Equality Saturation Guided by Large Language Models

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LLM for Program Optimization



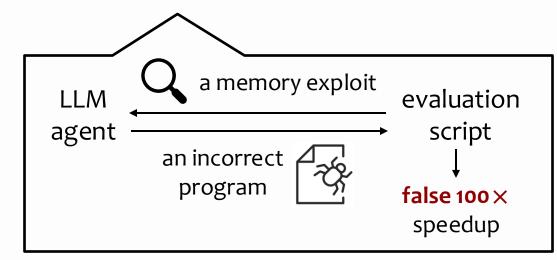




52% speedup [Lange et al. 2025]

fix 60% performance bugs [Garg et al. 2025]

2.65 × speedup [Chen et al. 2024]



Pros: rich knowledge, scalable

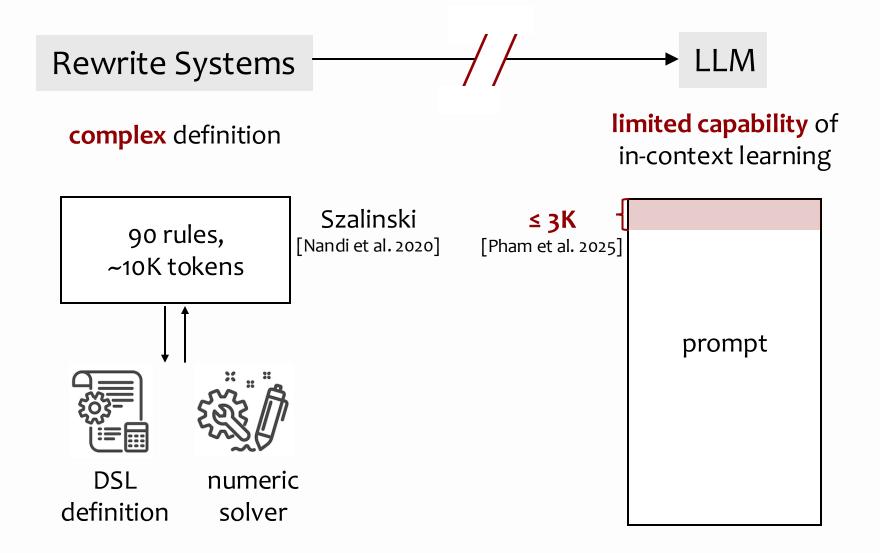
Cons: no soundness guarantee

Program Rewrite Systems

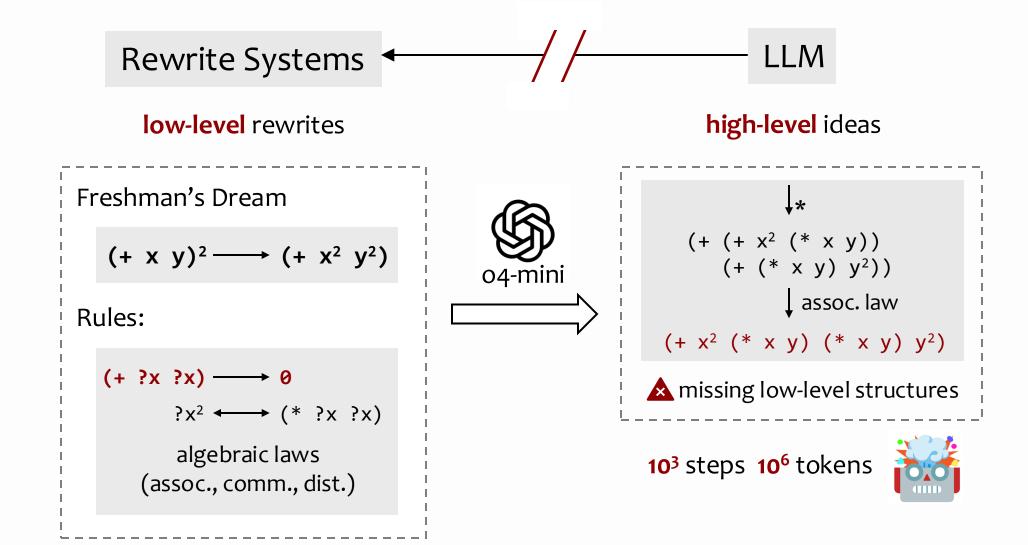
Pros: ensured soundness

Cons: limited scalability

Gaps Between Rewrite Systems and LLMs

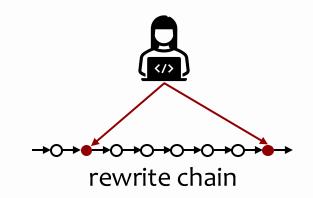


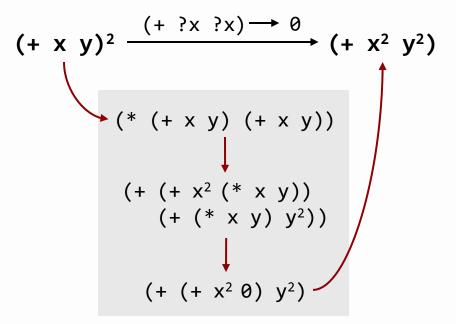
Gaps Between Rewrite Systems and LLMs



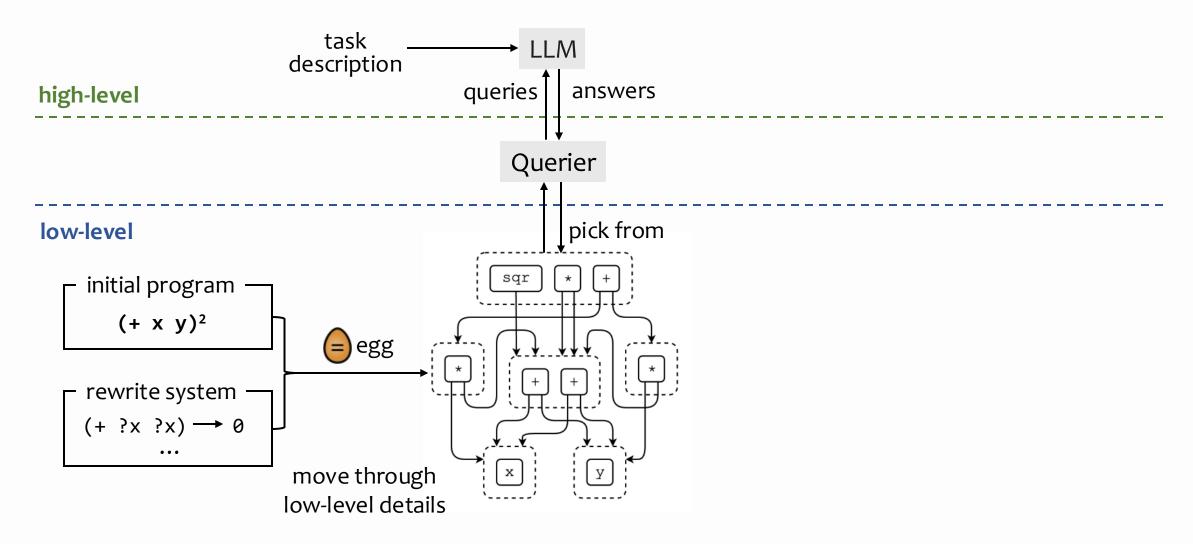
Background: Guided Equality Saturation [KŒHLER et al. 2024]

human users can find useful **checkpoints** w/o full details.

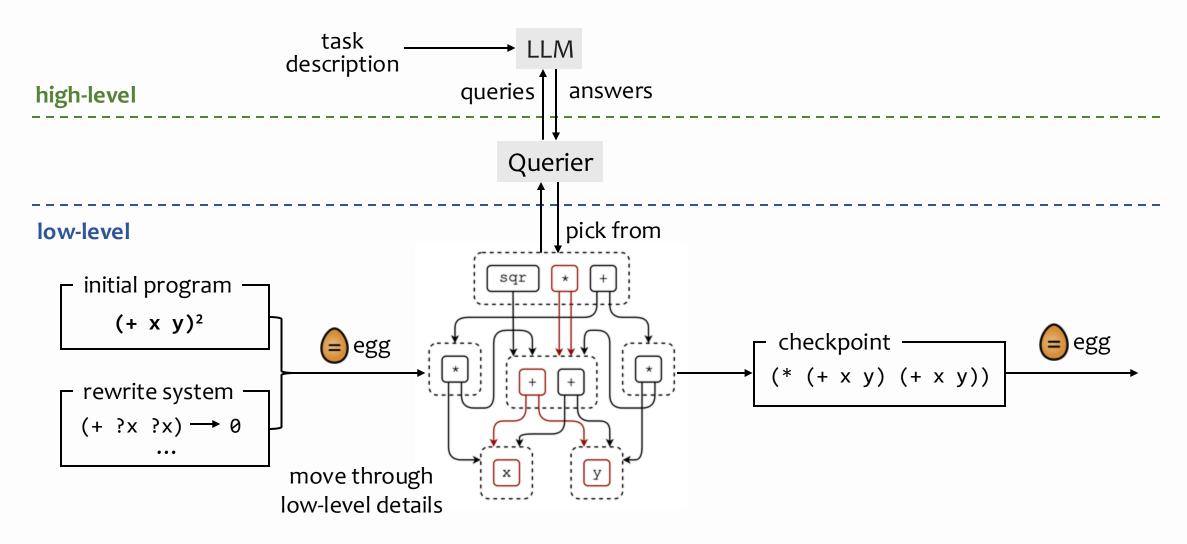




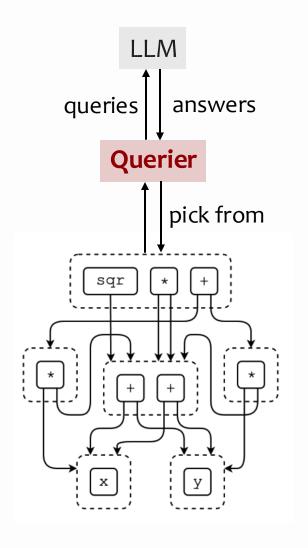
Our Approach: LLM-Guided Equality Saturation



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Core Component: Querier



Challenge 2:

LLM is error prone

more informative queries



Challenge 1:

numerous candidates

Direct Query

(a): show me a checkpoint.

 $= : (+ x^2 (* 2 x y) y^2)$

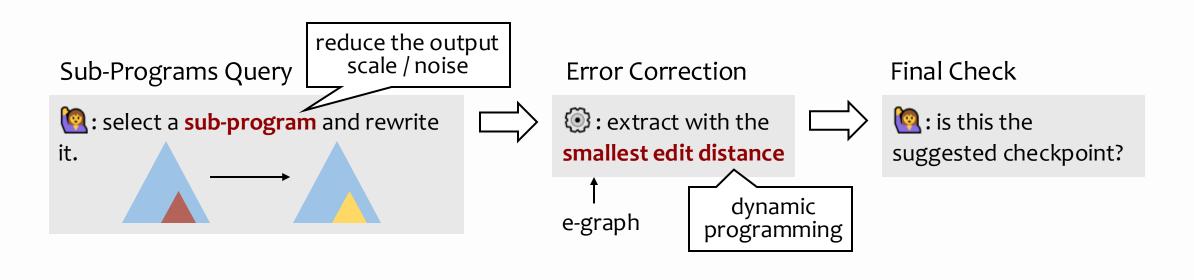
Cons: unreliable

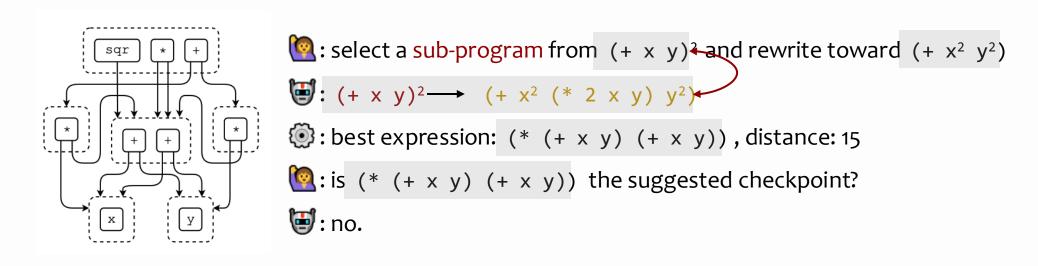
Pointwise Query

: no.

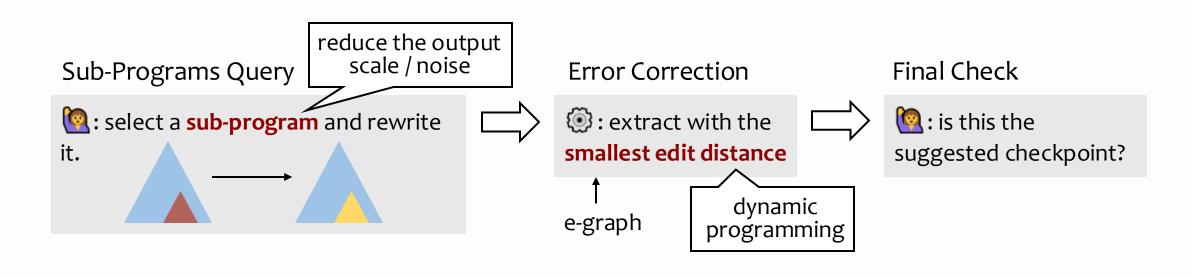
Cons: ineffective

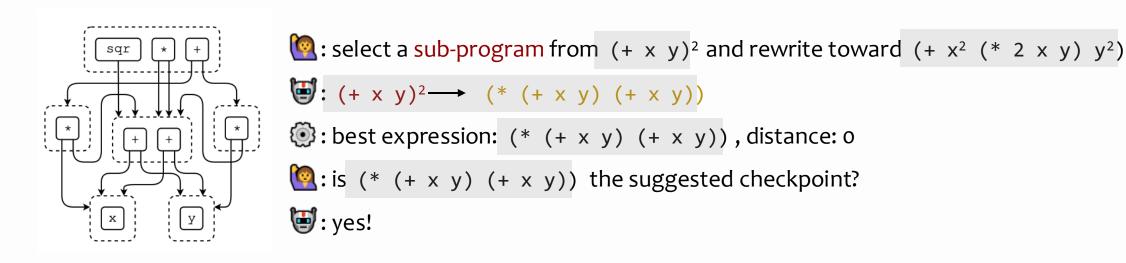
Our Design: Sub-Program Query + Correction





Our Design: Sub-Program Query + Correction





Our Design: Sub-Program Query + Correction

Full Trace (3 iteration limit for saturation)

Round	Initial Expression	S uggested	Extracted
1	(+ x y) ²	(* (+ x y) (+ x y))	

Case Study on Lean [KŒHLER et al. 2024]

Lean Theorem theorem inv_mul: $(a * b)^{-1} = b^{-1} * a^{-1}$ Goal: $(* a b)^{-1} \longrightarrow (* b^{-1} a^{-1})$ $(* ?x ?x^{-1}) \longleftrightarrow 1$ $(* ?x (* ?y ?z)) \longleftrightarrow (* (* ?x ?y) ?z)$... Rewrite Task Goal: $(* a b)^{-1} \longrightarrow (* b^{-1} a^{-1})$ $(* ?x (* ?y ?z)) \longleftrightarrow (* (* ?x ?y) ?z)$...

Case Study on Lean [KŒHLER et al. 2024]

Lean Theorem -

theorem inv_mul : $(a * b)^{-1} = b^{-1} * a^{-1}$



Rewrite Task

Goal: $(* \ a \ b)^{-1} \longrightarrow (* \ b^{-1} \ a^{-1})$ Rules: $(* \ ?x \ 1) \longleftrightarrow ?x$ $(* \ ?x \ ?x^{-1}) \longleftrightarrow 1$ $(* \ ?x \ (* \ ?y \ ?z)) \longleftrightarrow (* \ (* \ ?x \ ?y) \ ?z)$...





Equality Saturation





LLM-Guided

task description

Equality Saturation

Group Theory

theorem inv_mul_c_left:
 a⁻¹ * (a * b) = b
theorem mul_inv_c_left:
 a * (a⁻¹ * b) = b
theorem one_inv: 1⁻¹ = 1
theorem inv_mul:
 (a * b)⁻¹ = b⁻¹ * a⁻¹
theorem inv_inv: a⁻¹ = a

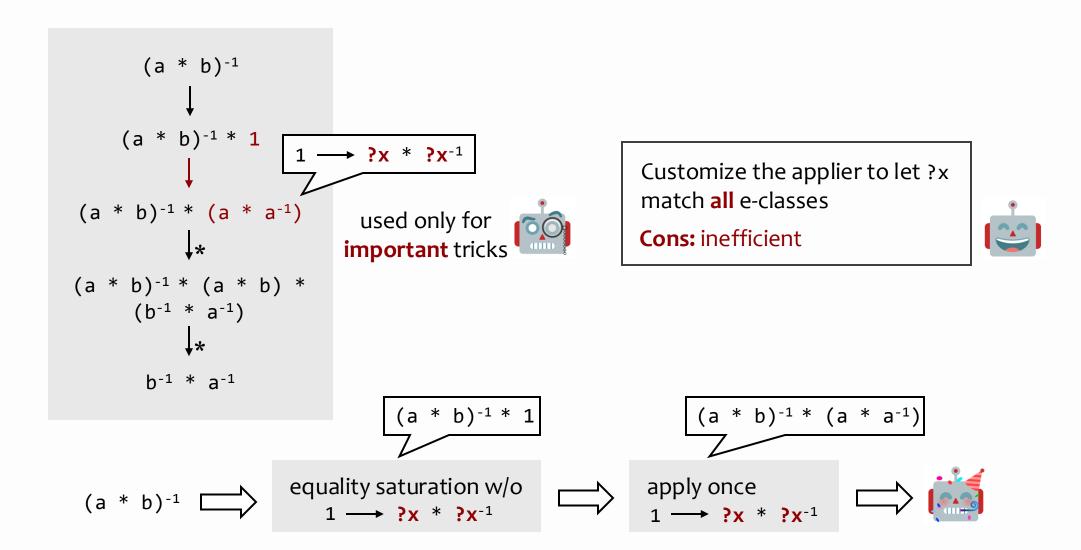
Ring with Char. 2

theorem freshman: $(x + y)^2 = (x^2 + y^2)$ theorem freshman3: $(x + y)^3 =$ $x^3 + x^2y + xy^2 + y^3$ $(+ ?x ?x) \longrightarrow 0$

Binomial Theorem

theorem binomial: $\frac{n!}{(r-1)!(n-r+1)!} + \frac{n!}{r!(n-r)!}$ $= \frac{(n+1)!}{r!(n+1-1)!}$

Case Study on Lean [KŒHLER et al. 2024]



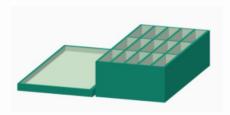
Future Work

Szalinski synthesize structured [Nandi et al. 2020] CAD models

Rise DSL optimize array-operating [KŒHLER et al. 2024] programs

- ~100 rules, 10 K tokens
 customized DSL
- entangled with solvers, type systems, heuristics, ...

Good News: LLMs are good at "understanding" new DSLs



```
(Binop Union
(Binop Diff

(Affine Trans (Vec3 30 60 15)

(Affine Scale (Vec3 60 120 30)

(Affine Trans (Vec3 -0.5 -0.5 -0.5)

(Cube (Vec3 1 1 1) false))))
(Binop Union
```



ing: Here is a DSL program describing a CAD model. What the model is?

: a rectangular box with a 3×5 array of rectangular through-slots cut into one face, plus a shallow backing box on the other side

105 LOC in a Szalinski's DSL

Summary

LLM

Pros: rich knowledge, scalable

Cons: no correctness guarantee



Rewrite System

Pros: ensured correctness

Cons: limited scalability

Thank you!



