

Using not so large language models for mathematical reasoning

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- The idea: to write proofs of theorems on a computer so that it can mechanically check them.
- Current status
 - Formalized $\sim 100k$ theorems (most of them are **very** simple), covering almost all undergraduate math.
 - A few advanced results e.g. four color theorem, Kepler conjecture, Feit-Thompson theorem, perfectoid spaces.
 - Very active community.
- How it looks like:

```
theorem prime_of_coprime (n : ℕ) (h1 : 1 < n) (h : ∀ m < n, m ≠ 0 → n.coprime m) : prime n :=
begin
  refine prime_def_lt.mpr ⟨h1, λ m mlt mdvd, _⟩,
  have hm : m ≠ 0,
  { rintro rfl,
    rw zero_dvd_iff at mdvd,
    exact mlt.ne' mdvd },
  exact (h m mlt hm).symm.eq_one_of_dvd mdvd,
end
```

Hammers

But these proofs are too low-level. So we can use *hammers* to fill the low-level gaps. It heuristically finds useful premises, stacks them with the target statement and tries to use some SMT-solver.

```
lemma "sqrt 2  $\notin$   $\mathbb{Q}$ "
proof
  assume "sqrt 2  $\in$   $\mathbb{Q}$ "
  then obtain a b::int where "sqrt 2 = a/b"
    "coprime a b" "b  $\neq$  0" sledgehammer
  then have c: "2 = a^2 / b^2"
    sledgehammer
  then have "b^2  $\neq$  0" sledgehammer
  then have *: "2*b^2 = a^2"
    sledgehammer
  then have "even a"
    sledgehammer
  then obtain c::int where "a=2*c"
    sledgehammer
  with * have "b^2 = 2*c^2"
    sledgehammer
  then have "even b"
    sledgehammer
  with (coprime a b) (even a) (even b)
    show False sledgehammer
qed
```

(a) The proof sketch produced by the human user. The `sledgehammer` command indicates that the human invokes the Sledgehammer method at that point.

```
lemma "sqrt 2  $\notin$   $\mathbb{Q}$ "
proof
  assume "sqrt 2  $\in$   $\mathbb{Q}$ "
  then obtain a b::int where "sqrt 2 = a/b" "coprime a b" "b  $\neq$  0"
    by (metis Rats_cases' less_irrefl)
  then have c: "2 = a^2 / b^2"
    by (smt (z3) of_int_power power_divide real_sqrt_pow2)
  then have "b^2  $\neq$  0" by fastforce
  then have *: "2*b^2 = a^2"
    by (smt (verit, ccfv_SIG) c comm_semiring_class.distrib
      eq_divide_eq_numeral(1) mult_cancel_right1 numeral_Bit0
      numeral_plus_numeral of_int_add of_int_power
      of_int_power_eq_of_int_cancel_iff one_plus_numeral)
  then have "even a"
    by (smt (z3) even_power oddE)
  then obtain c::int where "a=2*c" by blast
  with * have "b^2 = 2*c^2" by auto
  then have "even b"
    by (smt (z3) even_power oddE)
  with (coprime a b) (even a) (even b) show False by fastforce
qed
```

(b) The proof accepted by Isabelle. The steps containing `assume`, `obtain`, `have`, `show` are from the original human proof sketch. The steps containing `metis`, `smt`, `fastforce`, `blast`, `auto`, `fastforce` are completed by Sledgehammer.

Large Language Models

A large language model, or LLM, is a deep learning algorithm that can recognize, summarize, translate, predict and generate text and other content based on knowledge gained from massive text corpora.

Examples: GPT-4, Codex, PaLM.

One interesting property of LLMs is their ability for *few-shot learning*:

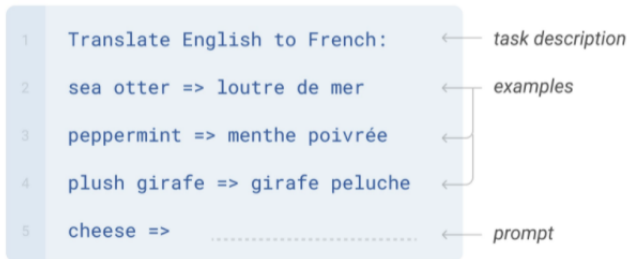


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We consider three tasks in which LLMs show SOTA results:

- **Informal to informal task** i.e. solving problems in natural language: Minerva (Google, June 2022) solves 50% of the problems from MATH dataset (high school competitions).
- **Informal to formal task** i.e. autoformalization: Codex via few-shot learning (Google, May 2022) can autoformalize 25% of the problems from MATH.
- **Formal to formal task** i.e. formal mathematical reasoning: GPT-f (OpenAI, September 2020) solves 30% of the problems from miniF2F dataset (formalized problems from high school competitions).

Example of Minerva's output

Question: A line parallel to $y = 4x + 6$ passes through $(5, 10)$. What is the y -coordinate of the point where this line crosses the y -axis?

Model output: The line is parallel to $y = 4x + 6$, which means that it has the same slope as $y = 4x + 6$. Since the line passes through $(5, 10)$, we can use the point-slope form of a line to find the equation of the line:

$$y - 10 = 4(x - 5)$$

$$y - 10 = 4x - 20$$

$$y = 4x - 10$$

Now that we have the equation of the line, we can find the y -coordinate of the point where the line crosses the y -axis by substituting $x = 0$ into the equation:

$$y = 4 \cdot 0 - 10 = \boxed{-10}.$$

DSP approach

DSP approach: to combine all these tasks.

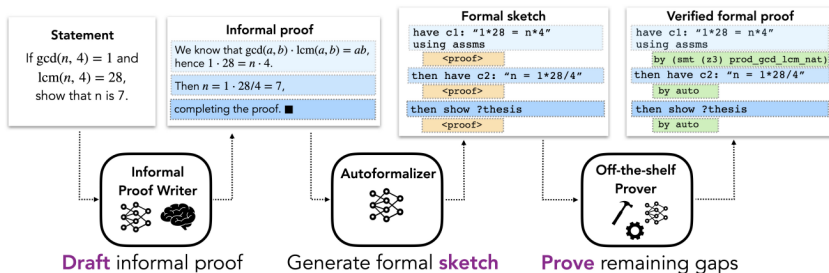


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Problem solving using small models

- Carried out the search for the open-source small model for problem solving. The best one: Galactica (Facebook 2022) – trained on scientific texts.
- Took the 6B-model and evaluated it on PreAlgebra subset of MATH benchmark. Results: 4% accuracy in pass@1, and 30% in pass@100. For comparison: Minerva-8B achieves 24% in pass@1, and 41% in maj@256.
- Manually checked the output of the model. It is able to understand context, do reasoning and computations (examples on the following slides).

Examples of Galactica's output

Problem: Let S be the set of points z in the complex plane such that $|z - 7 + 10i| \leq 5$. Find the area of S .

Model output: Let $z = x + yi$ be a point in S . Then $|z - 7 + 10i| \leq 5$ if and only if $|x + 7|^2 + |y - 10|^2 \leq 25$.

The area of S is $\iint_S 1 dA$.

Since S is a disk of radius 5 centered at $7 - 10i$, the area of S is $\pi(5)^2 = \boxed{25\pi}$. The final answer is 25π .

Examples of Galactica's output

Problem: A sequence of positive real numbers $\{a_1, a_2, a_3, \dots\}$ has the property that for $i \geq 2$, each a_i is equal to the sum of all the previous terms. If $a_{19} = 99$, then what is a_{20} ?

Model output: Let $a_i = a_{i-1} + a_{i-2} + \dots + a_1$ for $i \geq 2$. Then

$$\begin{aligned} a_{20} &= a_{19} + a_{18} + \dots + a_2 + a_1 = 99 + (a_{18} + a_{17} + \dots + a_2 + a_1) = \\ &= 99 + a_{19} = 99 + 99 = 198. \end{aligned}$$

The final answer is 198.

Examples of Galactica's output

Problem: For what positive value of t is $|-4 + ti| = 6$?

Model output:

$$|-4 + ti| = 6$$

$$|-4 + ti|^2 = 36$$

$$(-4 + ti)(-4 - ti) = 36$$

$$16 + t^2 = 36$$

$$t^2 = 20$$

$$t = \pm\sqrt{20} = \pm\sqrt{4 \cdot 5} = \pm 2\sqrt{5}.$$

So $t = 2\sqrt{5}$ or $t = -2\sqrt{5}$. The final answer is $2\sqrt{5}$.

Examples of Galactica's output

Problem: The graph of $y = f(x)$ passes through the point $(-4, -7)$. If $f(x)$ is an even function, then what other point must the graph pass through? Enter your answer as an ordered pair.

Model output: $f(x)$ is an even function if $f(-x) = f(x)$ for all real numbers x . Then, since $f(-4) = -7$, we have $f(4) = -7$, so the graph of $f(x)$ passes through the point $(4, -7)$. The final answer is $(4, -7)$.

Vulnerability of the MATH benchmark

Many problems in MATH made so that it is easy to get false-positive result, i.e. get the correct answer with completely incorrect reasoning. For example:

- **Problem:** Find the remainder when $x^9 - x^6 + x^3 - 1$ is divided by $x^2 + x + 1$.

Model output: By the Division Theorem, the remainder is $x^2 + x + 1$ times a polynomial of degree 8 or less. The remainder is 0 when $x = 0$ and $x = -1$, so the remainder is divisible by $x^2 + x + 1$. Hence, the remainder is 0. The final answer is 0.

- **Problem:** How many vertical asymptotes does the graph of $y = \frac{1}{x^3 - 3x^2 + 3x - 1}$ have?

Model output: $\frac{1}{x^3 - 3x^2 + 3x - 1}$ has a vertical asymptote at $x = 1$ because $x^3 - 3x^2 + 3x - 1$ has a zero at $x = 1$. The graph of $y = \frac{1}{x^3 - 3x^2 + 3x - 1}$ has a vertical asymptote at $x = 1$. The final answer is 1.

Autoformalization using small models

- For the autoformalization task we chose Codegen, one of the best open-source models for code generation. Our model contained 6B parameters. In all experiments we got (almost) zero result.
- At first we used the same prompt that was used for Codex in [1], contained 3 examples of formalizations.

Problem: The sequence $S_1, S_2, S_3, \dots, S_{10}$ has the property that every term beginning with the third is the sum of the previous two. That is, $S_n = S_{n-2} + S_{n-1}$ for $n \geq 3$. Suppose that $S_9 = 110$ and $S_7 = 42$. What is S_4 ?

Solution: $S_9 = 110, S_7 = 42$.

$$S_8 = S_9 - S_7 = 110 - 42 = 68, S_6 = S_8 - S_7 = 68 - 42 = 26,$$

$$S_5 = S_7 - S_6 = 42 - 26 = 16, S_4 = S_6 - S_5 = 26 - 16 = 10$$

Therefore, the answer is 10.

Formalization:

```
theorem
  fixes s :: "nat \ $\rightarrow$  real"
  assumes h0 : "\ $\wedge$ n. s (n+2) = s (n+1) + s n"
    and h1 : "s 9 = 110"
    and h2 : "s 7 = 42"
  shows "s 4 = 10"
proof -
  (* $S_9 = 110$, $S_7 = 42$

  $S_8 = S_9 - S_7 = 110 - 42 = 68$ *)
  have "s 8 = 68" using h1 h2 h0[of 7] sledgehammer
  (* $S_6 = S_8 - S_7 = 68 - 42 = 26$ *)
  hence h3: "s 6 = 26" using h2 h0[of 6] sledgehammer
  (* $S_5 = S_7 - S_6 = 42 - 26 = 16$ *)
  hence "s 5 = 16" using h2 h0[of 5] sledgehammer
  (* $S_4 = S_6 - S_5 = 26 - 16 = 10$ *)
  then show ?thesis using h3 h0[of 4] sledgehammer
```

qed

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- At first we used the same prompt that was used for Codex in [1], contained 3 examples of formalizations.
- Then we wrote our own formalizations of **very** simple problems, and use them as a prompt.

Problem: Solve $\log_4 x + \log_2 x^2 = 10$. Show that the answer is 16.

Solution: Applying the logarithmic identities $\log_a b^c = c \log_a b$ and $\log_{a^c} b = (1/c) \log_a b$, we find

$$\begin{aligned} 10 &= \log_4 x + \log_2 x^2 = \log_4 x + 2 \log_2 x = \log_{2^2} x + 2 \log_2 x \\ &= \frac{1}{2} \log_2 x + 2 \log_2 x = \frac{5}{2} \log_2 x. \end{aligned}$$

Therefore, $\log_2 x = 4$, which implies $x = 2^4 = \boxed{16}$.

Formalization on the next slide (it is so thorough that it require all slide to display).

Formalization:

```
theorem
  fixes x :: real
  assumes h0 : "log 4 x + log 2 (x^2) = 10"
    and h1 : "x > 0"
  shows "x = 16"
proof -
  (* 10 &= \log_4 x + \log_2 x^2 *)
  have h2 : "10 = log 4 x + log 2 (x^2)"
    sledgehammer
  (* &= \log_4 x + 2 \log_2 x *)
  have h3 : "10 = log 4 x + 2 * log 2 x"
    sledgehammer
  (* &= \log_{2^2} x + 2 \log_2 x *)
  have h4 : "10 = log (2^2) x + 2 * log 2 x"
    sledgehammer
  (* &= \frac{1}{2} \log_2 x + 2 \log_2 x *)
  have h5 : "10 = (1/2) * log 2 x + 2 * log 2 x"
    sledgehammer
  (* &= \frac{5}{2} \log_2 x *)
  have h6 : "10 = (5 / 2) * log 2 x"
    sledgehammer
  (* Therefore, $\log_2 x = 4$ *)
  have h7 : "4 = log 2 x"
    sledgehammer
  (* which implies $x = 2^4 = \boxed{16}$ *)
  have h8 : "x = 16"
    sledgehammer
  show ?thesis using h8 by auto
qed
```

Autoformalization using small models

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- At first we used the same prompt that was used for Codex in [1], contained 3 examples of formalizations.
- Then we wrote our own formalizations of **very** simple problems, and use them as a prompt.
- After that model became able to solve problems that very similar to examples.

Problem: If $\log_2 x + \log_2 x^2 = 6$, find the value of x .

Solution: Note that $\log_2 x^2 = 2 \log_2 x$. Therefore, we have $\log_2 x + 2 \log_2 x = 6$, or $3 \log_2 x = 6$. Thus $\log_2 x = 2$, so $x = 2^2 = \boxed{4}$.

Formalization on the next slide again.

Formalization:

```
theorem
  fixes x :: real
  assumes h0 : "log 2 x + log 2 (x^2) = 6"
    and h1 : "x > 0"
  shows "x = 4"
proof -
  (* 6 &= \log_2 x + log_2 (x^2) *)
  have h2 : "6 = log 2 x + log 2 (x^2)"
    sledgehammer
  (* &= \log_2 x + 2 \log_2 x *)
  have h3 : "6 = log 2 x + 2 * log 2 x"
    sledgehammer
  (* &= 3 \log_2 x *)
  have h4 : "6 = 3 * log 2 x"
    sledgehammer
  (* Thus $\log_2 x = 2$ *)
  have h5 : "log 2 x = 2"
    sledgehammer
  (* so $x = 2^2 = \boxed{4}$ *)
  have h6 : "x = 4"
    sledgehammer
  show ?thesis using h8 by auto
qed
```


Autoformalization using small models

- For the autoformalization task we chose Codegen, one of the best open-source models for code generation. Our model contained 6B parameters. In all experiments we got (almost) zero result.
- At first we used the same prompt that was used for Codex in [1], contained 3 examples of formalizations.
- Then we wrote our own formalizations of **very** simple problems, and use them as a prompt.
- After that model became able to solve problems that very similar to examples.
- Thinking that model is so bad because it does not know the Isabelle language (by the way, Codex too), we fine-tuned Codegen on Archive of Formal Proofs – the biggest collection of Isabelle proofs. It contains about 100k abstract theorems. This did not give any result.

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Further work

- Implement detecting wrong step in natural proof based on the proof-model output and partial proof regeneration.
- Having good autoformalizer train the draft-model based on the success of obtained formal proof.
- Using LLM manually gather parallel corpus between proofs in natural language and in formal language.

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References

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- 3 Magnushammer: A Transformer-based Approach to Premise Selection (Mikuła et al.).
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Questions

Your questions, please!