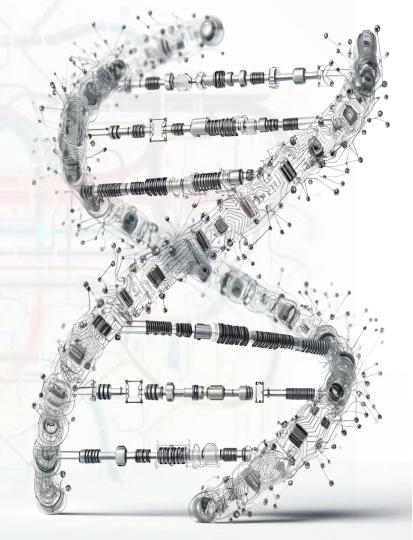


A practical guide to AI tools for life scientists.

Peering into the Future of Al-Augmented Scientific Research.



Michael Kuiper 8th May 2024.







Overview:

What:

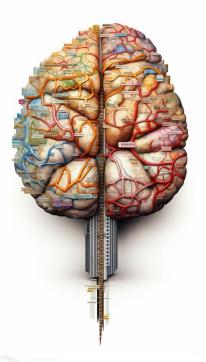
- The recent timeline of Large Language Models.
- What are LLMs anyway?

Why:

- Why use LLMS over search engines?
- The information challenge.

How:

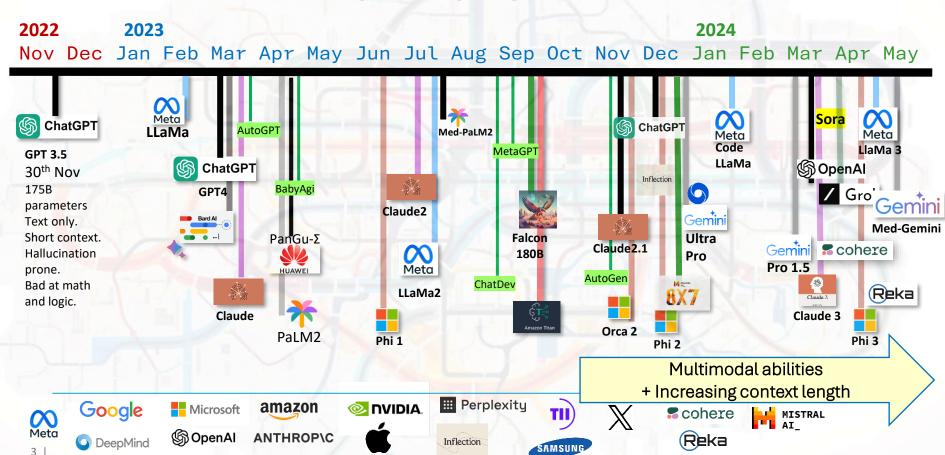
- LLM based Agents, Assistants and Tools.
- How we might use AI for research & dev. (demos)
- How to get the most from your 'cognitive co-pilot' and how to avoid the pitfalls.







The Rise of Large Language Models.







What are LLMs anyway?:

Evolution of Consumer Large Language Models:

- LLMs have evolved from earlier **chatbots**, which have been around for decades.
- Essentially, they are computer algorithms that are incredibly good at generating output for given input. No magic, just a lot of data, math and matrices!
- Trained on huge datasets, they learn associations between words, and in doing so build a broad 'general knowledge' about the world. They can be used to process natural language, understand context, and dynamically generate coherent responses from complex gueries.
- Increasingly they have become more functional with multimodal abilities. They can even run other programs and talk collaboratively to other LLMs.

Excellent overviews on Youtube:

Andrej Karpathy: Intro to Large Languange Models.

3Blue1Brown: But what is a GPT? and Attention in transformers.

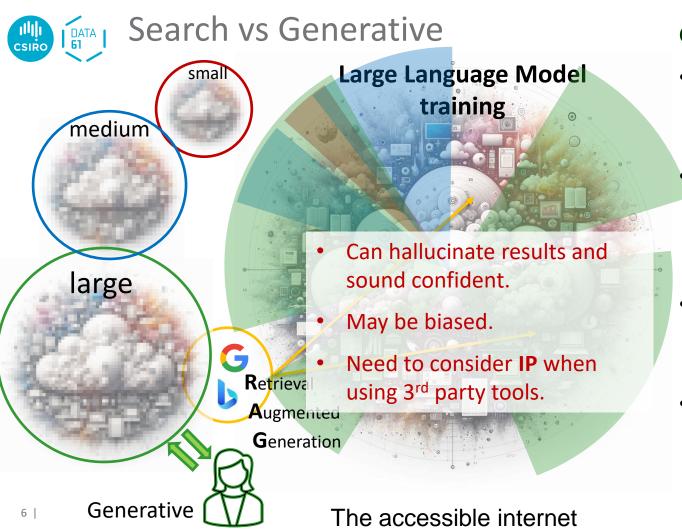


Search vs Generative



The problem with search:

- You have to know what you are looking for.
 - You must interpret information from the provided webpage. (Also, ads & click bait!)
 - The gatekeepers order search results based on algorithms/sponsorship. They watch you.
- Keyword and context ambiguity.



Generative AI:

- You can interact with LLMs using natural language (NL).
- LLMs leverage **NL** to access interdisciplinary knowledge **across diverse domains**.
- They are multimodal and multilingual. Can be domain specific.
- Ever increasing capabilities: higher quality and larger context windows.





Why use LLMs over Search Engines?

Typical search engine query:

'Influenza reviews'

'How to code pairwise alignment of proteins'

'Antibiotics and drug resistance'

Typical LLM interaction:

"What can you tell me about influenza? Be detailed and summarize the main points."

"How would I code pairwise alignment of proteins?

Can you write an app in python and explain the steps of how it works?"

"What is antibiotic resistance, - can you please give me a general overview? **Explain like I am 12**."





[DATA | It is all in how you ask: The power of Chain of Thought prompting

If it takes three hours to dry three towels on a line, how long would it take to dry nine towels? If it takes three hours to dry three towels on a line, how long would it take to dry nine towels? **Think about this** carefully before you answer and explain your reasoning.

ChatGPT:

If it takes three hours to dry three towels, then it takes 3 hours/3 towels = 1 hour/towel.

To dry nine towels at the same rate:

Time = (number of towels)×(time per towel) Time=(number of towels)×(time per towel)

It would take 9 hours to dry nine towels on the line, assuming the drying rate remains consistent.

ChatGPT:

Let's think about it step by step:

If it takes three hours to dry three towels, then it takes three hours for each towel to be fully dry. In other words, if you put three towels on a line, and come back in three hours, all three towels will be dry.

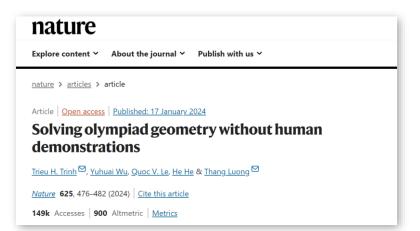
Now, if you add another three towels to the line, these new towels will also take three hours to dry. But since you're hanging them at the same time as the first three towels, all six towels will be dry in the same three hours.

Using the same reasoning, if you add yet another three towels (making it a total of nine towels), all nine towels will still take three hours to dry when hung up simultaneously.

So the answer is: it would still take three hours to dry nine towels when they're hung up at the same time on a long line.



LLMs are bad at math and logic. Usually.



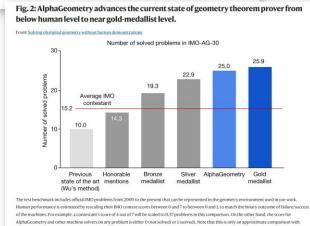




dolphin-2.7-mixtral-8x7b MoE



Fig. 1: Overview of our neuro-symbolic AlphaGeometry and how it solves both a simple problem and the IMO 2015 Problem 3. AlphaGeometry d Solution a A simple problem Symbolic deduce solved Construct "Let ABC be any triangle with AB = AC. AB=AC, BD + DC, AD=AD ⇒ ∠ABD+∠DCA [1] Prove that ∠ABC = ∠BCA. c Language model [1], B C D collinear → ∠ABC=∠BCA e IMO 2015 P3 f Solution Construct D: midpoint BH [a] *Let ABC be an acute triangle. Let [a], 0, midpoint HQ ⇒ BQ # 0,D [28] (O) be its circumcircle. Hits Construct G: midpoint HC [b] orthocenter, and F the foot of the \angle GMD = \angle GO,D \Rightarrow M O, G D cyclic [26] altitude from A. Let M be the midpoint of BC. Let Q be the point -Alpha-[a],[b] - BC / DG (38) on (O) such that QH \(\t \t QA \) and let K Geometry Construct E: midpoint MK [c] be the point on (O) such that KH \(\pm \) KQ. Prove that the circumcircles ..., [c] ⇒ ∠KFC = ∠KO,E [184] (O₄) and (O₄) of triangles FKM and ∠FKO,=∠FKO, → KO, // KO, [189] KQH are tangent to each other."

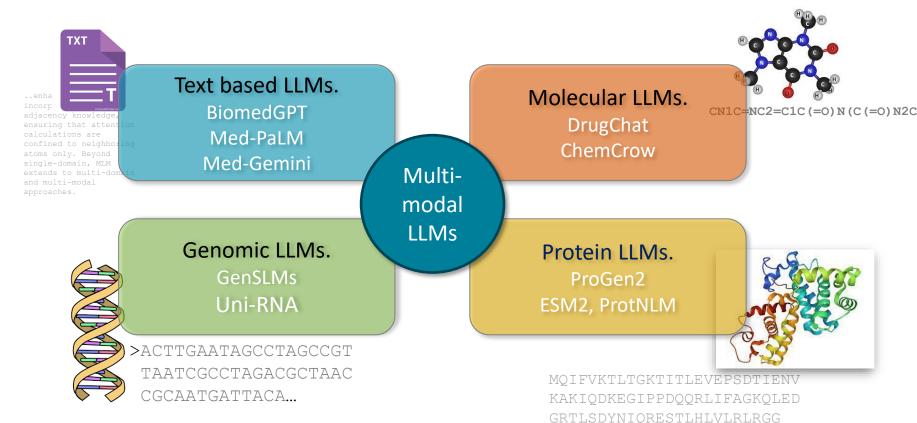


humans on classical geometry, who operate on natural-language statements rather than narrow, domain-specific translations, Further, the general IMO contest also includes other types of problem, such as geometric inequality or combinatorial geometry, and other domains of mathematics, such as algebra, number theory and combinatorics.





Scientific LLMs, not just words:







Why do we need AI?



- 1) Managing Information Overload
 - Analysing vast amounts of data and literature
- 2) Solving Complex, Interdisciplinary Problems
 - Uncovering new connections; your personal 'brainstorm buddy'
- 3) Al as a Personalized Tutor and Trainer
 - Adapting to individual learning needs
- 4) Enhancing Cognitive Capabilities
 - Al surpassing human performance



Garry Kasparov vs Deep Blue (1996)______



Lee Sedol
vs AlphaGo
(2016) DeepMino





Google DeepMind Google Research

2024-04-29

Capabilities of Gemini Models in Medicine

Khaled Saabo, Tao Tuo, Tuo, Wei-Hung Wengo, Ryutaro Tanno, David Stutz, Ellery Wulczyn, Ryutaro Tanno Fan Zhang*, Tim Strother*, Chunjong Park*, Elahe Vedadi*, Juanma Zambrano Chaves*, Szu-Yeu Hu*,1, Mike Schaekermann*,1, Aishwarya Kamath*,2, Yong Cheng*,2, David G.T. Barrett*,2, Cathy Cheung*,1, Basil Mustafa*,2, Anil Palepu*,1, Daniel McDuff*,1, Le Hou*,2, Tomer Golany*,4, Luyang Liu*,1, Jean-baptiste Alayrac*,2, Neil Houlsby*,2, Nenad Tomasev*,2, Jan Freyberg*,1, Charles Lau1, Jonas Kemp¹, Jeremy Lai¹, Shekoofeh Azizi², Kimberly Kanada¹, SiWai Man¹, Kavita Kulkarni¹, Ruoxi Sun³, Siamak Shakeri², Luheng He², Ben Caine², Albert Webson², Natasha Latysheva², Melvin Johnson², Philip Mansfield¹, Jian Lu¹, Ehud Rivlin⁴, Jesper Anderson¹, Bradley Green¹, Renee Wong¹, Jonathan Krause¹, Jonathon Shlens², Ewa Dominowska¹, S. M. Ali Eslami², Katherine Chou², Claire Cui², Oriol Vinyals², Koray Kavukcuoglu², James Manyika¹, Jeff Dean^{1,2}, Demis Hassabis², Yossi Matias¹, Dale Webster¹, Joelle Barral², Greg Corrado¹, Christopher Semturs¹, S. Sara Mahdavi^{*,2}, Juraj Gottweis^{*,3}, Alan Karthikesalingam*,1 and Vivek Natarajan†,1 °Co-first, *Core, †Technical Lead, †Senior Lead, ¹Google Research, ²Google DeepMind, ³Google Cloud, 4Verily

Excellence in a wide variety of medical applications poses considerable challenges for AI, requiring advanced reasoning, access to up-to-date medical knowledge and understanding of complex multimodal data. Gemini models, with their strong general capabilities in multimodal and long-context reasoning, offer exciting possibilities in medicine. Building on these core strengths of Gemini 1.0 and Gemini 1.5, we introduce Med-Gemini, a family of highly capable multimodal models that are specialized in medicine with the ability to seamlessly integrate the use of web search, and that can be efficiently tailored to novel modalities using custom encoders. We evaluate Med-Gemini on 14 medical benchmarks spanning text,

May





How: The Al toolbox. (as of early 2024*)

Generative Al.







S ChatGPT





Al research assistants.









Al coding assistants.





GitHub Copilot codeium

AlphaCode 2

Image/video/music Generative Al.













AI structural biology.

AlphaFold



RF diffusion

RosettaFold All Atom

Al multi-agents.









Custom GPTs Assistants.



A mentor bot encouraging critical thinking across various subjects.



A formal, academic guide for scientific innovation



I sharpen text for clarity and impact.



A chatbot that creates downloadable 'memory' files from chat sessions for continuity



How: Demo of ChatGPT / Gemini / Claude 3

- or, "What could go wrong?"





How: Demo of GPTs assistants



Aldeator: A tool for understanding and strengthening ideas with insightful dialogue. https://chat.openai.com/g/g-4UrZNpH50-aideator



Paper Mache: Reviews uploaded Scientific Papers and provides a summary and critique. https://chat.openai.com/g/g-uJvQlkk0d-paper-mache



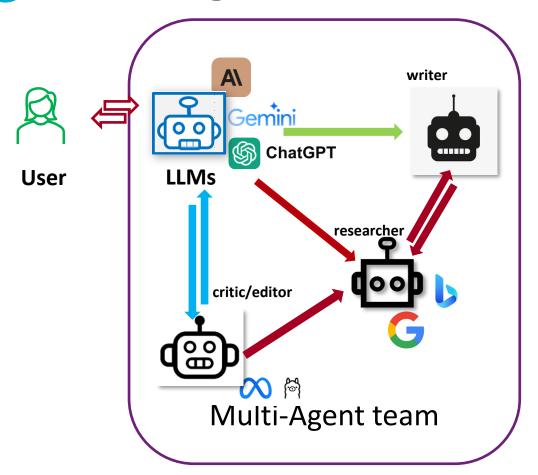
Meeting Scribe: I create meeting summaries and action items from transcripts. https://chat.openai.com/g/g-0aWDdYa28-meeting-scribe



Text Razor: I sharpen text for clarity and impact. https://chat.openai.com/g/g-FExztl0NU-text-razor



LLMs as Agents.



Multi-Agent Frameworks:

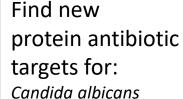


LLMs as Agents.

```
researcher = Agent(
 role='Senior Researcher',
goal='Perform comprehensive scientific research on the given topic: {topic}. Use reliable sources
and critically evaluate the information found.',
verbose=True,
 memory=True,
 backstory="""As a highly skilled researcher, you provide in-depth and comprehensive reports.
Your research begins with finding relevant articles for the subject matter and reading their
content. Provide at least TEN references which MUST be cited at the end of the report using
proper citation format.
The report should be structured with an introduction, background, main findings, discussion,
conclusion, and references sections. You will provide your report to the reviewer who will provide
constructive feedback, and you will use this feedback to improve your article.""",
tools=[search2, website rag],
 Ilm=gpt4,
allow delegation=True
```



Custom Al agent research teams.



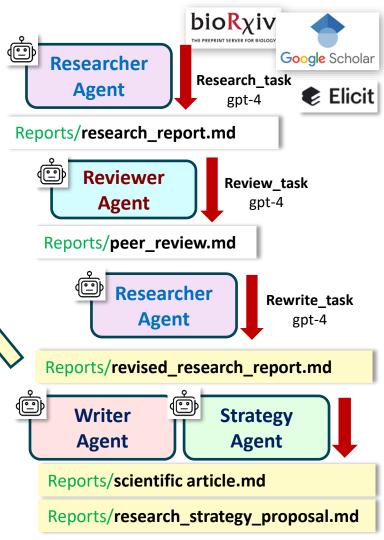
userproxy



Project Manager

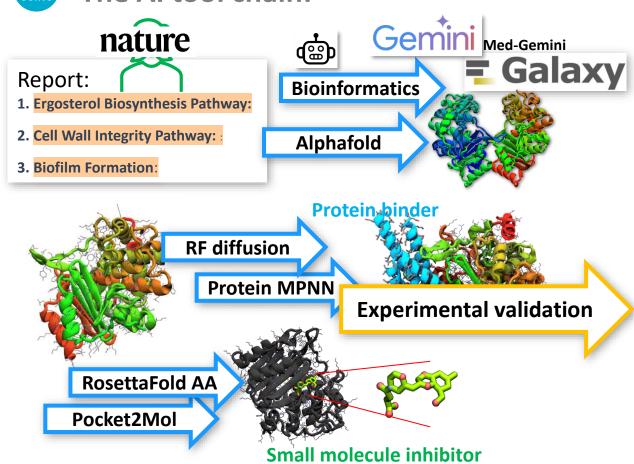


- **1. Ergosterol Biosynthesis Pathway:** This pathway is essential for maintaining fungal membrane integrity. While ERG11 is already targeted by azoles, ERG6, involved in sterol biosynthesis, shows promise based on preliminary studies (Watanabe et al., 2012).
- **2.** Cell Wall Integrity Pathway: Proteins such as glucan synthases, crucial for cell wall synthesis and remodeling, are potential targets. Their disruption can lead to increased susceptibility to osmotic stress and antifungal agents.
- **3. Biofilm Formation:** Proteins like Biofilm-associated protein (Bap) and adhesins such as Als3 are vital for biofilm formation and maintenance, contributing significantly to antifungal resistance
- * References legit and relevant. Pop science article and strategy report high level.





The AI tool chain:





The AI tool chain:

nature

Article Open access Published: 20 December 2023

Autonomous chemical research with large language models

Nature 624, 570–578 (2023) Cite this article

nature chemical engineering

Article Open access Published: 11 January 2024

Self-driving laboratories to autonomously navigate the protein fitness landscape

<u>Jacob T. Rapp</u>, <u>Bennett J. Bremer</u> & <u>Philip A. Romero</u> ✓

Nature Chemical Engineering 1, 97–107 (2024) Cite this article



Robotlabs?

Autonomous research?

Google RT-2?







What possibly could go wrong?



Contents lists available at ScienceDirect

Surfaces and Interfaces

journal homepage: www.sciencedirect.com/journal/surfaces-and-interfaces





Constant vigilance.

The three-dimensional porous mesh structure of Cu-based metal-organic-framework - aramid cellulose separator enhances the electrochemical performance of lithium metal anode batteries

Manshu Zhang ^{a,1}, Liming Wu ^{a,1}, Tao Yang ^b, Bing Zhu ^a, Yangai Liu ^{a,*}

1. Introduction

Certainly, here is a possible introduction for your topic: Lithium-metal batteries are promising candidates for high-energy-density rechargeable batteries due to their low electrode potentials and high theoretical capacities [1], [2]. However, during the cycle,

^a Beijing Key Laboratory of Materials Utilization of Nonmetallic Minerals and Solid Wastes, National Laboratory of Mineral Materials, School of Materials Science and Technology, China University of Geosciences, Beijing100083, China

^b College of Materials & Environmental Engineering, Hangzhou Dianzi University, Hangzhou 310036, China





Recap: How to get more from your 'cognitive co-pilots'.

- Take time to set the context of your questions:
 Be conversational. Use chain of thought prompting.
- If something is not clear, ask for clarification. Fill your gaps of knowledge. Ask to test your understanding. Don't be lazy.
- Check everything. Ask for sources.
- When brainstorming, remember to ask for additional ideas and critique of your own.
- Be mindful of IP and privacy when using tools.





The Future of Al in Science.

- Large commercial investments and rapid developments:
 - Hard to predict what new capabilities will exist next week: Be agile. Don't believe all the hype.
- Transformative potential:
 - New capabilities will radically alter how we process and interpret data. ie) multimodal and data analytics in OpenAI and Gemini models
 - Can we use AI to drive autonomous research (via agents/robotics) and software development?
- Big questions remain on long term consequences of using AI tools:
 - O Do we lose cognitive and critical thinking skills?
 - O Do we still need to be domain experts?
 - O How much do we trust the output?



Thank You

Where to next?

Resources:



- Al training
- Responsible AI
- Industry networking



- Matt Wolfe
- Matthew Berman
- AI explained

ANTHROP\C Prompt library:

https://docs.anthropic.com/claude/prompt-library



michael.kuiper@csiro.au