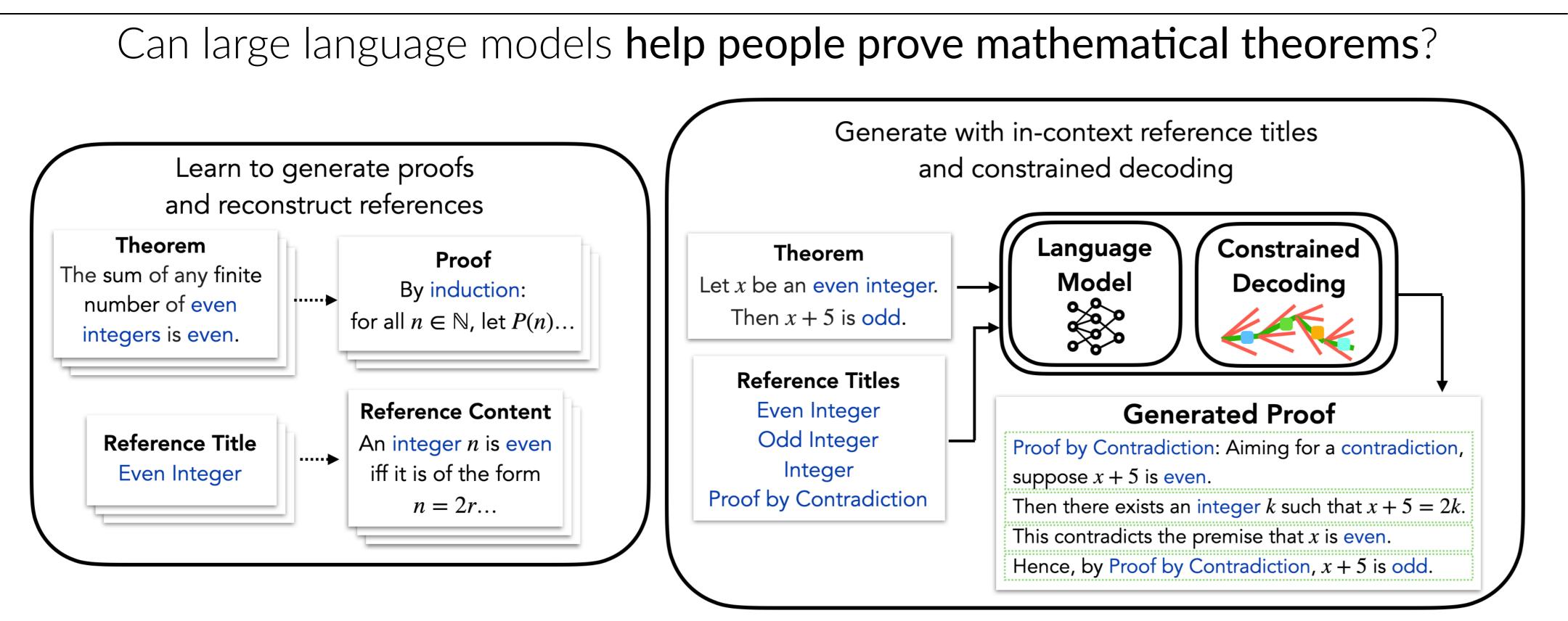
NaturalProver: Grounded Mathematical Proof Generation with Language Models

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NaturalProver



We present **NaturalProver**, a language model that generates mathematical proofs by conditioning on background references (e.g. theorems and definitions that are either retrieved or human-provided), and optionally enforces their presence with constrained decoding.

Grounding = references + constrained decoding

NaturalProver is an instance of GPT-3 fine-tuned on NaturalProofs [Welleck et al., Neurips 2021]. NaturalProver adds two components on top of GPT-3:

- In-context references: retrieved or provided theorems/definitions relevant to a correct proof.

Natural vs. formal theorem proving

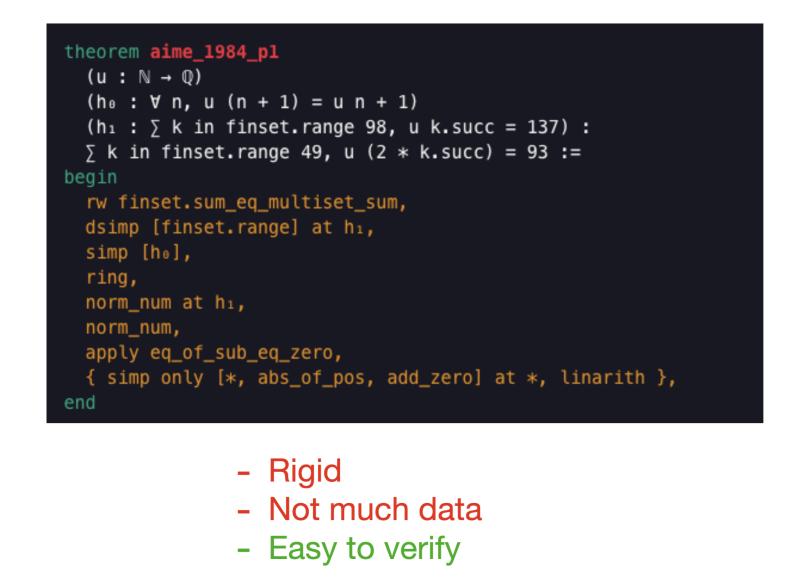


Figure 1. Classical provers use **rigid formal languages**.

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• **Constrained decoding**: samples multiple next-steps, retains steps in a beam based on constraints.

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Theorem:
Let x be an even integer. Then x + 5 is odd.
Proof by Contradiction: Aiming for a contradiction,
suppose x + 5 is even.
Then there exists an integer k such that x + 5 = 2k.
This contradicts the premise that x is even.
Hence, by Proof by Contradiction, x + 5 is odd.
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- Flexible
- Used in education, science, engineering
- Lots of language data
- Hard to verify!

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Capable of correct, useful proof generation

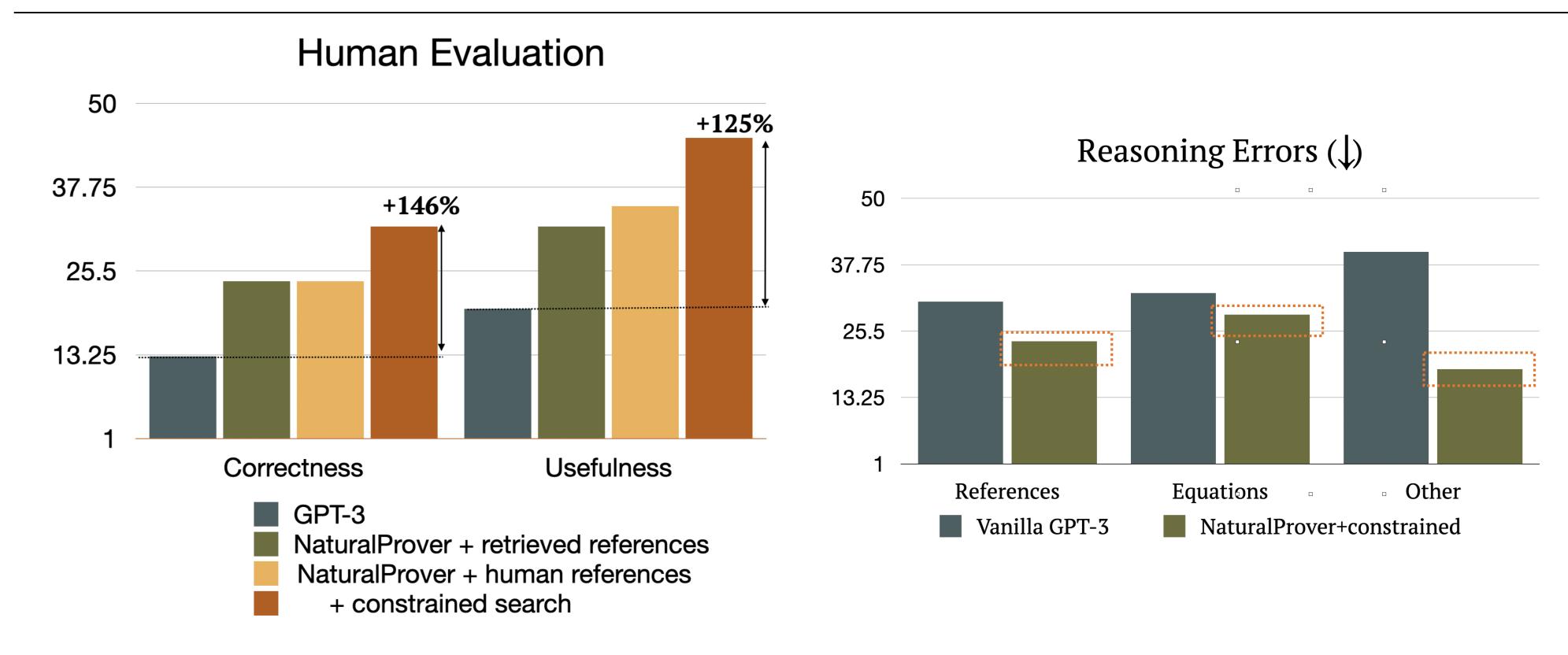


Figure 2. On theorems from the NaturalProofs benchmark, NaturalProver improves the quality of next-step suggestions and generated proofs over fine-tuned GPT-3, according to human evaluations from university-level mathematics students. NaturalProver is capable of proving some theorems that require short (2-6 step) proofs, and providing next-step suggestions that are rated as correct and useful over 40% of the time.

Human-machine collaboration

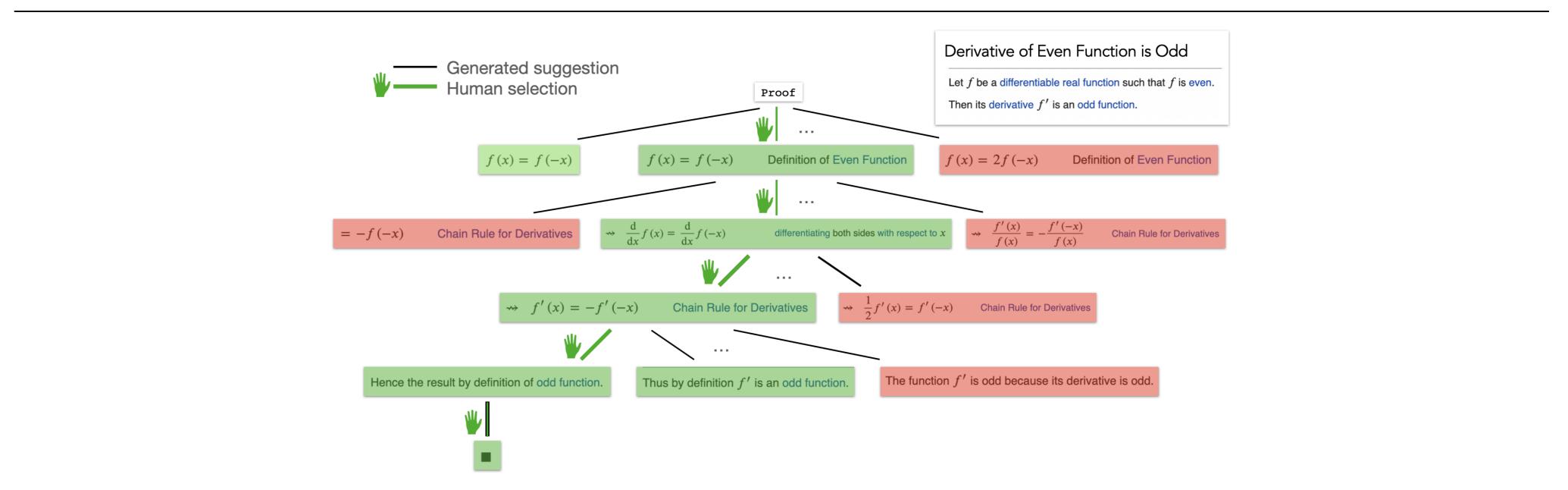
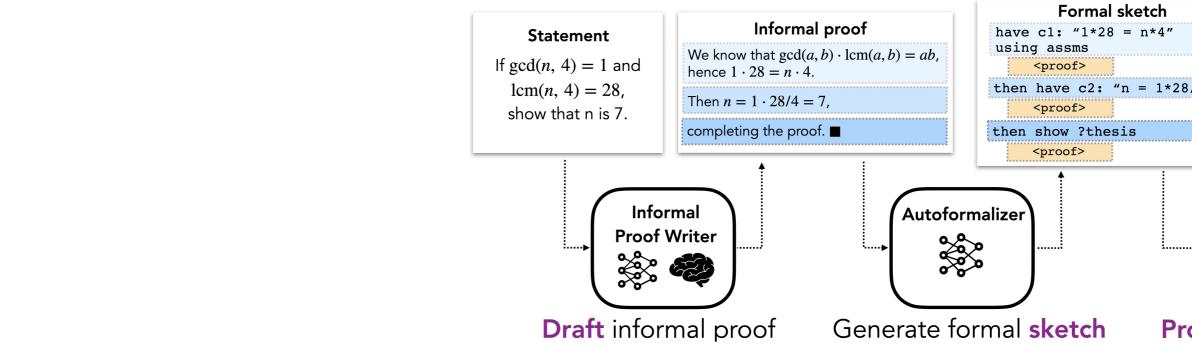


Figure 3. NaturalProver had > 40% correct and useful next-step predictions. These compound in full-proof generation. An exciting option is *human-machine collaboration* with *multiple* suggestions.

Towards verified natural proofs : come see us at the *MathAl workshop*!





	Verified formal proof
	have c1: "1*28 = n*4"
	using assms
	<pre>by (smt (z3) prod_gcd_lcm_nat)</pre>
/4″	then have c2: " $n = 1*28/4$ "
	by auto
	then show ?thesis
	by auto
	-the-shelf Prover
ove re	maining gaps