

# Relational E-Matching

## Simpler, Faster, and Optimal

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PLDI 2021 Student Research Competition

# E-Graphs are everywhere!

Spores [VLDB '20]

Herbie [PLDI '15]

Szalinski [PLDI '20]

TenSat [MLSys '21]

Diospyoros [ASPLOS '21]

Glenside [MAPS '21]

egg [POPL '20]

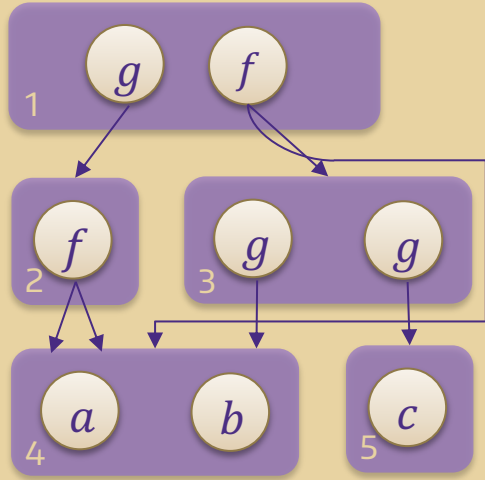
Z3

CVC4

Metatheory.jl

...

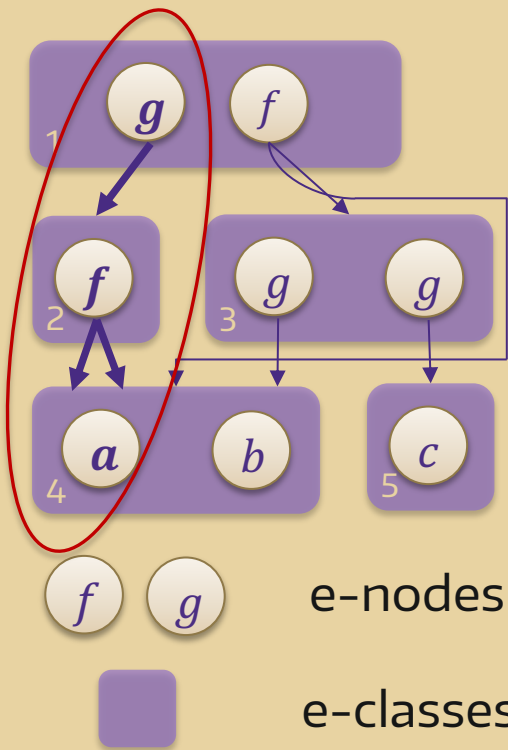
# E-Graphs



$$g(f(a, a))$$

e-class 1  
represents

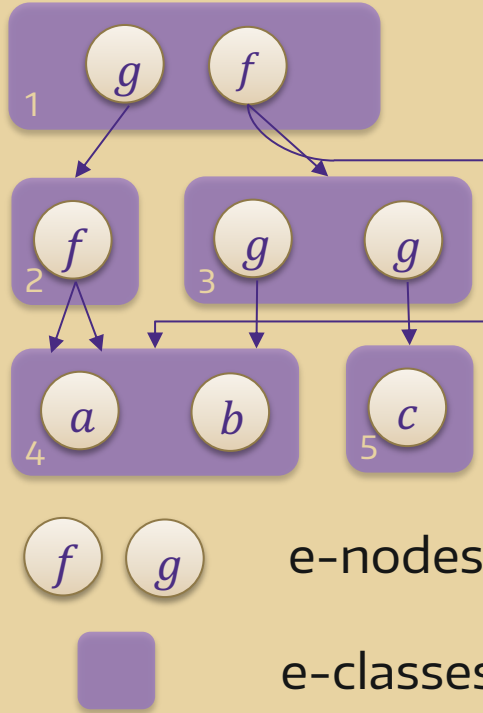
# E-Graphs



$g(f(a, a))$

e-class 1  
represents

# E-Graphs



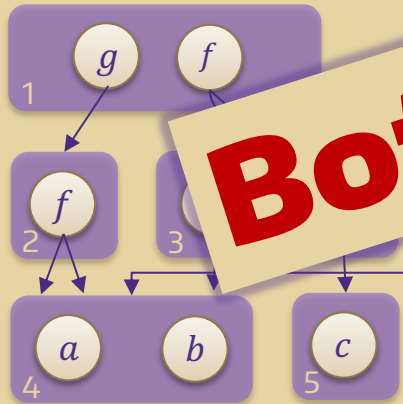
e-class 1  
represents

$$\begin{array}{ll}
 g(f(a, a)) & f(a, g(a)) \\
 g(f(a, b)) & f(a, g(b)) \\
 g(f(b, a)) & f(b, g(a)) \\
 g(f(b, b)) & f(b, g(b)) \\
 & f(a, g(c)) \\
 & f(b, g(c))
 \end{array}$$

exponentially many terms!

# E-Matching

- > E-matching is the query that finds patterns in an e-graph.



**Bottleneck!**

responds to

$$f(a, g(a)), f(a, g(b)) \\ f(b, g(a)), f(b, g(b)).$$

All witnessed by subst.  $\{\alpha \mapsto 4\}$ .

- > Responsible for 60-90% of the run time.

# Existing E-Matching Algorithms

$f(\alpha, g(\alpha))$



```
for all e-class  $c$  in e-graph  $E$ :  
  🙌 for all  $f$  node  $n_1$  in  $c$ :  
    subst = {  $\alpha \mapsto n_1.child_1$ , root  $\mapsto c$  }  
    🙌 for all  $g$  node  $n_2$  in  $n_1.child_2$ :  
      if subst[ $\alpha$ ] =  $n_2.child_1$ :  
        yield subst
```

**Quadratic runtime!**  
**Yet at most  $O(N)$  terms.**

# Relational E-Matching

E-Matching

E-Graphs

⊆

Conjunctive Queries

Relational Databases

Relational queries only involving  
joins e.g.,  
 $Q(a, c) \leftarrow R(a, b), S(b, c)$

Simpler



# Comparison to Existing E-Matching

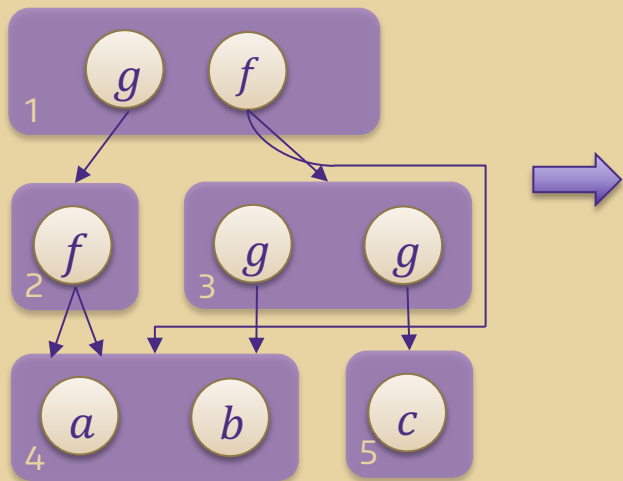
## Existing E-Matching

- ✗ Top-down backtracking search only.
- ✗ Exploits structural constraints only.
- ✗ No theoretical guarantee.

## Relational E-Matching

- ✓ Top-down, bottom-up, middle-out, etc. depending on the query optimizer.
- ✓ Exploits both structural constraints and equality constraints.
- ✓ Achieves optimality by adapting results from database research.

# E-Graphs → Relational Databases



Relations representing function symbols

$R_f$

<b>eclass-id</b>	<b>child<sub>1</sub></b>	<b>child<sub>2</sub></b>
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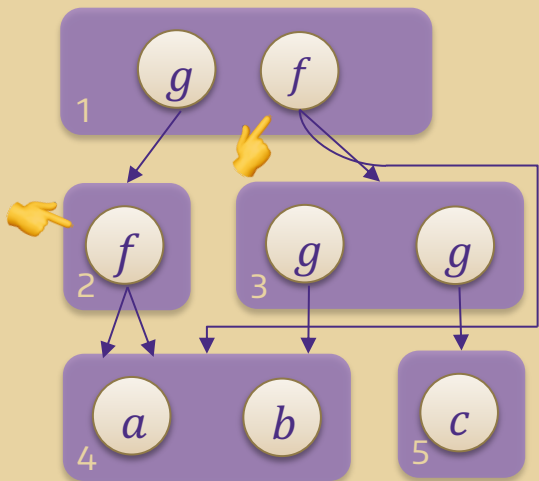
$R_g$

<b>eclass-id</b>	<b>child<sub>1</sub></b>
------------------	--------------------------

$R_a, R_b, R_c$

<b>eclass-id</b>
------------------

# E-Graphs → Relational Databases



Relations representing function symbols

$R_f$

eclass-id	child <sub>1</sub>	child <sub>2</sub>
1	4	3
2	4	4

**Faster**

# E-Matching $\rightarrow$ Conjunctive Queries

$f(\alpha, g(\alpha))$



for all e-class  $c$  in e-graph  $E$ :  
for all  $f$  node  $n_1$  in  $c$ :  
  subst =  $\{\alpha \mapsto n_1.child_1, \mathbf{root} \mapsto c\}$   
  for all  $g$  node  $n_2$  in  $n_1.child_2$ :  
    if subst[ $\alpha$ ] =  $n_2.child_1$ :  
      yield subst

**Quadratic**

$Q(\mathbf{root}, \alpha) \leftarrow$   
 $R_f(\mathbf{root}, \alpha, \mathbf{x}), R_g(\mathbf{x}, \alpha)$



for all  $r_f$  in  $R_f$ :  
  if  $(r_f.child_2, r_f.child_1)$  in  $R_g$ :  
    yield  $\{\mathbf{root} \mapsto R_f.eclass-id,$   
       $\alpha \mapsto R_f.child_1\}$

**Linear**

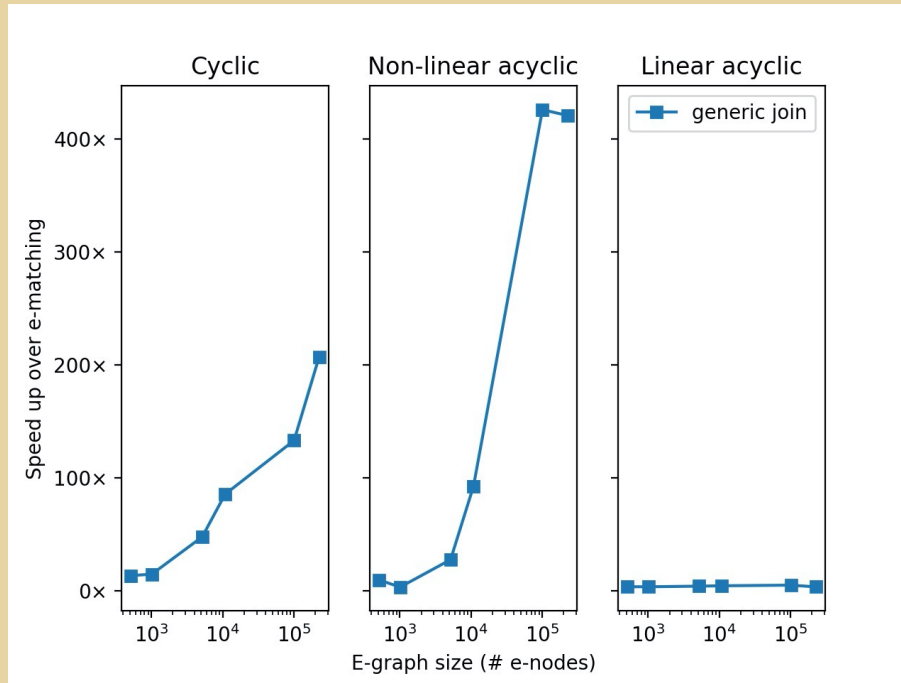
# Which CQ Algorithm to Use?

- > Traditional two-way join algorithms like hash join and merge-sort join are provably suboptimal.
- > We propose to use the generic join algorithm to solve the generated conjunctive query.
- > Fix a pattern  $p$ , let  $M(p, E)$  be the set of substitutions yielded by e-matching on an e-graph  $E$  with size  $n$ , relational e-matching runs in time

$$\tilde{O}\left(\max_E |M(p, E)|\right).$$

**Optimal**

# (Preliminary) Evaluations



Speed-ups over existing e-matching algorithms  
(de Moura and Bjørner)

- Asymptotic speed-ups on patterns with equality constraints (the cyclic and non-linear acyclic patterns).
- Similar performance on patterns with no equality constraints (the linear pattern).

**Thank you!**

# Why Faster?

$f(\alpha, g(\alpha))$

enumerates all terms of the form  $f(\alpha, g(\beta))$ , and check if  $\alpha = \beta$  only before yielding.

$Q(\text{root}, \alpha) \leftarrow$   
 $R_f(\text{root}, \alpha, \mathbf{x}), R_g(\mathbf{x}, \alpha)$

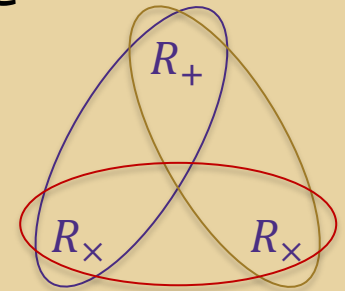
builds indices on both  $\alpha$  and  $\mathbf{x}$ ; only enumerates terms where constraints on  $\alpha$  and  $\mathbf{x}$  are both satisfied.

Equality constraints are exploited to prune the search space!



# Which CQ Algorithm to Use?

- > Traditional two-way join algorithms like hash join and merge-sort join are not optimal.
- > For example, for conjunctive query whose corresponding hypergraph is cyclic, two-way join algorithms will enumerate asymptotically more atoms than needed.
  - $(\alpha + \beta) \times (\alpha + \gamma)$ .



Hypergraph of  $(\alpha + \beta) \times (\alpha + \gamma)$

# Generic Join Algorithm

- > Multi-way join algorithm that avoids enumerating unnecessarily large intermediate relations.
- > Worst-case optimal & efficient for both cyclic and acyclic queries.