

ReFoRM Reading Group

Rethinking Foundations for Real-world ML

Amin Saberi & Andrew Ilyas

Welcome to ReFoRM!

Not a slight to other ML!
Just starts with R 🥰

What this is: an experimental reading group on foundations of “real-world” ML

What does this mean?

Idealized picture of ML: something like $\theta^* = \arg \min_{\theta \in \Theta} \mathbb{E}_{z_i \sim D}[\ell(z_i; \theta)]$

ML powering systems like Claude, DALL-E, Google Photos:

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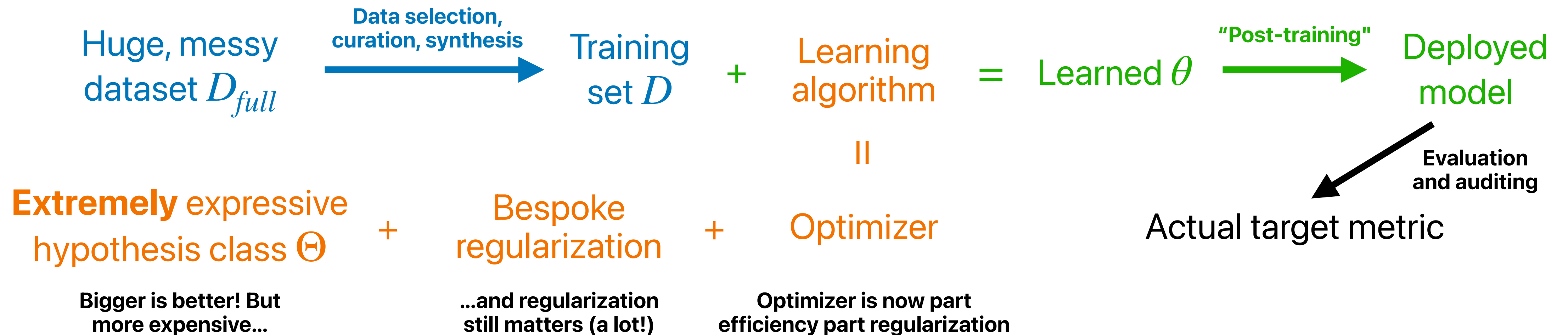
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What are the right questions to ask, and phenomena to explain—at what **level of abstraction** should we be aiming to explain them?

What theoretical models not only **explain** unexpected phenomena, but also **predict** new phenomena that we can verify experimentally?

Today's meeting

Logistics/plan for the quarter

Brief intro to this quarter's topic: reasoning

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Topics by weighted combination of {interest, coverage}:

Data selection, curation, and synthesis

Scaling laws & prediction

Post-training

Fine-tuning

LLM "Reasoning"

All past presentations are online: <https://andrewilyas.github.io/REFORM-reading-group/>

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Everyone else: Read the paper/watch a podcast/something! Try to come with some familiarity

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3. Main idea:
 1. We'll keep a **running document** throughout the quarter where people can dump open problems that they'd be open to thinking about, and other people can express interest
 2. Some meetings will be open problem sessions where people can present on problems they like from the doc

Introduction to reasoning

If a train is moving at 60 mph and travels for 3 hours, how far does it go?

The train travels 180 miles.

Plain response

To determine the distance traveled, use the formula:

Distance = Speed \times Time

Given that the speed is 60 mph and the time is 3 hours:

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So, the train travels 180 miles.

Response with intermediate reasoning steps

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Aside: There's lots of philosophical debate about what it means to "reason" (especially how it relates to human reasoning), but we'll mostly ignore this

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Evaluating reasoning

Mathematical reasoning: Benchmarks (GSM8k, MATH) use math word problems

Commonsense and logical reasoning: Benchmarks (Commonsense-QA, Strategy-QA, BigBench-Hard) evaluate everyday, multi-hop reasoning, and complex puzzles

Other reasoning tasks: Symbolic/algorithmic (e.g., list sorting, rubik's cube) and Code-based (e.g., programming problems)

Goal: High accuracy on benchmarks + faithful “reasoning traces”

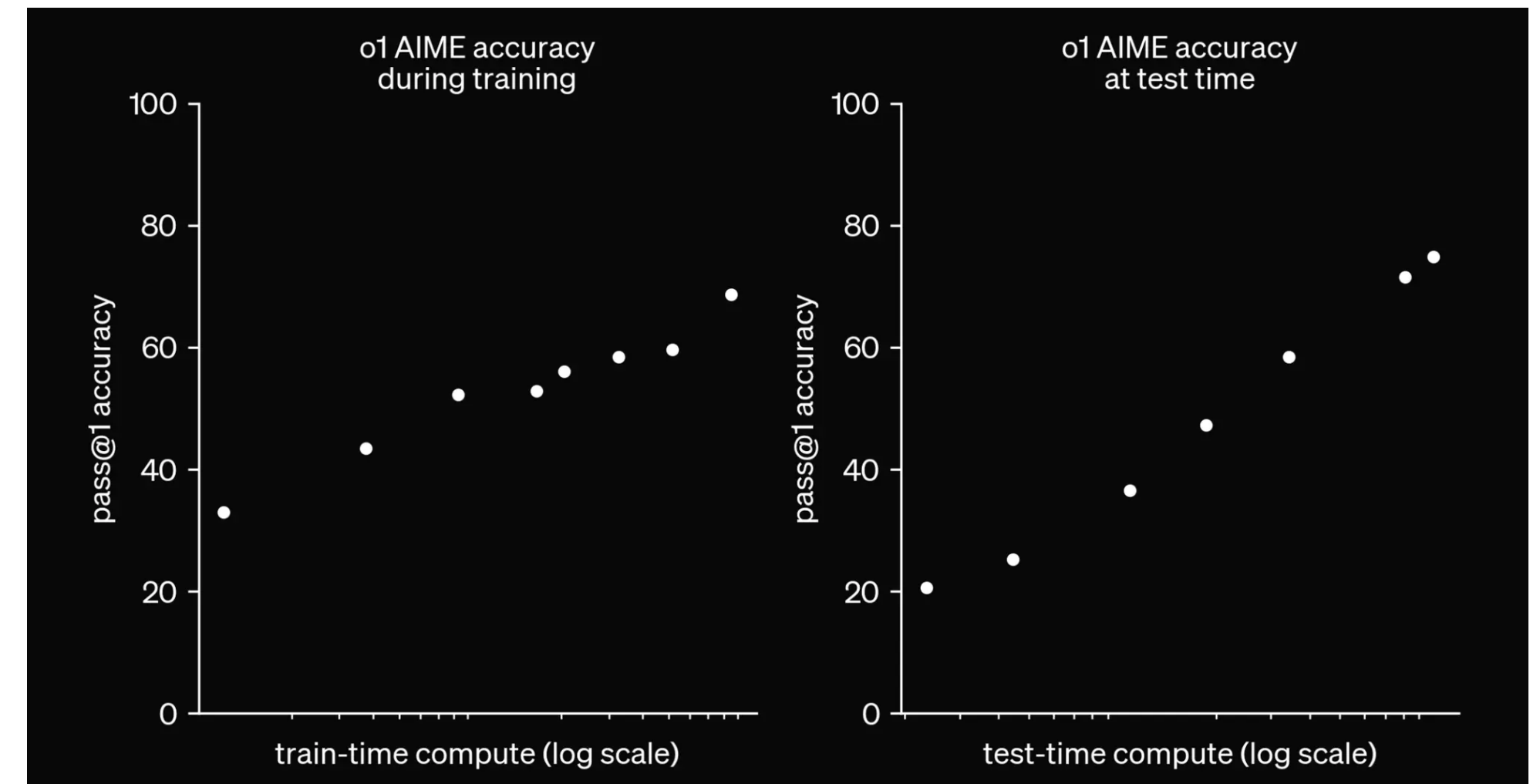
Basic idea: Generate a lot of stuff

Unifying theme of a lot of reasoning stuff is that the model “thinks” for a while before outputting an answer (i.e., test-time computation)

As a result, key questions are usually:

How do we encourage models to generate a lot of stuff?

How do we make sure that “generating a lot of stuff” actually helps reach the right answer?



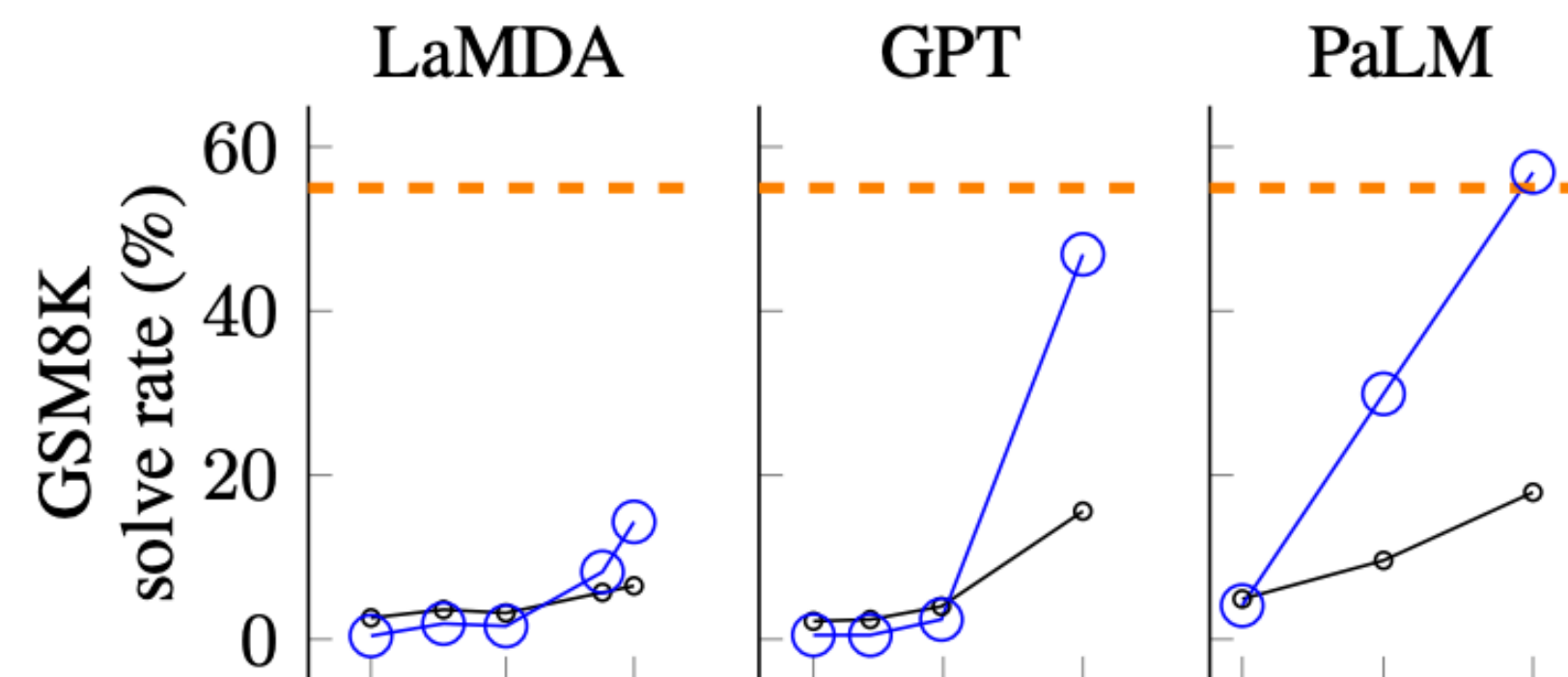
Simplest/Earliest forms of reasoning

Scratchpad/Chain of thought (Nye et al., 2021; Wei et al., 2022; Kojima et al. 2022)

Change nothing about the way the model is trained

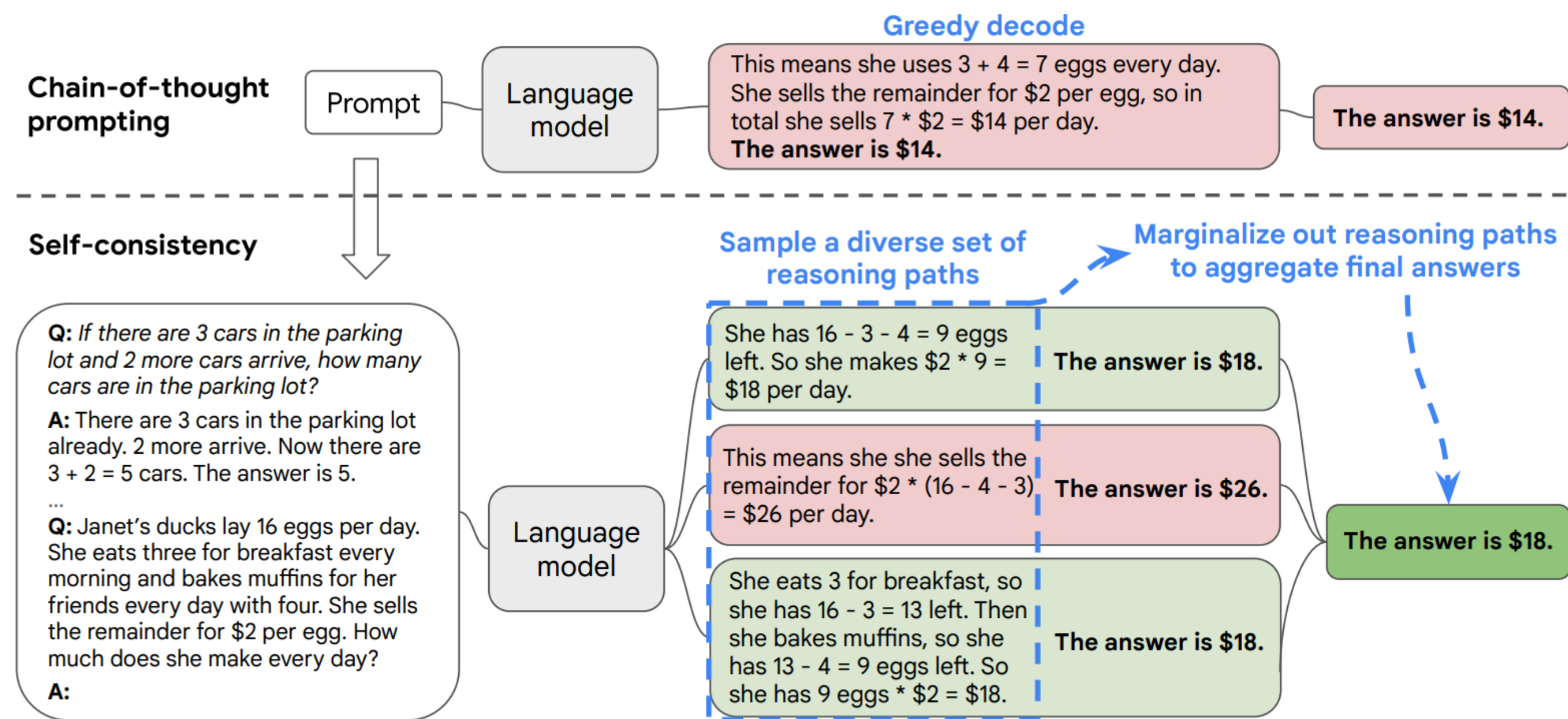
At test time, just ask the model to “show its work,” or “think step by step, or give it “scratch space” to perform big computations

Improved benchmark performance a lot but only worked for really big models



Refinement: Self-consistency

Idea: instead of asking the model to generate a lot of text once, ask it to generate a lot of text many times and take a majority vote (Wang et al., 2023)



PaLM-540B	Zero-shot CoT (Kojima et al., 2022) + Self-consistency (40 paths)	43.0 69.2
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More refinements

More complex search at test time

e.g., Tree of thought (Yao et al., 2023)

Encouraging models to reason at training time

Incorporating verifiers

e.g., STaR (Zelikman et al. 2022), ReST (Gulcehre et al., 2023), R1 (DeepSeek, 2025)

Rewarding thinking directly

e.g., Process Reward (Uesato et al., 2022; Lightman et al., 2023)

Teaching models to search

e.g., incorporating MCTS during preference learning (Xie et al., 2024)

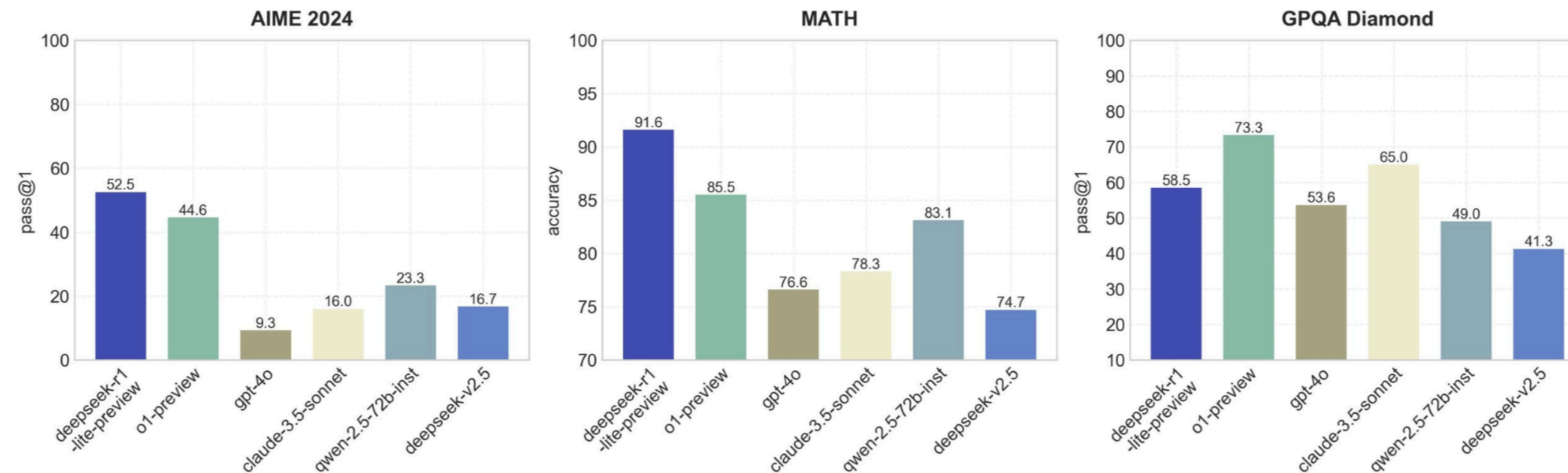
An unexpected success with verifiers

Reminder from Joyee and Wanqiao last quarter: **DeepSeek-R1**

Main idea: Direct RL on correctness

Model automatically learns to output a lot of stuff (even though not explicitly told to!)

Drastically improves performance on reasoning benchmarks



Thank you (and please sign up!)

Sign-up sheet: <https://tinyurl.com/reform-S25>

Mailing list: reform-ml-list@stanford.edu

Contact: andrewi@stanford.edu, saberi@stanford.edu



Presenter signup



Ideas doc & Reading list