# Equality Saturation Theory Exploration á la Carte

Anjali Pal, Brett Saiki, Ryan Tjoa\*, Cynthia Richey\*, Amy Zhu, Oliver Flatt, Max Willsey, Zachary Tatlock, Chandrakana Nandi











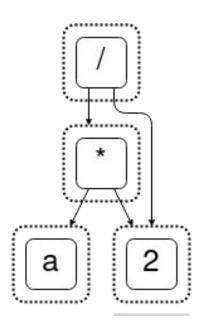


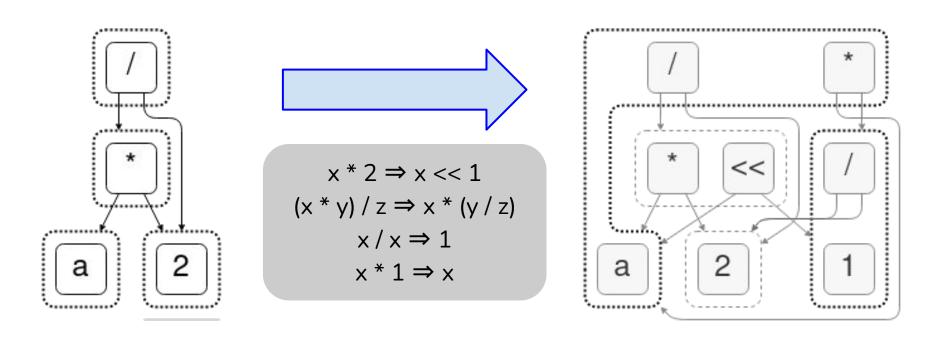


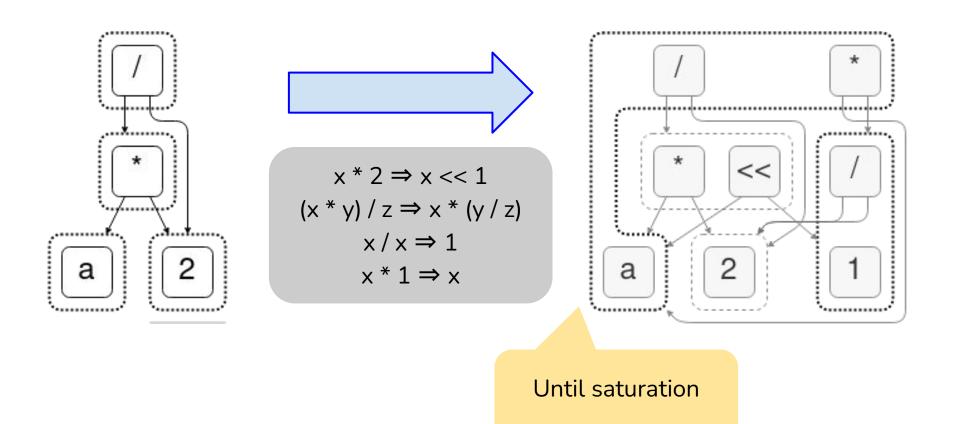




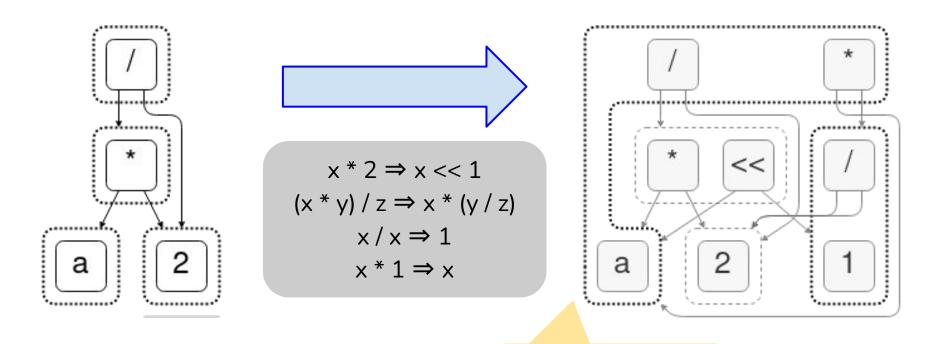








Tate et al. 2009; Willsey et al. 2021



Until saturation ... or resource limit

#### Equality Saturation is everywhere!

#### Automatic End-to-End Joint Optimization for

Equality Saturation for Datapath Synthesis:

A Pathway to Pareto Ontimality

**Automatic Generation of Vectorizing Compilers for Customizable Digital Signal Processors** 

> **Vectorization for Digital Signal Processors** via Equality Saturation

Abstract-VLIW-SIMD compute-heav mances for D written optim burden on pr pilation metho auto-vectoriza intact compila adjustments b

1970s for us advanced to tions in var optimization using rewrite saturation a optimization for achievin introduction and highligh in both RTI

Abstract-

#### Abstract

Embedded a trade-off fro extensive us writing the r DSP enginee produce effe and error-pro or application

Applications tar fast implementa ing auto-vector mance from larg movements nec performance, DS

#### ABSTRACT

#### Automating Constraint-Aware Datapath Optimization using E-Graphs

Samuel Coward Numerical Hardware Group Intel Corporation Email: samuel.coward@intel.com

George A. Constantinides Electrical and Electronic Engineering Imperial College London Email: g.constantinides@imperial.ac.uk

Theo Drane Numerical Hardware Group Intel Corporation Email: theo.drane@intel.com

Abstract-Numerical hardware design requires aggressive optimization, where designers exploit branch constraints, creating optimization opportunities that are valid only on a sub-domain of input space. We developed an RTL optimization tool that automatically learns the consequences of conditional branches and exploits that knowledge to enable deep optimization. The tool deploys custom built program analysis based on abstract interpretation theory, which when combined with a data-structure · evaluation on benchmarks showing the generality of the method.

#### II. BACKGROUND

E-graphs are a data structure that represents equivalence classes (e-classes) of expressions compactly [4], [5]. Nodes in the e-graph represent functions or arguments which are

Equality Saturation is only as powerful as the rules used

# Writing rewrite rules manually is hard

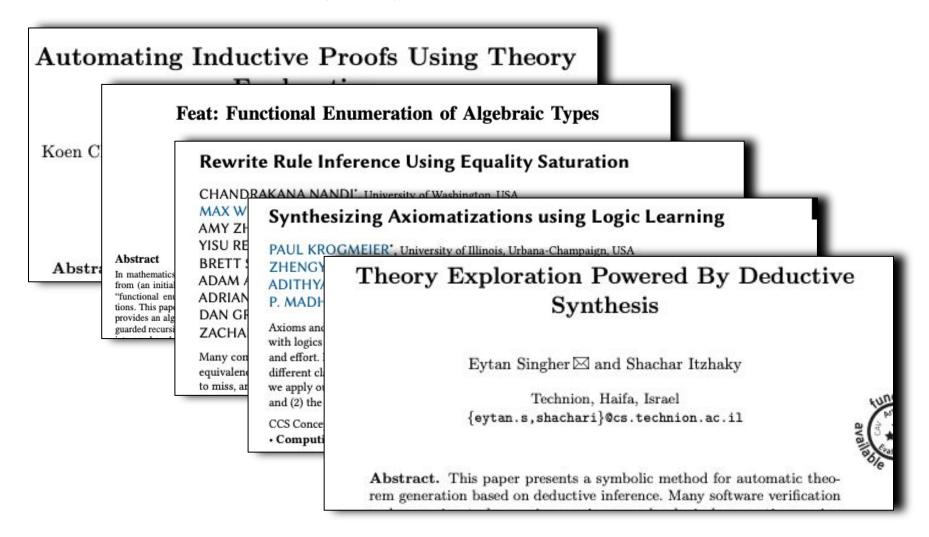








#### **Automated Theory Explorers**



**Problem**: Despite recent work in automated theory exploration, building and maintaining rulesets still requires significant engineering effort

Hypothesis: Traditional theory exploration is too inflexible

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Proposal: Programmable theory exploration using the ENUMO DSL

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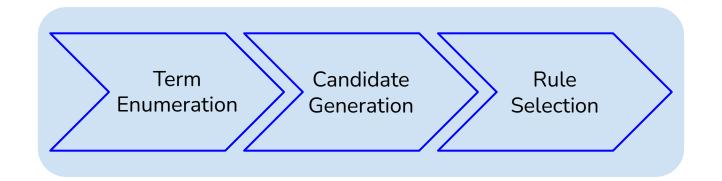
Benefits: Scales better and enables new strategies

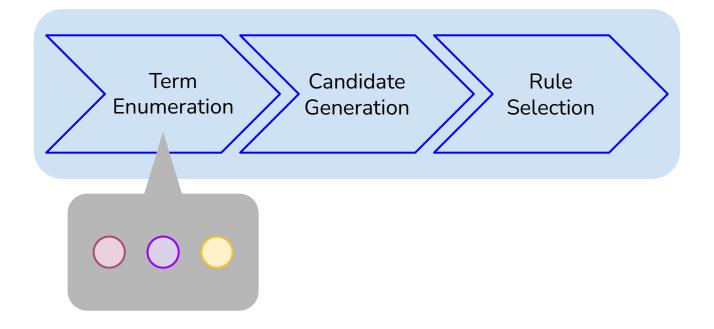
2. The ENUMO DSL

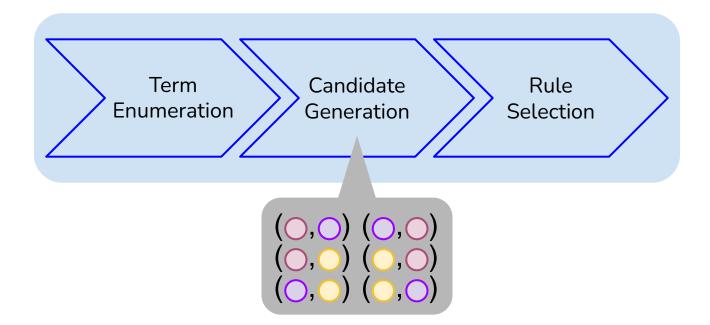
3. Novel scalable rule-finding strategies

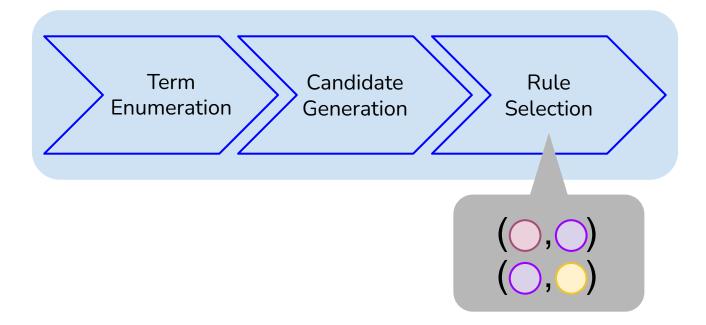
2. The ENUMO DSL

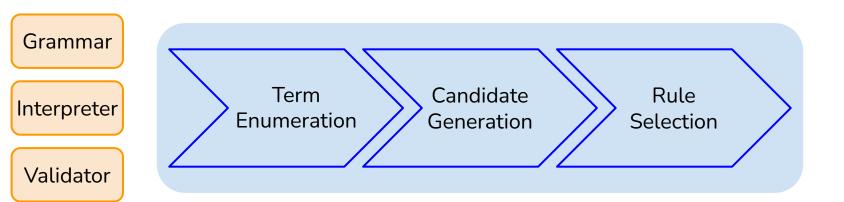
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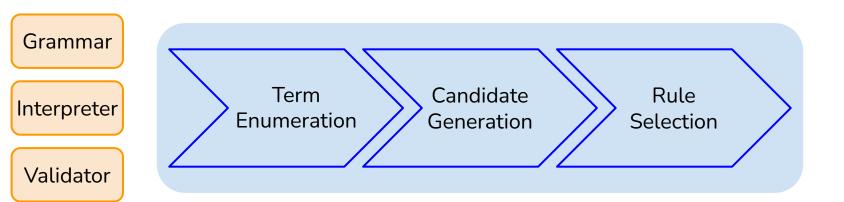


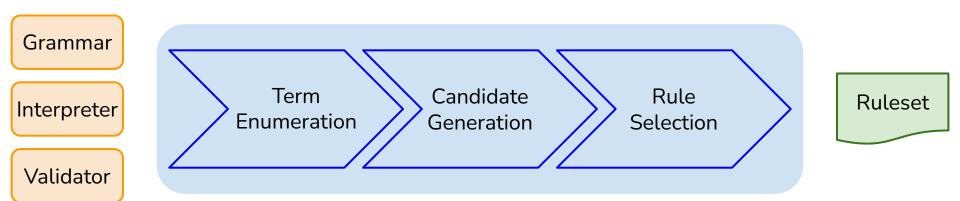


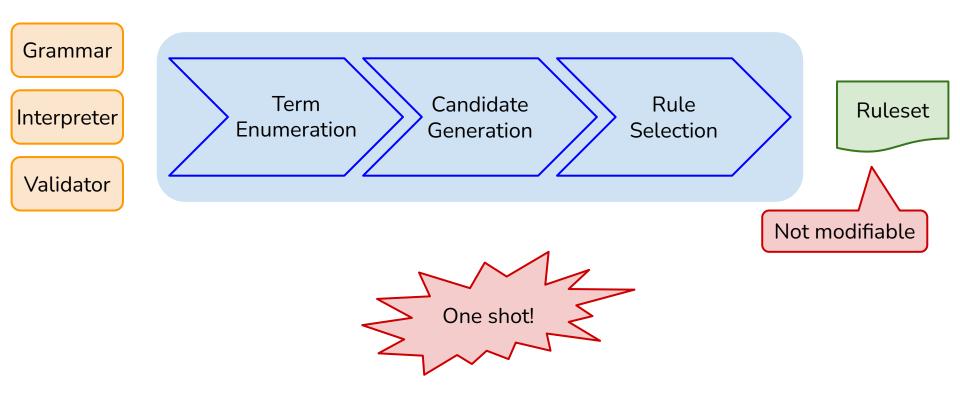
```
<EXPR> :=
Grammar
                            | (Lit n)
                            |(Var v)|
                            | (~ <EXPR>)
                            (+ <EXPR> <EXPR>)
                            | (* < EXPR> < EXPR>)
                            | (- <EXPR> <EXPR>)
                            (/ <EXPR> <EXPR>)
```

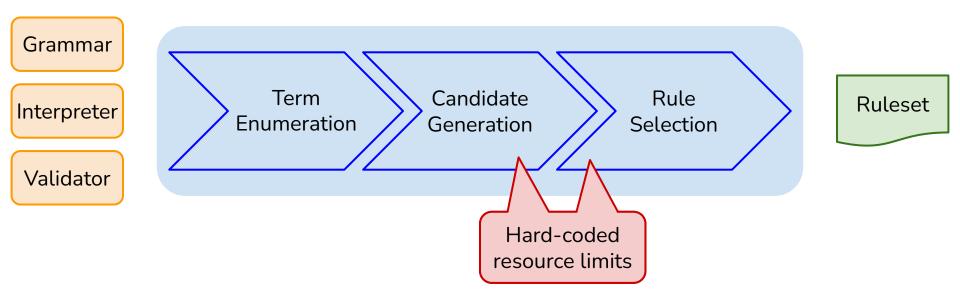
```
def eval(expr):
                           match expr
Grammar
                             |(Const n)| => n
                             |(Var v) => lookup(v)
|(~ e) => -1 * eval(e)
Interpreter
                             |(+ e1 e2)| => eval(e1) + eval(e2)
                             |(* e1 e2) => eval(e1) * eval(e2)
                             |(-e1 e2)| => eval(e1) - eval(e2)
                             |(/ e1 e2) => eval(e1) / eval(e2)
```

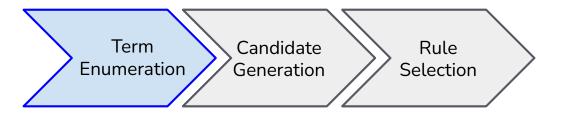
```
def is_valid(lhs, rhs):
Grammar
                         1 = lhs.to_z3()
                         r = rhs.to_z3()
                         z3.assert(1.eq(r).not())
                         match solver.check()
Interpreter
                            |Unsat => true
                            |Sat |Unknown => false
Validator
```





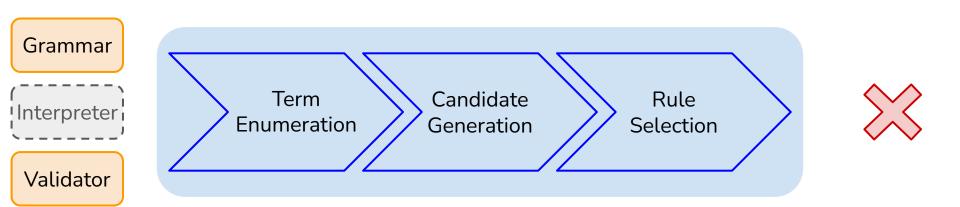






$$<$$
EXPR $>$  := (Lit n) | (Var v) | ( $\sim$   $<$ EXPR $>$ ) | (+  $<$ EXPR $>$ ,  $<$ EXPR $>$ ) | (\*  $<$ EXPR $>$ ,  $<$ EXPR $>$ )

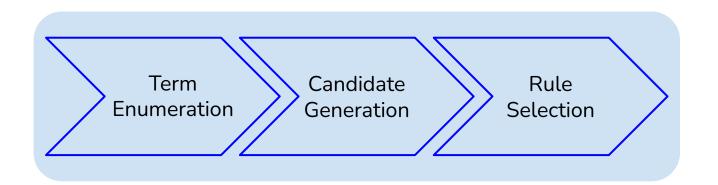
Size 1 Size 2 Size 3 Size 4 Size 5 Size 6  $(\sim (\sim 0))$ (+ a (~ a)) $(\sim a)$ (~ (~ (~ (~ 0)))) (~ (~ (~ (~ (~ 0))))) a (~ b)  $(\sim (\sim 1))$ (+ a (~ b))(~ (~ (~ (~ 1)))) b (~ (~ (~ (~ (~ 1))))) (~ c) (+ a a) (+ (~b) a)(~ (~ (+ a a))) (~ (~ (~ (+ a a)))) -1  $(\sim -1)$ (+ a b)(+ (~b) b)(~ (~ (+ a b))) (~ (~ (~ (+ a b)))) 0  $(\sim 0)$ (+ a 0)(+ b (~ a))(~ (~ (+ a 0))) (~ (~ (~ (+ a 0)))) (~ 1) (+a1)(+ b (~ b))(~ (~ (+ a 1))) (~ (~ (~ (+ a 1)))) 4902 52134 6 6 78 366



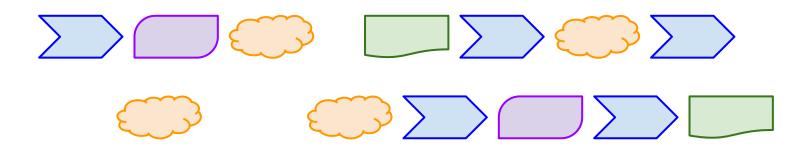
2. The ENUMO DSL

3. Novel scalable rule-finding strategies

# **Insight**: Users have intuition about which parts of the domain are worth exploring



We turn theory explorers inside out to expose a small set of useful operators for rule inference



#### **ENUMO DSL**

```
lits = Workload { a b c 0 1 }
```

#### **ENUMO DSL**

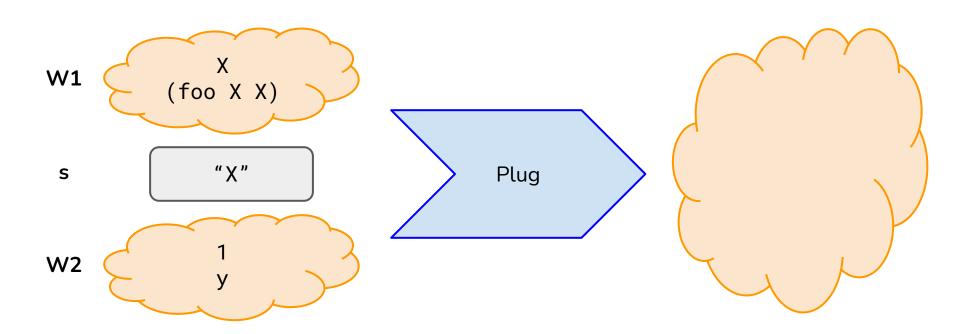
```
lits = Workload { a b c 0 1 }
exprs = Workload { LIT (~ EXPR) (+ EXPR EXPR) }
                                 <EXPR> :=
                                  (Lit n)
                                  (~ < EXPR>)
                                  (+ < EXPR > < EXPR >)
```

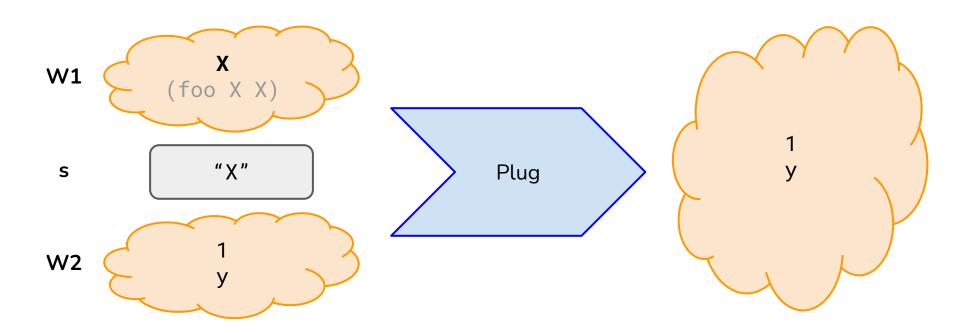
#### **ENUMO DSL**

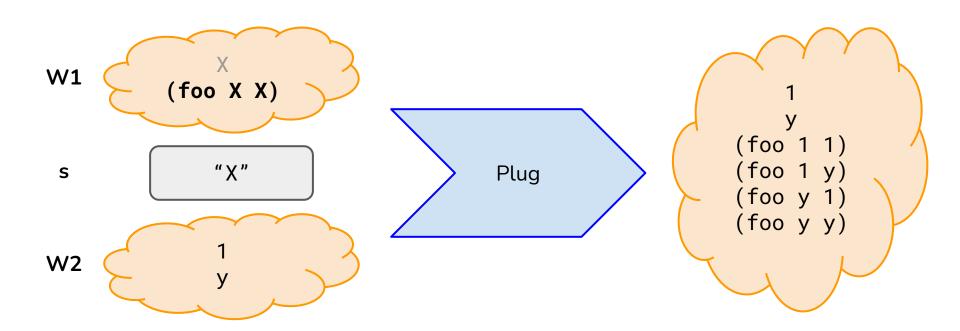
```
lits = Workload { a b c 0 1 }
exprs = Workload { LIT (~ EXPR) (+ EXPR EXPR) }
wkld = exprs.plug("EXPR", exprs)
```

# Plug $\mathcal{W}_1$ s $\mathcal{W}_2$

All combinations of replacing s in  $W_I$  with a term from  $W_2$ 







```
lits = Workload { a b c 0 1 }
sums = Workload { (+ EXPR EXPR) }
products = Workload { (* EXPR EXPR) }
sums_of_products =
  sums.plug("EXPR", products.plug("EXPR", lits))
```

```
lits = Workload { a b c 0 1 }
sums = Workload { (+ EXPR EXPR) }
products = Workload { (* EXPR EXPR) }
sums_of_products =
      sums.plug("EXPR", products.plug("EXPR", lits))

      (* a a)
      (* b a)
      (* c a)
      (* 0 a)
      (* 1 a)

      (* a b)
      (* b b)
      (* c b)
      (* 0 b)
      (* 1 b)

      (* a c)
      (* b c)
      (* c c)
      (* 0 c)
      (* 1 c)

      (* a 0)
      (* b 0)
      (* c 0)
      (* 1 0)

       (* a 1) (* b 1) (* c 1) (* 0 1) (* 1 1)
```

```
lits = Workload { a b c 0 1 }
sums = Workload { (+ EXPR EXPR) }
products = Workload { (* EXPR EXPR) }
sums_of_products =
     sums.plug("EXPR", products.plug("EXPR", lits))
     (+ (* a a) (* a a)) (+ (* a a) (* b a)) (+ (* a a) (* c a))
     (+ (* a a) (* a b)) (+ (* a a) (* b b)) (+ (* a a) (* c b))

(+ (* a a) (* a c)) (+ (* a a) (* b c)) (+ (* a a) (* c c))

(+ (* a a) (* a 0)) (+ (* a a) (* b 0)) . . .

(+ (* a a) (* a 1)) (+ (* a a) (* b 1)) (+ (* 1 1) (* 1 1))
```

```
lits = Workload { a b c 0 1 }
exprs = Workload { LIT (~ EXPR) (+ EXPR EXPR) }
wkld = exprs.plug("EXPR", exprs)
            .plug("LIT", lits)
```

```
lits = Workload { a b c 0 1 }
exprs = Workload { LIT (~ EXPR) (+ EXPR EXPR) }
wkld = exprs.plug("EXPR", exprs)
          .plug("LIT", lits)
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lits = Workload { a b c 0 1 }
exprs = Workload { LIT (~ EXPR) (+ EXPR EXPR) }
wkld = exprs.plug("EXPR", exprs)
            .plug("LIT", lits)
rules =
  wkld
    .to_egraph()
    .find_candidates()
    .select_rules(limits)
```

```
lits = Workload { a b c 0 1 }
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```

Term Enumeration

```
lits = Workload { a b c 0 1 }
exprs = Workload { LIT (~ EXPR) (+ EXPR EXPR) }
wkld = exprs.plug("EXPR", exprs)
            .plug("LIT", lits)
rules =
  wkld
    .to_egraph()
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lits = Workload { a b c 0 1 }
exprs = Workload { LIT (~ EXPR) (+ EXPR EXPR) }
wkld = exprs.plug("EXPR", exprs)
            .plug("LIT", lits)
rules =
  wkld
    .to_egraph()
    .find_candidates()
    .select_rules(limits)
```

Term Candidate Rule Enumeration Generation Selection

```
e1 = Workload { (~ EXPR) (+ EXPR EXPR) }
e2 = Workload { 1 (+ 2 3) (+ (+ 4 5) 6) }
e1.plug("EXPR", e2)
  .filter(\lambda t. size t < 4)
```

```
e1 = Workload { (~ EXPR) (+ EXPR EXPR) }
e2 = Workload \{ 1 (+ 2 3) (+ (+ 4 5) 6) \}
e1.plug("EXPR", e2)
   .filter(\lambda t. size t < 4)
    (~ 1)

(~ (+ 2 3) 1)

(~ (+ 2 3))

(~ (+ (+ 2 3) (+ 2 3))

(~ (+ (+ 3 4) 5))

(+ (+ 2 3) (+ (+ 4 5) 6))

(+ (+ (+ 4 5) 6) 1)

(+ (+ (+ 4 5) 6) (+ 2 3))
     (+1 (+ (+ 4 5) 6)) (+ (+ (+ 4 5) 6) (+ (+ 4 5) 6))
```

```
e1 = Workload { (~ EXPR) (+ EXPR EXPR) }
e2 = Workload { 1 (+ 2 3) (+ (+ 4 5) 6) }
e1.plug("EXPR", e2)
 .filter(\lambda t. size t < 4)
                (+ (+ 2 3) 1)
                (+ (+ 2 3) (+ 2 3))
    (+ (+ 3 4) 5)) (+ (+ 2 3) (+ (+ 4 5) 6))
                (+(+(+45)6)1)
```

```
e1 = Workload { (~ EXPR) (+ EXPR EXPR) }
e2 = Workload { 1 (+ 2 3) (+ (+ 4 5) 6) }
e1.plug("EXPR", e2.filter(\lambdat. size t < 4))—
   .filter(\lambda t. size t < 4)
```

```
e1 = Workload { (~ EXPR) (+ EXPR EXPR) }
e2 = Workload { 1 (+ 2 3) (+ (+ 4 5) 6) }
e1.plug("EXPR", e2.filter(λt. size t < 4))
   .filter(\lambda t. size t < 4)
   (~ 1)
(~ (+ 2 3))
(+ 1 1)
(+ 1 (+ 2 3))
(+ (+ 2 3) 1)
    (+ (+ 2 3) (+ 2 3))
```

```
e1 = Workload { (~ EXPR) (+ EXPR EXPR) }
e2 = Workload { 1 (+ 2 3) (+ (+ 4 5) 6) }
e1.plug("EXPR", e2.filter(λt. size t < 4))
  .filter(\lambda t. size t < 4)
```

**Optimization**: Pushing Filters through Plugs

[[ Filter f (Plug W1 s W2) ]] = [[ Filter f (Plug W1 s (Filter f W2))]]

[[ Filter f (Plug W1 s W2) ]] = [[ Filter f (Plug W1 s (Filter f W2))]]

Requires monotonicity of f

A filter f is monotonic if, for every term t satisfying f, every subterm  $s \subseteq t$ also satisfies f

A filter f is monotonic if, for every term t satisfying f, every subterm  $s \in t$ also satisfies f

Excludes(
$$(+ (* x x) (* y y)), "z")$$

Contains
$$((+ (* x x) (* y y)), "x")$$

Monotonic

A filter f is monotonic if, for every term t satisfying f, every subterm  $s \in t$ also satisfies f

Excludes((+ (\* x x) (\* y y)), "z")

Contains((+ (\* x x) (\* y y)), "x")

Monotonic

A filter f is monotonic if, for every term t satisfying f, every subterm  $s \in t$ also satisfies f

Excludes(
$$(+ (* x x) (* y y)), "z")$$

Contains(
$$(+ (* x x) (* y y)), "x")$$

Not monotonic

1. Traditional theory exploration

2. The ENUMO DSL

3. Novel scalable rule-finding strategies

# Comparison to Ruler

Domain	ENUMO LOC	# ENUMO	# Ruler	ENUMO → Ruler	Ruler → ENUMO
bool	44	64	51	100%	87.5%
bv4	21	180	84	100%	38.3%
bv32	20	120	78	100%	58.3%
rational	51	131	113	100%	62.6%

# Comparison to Ruler

Domain	Enumo LOC	# Enumo	# Ruler	ENUMO → Ruler	Ruler → Enumo
bool	44	64	51	100%	87.5%
bv4	21	180	84	100%	38.3%
bv32	20	120	78	100%	58.3%
rational	51	131	113	100%	62.6%

Check out the paper for more about derivability! (TLDR: More rules are not always better)

```
G = Workload {
        EXPR EXPR)
    <= EXPR EXPR)
    == EXPR EXPR)
    != EXPR EXPR)
        EXPR)
        EXPR)
    (&& EXPR EXPR)
       EXPR EXPR)
    (^ EXPR EXPR)
    (+ EXPR EXPR)
    – EXPR EXPR)
    (* EXPR EXPR)
    / EXPR EXPR)
    (min EXPR EXPR)
    (max EXPR EXPR)
    (select EXPR EXPR EXPR)
```

```
G = Workload {
        EXPR EXPR)
     <= EXPR EXPR)
     == EXPR EXPR)
         EXPR EXPR)
         EXPR)
         EXPR)
    (&& EXPR EXPR)
        EXPR EXPR)
        EXPR EXPR)
    (+ EXPR EXPR)
      EXPR EXPR)
        EXPR EXPR)
        EXPR EXPR)
    (min EXPR EXPR)
    (max EXPR EXPR)
    (select EXPR EXPR EXPR)
```

725 rules with no side conditions. unsupported operators, or unbound variables

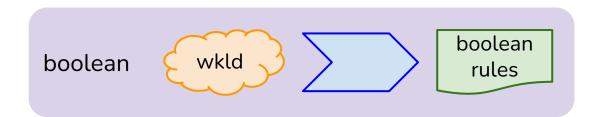
Term Size # Rules  $ENUMO \rightarrow Halide$ 

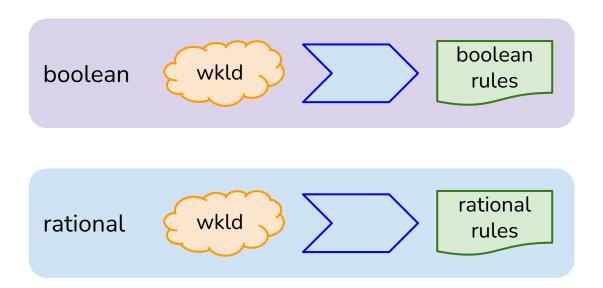
```
G = Workload {
        EXPR EXPR)
    <= EXPR EXPR)
    == EXPR EXPR)
        EXPR EXPR)
        EXPR)
        EXPR)
    (&& EXPR EXPR)
        EXPR EXPR)
    ^ EXPR EXPR)
    (+ EXPR EXPR)
    EXPR EXPR)
    (* EXPR EXPR)
      EXPR EXPR)
    (min EXPR EXPR)
    (max EXPR EXPR)
    (select EXPR EXPR EXPR)
```

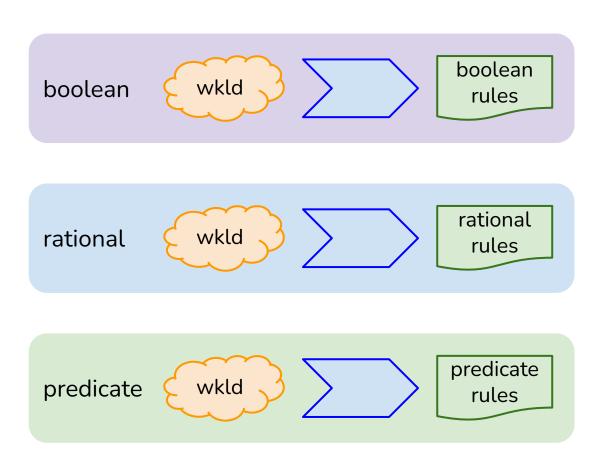
Term Size	# Rules	Enumo → Halide	
3	96	2.9%	
4	224	6.9%	
5	485	42.6%	
6	TIMEOUT	TIMEOUT	

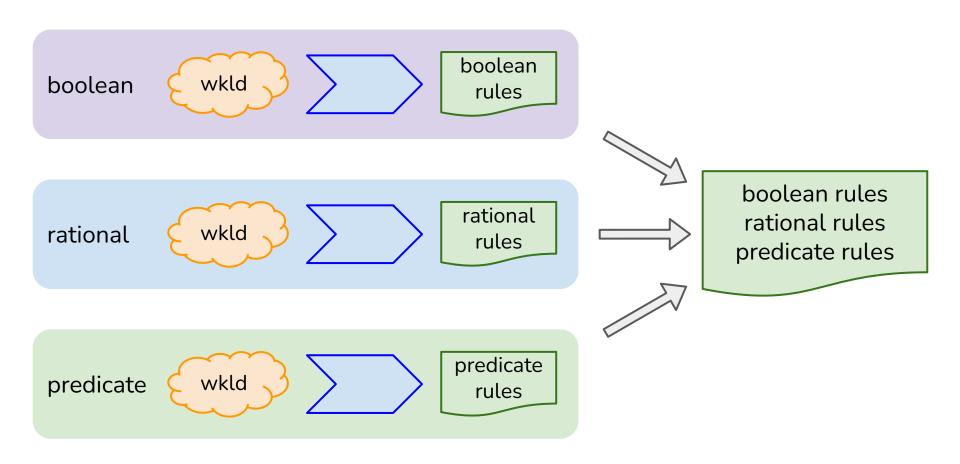
```
G = Workload {
        EXPR EXPR)
    <= EXPR EXPR)
    == EXPR EXPR)
     != EXPR EXPR)
        EXPR)
        EXPR)
    (&& EXPR EXPR)
       EXPR EXPR)
    ^ EXPR EXPR)
    (+ EXPR EXPR)
    - EXPR EXPR)
    * EXPR EXPR)
      EXPR EXPR)
    (min EXPR EXPR)
    (max EXPR EXPR)
    (select EXPR EXPR EXPR)
```

Domain expert know which operators are closely related









boolean rules rational rules predicate rules

boolean rules rational rules predicate rules



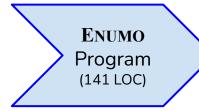
boolean rules rational rules predicate rules



**ENUMO** Program (141 LOC)

boolean rules rational rules predicate rules





Term Size	# Rules	<b>E</b> NUMO → <b>Halide</b>
Custom	845	90.6%

Guided search enables progress past exponential blowup

# ENUMO enables new rule inference strategies

# ENUMO enables new rule inference strategies

**Fast-forwarding** is a phased approach for finding "shortcut" rules

# ENUMO enables new rule inference strategies

Implemented with <10 lines of ENUMO

**Fast-forwarding** is a phased approach for finding "shortcut" rules

Doesn't require an interpreter!

```
def fast_forward(wkld, R, E):
 G = wkld.to_egraph()
  allowed = \{r \in R \mid r.is\_allowed()\}
 G' = G.compress(allowed, limits)
 G'' = G'.eqsat(E, limits)
 C = by_diff(G', G'')
 G''' = G''.compress(R, limits)
  C.union(by_diff(G'', G'''))
  return C.select_rules(allowed, limits)
```

Carefully crafted rewrite rules over real numbers



$$a+b \leadsto \frac{a^2-b^2}{a-b}$$

$$\cos^2(a) + \sin^2(a) \rightsquigarrow 1$$

Carefully crafted rewrite rules over real numbers

Learnable with traditional techniques



$$a+b \leadsto \frac{a^2-b^2}{a-b}$$

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Carefully crafted rewrite rules over real numbers

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No interpreter ⇒ Can't learn with traditional techniques

Carefully crafted rewrite rules over real numbers

Learnable with traditional techniques



$$a+b \leadsto \frac{a^2-b^2}{a-b}$$

$$\cos^2(a) + \sin^2(a) \rightsquigarrow 1$$

No interpreter ⇒ Can't learn with traditional techniques

$$cis(x) = cos(x) + i sin(x)$$
$$sin(x) \leadsto \frac{cis(x) - cis(-x)}{2i}$$

Carefully crafted rewrite rules over real numbers

Learnable with traditional techniques



$$a+b \leadsto \frac{a^2-b^2}{a-b}$$

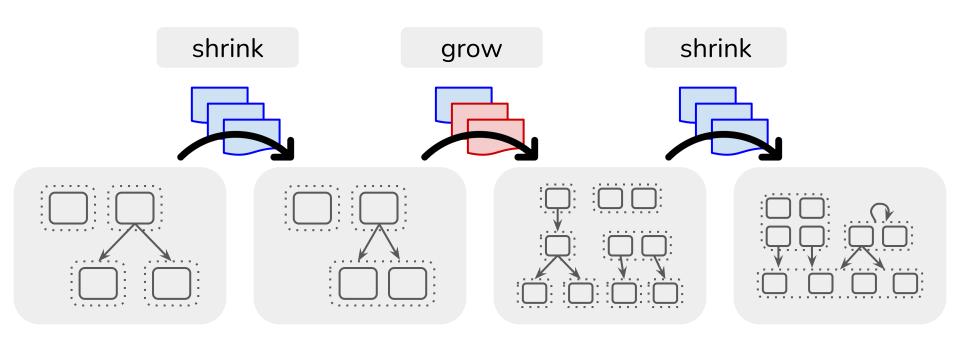
$$\cos^2(a) + \sin^2(a) \rightsquigarrow 1$$

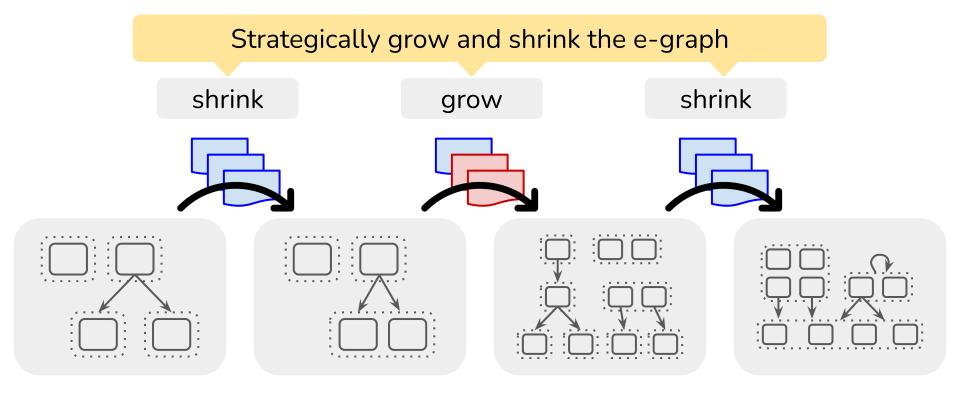
No interpreter ⇒ Can't learn with traditional techniques

$$\operatorname{cis}(x) = \cos(x) + i \sin(x)$$

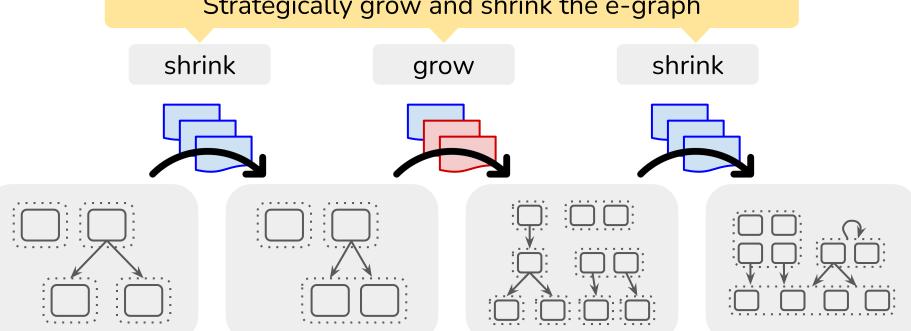
$$\sin(x) \leadsto \frac{\operatorname{cis}(x) - \operatorname{cis}(-x)}{2i}$$

Rewriting through complex terms is not feasible due to resource limits





#### Strategically grow and shrink the e-graph



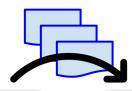
Extract rule candidates from merged e-classes

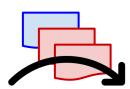
Strategically grow and shrink the e-graph

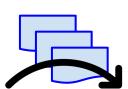
shrink

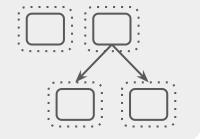
grow

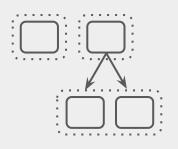
shrink

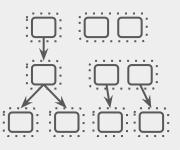


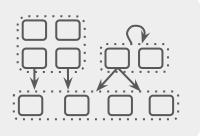






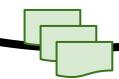






Extract rule candidates from merged e-classes

"Shortcut" rules improve results under given resource limits

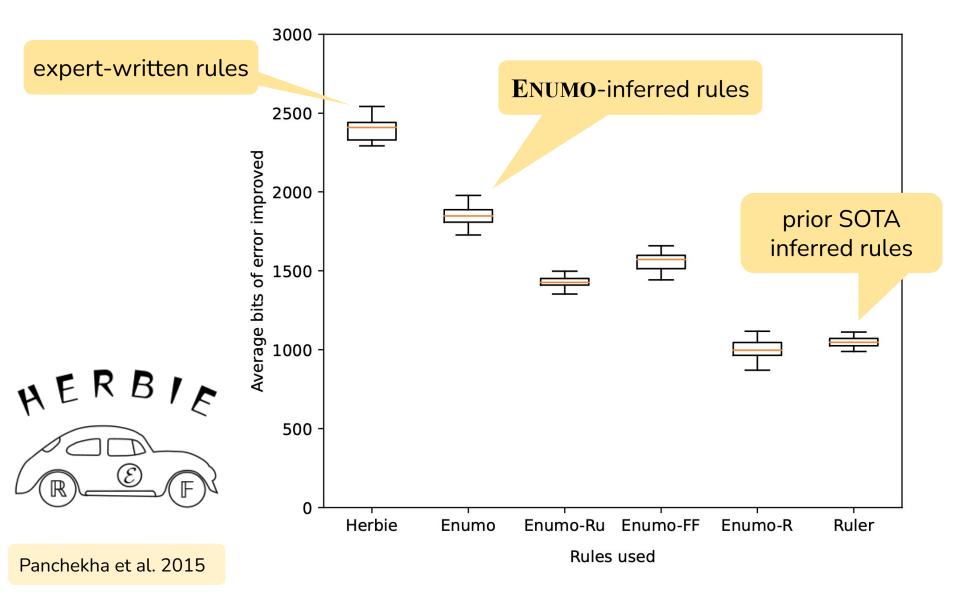


$$\sin(b+a) \leadsto \sin(b) \cdot \cos(a) + \sin(a) \cdot \cos(b)$$
$$\sin(b) \cdot \sin(a) \leadsto \frac{\cos(b-a) - \cos(b+a)}{2}$$

$$c^{ba} \leadsto (c^a)^b$$
$$(c^b)^{\log(a)} \leadsto (a^b)^{\log(c)}$$
$$\sqrt{b^a} \leadsto (\sqrt{b})^a$$

 $\mathsf{Scale}(a,b,c,\mathsf{Trans}(d,e,f,s)) \leadsto \mathsf{Trans}(da,eb,fc,\mathsf{Scale}(a,b,c,s))$  $\mathsf{Cube}(ad, be, cf) \leadsto \mathsf{Scale}(a, b, c, \mathsf{Cube}(d, e, f))$ 

#### End-to-end: Herbie

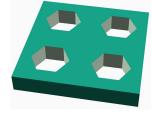


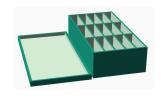
#### End-to-end: Szalinski

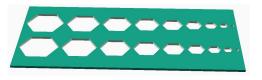
#### **Percent AST Shrinkage**

Program Id	<b>Handwritten Rules</b>	<b>Enumo-synthesized Rules</b>
TackleBox	91%	85%
SDCardRack	87%	86%
SingleRowHolder	92%	92%
CircleCell	80%	80%
CNCBitCase	89%	89%
CassetteStorage	89%	89%
RaspberryPiCover	96%	96%
ChargingStation	89%	82%
CardFramer	76%	76%
HexWrenchHolder	95%	84%

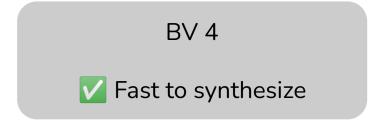


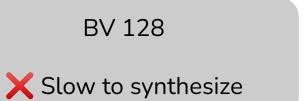




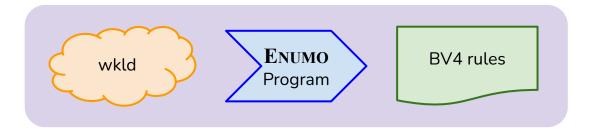


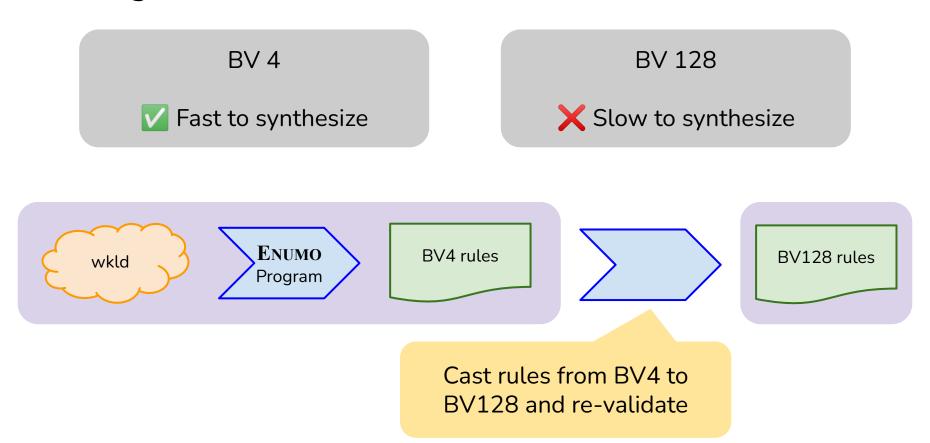


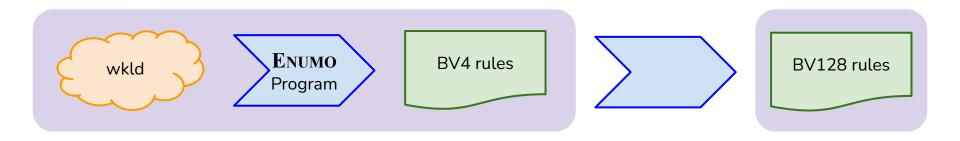












Generated Rules (Time) Ported BV4 Rules (Time) Ported → Generated

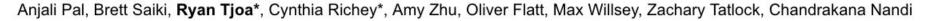
190 (1784.14) 210 (38.68) 91%

Directly synthesized BV128 rules

Of the 246 BV4 rules, 210 are sound for BV128 The ported rules have almost as much proving power at a fraction of the cost

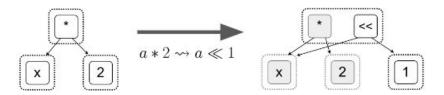
#### Equality Saturation Theory Exploration à la Carte

Synthesize better rewrite rulesets!



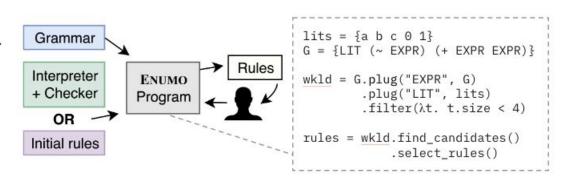
**Equality saturation** uses rewrite rules for program optimization, verification, synthesis, etc.

How do we discover rewrite rules?



**ENUMO:** a theory exploration DSL

Infer rules <u>incrementally</u> using <u>composable</u> search operators, even without an interpreter.



Metric for proving power; see §4.3

Derivability vs. Ruler (prior SOTA) for common theories:

Domain	$\textbf{Enumo} \rightarrow \textbf{Ruler}$	Ruler → Enumo
bool	100%	87.5%
bv4	100%	38.3%
bv32	100%	58.3%
rational	100%	62.6%

#### **Evaluation & Case Studies**

**Herbie**: 35% higher accuracy than with Ruler

**Szalinski**: shrink CAD programs by 87% (expert-written identities shrink by 90%)

$$a+b \leadsto \frac{a \cdot a - b \cdot b}{a-b} \qquad \text{Scale}(a,b,c,\operatorname{Trans}(d,e,f,s)) \leadsto \operatorname{Trans}(da,eb,fc,\operatorname{Scale}(a,b,c,s)) \\ \cos(b+a) \leadsto \cos b \cdot \cos a - \sin b \cdot \sin a \qquad \text{Cube}(ad,be,cf) \leadsto \operatorname{Scale}(a,b,c,\operatorname{Cube}(d,e,f))$$

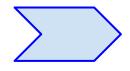
$$(c^b)^{\log a} \leadsto (a^b)^{\log c}$$





#### https://uwplse.org/ruler/

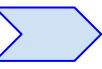




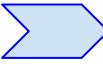












Composable Operators

Incremental Ruleset Building

ENUMO: A DSL for

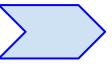
Programmable Theory Exploration

Leverage Domain Expertise

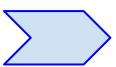
New Inference Techniques





























#### **ENUMO DSL**

#### Monotonic

Excludes

MetricLt

And

#### Not Monotonic

Contains

MetricEq

0r

Not

Canon