



# Does GenAI Write Good Science, and Does It Know Whether It Can?

Exploring the Ability of GenAI to Write and Evaluate Scientific Text

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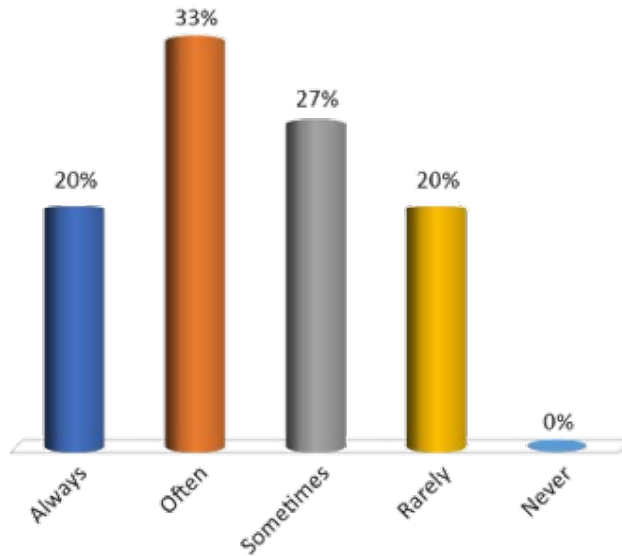
# Introduction

# The Rise of Generative Artificial Intelligence

- Large Language Models (LLMs) are impacting a wide variety of industries, including education and science
- LLMs are being used to draft, edit and refine outputs but can it evaluate the quality of scientific writing?
- Questions remain about their effectiveness, limitations, and potential biases

# Why do we need to worry?

16 to 18-year-olds are a lot more engaged with AI



How often do you use AI tools?

# What about current students?

## UG students know about it...

53% used AI to help with assessments

65% think institutions should not accept AI generated work

73% expect to use AI after graduation

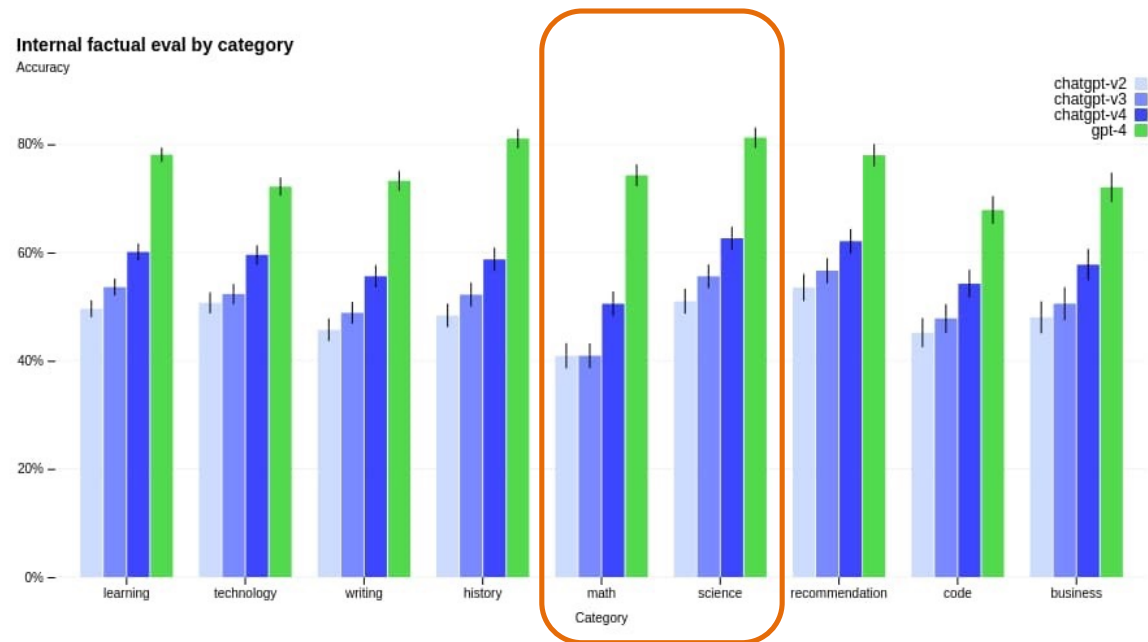
Only 9% said institutional approaches have changed

# Why this matters

- Educational impact: Can AI support or hinder learning in academic contexts
- Equity of access: Does the difference between free and premium models exacerbate inequalities
- AI as a tool: How can educators and students integrate AI effectively in teaching and learning?

# Purpose of the Study

- To assess whether GenAI can produce high quality scientific essays
- To evaluate whether GenAI can critique and mark essays it generates
- To explore the implications of free vs premium GenAI models in academic settings



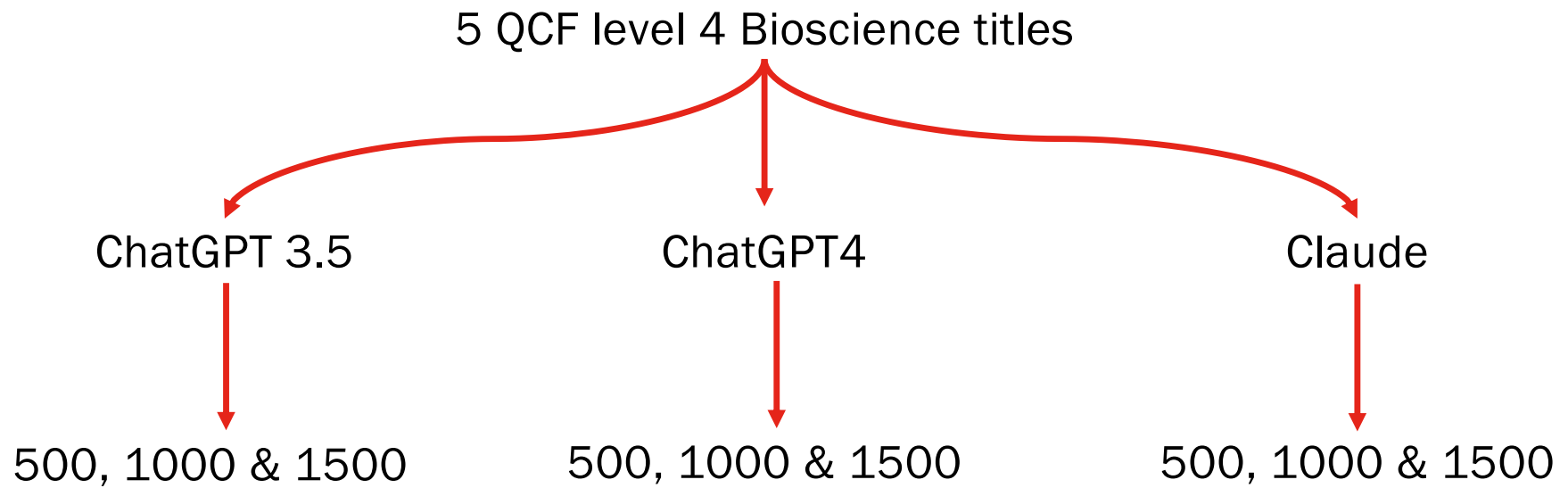




# Methods



# AI Generation of Essays



Total of 45 essays, 9 from each title.

# AI Generation of Essays - Prompting

- “imagine you are a 1<sup>st</sup> year university student; I want you to write an essay based on this title: *insert title*.”
- “Can you use Harvard referencing for your sources through the essay and also provide a reference list?”
- “Can you make the essay 500 words? (Excluding references)”
- “Can you make the essay 1000 words? (Excluding references)”
- “Can you make the essay 1500 words? (Excluding references)”
  
- All essays were generated on the same date

# Human Marking

- Each essay was 'blinded' so that the GenAI used was unknown to the markers
- Each essay was marked independently by 3 human markers who were final year undergraduate students
- Markers also provided qualitative feedback/justification of their mark
- Standardisation of marking was carried out as a group with guidance from an academic member of staff
- An established rubric/marketing criteria was used for mark generation – the criteria already in use for summative assessment of level 4 students using the same essay titles.

# Generative AI Marking

- “I am going to provide you with an essay marking rubric for a first-year essay at a university. I want you to analyze the document and then provide me with a summary of the five marking criteria and their weighting so I can check you have interpreted it correctly.”
- I am now going to provide you with an example of a full essay, and I want you to tell me within which grade level from ‘fail’ to ‘exceptional first’ you think it falls for each of the criteria: *pasted essay text*”
- Each model was reloaded after each input.
- The models were not able to provide quantitative marks like the human markers so the rubric was modified as follows:

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# Generative AI Marking – Adapted from a figure by Ahmed Al-Sammere

BI1001 - Spring Essay		Weight	0	15	25	35	42	45	48	52	55	58	62	65	68	72	75	78	85	95	100	
<b>KNOWLEDGE AND UNDERSTANDING and COGNITIVE SKILLS</b>																						
Relevant to the	<b>Knowledge and Understanding</b>	30%	Fail				3rd			2:2			2:1			1st			High First		Exceptional First	
	<b>Critical Analysis</b>	10%	18				45			55			65			75			85		98	
<b>TRANSFERABLE/EMPLOYABILITY SKILLS</b>																						
assessment).	<b>Organisation &amp; Communication</b>	20%																				
	<b>Presentation skills</b>	15%																				
<b>ACADEMIC SKILLS</b>																						
	<b>Literature and Referencing</b>	25%																				





# Results

# Three Way ANOVA results showing the impact of each factor and significant interactions (P<0.05)

Adapted from an original figure by Mollie Ridge

	Degrees of Freedom	Sum of Squares	Mean Squares	F value	P - Value
Human Marker	2	38.9	19.43	0.483	0.62120
Essay Length	2	122.7	61.34	1.526	0.23281
Essay Subject	4	471.3	117.83	2.931	0.03583*
AI Model	2	581.1	290.56	7.228	0.00257**
<b>Significant Interactions</b>					
Human marker & Essay Subject	8	961.6	120.19	2.990	0.01277**
AI Model & Essay Length	4	604.5	151.12	3.759	0.01287**

# Human Awarded Marks for AI generated essays

- There was a significant impact of AI model type on the awarded mark ( $P < 0.01$ )
- Average essay marks were 3<sup>rd</sup> class
- Claude averaged the lowest marks and ChatGPT 4.0 the highest marks
- Impact of different human markers not significant
- Significant interaction between humans and essay title ( $P < 0.05$ )
- Significant variation between essay title ( $P < 0.05$ ) but explained by Claude's lower performance in some titles.

# Impact of Essay length on Human Awarded Marks

- Effect of essay length was not significant
- Interaction between AI used and essay length was significant ( $P < 0.05$ )
- Claude gained lower marks in the 500 word essays
- Little other effect of essay length with ChatGPT 3.0 performing slightly better at 1000 words and relatively consistent results for ChatGPT 4.0
- AI generated surprisingly similar word counts regardless of the length prompt

# Qualitative Assessment of AI Generated Essays – content and presentation

- Lack of scientific detail
- Little or no discussion
- Reads like a list converted to prose
- No Figures
- Characteristic ‘awkward’ introductions

*“This essay will adhere to the Harvard referencing style and provide a reference list, while aiming to be informative and academically rigorous.”*

# Qualitative Assessment of AI Generated Essays – recognition of sources

- Frequent use of ‘imagined’ references
- Real references but from irrelevant work by a real author, but who had worked in the area.
- Genuine titles but imagined authors
- Missing authors in the reference list
- Citations only appearing at the end of a paragraph
- Longer essays increased accuracy of referencing and some 1500 word essays had no significant errors.



# Human and AI Awarded Marks for AI generated Essays.

- Average human awarded markers for AI generated essays (P<0.01)
- Claude - 41%
- ChatGPT 3.0 - 44%
- ChatGPT 4.0 - 46%
- Average AI awarded marks for AI generated essays
- Claude - 68%
- ChatGPT 3.0 - 68%
- Chat GPT 4.0 - 68%

The pattern of marks awarded to each essay was different for Human and AI awarded marks.

# Qualitative Assessment of AI Marking of AI Essays

- AI marks were higher in every criteria of the rubric
- Presentation Skills. May have assumed that figures were present and accurate even if not present
- Academic Skills – referencing, did not identify flaws in referencing



## **Discussion and Conclusions**

# GenAI Performance in Essay Writing

- GPT4 consistently outperformed other models (Claude and GPT3.5) across most variables
- Essays averaged 3rd class marks – human evaluation
- Limitations in GenAI's ability to produce high-quality scientific writing from zero-shot prompts

## **Essay Length and Subject**

- Essay length had limited impact on performance
- Longer essays showed slight improvements in referencing accuracy
- Certain essay subjects were handled better by GenAI
- High variability between models

## **GenAI vs Human Marking**

- GenAI consistently awarded higher marks compared to human markers
- GenAI marking failed to identify referencing and formatting flaws

# Strengths and Weaknesses of GenAI

## Strengths:

- Efficiency in generating coherent, structured prose
- Potential as a supplementary tool for students, particularly in generating drafts or structuring arguments

## Weaknesses:

- Lack of critical analysis and depth in content
- Over-reliance on fabricated or inappropriate references
- Inconsistent performance across topics and essay lengths



# Strengths and Weaknesses of GenAI

## Equity Considerations:

- Differences between free and premium models may widen educational inequalities

## Educational Potential:

- GenAI can serve as a teaching aid but cannot replace human expertise
- Encouraging transparent use of GenAI is essential to maintain academic integrity

# Conclusions

## Summary of findings

- GenAI is a promising tool, but not yet a replacement for human scientific writing or evaluation
- Current GenAI models produce basic scientific text but lack depth and detail for higher academic outcomes

## Recommendations

- **For Students:** Use AI for initial brainstorming and draft generation but review critically for accuracy and depth
- **For Educators:** Incorporate AI literacy into curricula to help students use these tools effectively and ethically

# Current and Future Work

## Current project:

- Can training enhance the ability of GenAI to evaluate and mark scientific text
- Is the accuracy of referencing improving through dedicated referencing tools?

## Future project:

- Long-term studies to evaluate the impact of AI-assisted writing on learning outcomes



## **Acknowledgements**

**Ahmed Al Samere**  
**Mollie Ridge**

# **OU and NCFE Evaluation Report**

<https://law-school.open.ac.uk/sites/law-school.open.ac.uk/files/files/OU%20NCFE%20report%20on%20GAI%20and%20assessment.pdf>