic review" OR "scoping review" OR "narrative review" OR "meta-analysis" OR "evidence synthesis" OR "meta-review" OR "evidence map" OR "rapid review" OR "umbrella review" OR "qualitative synthesis" OR "configurative review" OR "aggregative review" OR "thematic synthesis" OR "framework synthesis" OR "mapping review" OR "meta-synthesis" OR "qualitative evidence synthesis" OR "critical review" OR "integrative review" OR "integrative synthesis" OR "narrative summary" OR "state of the art review" OR "rapid evidence assessment" OR "qualitative research synthesis" OR "qualitative meta-summary" OR "meta-ethnography" OR "meta-narrative review" OR "mixed methods synthesis" OR "scoping study" OR "systematic map"

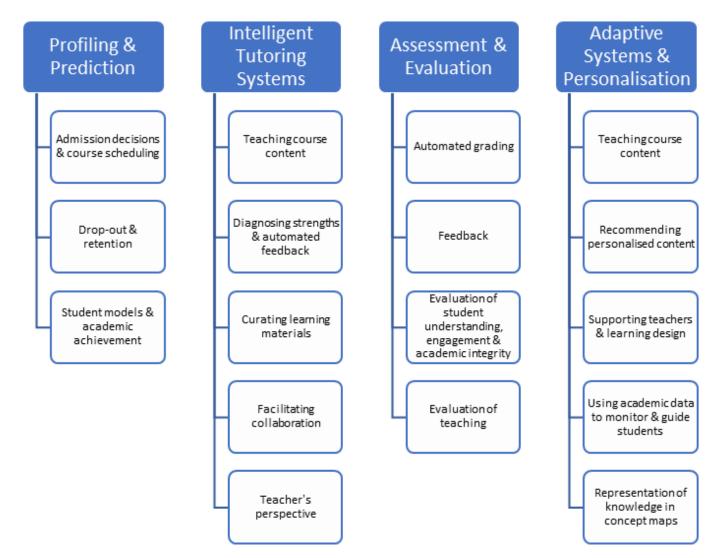
- Kitchenham et al. (2009); Lai & Bower (2020)
- 2. Daudt et al. (2013, p. 8)



Inclusion	Exclusion
Jan 2018 – July 2023	Published before 2018
Applications of Al in education	Not about Al
Formal teaching & learning	Informal learning
Journal article or conference paper	Chapters, editorials, theses
Secondary research	Primary research
English language	No method section



Data extraction





Zawacki-Richter et al. (2019)

Quality assessment

AMSTAR 21 / DARE2

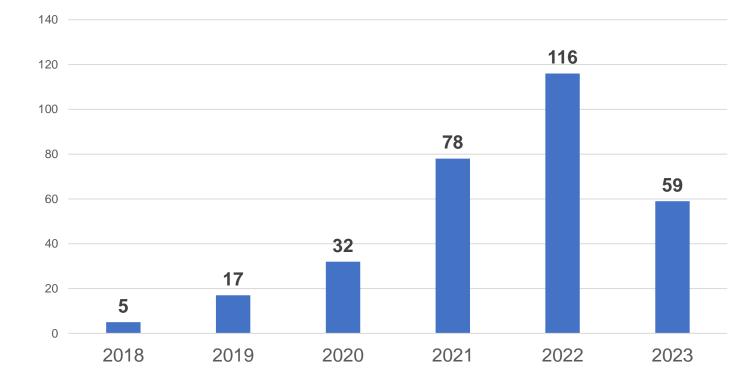
- 1. Are there any research questions, aims or objectives?
- 2. Were inclusion/exclusion criteria reported in the review and are they appropriate?
- 3. Are the publication years included defined?
- 4. Was the search adequately conducted and likely to have covered all relevant studies?
- 5. Was the search string provided in full?
- 6. Do they report inter-rater reliability?
- 7. Was the data extraction coding scheme provided?
- 8. Was a quality assessment undertaken?
- 9. Are sufficient details provided about the individual included studies?
- 10. Is there a reflection on review limitations?

- Shea et al. (2017)
- Centre for Reviews and Dissemination, 199



Key Findings

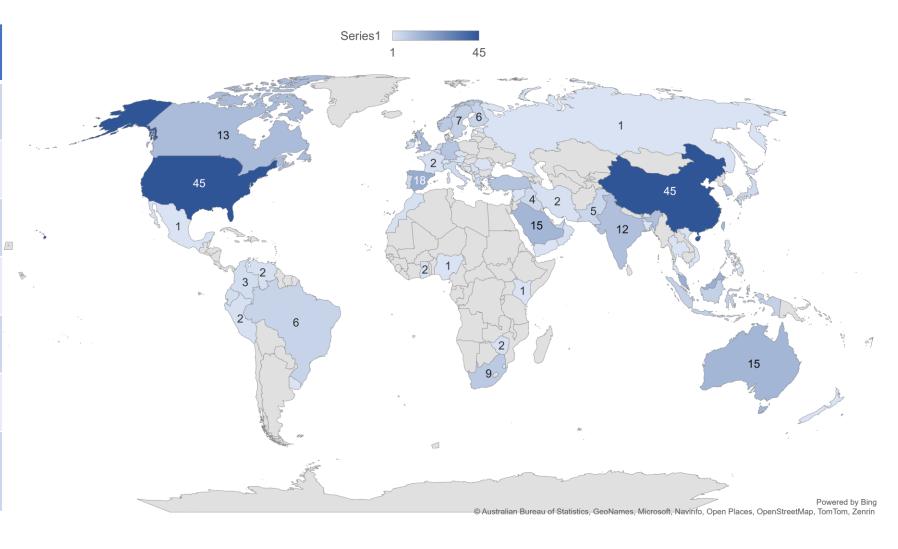
- 84% Journal articles
 - Computers and Education: AI
 - ➤ IEEE Frontiers in Education conference
- 61% open access
- 66% systematic reviews
- 32.9% Education,
 31.6% Computer
 Science & IT



- > 9.1% STEM
- > 6.5% online/BL/distance learning
- > 5.2% Health & Welfare
- > 4.2% Foreign language learning

Geographical distribution

Continent	%
Asia	47.6
Europe	26.7
North America	17.6
Middle East	9.8
Africa	5.2
Oceania	5.2
South & Central America	5.2

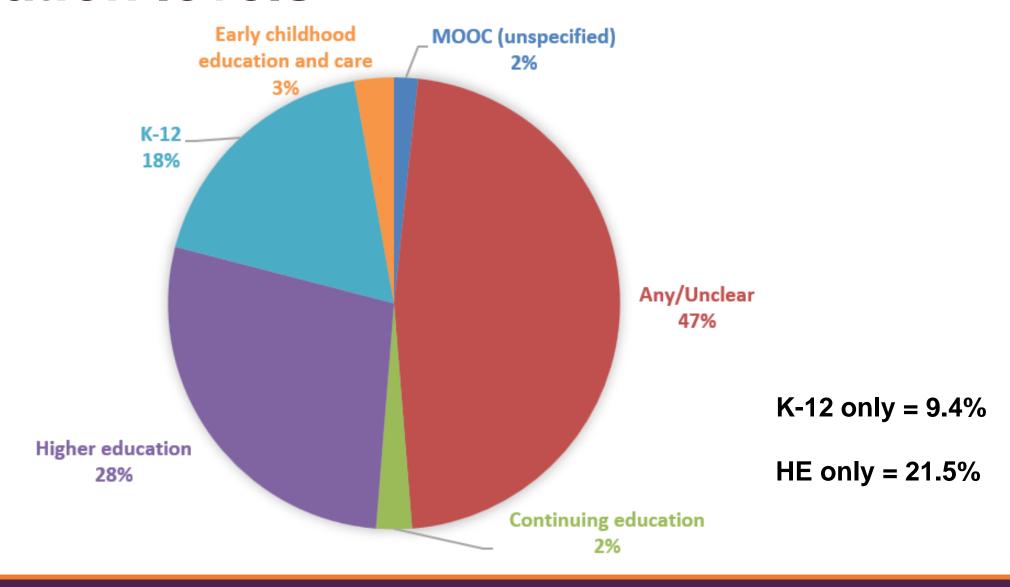


Authorship

- 92.2% collaborative, mostly teams of 2, 3 or 4 authors
- Africa and Middle East more solo authorship
- 66.8% domestic-only collaborations
- Middle East, Oceania and S & C America highest rates of international collaboration

Type of collaboration	Africa	Asia	Europe	Oceania	N.America	S. America	Middle E
	16	146	82	16	54	16	30
No collaboration	19%	8%	4%	0%	7%	0%	10%
Domestic collaboration only	50%	66%	55%	50%	57%	50%	30%
Total IRC	31%	27%	41%	50%	35%	50%	60%
International collaboration only	19%	2%	9%	0%	4%	13%	3%
Both domestic and international collaboration	13%	25%	33%	50%	31%	38%	57%

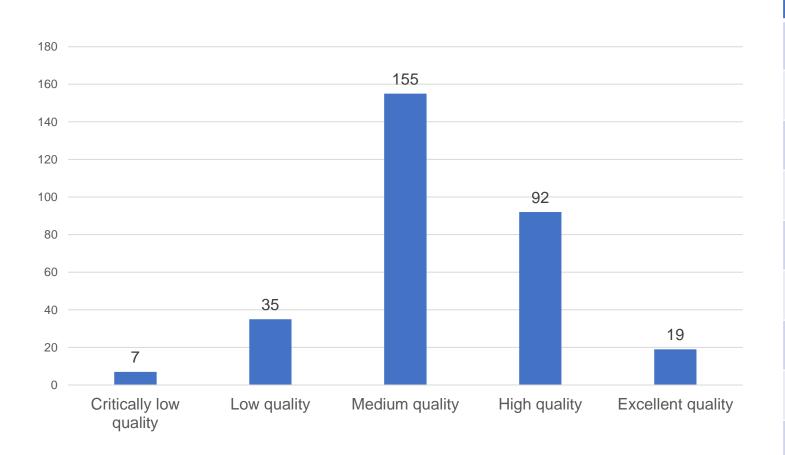
Education levels



Types of evidence synthesis

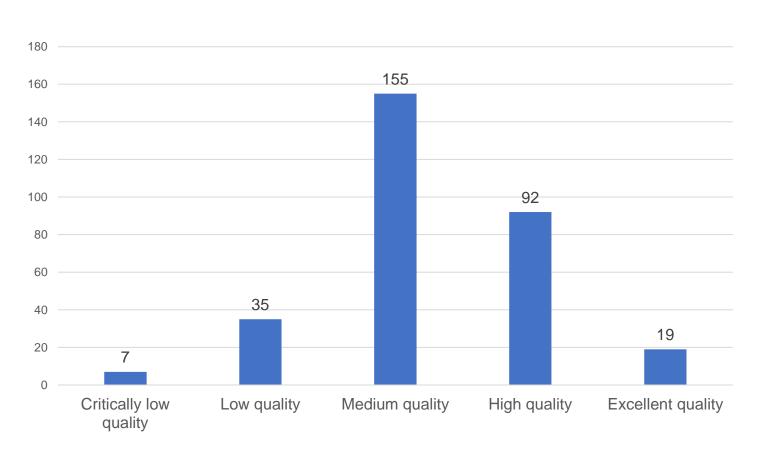
Review type	n	%
Systematic review	202	65.8%
Bibliometric review	44	14.3%
Literature review	23	7.5%
Meta-analysis	21	6.8%
Scoping review	16	5.2%
Mapping review	12	3.9%

Evidence synthesis quality



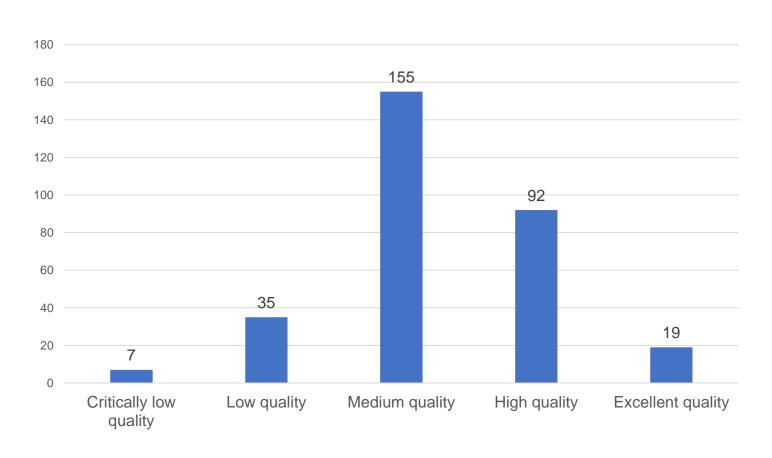
Quality Indicator	Yes
Research questions?	93.2%
Inclusion/exclusion criteria	78.2%
Publication years	81.4%
Search scope	43.0%
Search string	68.7%
Inter-rater reliability	22.5%
Data extraction coding	32.6%
Quality assessment (SRs)	12.7%
Sufficient details	35.8%
Review limitations	42.3%

Evidence synthesis quality



Quality Indicator	Yes
Research questions?	93.2%
Inclusion/exclusion criteria	78.2%
Publication years	81.4%
Search scope	43.0%
Search string	68.7%
Inter-rater reliability	22.5%
Data extraction coding	32.6%
Quality assessment (SRs)	12.7%
Sufficient details	35.8%
Review limitations	42.3%

Evidence synthesis quality

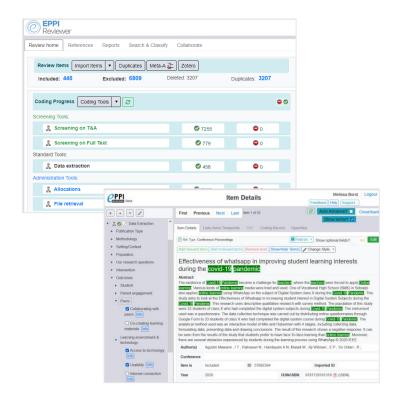


Quality Indicator	Yes
Research questions?	93.2%
Inclusion/exclusion criteria	78.2%
Publication years	81.4%
Search scope	43.0%
Search string	68.7%
Inter-rater reliability	22.5%
Data extraction coding	32.6%
Quality assessment (SRs)	12.7%
Sufficient details	35.8%
Review limitations	42.3%

Prevalence of DEST use

54.1% did not report using a tool

- > 5.2% used evidence synthesis software
 - Rayyan¹ (n = 7)
 - EPPI Reviewer² (n = 3)
 - Covidence³ (n = 3)
 - ASReview⁴ (n = 2)
 - DistillerSR 5 (n = 1)



- 1. https://www.rayyan.ai/
- 2. https://eppi.ioe.ac.uk/cms/Default.aspx?alias=eppi.ioe.ac.uk/cms/er4&
- 3. https://www.covidence.org/
- 4. https://asreview.nl/
- https://www.distillersr.com/products/distillersr-systematic-review-software

Prevalence of DEST use

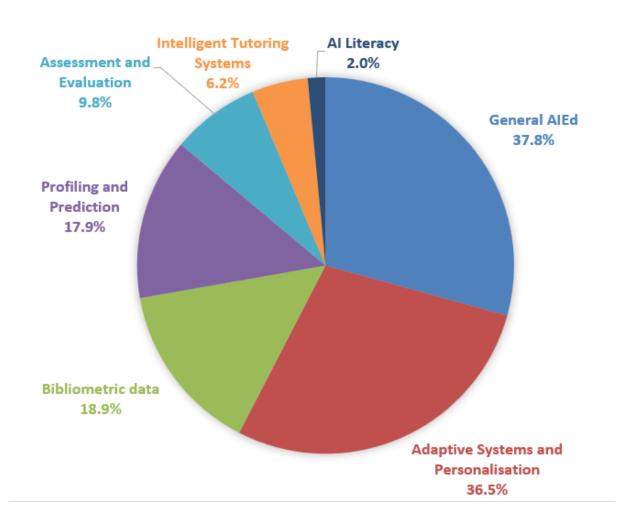
Most used tools		
Spreadsheet (Excel)	15.0%	
Reference management software	12.1%	
VOSViewer	6.5%	
R	5.2%	
Python	2.9%	
CiteSpace	2.6%	
Rayyan	2.3%	
Comprehensive Meta Analysis	1.6%	
Word	1.6%	
Gephi	1.3%	
STATA	1.3%	

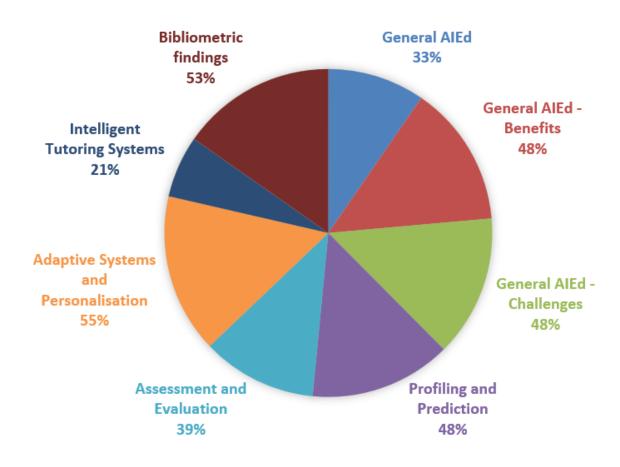
Most used tool by review type			
Systematic review	Spreadsheet		
Bibliometric analysis	VOSViewer		
Meta-analysis	CMA		
Literature review	RMS, Word		
Mapping review Spreadsheet			
Scoping review	RMS		

RMS = Reference Management Software

AlEd as a topic

Al applications in HE







Alhed Benefits & Challenges n = 31 General Alhed reviews

Top five benefits				
Personalised learning	12	38.7%		
Positive influence on learning	10	32.3%		
Reduced planning & admin	10	32.3%		
Greater insight into student understanding	10	32.3%		
Precise assessment / Greater equity	7	22.6%		

Top five challenges				
Lack of ethical consideration	9	29.0%		
Curriculum development	7	22.6%		
Infrastructure	7	22.6%		
Lack of teacher technical knowledge	7	22.6%		
Shifting authority	7	22.6%		

AIHEd Research Gaps (n = 66)

Top ten research gaps		
Ethical implications	27	40.9%
More methodological approaches	24	36.4%
More research needed (specific topics in Education)	22	33.3%
More research with a wider range of stakeholders	14	21.2%
Interdisciplinary approaches required	11	16.7%
Research limited to specific discipline areas	11	16.7%
More research in a wider range of countries	10	15.2%
Greater emphasis on theoretical foundations	9	13.6%
Longitudinal studies recommended	8	12.1%
Research limited to a few topics	8	12.1%

Implications

Call for ethics

- Data privacy and security
- Data bias
- Ethical AI in the curriculum

Call for collaboration

- Al development
- Curriculum development
- Researching AIEd

Call for rigour

- Rigorous primary & secondary research
- Need for updated evidence synthesis guidance for Educational research

AIHEd pre-print

A Meta Systematic Review of Artificial Intelligence in Higher Education: A call for increased ethics, collaboration, and rigour

Melissa Bond, Hassan Khosravi, Maarten De Laat, Nina Bergdahl, Violeta Negrea, Emily Oxley, Phuong Pham, Sin Wang Chong, and George Siemens

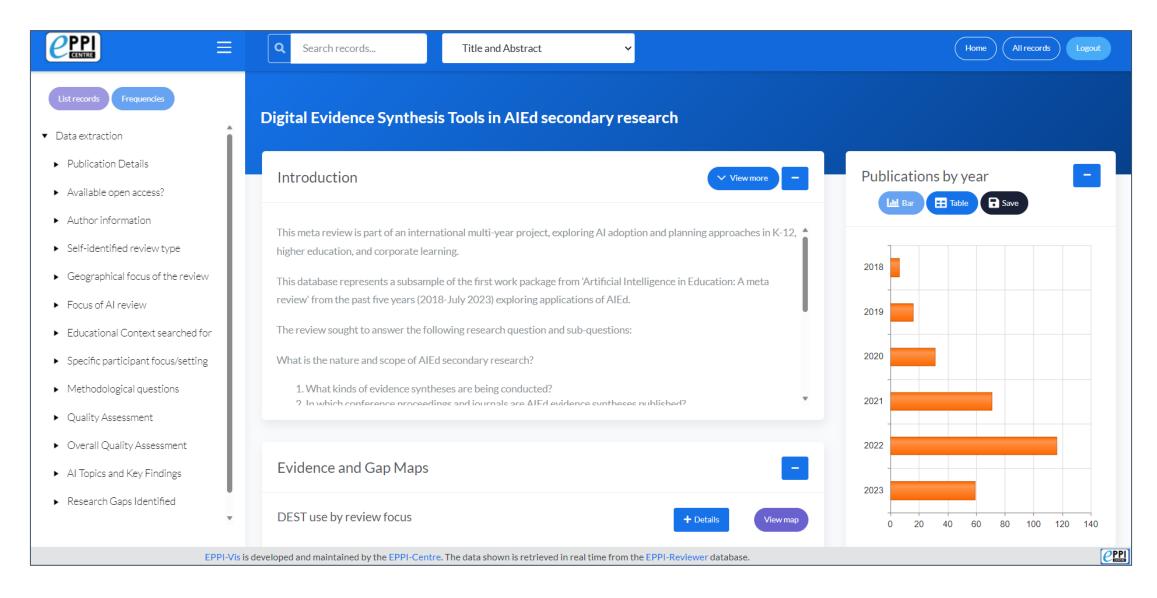
Available to read now:

http://dx.doi.org/10.13140/R G.2.2.31921.56162/1

Abstract.

Although the field of Artificial Intelligence in Education (AIEd) has a substantial history as a research domain, never before has the rapid evolution of AI applications in education sparked such prominent public discourse. Given the growing AIEd literature base in higher education, it is important to ensure that the field has a solid research and conceptual grounding as AI adoption increases. This review of reviews is the first comprehensive meta review to explore the scope and nature of AIEd in higher education (AIHEd) research, by synthesising secondary research (e.g., systematic reviews) 66 publications were included for data extraction and synthesis in EPPI Reviewer, which were predominantly systematic reviews (66.7%), published by authors from North America (27.3%), conducted in teams (89.4%) in mostly domestic-only collaborations (71.2%). Findings show that these reviews mostly focused on AIHEd generally (47.0%) or Profiling and Prediction (28.8%) as thematic foci, however key findings indicated a predominance of the use of Adaptive Systems and Personalisation in higher education. Research gaps identified suggest a need for greater ethical, methodological, and contextual considerations within future research, alongside interdisciplinary approaches to AIHEd application. Suggestions are provided to guide future primary and secondary research.



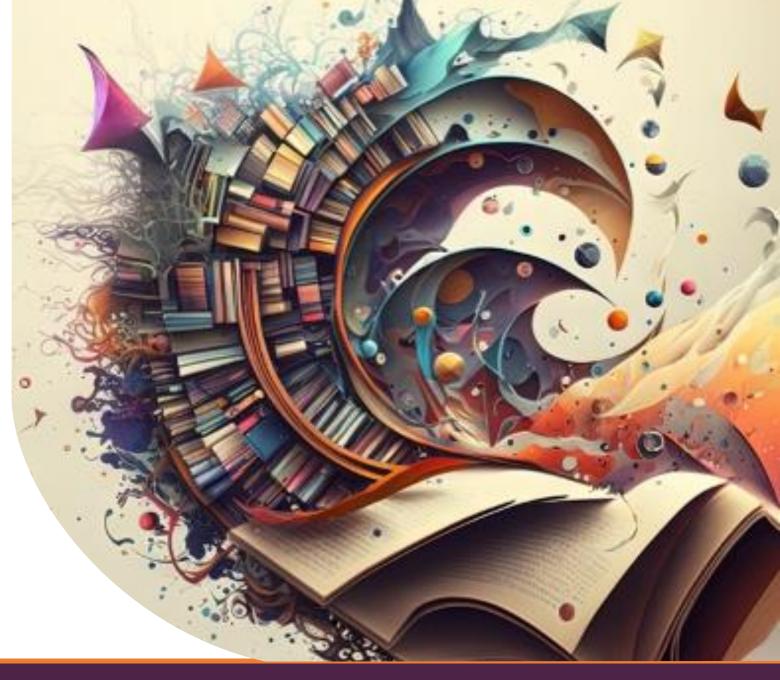


HE reviews - https://eppi.ioe.ac.uk/eppi-vis/login/open?webdbid=322



Generative Al in Education

The Case of Provisioning Immediate Feedback



Unpacking the Excitement Around Generative Al

Artificial Narrow Intelligence (ANI): Designed to perform a narrowly defined set of structured task. ANI has been successfully implemented in numerous fields.

IBM's Deep Blue beat world chess champion Garry Kasparov in the 1997 match

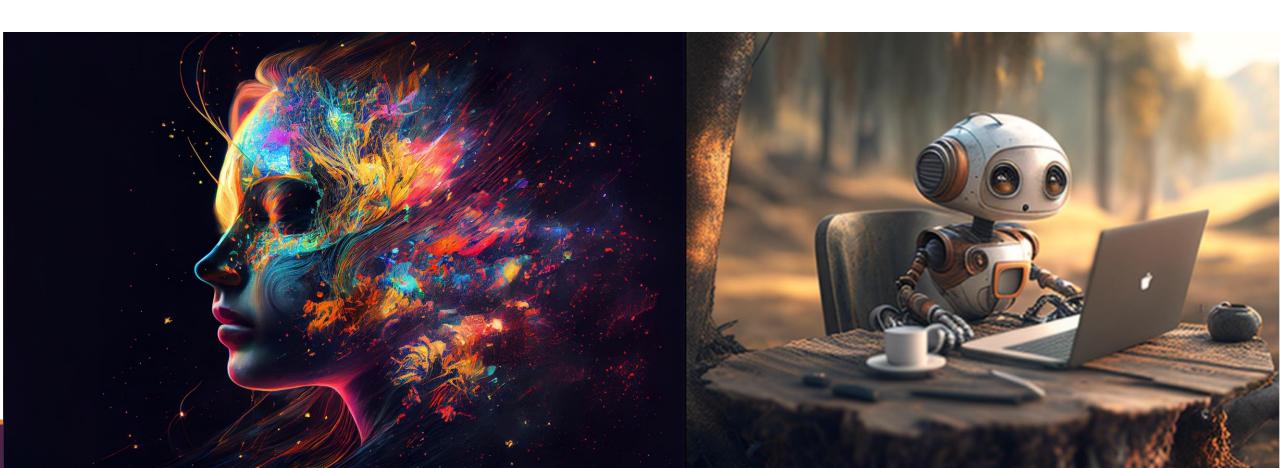
DeepMind's AlphaGo beat world Go Champion Lee Sedol in the 2022 match





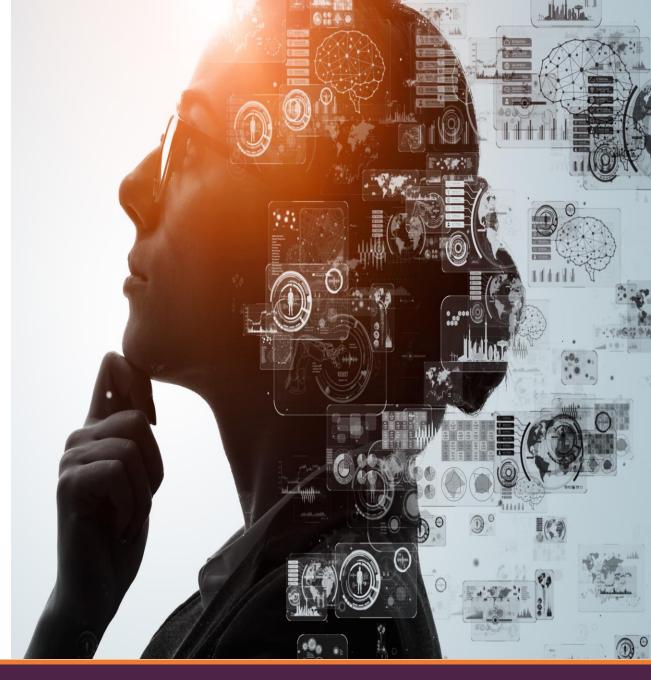
Unpacking the Excitement Around Generative Al

Artificial General Intelligence (AGI): Designed to perform across intellectual tasks similar to humans. Generative AI models represent our most significant progress towards achieving AGI.



Generative AI in Education

- Develop AI-powered tools that empower teachers to cater to diverse student needs, enhance learning experiences while reducing their workload
- Minimise the associated concerns with incorporating AI in education.
- Prepare students for an uncertain, turbulent, and networked world.



A Case Study from The University of Queensland

RiPPLE is an award-winning, UQ-developed learning tool that employs ethical and practical applications of AI to help teachers meet the learning needs of students while reducing their overall workload.



For more info see http://itali.uq.edu.au/ripple

How RiPPLE Works

RiPPLE leverages the science of learning to enhance student learning and experiences with three interconnected activities:



Empower students to craft study resources with AI assistance, nurturing their creativity and critical thinking skills

Facilitate peer evaluation of study resources with the support of AI, nurturing their metacognition and mentorship skills

Personalised practice

Help your students thrive as Al generates a unique study plan for each student based on their learning need

Why Use RiPPLE



Enhance learning

By engaging students in content creation, peer feedback and differentiated learning



Students as Partners

Recognises students as partners in learning.



Real-time feedback

by employing state-of-the-art generative AI systems



Empower learners

to become knowledge creators and critical evaluators



Metacognitive & Employability skills

Promotes evaluative judgment collaboration and communication.

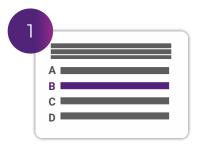


Reduce workload

by leveraging AI assistance and partnering with students

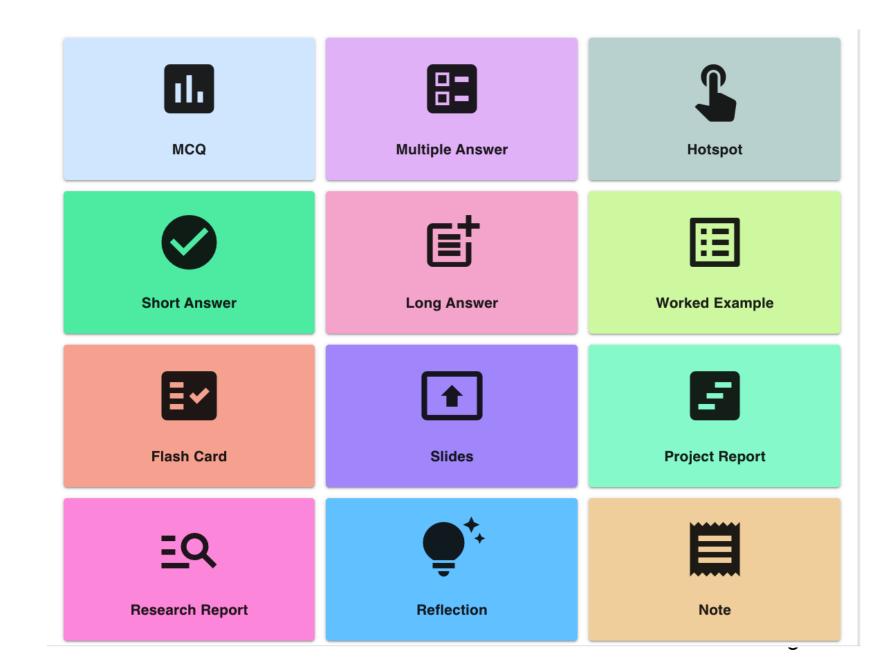


Create in RiPPLE

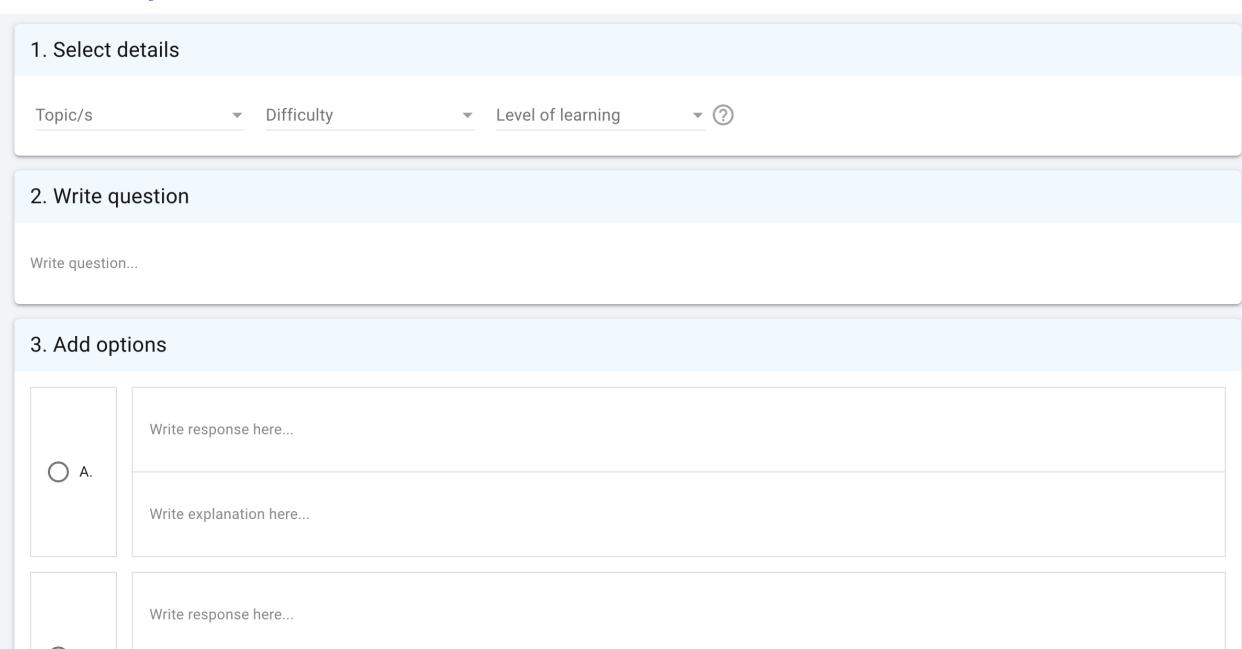


Creation

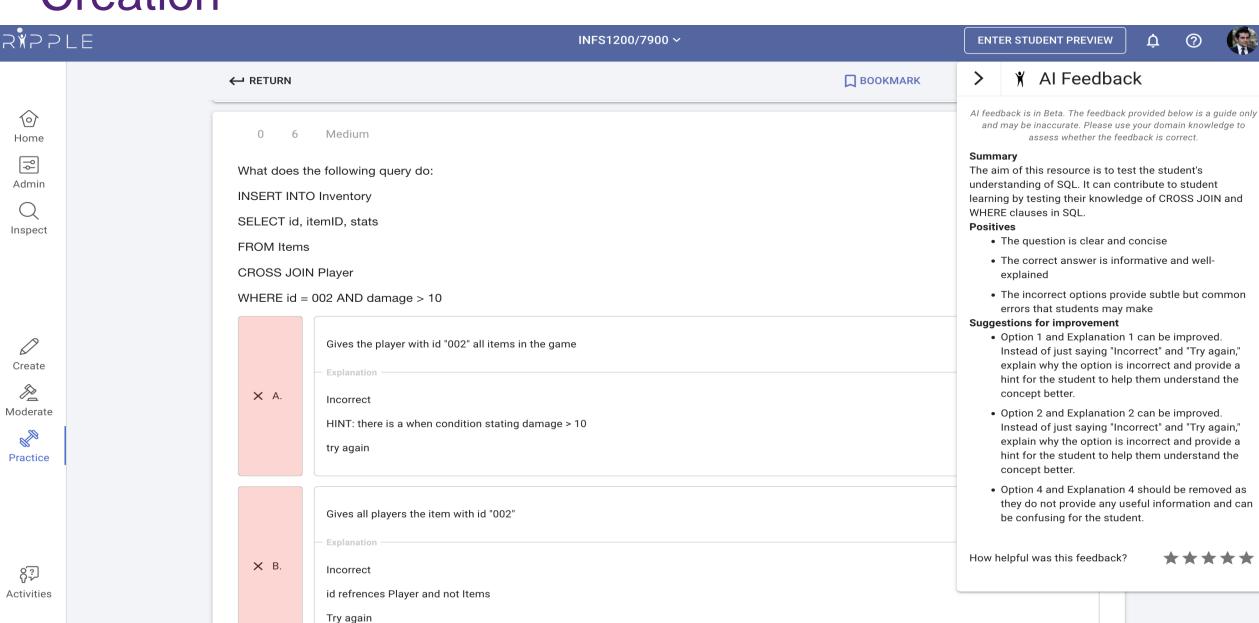
Empower students to craft study resources with AI assistance, nurturing their metacognition and critical thinking skills



Multiple Choice Question Creation



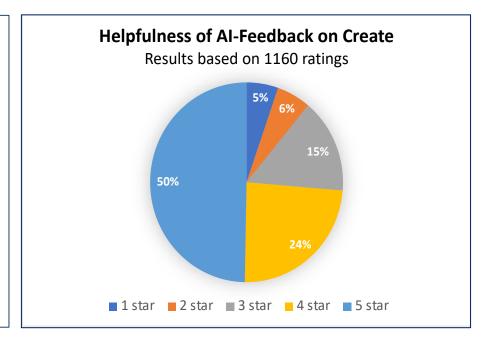
Creation



Student Perception of AI-Feedback on Create

Key highlights

- Al-feedback on create received 1,163 ratings with a 4.3 average.
- 74% of students assigned a 4 or 5-star rating, revealing its substantial utility among the majority.
- 15% of users expressed neutrality with a 3-star rating.
- 11% found it less beneficial, assigning a 1- or 2-star rating.



Representative comments:

- **Positive feedback**: "The suggestions are very reasonable and feasible, and the explanations are very specific, which helps me to improve the problems raised".
- **Neutral feedback**: "Feedback points 2 and 3 were useful and immediately corrected in the resource but feedback point 1 defeats the point of the question as the relationship degree consisting of three entities is the question being asked".
- Negative feedback: "Positives and suggestions contradict each other; suggestions make it less concise".

Moderate in RiPPLE

AI Feedback

The feedback provided below is a guide only and may be inaccurate. Please use your domain knowledge to assess whether the feedback is correct.

Hi there! I am The RiPPLE AI assistant, and I am here to help you with providing effective feedback.

Suggestions for improvement

- Instead of saying "I like how you explained your point," you can be more specific about what you liked. For example, you can say "I appreciate how you provided real-life examples to support your argument."
- When mentioning areas of improvement, try to phrase it in a constructive and positive manner. Instead of saying "Your writing is confusing," you can say "Your writing can be clearer if you organize your ideas in a more logical manner."
- Consider providing specific suggestions for improvement. For instance, instead of saying "Your conclusion could be better," you can say "To improve your conclusion, you can summarize your main points and provide a thought-provoking ending."

How helpful was this feedback?

I don't want to moderate this resource

Resource Feedback

Please evaluate the resource based on the following criteria:

Correctness and precision:		Poor	Needs Improvement	Satisfactory	Great	Outstanding
Quality of Question:	②	Poor	Needs Improvement	Satisfactory	Great	Outstanding
Quality of Options:	②	Poor	Needs Improvement	Satisfactory	Great	Outstanding
Quality of Explanation:	?	Poor	Needs Improvement	Satisfactory	Great	Outstanding

What is good about this resource?

Please list each aspect as a separate dot point and align feedback to the rubric. Be specific & detailed. Use constructive language. Leave blank if not applicable.

It is a comprehensive resource with a code sample and output.
 covers an important topic

+ Add feedback...

How can this resource be improved?

Please list each aspect as a separate dot point and align feedback to the rubric. Be specific & detailed. Use constructive language. Leave blank if not applicable.

It is already a pretty good resource
 Add suggestion...

×

Further comments

keep up the good work



Decision

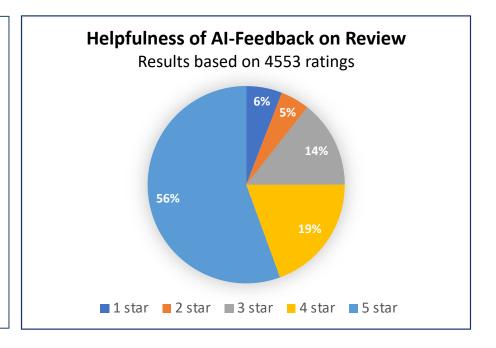
Please rate the overall quality of this resource based on the criteria above.

The overall quality of this resource is:	Poor	Needs Improvement	Satisfactory	Great	Outstanding
Rate your confidence in assessing this resource:	Very Iow	Low	Medium	High	Very High

Student Perception of Al-Feedback on Review

Key highlights

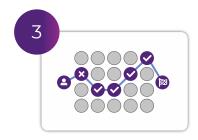
- Al-feedback on create received 4,553 ratings with a 4.4 average.
- 75% of students assigned a 4 or 5-star rating, revealing its substantial utility among the majority.
- 14% of users expressed neutrality with a 3-star rating.
- 11% found it less beneficial, assigning a 1- or 2-star rating.



Representative comments:

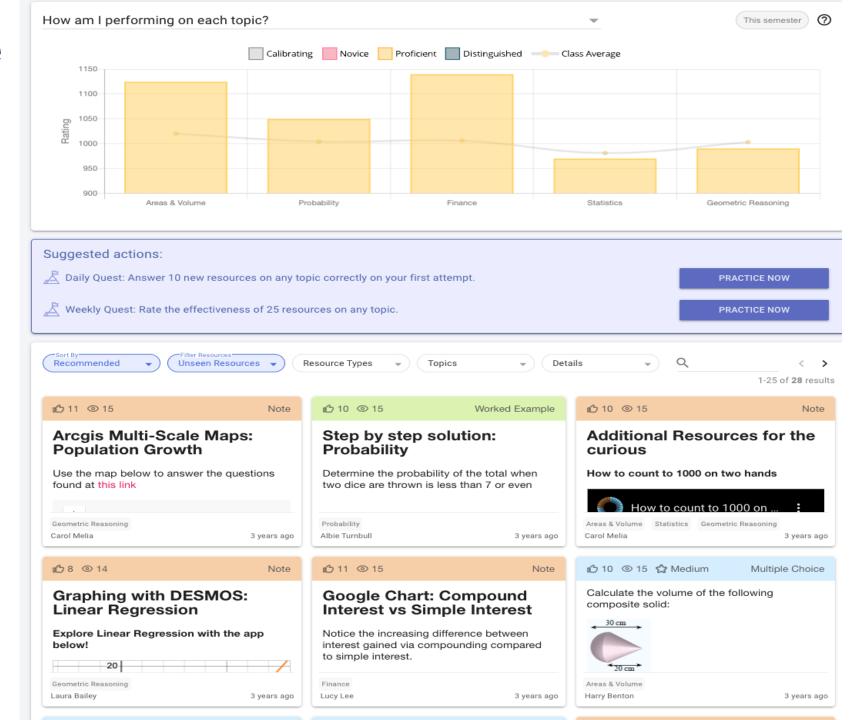
- **Positive feedback**: "This feedback was very helpful as it gave me the answer I was looking for but didn't know how to put into words".
- **Neutral feedback**: "Dot point 2 would be good to improve my feedback and make it more specific. I think the AI did not quite understand dot points 1 and 3, as I feel that I clarified these in my feedback".
- **Negative feedback**: "The suggestions for improvement was the same wording as my response with no explanation for why this needs improvement".

Adaptive Practice



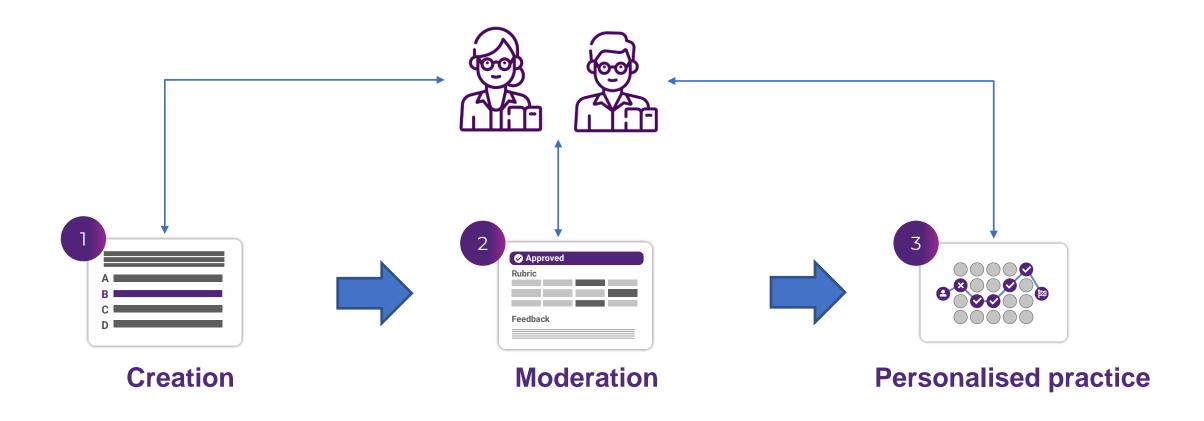
Personalised practice

Help your students thrive as Al generates a unique study plan for each student based on their learning need



Educator Oversight

RiPPLE leverages AI to empowers educators to effectively oversee student learning across creation, moderation and personalised practice activities



Inspect in RiPPLE



Inspect

The inspect page empowers instructors to effectively use their time by leveraging out AI algrithms to find and examine resources where incorrect decisions may have been made or peer reviews are of poor quality

Inspect resources

Inspect student resources to verify their correctness, helpfulness and quality.

Resource inspection history

View a log of your resource inspections

Inspect moderations

Identify & manage the quality of student moderations.

Moderation inspection history

View a log of your moderation inspections

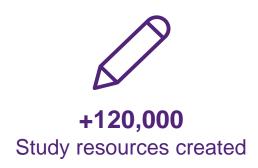
Inspect students

View the user profile of your students

Adoption and Recognition









RiPPLE is increasingly being recognised as an exemplar tool for using AI in education.







Academics from **over 50**Universities have created
RiPPLE accounts



Provisional **patent** submitted



Featured as an exemplar educational technology by **EDUCAUSE**

Effects on Student Learning and Experience

Enhance learning

Students who use RiPPLE gain a 10% improvement in grades.



A survey of 500 students showed enjoyment, learning benefits, and a desire to use RiPPLE in other courses.



RiPPLE's AI assistance improves student authorship, feedback literacy and personal practice



References

Bond, M., Khosravi, H., De Laat, M., Bergdahl, N., Negrea, V., Oxley, E., Pham, P., Chong, S.W., & Siemens, G. (2023). A meta systematic review of Artificial Intelligence in Higher Education: A call for increased ethics, collaboration, and rigour. *Pre-print*. http://dx.doi.org/10.13140/RG.2.2.31921.56162/1

Centre for Reviews and Dissemination (UK). (1995). *Database of Abstracts of Reviews of Effects (DARE): Quality-assessed Reviews*. https://www.ncbi.nlm.nih.gov/books/NBK285222/. Accessed 4 January 2023.

Daudt, H. M. L., van Mossel, C., & Scott, S. J. (2013). Enhancing the scoping study methodology: A large, inter-professional team's experience with Arksey and O'Malley's framework. *BMC Medical Research Methodology*, 13, 48. https://doi.org/10.1186/1471-2288-13-48

Jardim, P.S.J., Rose, C.J., Ames, H.M. *et al.* (2022). Automating risk of bias assessment in systematic reviews: a real-time mixed methods comparison of human researchers to a machine learning system. *BMC Medical Research Methodology*, 22, 167. https://doi.org/10.1186/s12874-022-01649-y

Kitchenham, B., Pearl Brereton, O., Budgen, D., Turner, M., Bailey, J., & Linkman, S. (2009). Systematic literature reviews in software engineering – A systematic literature review. *Information and Software Technology*, *51*(1), 7–15. https://doi.org/10.1016/j.infsof.2008.09.009

Lai, J. W., & Bower, M. (2020). Evaluation of technology use in education: Findings from a critical analysis of systematic literature reviews. *Journal of Computer Assisted Learning*, 36(3), 241–259. https://doi.org/10.1111/jcal.12412

Liu, B. (2023). Chinese University Students' Attitudes and Perceptions in Learning English Using ChatGPT. *International Journal of Education and Humanities*, 3(2), 132–140. https://doi.org/10.58557/(ijeh).v3i2.145

Marshall, C., Kitchenham, B., & Brereton, P. (2018). Tool features to support systematic reviews in software engineering – A cross domain study. *E-Informatica Software Engineering Journal*, 12(1), 79-115. https://doi.org/10.5277/e-Inf180104



References

Shea, B. J., Reeves, B. C., Wells, G., Thuku, M., Hamel, C., Moran, J., Moher, D., Tugwell, P., Welch, V., Kristjansson, E., & Henry, D. A. (2017). Amstar 2: A critical appraisal tool for systematic reviews that include randomised or non-randomised studies of healthcare interventions, or both. *BMJ (Clinical Research Ed.)*, 358, j4008. https://doi.org/10.1136/bmj.j4008

Smolansky, A., Cram, A., Raduescu, C., Zeivots, S., Huber, E., & Kizilcec, R. (2023). Educator and Student Perspectives on the Impact of Generative AI on Assessments in Higher Education. In Proceedings of the Tenth ACM Conference on Learning @ Scale (L@S '23), July 20 1–22, Copenhagen, Denmark. ACM, New York, NY, USA, 5 pages. https://doi.org/10.1145/3573051.3596191

Stansfield, C., Stokes, G., & Thomas, J. (2022). Applying machine classifiers to update searches: Analysis from two case studies. *Research Synthesis Methods*, *13*(1), 121-133. https://doi.org/10.1002/jrsm.1537

Tsou, A.Y., Treadwell, J.R., Erinoff, E. *et al.* (2020). Machine learning for screening prioritization in systematic reviews: comparative performance of Abstrackr and EPPI-Reviewer. *Systematic Reviews*, 9, 73. https://doi.org/10.1186/s13643-020-01324-7

Zawacki-Richter, O., Marín, V.I., Bond, M. *et al.* (2019). Systematic review of research on artificial intelligence applications in higher education – where are the educators?. *International Journal of Educational Technology in Higher Education, 16,* 39, 1-27. https://doi.org/10.1186/s41239-019-0171-0

Contact Details

Dr Melissa Bond

Email: <u>melissa.bond@ucl.ac.uk</u>

Twitter: https://twitter.com/misc_nerd

Website: http://drmelissabond.weebly.com/

ResearchGate: https://www.researchgate.net/profile/Melissa-Bond-5

LinkedIn: https://www.linkedin.com/in/bondmelissa/



Dr Hassan Khosravi

Email: <u>h.khosravi@uq.edu.au</u>

Twitter: https://twitter.com/haskhosravi

ResearchGate: https://www.researchgate.net/profile/Hassan-Khosravi-5

LinkedIn: https://www.linkedin.com/in/hassan-khosravi/

