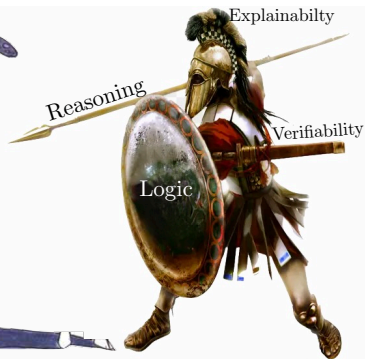
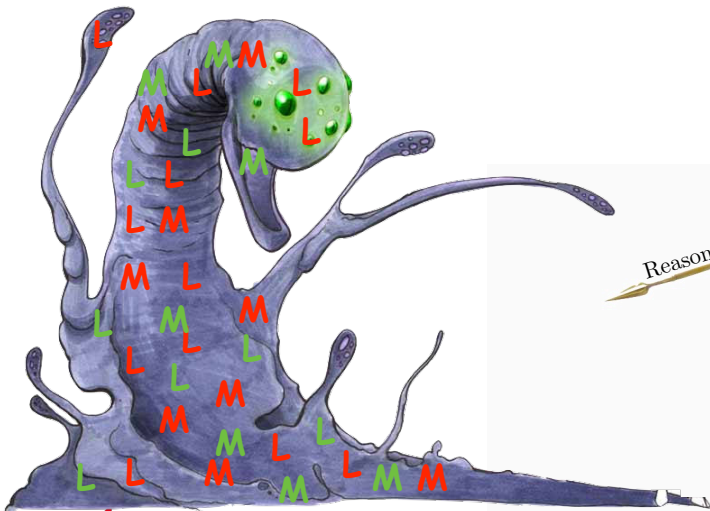
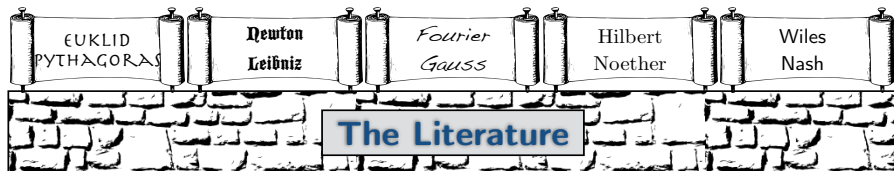


Theorem Provers and the Future AI Math Ecosystem

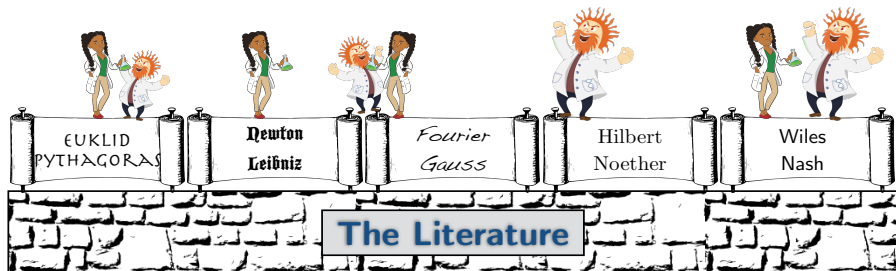
Stephan Schulz



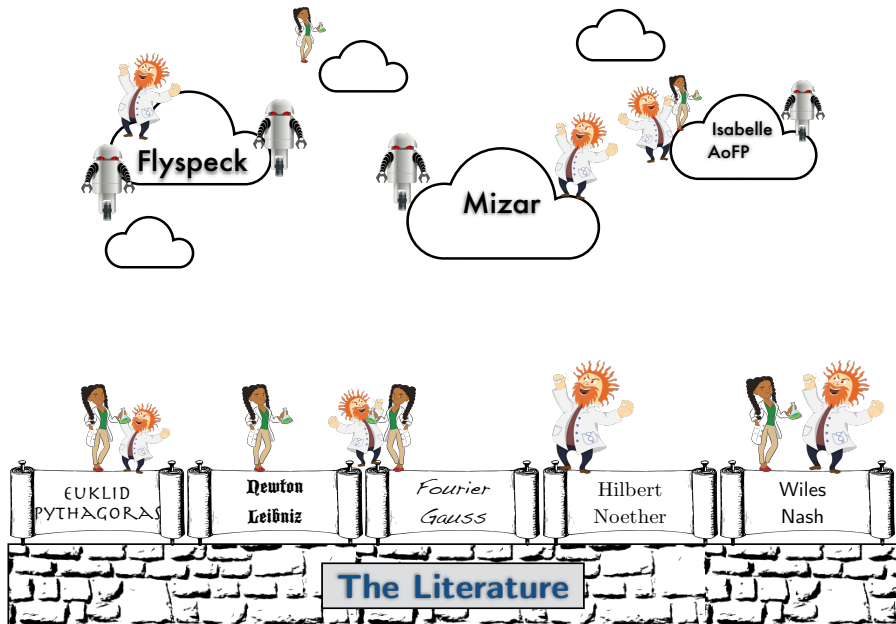
The math ecosystem (ca. now)



The math ecosystem (ca. now)

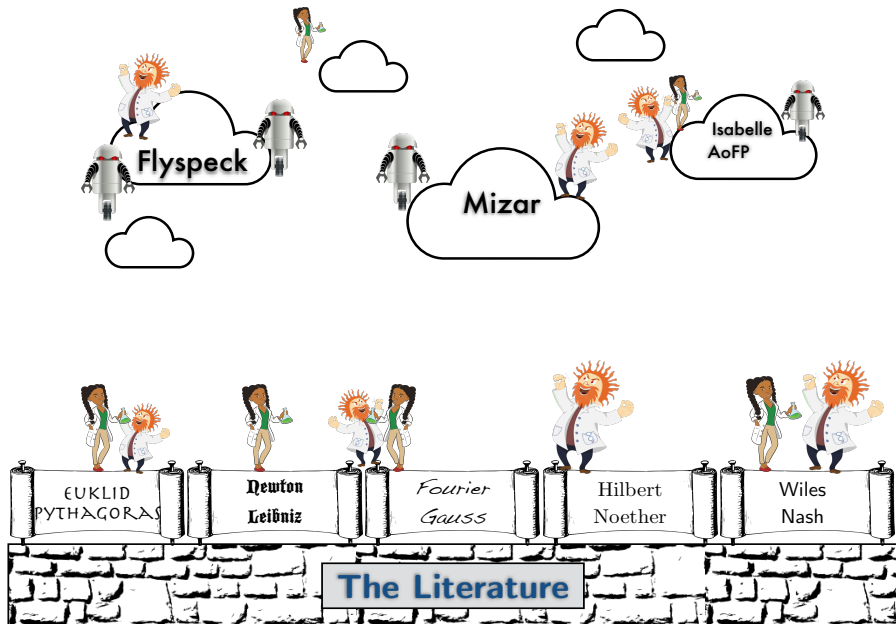


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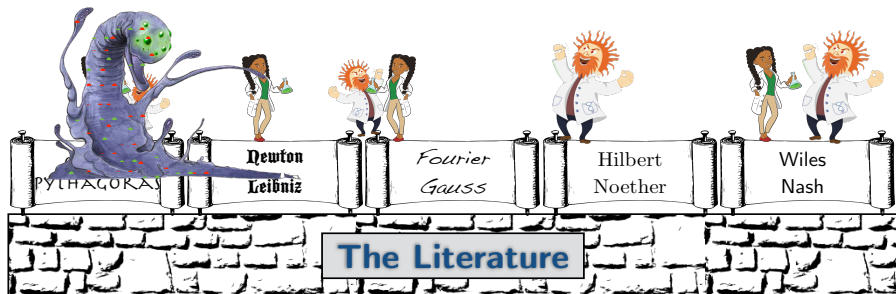
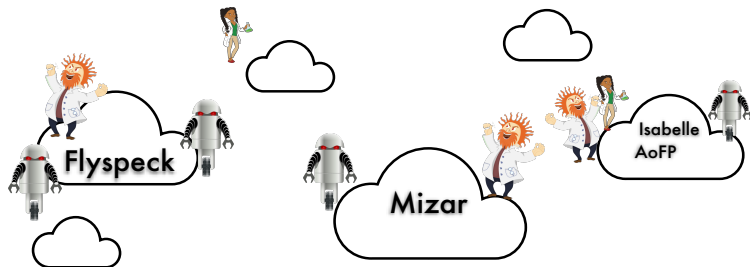


Enter
THE AGE OF ARTIFICIAL INTELLIGENCE

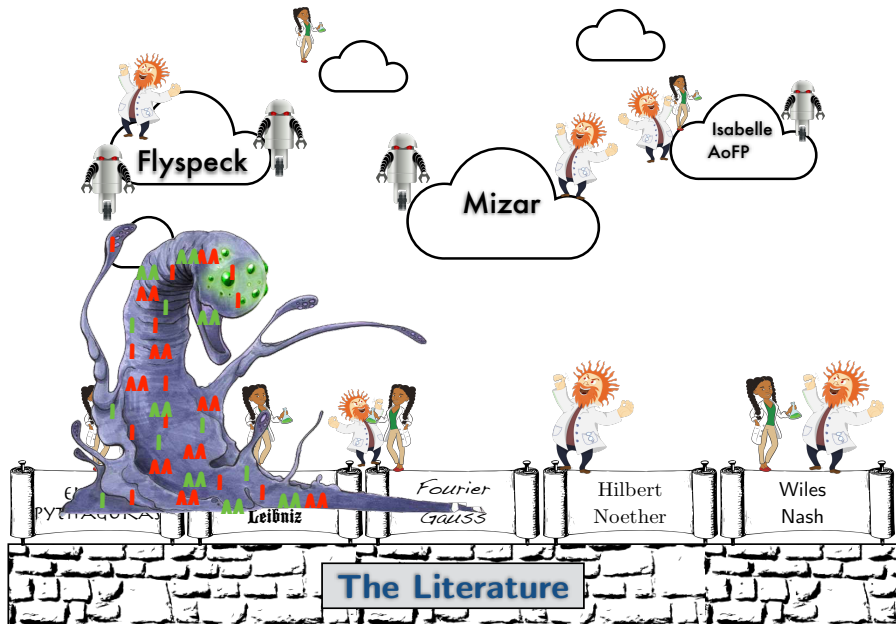
The future math ecosystem (OpenAI/xAI/... version)



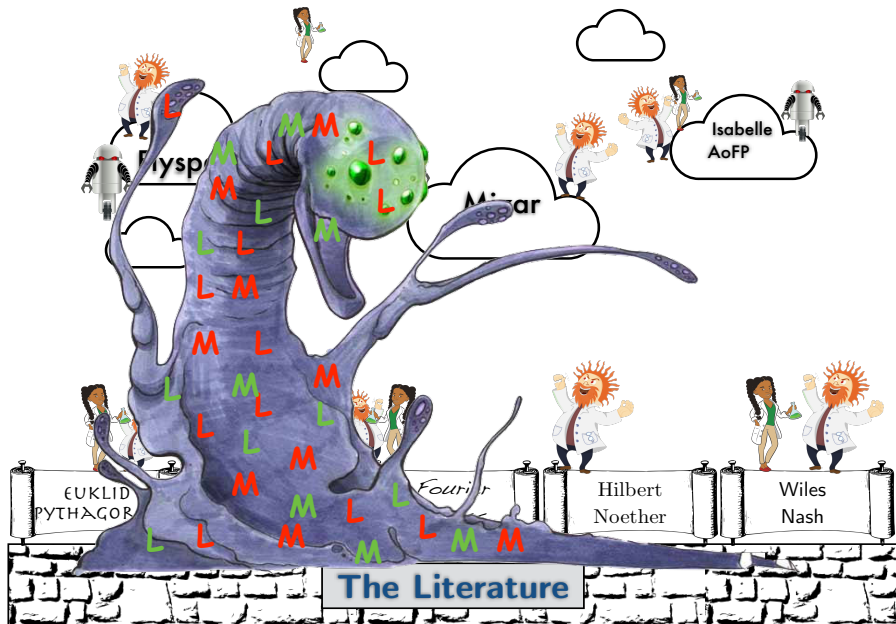
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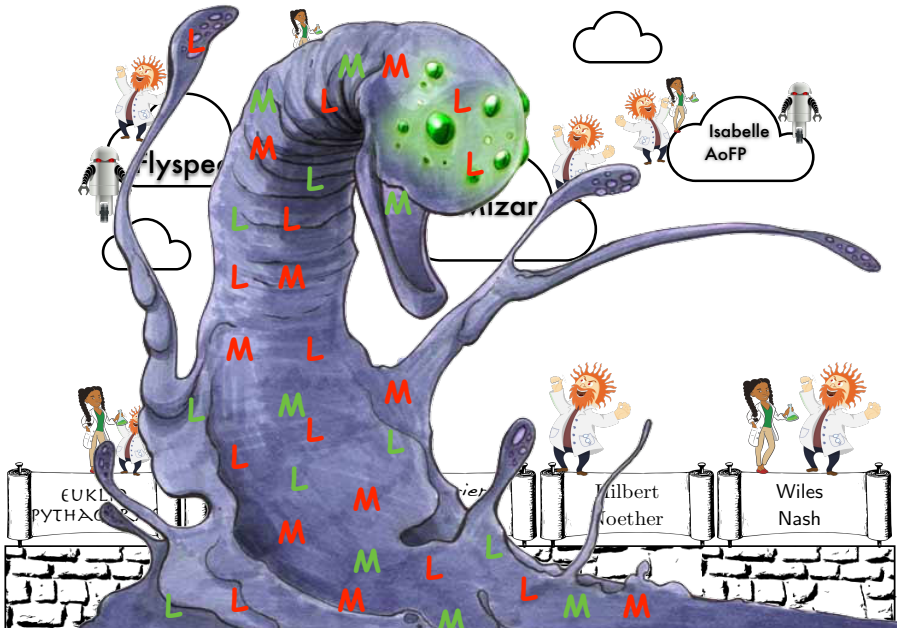
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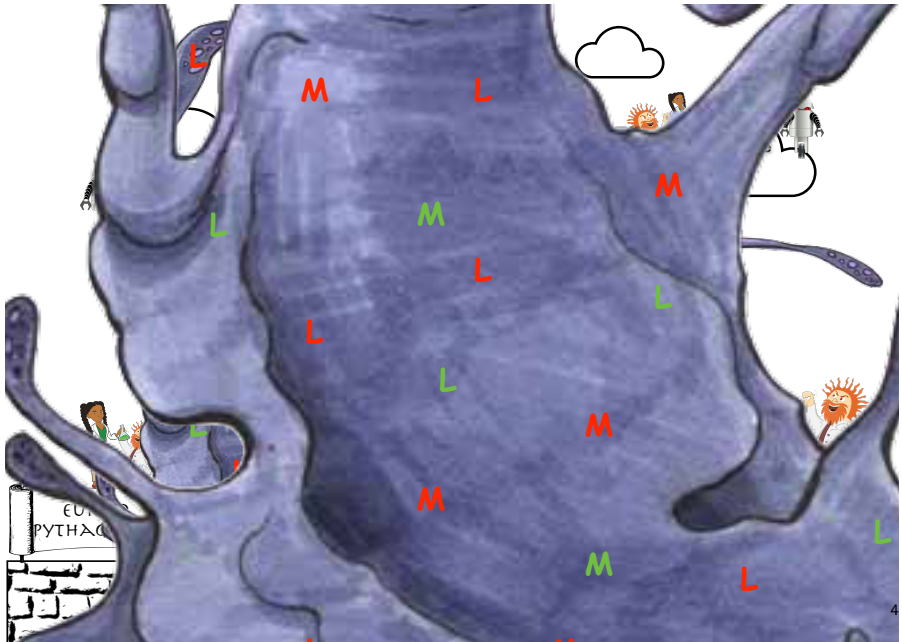
The future math ecosystem (OpenAI/xAI/... version)



The future math ecosystem (OpenAI/xAI/... version)



The future math ecosystem (OpenAI/xAI/... version)



This is not the future!

Large language models are amazing. . .

Large language models are amazing...

- ▶ ...at processing *language*

Bold claims

Mathematics is *not* a language

Claiming otherwise is the same confusion as that between being able to program, and knowing a programming language (which gave us the horror that is COBOL). Or between knowing the rules of chess and being a competent chess player.

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- ▶ Of course there is a mathematical language
 - ▶ ... with a lot of national dialects
 - ▶ ... with a lot of field-specific dialects

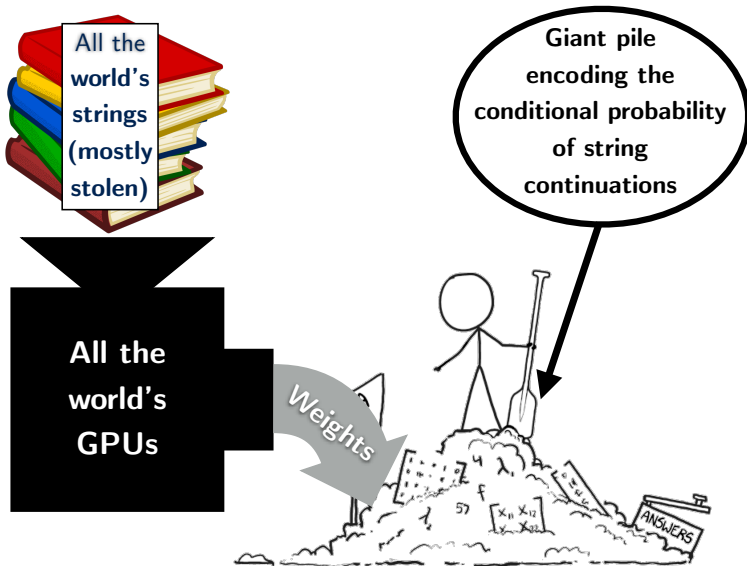
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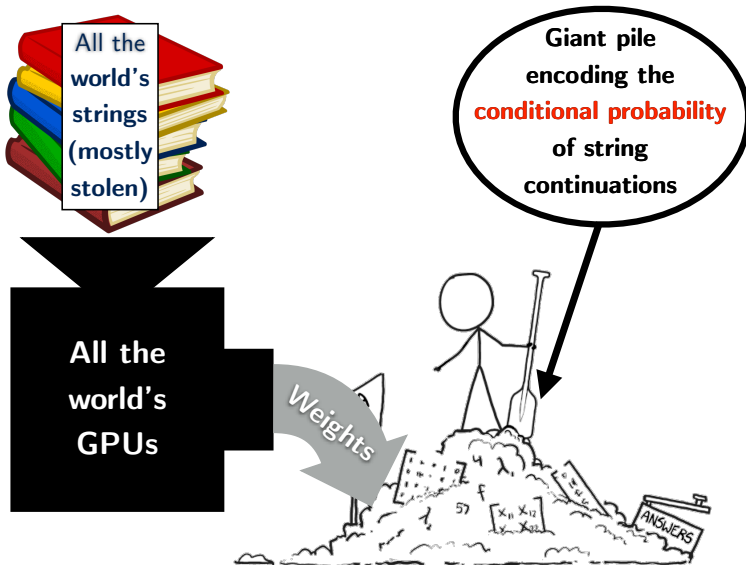
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Mathematics is, at the core, the definition and exploration of new structures and their properties.

Genesis of an LLM



Genesis of an LLM



- ▶ LLMs learn conditional probability distributions on language
 - ▶ They do not learn abstract theory
 - ▶ They do not learn abstract reasoning
 - ▶ They do not learn calculations (except by memorization)
- ▶ LLMs are essentially reproductive, not creative

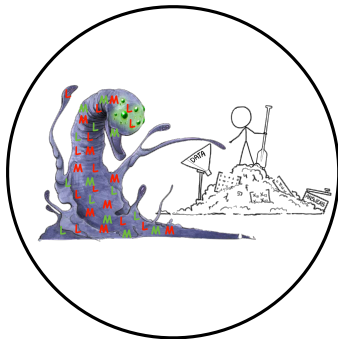
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LLMs can't do (new) mathematics!

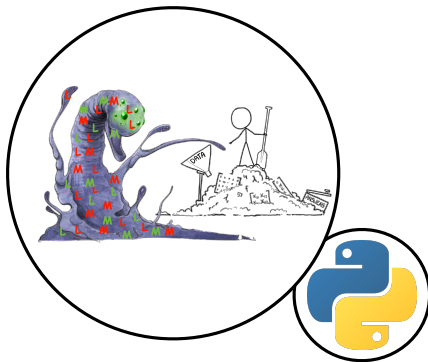
LLMs can't handle abstract reasoning

- ▶ Systematically renaming terms kills performance (*A chat with Bard*, [SMS25])
- ▶ Adding unrelated information leads to failure (*GSM-Symbolic: Understanding the Limitations of Mathematical Reasoning in Large Language Models*, [MAS⁺24])
- ▶ Answers are not robust (*Fidelity of Medical Reasoning in Large Language Models*, [BJC⁺25])
- ▶ Hallucinations are likely unavoidable (*LLMs Will Always Hallucinate, and We Need to Live With This*, [BAS24])
- ▶ Models only “fake” understanding (*Potemkin Understanding in Large Language Models*, citeMWVM:Potemkin-2025)
- ▶ Even so-called *reasoning models* break gracefully (*The Illusion of Thinking : Understanding the Strengths and Limitations of Reasoning Models via the Lens of Problem Complexity*, [SMA⁺25])

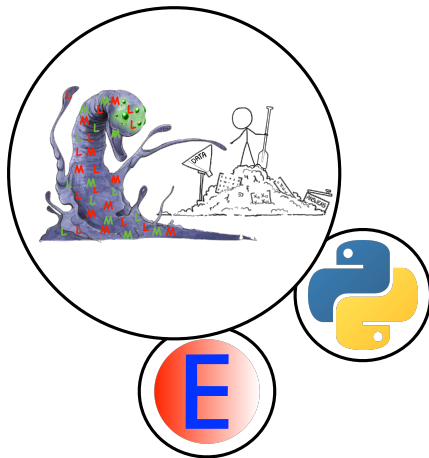
Fixing LLMs (?): The LLM+ Architecture



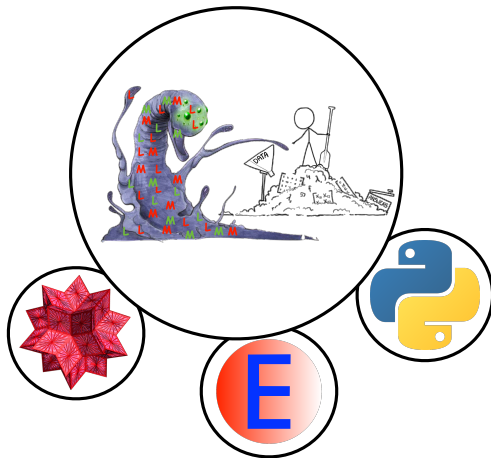
Fixing LLMs (?): The LLM+ Architecture



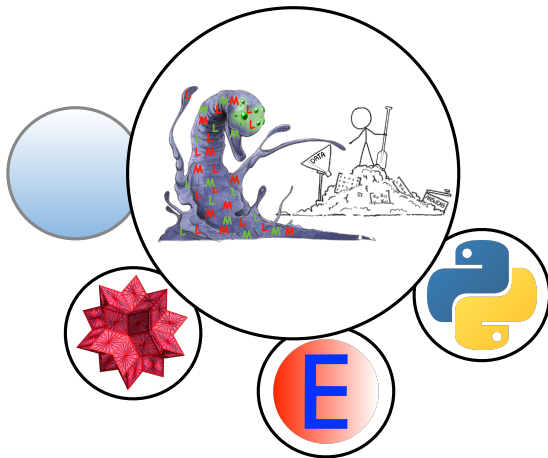
Fixing LLMs (?): The LLM+ Architecture



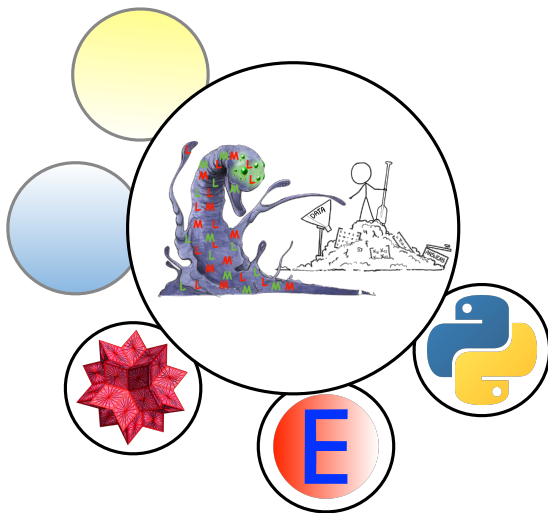
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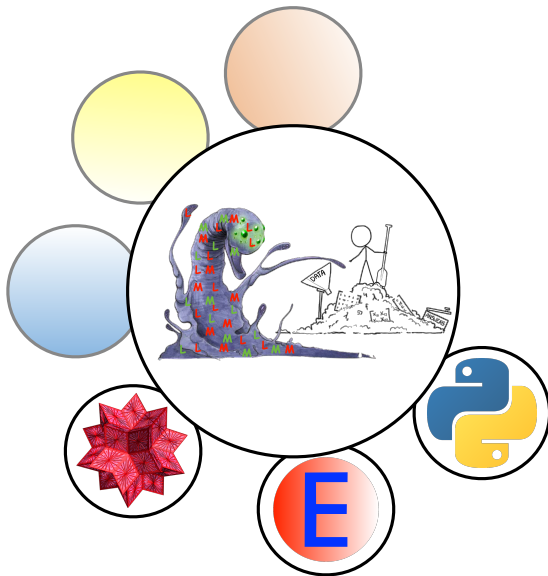
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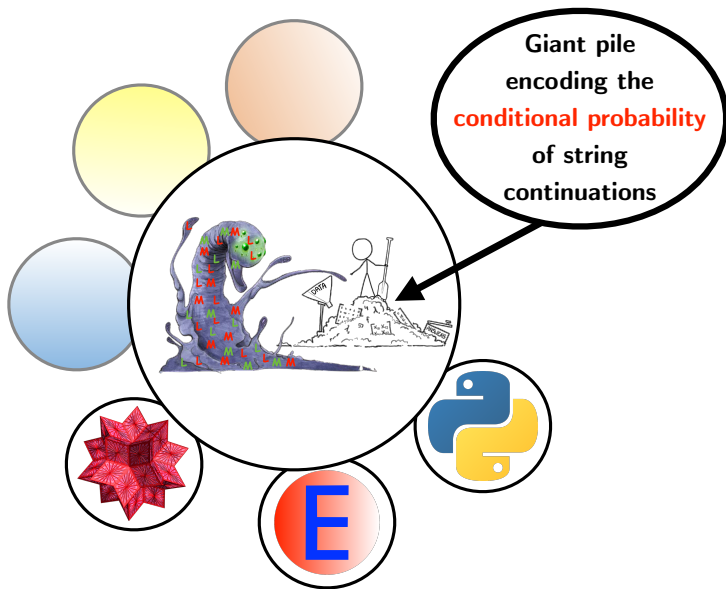
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Fixing LLMs (?): The LLM+ Architecture



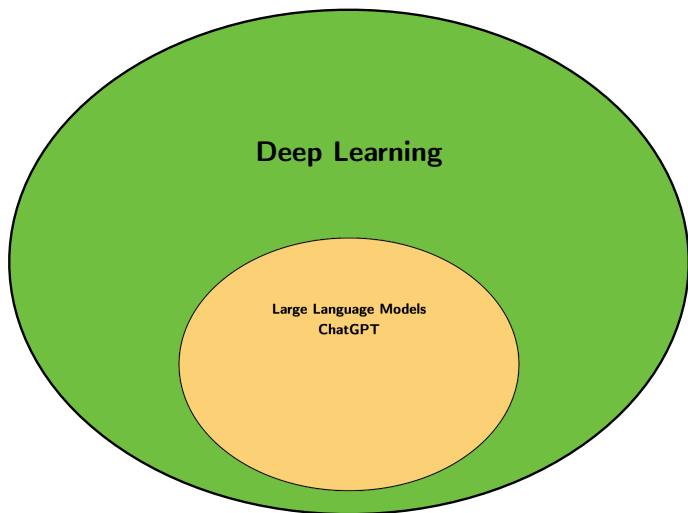
If LLM-based systems are not the future, what is?

ChatGPT is not AI

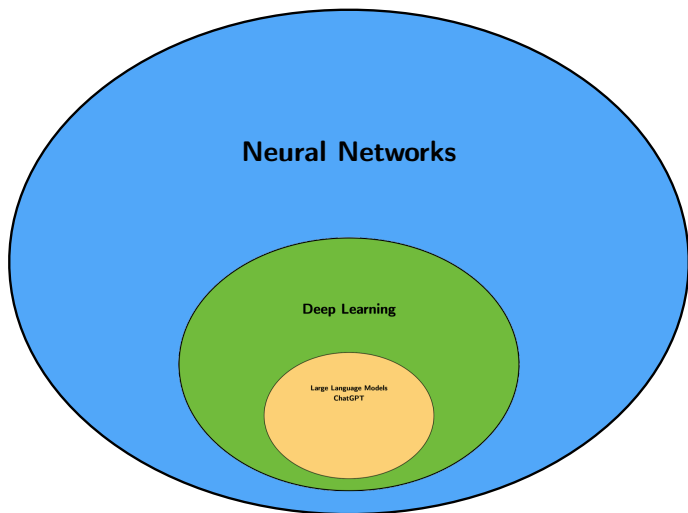


Large Language Models
ChatGPT

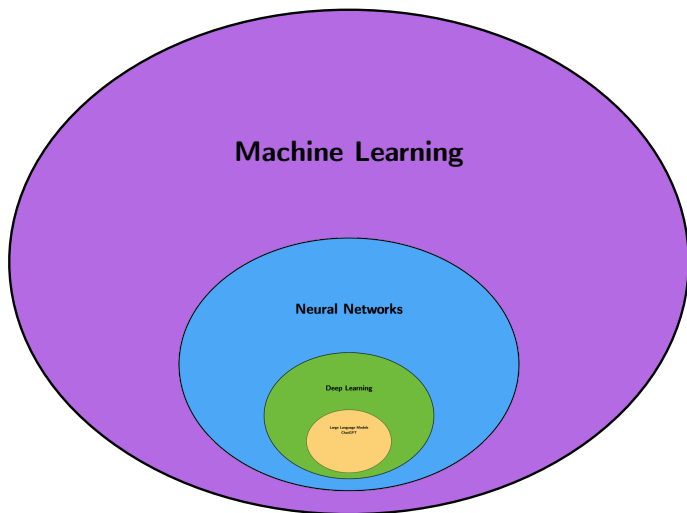
ChatGPT is not AI



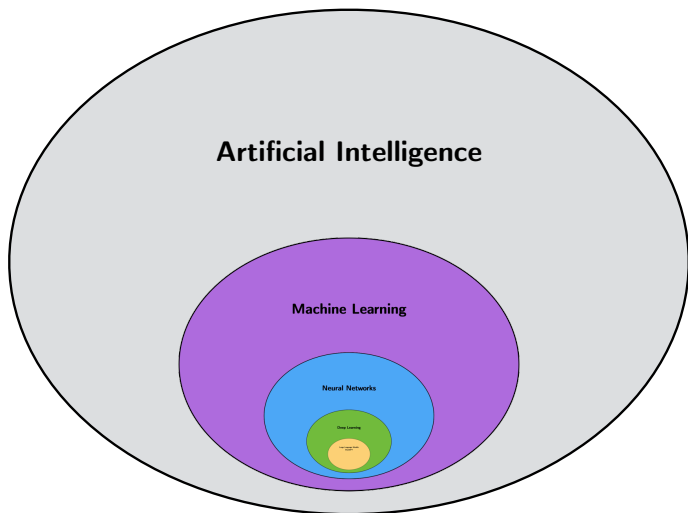
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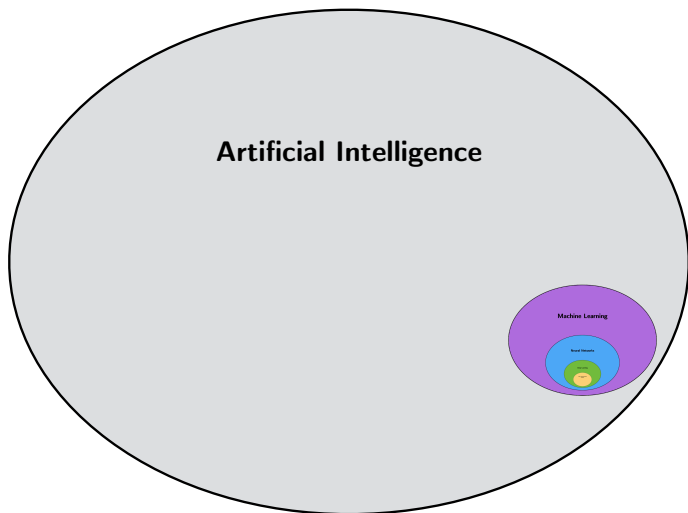
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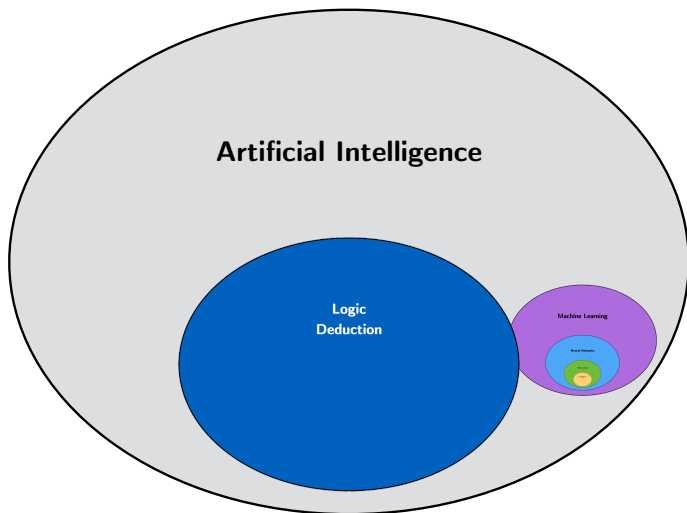
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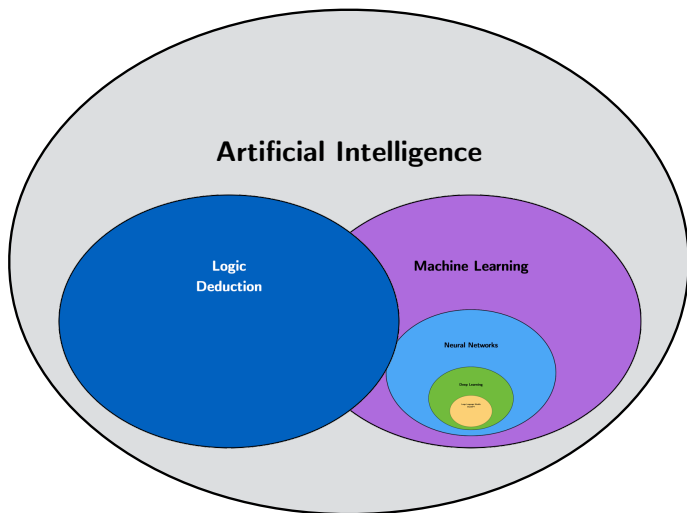
ChatGPT is not AI



ChatGPT is not AI



ChatGPT is not AI



2500 Years of History

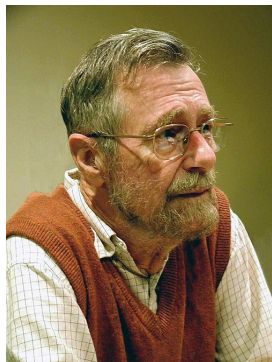
- ▶ LLMs are great at processing (natural) language
 - ▶ Natural language is ambiguous
 - ▶ Natural language is imprecise
- ▶ Mathematics has moved towards more formal language for its entire history
 - ▶ Euklid (formal definitions and arguments)
 - ▶ Al-Khwarizmi et al. (Algebraic formula language)
 - ▶ Newton and Leibniz (Formalisms for infinitesimals, *Calculus*)
 - ▶ Bool, Frege, Church, ... (Formal logics)
 - ▶ Russell and Whitehead (*Principia Mathematica*)
 - ▶ Mizar/Isabelle/HOL/Lean/FlySpeck ...

2500 Years of History

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Future Mathematics will be based on formal logic!

Argument by Authority (and analogy)

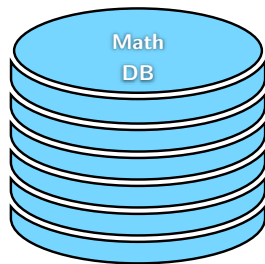


Edsger W.
Dijkstra [Dij78]:
*On the foolishness of
“natural language
programming”*

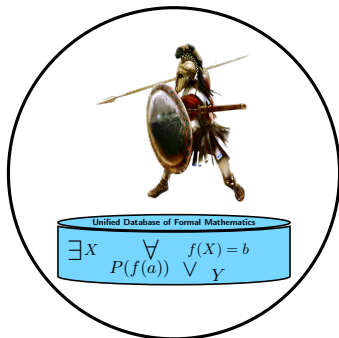
It may be illuminating to try to imagine what would have happened if, right from the start our native tongue would have been the only vehicle for the input into and the output from our information processing equipment. My considered guess is that history would, in a sense, have repeated itself, and that computer science would consist mainly of the indeed black art how to bootstrap from there to a sufficiently well-defined formal system.

The Future Formal Mathematics Database

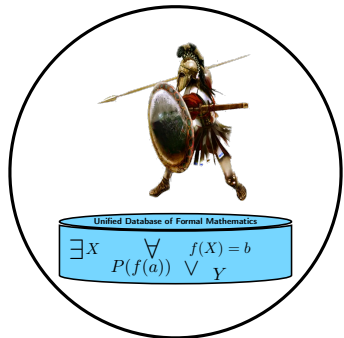
- ▶ Expressed in symbolic logic
 - ▶ Clear syntax
 - ▶ Clear semantics
- ▶ Unified (I can dream!)
 - ▶ One format (or compatible formats)
 - ▶ Globally consistent
- ▶ Structured
 - ▶ Different domains
 - ▶ Compositional
 - ▶ Searchable
- ▶ (Increasingly) comprehensive
 - ▶ *"The place to be"*
 - ▶ One size fits all (because it's flexible)



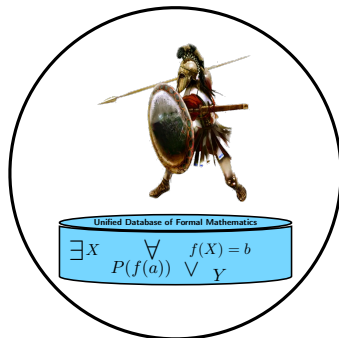
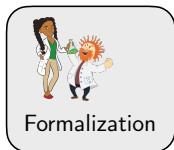
The Logic+ Architecture



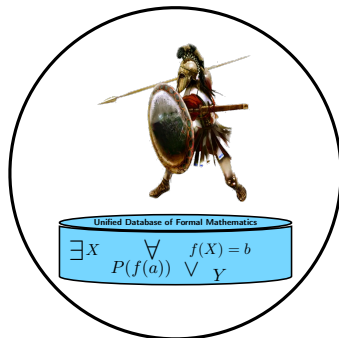
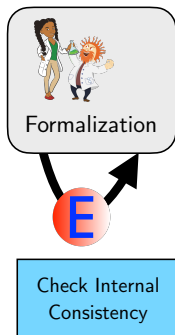
The Logic+ Architecture



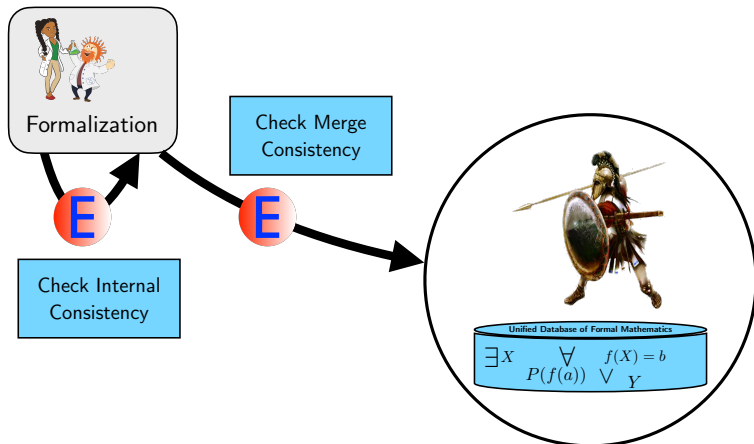
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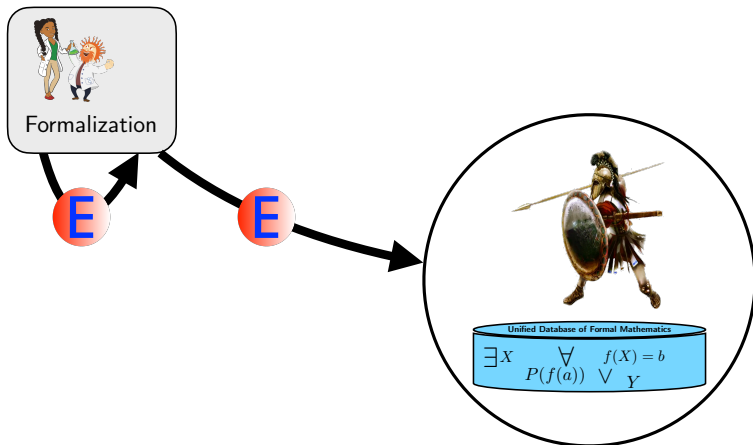
The Logic+ Architecture



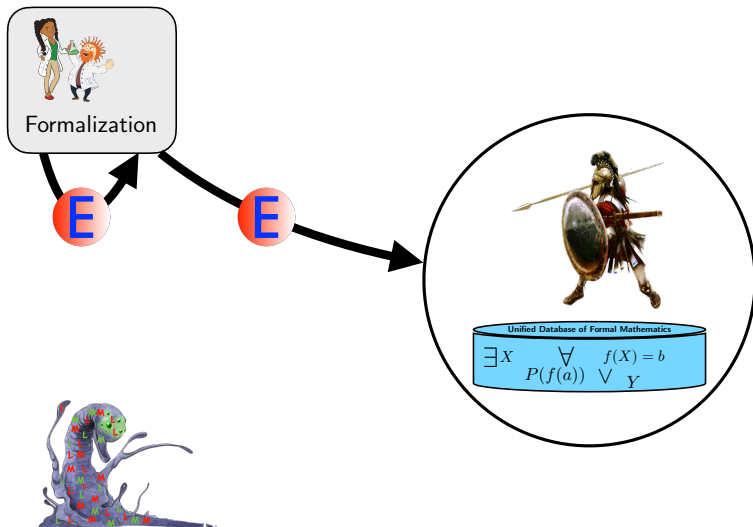
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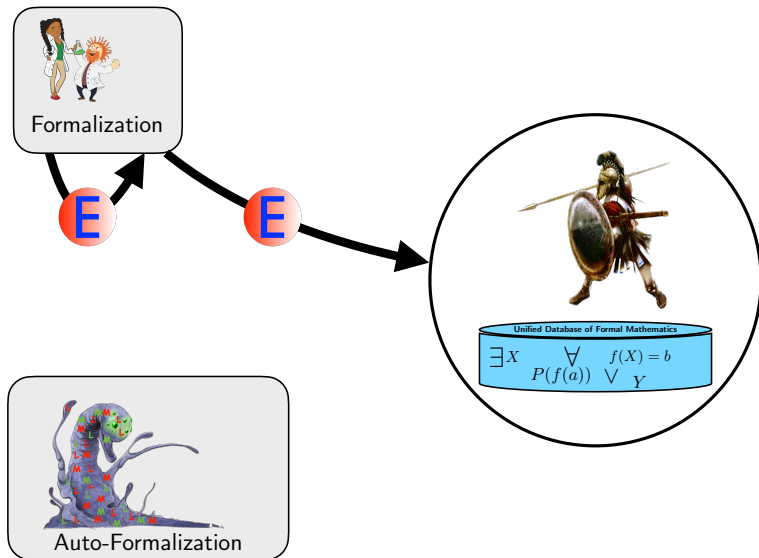
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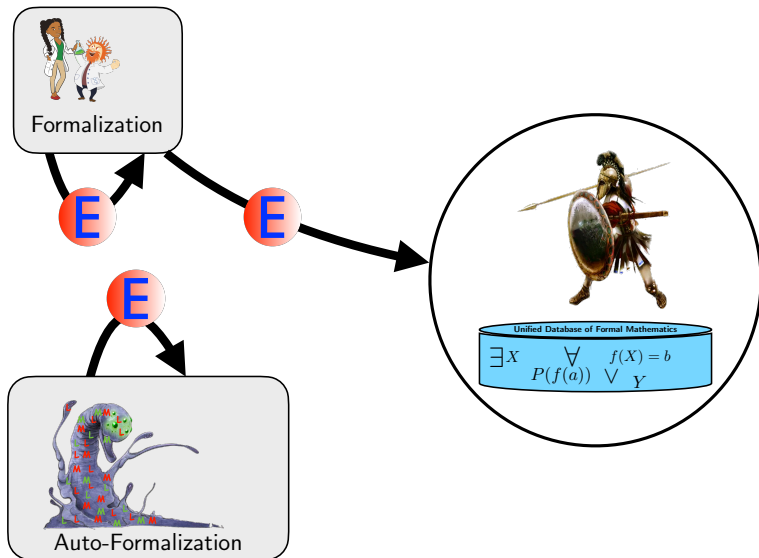
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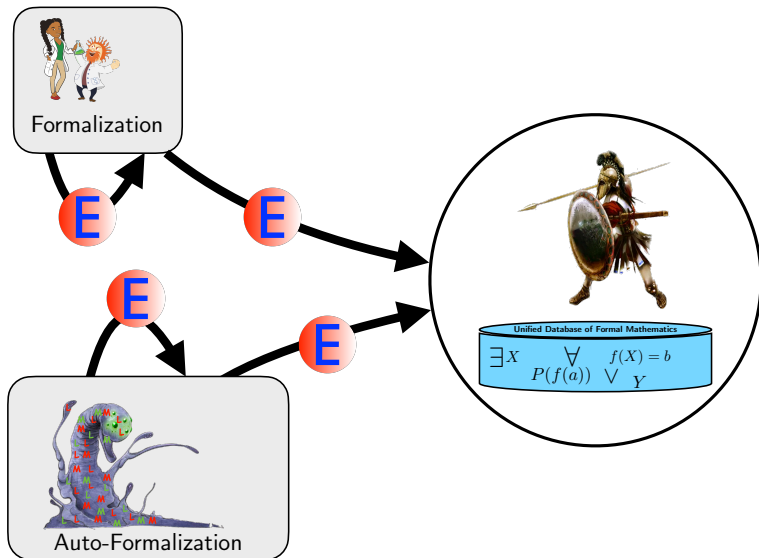
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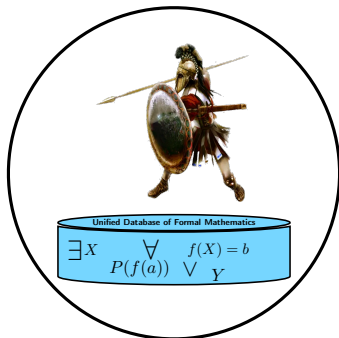
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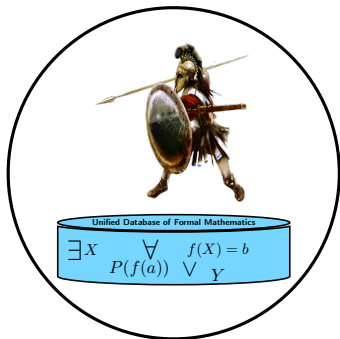
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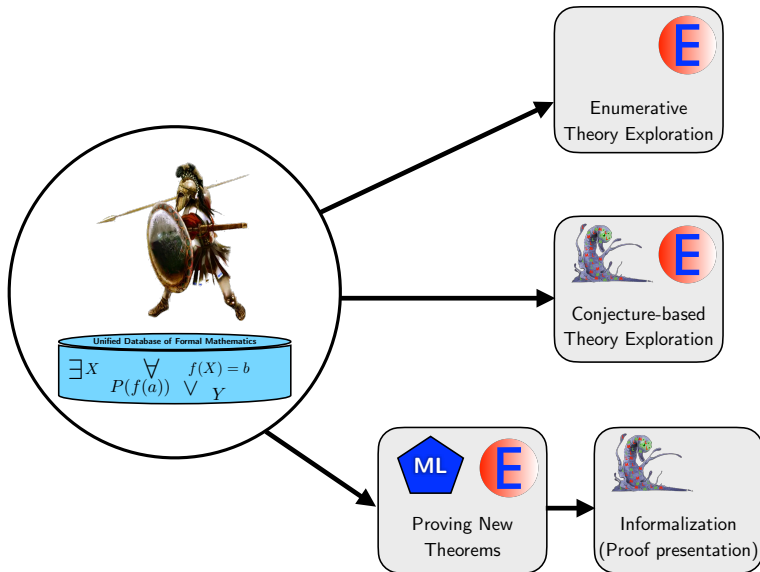
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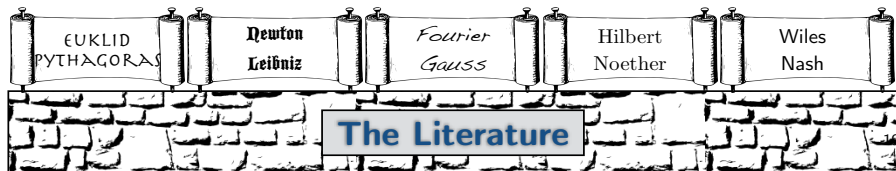
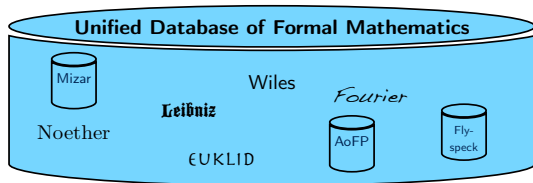
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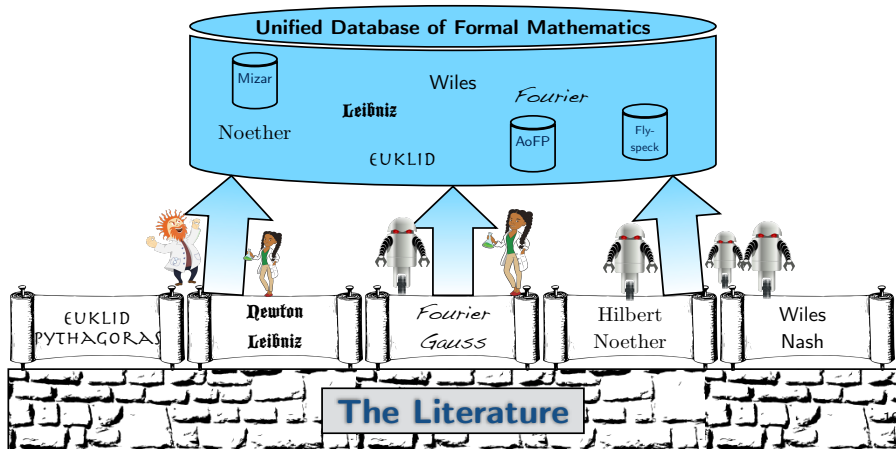
The future belongs to hybrid systems

- ▶ Interactive theorem provers are the main front-end for human mathematicians
 - ▶ User-friendly syntax and editors
 - ▶ Access to mathematical libraries
- ▶ Automatic theorem provers boost productivity
 - ▶ ... e.g. as Hammers for ITPs
 - ▶ ... supporting theory exploration
- ▶ Automatic theorem provers perform quality control
 - ▶ Internal consistency
 - ▶ Merge consistency
- ▶ Machine learning improves the power of automated systems
 - ▶ Proof guidance
- ▶ LLM-based systems translate to and from symbolic logic
 - ▶ Auto-formalization

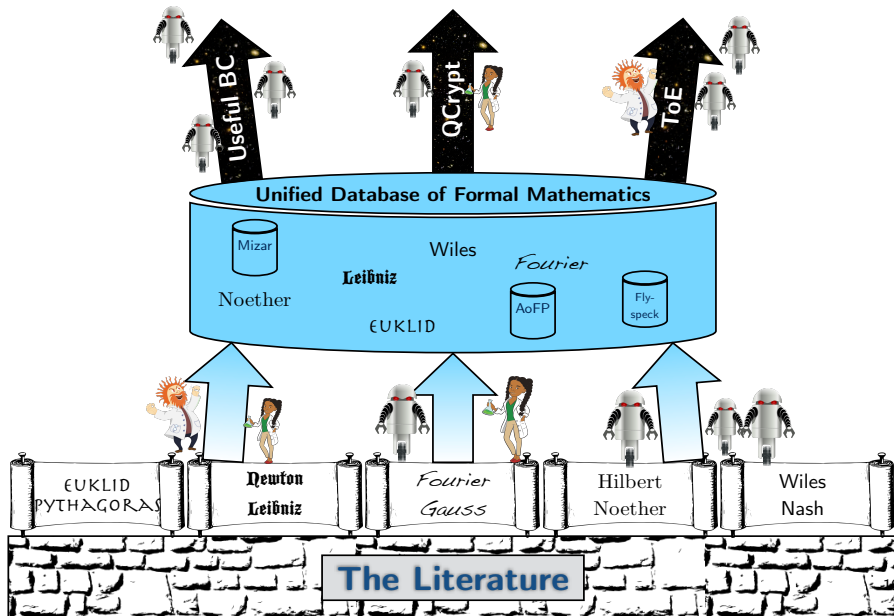
The Future Math Ecosystem (Utopia)



The Future Math Ecosystem (Utopia)

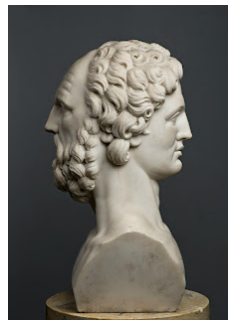


The Future Math Ecosystem (Utopia)



Conclusion

- ▶ Automated Reasoning and Machine Learning are complementary in many ways
 - ▶ The main formalism of computer maths will be symbolic
 - ▶ Deduction will use formal symbolic rules
- ▶ LLMs are not *the* way, but may contribute
 - ▶ ...especially interfacing humans and machine
 - ▶ ...especially translating different formalisms
- ▶ Other machine learning techniques may be more central
 - ▶ Nearest neighbors
 - ▶ Decision trees
 - ▶ Genetic algorithms
 - ▶ Tree-based ANN architectures



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Geoff Sutcliffe, Jack McKeown, and Alexander Steen.
A chat with bard.
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The END!