HUAJIAN XIN

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Homepage: https://xinhuajian.wordpress.com

https://scholar.google.com/citations?hl=en&user=E5M9x8wAAAAJ Google scholar:

EDUCATION

Ph. D. Student | Artificial Intelligence

School of Informatics, the University of Edinburgh

• Supervisor: Wenda Li

Bachelor of Philosophy | Logic

Department of Philosophy, Sun Yat-sen University

- GPA: 91% (ranking 2/32)
- Selected Coursework:

Logic: Mathematical Logic, Naive Set Theory, Proof Theory, Model Theory, Computation Theory, Modal Logic, Non-classical Logic, Formal Semantics, Informal Logic

Mathematics: Mathematical Analysis, Linear Algebra, Abstract Algebra (Lattice Theory and Universal Algebra), Probability theory

Computer Science: Data Structure and Algorithm, Artificial Intelligence (Logic Programming and Machine Learning) Philosophy: Metaphysics, Epistemology, Philosophy of Language, Philosophy of Mind

· Honor:

National Scholarship (2021, top prize for university students in mainland China, funded by the central government) Honors Graduate of Sun Yat-sen University (2023)

Undergraduate (Minor) | *Mathematics and Applied Mathematics*

School of Mathematics, Sun Yat-sen University

• Selected Coursework:

Real Variable Function, Complex Variable Function, Mathematical Statistics

RESEARCH AND WORKING EXPERIENCES

Intern | DeepSeek AI Supervisor: Chong Ruan and Daya Guo Research Assistant | Sun Yat-sen University Supervisor: Xiaodan Liang and Zhengying Liu

Jan. 2024 – Aug. 2024 Beijing, China Sept. 2022 - Jan. 2024 Shenzhen, China

Feb. 2021 – Jun. 2021

Guangzhou, China

Sep. 2024 – Apr. 2027 (expected)

Edinburgh, United Kingdom

Sep. 2019 - Jun. 2023

Guangzhou, China

SELECTED PUBLICATIONS AND PREPRINTS

DeepSeek-Prover-V1.5: Harnessing Proof Assistant Feedback for Reinforcement Learning and Monte-Carlo Tree Search

arXiv & GitHub & Huggingface & X.com

- Introduced informal chain-of-thought augmented whole-proof generation, reinforcement learning from proof assistant feedback (RLPAF), and the intrinsic-reward-driven Monte Carlo tree search algorithm (RMaxTS).
- Set a new state-of-the-art in theorem proving for Lean 4, achieving a 63.5% pass rate on the high school-level miniF2F benchmark and a 25.3% pass rate on the undergraduate-level ProofNet benchmark.

DeepSeek-Prover-V1: Advancing Theorem Proving in LLMs through Large-Scale Synthetic Data

arXiv

- Developed a data synthesis pipeline that automatically formalizes natural language math problems into 8 million formal statements with proofs in Lean 4.
- Set a new state-of-the-art in theorem proving for Lean 4, achieving a 50.0% pass rate on the miniF2F benchmark.

Proving Theorems Recursively

· Introduced a recursive, level-by-level theorem proving method, improving on traditional step-by-step approaches by focusing on high-level proof sketches and deferring intermediate conjectures to later stages.

LEGO-Prover: Neural Theorem Proving with Growing Libraries

ICLR 2024 Oral

- · Developed a modular theorem-proving approach with an integrated skill library, enabling large language models to propose, prove, and store lemmas for reuse in proving target theorems.
- Set a new state-of-the-art in theorem proving for Isabelle, achieving a 47.1% pass rate on the miniF2F benchmark.

MUSTARD: Mastering Uniform Synthesis of Theorem and Proof Data

• Developed a data synthesis framework that prompts the language model with specific keywords to generate and solve mathematical problems in natural language, followed by translating the problems and solutions into Lean 3 to verify their correctness.