

Model-Based Reasoning about Quantified Formulas in CVC4

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Outline

- Introduction to SMT and applications
- Model-Based approach for handling quantifiers
- How can we construct good models?
- Experimental Results

Satisfiability Modulo Theories (SMT)

- SMT solvers are powerful tools
 - Used in many formal methods applications
 - Support many background *theories*
 - Arithmetic, bitvectors, arrays, datatypes, ...
 - May generate:
 - Proofs
 - Theorem proving, software/hardware verification
 - Models
 - Failing instances of aforementioned applications
 - Invariant synthesis, scheduling, test case generation

Satisfiability Modulo Theories

$$(f(a) = b \vee f(a) = c) \wedge c+1 = b \wedge \forall x. f(x) = g(x)$$

Satisfiability Modulo Theories

$$(f(a) = b \vee f(a) = c) \wedge c+1 = b \wedge \forall x. f(x) = g(x)$$

↓ Abstract to propositional logic

$$(A \vee B) \wedge C \wedge D$$

Satisfiability Modulo Theories

$$(f(a) = b \vee f(a) = c) \wedge c+1 = b \wedge \forall x. f(x) = g(x)$$

$$\underbrace{(A \vee B)}_{\text{true}} \wedge \underbrace{C}_{\text{true}} \wedge \underbrace{D}_{\text{true}}$$

Find satisfying assignment: A , C , D

Satisfiability Modulo Theories

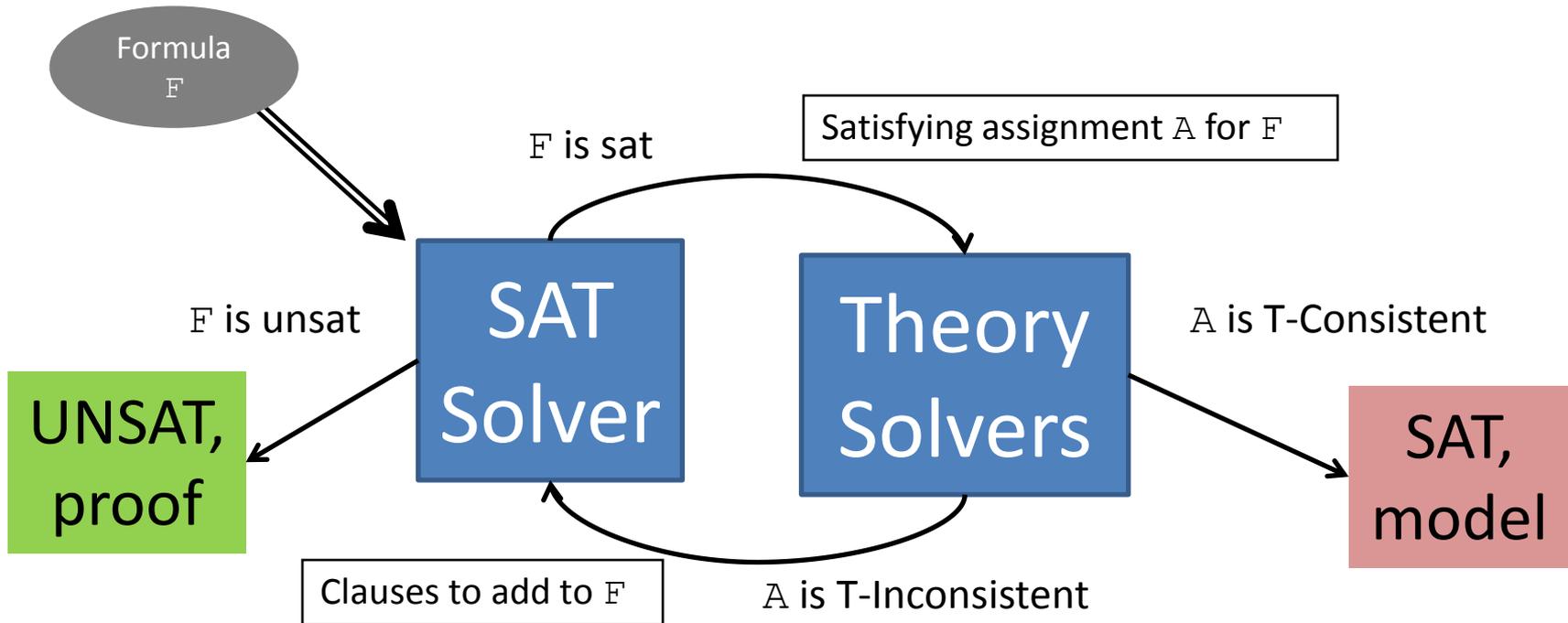
$$(f(a) = b \vee f(a) = c) \wedge c+1 = b \wedge \forall x. f(x) = g(x)$$

$$\underbrace{(A \vee B)}_{\text{true}} \wedge \underbrace{C}_{\text{true}} \wedge \underbrace{D}_{\text{true}}$$

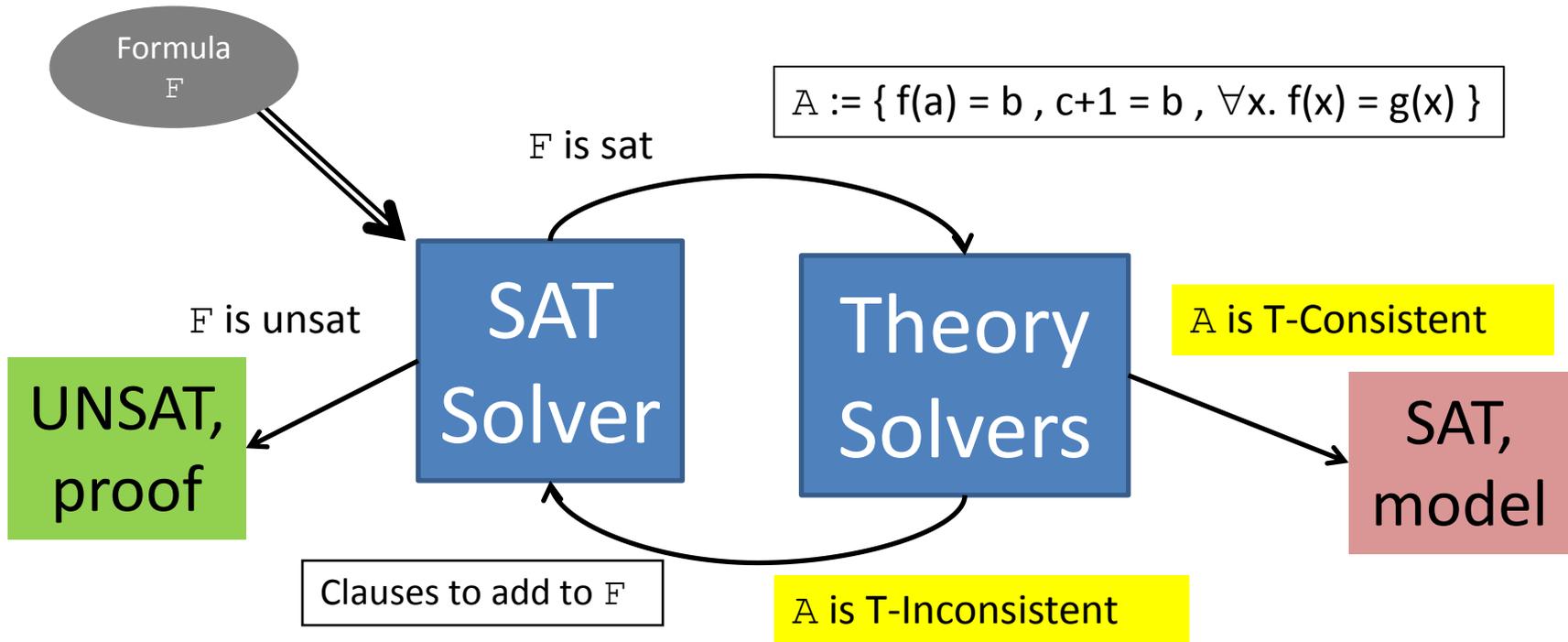
Find satisfying assignment: A , C , D

Check T-consistency: $f(a) = b$, $c+1 = b$, $\forall x. f(x) = g(x)$

DPLL(T) Architecture



DPLL(T) Architecture : Challenge



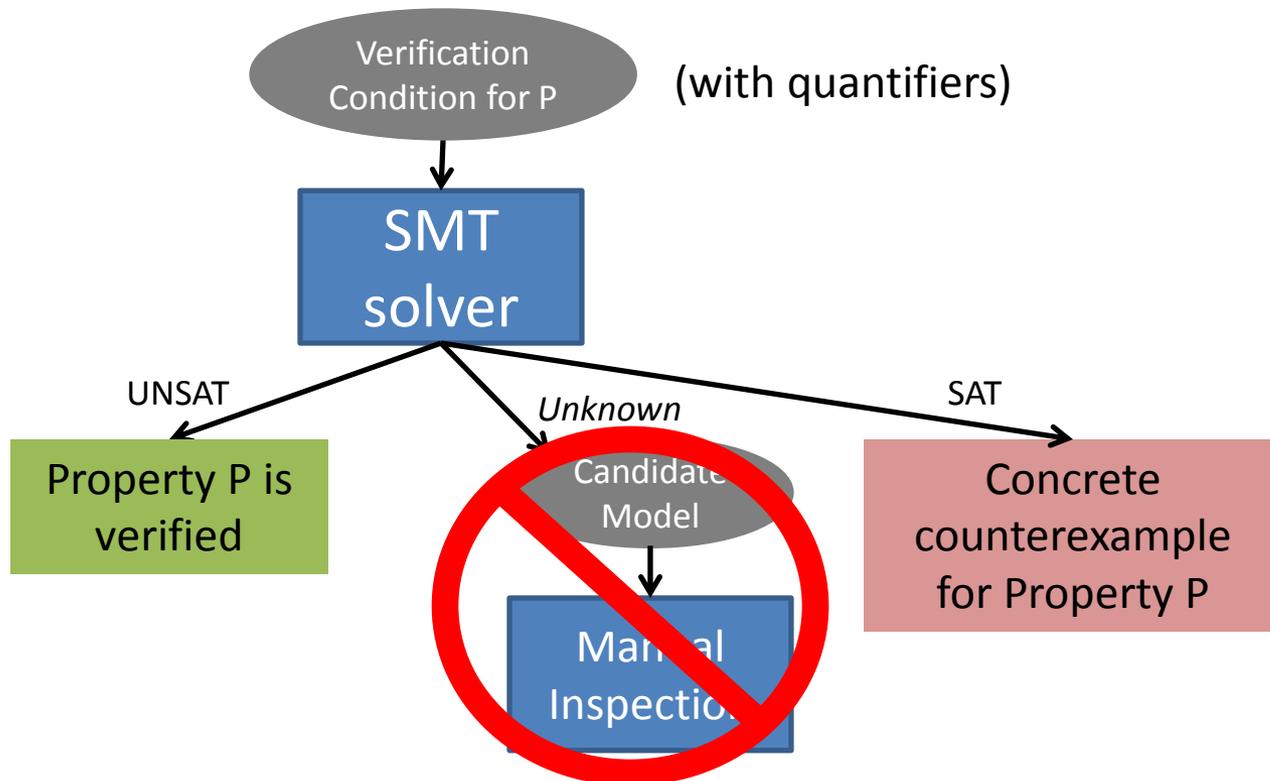
- Challenge: What if determining the consistency of \mathbb{A} is difficult?
- For quantified formulas, determining consistency is *undecidable*

SMT with Quantified Formulas

- When quantified formulas are asserted
 - Most SMT solvers will:
 - Answer unsat, if they happen to find a proof
 - Run indefinitely
 - Give up, reporting “unknown”

Why Models are Important

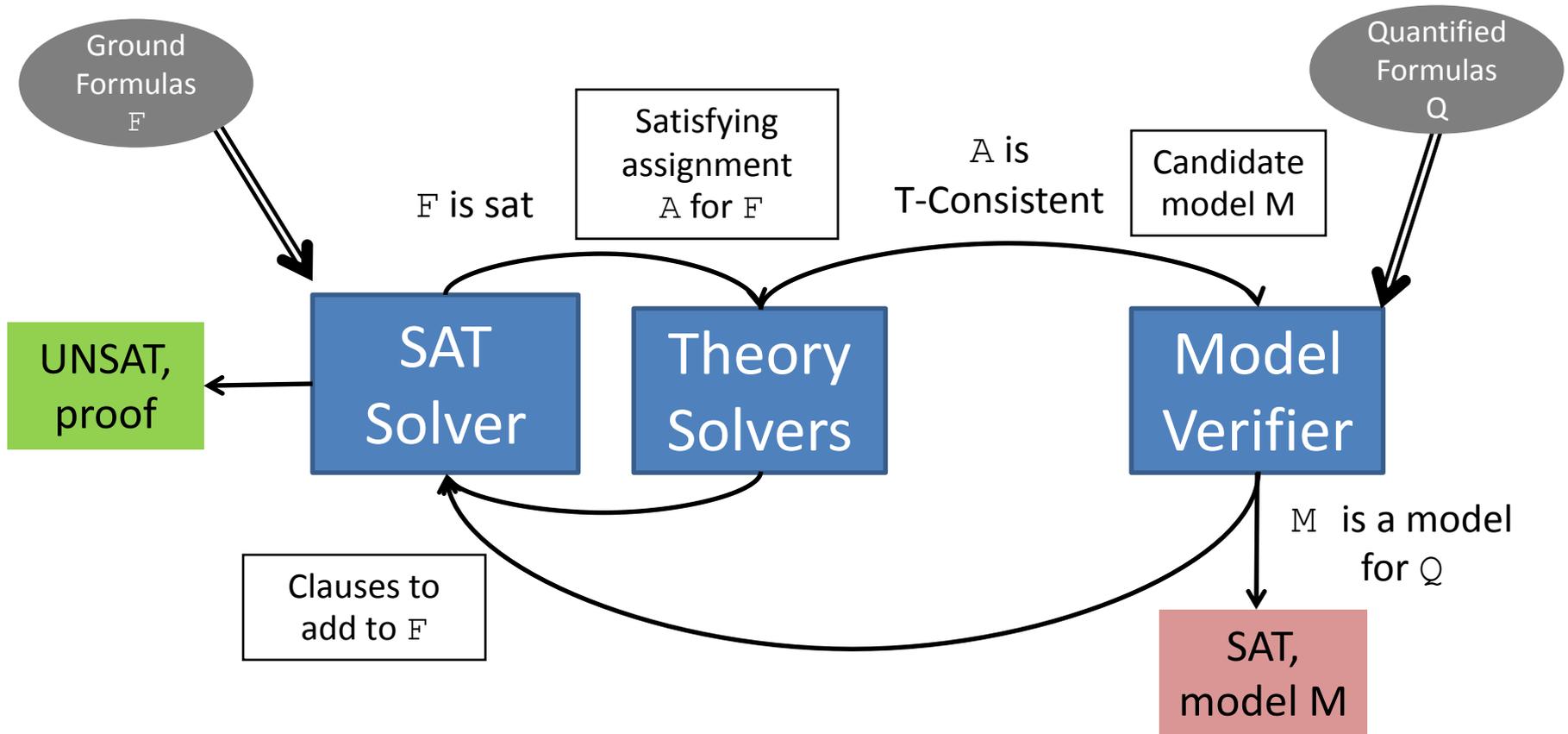
⇒ *Solver needs way of answering satisfiable when quantified formulas are asserted*



Model-Based Approach for Quantifiers

- Given:
 - Set of ground formulas \mathbb{F}
 - Set of quantified axioms \mathbb{Q}
- Determine the satisfiability of $\mathbb{F} \wedge \mathbb{Q}$
- Idea:
 - Construct candidate models for \mathbb{Q} based on satisfying assignments for \mathbb{F}
 - Ge and deMoura [2009]

Model-Based Approach for Quantifiers



Running Example

$\text{person}_1, \text{person}_2, \text{person}_3 : \text{Person}$

$\text{blue}, \text{brown}, \text{blonde} : \text{Color}$

$\text{eye}, \text{hair} : \text{Person} \rightarrow \text{Color}$

$\text{distinct}(\text{blue}, \text{brown}, \text{blonde})$

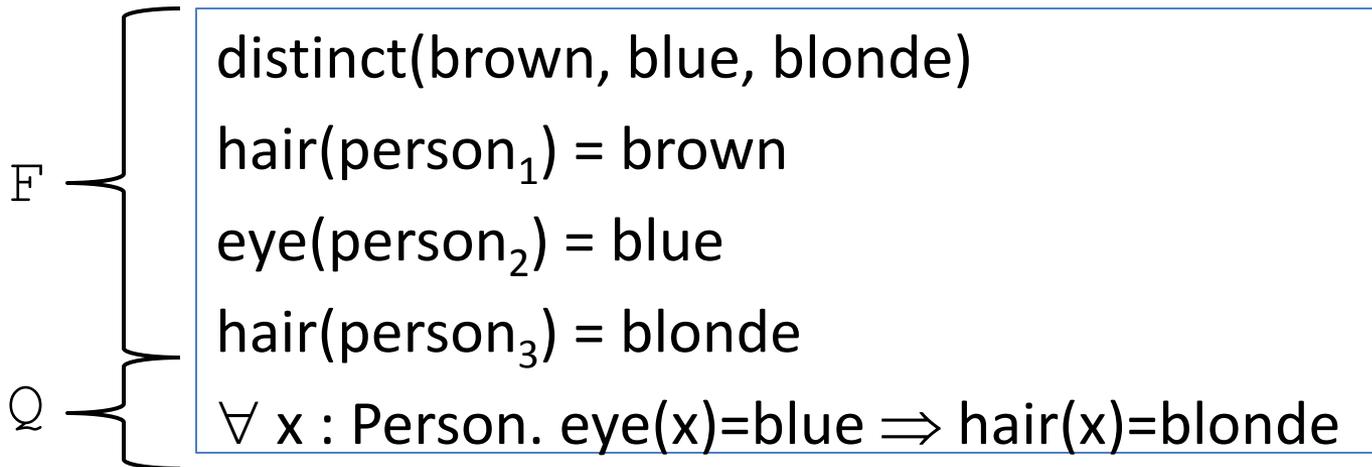
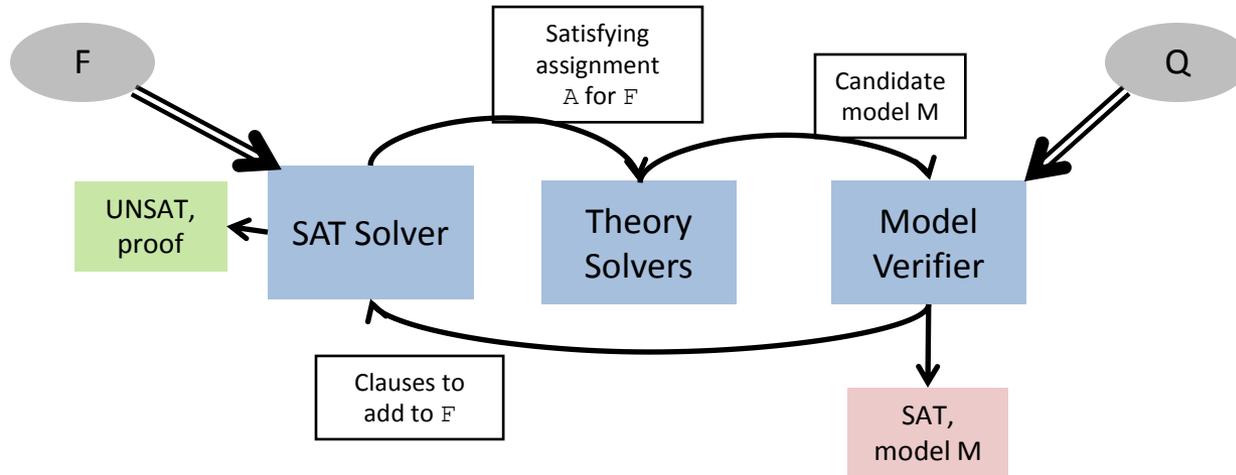
$\text{hair}(\text{person}_1) = \text{brown}$

$\text{eye}(\text{person}_2) = \text{blue}$

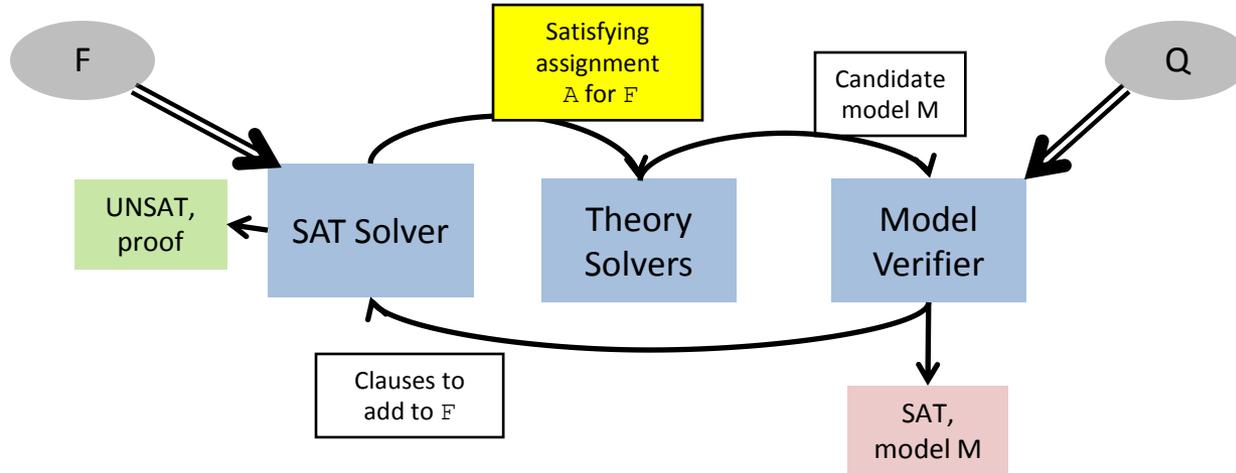
$\text{hair}(\text{person}_3) = \text{blonde}$

$\forall x : \text{Person}. \text{eye}(x) = \text{blue} \Rightarrow \text{hair}(x) = \text{blonde}$

Running Example



Find Satisfying Assignment A for F



true { **distinct(brown, blue, blonde)**

true { **hair(person₁) = brown**

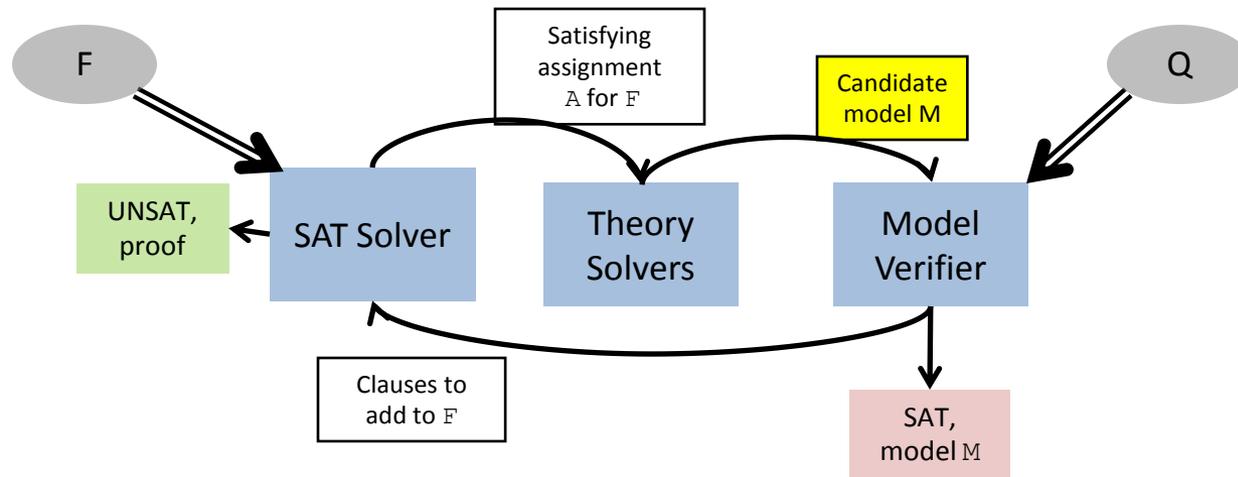
true { **eye(person₂) = blue**

true { **hair(person₃) = blonde**

$\forall x : \text{Person. eye}(x)=\text{blue} \Rightarrow \text{hair}(x)=\text{blonde}$

- A is Theory-Consistent according to the theory of equality

From \mathbb{A} , construct candidate model \mathbb{M}



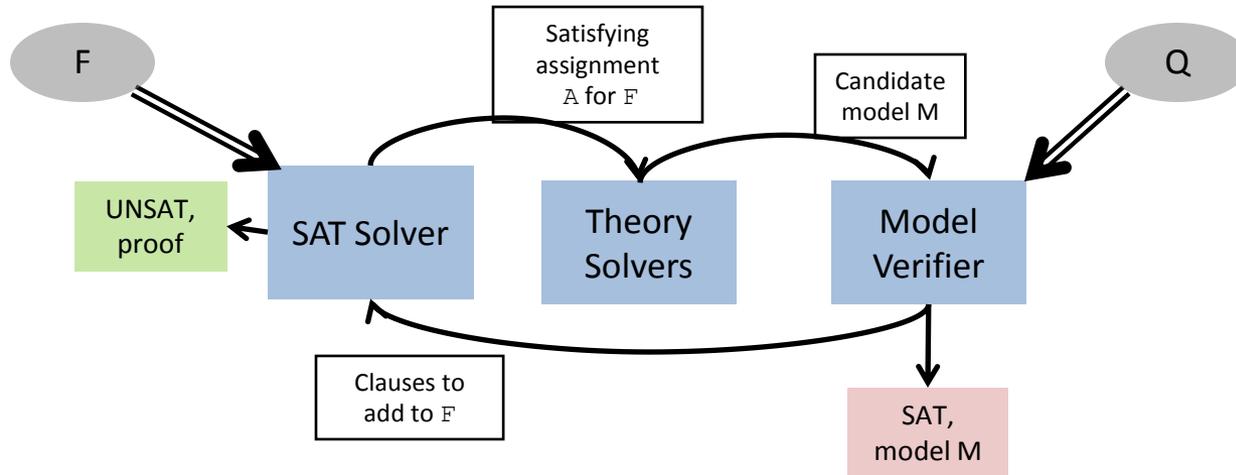
$\mathbb{A} :=$

{ distinct(brown, blue, blonde),
hair(person₁) = brown,
eye(person₂) = blue,
hair(person₃) = blonde }

$\mathbb{M} :=$

hair : person₁ -> brown
 person₃ -> blonde
 else -> brown
eye : person₂ -> blue
 else -> blue

Check whether M is a model of Q



$M :=$

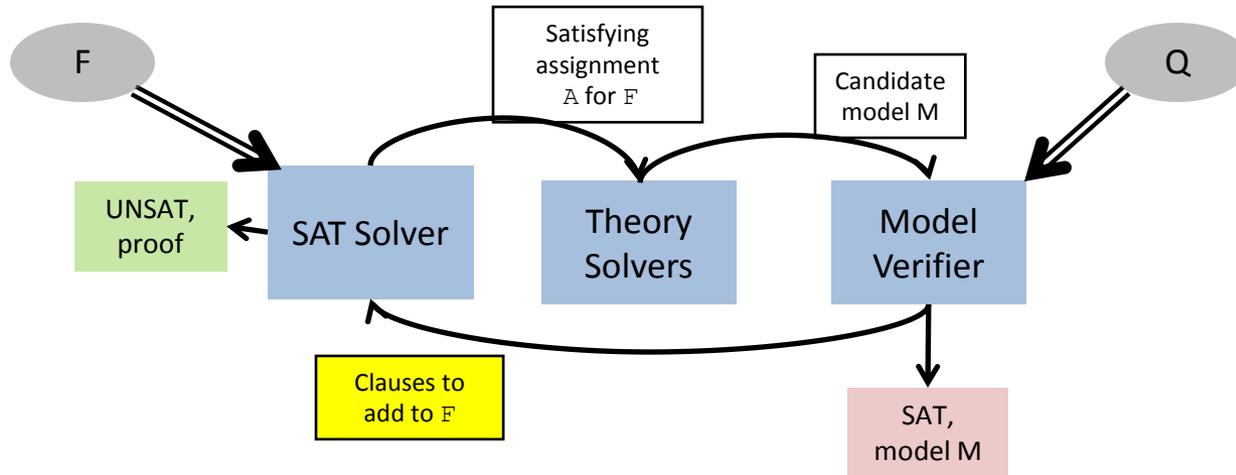
hair : person₁ -> brown
 person₃ -> blonde
 else -> brown
eye : person₂ -> blue
 else -> blue

$Q :=$

$\forall x : \text{Person. eye}(x)=\text{blue} \Rightarrow \text{hair}(x)=\text{blonde}$

- Q is false for person₂

Check whether M is a model of Q



$M :=$

hair : person₁ -> brown
 person₃ -> blonde
 else -> brown
eye : person₂ -> blue
 else -> blue

$Q :=$

$\forall x : \text{Person. eye}(x)=\text{blue} \Rightarrow \text{hair}(x)=\text{blonde}$

- Q is false for person₂

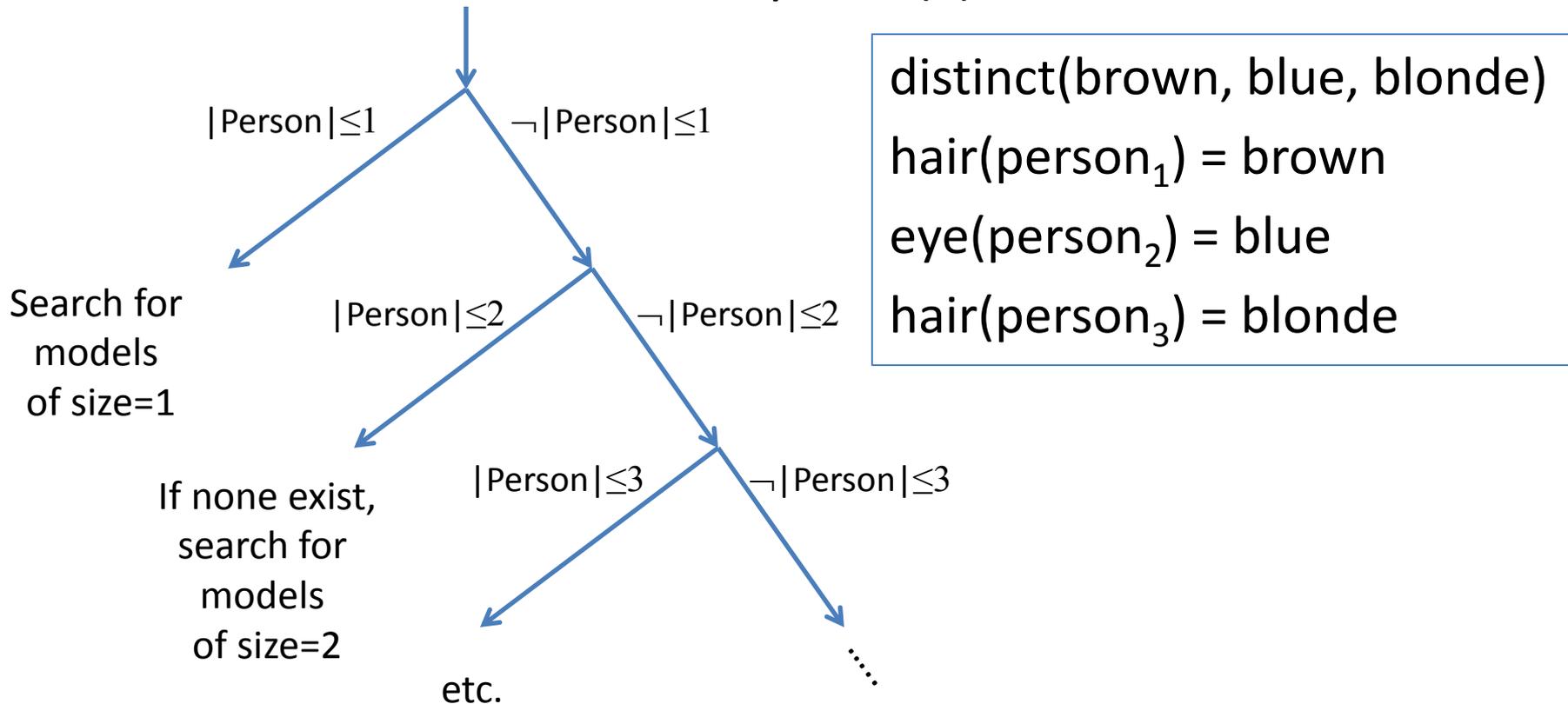
- Add $(\text{eye}(\text{person}_2)=\text{blue} \Rightarrow \text{hair}(\text{person}_2)=\text{blonde})$ to F
- Will rule out M on next iteration
 - Can be thought of as model “refinement”

What are good candidate models?

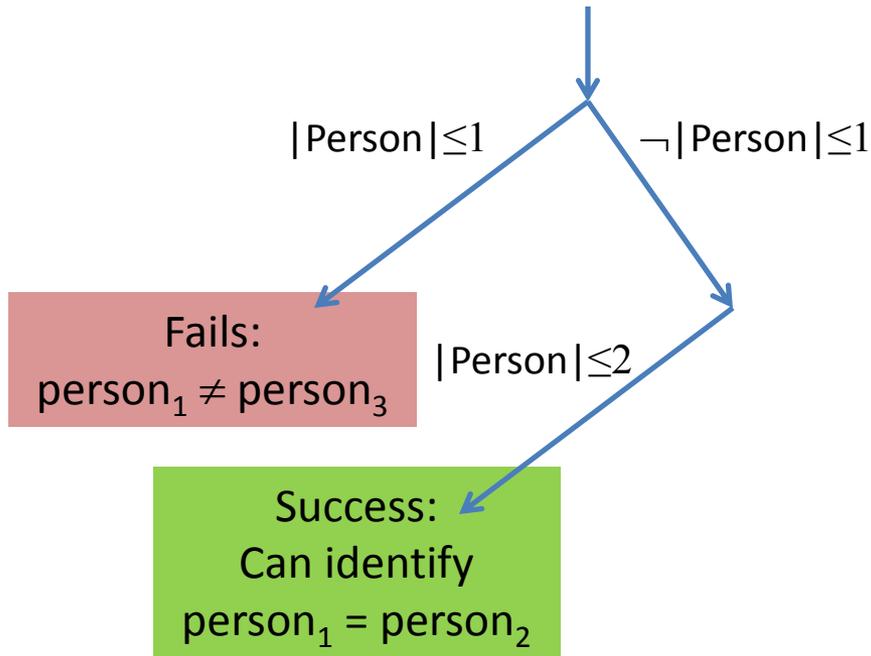
- Good candidate models
 - Have small domain sizes
 - Most instances of axioms \mathcal{Q} are likely to be true
- For small domain sizes,
 - Use specialized theory solver within DPLL(T)
- For making most instances true,
 - Use ground solver to guide model construction
- These features are implemented in SMT solver CVC4

Finding Minimal Models in DPLL(T) Search

- Idea: try to fix domain sizes 1,2,3,....
 - Prioritize decisions made by DPLL(T) search



Finding Minimal Models in DPLL(T) Search



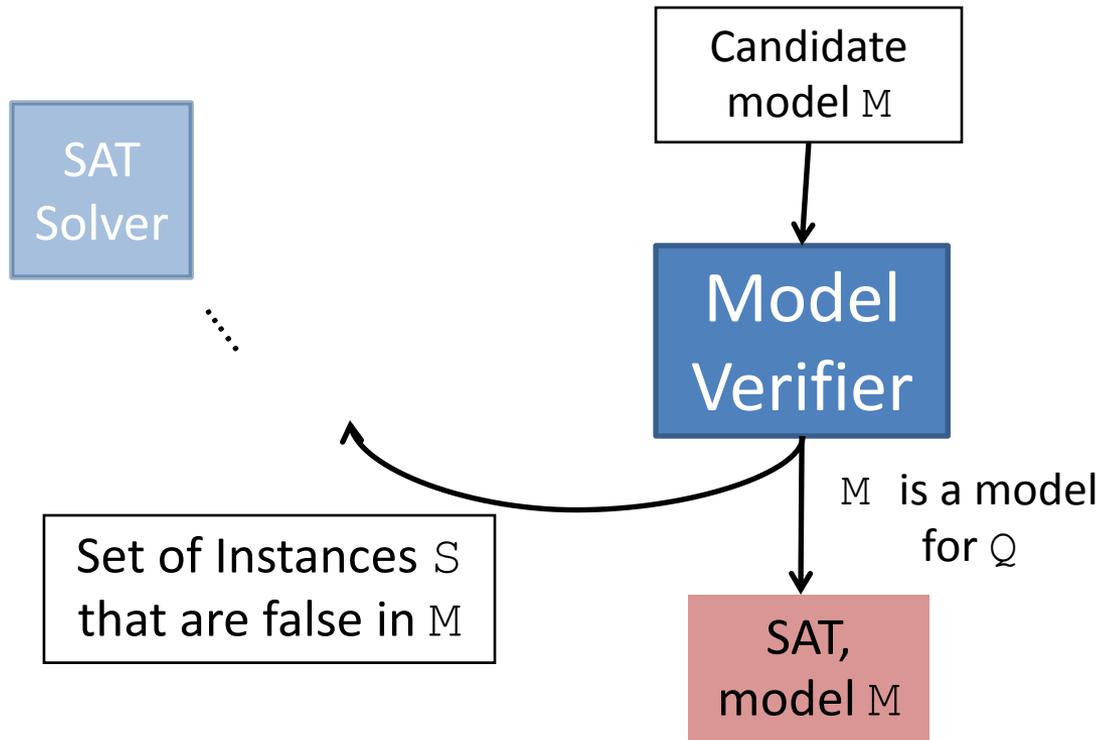
distinct(brown, blue, blonde)
hair(person₁) = brown
eye(person₂) = blue
hair(person₃) = blonde

- Implementation in CVC4 uses:

- Splitting on demand to shrink model sizes
- Efficient methods for clique detection

⇒ *Theory of finite cardinality constraints [CAV 2013]*

Constructing Good Candidate Models



- Set S can be very large
 - For Q with n variables with domain size d , $|S|$ can be $O(d^n)$
- Would prefer if most instances of Q are true in M

Constructing Good Candidate Models

- Idea for axiom Q :
 - Chose default values in model M based on one satisfying ground instance of Q

Q {

distinct(brown, blue, blonde)
hair(person₁) = brown
eye(person₂) = blue
hair(person₃) = blonde
 $\forall x : \text{Person. eye}(x)=\text{blue} \Rightarrow \text{hair}(x)=\text{blonde}$

- See how Q is satisfied for one instance, then generalize this [CADE 2013]

Constructing Good Candidate Models

distinct(brown, blue, blonde)

hair(person₁) = brown

eye(person₂) = blue

hair(person₃) = blonde

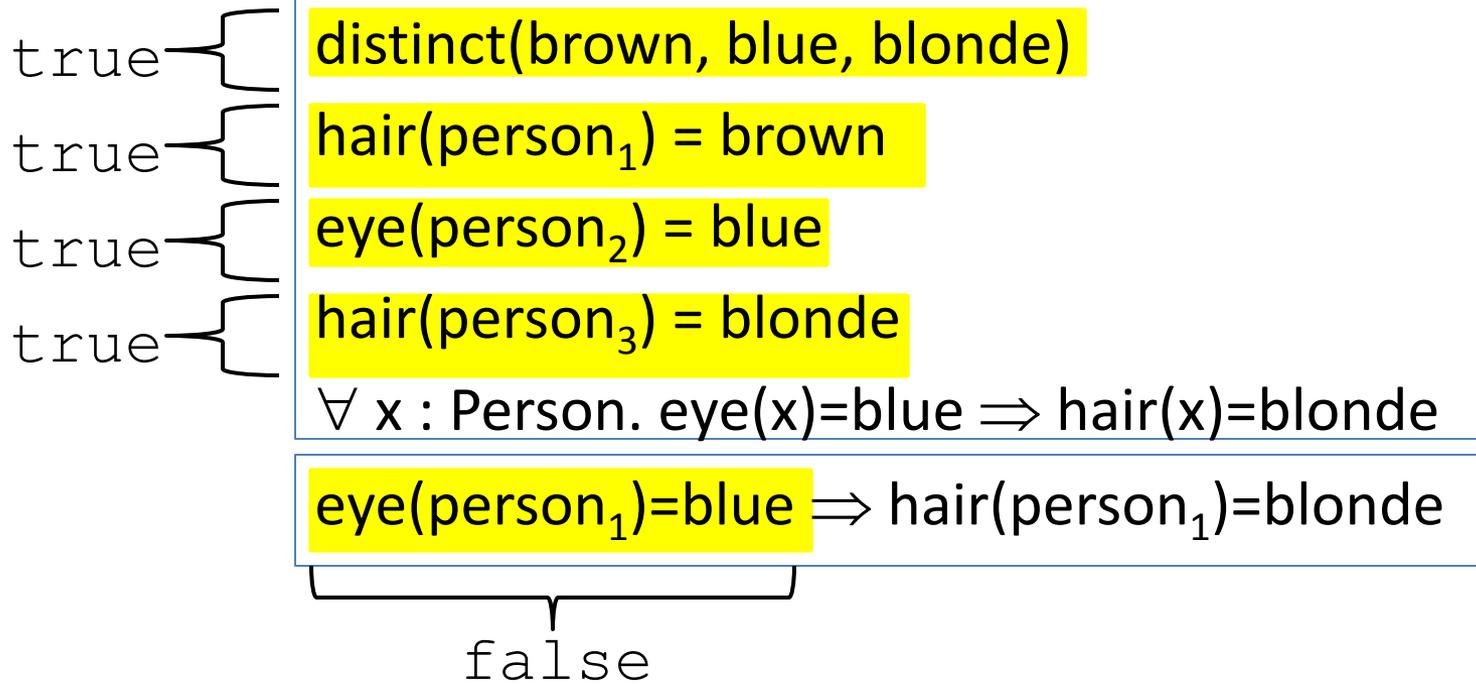
$\forall x : \text{Person}. \text{eye}(x)=\text{blue} \Rightarrow \text{hair}(x)=\text{blonde}$

$\text{eye}(\text{person}_1)=\text{blue} \Rightarrow \text{hair}(\text{person}_1)=\text{blonde}$

Q

- Consider $Q[\text{person}_1/x]$

Constructing Good Candidate Models



- Find satisfying assignment

Constructing Good Candidate Models

distinct(brown, blue, blonde)

hair(person₁) = brown

eye(person₂) = blue

hair(person₃) = blonde

$\forall x : \text{Person}. \text{eye}(x)=\text{blue} \Rightarrow \text{hair}(x)=\text{blonde}$

$\text{eye}(\text{person}_1)=\text{blue} \Rightarrow \text{hair}(\text{person}_1)=\text{blonde}$

- Construct candidate model

A :=

{ distinct(brown, blue, blonde),

hair(person₁) = brown,

eye(person₂) = blue,

hair(person₃) = blonde,

eye(person₁) ≠ blue }

M := hair : person₁ -> brown

person₃ -> blonde

else -> ...

eye : person₁ -> brown

person₂ -> blue

else -> ...

Constructing Good Candidate Models

distinct(brown, blue, blonde)

hair(person₁) = brown

eye(person₂) = blue

hair(person₃) = blonde

$\forall x : \text{Person. eye}(x)=\text{blue} \Rightarrow \text{hair}(x)=\text{blonde}$

eye(person₁)=blue \Rightarrow hair(person₁)=blonde

A :=

{ distinct(brown, blue, blonde),

hair(person₁) = brown,

eye(person₂) = blue,

hair(person₃) = blonde,

eye(person₁) \neq blue }

M := hair : person₁ -> brown

person₃ -> blonde

else -> brown

eye : person₁ -> **brown**

person₂ -> blue

else -> **brown**

Model-Based Approach in CVC4

- CVC4 is state of the art SMT solver with
 - Support for many theories
- Features implemented in CVC4:
 - Theory solver for handling cardinality constraints
 - Techniques for constructing candidate models
 - Efficient methods for verifying candidate models
 - Not mentioned in this talk

Experiments

- DVF Benchmarks
 - Taken from verification tool DVF used by Intel
 - Both SAT/UNSAT benchmarks
 - SAT benchmarks generated by removing necessary pf assumptions
 - Many theories: UF, arithmetic, arrays, datatypes
 - Quantifiers only over free sorts
 - Memory addresses, Values, Sets, ...
- TPTP Benchmarks
 - Unsorted, equality, function symbols
 - Heavy use of quantifiers

Experiments: DVF

SAT	german	refcount	agree	apg	bmh	Total
#	45	6	42	19	37	149
cvc3	0	0	0	0	0	0
yices	2	0	0	0	0	2
z3	45	1	0	0	0	46
cvc4	2	0	0	0	0	2
cvc4+f	45	6	42	19	37	149

UNSAT	german	refcount	agree	apg	bmh	Total
#	145	40	488	304	244	1221
cvc3	145	40	457	267	229	1138
yices	145	40	488	304	244	1221
z3	145	40	488	304	244	1221
cvc4	145	40	484	304	244	1217
cvc4+f	145	40	471	300	242	1198

- Configurations :
 - cvc4 : heuristic inst.
 - cvc4+f : model-based

- cvc4+f effective for sat
- cvc4+f solves 4 unsat that cvc4 cannot

Experiments: TPTP

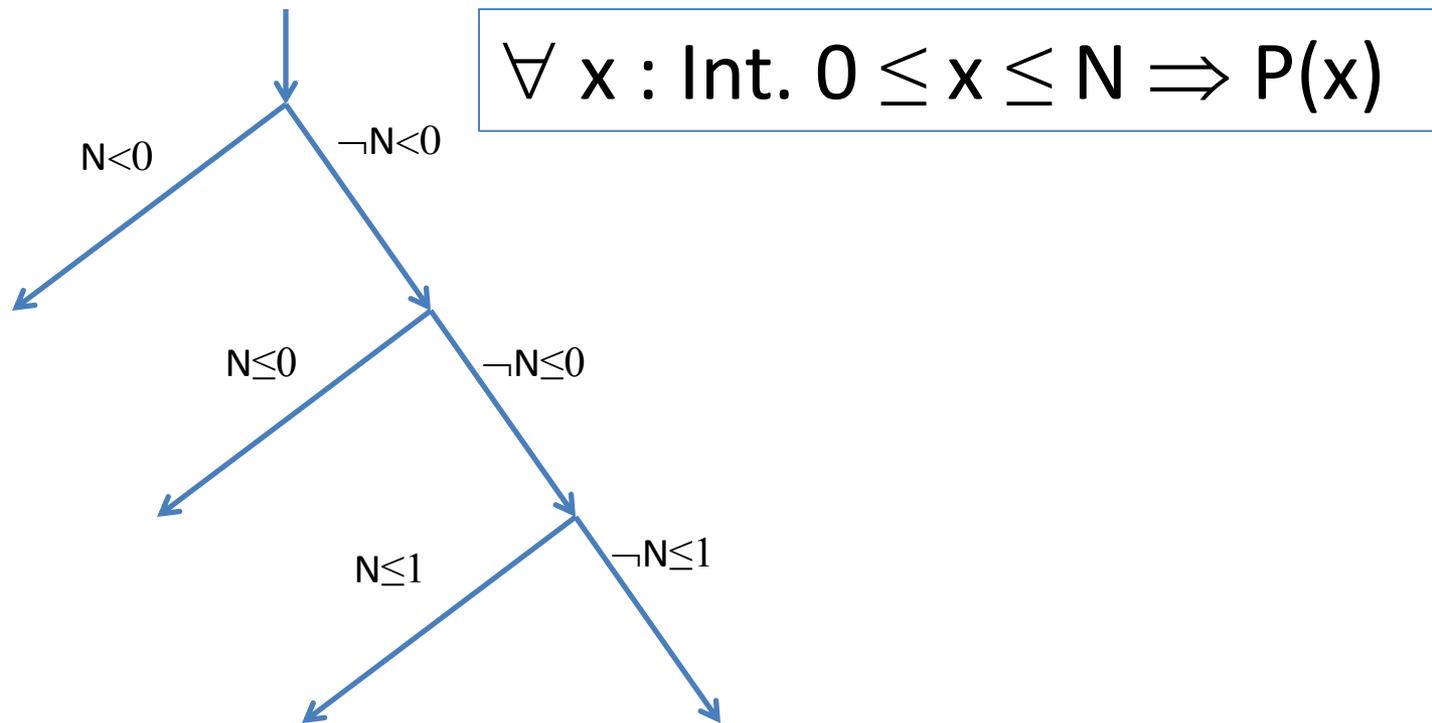
- For 1995 satisfiable benchmarks:
 - Paradox solves 1305
 - iProver solves 1231
 - z3 solves 887
 - cvc4+f solves 1186
 - Includes 3 problems with rating 1.0
- For 12568 unsatisfiable benchmarks:
 - z3 solves 5934
 - iProver solves 5556
 - cvc4 solves 5415
 - cvc4+f solves 3028
 - Orthogonal to other approaches
 - 282 cannot be solved by z3

Summary

- Completed work in CVC4:
 - Ground solver for finding small models
 - Methods for constructing and verifying candidate models
- Publicly available : <http://cvc4.cs.nyu.edu/>
- Current work:
 - Fair strategies for minimizing models for multiple sorts
 - Improve existing approaches for answering UNSAT
 - Other applications
 - Theory of Strings : bounded length
 - Integer quantification within bounded ranges

Current Work

- Extension to bounded integer quantification
 - Can use similar approach



Thanks

- Collaborators:
 - Cesare Tinelli, Amit Goel, Sava Krstic, Clark Barrett, Morgan Deters, Leonardo de Moura
- Questions?