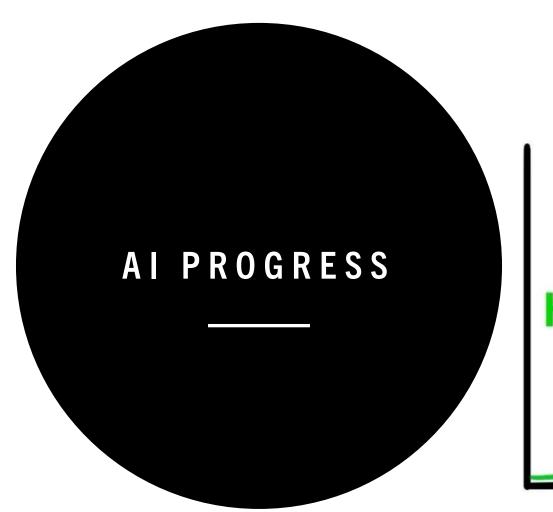


CONTENTS

- I. Overview of AI Progress
- 2. Advanced Uses of Al and Implications for Research
- 3. Understanding Your Generative Al Options
- 4. Ethical Al Usage
- 5. Systems I vs Systems 2 Prompting
- 6. Available Generative Al Tools
- 7. Qualitative Analysis Exercises
- Sources of Good Information





TWO AI LEADERS AWARDED NOBEL PRIZES IN 2024

GEOFFREY HINTON - PHYSICS

DEMIS HASSABIS - CHEMISTRY



OPEN AI DEFINES THE PATH TO AGI

- Open Al Released the OI Preview Reasoning Model in September 2024
- 2. AGI = "highly autonomous systems that outperform humans at most economically valuable work"

5. Organisational

4. Innovator

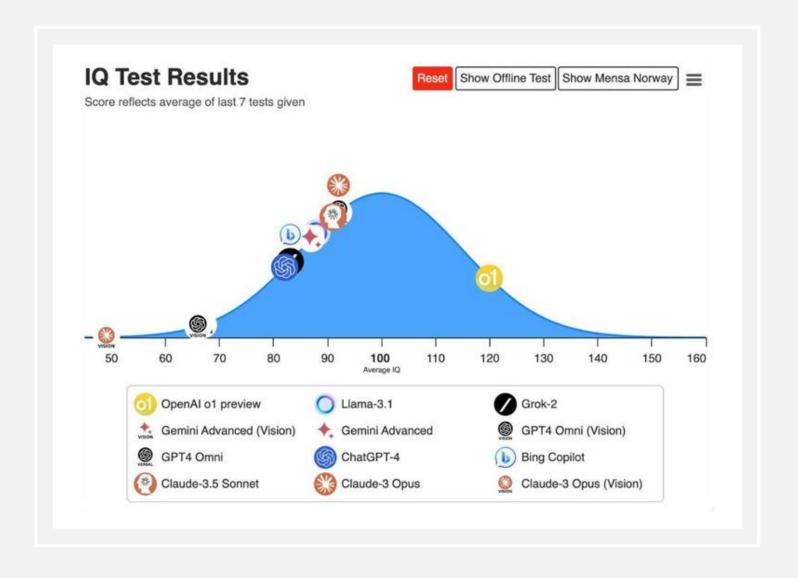
3. Agent

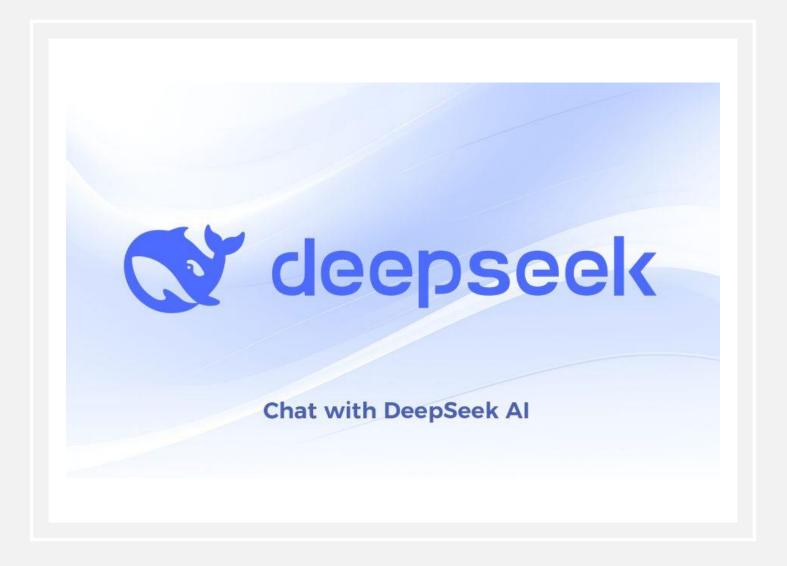
2. Reasoning

I. Conversational

OI ACHIEVES IQ OF 120

- Using the Norway Mensa IQ Test
- Median Human
 Score of 100





CHINA RELEASES RI

- Chinese Lab Deepseek releases Deepseek-RI Reasoning Model last week
- R1 Surpasses O1 on many benchmarks
- Model released in Open Source Format
- Deepseek Registered on 17 July 2023
- https://www.deepseek.com/

FROM AGI TO ASI

- Level above AGI is Artificial Superintelligence (ASI)
- When AI Capabilities exceed the combined abilities of all Humans on the plannet

6.All Humans

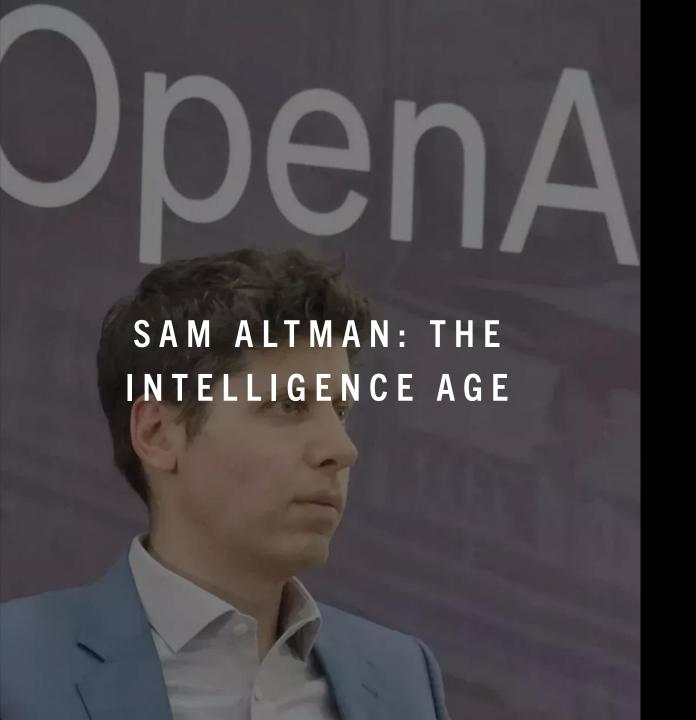
5. Organisational

4. Innovator

3. Agent

2. Reasoning

I. Conversational



- Super Intelligence will arrive in "a few thousand days"
- Drastic Advances in Healthcare, Education, Climate Change and Innovation
- Open AI has unlocked reasoning leading to advances in Physics and Maths
- Depends on Computation Power and Energy



- Super Intelligence is Approaching
- "Marginal Returns to Intelligence"
 We need to consider limiting factors such as infrastructure, data costs, social barriers.
- Expands benefits in Health,
 Economic Development,
 Governance
- Incremental but rapid changes to work. E.g. No more call centres in 2025



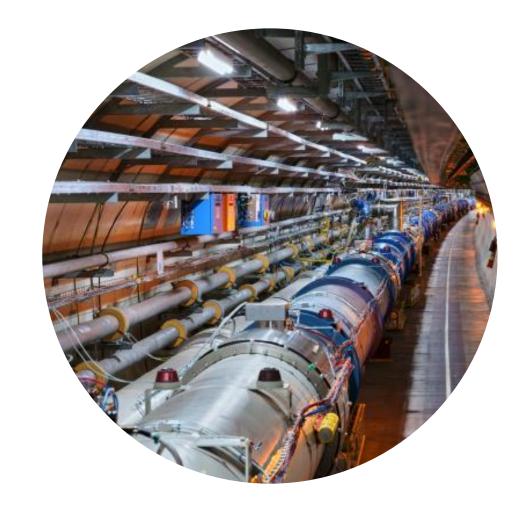
- Al to fundamentally reshape society
- Al reduces the cost of intelligence
- Al control is monopolized
- Democratising Al is critical
- Open and Collaborative Al Development is needed
- Need to rethink our Economic Model
- Need Al Awareness

ADVANCED USES OF AI



CERN'S LARGE HADRON COLLIDER

- Use AI Algorithms to examine Petabytes of data
- Each particle collision generates 1 MB of data
- The model generates 40 million collisions a second



LADDERBOT - AI INTERVIEW AGENT

- Developed in the Karlsruhe Institute of Technology (KIT) in Germany
- Al Agent that is taught to ask relevant probing questions, mimic expert human interviewers
- Reach a larger audience
- Found to increase individual survey responses by 30%
- More intuitive and deeper insights
- Interviews are more reproducible



Go to International Journal of Human-Computer Studies on ScienceDirect an-Computer

Studies

Volume 171, March 2023, 102969



Ladderbot—A conversational agent for human-like online laddering interviews

Tim Rietz ^a ス ⋈, Alexander Maedche ^b

Show more V

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https://doi.org/10.1016/j.ijhcs.2022.102969 7

Get rights and content .

Highlights

- Chatbots may extend the laddering technique by providing an alternative to surveys.
- · Ladderbot encourages participants to provide twice as many and significantly longer answers.
- Interacting with Ladderbot provides participants with higher learnability.
- Survey-based laddering may produce more structured results than chatbots.

GENERATIVE AGENT SIMULATIONS OF 1000 PEOPLE

- Al Based Interview tool collects data from 1000 people using the laddering technique through a single 2 hour interview
- Applies deep introspection and follow up questions
- Found simulated answer of AI model is similar to the human answer (85% Accuracy)
- Can now model human behaviour
- Allows mass testing

Generative Agent Simulations of 1,000 People

Authors: Joon Sung Park^{1*}, Carolyn Q. Zou^{1,2}, Aaron Shaw², Benjamin Mako Hill³, Carrie Cai⁴, Meredith Ringel Morris⁵, Robb Willer⁶, Percy Liang¹, Michael S. Bernstein¹

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Abstract:

The promise of human behavioral simulation—general-purpose computational agents that replicate human behavior across domains—could enable broad applications in policymaking and social science. We present a novel agent architecture that simulates the attitudes and behaviors of 1,052 real individuals—applying large language models to qualitative interviews about their lives, then measuring how well these agents replicate the attitudes and behaviors of the individuals that they represent. The generative agents replicate participants' responses on the General Social Survey 85% as accurately as participants replicate their own answers two weeks later, and perform comparably in predicting personality traits and outcomes in experimental replications. Our architecture reduces accuracy biases across racial and ideological groups compared to agents given demographic descriptions. This work provides a foundation for new tools that can help investigate individual and collective behavior.

CHATGPT OUTPERFORMS DOCTORS IN DIAGNOSIS

- Study in JAMA Network
- Sample of 50 Doctors
- Al Outperformed Doctors in Diagnosing Diseases and Medical Conditions

Results

- Doctors without AI 74% Accuracy
- Doctors with AI 76% Accuracy
- ChatGPT Alone 90% Accuracy
- Human Bias?





Original Investigation | Health Informatics

Large Language Model Influence on Diagnostic Reasoning A Randomized Clinical Trial

Ethan Goh, MBBS, MS; Robert Gallo, MD; Jason Hom, MD; Eric Strong, MD; Yingjie Weng, MHS; Hannah Kerman, MD; Joséphine A. Cool, MD; Zahir Kanjee, MD, MPH; Andrew S. Parsons, MD, MPH; Neera Ahuja, MD; Eric Horvitz, MD, PhD; Daniel Yang, MD; Arnold Milstein, MD; Andrew P. J. Olson, MD; Adam Rodman, MD, MPH; Jonathan H. Chen, MD, PhD

Abstract

IMPORTANCE Large language models (LLMs) have shown promise in their performance on both multiple-choice and open-ended medical reasoning examinations, but it remains unknown whether the use of such tools improves physician diagnostic reasoning.

OBJECTIVE To assess the effect of an LLM on physicians' diagnostic reasoning compared with conventional resources.

DESIGN, SETTING, AND PARTICIPANTS A single-blind randomized clinical trial was conducted from November 29 to December 29, 2023. Using remote video conferencing and in-person participation across multiple academic medical institutions, physicians with training in family medicine, internal medicine, or emergency medicine were recruited.

INTERVENTION Participants were randomized to either access the LLM in addition to conventional diagnostic resources or conventional resources only, stratified by career stage. Participants were allocated 60 minutes to review up to 6 clinical vignettes.

MAIN OUTCOMES AND MEASURES The primary outcome was performance on a standardized rubric of diagnostic performance based on differential diagnosis accuracy, appropriateness of supporting and opposing factors, and next diagnostic evaluation steps, validated and graded via blinded expert consensus. Secondary outcomes included time spent per case (in seconds) and final diagnosis accuracy. All analyses followed the intention-to-treat principle. A secondary exploratory analysis evaluated the standalone performance of the LLM by comparing the primary outcomes between the LLM alone group and the conventional resource group.

RESULTS Fifty physicians (26 attendings, 24 residents; median years in practice, 3 [IQR, 2-8]) participated virtually as well as at 1 in-person site. The median diagnostic reasoning score per case was 76% (IQR, 66%-87%) for the LLM group and 74% (IQR, 63%-84%) for the conventional resources-only group, with an adjusted difference of 2 percentage points (95% CI, -4 to 8 percentage points; P = .60). The median time spent per case for the LLM group was 519 (IQR, 371-668) seconds, compared with 565 (IQR, 456-788) seconds for the conventional resources group, with a time difference of -82 (95% CI, -195 to 31; P = .20) seconds. The LLM alone scored 16 percentage points (95% CI, 2-30 percentage points; P = .03) higher than the conventional

Key Points

Question Does the use of a large language model (LLM) improve diagnostic reasoning performance among physicians in family medicine, internal medicine, or emergency medicine compared with conventional resources?

Findings In a randomized clinical trial including 50 physicians, the use of an LLM did not significantly enhance diagnostic reasoning performance compared with the availability of only conventional resources.

Meaning In this study, the use of an LLM did not necessarily enhance diagnostic reasoning of physicians beyond conventional resources; further development is needed to effectively integrate LLMs into clinical practice.

- Visual Abstract
- Invited Commentary
- Supplemental content

Author affiliations and article information are listed at the end of this article.

IMPLICATIONS FOR ALIN RESEARCH

- MUST BUILD AWARENESS AMONG STAFF AND PUBLIC!
- Use AI for Public Engagement
- Data collection is critical

Expand our data collection processes

Larger Samples

More Frequent Data Collection

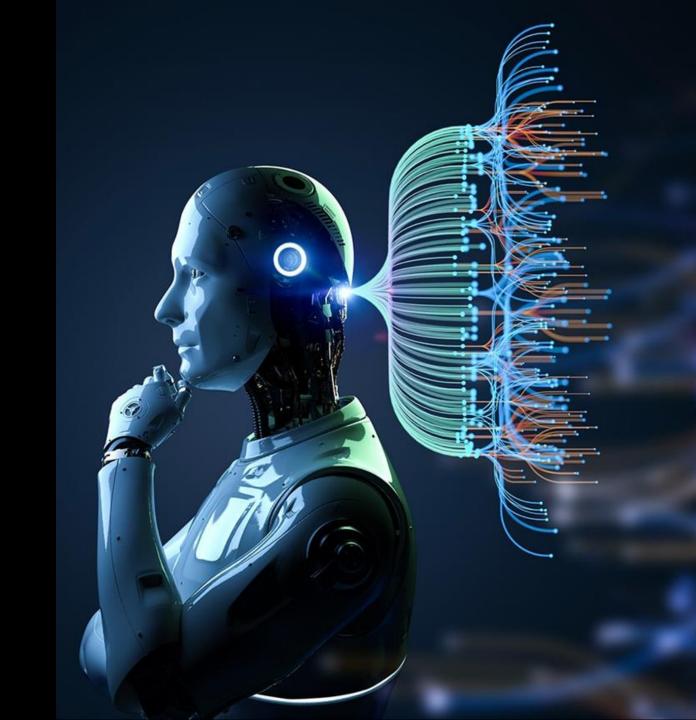
Iterations

 Al can be used to accelerate data analysis

Use AI to analyse data

- Natural Language Processing to automate the analysis of Qualitative data
 Allows Qualitative data to be paired with Quantitative data
- Dynamic AI Enhanced Public Surveys
 Automatically analyse data giving
 respondents immediate feedback.

UNDERSTANDING YOUR AI OPTIONS?



AVAILABLE CHATBOTS

	ChatGPT 40	Anthropic Claude	Google Gemini	Google's Notebook LM	Meta.Al	Poe.com	Le Chat from Mistral	Deepseek
Cost	Free: Limited Plus: \$20	Free – limited Pro = \$20	Free Pro = \$20	Free	Free	Based on Points System	Free (Open Source)	Free
Features	Document Analysis + GPTs + Advanced Data Analytics +OI Preview + Canvas + Search + Advanced Voice	Document analysis + Quantitative Analysis +Artefacts	Linked to the Internet +Doc Analysis	Audio Transcription +Podcasting +Deep analysis of 50 Sources	Immediate image creation	Links to different models Create Custom Bots like GPT	Image Generation +Search +Canvas	Free Reasoning model – 50 Questions a day
Languages	85	10	265	265	8	Depends	13	30+
Token Window	128k	Free = 100k Paid = 200k	Free = 32k Paid = 2m	25 million	128k	Depends	32k	128k









Living
guidelines
on the
RESPONSIBLE
USE OF
GENERATIVE AI
IN RESEARCH

ERA Forum Stakeholders' document

European Commission

UCT Senate Ethics in Research Committee

Senate Ethics in Research Committee (EiRC) Guidelines and recommendations for the use of generative artificial intelligence (AI) tools in research

Title	Guidelines and recommendations for the use of generative artificial intelligence (Al) tools in research		
Approval date	31 October 2023 (via EiRC Chair's Circular)		
Person(s) responsible for	Research Integrity Manager		
drafting, review and revision	Research Integrity Director		
Next review date	October 2024		
Status	Guideline		
Approval route	Office of Research Integrity (ORI) > Senate Ethics in Research Committee (EiRC)		
Approving body	EIRC		
Relevant related policies, procedures and guidelines	UCT Policy for Responsible Conduct of Research UCT Research Ethics Code for Research Involving Human Participants UCT Research Ethics Code for Use of Animals in Research and Teaching Authorship Practices Policy Conflict of Interest Policy UCT Policy and Procedures for Breach of Research Ethics Codes and Allegations of Misconduct in Research (colloquially known as the 'Research Misconduct' policy)		

Table of Contents

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0	utputs are only as good as inputs3
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7.	Concluding remarks
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NWU Guidelines on the Responsible and Ethical Use of Artificial Intelligence

The following information will appear as a declaration when assignments/tests are to be submitted to eFundi:

Honour Pledge: I declare that this assignment/work was done according to the NWU Guidelines on Responsible and ethical use of Artificial Intelligence (NWU Current Students)

Al, as Large Language Models (LLMs) or Generative AI, is primarily defined as "a type of artificial intelligence that can learn from and mimic large amounts of data to create content such as text, images, music, videos, code and more, based on inputs and prompts" (https://huit.harvard.edu/ai#block-boxes-1687273052). Al may also include other AI tools such as research summarising tools, automatic data analysis tools, and creation of synthetic data tools (https://sajs.co.za/article/view/17147). Al is not restricted to this definition and with new AI tools constantly being developed, the use of other or new forms of AI should not be excluded from this definition and it remains the responsibility of the student to get permission for the use of it from the relevant lecturer or supervisor.

Ethical use of Al

The NWU is committed to the ethical use of AI technologies and applies the following basic principles to which you as a student should adhere:

- Transparency. You should keep a record and be open and clear about when and how you use AI. Declare your use of AI and keep track of your use of it.
- 2. Honesty. You should be honest about the use of AI in all your academic work. You need to follow the rules and guidelines given by the lecturer/supervisor for the use of AI, otherwise you are academically dishonest. Any work submitted by you that is not your own, and not clearly declared as such, might lead to a report on AITSA (Academic Integrity Information Technology System Application).
- Integrity. Do not lose your trustworthiness through the dishonest use of Al. Integrity
 means wholeness or the consistent and coherent ability to live as an honest person.
 Honest and transparent use of Al is needed to keep your academic work's integrity
 integrity.
- 4. Accountability and Ownership. With the submission of an assignment, you as a student are seen as the owner of the work. You should therefore take ownership of what is written since you are accountable for it. You cannot blame Al if something false, biased, or inappropriate is stated in the work you present as your own. The

North West University

KEY ETHICS PRINCIPLES

- I. Reliability: Ensuring the quality and integrity of research, including the verification and reproducibility of Al-generated information.
- 2. Honesty: Maintaining transparency in all aspects of the research process, including the disclosure of generative Al use.
- 3. Respect: Consider the impact of Al use on research participants, colleagues, society, and the environment.
- 4. Accountability: Taking responsibility for all aspects of the research process and its outcomes, including those involving generative Al.

PRIVACY AND INTELLECTUAL CAPITAL

- Everything uploaded into ChatGPT (free), ChatGPT+ is introduced into ChatGPT's training data
 - Potentially could be accessed by someone else with a specific prompt
- ChatGPT Teams & Notebook LM
 -Claimed to protect your data –
 not introduced into Training Data
- Open Al API data not used in training since I Mar 2023
- Similar Issues with Claude, Bing, Bard, Perplexity, etc

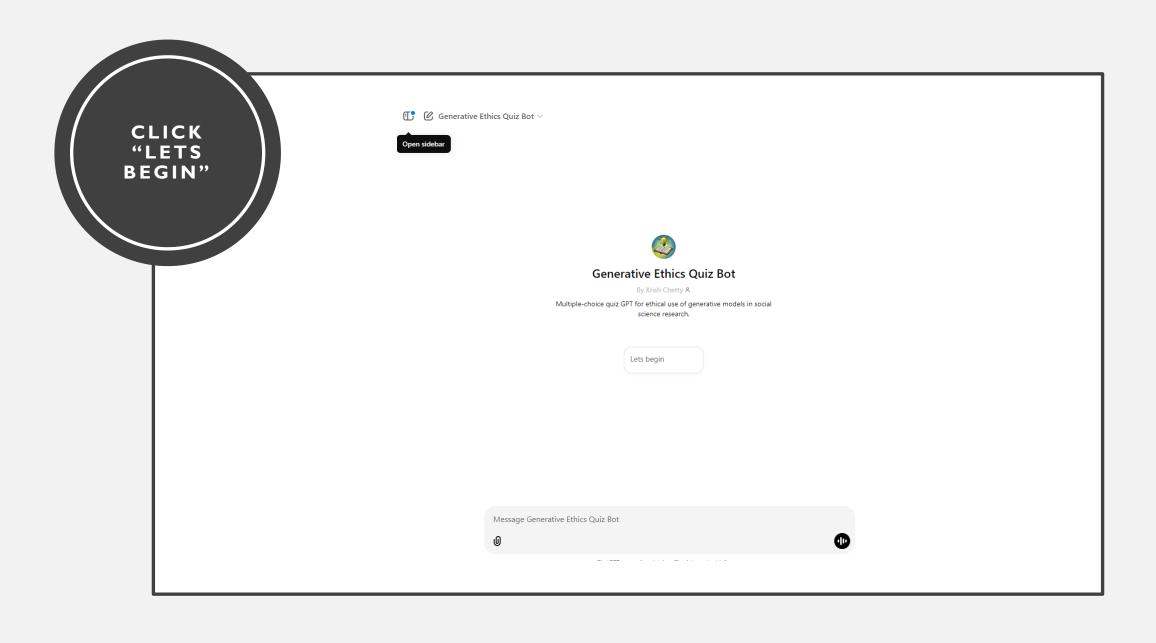
Do	Don't
 Safeguard personal and sensitive data, Be transparent on how you use Al Verify your outputs Respect the Intellectual Property of others Stay informed 	 Don't overely on AI Don't share secrets. Do not attribute AI- generated content as your own Don't use brand names or trademarks. Do not share valuable IP

RESEARCH ETHICS QUIZ

 https://chatgpt.com/g/guDPGP8pgK-generative-ethicsquiz-bot

Click link and Sign Into ChatGPT









Lets begin



Great! Let's dive into this interactive learning session about the ethical use of generative AI in social science research. I'll start with a question, provide options, and then explain the correct answer after you respond.

Question 1: Which of the following best captures the principle of "Honesty" as outlined in the guidelines for ethical generative AI use in social science research?

- A) Ensuring all AI outputs are free from bias and inaccuracies.
- B) Disclosing the use of generative AI tools in the research process.
- C) Avoiding the use of generative AI tools in sensitive research activities.
- D) Holding researchers accountable for Al-generated outputs.

What do you think is the correct answer?







PROMPTING



WRITING A GOOD PROMPT

Prompt Structure:

- Instruction
- Layout
- Useful information

- I. Critically engage
 - 2. Refine
 - 3. Repeat

The more specific, the better the output

DANIEL KAHNEMAN'S MACHINERY OF THOUGHT



System I Thinking – Immediate

- Operates
 automatically and
 quickly,
- Little or no effort
- No sense of voluntary control.

System 2 Thinking – Deliberate

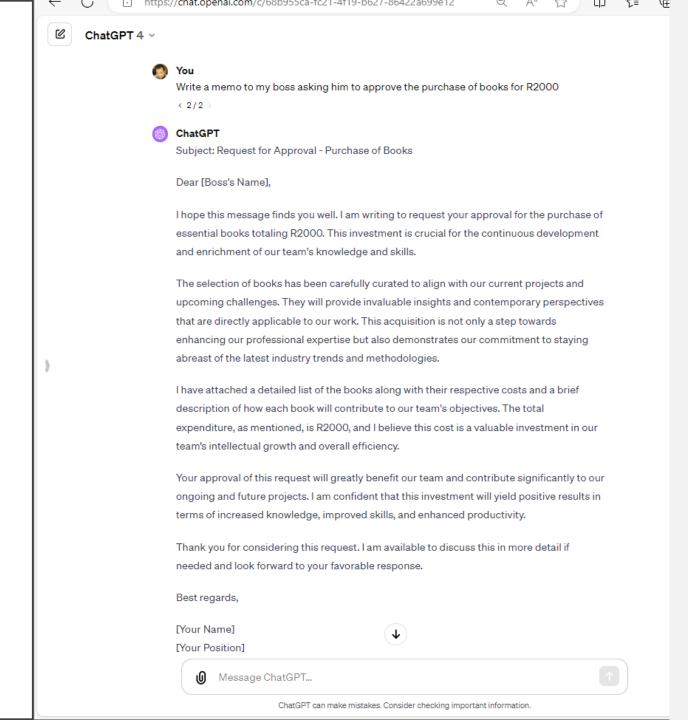
- Effortful mental activities
- Demand complex computations.
- More Subjective
- Greater Agency, choice and concentration.



SYSTEMS 1: MEMO WRITING

Prompt:

Write a memo to my boss asking him to approve the purchase of books for R2000





Prompt:

- No structure
- No context
- Not useful



SYSTEMS 2 MEMO WRITING

Prompt:

#Role

Expert Memo Writer

#Mission

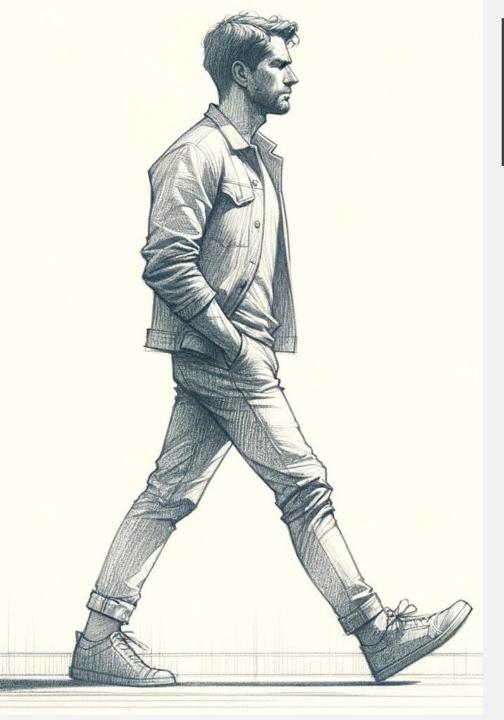
Craft a detailed memo for the CEO to approve the [your issue].

#Approach

Persuasively articulate the necessity and benefits of [your issue], ensuring alignment with organizational goals and compliance with legal and financial constraints.

#Structure

- 1. Purpose: Summarise the key arguments of the memo
- 2. Background and Motivation: [capture critical points related to your issue].
- Financial Implications: The items cost [Rxxx]
- Legal Implications: The issue requires special CEO consent, [expand with legal concerns]
- Recommendation: Summarise the central recommendation of the Memo



SYSTEMS 2 MEMO WRITING

Results:

https://chat.openai.com/share/ddd311a5-765a-46e9-9c8e-c0ca033dae5f

Structured

Context driven

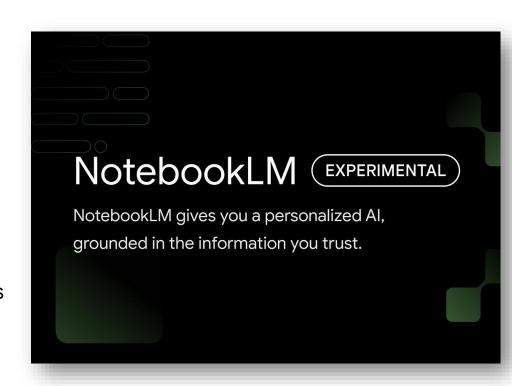
Opportunity to make edits

More thoughtful

AVAILABLE AI TOOLS

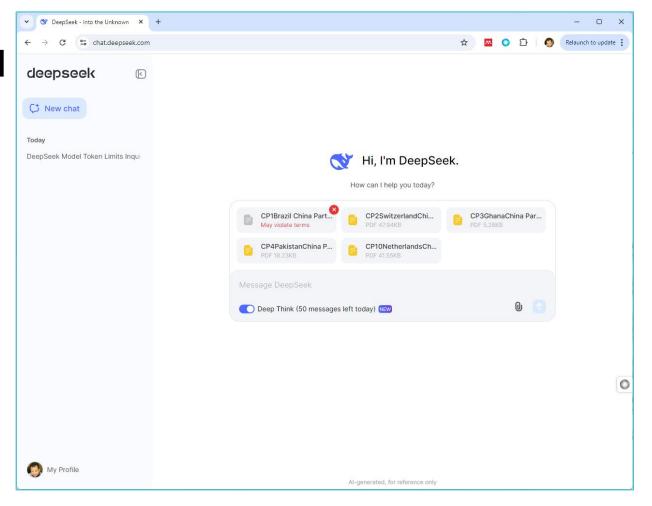
NOTEBOOK LM FROM GOOGLE

- Access: https://notebooklm.google.com/
- Google has several "experiments" https://labs.google/
- Built with Google's Gemini 1.5 Model
- Free
- Claim not to use data for training
- Can input Documents, Audio Files, Web Links, YouTube Links (with a transcript)
- Can accept 50 Sources of Content
- Can process 25 million words
 - For comparison: ChatGPT+ processes 100 thousand words



CHAT. DEEPSEEK. COM

- Access: https://chat.deepseek.com/
- Accepts up to 50 Attachments
- Integrates Reasoning with Document Analysis
- Free
- Chinese rules of acceptable content

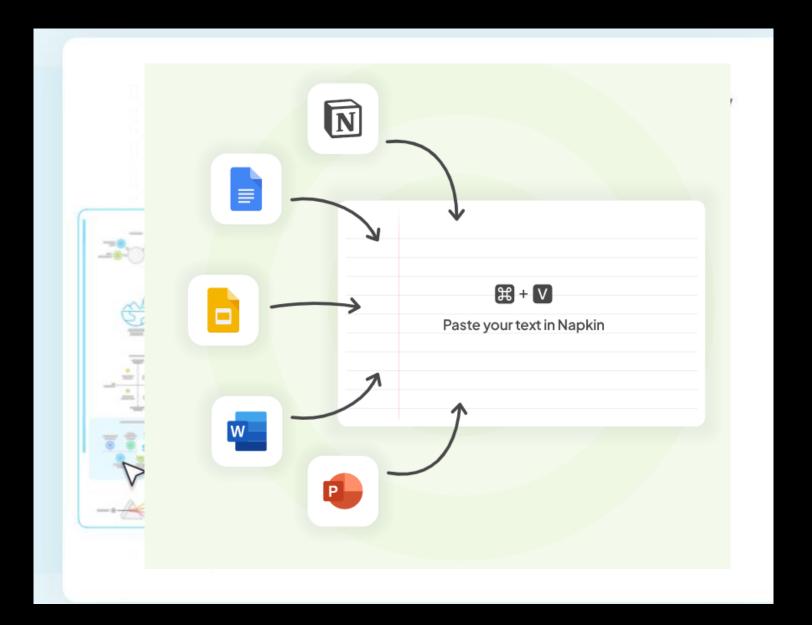


NAPKIN.AI

Access:

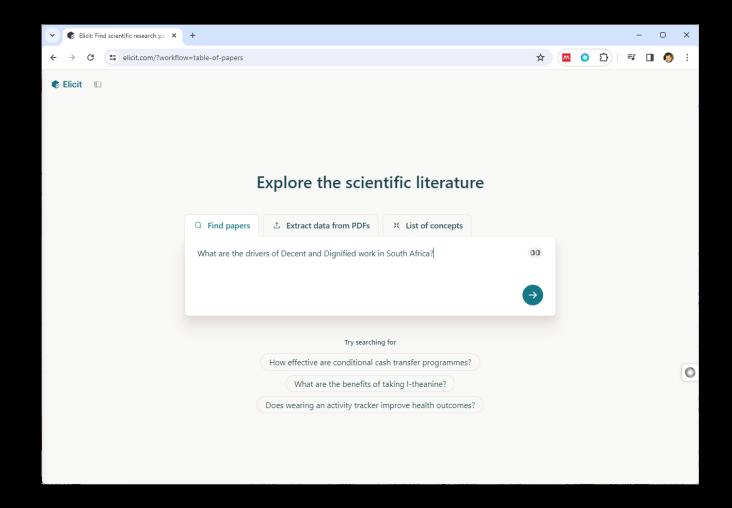
https://www.napkin.ai/

- Free (for now)
- Analyses text
 - → Identifies Insights
 - > Creates Graphics



ELICIT.COM

- https://elicit.com
- Semantic Search
- \$5 per month for the paid version
- Search for literature based on a research question, not keywords.



CLAUDE -COMPUTER USE

- https://docs.anthropic.com/en/d ocs/build-with-claude/computeruse
- Claude is used to manage a virtual computer as your worker/assistant
- Test Version
- Costs you per token and time of task
- Uses a lot of system resources
- WATCH: <u>https://www.youtube.com/watch</u> <u>?v=ODaHJzOyVCQ</u>



QUALITATIVE ANALYSIS

Exercise



PRECIS WRITING

- Use Chatbot of Choice (requires doc upload)
- Upload your paper
- Craft a Prompt to produce a Precis Summary

AI PRECIS WRITING

#Role

Expert Precis Writer and Academic Summarizer

#Mission

Create a concise and comprehensive summary of the uploaded academic paper, capturing its essence in a clear and accessible format.

#Approach

Carefully analyze the paper's structure, arguments, and key findings. Distill the main ideas and supporting evidence into a coherent summary that maintains the original's intent and significance.

#Structure

Title and Author(s): Provide the paper's title and author(s) name(s).

Research Objective: Clearly state the main purpose or research question of the paper.

Methodology: Briefly describe the research methods or approach used in the study.

Key Findings: Summarize the most important results or discoveries presented in the paper.

Main Arguments: Outline the central arguments or theoretical contributions of the paper.

Significance: Explain the importance of this research within its field or broader context.

Limitations: Mention any significant limitations or areas for future research noted in the paper.

Conclusion: Summarize the paper's overall conclusion and its implications.

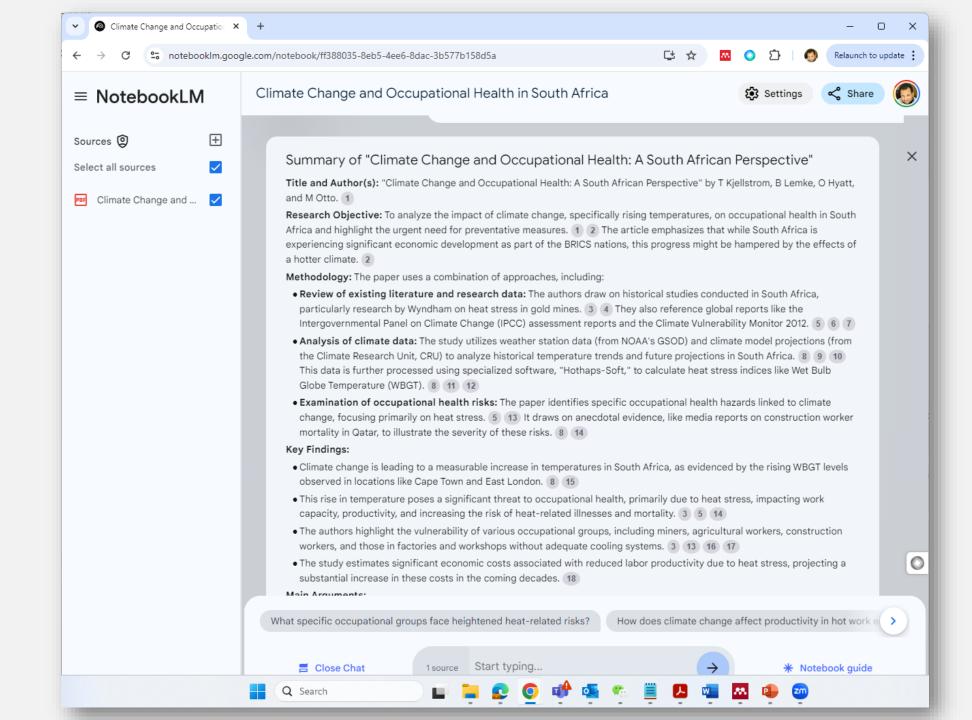
#Guidelines

Aim for a summary length of approximately 10-15% of the original paper.

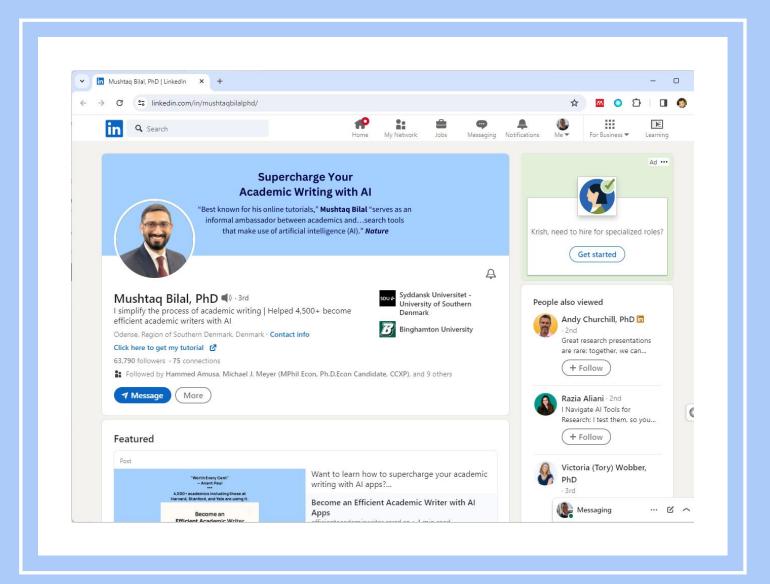
Use clear, concise language while maintaining academic integrity.

Avoid personal opinions or critiques; focus on objectively presenting the paper's content.

OUTPUT

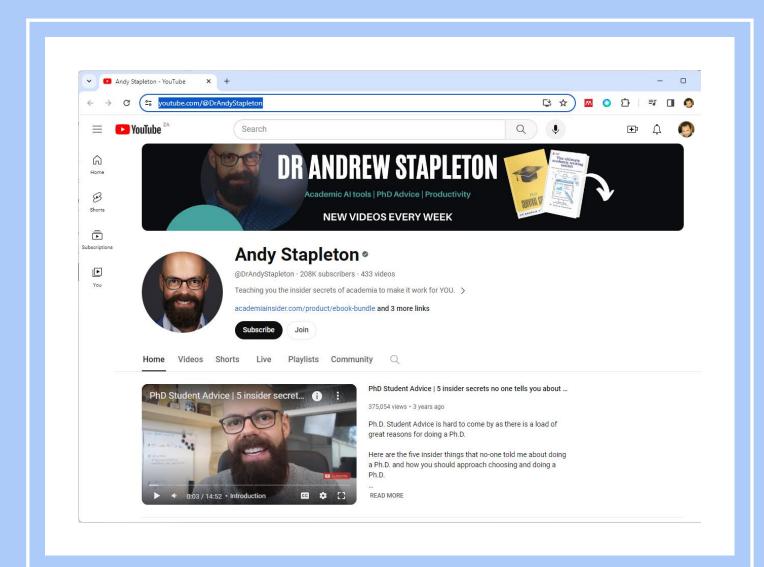


GOOD INFORMATION



MUSHTAQ BILAL ON LINKEDIN

https://www.linkedin.com/in/mushtaqbilalphd/



ANDY STAPLETON ON YOUTUBE

https://www.youtube.com/@DrAndyStapleton



Krish Chetty, Senior Research Manager, HSRC

kchetty@hsrc.ac.za