

# Introduction to AI-Enabled Teaching

Facilitated by  
Cassie Collier, Assistant Professor  
Computer Management and Information Systems

*Adapted from Harvard Business School's Teaching with AI Program*

# Agenda



What is AI?



Guidelines for the use of AI for teaching



Activity: Finding the jagged frontier of AI in education tasks



Retrospective & next steps

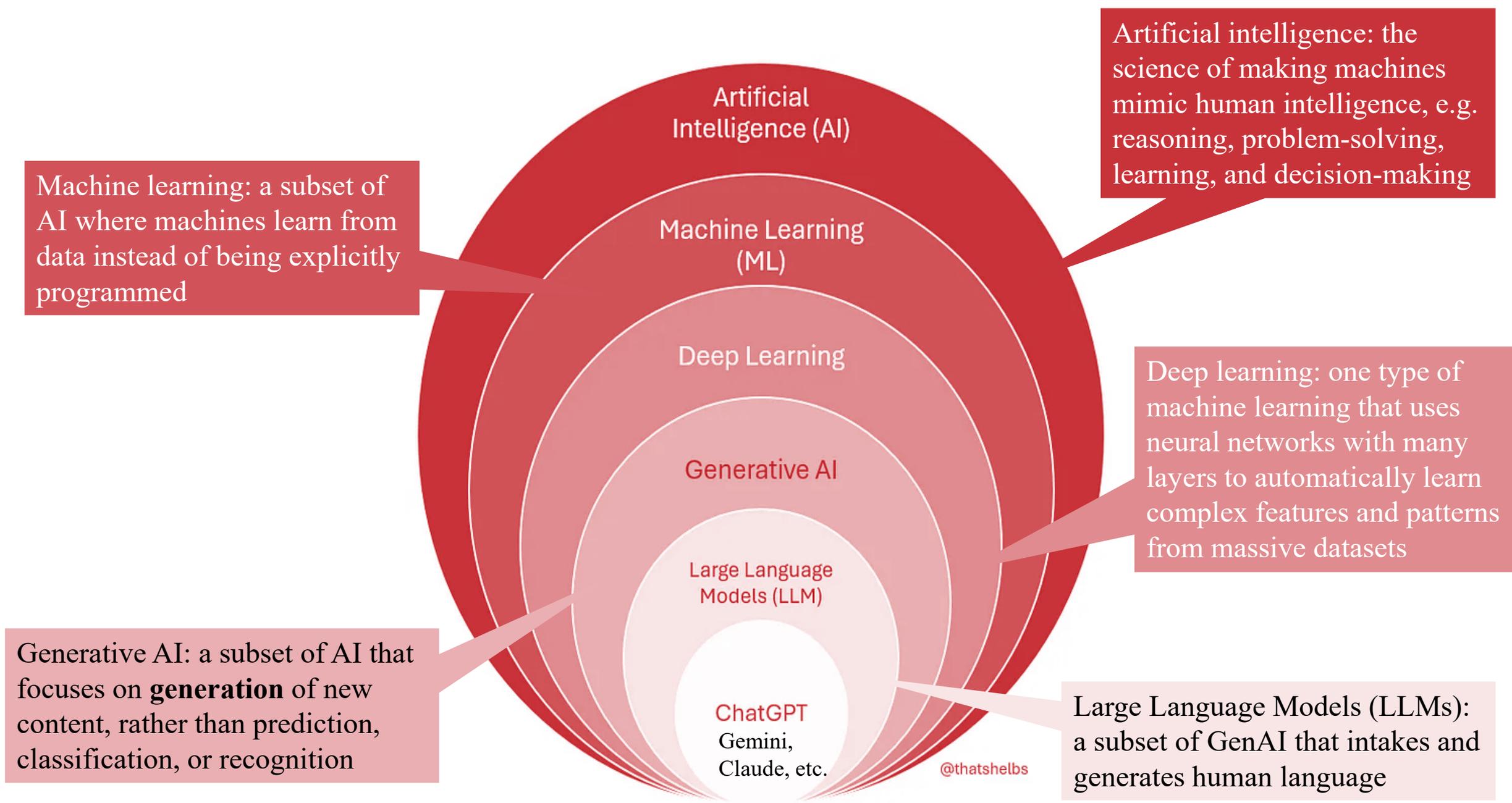
# A couple disclaimers

Everyone in this room has experience with / knowledge of AI. I'm here as a facilitator to help us all gain value from our collective experience and knowledge; not as a lecturer to purely deliver content

The ethics and values of AI use is outside the scope for this conversation. There are many ethical reasons to be resistant to the use of AI. Do not take my excitement for the topic as a wholehearted endorsement of all things AI.



# What is AI?



Machine learning: a subset of AI where machines learn from data instead of being explicitly programmed

Artificial intelligence: the science of making machines mimic human intelligence, e.g. reasoning, problem-solving, learning, and decision-making

Deep learning: one type of machine learning that uses neural networks with many layers to automatically learn complex features and patterns from massive datasets

Generative AI: a subset of AI that focuses on **generation** of new content, rather than prediction, classification, or recognition

Large Language Models (LLMs): a subset of GenAI that intakes and generates human language

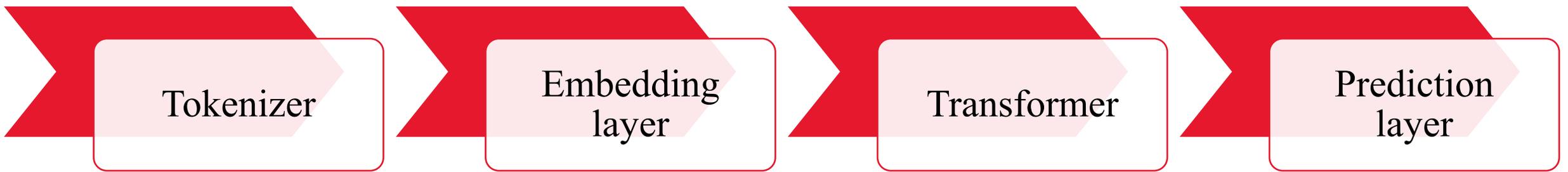
@thatshelbs

*Large Language Models are  
Next Token Predictors*

Roses are **red**, violets are blue

*How did you know that?*

# Four Elements of an LLM



Tokenizer

Embedding  
layer

Transformer

Prediction  
layer

# Four Elements of an LLM



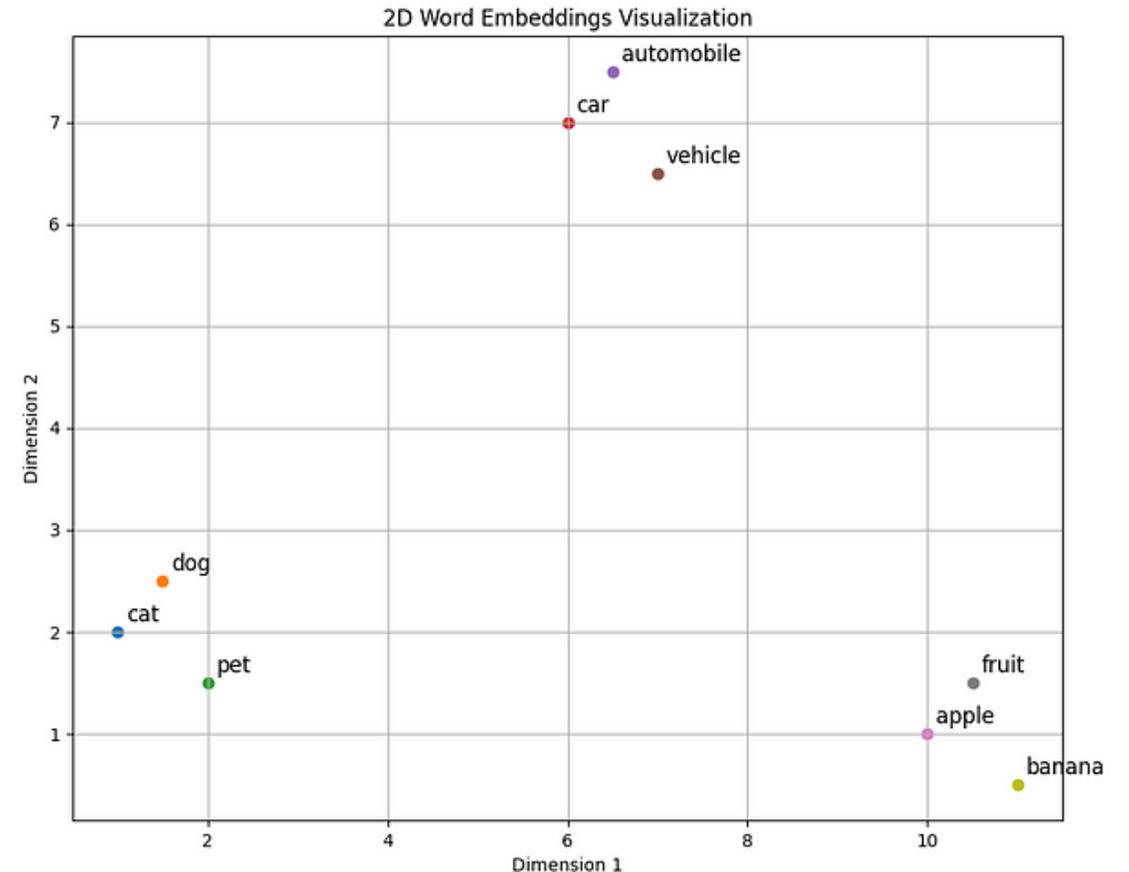
Input text is broken up into individual **tokens** to be analyzed (words, punctuation, brief phrases)

The screenshot shows a web interface for text analysis. At the top, a text input field contains the question "What is a Large Language Model?". Below the input field are two buttons: "Clear" and "Show example". Underneath the buttons, the statistics "Tokens 7" and "Characters 31" are displayed. At the bottom, a text area shows the input text "What is a Large Language Model?" with each word highlighted in a different color (green, yellow, orange, red, blue, purple, green). Below this text area are two buttons: "Text" and "Token IDs".

# Four Elements of an LLM



Tokens are evaluated for their **relatedness** based on how often (statistically) they occur together in training data. GPTs are pre-trained in these embeddings based on data from the internet and other sources.



# Four Elements of an LLM



The **bat** flew out of the **cave**.

Vs.

He hit the **ball** with his **bat**.



Embeddings for each token are mixed together to provide **context** (resulting in better embeddings called contextualized embeddings).

# Four Elements of an LLM



*The best thing about AI is its ability to*

learn	4.5%
predict	3.5%
make	3.2%
understand	3.1%
do	2.9%

Produces a huge list of probabilities for the next token based on contextualized embeddings and makes a selection.

# Four Elements of an LLM

## Tokenizer

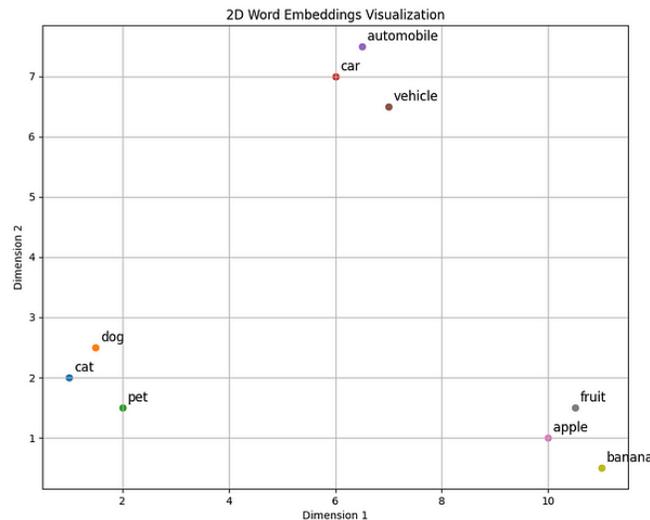
Input text is broken up into individual **tokens** to be analyzed (words, punctuation, brief phrases)

Tokens: 7  
Characters: 31

What is a Large Language Model?

## Embedding layer

Tokens are evaluated for their **relatedness** in an embedding layer. More related = closer together. GPTs are pre-trained in these embeddings



## Transformer

Embeddings for each token are mixed together to provide **context** (resulting in better embeddings called contextualized embeddings).



*The best thing about AI is its ability to*

## Prediction layer

Produces a huge list of probabilities for the next token based on contextualized embeddings. What we ultimately see produced by an LLM is the token with an acceptably high **probability** of being the “right” next token.

learn	4.5%
predict	3.5%
make	3.2%
understand	3.1%
do	2.9%

# It's not magic or human thinking – it's math!

... which is why sometimes it doesn't work

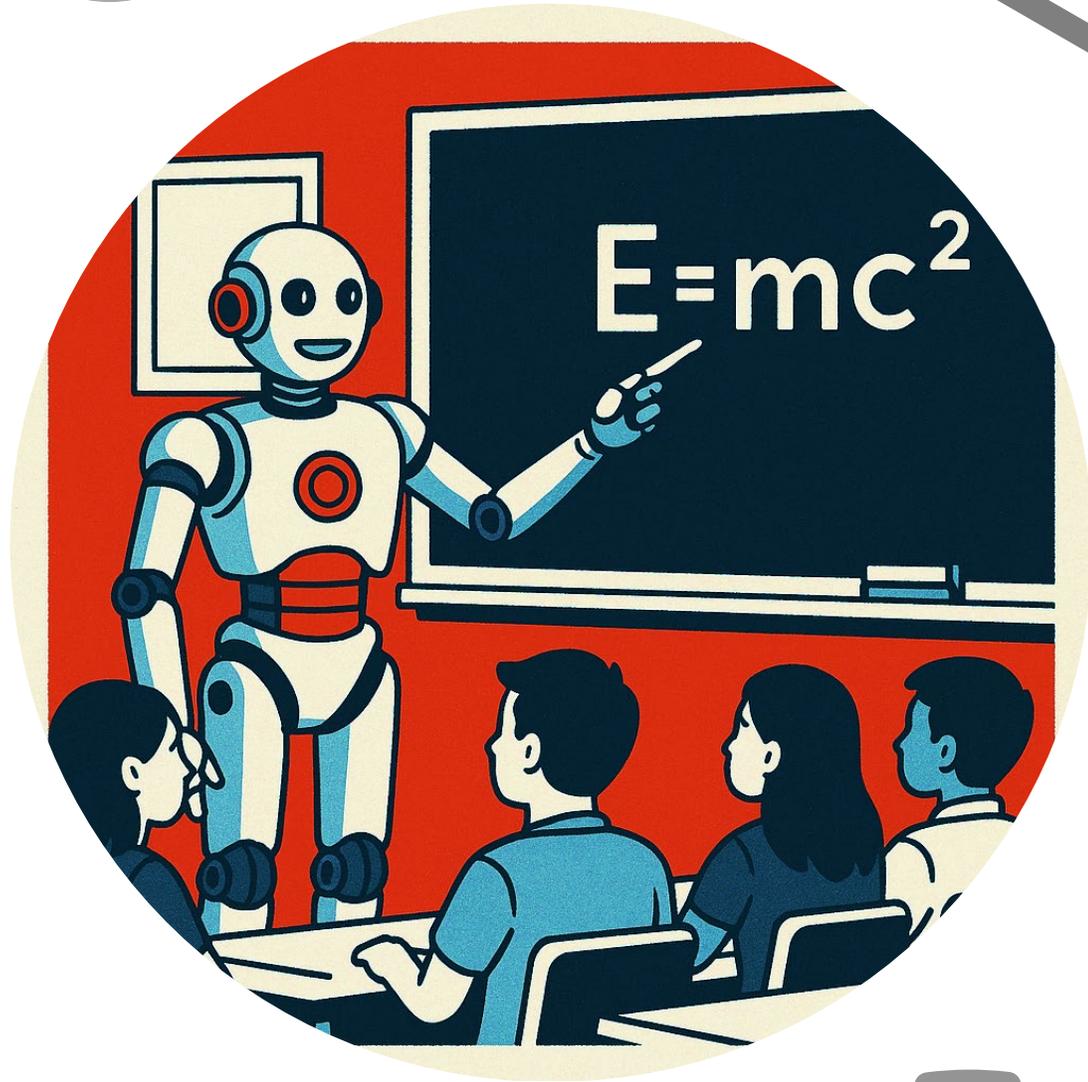
## Exposing The Brittleness Of Generative AI As Exemplified By The Recent Gibberish Meltdown Of ChatGPT

By [Lance Eliot](#), Contributor. © Dr. Lance B. Eliot is a world-renowned AI scienti... ▼

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Published Feb 25, 2024, 07:00am EST

- “When one user inquired what a computer was, ChatGPT allegedly replied, “It does this as the good work of a web of art for the country, a mouse of science, an easy draw of a sad few, and finally, the global house of art, just in one job in the total rest.” (Source: “ChatGPT glitches out: Rogue AI responding in nonsensical Spanglish, gibberish” by Ben Cost, *New York Post*, February 20, 2024).



# Using AI in Teaching

*Photo generated by ChatGPT*

# If students use AI, will they still be learning?

## The Impact of Generative AI on Critical Thinking: Self-Reported Reductions in Cognitive Effort and Confidence Effects From a Survey of Knowledge Workers

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Microsoft Research Cambridge  
Cambridge, United Kingdom  
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Publication

**Your Brain on ChatGPT: Accumulation of Cognitive Debt when Using an AI Assistant for Essay Writing Task**

# Every new technology has brought concerns that it will lead humans to stop using our brains



Writing

Plato: *They will cease to exercise memory because they rely on that which is written, calling things to remembrance no longer from within themselves, but by means of external marks.*



Cell phones

*New York Times, 2005: **HEY, cellphone user, when was the last time you memorized a phone number?***



Calculators

*New York Times, 1985: ...calculators could lead to "speed and ease, but not to learning."*



# The 10x Educator

The goal: **Proper** use of AI can increase our efficiency and effectiveness by 10 times

# What is the “proper” use of AI?



Demonstrating  
good values /  
adhering to policy



Technically  
proficient  
interaction with AI  
(crafting input +  
evaluating output)

# (A few) Considerations Regarding AI Use in Education

1. Research & adhere to legal limitations – copyright, FERPA, etc. ([\*Future FDCI AI seminars\*](#) will address fair use & intellectual property etc.)
2. Enact explicit policies regarding acceptable student use of AI for a given program, class, and/or assignment (*CFDI has sample policies available for use online now in their [AI Resources](#)*)

# GenAI Usage Policy Example

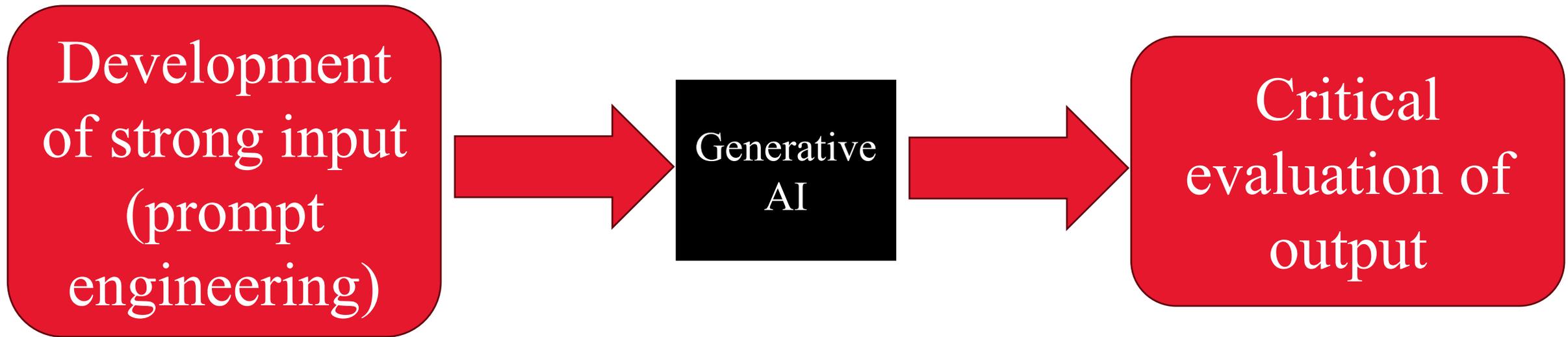
	Level descriptor:	What this looks like in practice:
Level 0	No use of GenAI	Students will create their own, original work without the use of GenAI for any manner.
Level 1	Organizational use of GenAI	Students will create their own, original work without the use of GenAI; however, the use of GenAI for personal efficiency (i.e., summarizing notes/readings, clarifying content) is acceptable.
Level 2	Use of GenAI for brainstorming or idea generation	Students can consult GenAI as a tool for brainstorming or idea generation, but are expected to create their own, original work without the use of GenAI.
Level 3	Use of GenAI for feedback	Students create their own work, then use GenAI as a tool to provide feedback on their work. Students are expected to use feedback from GenAI to conduct their own revisions of their own work, so any work submitted should be GenAI-supported, not Gen-AI created.
Level 4	Use of GenAI to co-create and revise work	Students can use GenAI to develop drafts/outlines of their work but are expected to carefully edit and revise GenAI-created content as appropriate for their learning context. It is expected that any use of GenAI-created content is properly disclosed and attributed.
Level 5	Unrestricted, attributed use of GenAI	Students can freely use GenAI if the use of any GenAI-created content is properly disclosed and attributed.
Level 6	Unrestricted, unattributed use of GenAI	Students can freely use GenAI in any form. Attribution is not necessary. <i>Note: While it is possible to use GenAI in an unrestricted manner without attribution, any who apply this level of GenAI use in their course should carefully consider ethical and legal implications of such AI use.</i>

# (A few) Considerations Regarding AI Use in Education

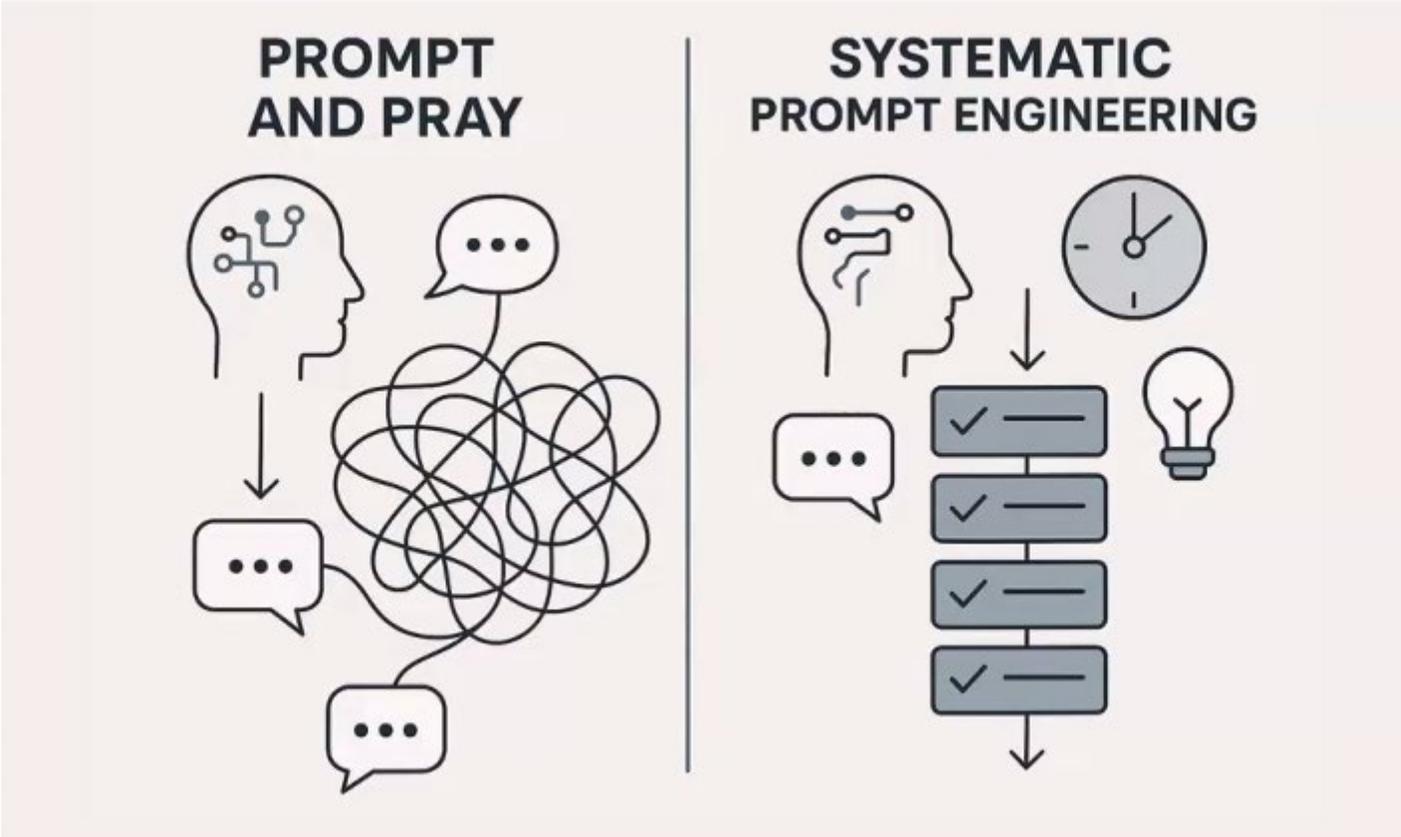
1. Research & adhere to legal limitations – copyright, FERPA, etc. (*Future FDCI AI seminars will address fair use & intellectual property etc.*)
2. Enact explicit policies regarding acceptable student use of AI for a given program or class (*CFDI has sample policies available for use online now in their AI Resources*)
3. Avoid the trap of “policing” AI use – aim instead to develop course content that *embraces* AI use (*main focus of 2/26 & 3/4 FDCI seminars*)
4. **Use AI well, ethically, and transparently, to increase your own effectiveness as an educator and provide a model of “good” AI use for your students.**

*... but how?*

# Technically Proficient AI interaction



# Prompt Engineering



# Prompting Basics



Be specific



Assign AI a role



Describe preferred output



Use “do” and  
“don’t”



Use examples



Identify key sources

# Prompting Basics, Cont'd



Consider tone and audience



Build on previous prompts



Correct mistakes and give feedback



Ask AI what else it needs

Or...  
“ChatGPT, help  
me develop an  
excellent  
prompt to teach  
X topic”



Output  
Evaluation:  
(A Few)  
Generative  
AI Pitfalls  
& Concerns

---

Hallucinations

---

Sycophancy

---

Vague & verbose output

---

Environmental concerns

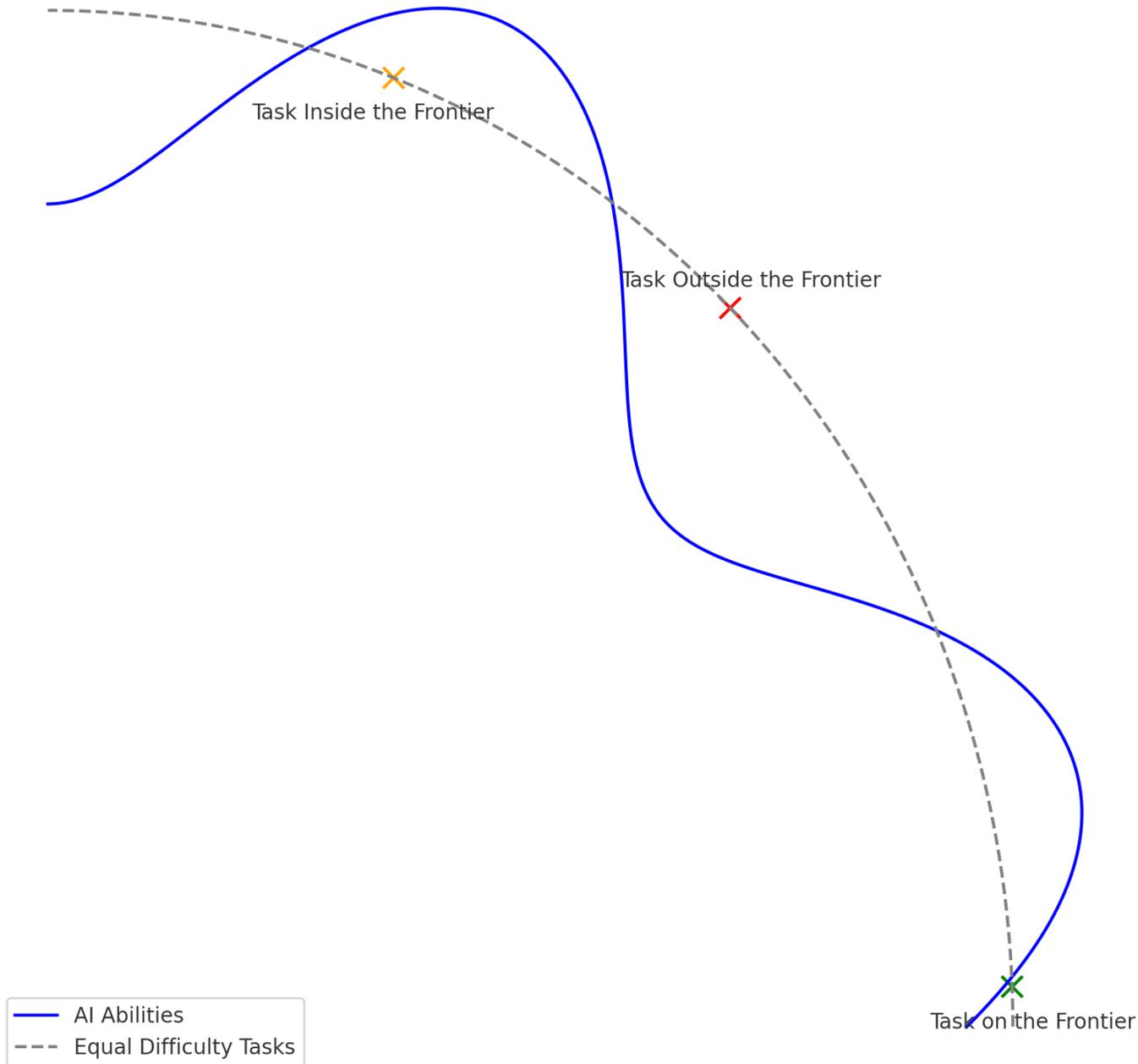
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Intellectual property concerns

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Potential biases

# Identifying the Jagged Frontier of AI in Education



*Dell'Acqua et al. 2023*

# AI Capabilities are Not Intuitive

## GPT-4 Passes the Bar Exam: What That Means for Artificial Intelligence Tools in the Legal Profession

April 19, 2023 | By Pablo Arredondo, Q&A with Sharon Driscoll and Monica Schreiber

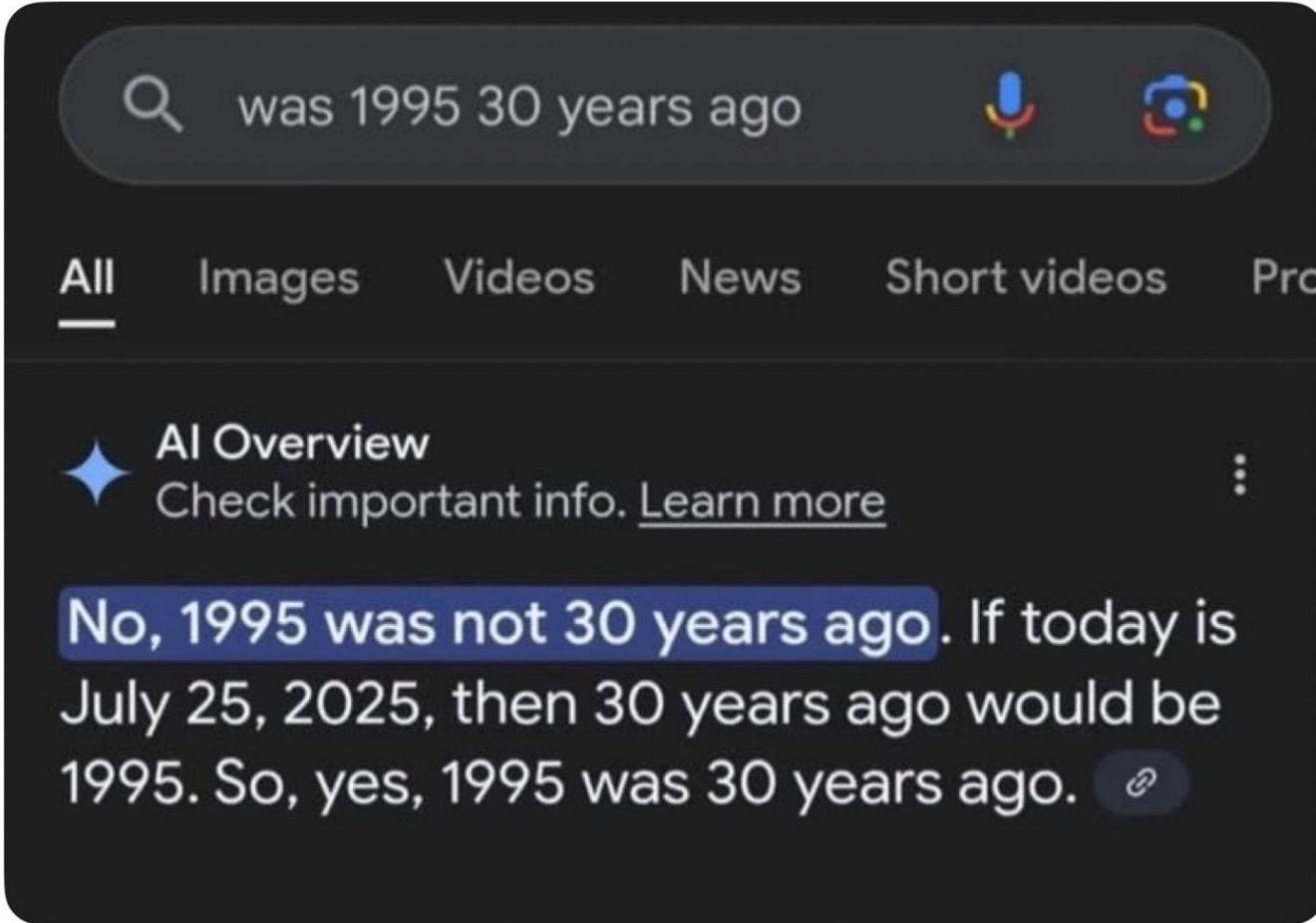
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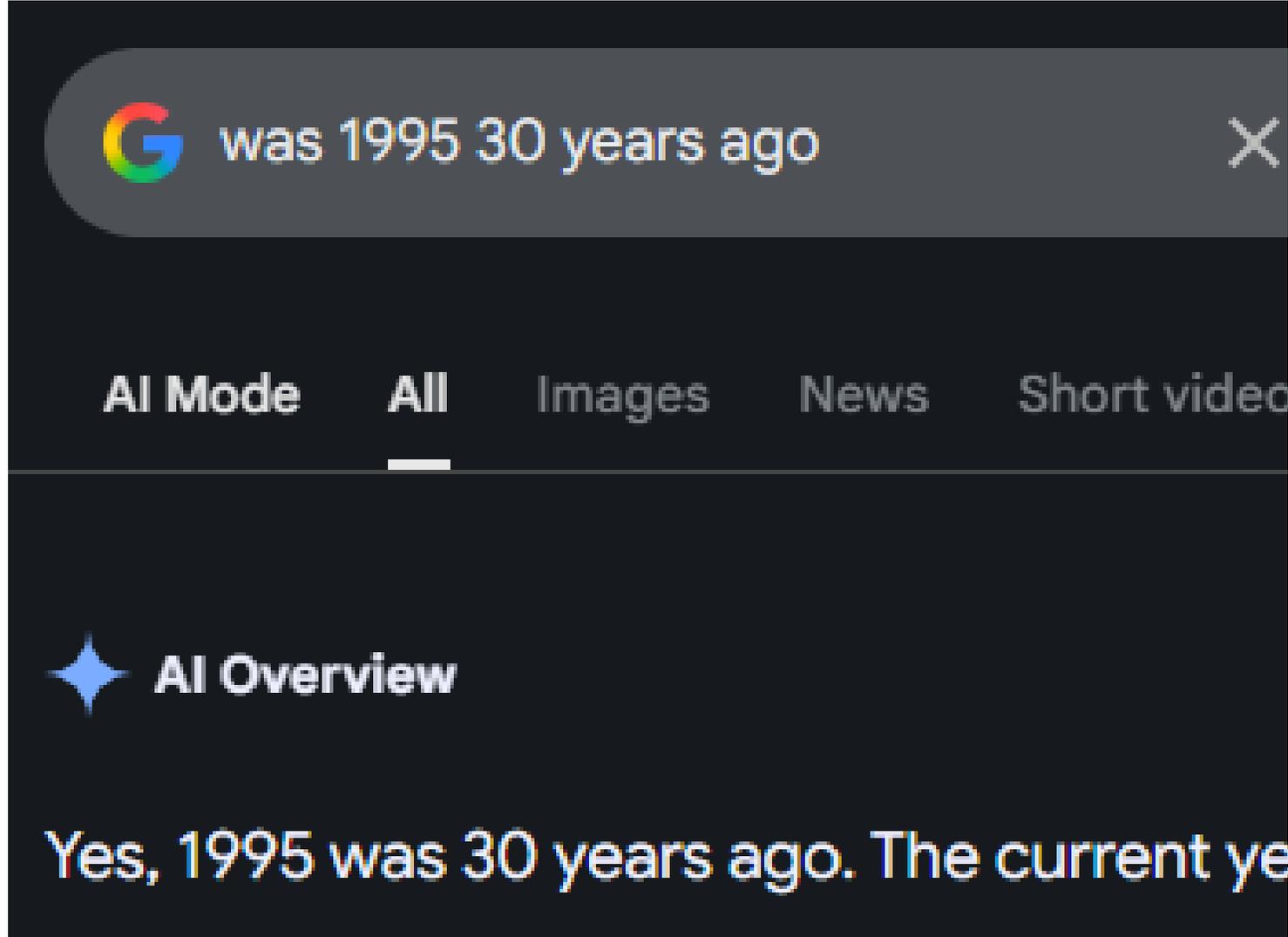


AI Capabilities  
are Not  
Intuitive...

(July 25, 2025)

9:35 AM · 8/23/25 · **8.4M** Views

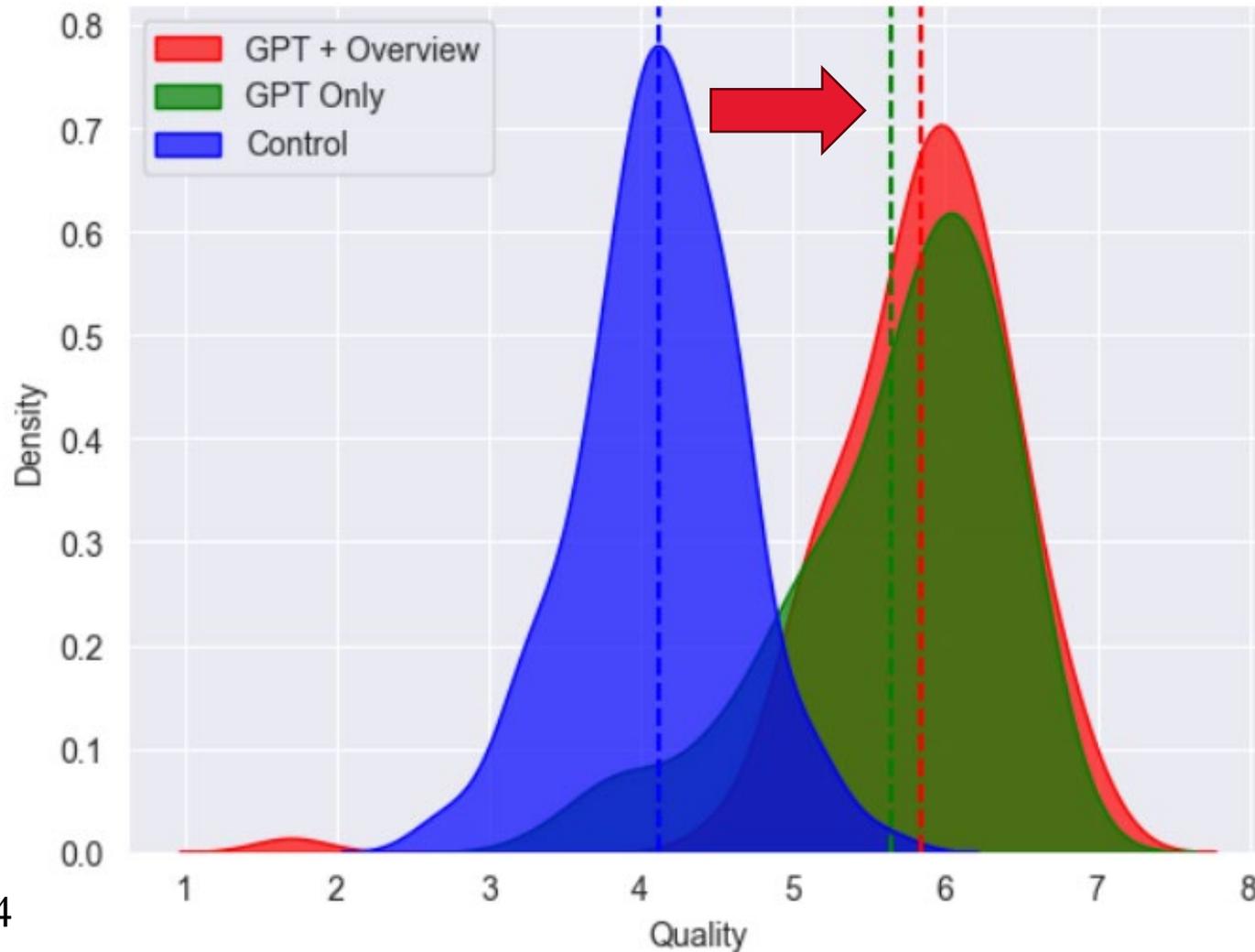
 1.4K    13K    267K    11K   



Nor are  
they  
permanent

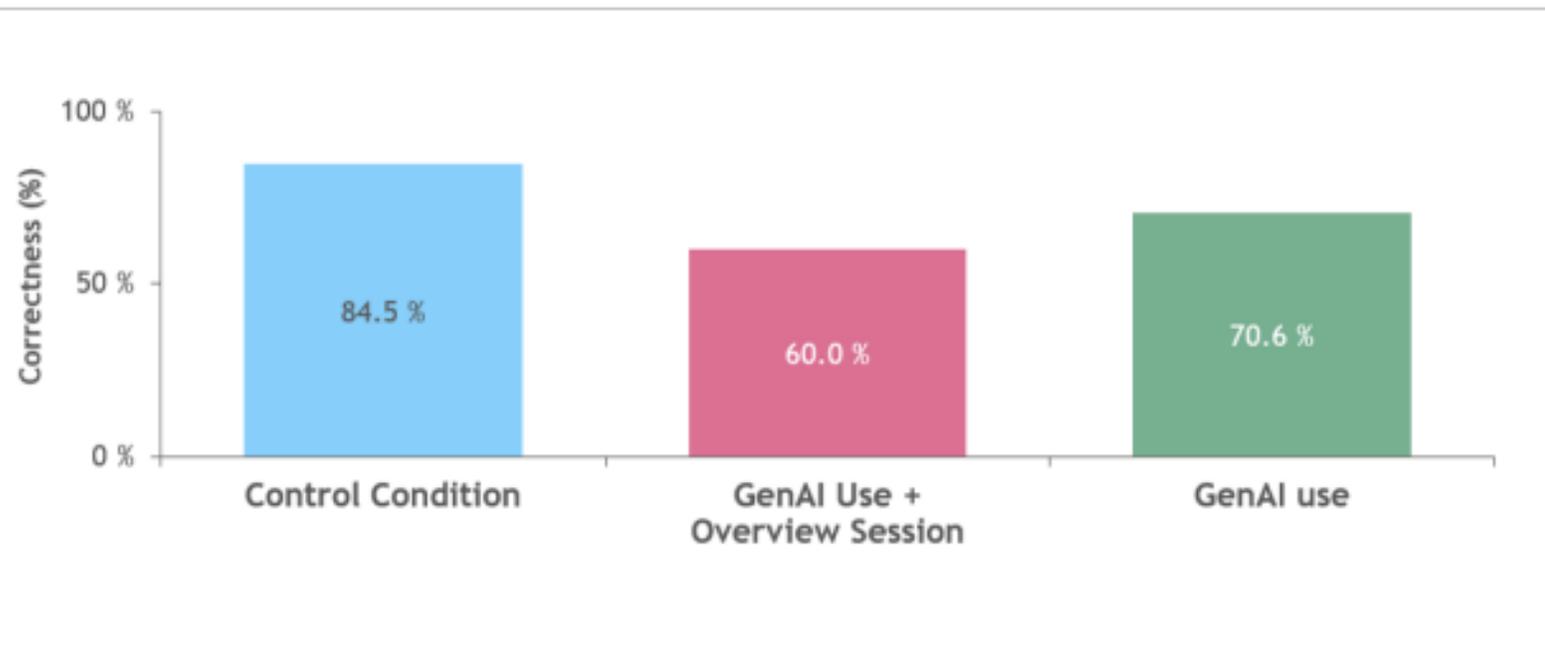
(September 3, 2025)

# Using AI for “good fit” tasks increases quality of the output



# Using AI for “bad fit” tasks decreases quality of the output

Figure 7: Performance - Outside the Frontier



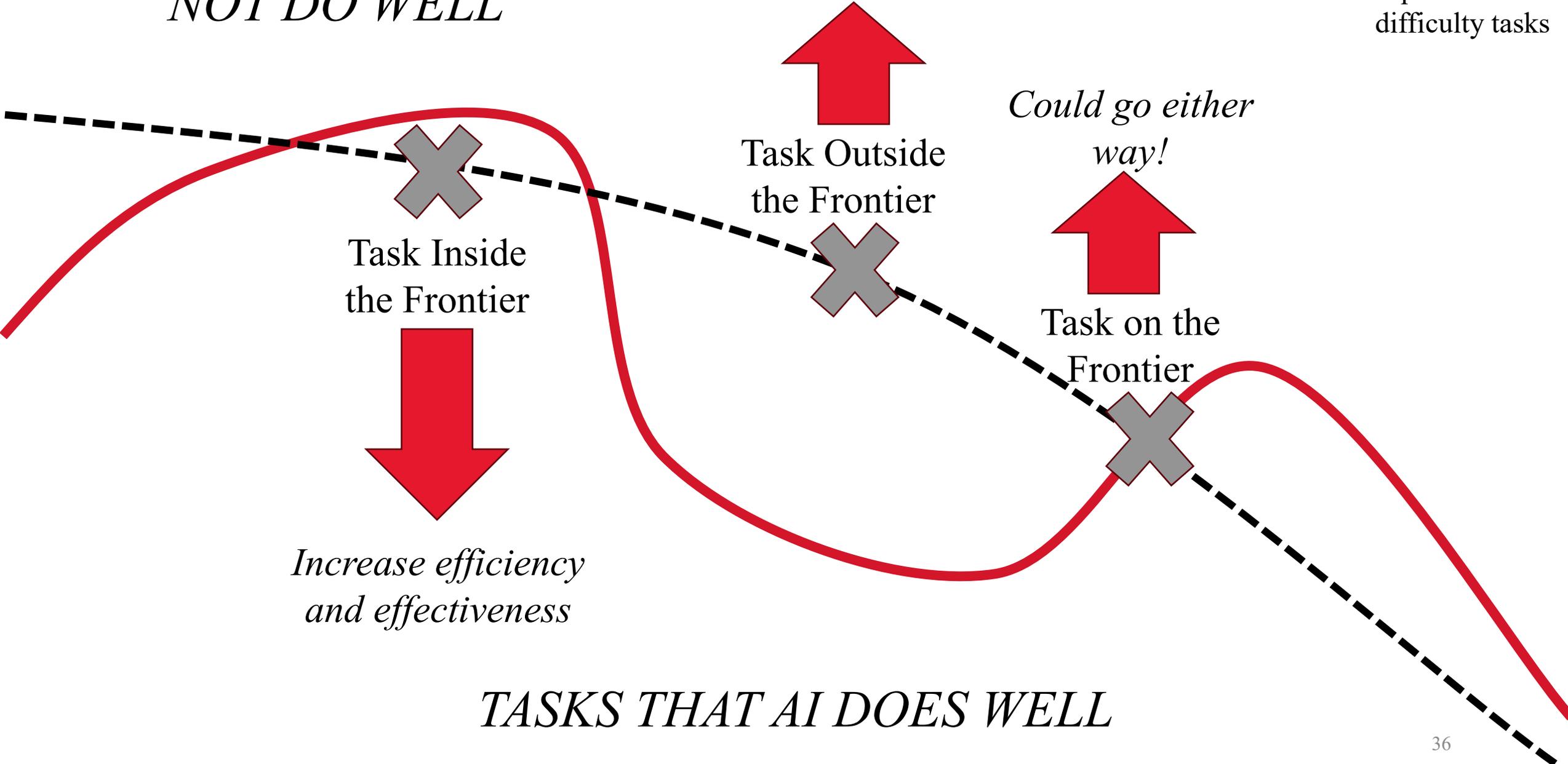
# In summary...

The experiments show that the shape and position of the frontier are vital to understanding the impact of AI on work. On tasks within the frontier, AI significantly improved human performance. Outside of it, humans relied too much on the AI and were more likely to make mistakes. Not all users navigated the jagged frontier with equal adeptness. While some completed their task incorrectly, others showcased a remarkable ability to harness the power of AI effectively. We conducted further analyses of the

*TASKS THAT AI DOES  
NOT DO WELL*

*Slow us down and  
waste our time*

— AI Abilities  
- - - Equal difficulty tasks



Task Inside the Frontier

Task Outside the Frontier

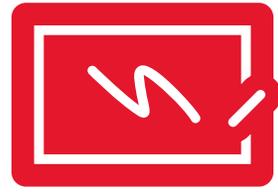
Could go either way!

Task on the Frontier

*Increase efficiency  
and effectiveness*

*TASKS THAT AI DOES WELL*

# Let's find the jagged frontier of AI in education



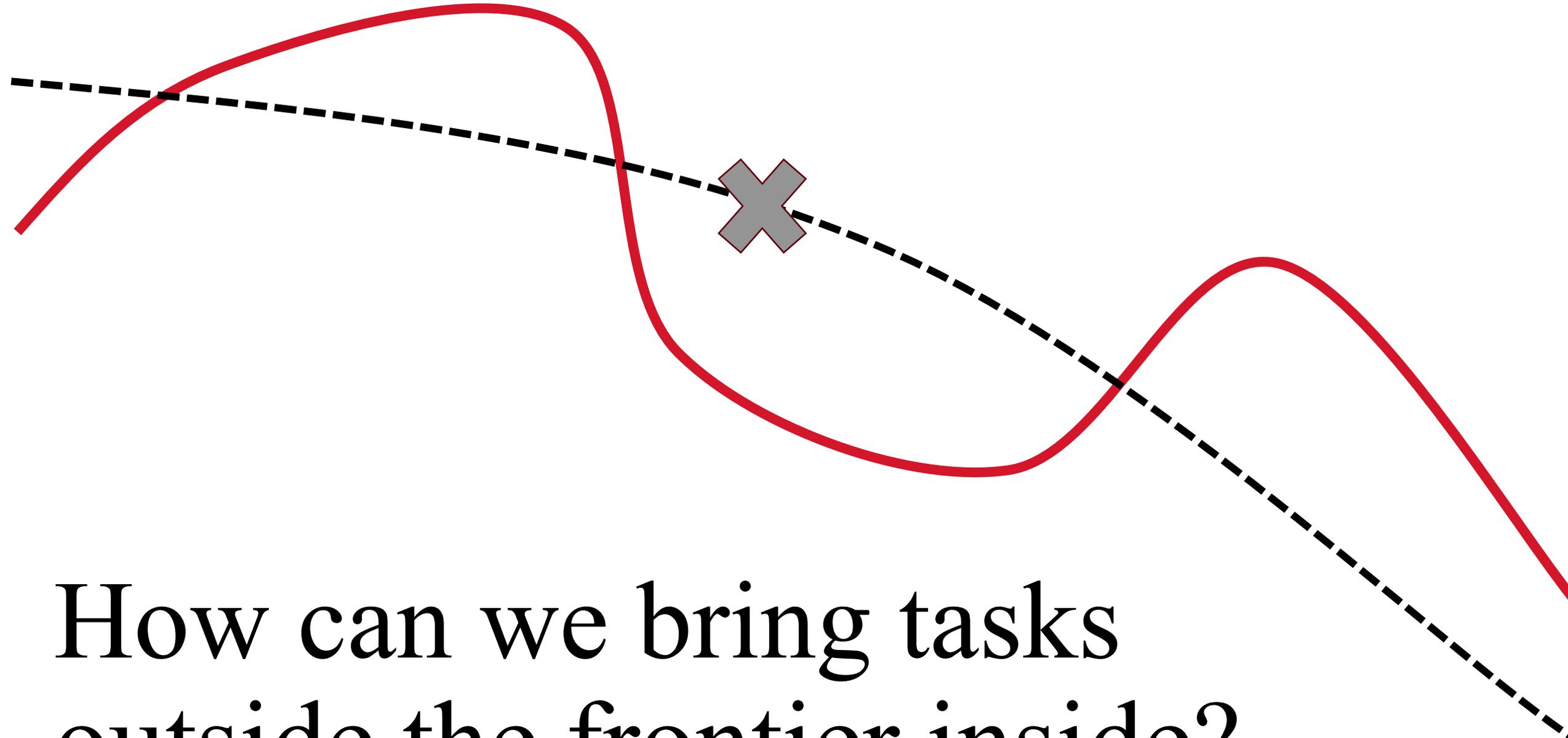
Course Design and  
Content

Classroom Prep and  
Facilitation

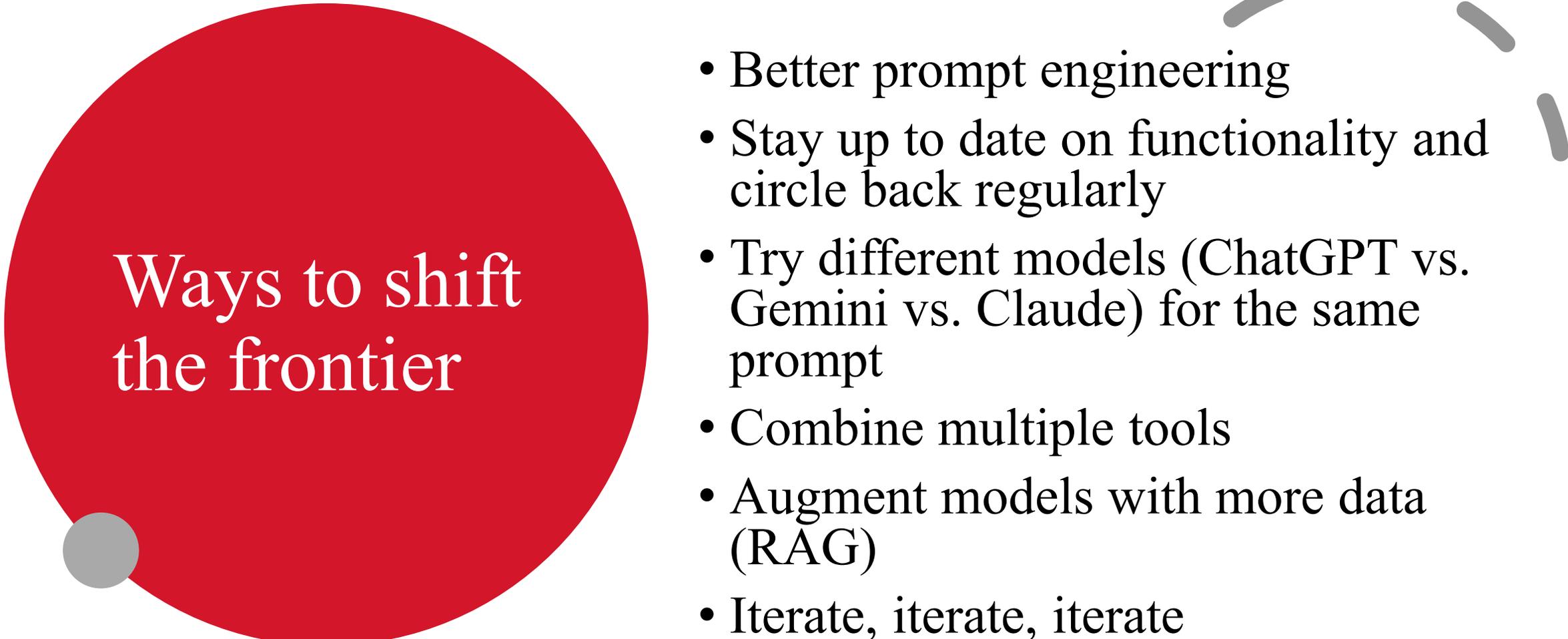
Student Assessment  
and Support

# Instructions

1. Think of a task that you do for each of the three areas: **course design and content, class prep and facilitation, student assessment & support.**
2. Prompt an AI tool of your choice (ChatGPT, Copilot, Gemini, Claude, Anthropic, etc.) to help you with that task
3. Evaluate output and iterate
4. Keep notes and share with others



How can we bring tasks  
outside the frontier inside?



## Ways to shift the frontier

- Better prompt engineering
- Stay up to date on functionality and circle back regularly
- Try different models (ChatGPT vs. Gemini vs. Claude) for the same prompt
- Combine multiple tools
- Augment models with more data (RAG)
- Iterate, iterate, iterate
- Human Q&A and testing

*just keep trying*

*You are the worst at AI you will ever be right now. The models and tools you are using are the worst they will ever be right now. Anticipate a learning curve that will delay the positive impacts of AI use on your work.*



# What's next?

- Provide your feedback on this session
- Future seminars:
  - [Architecting Assignments for an AI World](#), 2/26 & 3/04
  - [Leveling up the research process with generative AI](#), 3/19 & 3/26
  - [Who Owns the Output? Fair Use and IP in Generative AI](#), 3/26 & 4/01

# Resources

- Resources specific to each slide are available in the Notes of the slide
- *Co-Intelligence: Living and Working with AI*, Ethan Mollick
- [One Useful Thing](#) – Ethan Mollick’s blog
- [Everyday AI](#) – podcast and newsletter
- *Ethical Machines*, Reid Blackman
- Harvard Business Impact: [Teaching with AI](#) (requires creation of a free HBI Educator account)
- [CFDI AI Resources Page](#)
- [Harvard University Generative AI Resources](#)

# Backup Slides

It's not magic or human thinking – it's math!

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... and why sometimes it “hallucinates” – if there is not enough data to suggest an appropriately high probability, it will *generate* a fictional response that *looks* like an appropriate response.

The New York Times

***A.I. Is Getting More Powerful,  
but Its Hallucinations  
Are Getting Worse***

A new wave of “reasoning” systems from companies like OpenAI is producing incorrect information more often. Even the companies don't know why.

# Scientists Increasingly Can't Explain How AI Works

By Chloe Xiang November 1, 2022, 9:00am

The specific engineering of ChatGPT has made it quite compelling. But ultimately (at least until it can use outside tools) ChatGPT is “merely” pulling out some “coherent thread of text” from the “statistics of conventional wisdom” that it’s accumulated. But it’s amazing how human-like the results are. And as I’ve discussed, this suggests something that’s at least scientifically very important: that human language (and the patterns of thinking behind it) are somehow simpler and more “law like” in their structure than we thought. ChatGPT has implicitly discovered it. But we can potentially explicitly expose it, with semantic grammar, computational language, etc.

What ChatGPT does in generating text is very impressive—and the results are usually very much like what we humans would produce. So does this mean ChatGPT is working like a brain? Its underlying artificial-neural-network architecture is based on an idealization of what we know about the brain, and it’s clear that when we have a good understanding of what’s going on are quite similar.

Stephen Wolfram, 2023

# Better prompting

Ask	Ask it to do (better, newer, more specific)
Explain	Ask it to explain itself (it won't – but could still be illuminating)
Invite	Invite it to take a specific perspective or persona
Organize	Ask it to organize its response
Decompose	Break your question into parts
Reframe	Reframe or rephrase the question
Constraints	Give constraints
Examples	Give examples of good answers, guidelines for success, rubrics
Provide	Provide data and ask it to hypothesize
Apply	Ask it to apply methods of frameworks you use

# AI is replacing junior (but not senior) employees

## Generative AI as Seniority-Biased Technological Change: Evidence from U.S. Résumé and Job Posting Data\*

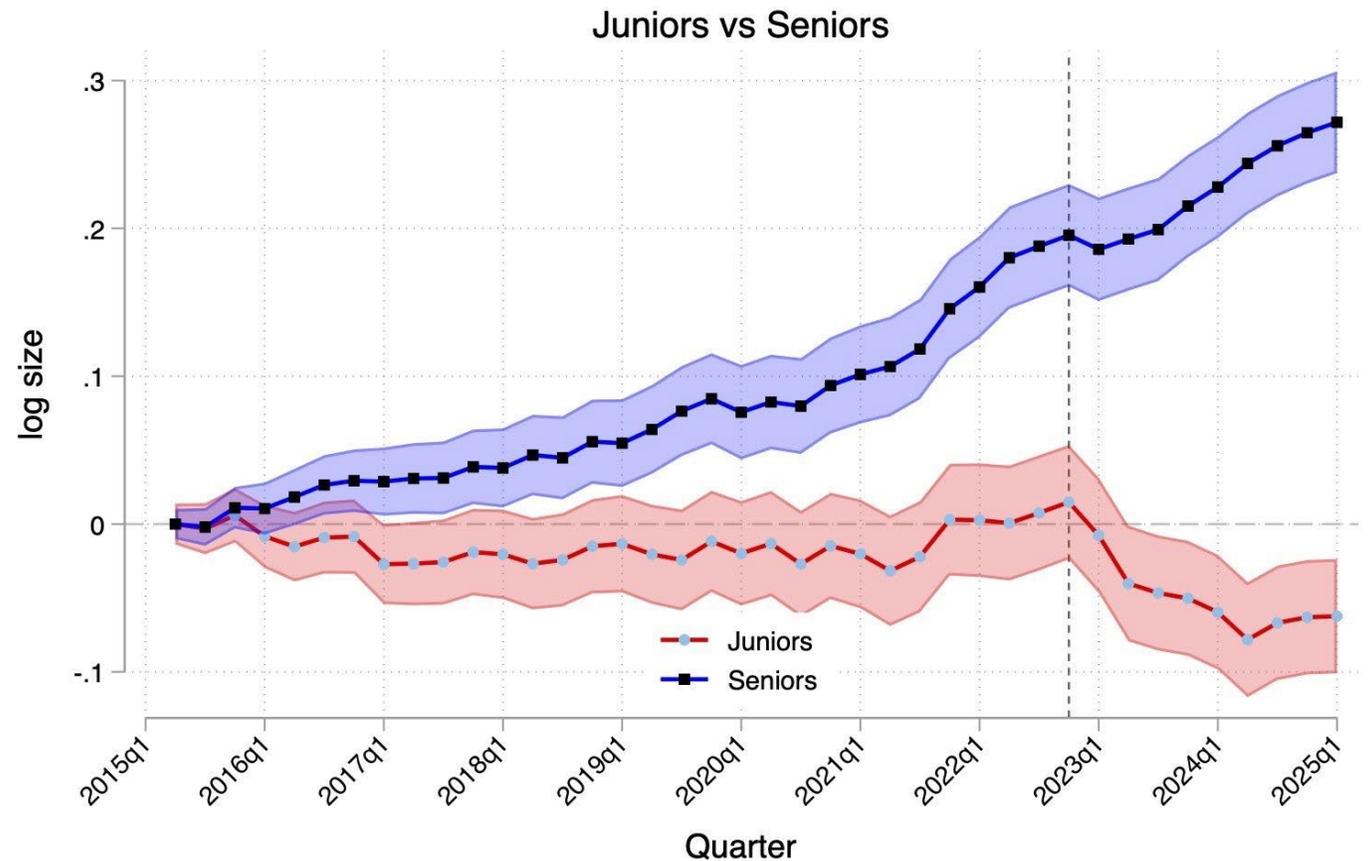
Seyed M. Hosseini<sup>†</sup>

Guy Lichtinger<sup>†</sup>

Preliminary  
August 2025

### Abstract

We study whether generative artificial intelligence (AI) constitutes *seniority-biased technological change*, disproportionately affecting junior reworkers. Using U.S. résumé and job posting data covering nearly 60 years in 285,000 firms (2015–2025), we track within-firm employment dynamics by seniority. We identify AI adoption through a text-analysis approach looking for dedicated “AI integrator” roles, signaling active implementation of generative AI. Difference-in-differences and triple-difference estimates show that in 2023Q1, junior employment in adopting firms declined sharply in firms that adopted AI, while senior employment continued to rise. The junior decline is driven primarily by slower hiring rather than increased separations, with the



# A final note on prompts...



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Technology And Analytics

## AI Prompt Engineering Isn't the Future

by Oguz A. Acar

June 6, 2023

# Environmental Impact of AI



## What Uses More?



Free to share or adapt  
(see **About this app** for credits)

### Compare the environmental footprint of digital tasks

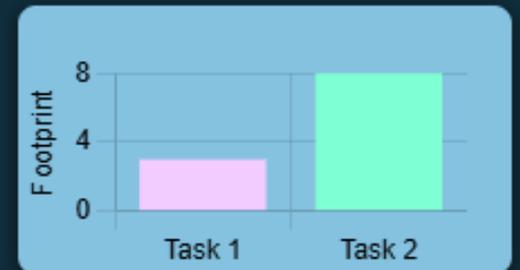
#### ▼ About this app

It's hard to know what to believe about the ecological impact of generative AI when the Internet seems evenly divided between "Stop worrying" and "It's the apocalypse." This nonpartisan calculator is an attempt to provide a more objective comparison of the energy and water usage of various AI tasks versus other common digital activities, like watching Netflix or searching Google. The calculations are based on academic papers, industry reports, and expert estimates.

Paragraph of AI text x 1: 💡 3 lightbulb-minutes (Wh) [🌊 0.02 liters]

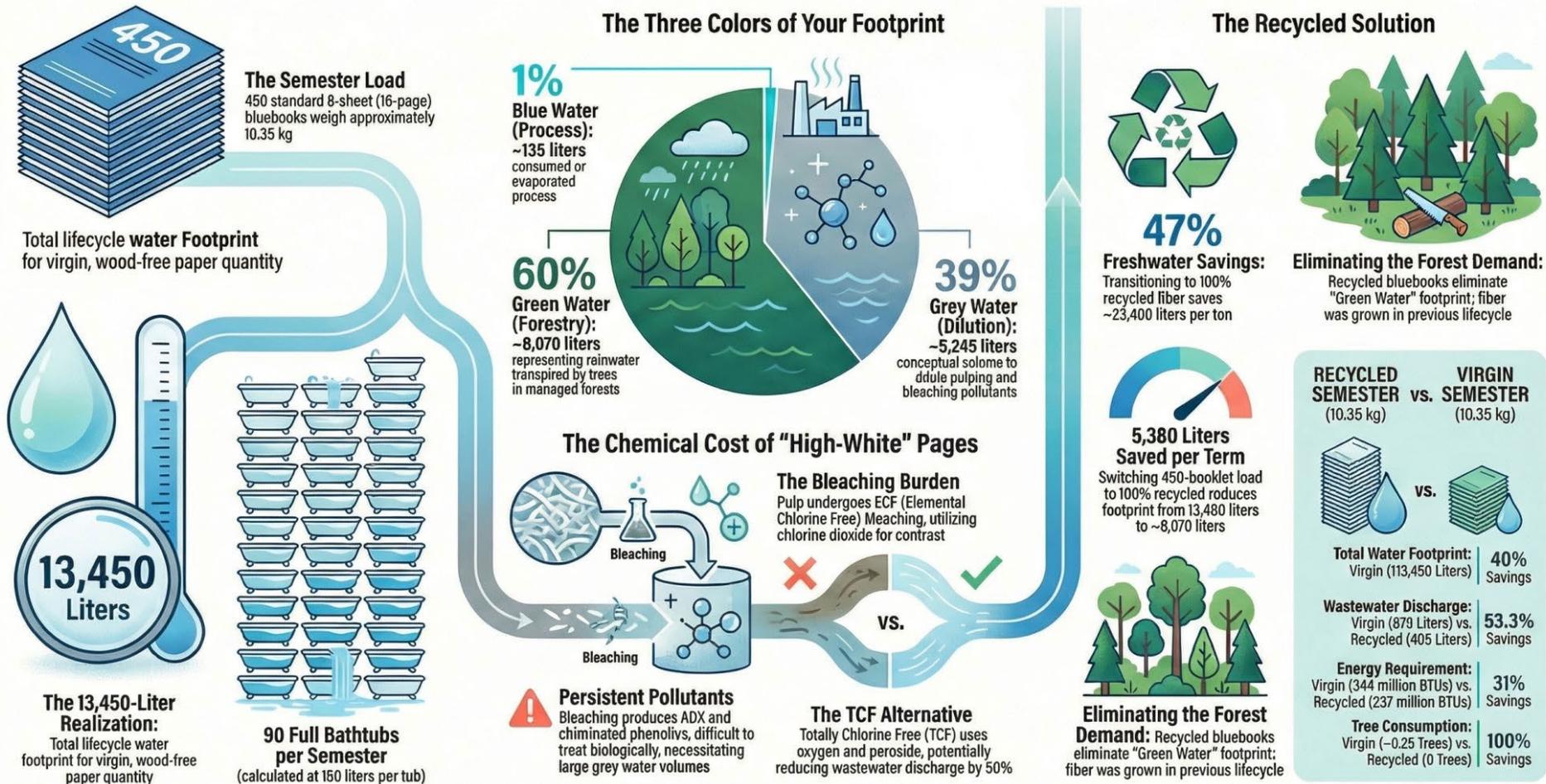
versus

Storing 5 GB in the cloud for 1 month x 1: 💡 8 lightbulb-minutes (Wh) [🌊 0.06 liters]



# Environmental Impact of Blue Books

## The Hidden Hydraulic Cost of a Semester: A 450-Bluebook Water Audit



NotebookLM

Source: Michelle Kassorla, Ph.D. LinkedIn (link in slide notes)