

Synthesis of Domain Specific Encoders for Bit- Vector Solvers

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with

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Appears in SAT'16

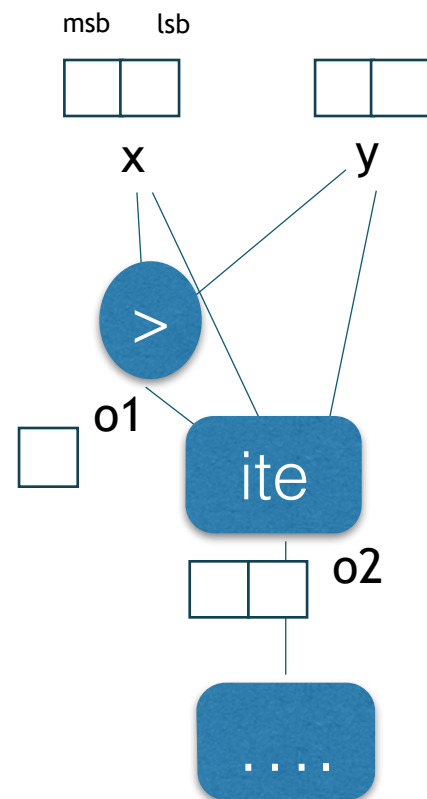
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High-level constraint to CNF clauses

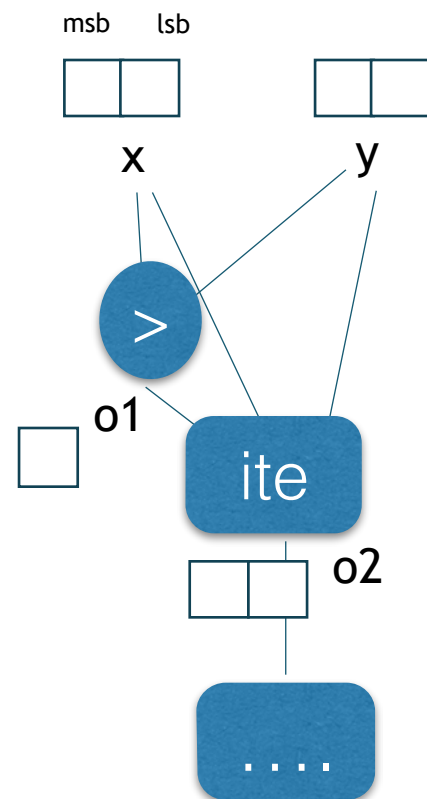
SMT solver
High-level constraint

SAT solver
CNF clauses



High-level constraint to CNF clauses

SMT solver
High-level constraint



Is this the “best”
encoding?

SAT solver
CNF clauses

$$\begin{aligned} \overline{y_1} \vee x_1 \vee \overline{t_1} \\ y_1 \vee \overline{x_1} \vee \overline{t_1} \\ \overline{y_1} \vee \overline{x_1} \vee t_1 \\ y_1 \vee x_1 \vee t_1 \\ \overline{t_2} \vee t_1 \\ \overline{t_2} \vee \overline{y_0} \\ \overline{t_2} \vee x_0 \end{aligned}$$

$$\begin{aligned} \overline{t_1} \vee y_0 \vee \overline{x_0} \vee t_2 \\ \overline{t_3} \vee \overline{y_1} \\ \overline{t_3} \vee x_1 \\ y_1 \vee \overline{x_1} \vee t_3 \\ o_1 \vee \overline{t_2} \\ o_1 \vee \overline{t_3} \\ t_2 \vee t_3 \vee \overline{o_1} \end{aligned}$$

$$\begin{aligned} \overline{o_1} \vee x_1 \vee \overline{o_{2_1}} \\ \overline{o_1} \vee \overline{x_1} \vee o_{2_1} \\ o_1 \vee y_1 \vee \overline{o_{2_1}} \\ o_1 \vee \overline{y_1} \vee o_{2_1} \\ x_1 \vee y_1 \vee \overline{o_{2_1}} \\ \overline{x_1} \vee \overline{y_1} \vee o_{2_1} \end{aligned}$$

$$\begin{aligned} \overline{o_1} \vee x_0 \vee \overline{o_{2_0}} \\ \overline{o_1} \vee \overline{x_0} \vee o_{2_0} \\ o_1 \vee y_0 \vee \overline{o_{2_0}} \\ o_1 \vee \overline{y_0} \vee o_{2_0} \\ x_0 \vee y_0 \vee \overline{o_{2_0}} \\ \overline{x_0} \vee \overline{y_0} \vee o_{2_0} \end{aligned}$$

....

Goal: Synthesize better code for this translation

How SAT solvers work?

- SAT solvers use unit propagation to infer variable assignments

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- SAT solvers use unit propagation to infer variable assignments

Current variables assignment

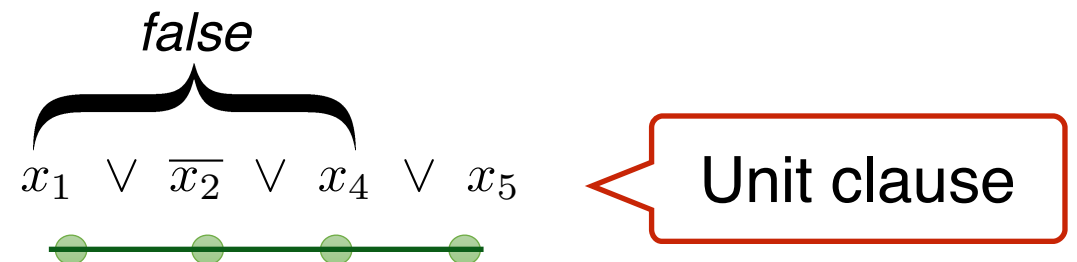
$$x_1 = F, x_2 = T, x_3 = F, x_4 = F$$

How SAT solvers work?

- SAT solvers use unit propagation to infer variable assignments

Current variables assignment

$$x_1 = F, x_2 = T, x_3 = F, x_4 = F$$

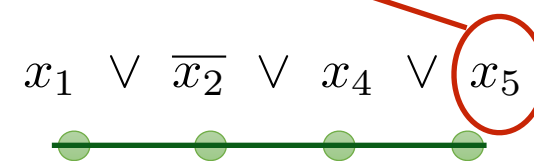


How SAT solvers work?

- SAT solvers use unit propagation to infer variable assignments

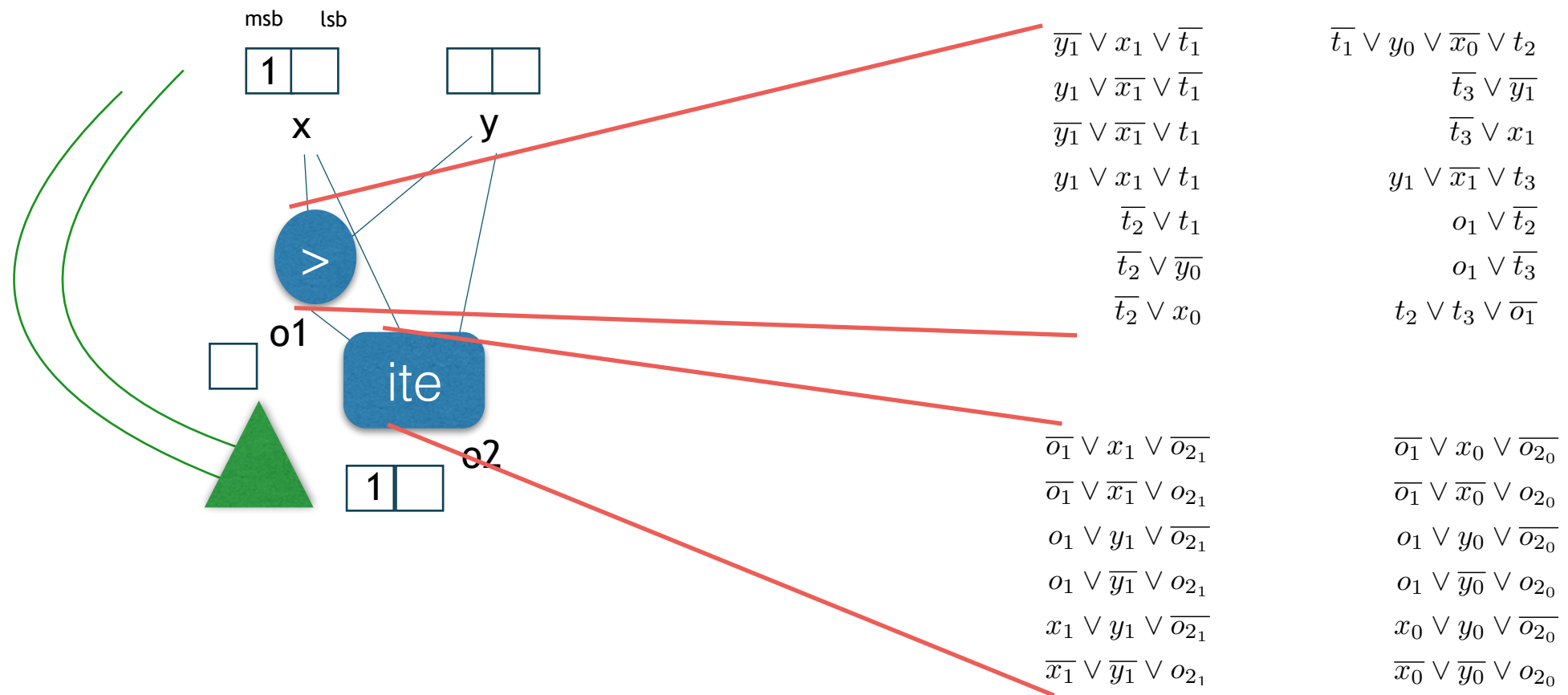
Current variables assignment

$$x_1 = F, x_2 = T, x_3 = F, x_4 = F, x_5 = T$$



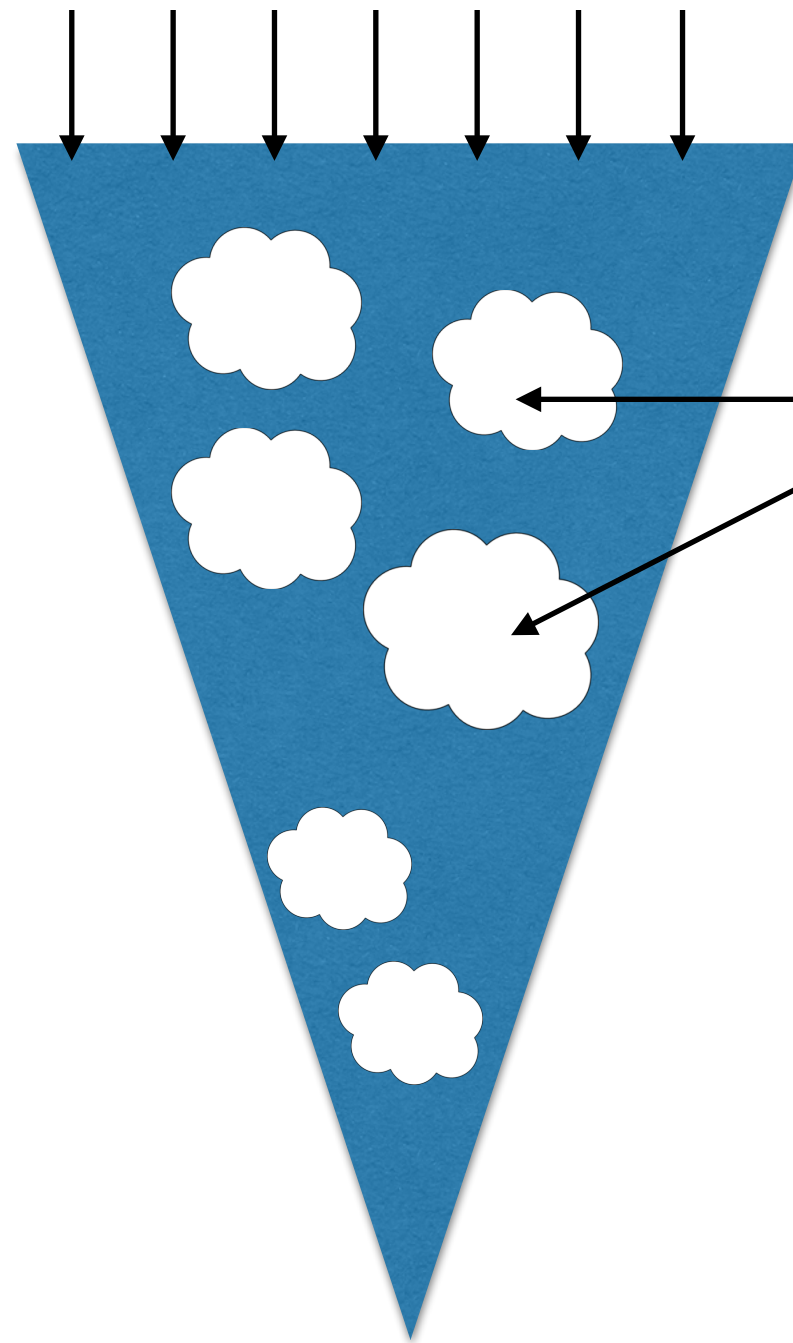
Criteria for a good encoding

- Maximal propagation
 - Maximize what we learn through unit propagations
- Fewer clauses
- Fewer variables

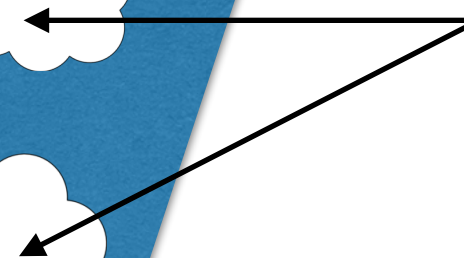


$$x_1 = 1 \xrightarrow{\text{Unit prop}} o_{2_1} = 1$$

Composing encodings does not preserve optimality



**Focus on
optimizing
encodings for
these patterns**

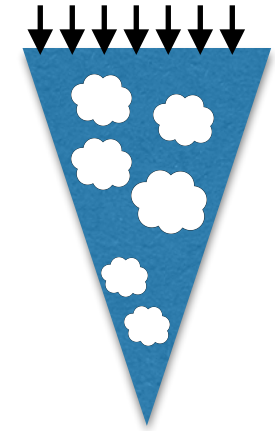




What patterns to
target?

How do we come
up with “optimal”
encoding for a
pattern?

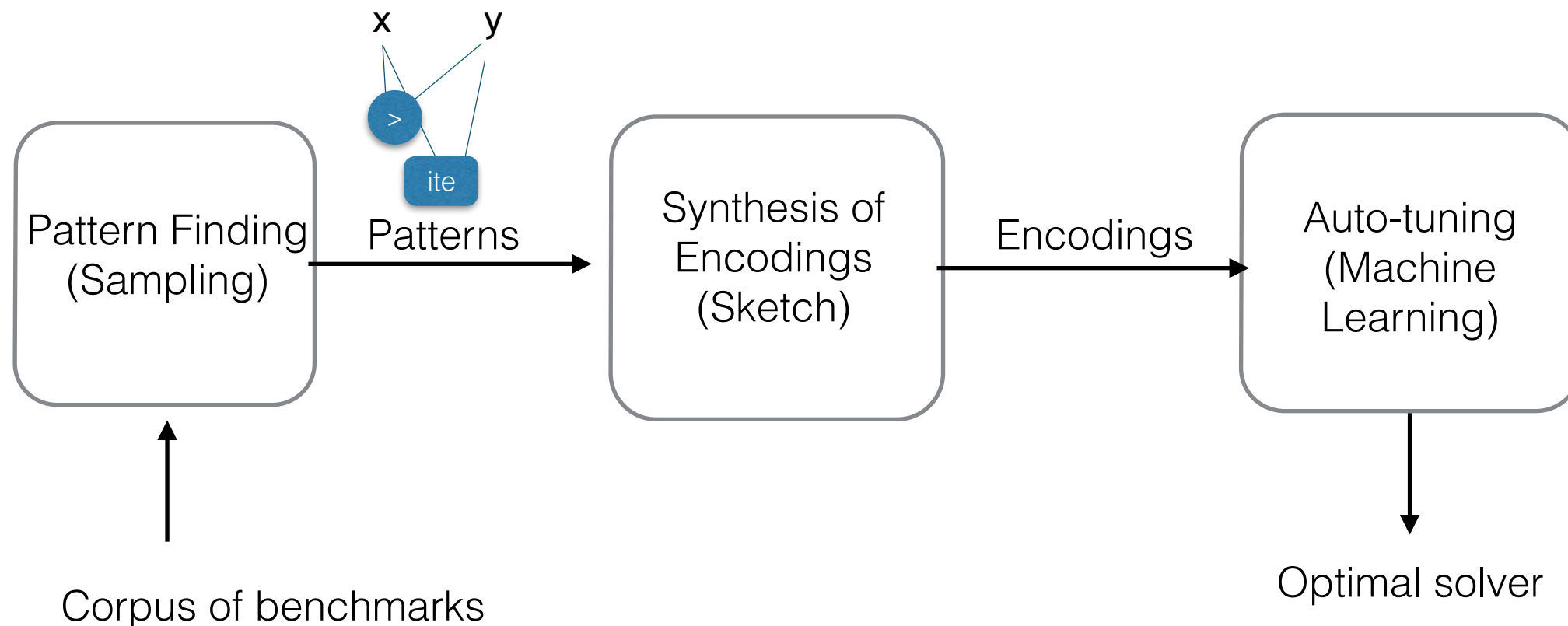
Do these encodings
actually improve the
performance?



What patterns to target?

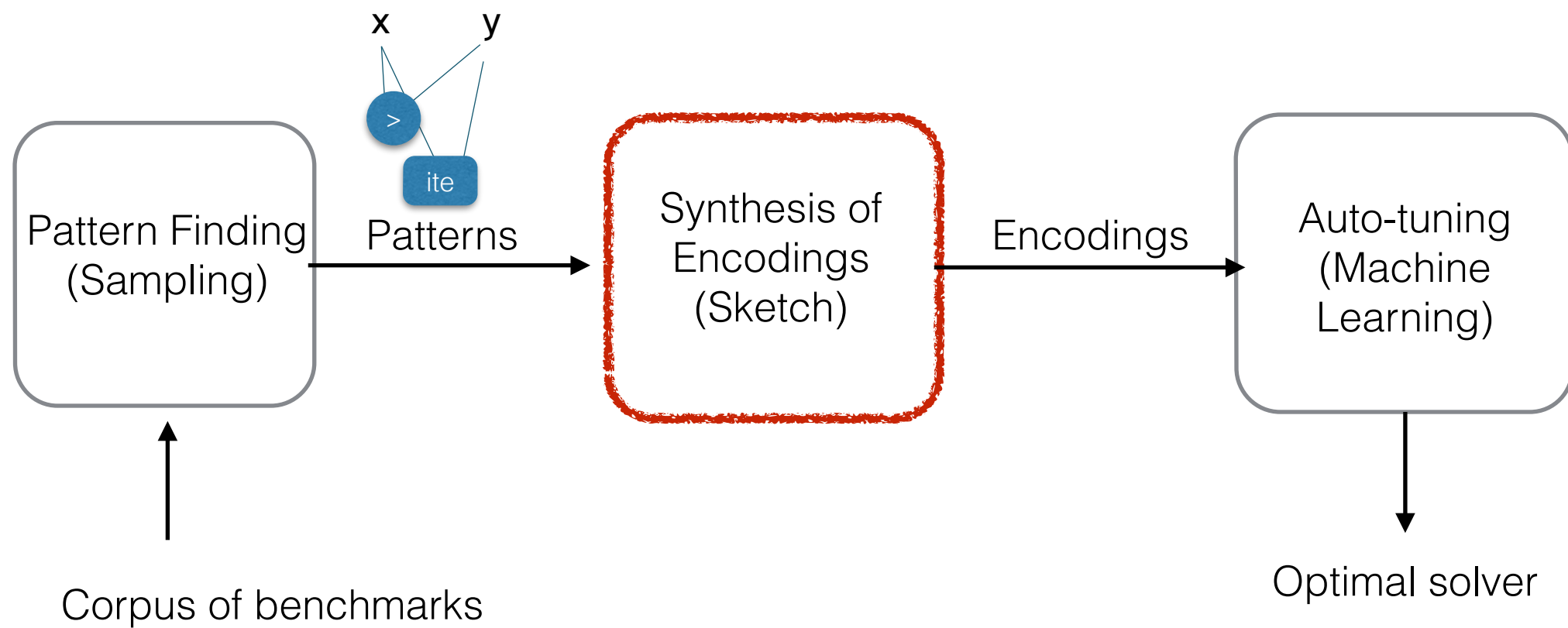
How do we come up with “optimal” encoding for a pattern?

Do these encodings actually improve the performance?



Related Work

- Automatic Generation of Propagation Complete SAT Encodings (VMCAI'16)
- Algorithm configuration for solver parameters
- Logic synthesis based SMT solvers



Synthesis as a SyGus problem

Boolean predicate P



CNF clauses C



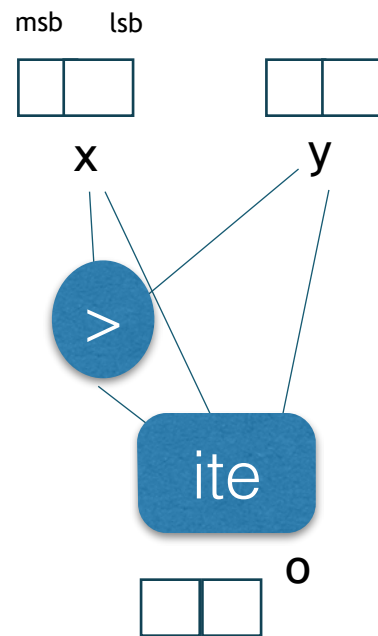
Templates

Boolean predicate P



CNF clauses C

$$o = ITE_N((GT_N x, y), x, y)$$



```

t1 = true
t2 = true
for i from N to 1 :
    t3 = newVar
    t4 = newVar
    clause({x[i], y[i], o[i]})
    clause({x[i], t1, t3})
    clause({x[i], t2, o[i], t4})
    clause({x[i], o[i], t3})
    clause({x[i], y[i], o[i]})
    clause({x[i], t2, o[i]})
    clause({x[i], t2, t4})
    clause({y[i], t2, t4})
    clause({y[i], o[i], t3})
    clause({y[i], t1, t3})
    clause({y[i], o[i], t3})
    clause({t1, t3})
    clause({t1, o[i], t3})
    clause({t2, t4})
    clause({t3, t4})
    t1 = t3
    t2 = t4
    
```

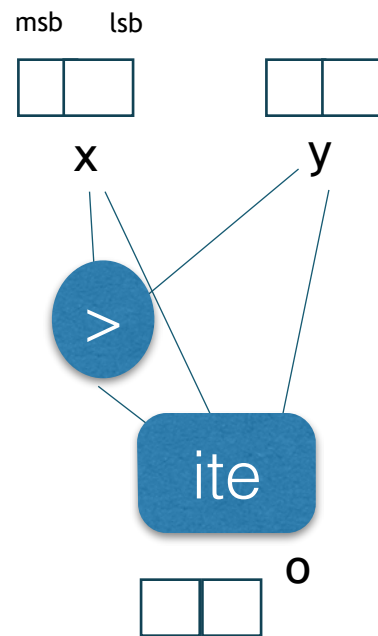

Templates

Boolean predicate P



CNF clauses C

$$o = ITE_N((GT_N, x, y), x, y)$$



$t1 = true$

$t2 = true$

for i from N to 1 :

$t3 = newVar$

$t4 = newVar$

$clause(\{x[i], y[i], \overline{o[i]}\})$

$clause(\{x[i], \overline{t1}, t3\})$

$clause(\{x[i], \overline{t2}, o[i], t4\})$

$clause(\{x[i], \overline{o[i]}, \overline{t3}\})$

$clause(\{x[i], y[i], o[i]\})$

$clause(\{x[i], \overline{t2}, o[i]\})$

$clause(\{x[i], \overline{t2}, t4\})$

$clause(\{y[i], \overline{t2}, t4\})$

$clause(\{y[i], \overline{o[i]}, \overline{t3}\})$

$clause(\{y[i], \overline{t1}, t3\})$

$clause(\{y[i], o[i], \overline{t3}\})$

$clause(\{t1, \overline{t3}\})$

$clause(\{\overline{t1}, o[i], t3\})$

$clause(\{t2, \overline{t4}\})$

$clause(\{t3, \overline{t4}\})$

$t1 = t3$

$t2 = t4$

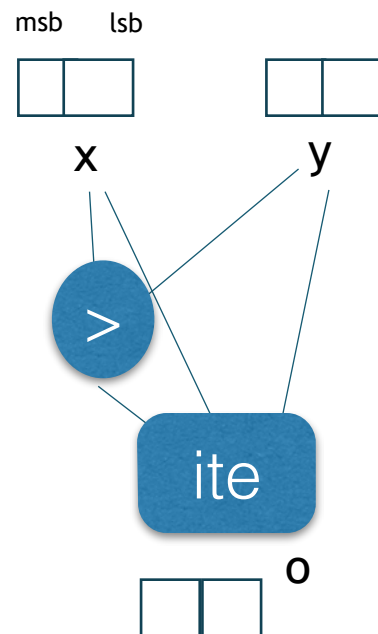
Templates

Boolean predicate P



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$t1 = true$

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$clause(\{x[i], \overline{t1}, t3\})$

$clause(\{x[i], \overline{t2}, o[i], t4\})$

$clause(\{x[i], \overline{o[i]}, t3\})$

$clause(\{x[i], y[i], o[i]\})$

$clause(\{x[i], \overline{t2}, o[i]\})$

$clause(\{x[i], \overline{t2}, t4\})$

$clause(\{y[i], \overline{t2}, t4\})$

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$clause(\{t1, \overline{t3}\})$

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$t1 = t3$

$t2 = t4$

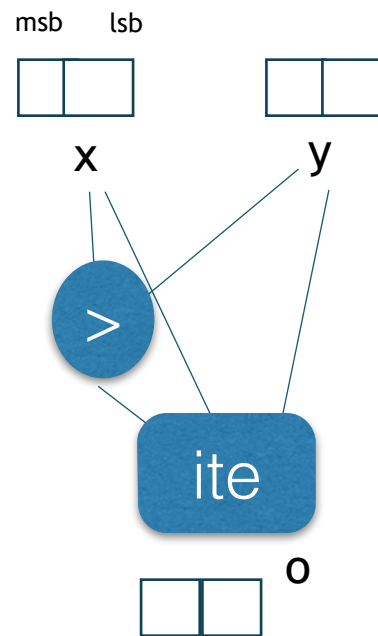
Templates

Boolean predicate P



CNF clauses C

$$o = ITE_N((GT_N, x, y), x, y)$$



$t1 = true$
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 $clause(\{t1, \overline{t3}\})$
 $clause(\{\overline{t1}, o[i], t3\})$
 $clause(\{t2, \overline{t4}\})$
 $clause(\{t3, \overline{t4}\})$

$t1 = t3$
 $t2 = t4$

Templates

```
t1 = ??  
t2 = ??  
for i from ?? to ?? :  
    t3 = newVar  
    t4 = newVar  
    genClauses(x[i], y[i], o[i], t1, t2, t3, t4)  
    t1 = t3  
    t2 = t4
```

Synthesis as a SyGus problem

Boolean predicate P



CNF clauses C



Specification

Boolean predicate P



CNF clauses C

Correctness:

$$\forall \sigma. P(\sigma) = C(\sigma)$$

σ = variables assignment

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$a = T, b = T, c = F, d = T, e = T$$

$$a \vee c \vee \bar{d}$$

$$T \vee F \vee F$$

$$a \vee \bar{c} \vee d$$

$$T \vee T \vee T$$

$$\bar{a} \vee b \vee \bar{d}$$

$$F \vee T \vee F$$

$$\bar{a} \vee \bar{b} \vee d$$

$$F \vee F \vee T$$

$$b \vee c \vee \bar{d}$$

$$T \vee F \vee F$$

$$\bar{b} \vee \bar{c} \vee d$$

$$F \vee T \vee T$$

$$d \vee \bar{e}$$

$$T \vee F$$

$$\bar{d} \vee e$$

$$F \vee T$$

$$P(\sigma) = C(\sigma) = T$$

Specification

Boolean predicate P



CNF clauses C

Correctness:

$$\forall \sigma. P(\sigma) = C(\sigma)$$

σ = variables assignment

$$d = ite(a, b, c) \wedge e = d$$

$$a = F, b = T, c = F, d = T, e = T$$

$$a \vee c \vee \bar{d}$$

$$F \vee F \vee F$$

$$a \vee \bar{c} \vee d$$

$$F \vee T \vee T$$

$$\bar{a} \vee b \vee \bar{d}$$

$$T \vee T \vee F$$

$$\bar{a} \vee \bar{b} \vee d$$

$$T \vee F \vee T$$

$$P(\sigma) = C(\sigma) = F$$

$$b \vee c \vee \bar{d}$$

$$T \vee F \vee F$$

$$\bar{b} \vee \bar{c} \vee d$$

$$F \vee T \vee T$$

$$d \vee \bar{e}$$

$$T \vee F$$

$$\bar{d} \vee e$$

$$F \vee T$$

Specification

Boolean predicate P



CNF clauses C

**Maximal
Propagation**

$$\forall \sigma. \text{ satisfiable}(\sigma, P) \implies$$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$b = T, c = T$$

$$a \vee c \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee \bar{b} \vee d$$

$$b \vee c \vee \bar{d}$$

$$\bar{b} \vee \bar{c} \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

Specification

Boolean predicate P



CNF clauses C

**Maximal
Propagation**

$$\begin{aligned} \forall \sigma. \text{ satisfiable}(\sigma, P) &\implies \\ \forall x_i, b_i. (\text{forces}(\sigma, P, x_i, b_i) &\implies \end{aligned}$$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$b = T, c = T \implies e = T$$

$$a \vee c \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee \bar{b} \vee d$$

$$b \vee c \vee \bar{d}$$

$$\bar{b} \vee \bar{c} \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

Specification

Boolean predicate P



CNF clauses C

**Maximal
Propagation**

$$\forall \sigma. \text{ satisfiable}(\sigma, P) \implies \forall x_i, b_i. (\text{forces}(\sigma, P, x_i, b_i) \implies UP(C, \sigma) \sqsubseteq \text{extend}(\sigma, x_i, b_i))$$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$b = T, c = T \implies e = T$$

$$a \vee c \vee \bar{d}$$

$$a \vee T \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

$$a \vee F \vee d$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee T \vee \bar{d}$$

$$\bar{a} \vee \bar{b} \vee d$$

$$\bar{a} \vee F \vee d$$

$$b \vee c \vee \bar{d}$$

$$T \vee T \vee \bar{d}$$

$$\bar{b} \vee \bar{c} \vee d$$

$$F \vee F \vee d$$

$$d \vee \bar{e}$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

$$\bar{d} \vee e$$

Specification

Boolean predicate P



CNF clauses C

**Maximal
Propagation**

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$$d = \text{ite}(a, b, c) \wedge e = d$$

$$b = T, c = T, d = T \implies e = T$$

$$a \vee c \vee \bar{d}$$

$$a \vee T \vee F$$

$$a \vee \bar{c} \vee d$$

$$a \vee F \vee T$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee T \vee F$$

$$\bar{a} \vee \bar{b} \vee d$$

$$\bar{a} \vee F \vee T$$

$$b \vee c \vee \bar{d}$$

$$T \vee T \vee F$$

$$\bar{b} \vee \bar{c} \vee d$$

$$F \vee F \vee T$$

$$d \vee \bar{e}$$

$$T \vee \bar{e}$$

$$\bar{d} \vee e$$

$$F \vee e$$

Specification

Boolean predicate P



CNF clauses C

**Maximal
Propagation**

$$\forall \sigma. \text{ satisfiable}(\sigma, P) \implies \forall x_i, b_i. (\text{forces}(\sigma, P, x_i, b_i) \implies UP(C, \sigma) \sqsubseteq \text{extend}(\sigma, x_i, b_i))$$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$b = T, c = T, d = T, e = T \implies e = T$$

$$a \vee c \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

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$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

$$a \vee T \vee F$$

$$a \vee F \vee T$$

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$$T \vee T \vee F$$

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$$T \vee \bar{e}$$

$$F \vee e$$

Specification

Boolean predicate P



CNF clauses C

**Maximal
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$$\forall \sigma. \text{ satisfiable}(\sigma, P) \implies \forall x_i, b_i. (\text{forces}(\sigma, P, x_i, b_i) \implies UP(C, \sigma) \sqsubseteq \text{extend}(\sigma, x_i, b_i))$$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$b = T, c = T, d = T, e = T \implies e = T$$

$$a \vee c \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee \bar{b} \vee d$$

$$b \vee c \vee \bar{d}$$

$$\bar{b} \vee \bar{c} \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

$$a \vee T \vee F$$

$$a \vee F \vee T$$

$$\bar{a} \vee T \vee F$$

$$\bar{a} \vee F \vee T$$

$$T \vee T \vee F$$

$$F \vee F \vee T$$

$$T \vee \bar{e}$$

$$F \vee e$$

Specification

Boolean predicate P



CNF clauses C

**Maximal
Propagation**

$$\forall \sigma. \text{ *satisfiable* }(\sigma, P) \implies \forall x_i, b_i. (\text{ *forces* }(\sigma, P, x_i, b_i) \implies UP(C, \sigma) \sqsubseteq \text{ *extend* }(\sigma, x_i, b_i))$$

Difficult for synthesis tools
to reason about

Is there a better specification for maximal
propagation?

Synthesis Friendly Spec

Boolean predicate P



CNF clauses C

$$1. \forall \sigma. \text{ satisfiable}(\sigma, P) \implies C(\sigma) \neq \text{false}$$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$a \vee c \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee \bar{b} \vee d$$

$$b \vee c \vee \bar{d}$$

$$\bar{b} \vee \bar{c} \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

$$b = T, c = T$$

$$a \vee T \vee \bar{d}$$

$$a \vee F \vee d$$

$$\bar{a} \vee T \vee \bar{d}$$

$$\bar{a} \vee F \vee d$$

$$T \vee T \vee \bar{d}$$

$$F \vee F \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

is satisfiable

No false clauses

Synthesis Friendly Spec

Boolean predicate P



CNF clauses C

1. $\forall \sigma. \text{satisfiable}(\sigma, P) \implies C(\sigma) \neq \text{false}$
2. $\forall \sigma. \text{maypropagate}(\sigma, P) \implies C(\sigma) \text{ has a unit clause}$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$a \vee c \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee \bar{b} \vee d$$

$$b \vee c \vee \bar{d}$$

$$\bar{b} \vee \bar{c} \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

$$b = T, c = T$$

$$a \vee T \vee \bar{d}$$

$$a \vee F \vee d$$

$$\bar{a} \vee T \vee \bar{d}$$

$$\bar{a} \vee F \vee d$$

$$T \vee T \vee \bar{d}$$

$$F \vee F \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

Can propagate e to T

Has a unit clause

Synthesis Friendly Spec

Boolean predicate P



CNF clauses C

1. $\forall \sigma. \text{satisfiable}(\sigma, P) \implies C(\sigma) \neq \text{false}$
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$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

$$a \vee T \vee F$$

$$a \vee F \vee T$$

$$\bar{a} \vee T \vee F$$

$$\bar{a} \vee F \vee T$$

$$T \vee T \vee F$$

$$F \vee F \vee T$$

$$T \vee \bar{e}$$

$$F \vee e$$

Can still propagate e to T

Has a unit clause

Synthesis Friendly Spec

Boolean predicate P



CNF clauses C

1. $\forall \sigma. \text{satisfiable}(\sigma, P) \implies C(\sigma) \neq \text{false}$
2. $\forall \sigma. \text{maypropagate}(\sigma, P) \implies C(\sigma) \text{ has a unit clause}$
3. $\forall \sigma. \text{unsatisfiable}(\sigma, P) \implies C(\sigma) = \text{false} \text{ (or) } C(\sigma) \text{ has a unit clause}$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$b = T, c = T, e = F$$

$$a \vee c \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee \bar{b} \vee d$$

$$b \vee c \vee \bar{d}$$

$$\bar{b} \vee \bar{c} \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

$$a \vee T \vee \bar{d}$$

$$a \vee F \vee d$$

$$\bar{a} \vee T \vee \bar{d}$$

$$\bar{a} \vee F \vee d$$

$$T \vee T \vee \bar{d}$$

$$F \vee F \vee d$$

$$d \vee T$$

$$\bar{d} \vee F$$

Unsatisfiable

Has unit clauses

Synthesis Friendly Spec

Boolean predicate P



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2. $\forall \sigma. \text{maypropagate}(\sigma, P) \implies C(\sigma) \text{ has a unit clause}$
3. $\forall \sigma. \text{unsatisfiable}(\sigma, P) \implies C(\sigma) = \text{false} \text{ (or) } C(\sigma) \text{ has a unit clause}$

$$d = \text{ite}(a, b, c) \wedge e = d$$

$$b = T, c = T, e = F, d = F$$

$$a \vee c \vee \bar{d}$$

$$a \vee \bar{c} \vee d$$

$$\bar{a} \vee b \vee \bar{d}$$

$$\bar{a} \vee \bar{b} \vee d$$

$$b \vee c \vee \bar{d}$$

$$\bar{b} \vee \bar{c} \vee d$$

$$d \vee \bar{e}$$

$$\bar{d} \vee e$$

$$a \vee T \vee T$$

$$a \vee F \vee F$$

$$\bar{a} \vee T \vee T$$

$$\bar{a} \vee F \vee F$$

$$T \vee T \vee T$$

$$F \vee F \vee F$$

$$F \vee T$$

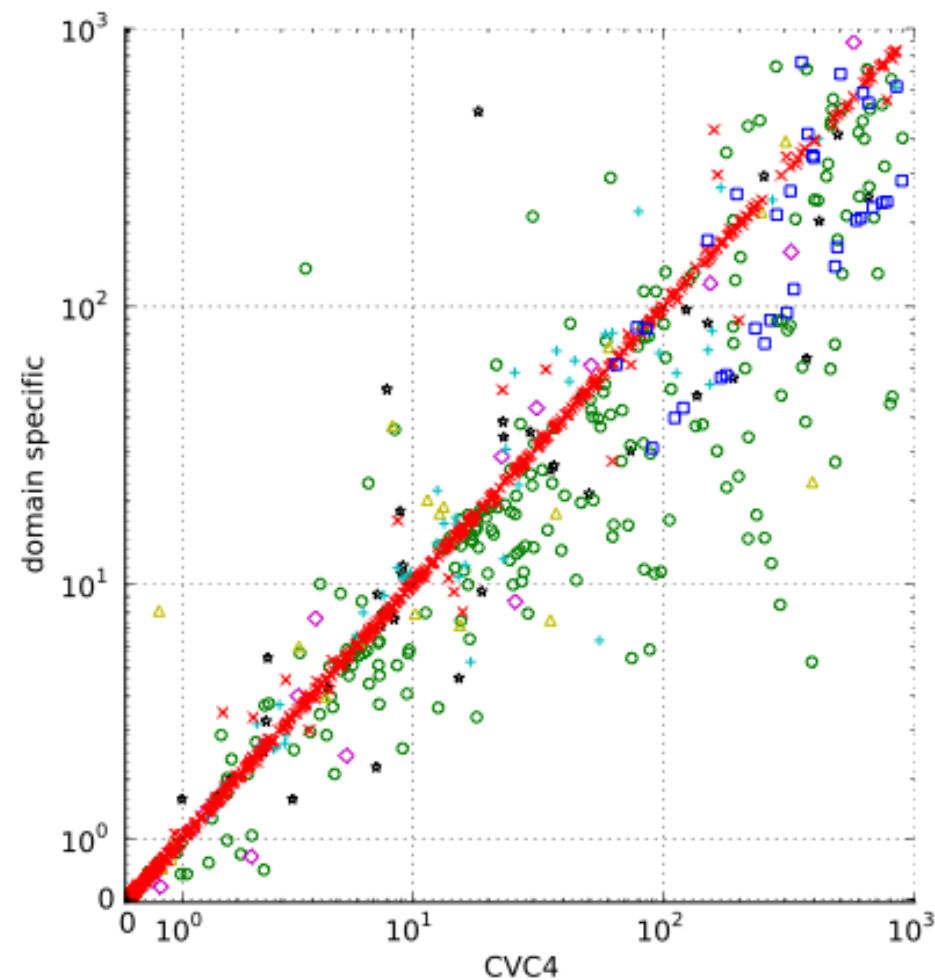
$$T \vee F$$

Also unsatisfiable

Has a false clause

Does this buy you anything?

- 7 domains from bit-vector category in SMT Comp'15



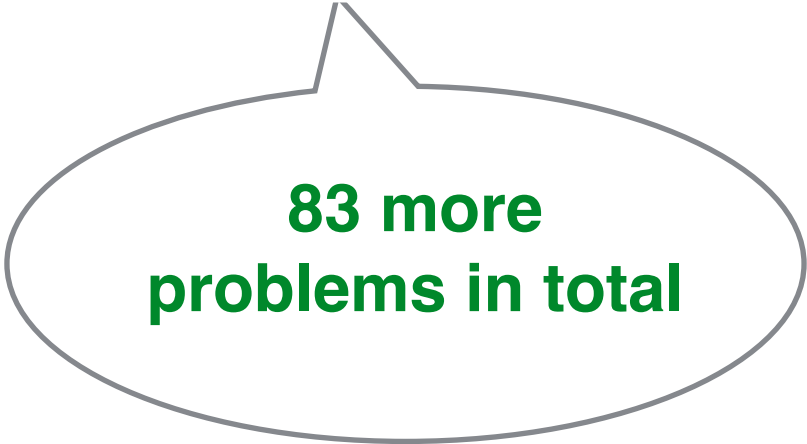
* Excluding the benchmarks used for training

Solve more problems

Benchmark Family	Solved by CVC4 → Our Solver
<i>Log-slicing</i> (79)	33 → 62
<i>ASP</i> (365)	240 → 288
<i>Mcm</i> (61)	40 → 43
<i>Brummayerbiere2</i> (33)	28 → 29
<i>Float</i> (62)	59 → 60
<i>Brummayerbiere3</i> (40)	23 → 24
<i>Bruttomesso</i> (676)	623 → 623
TOTAL	1046 → 1129

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TOTAL	1046 → 1129



**83 more
problems in total**

Cross domain performance

Solver Domain	→ ↓	log- slicing	asp	mcm	brumma2	float	brumma3	brutto
<i>log-slicing</i>		62	58	36	59	32	35	35
<i>asp</i>		227	288	255	227	236	253	240
<i>mcm</i>		39	38	43	40	39	39	41
<i>brumma2</i>		29	28	28	29	29	29	29
<i>float</i>		57	57	59	57	60	60	59
<i>brumma3</i>		22	22	25	22	23	24	23
<i>brutto</i>		607	606	623	609	623	623	623

 
Best Worst

What did it take?

- Around 2000 benchmarks across 7 domains
- Over 200 million nodes in the high level SMT constraints
- Sampling generated ~2000 patterns (size ≤ 5)
- ~2000 SyGus problems to solve
- Generated ~40k to 160k lines of code per domain (30 lines per encoding)
- 8 hours of auto-tuning per domain
- On total, took 10-20 hours per domain with parallelism of 30

Conclusion

- A framework to generate the code for translating high-level constraints to CNF
- Combined synthesis with pattern finding and auto-tuning to generate domain specific solvers
- Significant performance improvement compared to CVC4

THANK YOU