

Enabling Robust Equality Saturation Through Flexible Theory Exploration

Anjali Pal

1. Fast, Flexible, Robust **Term Rewriting**
via Equality Saturation

2. Fast, Flexible, Robust **Rule Inference**
for Equality Saturation

Term Rewriting

Synthesis

Canonicalization

Equivalence Checking

SMT Solvers

Optimization

Verification

Code Generation

Symbolic Evaluation

And more!

Term Rewriting: Optimization

Task: Compile the program

```
if True:
    y = x + x
    r = y * 2 + 0
else:
    r = x * 1

return r
```

Term Rewriting: Optimization

Task: Compile the program

```
if True:
```

```
    y = x + x
```

```
    r = y * 2 + 0
```

```
else:
```

```
    r = x * 1
```

```
return r
```

```
return 4 * y
```

$x + x \Rightarrow 2 * x$

$\text{if True then } x \text{ else } y \Rightarrow x$

$x + 0 \Rightarrow x$

$2 * 2 \Rightarrow 4$

Term Rewriting: Equivalence Checking

Task: Determine if refactor is safe

```
def is_even(n):  
    return n % 2 == 0
```

```
def foo(x, y):  
    if is_even(x):  
        return x + y  
    else:  
        return x - y
```

```
def is_odd(n):  
    return n % 2 != 0
```

```
def foo(x, y):  
    if is_odd(x):  
        return x - y  
    else:  
        return x + y
```

Term Rewriting: Equivalence Checking

Task: Determine if refactor is safe

```
def is_even(n):  
    return n % 2 == 0
```

```
def foo(x, y):  
    if is_even(x):  
        return x + y
```

```
else:
```

```
    return
```

```
def is_odd(n):  
    return n % 2 != 0
```

```
def foo(x, y):  
    if is_odd(x):  
        return x - y
```

```
    return x + y
```

$\text{is_even}(a) \implies a \% 2 = 0$

$\text{is_odd}(a) \implies a \% 2 \neq 0$

$x = y \implies !(x \neq y)$

$\text{if } A \text{ then } B \text{ else } C \implies$

$\text{if } !A \text{ then } C \text{ else } B$



Term Rewriting: Canonicalization

Task: Put in Polynomial Normal Form

$$2(x + 1) + 3xy + xy + 4$$

Term Rewriting: Canonicalization

Task: Put in Polynomial Normal Form

$$2(x + 1) + 3xy + xy + 4$$

$$4xy + 2x + 6$$

$$a + (b + c) \implies (a + b) + c$$

$$a * m + b * m \implies (a + b) * m$$

$$a + b \implies b + a$$

$$c * (a + b) \implies c * a + c * b$$

...

$$(a * 2) / 2 \Rightarrow a$$

$$(a * 2) / 2 \Rightarrow a$$

Rewrite Rules

$$(x * y) / z \Rightarrow x * (y / z)$$

$$x / x \Rightarrow 1$$

$$x * 1 \Rightarrow x$$

$$x * 2 \Rightarrow x << 1$$

$$x * y \Rightarrow y * x$$

$$x \Rightarrow x * 1$$

$$(a * 2) / 2 \longrightarrow a$$

$$(a * 2) / 2 \implies a * (2 / 2) \implies a * 1 \implies a$$



Rewrite Rules

$$(x * y) / z \implies x * (y / z)$$

$$x / x \implies 1$$

$$x * 1 \implies x$$

$$x * 2 \implies x \ll 1$$

$$x * y \implies y * x$$

$$x \implies x * 1$$

$$(a * 2) / 2 \longrightarrow a$$

$$(a * 2) / 2 \implies (a << 1) / 2$$



stuck

Rewrite Rules

$$(x * y) / z \implies x * (y / z)$$

$$x / x \implies 1$$

$$x * 1 \implies x$$

$$x * 2 \implies x << 1$$

$$x * y \implies y * x$$

$$x \implies x * 1$$

$$(a * 2) / 2 \longrightarrow a$$

$$(a * 2) / 2 \Longrightarrow (2 * a) / 2 \Longrightarrow (a * 2) / 2 \Longrightarrow \dots$$



diverge

Rewrite Rules

$$(x * y) / z \Longrightarrow x * (y / z)$$

$$x / x \Longrightarrow 1$$

$$x * 1 \Longrightarrow x$$

$$x * 2 \Longrightarrow x << 1$$

$$x * y \Longrightarrow y * x$$

$$x \Longrightarrow x * 1$$

$$(a * 2) / 2 \Rightarrow a$$

$$a \Rightarrow a * 1 \Rightarrow (a * 1) * 1 \Rightarrow \dots$$



**infinite
size**

Rewrite Rules

$$(x * y) / z \Rightarrow x * (y / z)$$

$$x / x \Rightarrow 1$$

$$x * 1 \Rightarrow x$$

$$x * 2 \Rightarrow x << 1$$

$$x * y \Rightarrow y * x$$

$$x \Rightarrow x * 1$$

$$(a * 2) / 2 \Rightarrow a$$

USEFUL

$$(x * y) / z \Rightarrow x * (y / z)$$

$$x / x \Rightarrow 1$$

$$x * 1 \Rightarrow x$$

NOT SO USEFUL

$$x * 2 \Rightarrow x << 1$$

$$x * y \Rightarrow y * x$$

$$x \Rightarrow x * 1$$

$$(a * 2) / 2 \Rightarrow a$$

But critical for other inputs!

USEFUL

$$(x * y) / z \Rightarrow x * (y / z)$$

$$x / x \Rightarrow 1$$

$$x * 1 \Rightarrow x$$

NOT SO USEFUL

$$x * 2 \Rightarrow x << 1$$

$$x * y \Rightarrow y * x$$

$$x \Rightarrow x * 1$$

$$(a * 2) / 2 \Rightarrow a$$

Which rewrite? When?

USEFUL

$$(x * y) / z \Rightarrow x * (y / z)$$

$$x / x \Rightarrow 1$$

$$x * 1 \Rightarrow x$$

NOT SO USEFUL

$$x * 2 \Rightarrow x << 1$$

$$x * y \Rightarrow y * x$$

$$x \Rightarrow x * 1$$

$$(a * 2) / 2 \Rightarrow a$$

All of them at once!

USEFUL

$$(x * y) / z \Rightarrow x * (y / z)$$

$$x / x \Rightarrow 1$$

$$x * 1 \Rightarrow x$$

NOT SO USEFUL

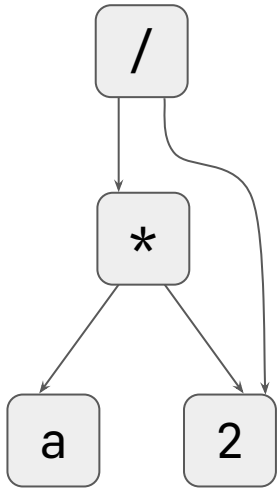
$$x * 2 \Rightarrow x << 1$$

$$x * y \Rightarrow y * x$$

$$x \Rightarrow x * 1$$

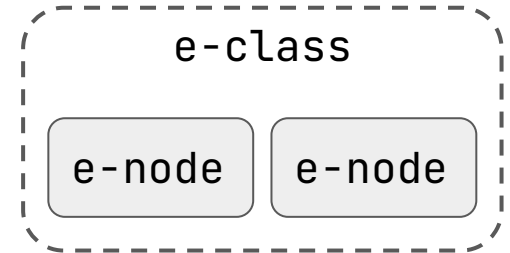
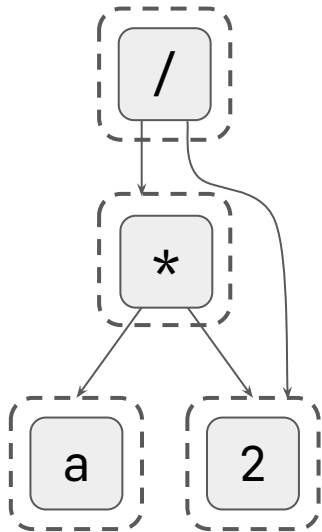
Equivalence Graphs (e-graphs)

$(a * 2) / 2$

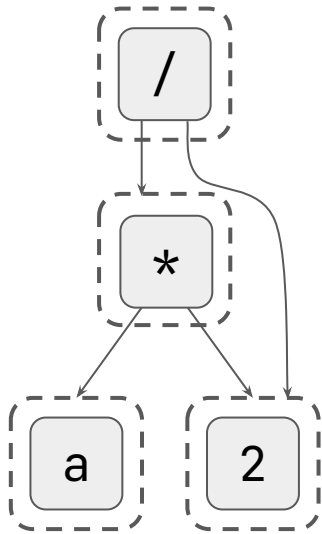


Equivalence Graphs (e-graphs)

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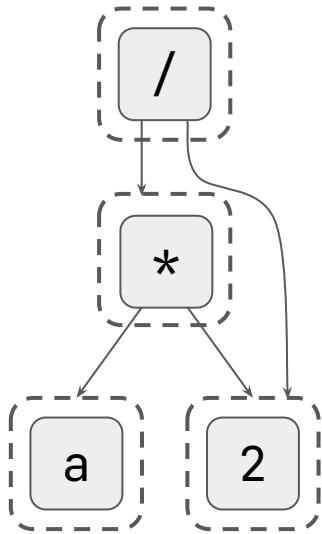


Equality Saturation



$$x * 2 \implies x \ll 1$$

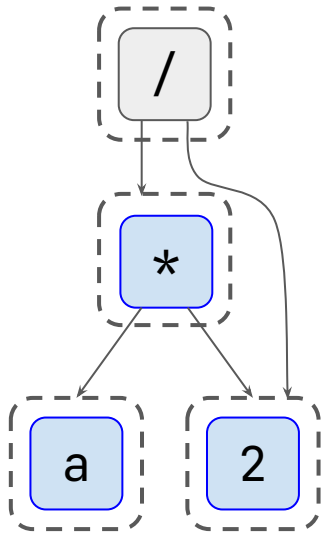
Equality Saturation



$x * 2 \implies x \ll 1$

Find a term that looks like the left,
Add a term that looks like the right,
And mark them equivalent

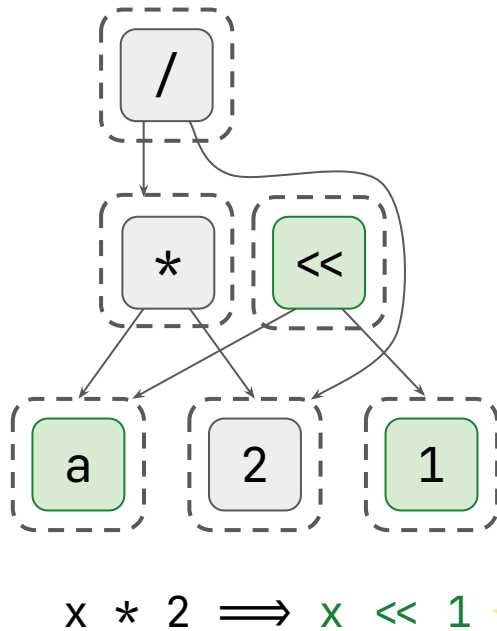
Equality Saturation



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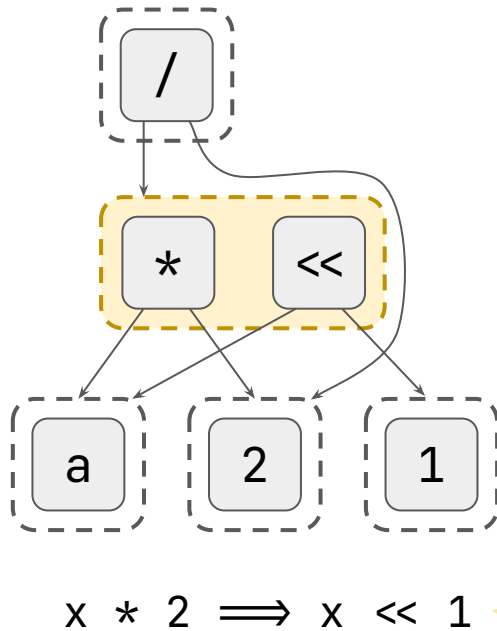
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Equality Saturation



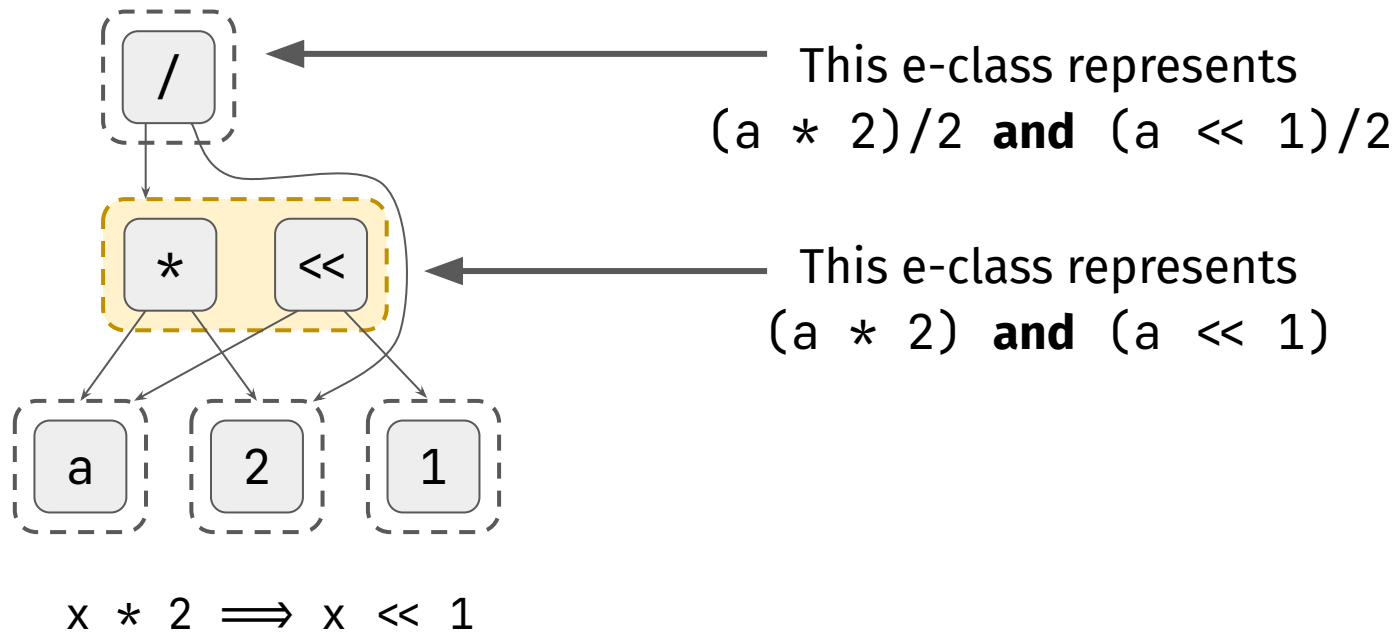
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Equality Saturation

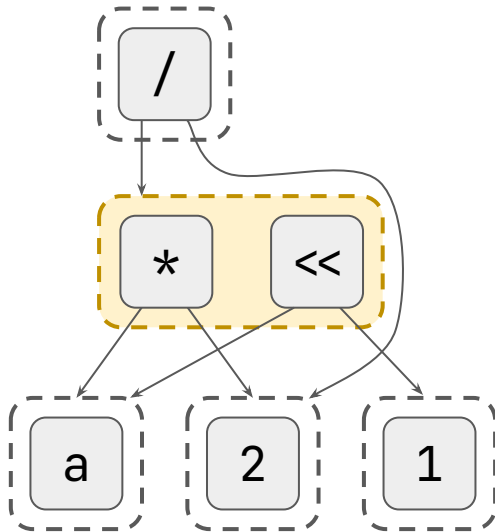


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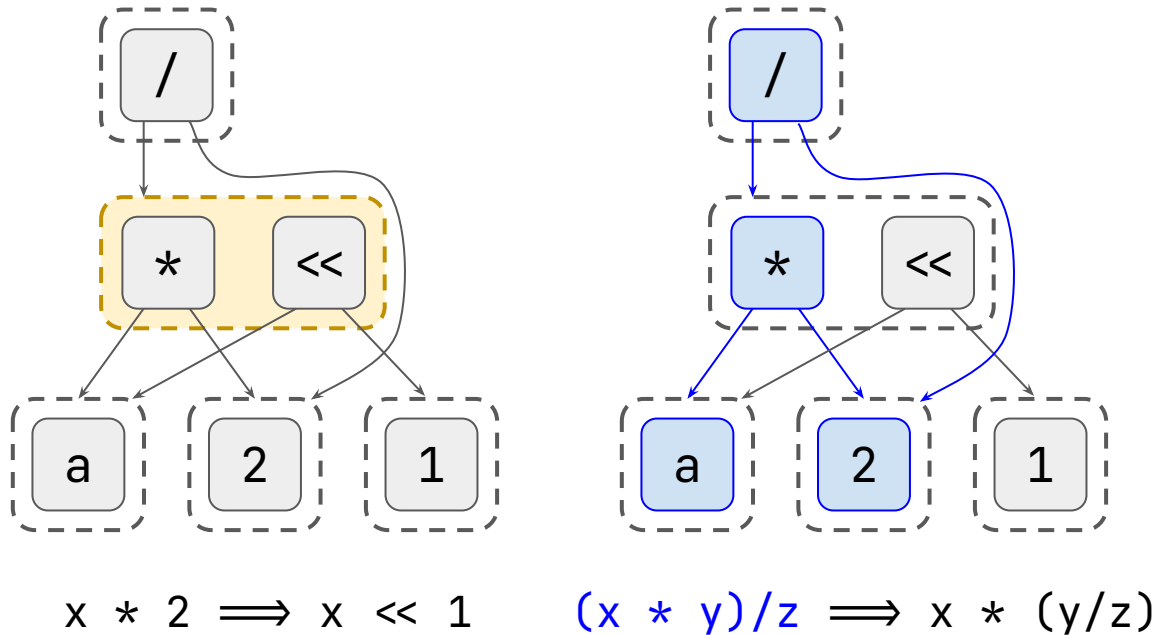


Equality Saturation

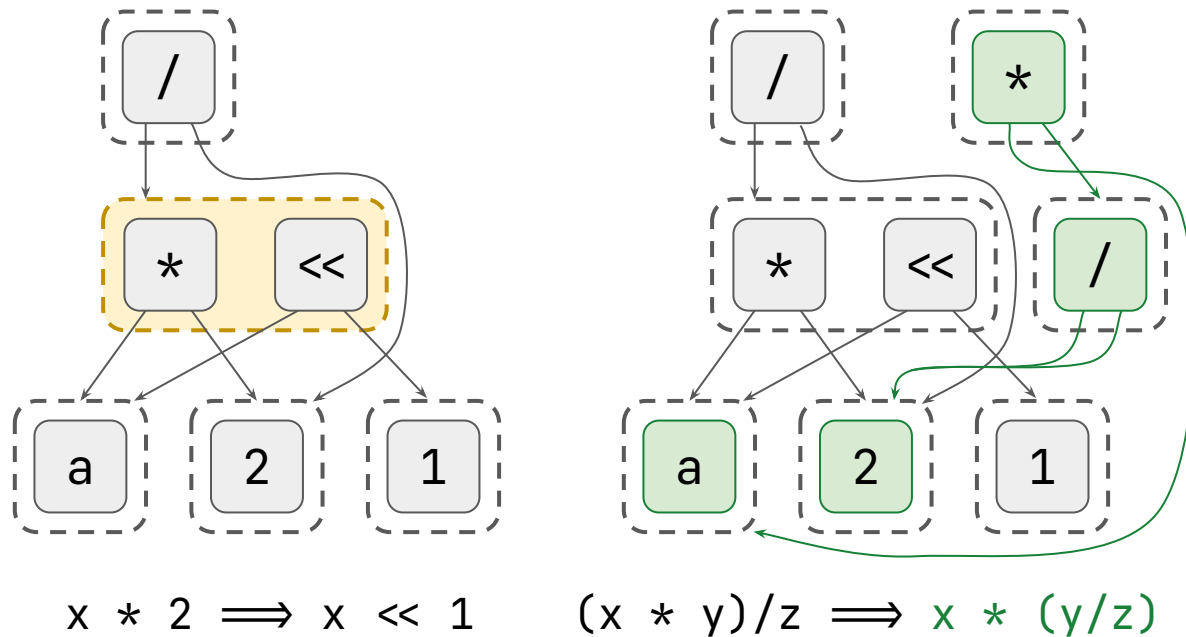


$$x * 2 \implies x \ll 1 \quad (x * y) / z \implies x * (y / z)$$

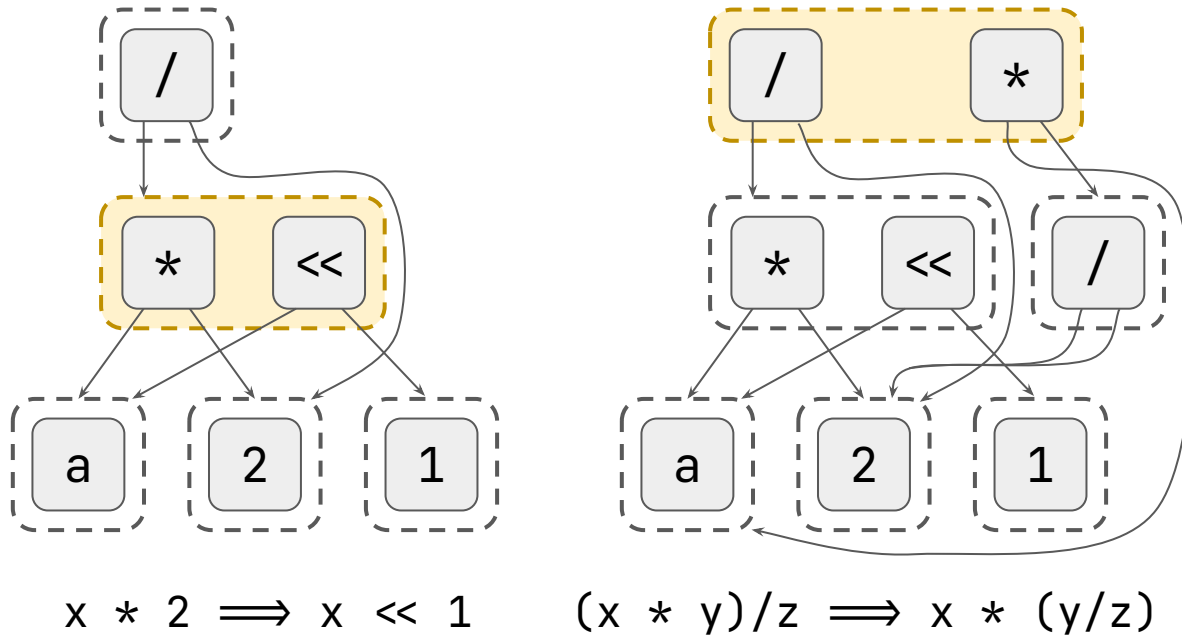
Equality Saturation



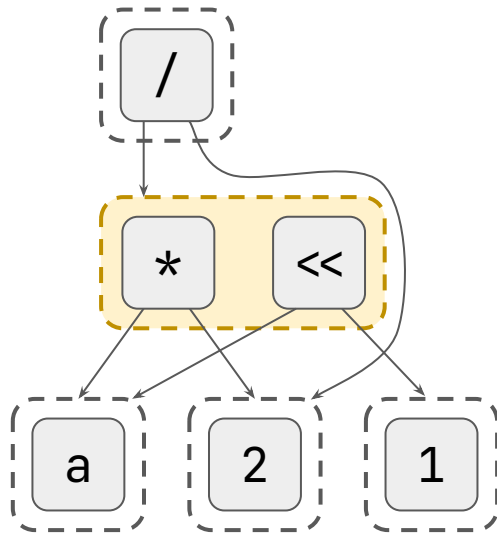
Equality Saturation



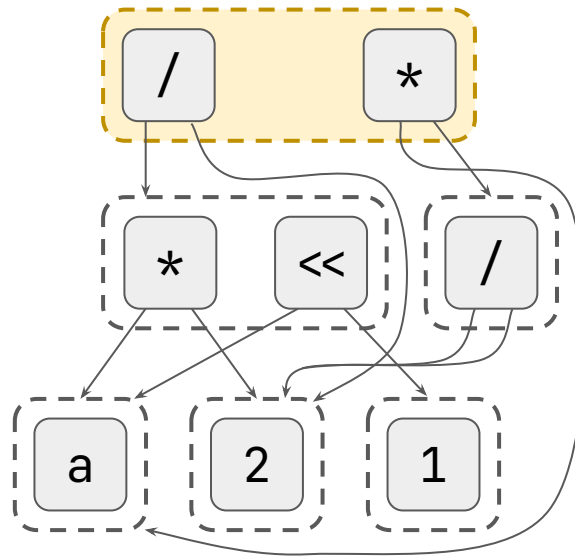
Equality Saturation



Equality Saturation



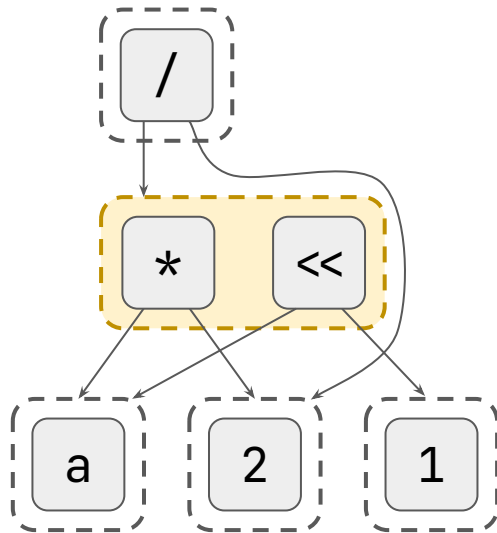
$$x * 2 \implies x \ll 1$$



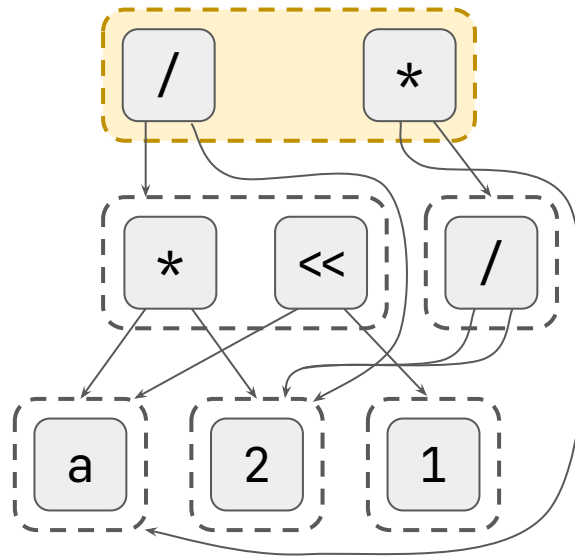
$$(x * y) / z \implies x * (y / z)$$

$$\begin{aligned} x / x &\implies 1 \\ x * 1 &\implies x \end{aligned}$$

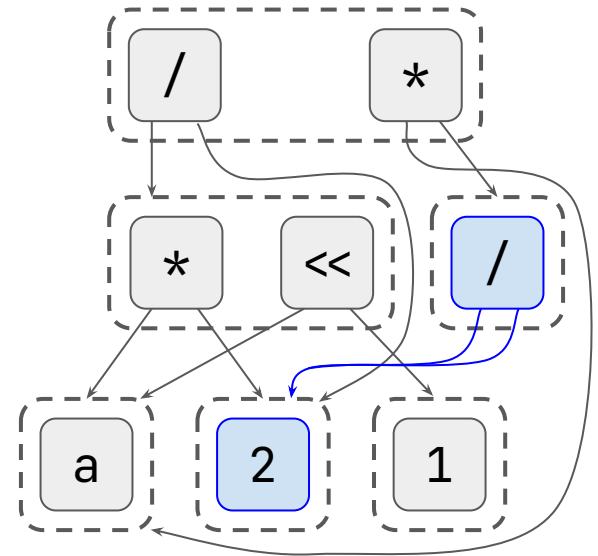
Equality Saturation



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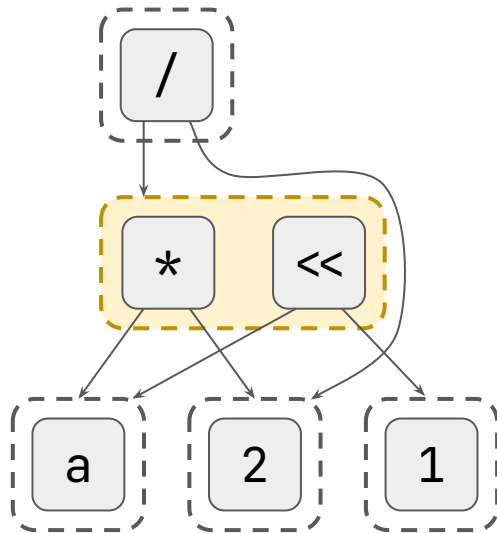


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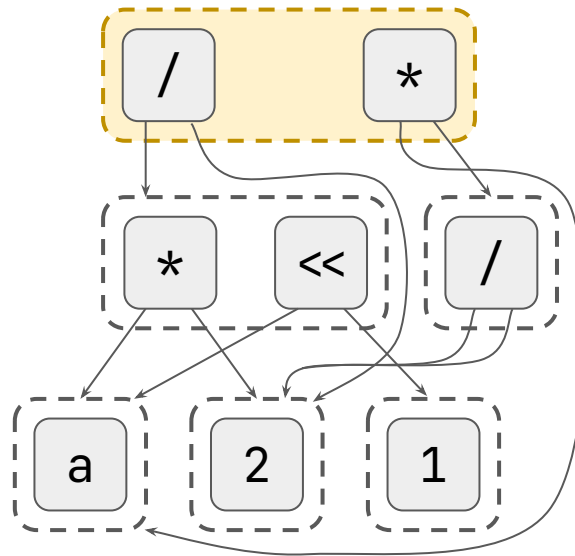


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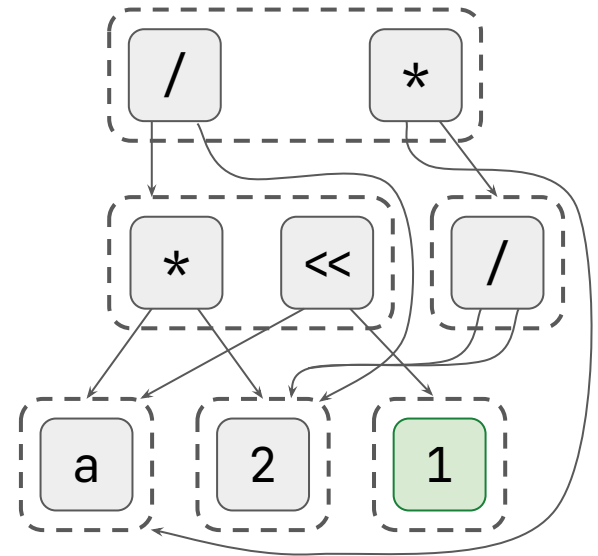
Equality Saturation



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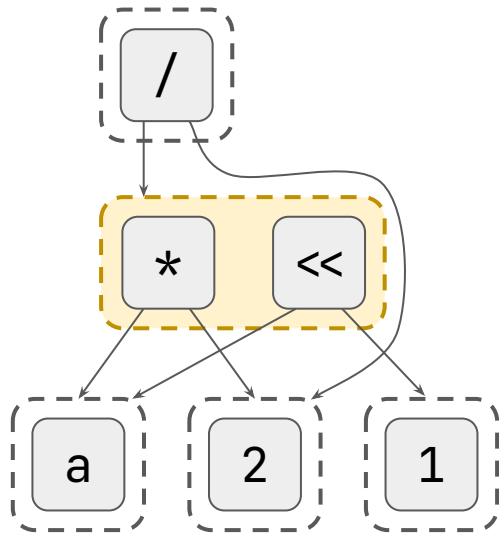


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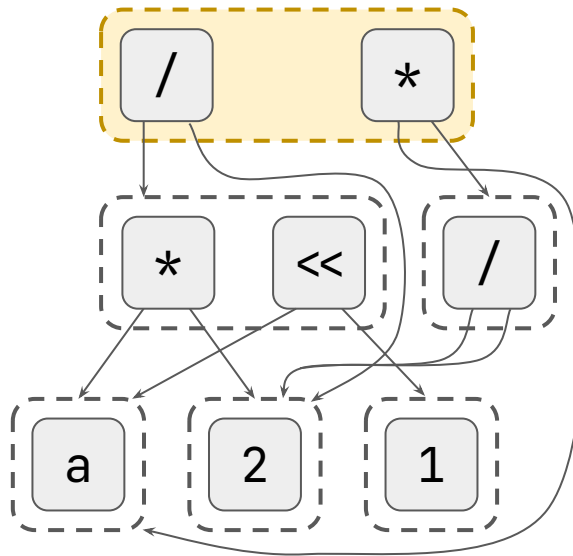


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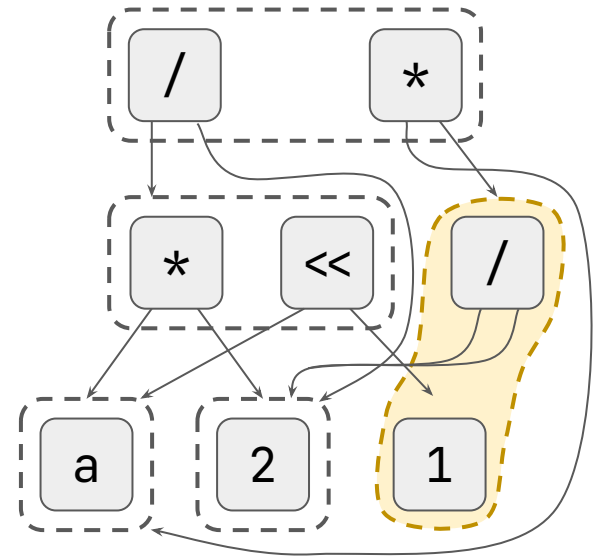
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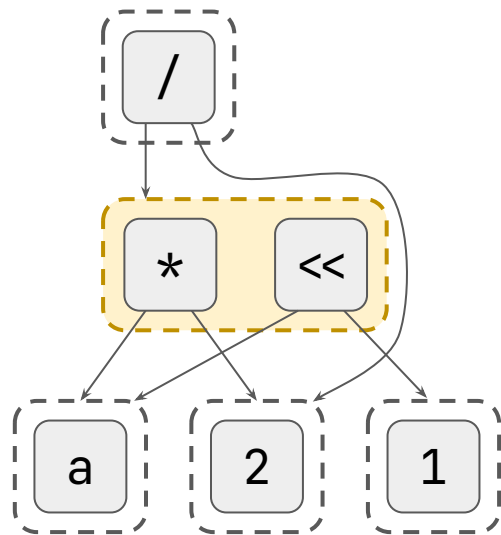


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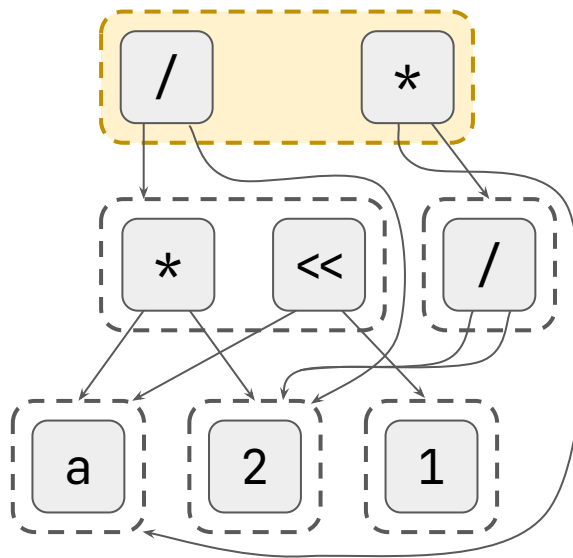


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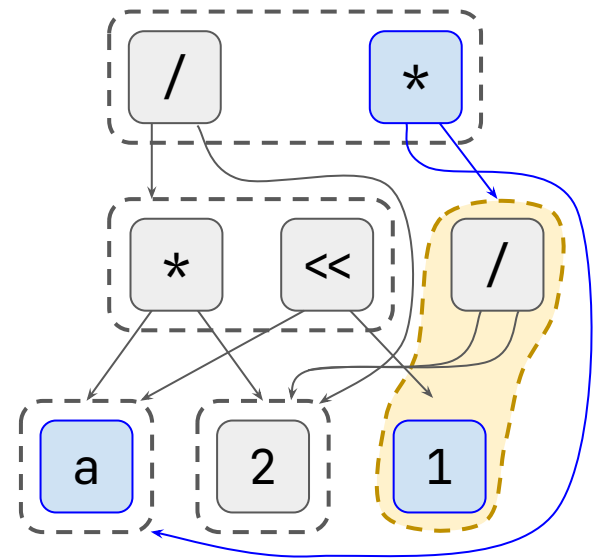
Equality Saturation



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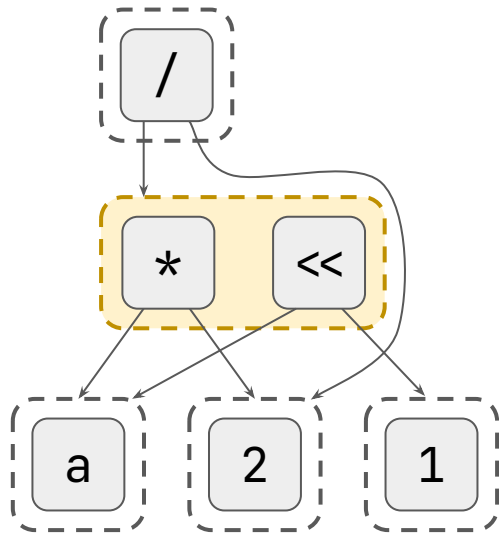
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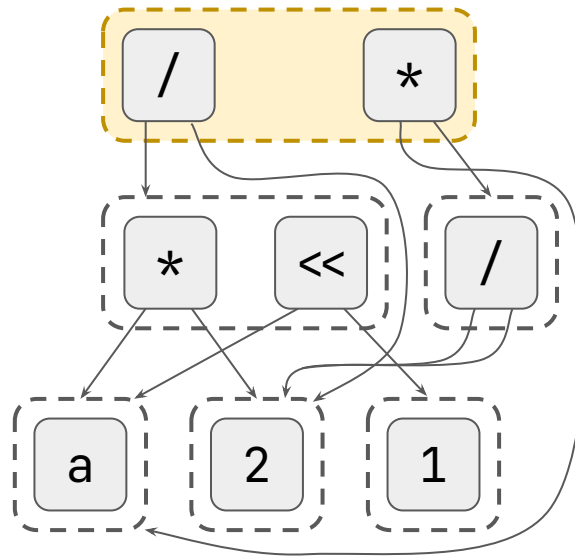
$$x / x \implies 1$$

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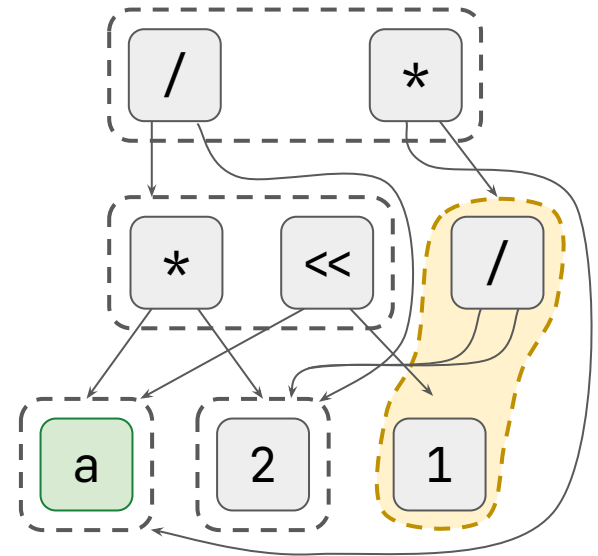
Equality Saturation



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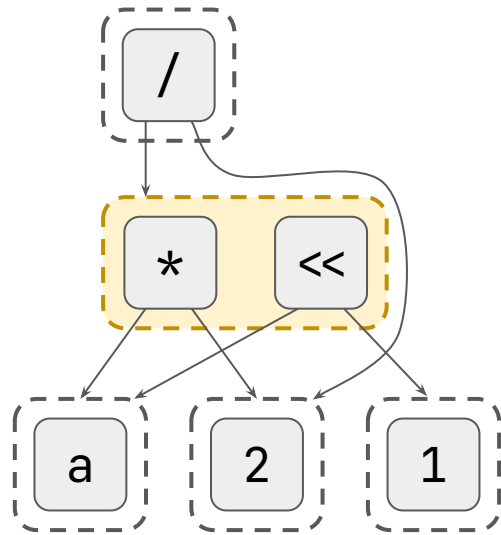


$$(x * y) / z \implies x * (y / z)$$

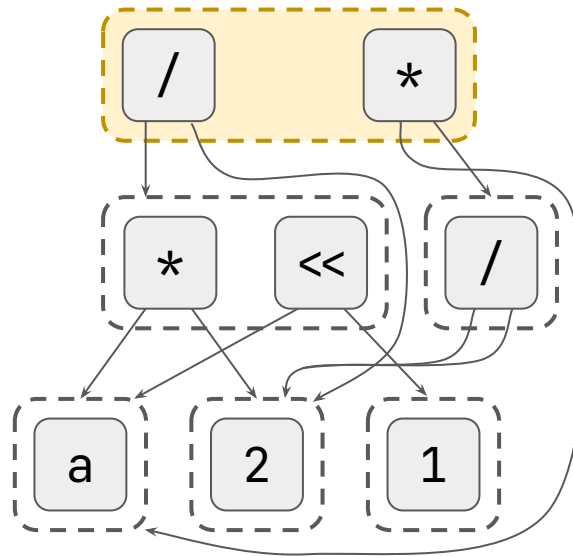


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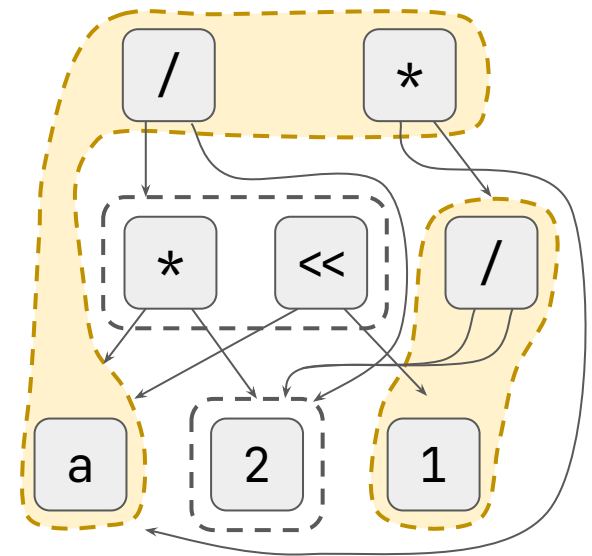
Equality Saturation



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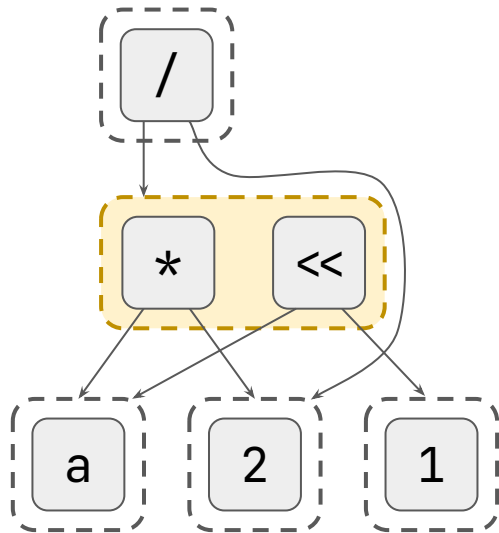


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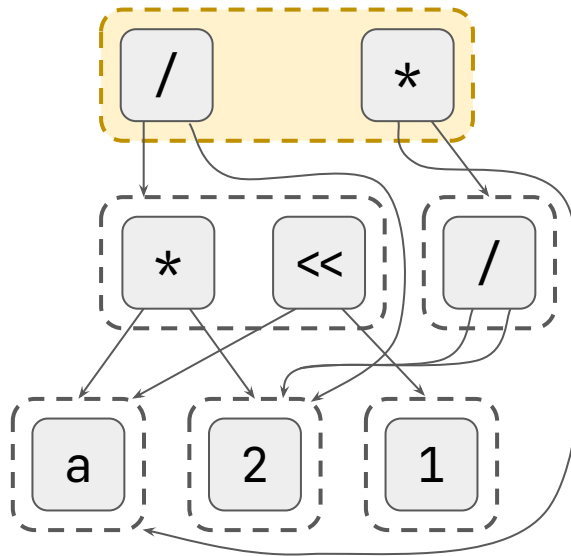


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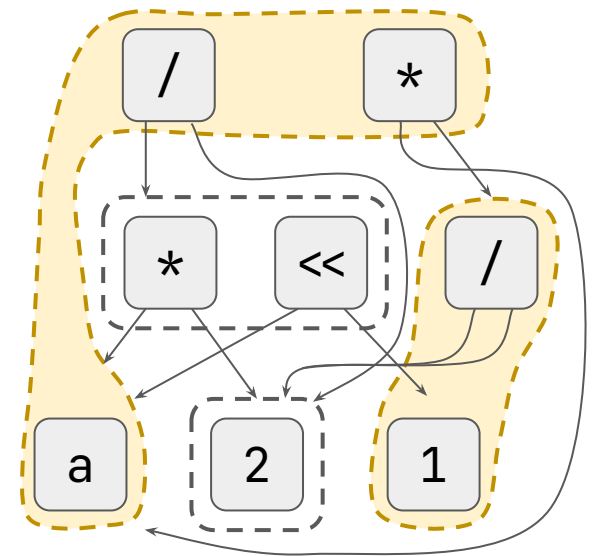
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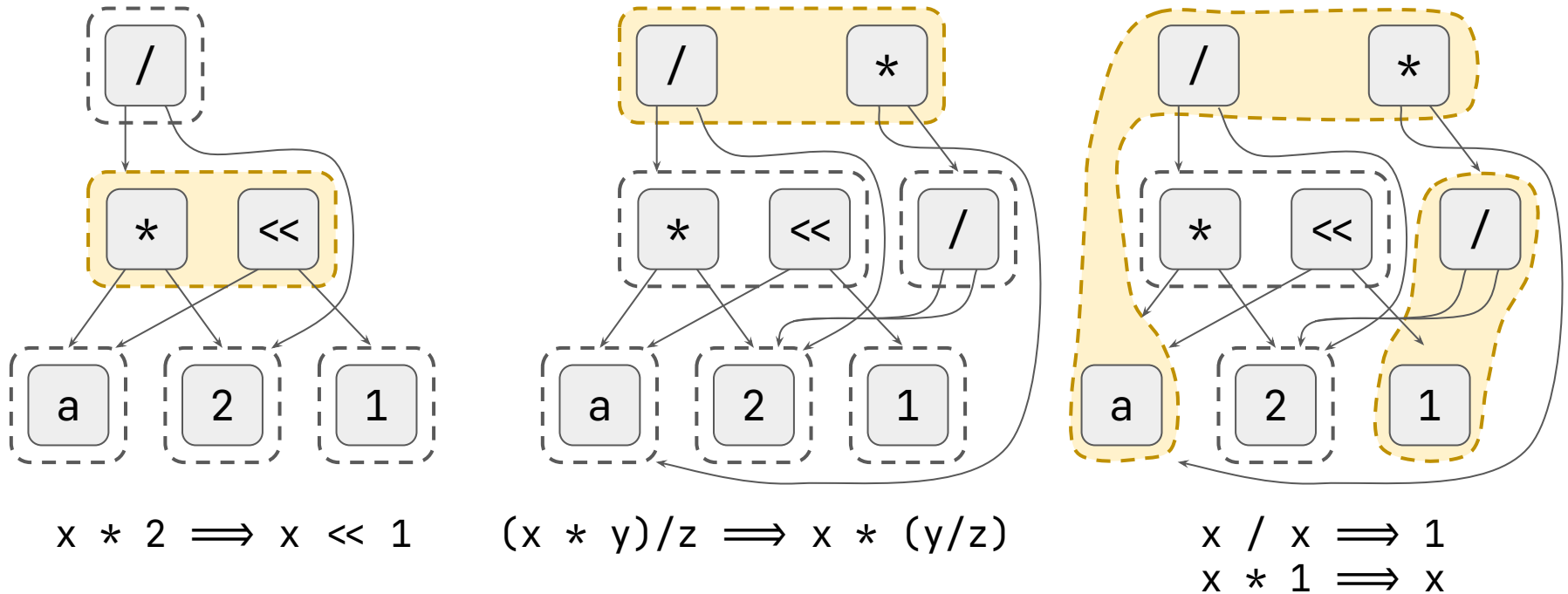


$$(x * y) / z \implies x * (y / z)$$



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Equality Saturation



Keep going until saturation or timeout

Equality Saturation is everywhere!

Automatic End-to-End Joint Optimization for

Equality Saturation for Datapath Synthesis:

A Pathway to Pareto Optimality

Automatic Generation of Vectorizing Compilers for Customizable Digital Signal Processors

Vectorization for Digital Signal Processors via Equality Saturation

Automating Constraint-Aware Datapath Optimization using E-Graphs

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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 Exploration

Grammar

Interpreter

Validator

Term
Enumeration

Candidate
Generation

Rule
Selection

Ruleset

Theory Exploration Inputs

Grammar

```
EXPR :=  
| Num(n)  
| Var(v)  
| Add(EXPR, EXPR)
```

Interpreter

```
def eval(expr):  
    match expr  
    | Num(n) => n  
    | Var(v) => lookup(v)  
    | Add(e1, e2) =>  
        eval(e1) + eval(e2)
```

Validator

```
def is_valid(lhs, rhs):  
    l = lhs.to_z3()  
    r = rhs.to_z3()  
    z3.assert(l.eq(r).not())  
    return  
        z3.check() == Unsat
```

What terms in the language **look like**

Num(5)

Add(Num(1), Num(2))

Add(Add(Var("x"), Num(1)), Num(2))

Theory Exploration Inputs

Grammar

```
EXPR :=  
| Num(n)  
| Var(v)  
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```

Interpreter

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```

What terms in the language **mean**

$\text{eval}(\text{Num}(5)) = 5$

$\text{eval}(\text{Add}(\text{Num}(1), \text{Num}(2))) = 3$

$\text{eval}(\text{Add}(\text{Num}(1), \text{Add}(\text{Num}(2), \text{Num}(3)))) = 6$

Theory Exploration Inputs

Grammar

```
EXPR :=  
| Num(n)  
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```
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```

Whether a candidate rewrite rule is **correct**

Add(x, y) \Rightarrow Add(y, x) 


Add(x, Num(1)) \Rightarrow x 

Add(x, Add(y, z)) \Rightarrow Add(Add(x, y), z)

Term Enumeration

Grammar

```
EXPR :=  
| Num(n)  
| Var(v)  
| Add(EXPR, EXPR)
```


Term Enumeration

Grammar

```
EXPR :=  
| Num(n)  
| Var(v)  
| Add(EXPR, EXPR)
```

a

b

0

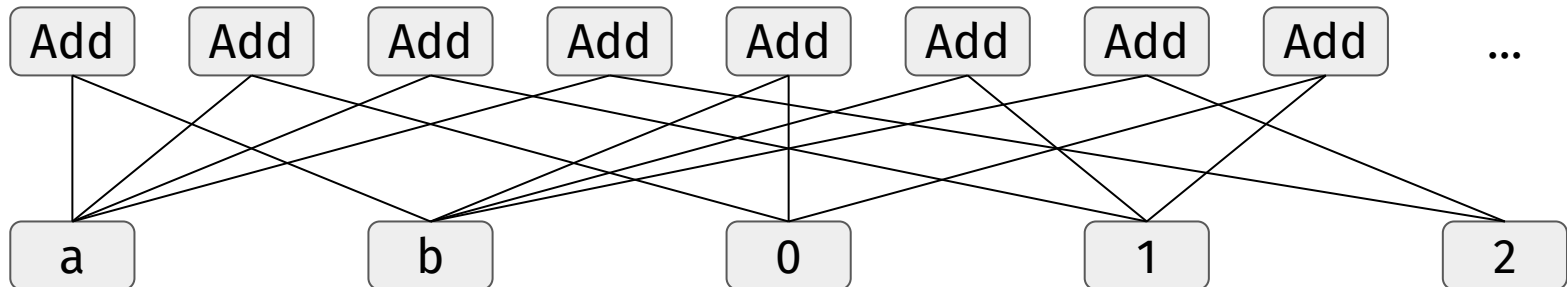
1

2

Term Enumeration

Grammar

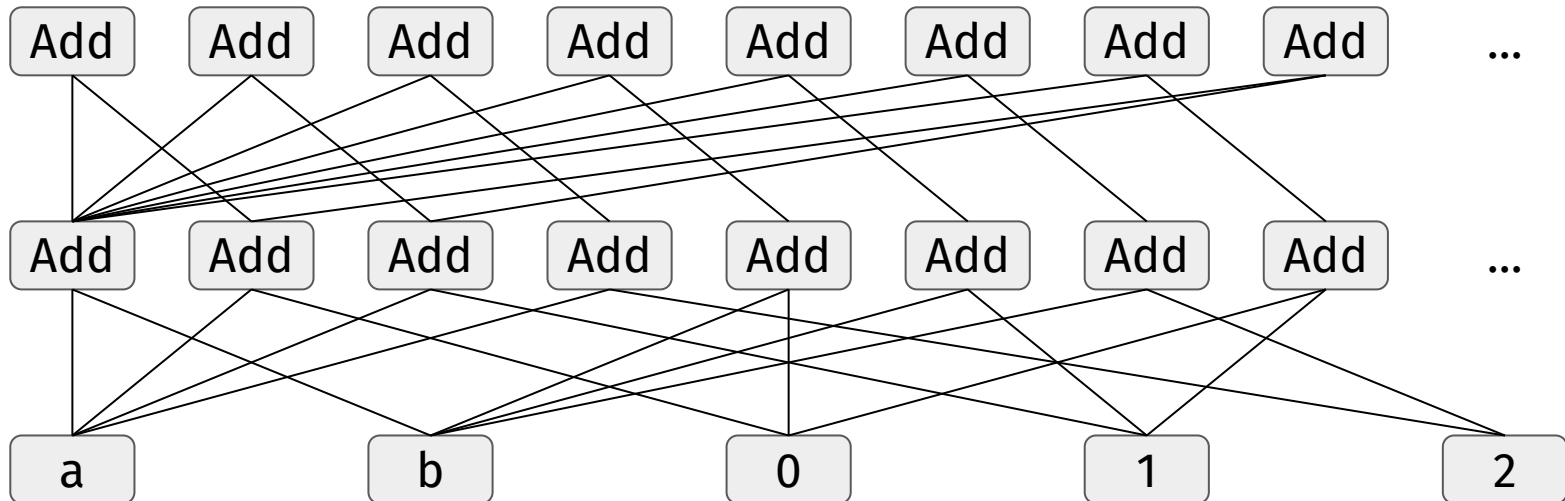
```
EXPR :=  
| Num(n)  
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```



Term Enumeration

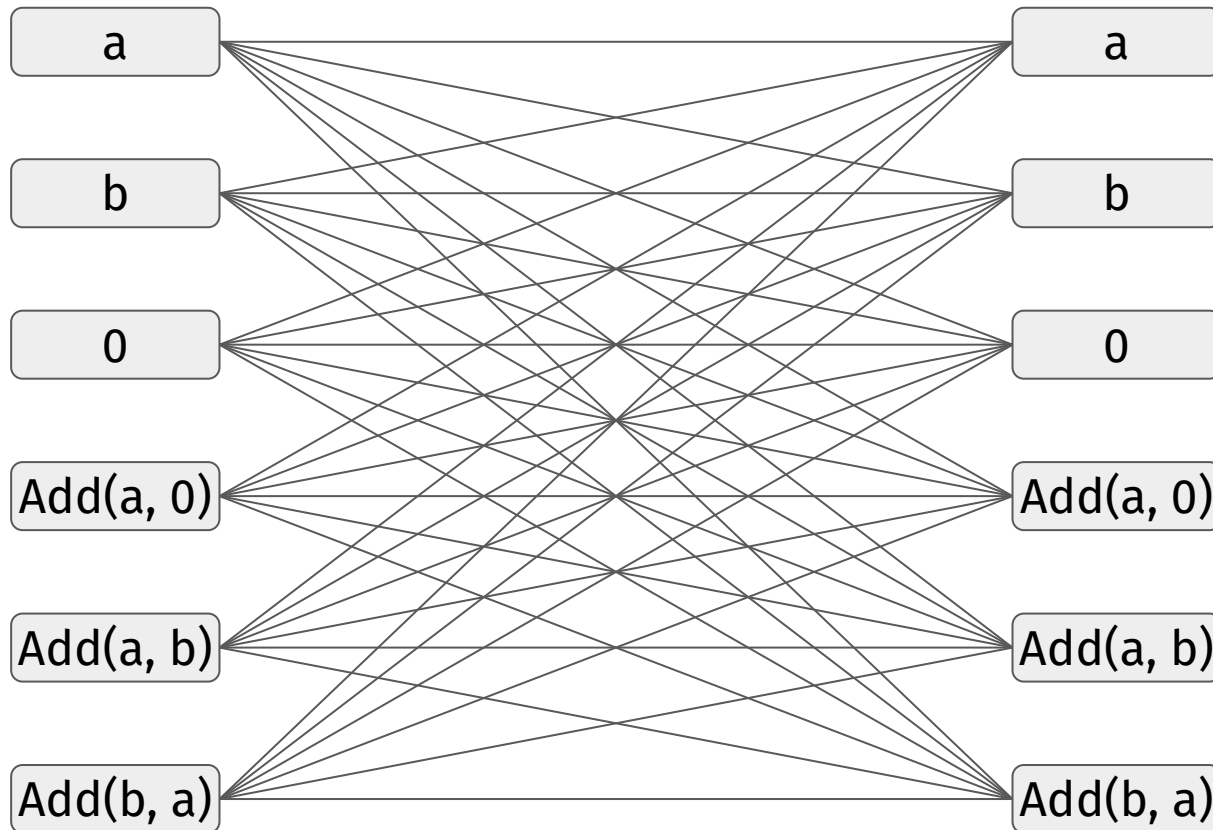
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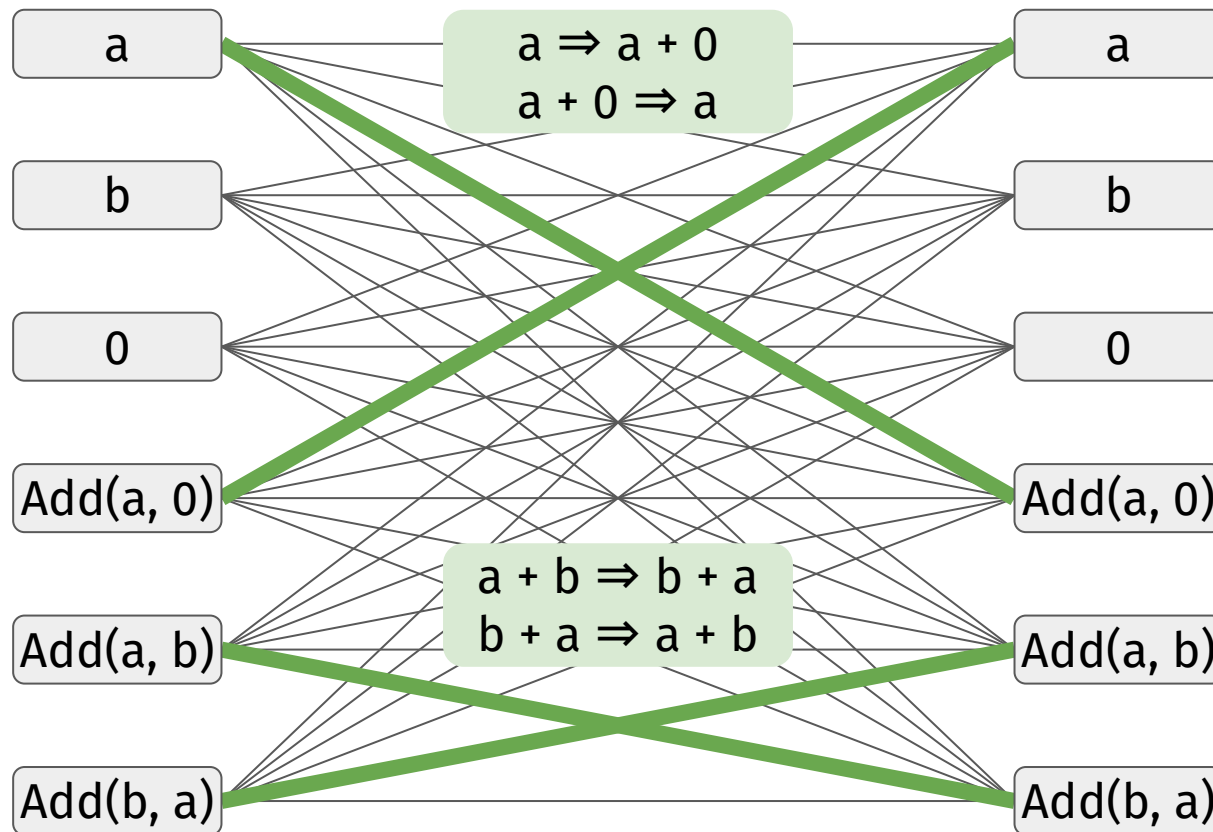


Candidate Generation

Candidate Generation



Candidate Generation

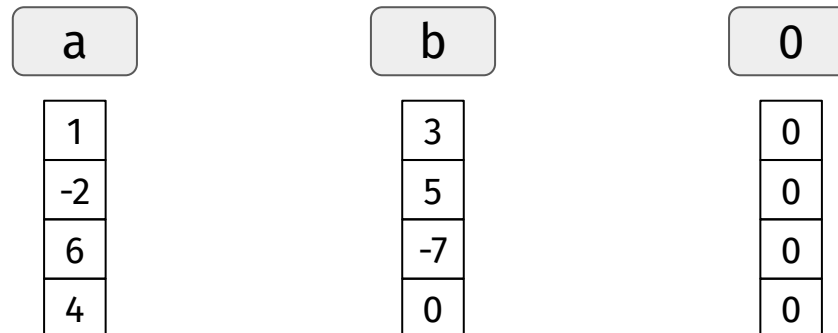


Candidate Generation

Interpreter

```
def eval(expr):  
    match expr  
    | ...
```

Tag each e-class with an array of possible values



Candidate Generation

Interpreter

```
def eval(expr):  
    match expr  
    | ...
```

Tag each e-class with an array of possible values

a

1

-2

6

4

b

3

5

-7

0

0

0

0

0

Sample values from
the domain for
variables

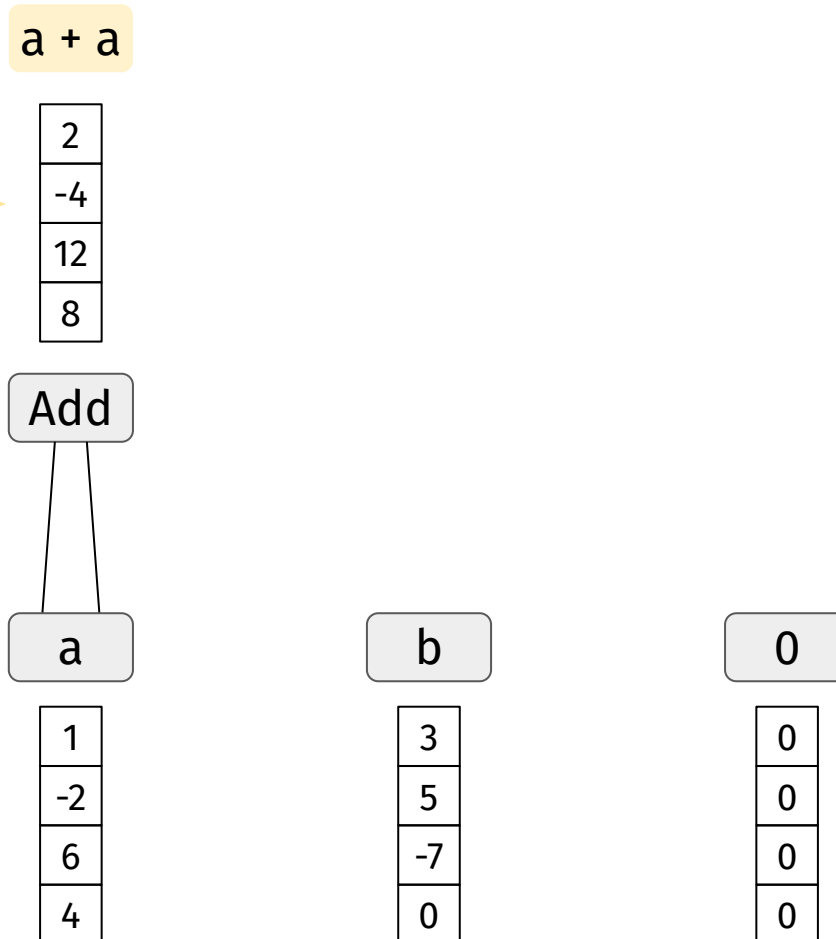
Constants always
have the same value

Candidate Generation

Interpreter

```
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    | ...
```

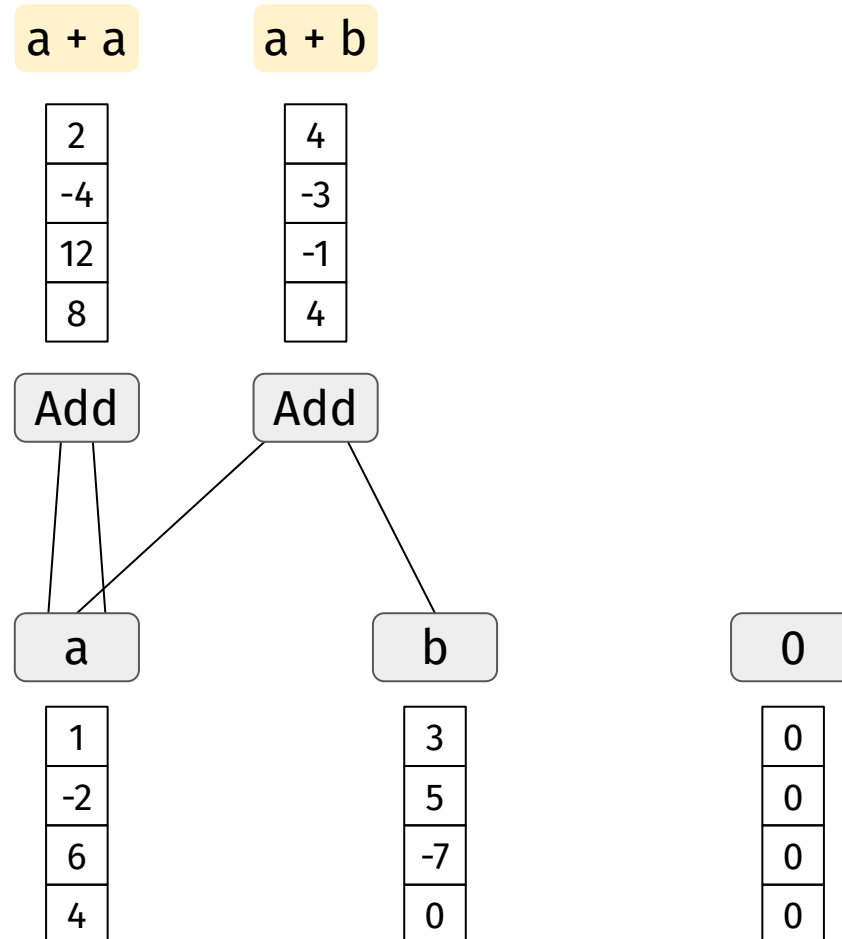
Compute values
from children



Candidate Generation

Interpreter

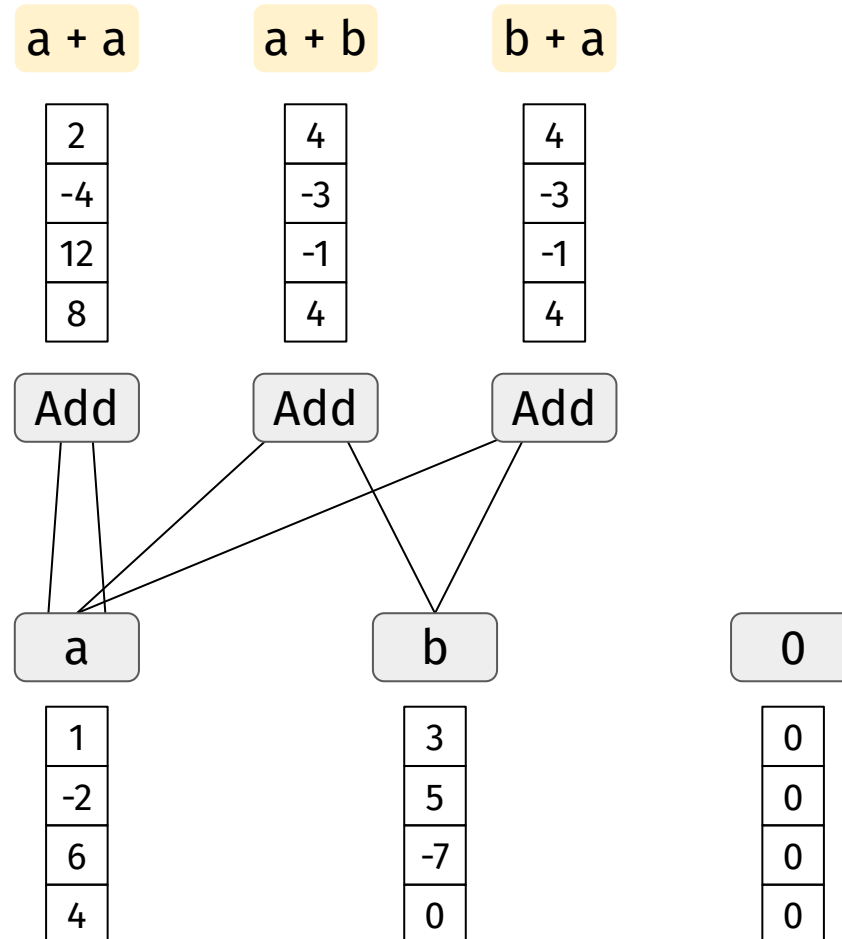
```
def eval(expr):  
    match expr  
    | ...
```



Candidate Generation

Interpreter

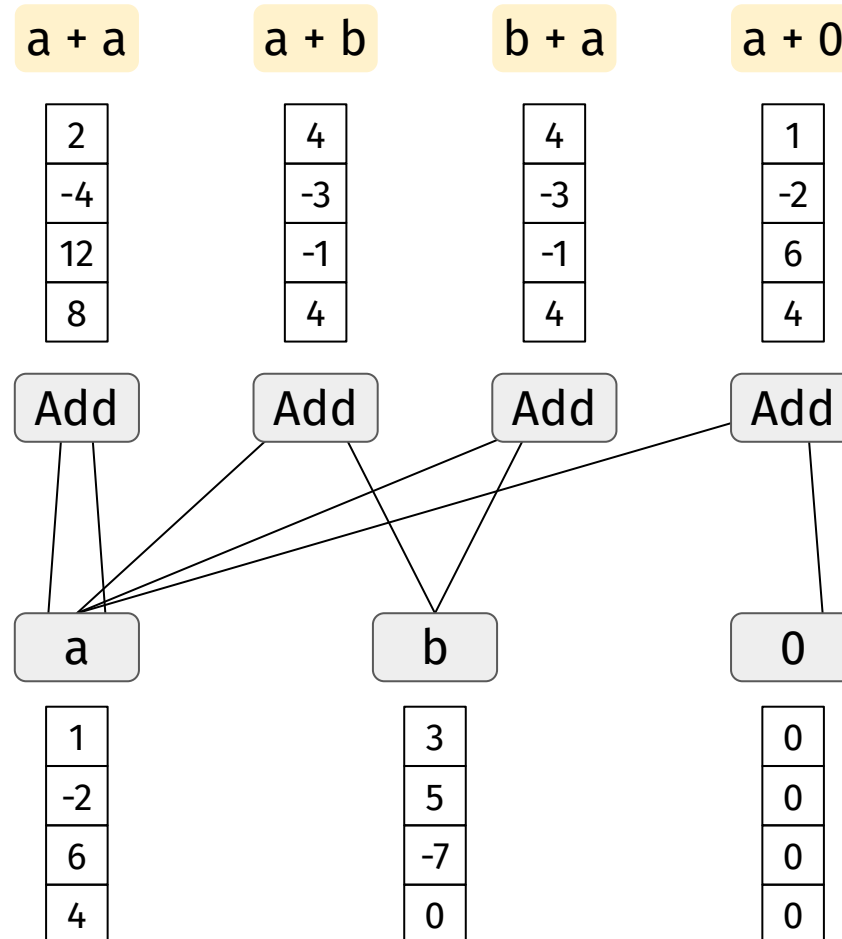
```
def eval(expr):  
    match expr  
    | ...
```



Candidate Generation

Interpreter

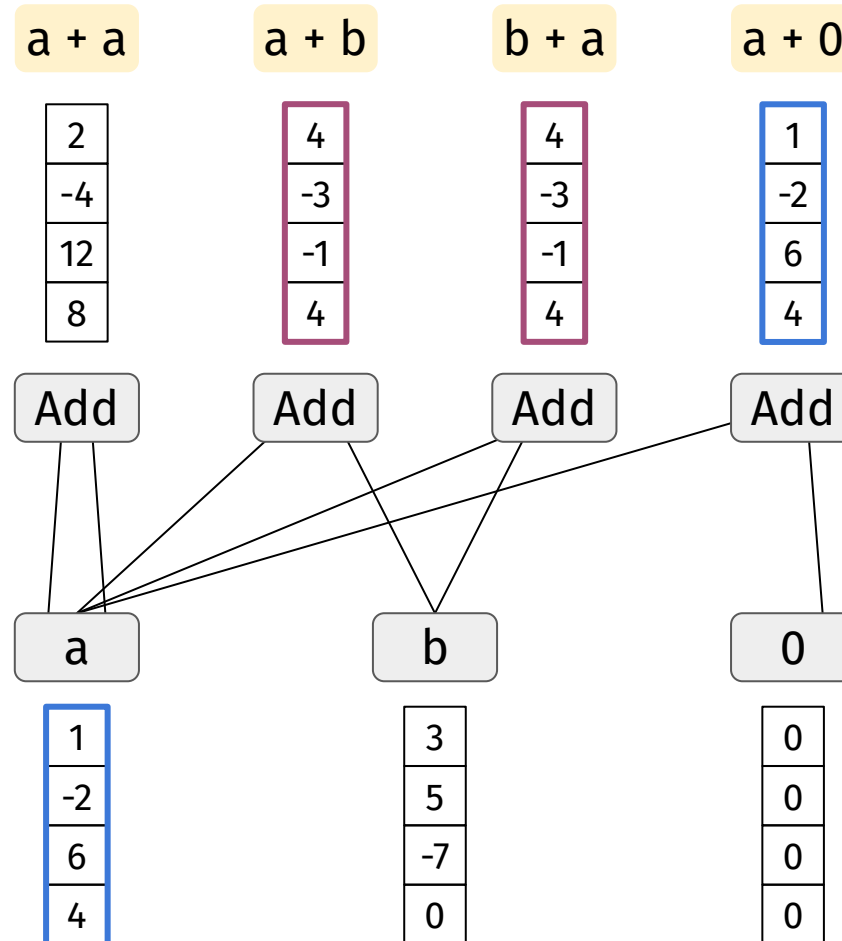
```
def eval(expr):  
    match expr  
    | ...
```



Candidate Generation

Interpreter

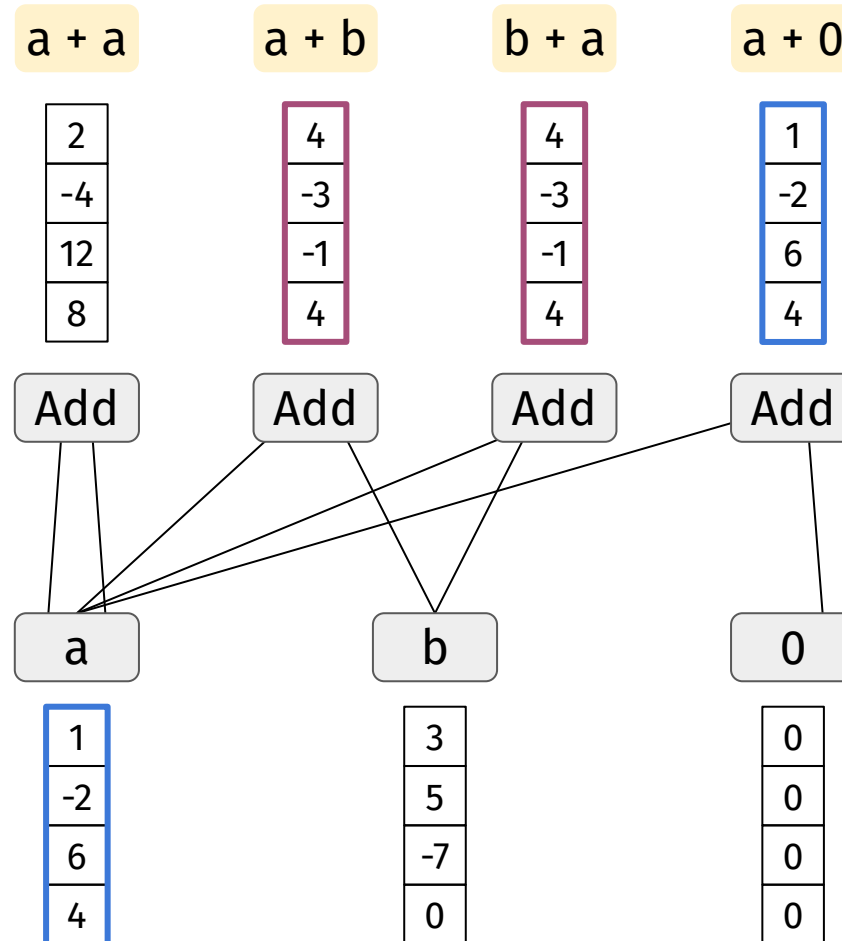
```
def eval(expr):  
    match expr  
    | ...
```



Candidate Generation

Interpreter

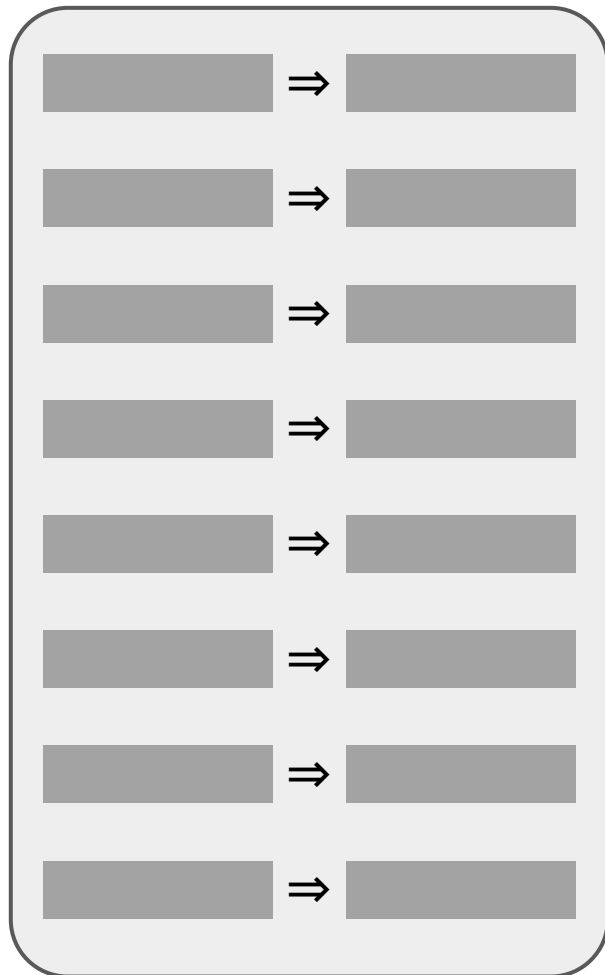
```
def eval(expr):
    match expr
    | ...
```



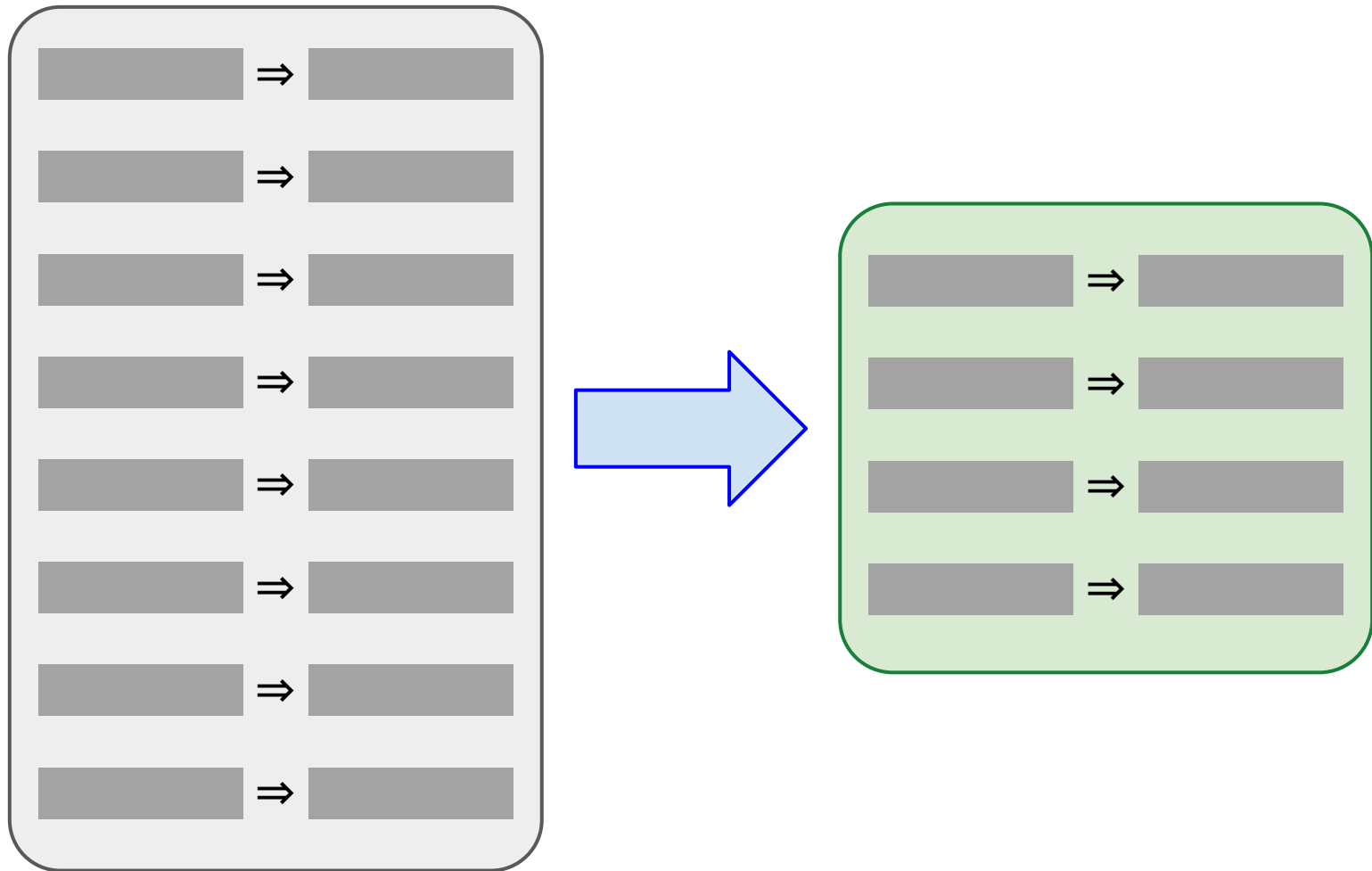
$a + b \Rightarrow b + a$
 $b + a \Rightarrow a + b$

$a + 0 \Rightarrow a$
 $a \Rightarrow a + 0$

Rule Selection



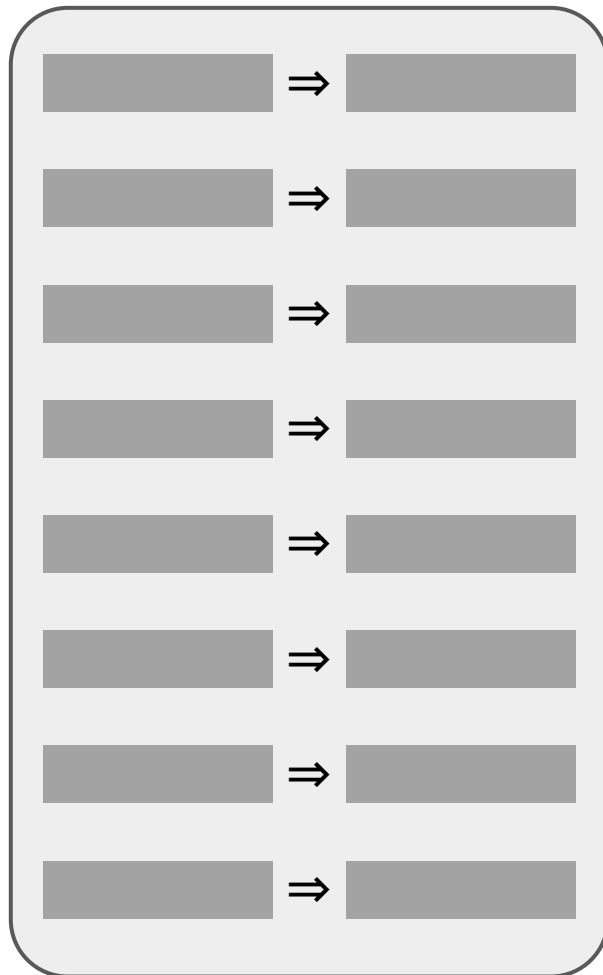
Rule Selection



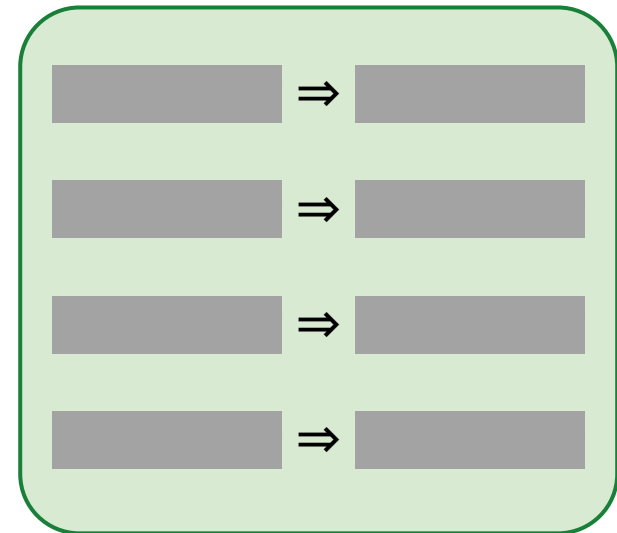
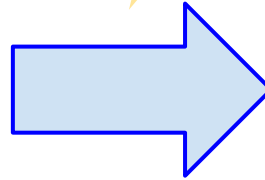
Rule Selection

Validator

```
def is_valid(l, r):  
    ...
```



Filter out unsound rule candidates with validator



Pick the best rules using heuristics and equality saturation to eliminate redundant rules

Rule Selection

Candidates:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$

$$x + 0 \Rightarrow 0 + x$$

$$y + 0 \Rightarrow 0 + y$$

$$x * 1 \Rightarrow 1 * x$$

$$y * 1 \Rightarrow 1 * y$$

Chosen Rules:

Sort rule candidates using heuristics:
More general is better

Rule Selection

Candidates:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$

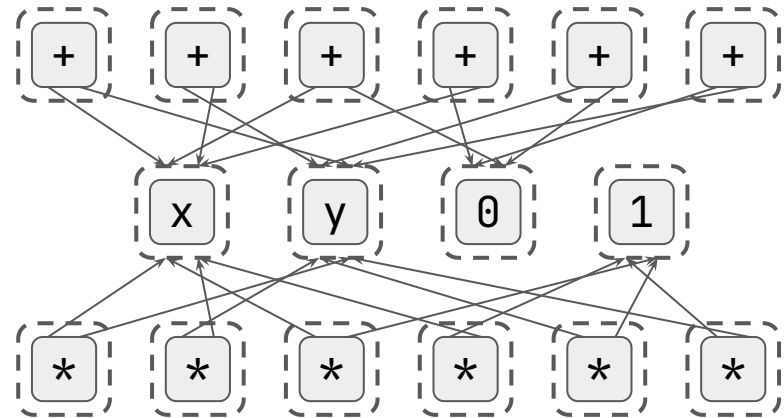
$$x + 0 \Rightarrow 0 + x$$

$$y + 0 \Rightarrow 0 + y$$

$$x * 1 \Rightarrow 1 * x$$

$$y * 1 \Rightarrow 1 * y$$

Chosen Rules:



Initialize e-graph with all candidates

Rule Selection

Candidates:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$

$$x + 0 \Rightarrow 0 + x$$

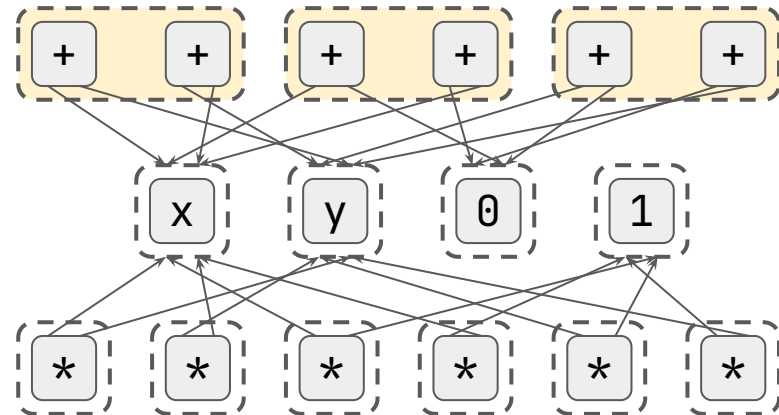
$$y + 0 \Rightarrow 0 + y$$

$$x * 1 \Rightarrow 1 * x$$

$$y * 1 \Rightarrow 1 * y$$

Chosen Rules:

$$x + y \Rightarrow y + x$$



Pick a rule; Run equality saturation

Rule Selection

Candidates:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$

~~$$x + 0 \Rightarrow 0 + x$$~~

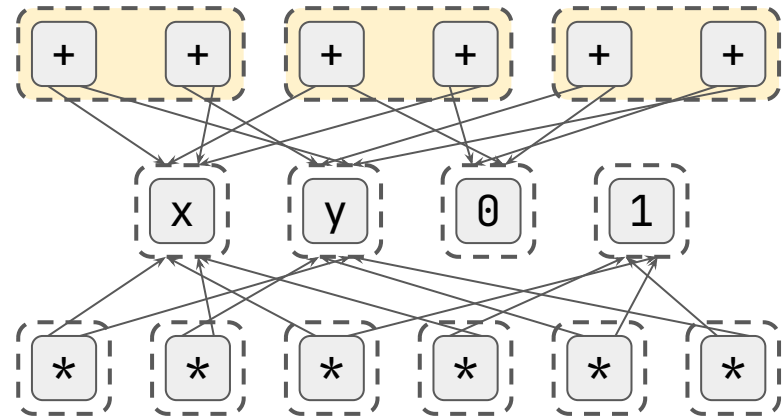
~~$$y + 0 \Rightarrow 0 + y$$~~

$$x * 1 \Rightarrow 1 * x$$

$$y * 1 \Rightarrow 1 * y$$

Chosen Rules:

$$x + y \Rightarrow y + x$$



Eliminate redundant candidates

Rule Selection

Candidates:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$

~~$$x + 0 \Rightarrow 0 + x$$~~

~~$$y + 0 \Rightarrow 0 + y$$~~

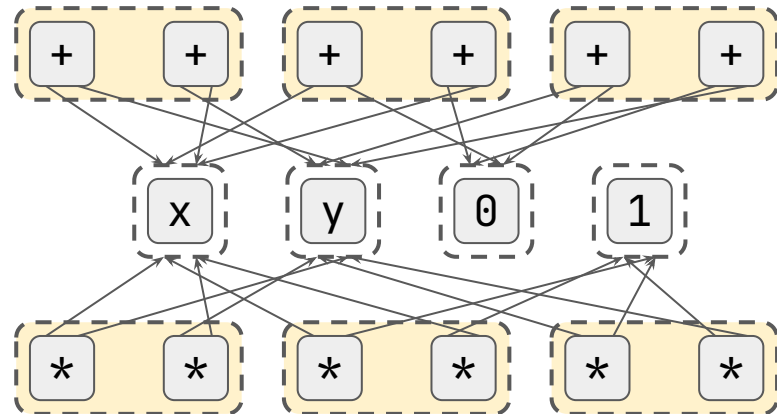
$$x * 1 \Rightarrow 1 * x$$

$$y * 1 \Rightarrow 1 * y$$

Chosen Rules:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$



Pick a rule; Run equality saturation

Rule Selection

Candidates:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$

~~$$x + 0 \Rightarrow 0 + x$$~~

~~$$y + 0 \Rightarrow 0 + y$$~~

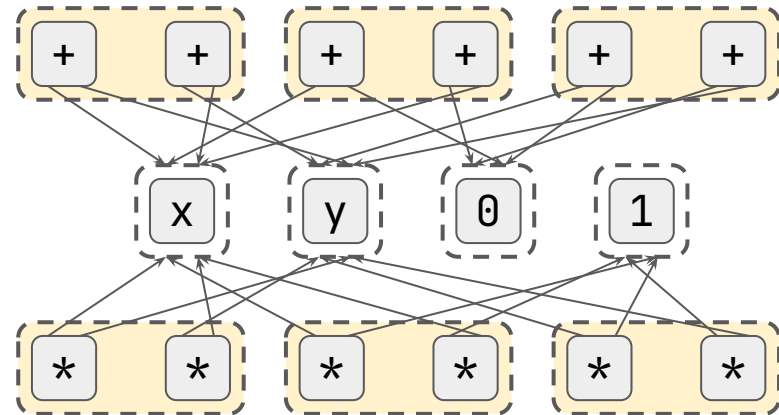
~~$$x * 1 \Rightarrow 1 * x$$~~

~~$$y * 1 \Rightarrow 1 * y$$~~

Chosen Rules:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$



Eliminate redundant candidates

Rule Selection

Candidates:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$

~~$$x + 0 \Rightarrow 0 + x$$~~

~~$$y + 0 \Rightarrow 0 + y$$~~

~~$$x * 1 \Rightarrow 1 * x$$~~

~~$$y * 1 \Rightarrow 1 * y$$~~

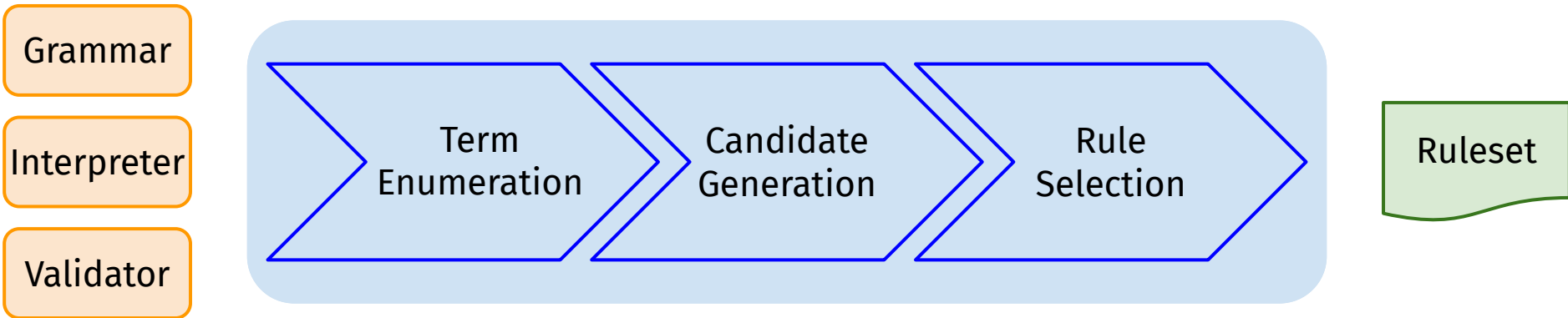
Final Ruleset:

$$x + y \Rightarrow y + x$$

$$x * y \Rightarrow y * x$$

Repeat until no more candidates

Traditional Theory Exploration



Traditional Theory Exploration



Grammar

Interpreter

Validator

Term
Enumeration

Candidate
Generation

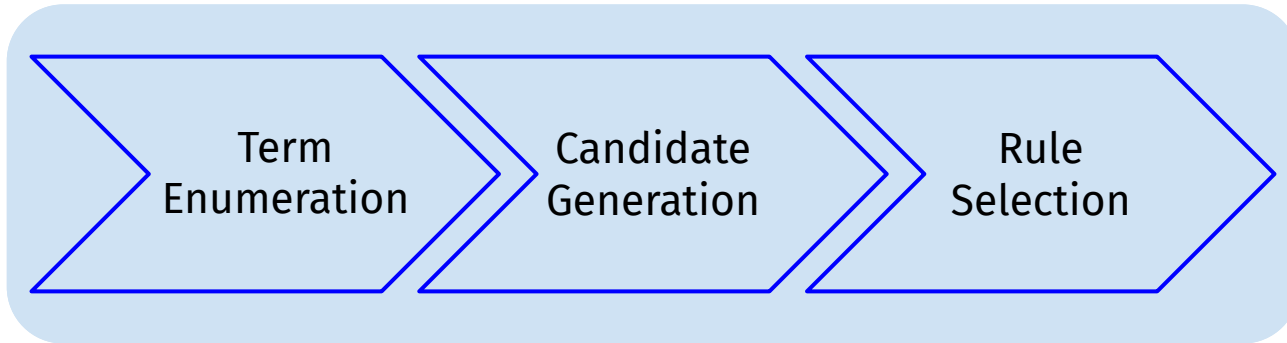
Rule
Selection

Ruleset

Traditional Theory Exploration

One shot!

Grammar
Interpreter
Validator



Ruleset

Not modifiable

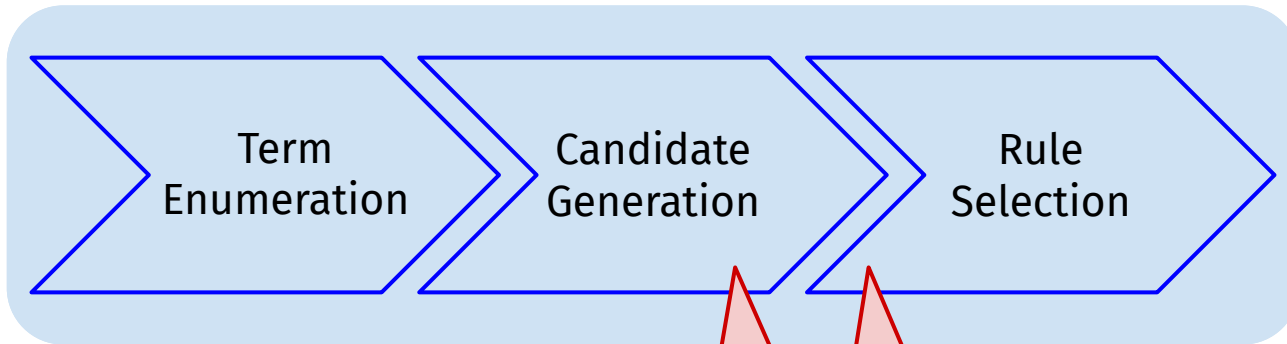
Traditional Theory Exploration

One shot!

Grammar

Interpreter

Validator



Hard-coded
resource limits

Not modifiable

Traditional Theory Exploration

One shot!

Grammar

Interpreter

Validator

Term
Enumeration

Candidate
Generation

Rule
Selection

Ruleset

Exponential
Blowup

Hard-coded
resource limits

Not modifiable

Traditional Theory Exploration

One shot!

Grammar

Interpreter

Validator

Term
Enumeration

Candidate
Generation

Rule
Selection

Ruleset

Impossible for
some domains

Exponential
Blowup

Hard-coded
resource limits

Not modifiable

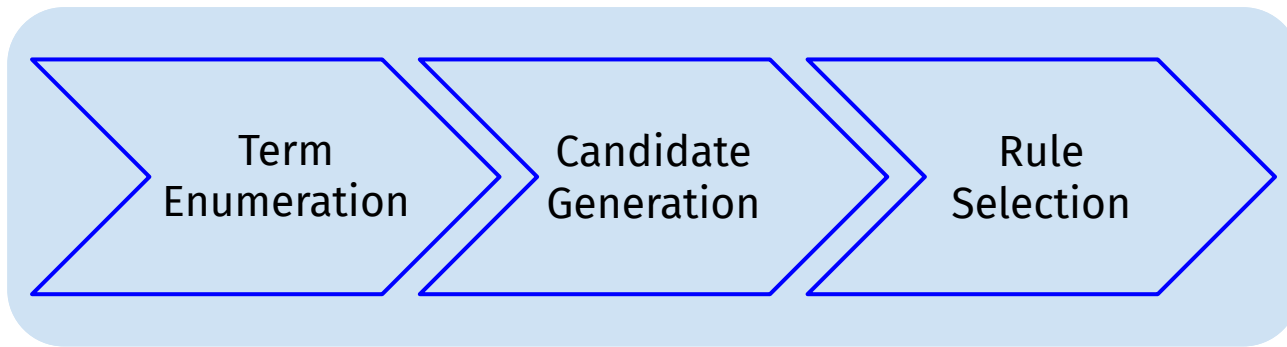
The **ENUMO** DSL:

A more flexible and
extensible approach
to rule inference

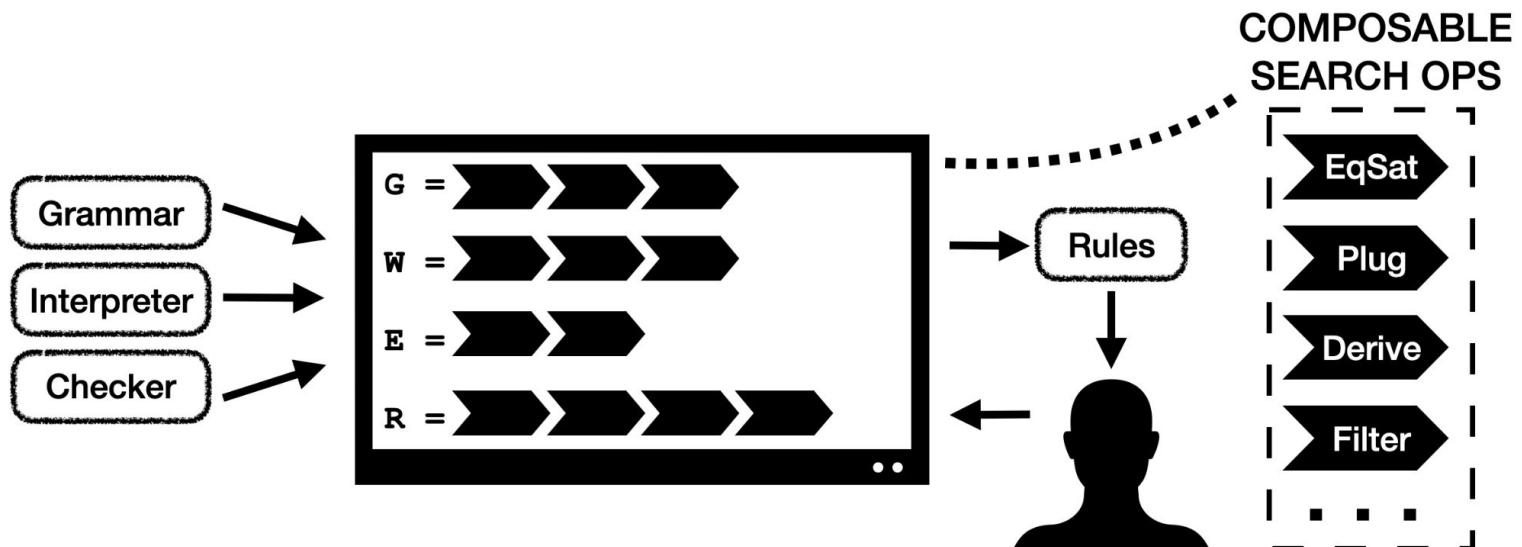
Insight: Users have intuition about which parts of the domain are worth exploring

$(\sim (\sim (\sim (\sim (\sim a))))))$

$(- (* a a) (* b b))$



We enable experts to leverage their expertise by exposing a small set of useful operators



ENUMO DSL

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

ENUMO DSL

```
lits = Workload { a b 0 1 }  
exps = Workload { EXP (~ EXP) (+ EXP EXP) }
```

```
EXP :=  
| Num(n)  
| Var(v)  
| Neg(EXP)  
| Add(EXP, EXP)
```

ENUMO DSL

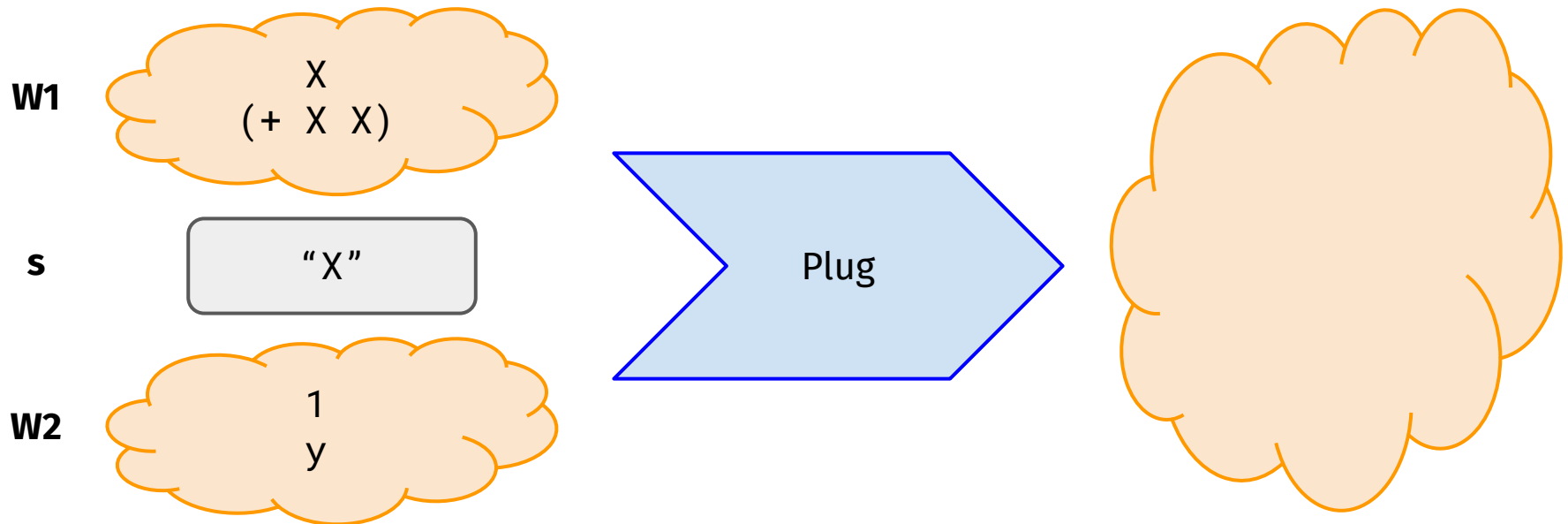
```
lits = Workload { a b 0 1 }  
exps = Workload { EXP (~ EXP) (+ EXP EXP) }  
  
wkld = exps.plugin("EXP", exps)
```

ENUMO DSL

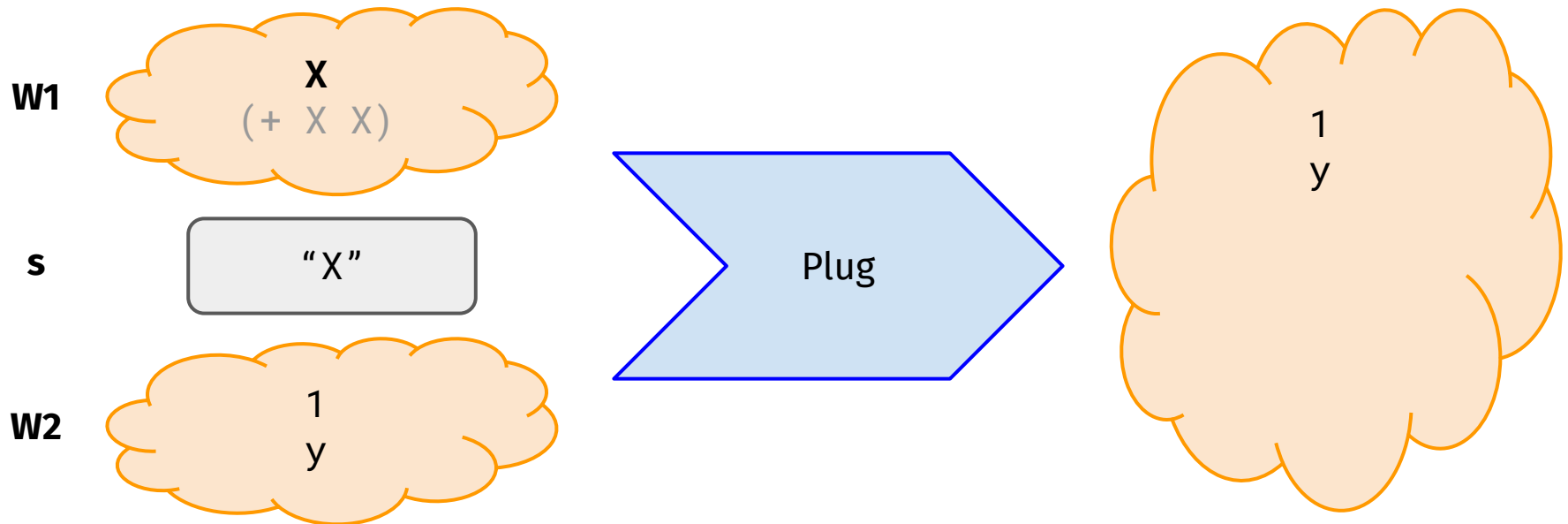
Plug \mathcal{W}_1 s \mathcal{W}_2

All combinations of
replacing s in \mathcal{W}_1 with
a term from \mathcal{W}_2

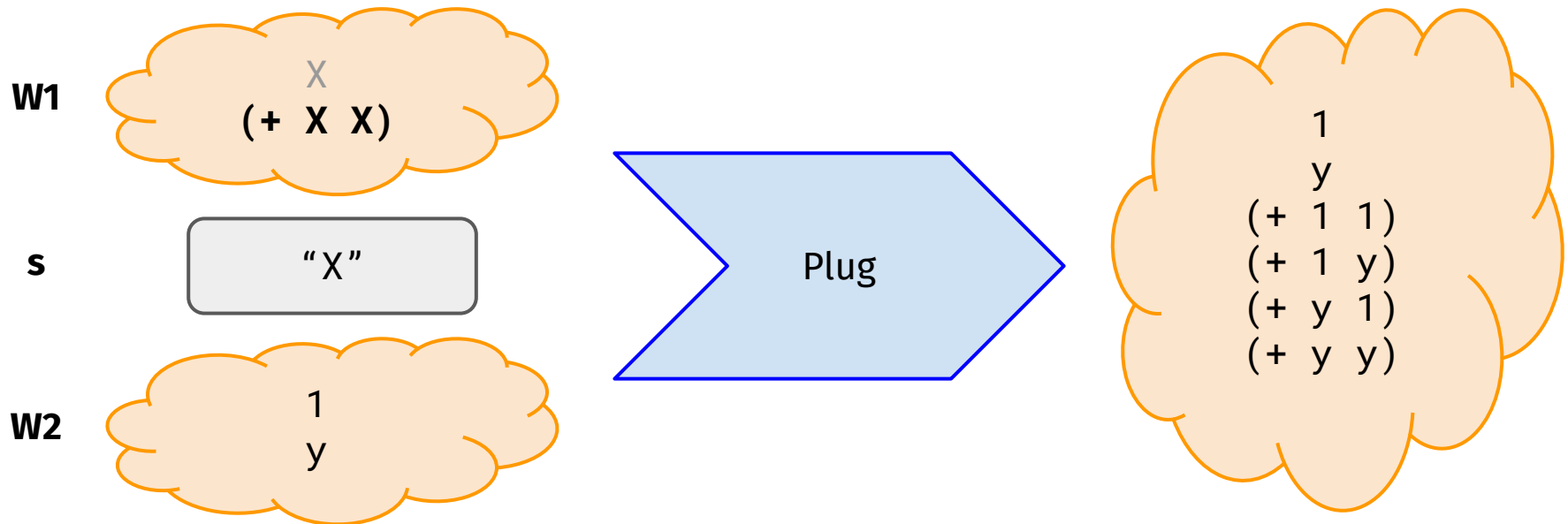
ENUMO DSL



ENUMO DSL



ENUMO DSL



ENUMO DSL

```
lits = Workload { a b 0 1 }  
exps = Workload { EXP (~ EXP) (+ EXP EXP) }  
  
wkld = exps.plugin("EXP", exps)  
        .plugin("EXP", lits)
```

ENUMO DSL

```
lits = Workload { a b 0 1 }  
exps = Workload { EXP (~ EXP) (+ EXP EXP) }
```

```
wkld = exps.plugin("EXP", exps)  
      .plugin("EXP", lits)
```

EXP	(+ (~ EXP) EXP)
(~ EXP)	(+ EXP (+ EXP EXP))
(+ EXP EXP)	(+ (~ EXP) (~ EXP))
(~ (~ EXP))	(+ (+ EXP EXP) EXP)
(+ EXP EXP)	(+ (~ EXP) (+ EXP EXP))
(~ (+ EXP EXP))	(+ (+ EXP EXP) (~ EXP))
(+ EXP (~ EXP))	(+ (+ EXP EXP) (+ EXP EXP))

ENUMO DSL

```
lits = Workload { a b 0 1 }  
exps = Workload { EXP (~ EXP) (+ EXP EXP) }
```

```
wkld = exps.plugin("EXP", exps)  
      .plugin("EXP", lits)
```

a	(~ (~ a))	(~ (+ b a))
b	(~ (~ b))	(~ (+ b b))
0	(~ (~ 0))	(~ (+ b 0))
1	(~ (~ 1))	(~ (+ b 1))
(~ a)	(~ (+ a a))	(~ (+ 0 a))
(~ b)	(~ (+ a b))	(~ (+ 0 b))
(~ 0)	(~ (+ a 0))	...
(~ 1)	(~ (+ a 1))	(+ (+ 1 1) (+ 1 1))

ENUMO DSL

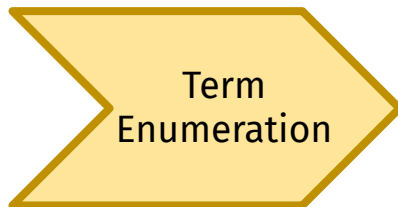
```
lits = Workload { a b 0 1 }
exps = Workload { EXP (~ EXP) (+ EXP EXP) }

wkld = exps.plugin("EXP", exps)
      .plugin("EXP", lits)

rules =
  wkld
    .to_egrph()
    .find_candidates()
    .select_rules(limits)
```

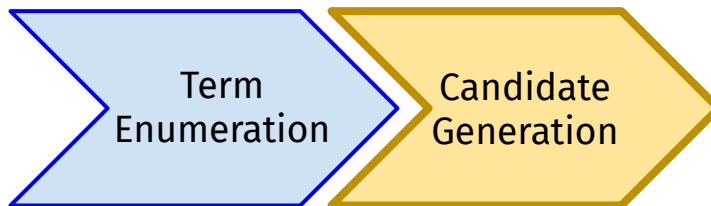
ENUMO DSL

```
lits = Workload { a b 0 1 }  
exps = Workload { EXP (~ EXP) (+ EXP EXP) }  
  
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    wkld  
      .to_egrph()  
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      .select_rules(limits)
```



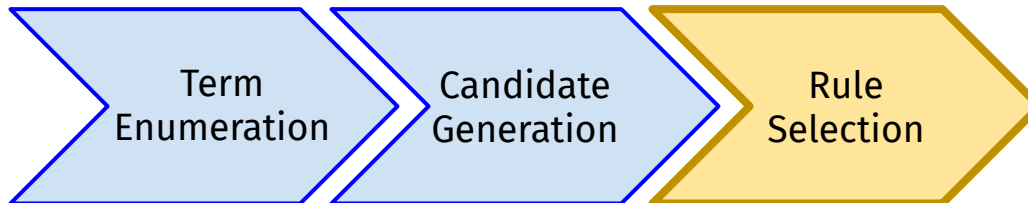
ENUMO DSL

```
lits = Workload { a b 0 1 }  
exps = Workload { EXP (~ EXP) (+ EXP EXP) }  
  
wkld = exps.plugin("EXP", exps)  
      .plugin("EXP", lits)  
  
rules =  
  wkld  
    .to_egrph()  
    .find_candidates()  
    .select_rules(limits)
```



ENUMO DSL

```
lits = Workload { a b 0 1 }  
exps = Workload { EXP (~ EXP) (+ EXP EXP) }  
  
wkld = exps.plugin("EXP", exps)  
      .plugin("EXP", lits)  
  
rules =  
  wkld  
    .to_egrph()  
    .find_candidates()  
    .select_rules(limits)
```



ENUMO DSL

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


ENUMO DSL

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

ENUMO DSL

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

ENUMO DSL

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

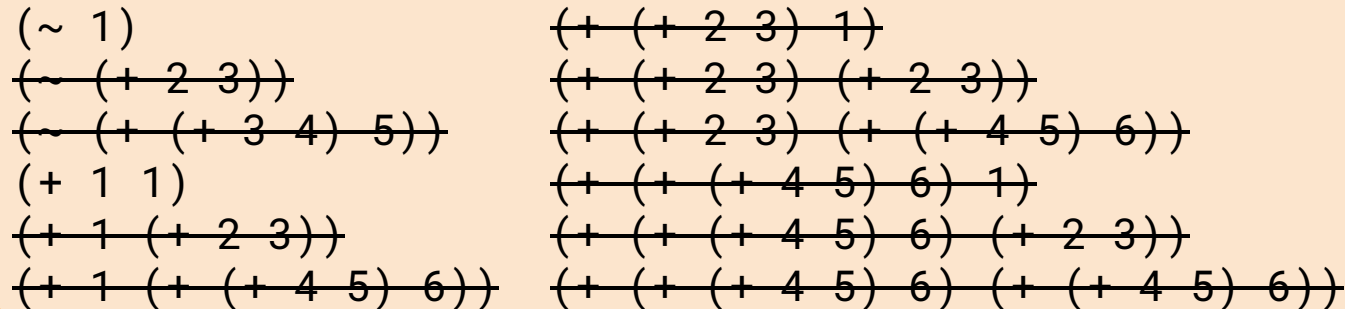
ENUMO DSL

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

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

ENUMO DSL

```
e1 = Workload { (~ EXP) (+ EXP EXP) }  
e2 = Workload { 1 (+ 2 3) (+ (+ 4 5) 6) }  
e1.plugin("EXP", 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))
(+ 1 1)	(+ (+ (+ 4 5) 6) 1)
(+ 1 (+ 2 3))	(+ (+ (+ 4 5) 6) (+ 2 3))
(+ 1 (+ (+ 4 5) 6))	(+ (+ (+ 4 5) 6) (+ (+ 4 5) 6))

ENUMO DSL

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

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

ENUMO DSL

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

1
(+ 2 3)

ENUMO DSL

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

1
(+ 2 3)

(~ 1)
(~ (+ 2 3))
(+ 1 1)
(+ 1 (+ 2 3))
(+ (+ 2 3) 1)
(+ (+ 2 3) (+ 2 3))

ENUMO DSL

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

1
(+ 2 3)

(~ 1)
~~(~ (+ 2 3))~~
(+ 1 1)
~~(+ 1 (+ 2 3))~~
~~(+ (+ 2 3) 1)~~
~~(+ (+ 2 3) (+ 2 3))~~

ENUMO DSL

Optimization: Pushing Filters through Plugs

$$\llbracket \text{Filter } f (\text{Plug } W1 \text{ s } W2) \rrbracket = \llbracket \text{Filter } f (\text{Plug } W1 \text{ s } (\text{Filter } f W2)) \rrbracket$$

ENUMO DSL

Optimization: Pushing Filters through Plugs

$$\llbracket \text{Filter } f (\text{Plug } W1 \text{ s } W2) \rrbracket = \llbracket \text{Filter } f (\text{Plug } W1 \text{ s } (\text{Filter } f W2)) \rrbracket$$

Requires monotonicity of f

ENUMO DSL

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

ENUMO DSL

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

Monotonic

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

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

Not monotonic

Comparison to Ruler

Domain	ENUMO LOC	ENUMO Time (s)	# ENUMO	# Ruler	ENUMO \rightarrow Ruler	Ruler \rightarrow ENUMO
bool	44	0.25	64	51		
bv4	21	7.10	180	84		
bv32	20	48.44	120	78		
rational	51	6.37	131	113		

Comparison to Ruler

Domain	ENUMO LOC	ENUMO Time (s)	# ENUMO	# Ruler	ENUMO \rightarrow Ruler	Ruler \rightarrow ENUMO
bool	44	0.25	64	51		
bv4	21	7.10	180	84		
bv32	20	48.44	120	78		
rational	51	6.37	131	113		

To test whether a ruleset, R ,
can derive a rule, $e_1 \Rightarrow e_2$:

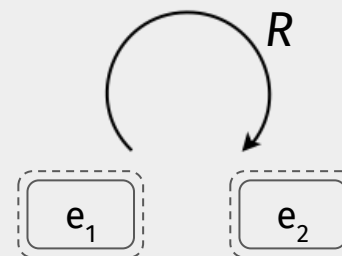


Initialize e-graph with e_1 and e_2

Comparison to Ruler

Domain	ENUMO LOC	ENUMO Time (s)	# ENUMO	# Ruler	ENUMO \rightarrow Ruler	Ruler \rightarrow ENUMO
bool	44	0.25	64	51		
bv4	21	7.10	180	84		
bv32	20	48.44	120	78		
rational	51	6.37	131	113		

To test whether a ruleset, R ,
can derive a rule, $e_1 \Rightarrow e_2$:



Run equality saturation with R

Comparison to Ruler

Domain	ENUMO LOC	ENUMO Time (s)	# ENUMO	# Ruler	ENUMO \rightarrow Ruler	Ruler \rightarrow ENUMO
bool	44	0.25	64	51		
bv4	21	7.10	180	84		
bv32	20	48.44	120	78		
rational	51	6.37	131	113		

To test whether a ruleset, R , can derive a rule, $e_1 \Rightarrow e_2$:



Derivable if the e-classes merge

Comparison to Ruler

Key Idea:
Enumo outperforms prior work

Domain	ENUMO LOC	ENUMO Time (s)	# ENUMO	# Ruler	ENUMO → Ruler	Ruler → ENUMO
bool	44	0.25	64	51	100%	87.5%
bv4	21	7.10	180	84	100%	38.3%
bv32	20	48.44	120	78	100%	58.3%
rational	51	6.37	131	113	100%	62.6%

Comparison to Ruler

Key Idea:
Rule Inference is fast

Domain	ENUMO LOC	ENUMO Time (s)	# ENUMO	# Ruler	ENUMO → Ruler	Ruler → ENUMO
bool	44	0.25	64	51	100%	87.5%
bv4	21	7.10	180	84	100%	38.3%
bv32	20	48.44	120	78	100%	58.3%
rational	51	6.37	131	113	100%	62.6%

Case Study: Large Grammar

```
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

Term Size	# Rules	ENUMO → Halide
-----------	---------	----------------

Case Study: Large Grammar

```
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)  
}
```

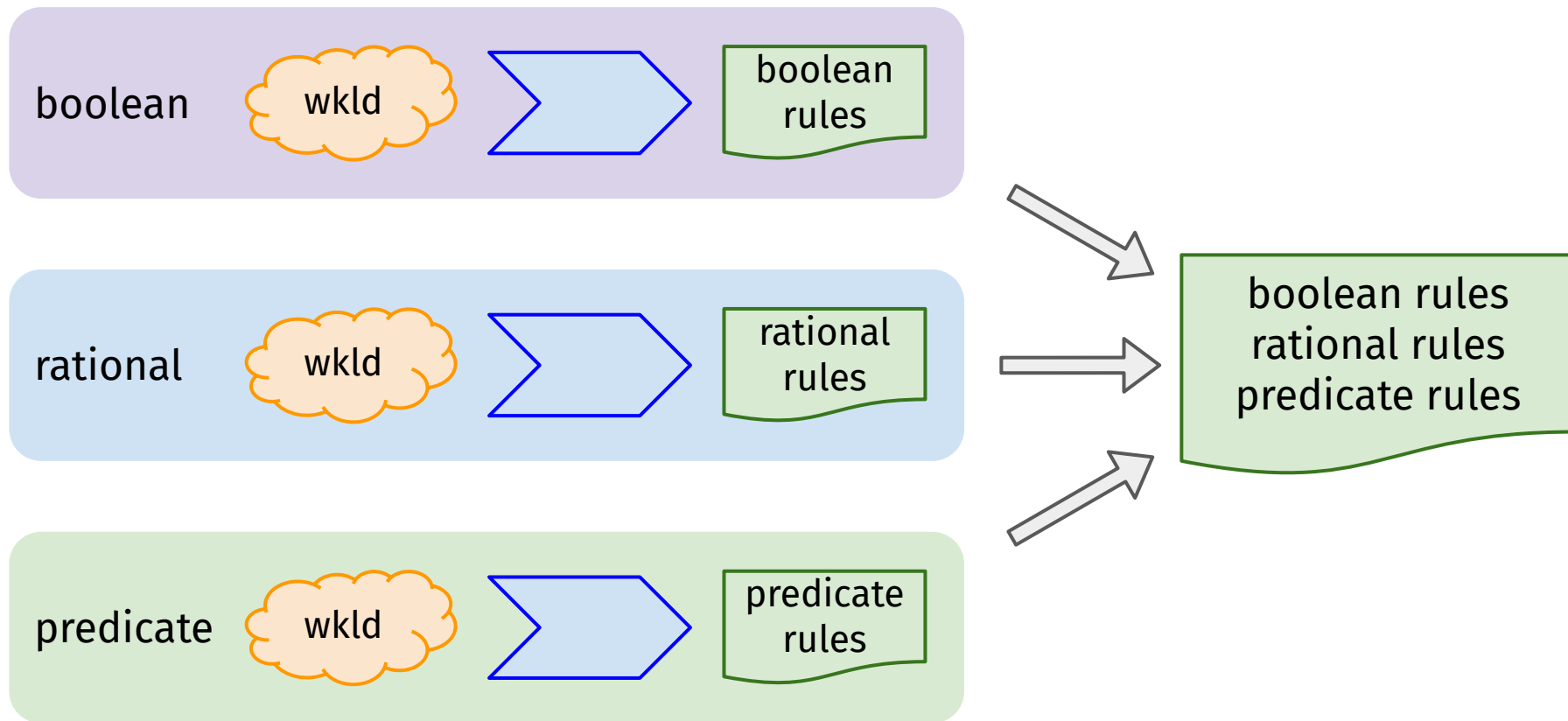
Term Size	# Rules	ENUMO → Halide
3	96	2.9%
4	224	6.9%
5	485	42.6%
6	TIMEOUT	TIMEOUT

Case Study: Large Grammar

```
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 experts know
which operators are
closely related

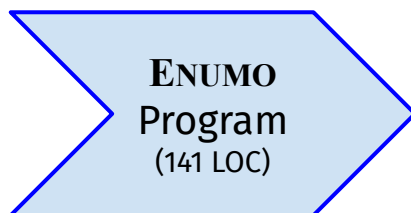
Case Study: Large Grammar



Case Study: Large Grammar

Key Idea:
Guided search enables progress
past exponential blowup

boolean rules
rational rules
predicate rules



Term Size	Time (s)	# Rules	ENUMO → Halide
Custom	51.76	845	90.6%

Case Study: Cross-Domain Inference

BV 4



Fast to synthesize

BV 128



Slow to synthesize

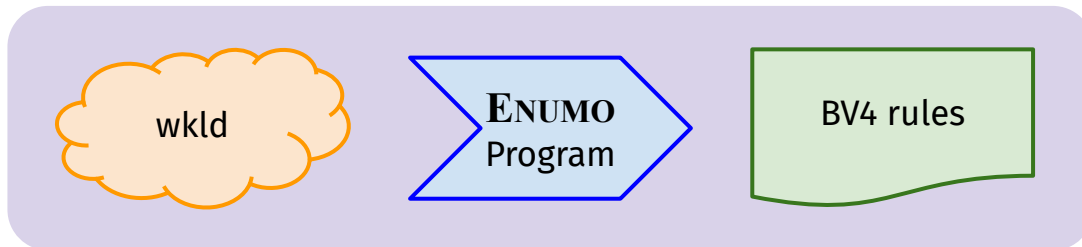
Case Study: Cross-Domain Inference

BV 4

✓ Fast to synthesize

BV 128

✗ Slow to synthesize



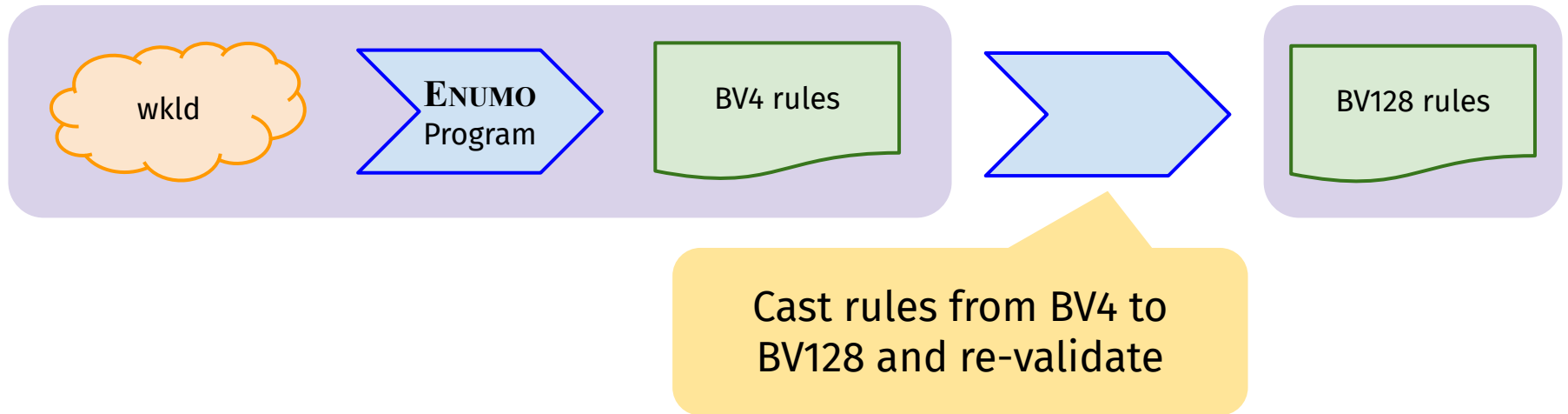
Case Study: Cross-Domain Inference

BV 4

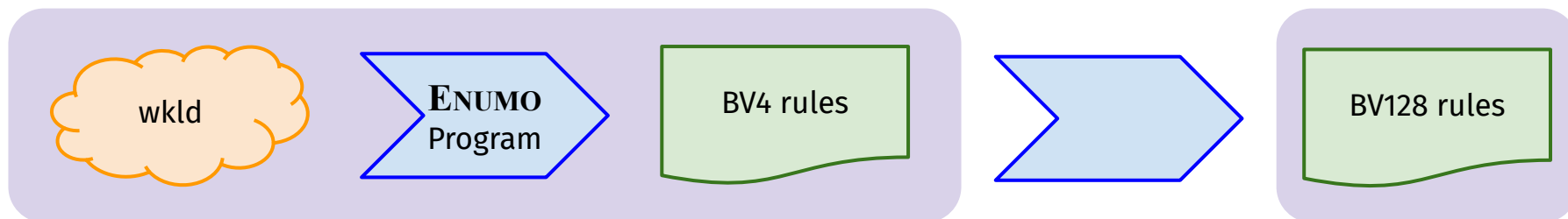
✓ Fast to synthesize

BV 128

✗ Slow to synthesize



Case Study: Cross-Domain Inference



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

Case Study: No Interpreter

`sin(0)`



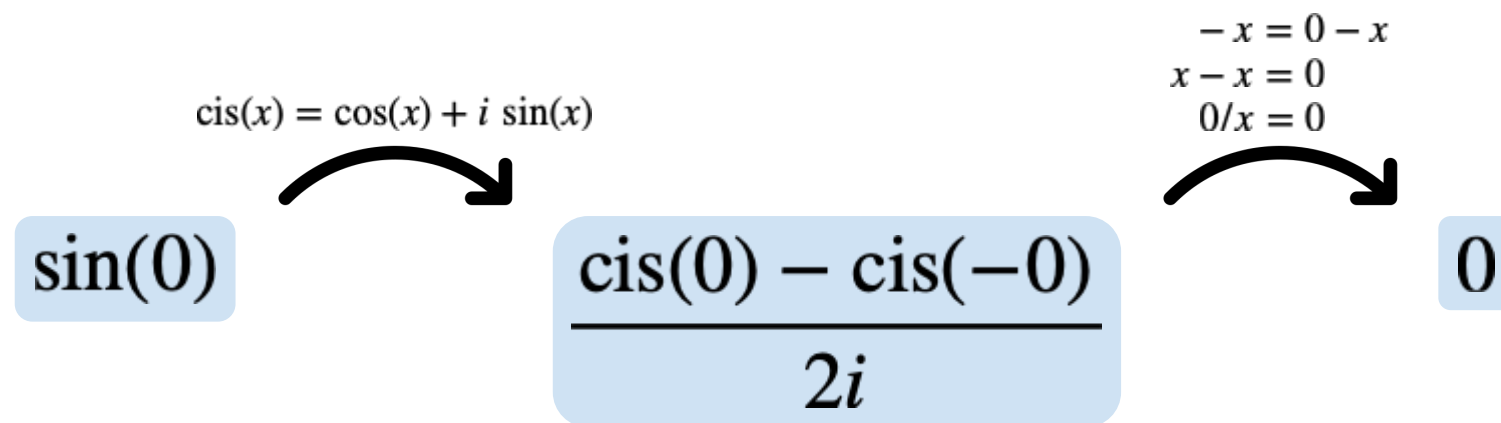
Case Study: No Interpreter

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


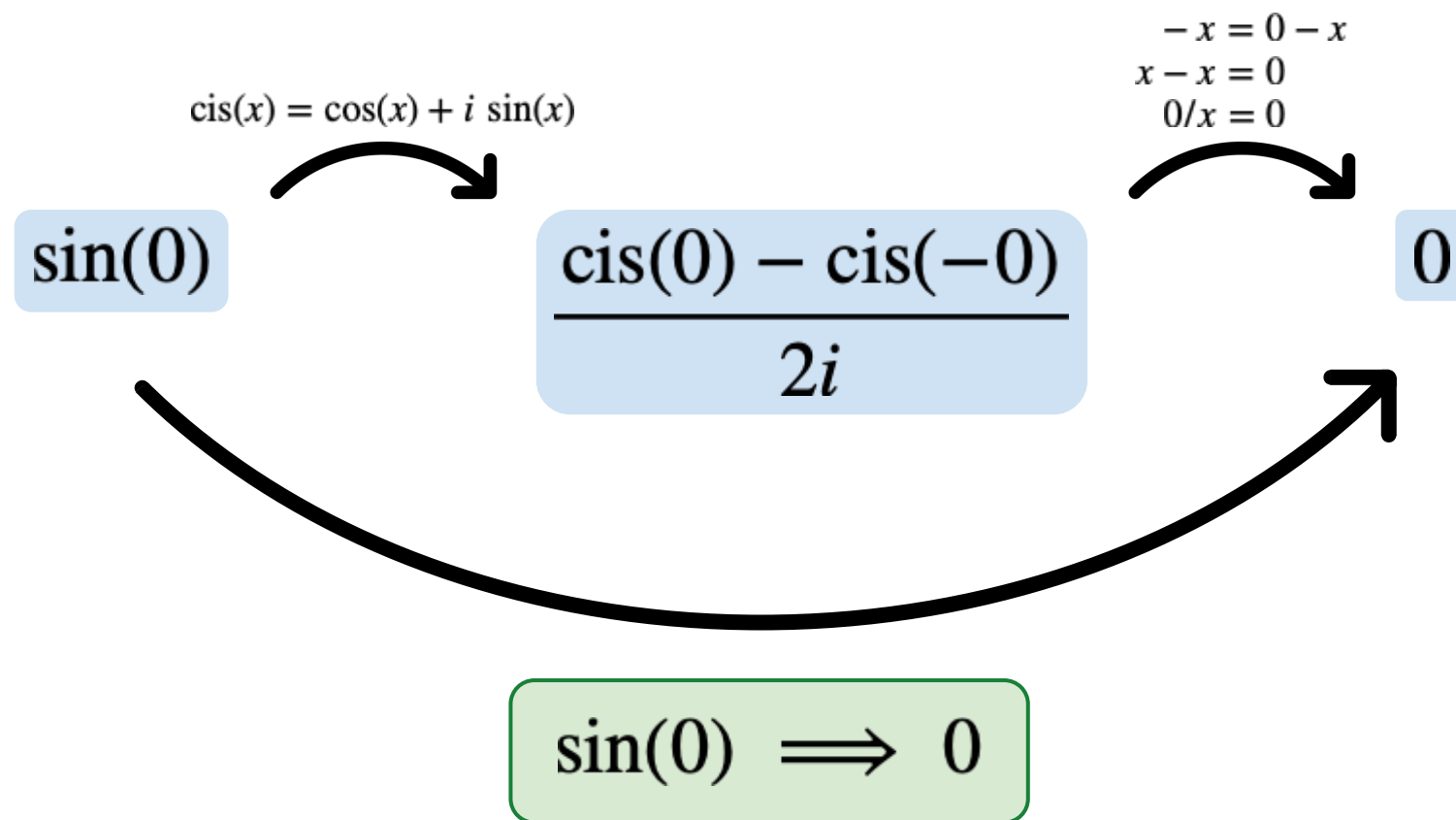
$\sin(0)$

$$\frac{\text{cis}(0) - \text{cis}(-0)}{2i}$$

Case Study: No Interpreter

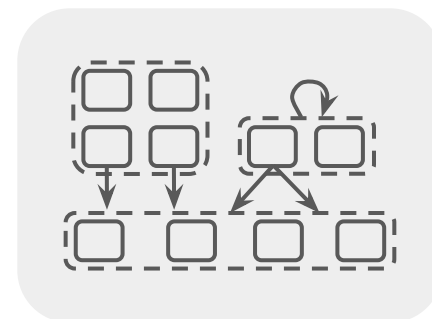
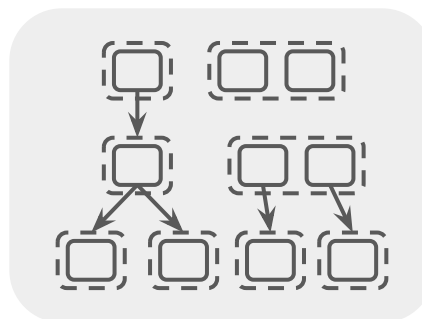
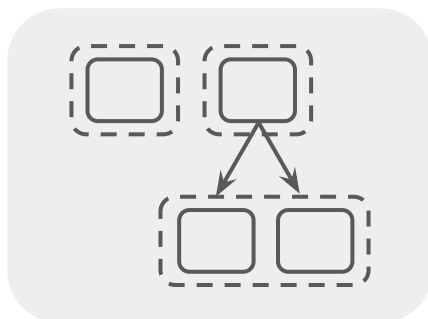
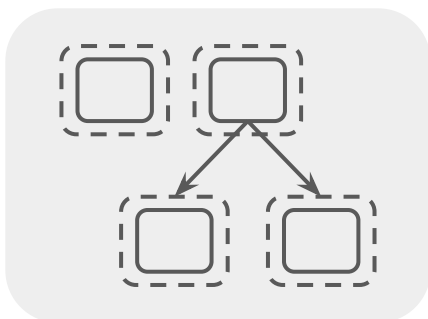
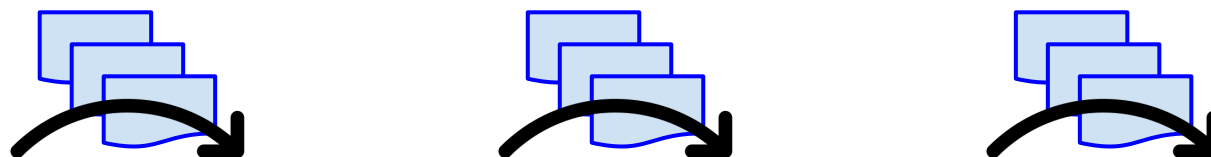


Case Study: No Interpreter

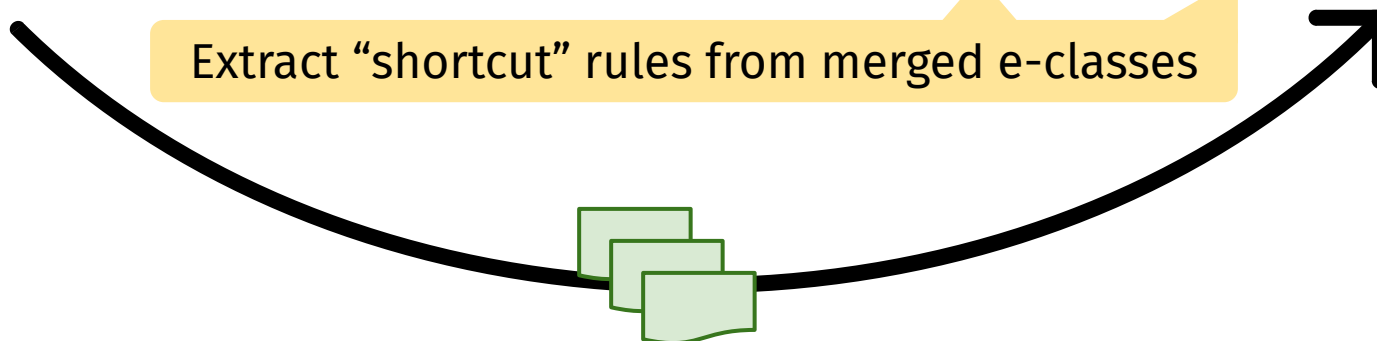


Case Study: No Interpreter

Multiple phases with different rulesets



Extract “shortcut” rules from merged e-classes



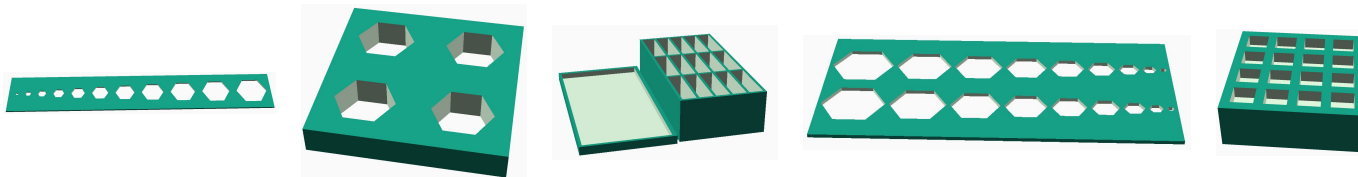
Case Study: No Interpreter

Key Idea:
New Domains for Rule Inference

$$\sin(b + a) \Rightarrow \sin(b) \cdot \cos(a) + \sin(a) \cdot \cos(b)$$
$$\sin(b) \cdot \sin(a) \Rightarrow (\cos(b - a) - \cos(b + a)) / 2$$

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

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



What's next?

Conditional Rule Inference

Verified Optimization

Large Language Models

What's next?

Conditional Rule Inference

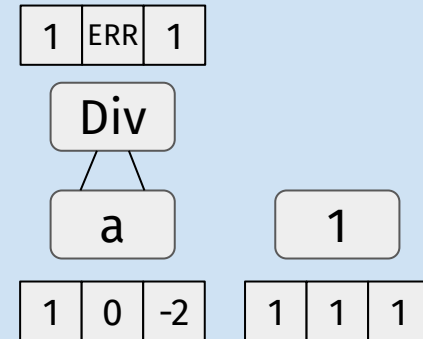
Verified Optimization

Large Language Models

Most rules depend on context

$x / x \Rightarrow 1$ **when $x \neq 0$**

Candidate generation will miss this because the arrays for the two e-classes won't match



Can we infer useful, simple side conditions for partial rules?

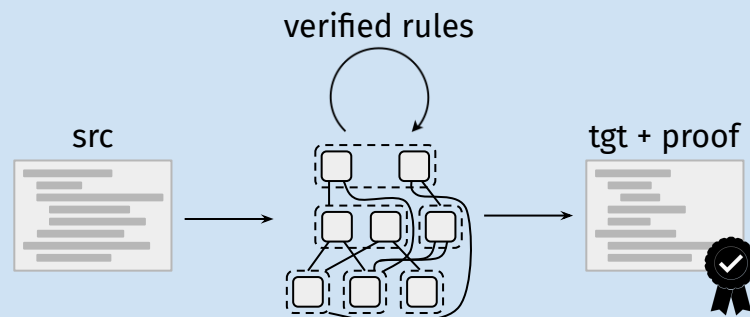
What's next?

Conditional Rule Inference

Verified Optimization

Large Language Models

Can we build a verified compiler using equality saturation?



Can we automatically infer rules for program optimization?

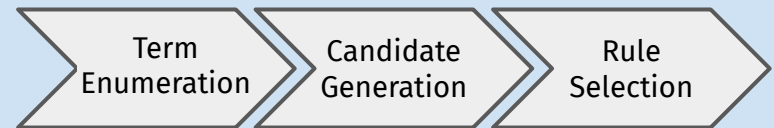
What's next?

Conditional Rule Inference

Verified Optimization

Large Language Models

Which parts of rule inference can language models help with?



Does it vary by domain?

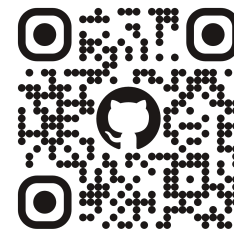
$x \ \&\& \ \text{true} \implies x$

$\sin 0 \implies 0$

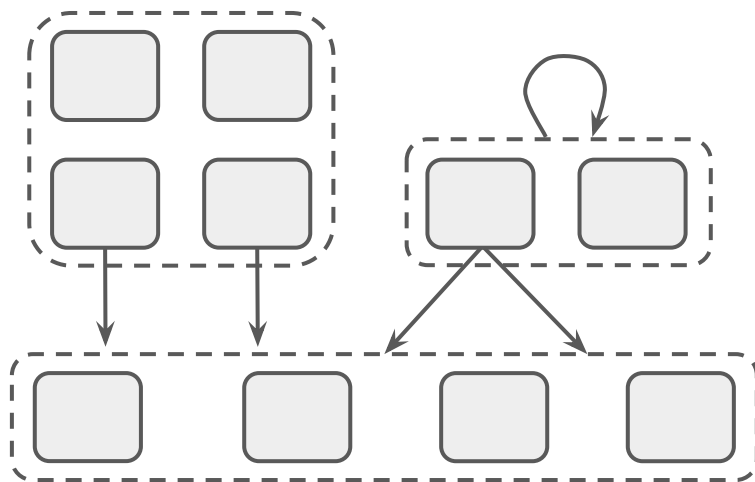
$c^{ba} \implies (c^a)^b$

$\text{Cube}(ad, be, cf) \implies \text{Scale}(a, b, c, \text{Cube}(d, e, f))$

Thank you!



Fast, Flexible, Robust Term Rewriting via Equality Saturation



Fast, Flexible, Robust Rule Inference via Programmable Theory Exploration

