

TuringQ: Benchmarking AI Comprehension in Theory of Computation

Pardis Sadat Zahraei^{✦✦} (✉ Paradise2001@gmail.com) Sharif University of Technology[✦]
Ehsaneddin Asgari[✦] (✉ easgari@hbku.edu.qa) Qatar Computing Research Institute - QCRI, Qatar[✦]



QCRI
معهد قطر لبحوث الحوسبة
Qatar Computing Research Institute
جامعة حمد بن خليفة
HAMAD BIN KHALIFA UNIVERSITY



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Paper

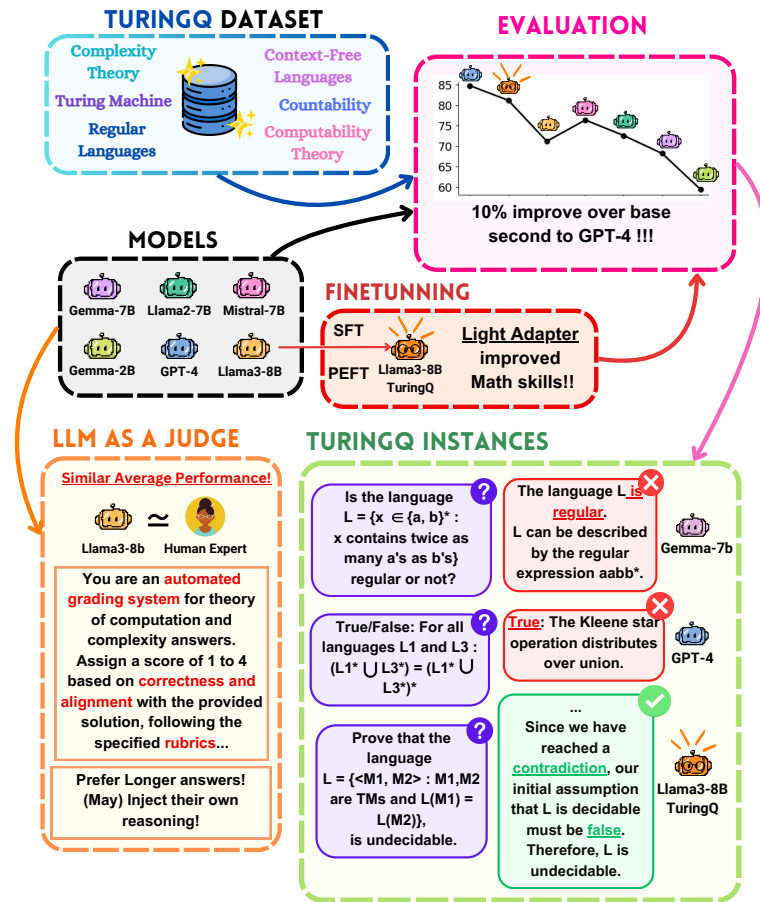


HuggingFace

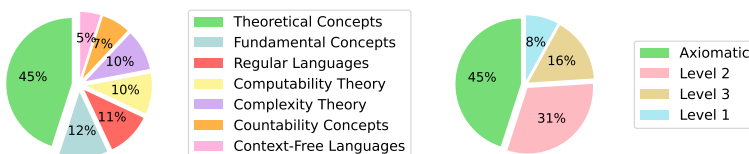


Github

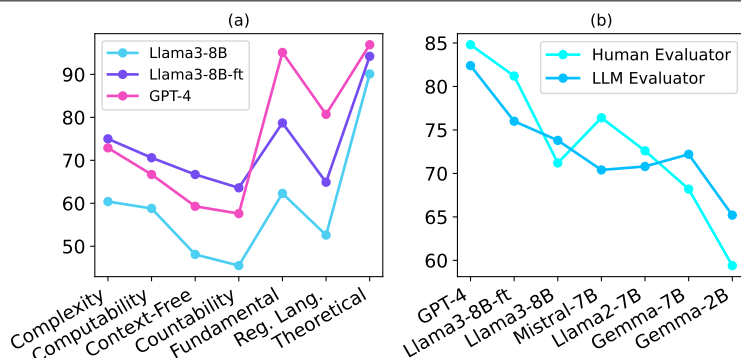
APPROACH



TuringQ Overview



Category and Difficulty Level Distribution in the TuringQ Dataset



a) LLM Performance Across TuringQ Categories, b) Performance on TuringQ: Evaluated by Human vs. LLM

SUMMARY

TuringQ is the **first benchmark** designed to evaluate the **reasoning capabilities** of Large Language Models (LLMs) in the field of **Computation Theory**.

- Sourced from **top university exams**, it comprises **4,006** question-answer pairs.
- Employs **Chain of Thought** prompting to assess the performance of **open-source LLMs** and **GPT-4**.

Automated Evaluation by LLMs

- Generates results **similar to human** judgment.
- Exhibits **surprising differences** compared to human evaluations.

Fine-Tuning Llama3-8B

- Finetuned using **SFT** and **PEFT**.
- **Substantial improvements** in reasoning accuracy and **out-of-domain tasks**.

Why Choose TuringQ?

- Evaluating LLM performance in **formal languages**.
- Advancing **complex computational reasoning** capabilities.

RESULTS

- Our fine-tuned model, **Llama3-8B-ft-TuringQ**, achieved a binary accuracy of **81.2%** on the TuringQ test set, indicating a **10%** improvement over the base model. Notably, it **enhanced performance across all categories**, approaching **GPT-4's** accuracy of **84.8%**.
- The LLM evaluator shows the **opposite trend** compared to human evaluators; **giving higher ratings to answers for harder questions**.
- LLM evaluators tend to **overrate weaker models** and **underrate stronger ones** compared to humans.
- LLMs prefer **longer, more complex answers, even if they are incorrect**, while humans evaluate more holistically.
- Evaluation on the **MATH dataset** showed a **0.6% accuracy increase** for the fine-tuned model.
- **Performance varied** across TuringQ categories, with the models performing best in **Theoretical Concepts** and worst in **Countability Concepts**.