Satisfiability Modulo Theories Summer School on Formal Methods

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Roadmap

- Logic Background
- Modern SAT Solvers
- DPLL with Theory Solvers
- Theory Combination
- Equality
- Arithmetic
- Applications

Satisfiability Modulo Theories (SMT)

- In SMT solving, the Boolean atoms represent constraints over individual theory variables (ranging over integer, reals, bit-vectors, datatypes, arrays, etc.).
- The constraints can involve theory operations, equality, and inequality.
- Now, the SAT solver has to interact with theory solvers.
- The constraint solver can detect conflicts involving theory reasoning, e.g., f(x) ≠ f(y), x = y, or x - y ≤ 2, y - z ≤ -1, z - x ≤ -3.
- The theory solver must support incremental assertions, efficient backtracking and propagation, and produce efficient explanations of unsatisfiability.

Theory Solver: Examples

- Equality: x = y (union-find), and offset equalities x = y + k.
- Term equality: congruence closure for uninterpreted function symbols.
- Difference constraints: incremental negative cycle detection for inequality constraints of the form $x y \le k$.
- Linear arithmetic: Fourier's method, Simplex.

Theory Solver: Rules

- We use $F \models_T G$ to denote the fact that F entails G in theory T.
- Abstract DPLL can be extended with two new rules to deal with theory T:

T-Propagate

$$M \parallel F \implies M \, l_{(\neg l_1 \lor \ldots \lor \neg l_n \lor l)} \parallel F \quad \text{if} \begin{cases} l \text{ occurs in } F, \\ l \text{ is undefined in } M, \\ l_1 \land \ldots \land l_n \models_T l, \\ l_1, \ldots, l_n \in \textit{lits}(M) \end{cases}$$

T-Conflict

 $M \parallel F \implies M \parallel F \parallel \neg l_1 \lor \ldots \lor \neg l_n \quad \text{if} \begin{cases} l_1 \land \ldots \land l_n \models_T \text{ false,} \\ l_1, \ldots, l_n \in \text{lits}(M) \end{cases}$

$$p \equiv 3 < x$$
$$q \equiv x < 0$$
$$r \equiv x < y$$

 $s \equiv y < 0$

 $\| \quad p, \ q \lor r, \ s \lor \neg r$

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$$p_p \ \| \quad p, \ q \lor r, \ s \lor \neg r \quad \Rightarrow \quad \text{(T-Propagate)}$$

$$p_p \neg q_{\neg p \lor \neg q} \ \| \quad p, \ q \lor r, \ s \lor \neg r$$

$$\underbrace{3 < x}_{p} \text{ implies } \neg \underbrace{x < 0}_{q}$$

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$$\begin{array}{cccc} \| & p, q \lor r, s \lor \neg r \Rightarrow & (\text{UnitPropagate}) \\ p_p \| & p, q \lor r, s \lor \neg r \Rightarrow & (\text{T-Propagate}) \\ p_p \neg q_{\neg p \lor \neg q} \| & p, q \lor r, s \lor \neg r \Rightarrow & (\text{UnitPropagate}) \\ p_p \neg q_{\neg p \lor \neg q} r_{q \lor r} \| & p, q \lor r, s \lor \neg r \end{array}$$

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- Trade-off between precision and performance.
- What is the minimal functionality of a theory solver?
 - Check the unsatisfiability of conjunction of literals.
- Efficiently mining T-justifications

T-Propagate

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The Ideal Theory Solver

- Incremental
- Efficient Backtracking
- Efficient T-Propagate
- Precise T-Justifications

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Combination of Theories

- In practice, we need a combination of theories.
- Example:

$$x + 2 = y \Rightarrow f(\textit{read}(\textit{write}(a, x, 3), y - 2)) = f(y - x + 1)$$

Given

$$\begin{split} \Sigma &= \Sigma_1 \cup \Sigma_2 \\ \mathcal{T}_1, \mathcal{T}_2 &: \text{ theories over } \Sigma_1, \Sigma_2 \\ \mathcal{T} &= \textit{DC}(\mathcal{T}_1 \cup \mathcal{T}_2) \end{split}$$

- Is \mathcal{T} consistent?
- Given satisfiability procedures for conjunction of literals of \mathcal{T}_1 and \mathcal{T}_2 , how to decide the satisfiability of \mathcal{T} ?

Preamble

- Disjoint signatures: $\Sigma_1 \cap \Sigma_2 = \emptyset$.
- Purification
- Stably-Infinite Theories.
- Convex Theories.

- Goal: convert a formula φ into φ₁ ∧ φ₂, where φ₁ is in T₁'s language and φ₂ is in T₂'s language.
 So φ₁ and φ₂ have no common symbols, except variables.
- Purification step: replace term t by a fresh variable x $\phi \wedge F(\dots, s[t], \dots) \rightsquigarrow \phi \wedge F(\dots, s[x], \dots) \wedge x = t$,
- Purification is satisfiability preserving and terminating.

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After Purification

$x = f(z), f(x) \neq f(y), 0 \le x \le 1, 0 \le y \le 1, z = y - 1$

$x = f(z), f(x) \neq f(y), 0 \le x \le 1, 0 \le y \le 1, z = y - 1$

Red Model	Blue Model
$ R = \{*_1, \dots, *_6\}$	$ B = \{\dots, -1, 0, 1, \dots\}$
$R(x) = *_1$	B(x) = 0
$R(y) = *_2$	B(y) = 0
$R(z) = *_3$	B(z) = -1
$R(f) = \{ *_1 \mapsto *_4, $	
$*_2 \mapsto *_5,$	
$*_3 \mapsto *_1,$	
$else \mapsto *_6\}$	

Stably-Infinite Theories

- A theory is stably infinite if every satisfiable QFF is satisfiable in an infinite model.
- Example. Theories with only finite models are not stably infinite. $T_2 = DC(\forall x, y, z. (x = y) \lor (x = z) \lor (y = z)).$
- The union of two consistent, disjoint, stably infinite theories is consistent.

• A theory \mathcal{T} is convex iff

for all finite sets Γ of literals and for all non-empty disjunctions $\bigvee_{i \in I} x_i = y_i$ of variables, $\Gamma \models_{\mathcal{T}} \bigvee_{i \in I} x_i = y_i$ iff $\Gamma \models_{\mathcal{T}} x_i = y_i$ for some $i \in I$.

- Every convex theory \mathcal{T} with non trivial models (i.e., $\models_T \exists x, y. \ x \neq y$) is stably infinite.
- All Horn theories are convex this includes all (conditional) equational theories.
- Linear rational arithmetic is convex.

- Many theories are not convex:
 - Linear integer arithmetic.

$$y = 1, z = 2, 1 \le x \le 2 \models x = y \lor x = z$$

Nonlinear arithmetic.

$$x^2=1, y=1, z=-1 \models x=y \lor x=z$$

- Theory of Bit-vectors.
- Theory of Arrays.

$$v_1 = \operatorname{read}(\operatorname{write}(a, i, v_2), j), v_3 = \operatorname{read}(a, j) \models v_1 = v_2 \lor v_1 = v_3$$

Nelson-Oppen Combination

- Let T₁ and T₂ be consistent, stably infinite theories over disjoint (countable) signatures. Assume satisfiability of conjunction of literals can decided in O(T₁(n)) and O(T₂(n)) time respectively. Then,
 - 1. The combined theory ${\mathcal T}$ is consistent and stably infinite.
 - 2. Satisfiability of quantifier free conjunction of literals in \mathcal{T} can be decided in $O(2^{n^2} \times (T_1(n) + T_2(n)))$.
 - 3. If \mathcal{T}_1 and \mathcal{T}_2 are convex, then so is \mathcal{T} and satisfiability in \mathcal{T} is in $O(n^3 \times (T_1(n) + T_2(n)))$.

Nelson-Oppen Combination Procedure

- The combination procedure:
 - **Initial State:** ϕ is a conjunction of literals over $\Sigma_1 \cup \Sigma_2$.
 - **Purification:** Preserving satisfiability transform ϕ into $\phi_1 \wedge \phi_2$, such that, $\phi_i \in \Sigma_i$.
 - Interaction: Guess a partition of $\mathcal{V}(\phi_1) \cap \mathcal{V}(\phi_2)$ into disjoint subsets. Express it as conjunction of literals ψ . Example. The partition $\{x_1\}, \{x_2, x_3\}, \{x_4\}$ is represented as $x_1 \neq x_2, x_1 \neq x_4, x_2 \neq x_4, x_2 = x_3$.
 - Component Procedures : Use individual procedures to decide whether $\phi_i \wedge \psi$ is satisfiable.
 - **Return:** If both return yes, return yes. No, otherwise.

NO procedure: soundness

- Each step is satisfiability preserving.
- Say ϕ is satisfiable (in the combination).
 - Purification: $\phi_1 \wedge \phi_2$ is satisfiable.

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 - Iteration: for some partition ψ , $\phi_1 \wedge \phi_2 \wedge \psi$ is satisfiable.

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 - Component procedures: $\phi_1 \wedge \psi$ and $\phi_2 \wedge \psi$ are both satisfiable in component theories.
 - Therefore, if the procedure return unsatisfiable, then ϕ is unsatisfiable.

- Suppose the procedure returns satisfiable.
 - Let ψ be the partition and A and B be models of $\mathcal{T}_1 \wedge \phi_1 \wedge \psi$ and $\mathcal{T}_2 \wedge \phi_2 \wedge \psi$.

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 - Let *h* be a bijection between |A| and |B| such that h(A(x)) = B(x) for each shared variable.
 - Extend B to \overline{B} by interpretations of symbols in Σ_1 : $\overline{B}(f)(b_1, \ldots, b_n) = h(A(f)(h^{-1}(b_1), \ldots, h^{-1}(b_n)))$

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 - \bar{B} is a model of:

 $\mathcal{T}_1 \wedge \phi_1 \wedge \mathcal{T}_2 \wedge \phi_2 \wedge \psi$

NO deterministic procedure

Instead of guessing, we can deduce the equalities to be shared.
Purification: no changes.
Interaction: Deduce an equality x = y:

$$\mathcal{T}_1 \vdash (\phi_1 \Rightarrow x = y)$$

Update $\phi_2 := \phi_2 \wedge x = y$. And vice-versa. Repeat until no further changes.

- **Component Procedures** : Use individual procedures to decide whether ϕ_i is satisfiable.
- Remark: $\mathcal{T}_i \vdash (\phi_i \Rightarrow x = y)$ iff $\phi_i \land x \neq y$ is not satisfiable in \mathcal{T}_i .

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 - By convexity, $\mathcal{T}_i \not\vdash \phi_i \Rightarrow \bigvee_E x_j = x_k$.

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$$\phi_i \wedge \bigwedge_E x_j \neq x_k$$
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 - By convexity, $\mathcal{T}_i \not\vdash \phi_i \Rightarrow \bigvee_E x_j = x_k$.
 - $\phi_i \wedge \bigwedge_E x_j \neq x_k$ is satisfiable.
 - The proof now is identical to the nondeterministic case.
 - Sharing equalities is sufficient, because a theory T₁ can assume that x^B ≠ y^B whenever x = y is not implied by T₂ and vice versa.

 $x + 2 = y \land f(\mathit{read}(\mathit{write}(a, x, 3), y - 2)) \neq f(y - x + 1)$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$

NO procedure: example

$$f(\mathit{read}(\mathit{write}(a, x, \mathbf{3}), y - 2)) \neq f(y - x + 1)$$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
	x + 2 = y	

NO procedure: example

 $f(\textit{read}(\textit{write}(a, x, u_1), y - 2)) \neq f(y - x + 1)$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
	x + 2 = y	
	$u_1 = 3$	

NO procedure: example

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${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
	x + 2 = y	
	$u_1 = 3$	
	$u_2 = y - 2$	

$$f(u_3) \neq f(y - x + 1)$$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
	x + 2 = y	$u_3 =$
	$u_1 = 3$	$\mathit{read}(\mathit{write}(a,x,u_1),u_2)$
	$u_2 = y - 2$	

 $f(u_3) \neq f(u_4)$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
	x + 2 = y	$u_3 =$
	$u_1 = 3$	$\mathit{read}(\mathit{write}(a,x,u_1),u_2)$
	$u_2 = y - 2$	
	$u_4 = y - x + 1$	

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
$f(u_3) \neq f(u_4)$	x + 2 = y	$u_3 =$
	$u_1 = 3$	$\mathit{read}(\mathit{write}(a,x,u_1),u_2)$
	$u_2 = y - 2$	
	$u_4 = \mathbf{y} - x + 1$	

Solving ${\mathcal T}_A$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
$f(u_3) \neq f(u_4)$	y = x + 2	$u_3 =$
	$u_1 = 3$	$\mathit{read}(\mathit{write}(a,x,u_1),u_2)$
	$u_2 = x$	
	$u_4 = 3$	

Propagating $u_2 = x$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
$f(u_3) \neq f(u_4)$	y = x + 2	$u_3 =$
$u_2 = x$	$u_1 = 3$	$\mathit{read}(\mathit{write}(a, \pmb{x}, u_1), \pmb{u_2})$
	$u_2 = x$	$u_2 = x$
	$u_4 = 3$	

Solving ${\mathcal T}_{Ar}$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
$f(u_3) \neq f(u_4)$	y = x + 2	$u_3 = u_1$
$u_2 = x$	$u_1 = 3$	$u_2 = x$
	$u_2 = x$	
	$u_4 = 3$	

Propagating $u_3 = u_1$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
$f(u_3) \neq f(u_4)$	y = x + 2	$u_3 = u_1$
$u_2 = x$	$u_1 = 3$	$u_2 = x$
$u_3 = u_1$	$u_2 = x$	
	$u_4 = 3$	
	$u_3 = u_1$	

Propagating $u_1 = u_4$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
$f(u_3) \neq f(u_4)$	y = x + 2	$u_3 = u_1$
$u_2 = x$	$u_1 = 3$	$u_2 = x$
$u_3 = u_1$	$u_2 = x$	
$u_4 = u_1$	$u_4 = 3$	
	$u_3 = u_1$	

Congruence $u_3 = u_1 \land u_4 = u_1 \Rightarrow f(u_3) = f(u_4)$

${\cal T}_E$	${\cal T}_A$	${\cal T}_{Ar}$
$f(u_3) \neq f(u_4)$	y = x + 2	$u_3 = u_1$
$u_2 = x$	$u_1 = 3$	$u_2 = x$
$u_3 = u_1$	$u_2 = x$	
$u_4 = u_1$	$u_4 = 3$	
$f(u_3) = f(u_4)$	$u_3 = u_1$	

Unsatisfiable!

NO deterministic procedure

- Deterministic procedure does not work for non convex theories.
- Example (integer arithmetic):

 $0 \leq x, y, z \leq 1, f(x) \neq f(y), f(x) \neq f(z), f(y) \neq f(z)$

(Expensive) solution: deduce disjunctions of equalities.

Combining theories in practice

- Propagate all implied equalities.
 - Deterministic Nelson-Oppen.
 - Complete only for convex theories.
 - It may be expensive for some theories.
- Delayed Theory Combination.
 - Nondeterministic Nelson-Oppen.
 - Create set of interface equalities (x = y) between shared variables.
 - Use SAT solver to guess the partition.
 - Disadvantage: the number of additional equality literals is quadratic in the number of shared variables.

Combining theories in practice (cont.)

- Common to these methods is that they are pessimistic about which equalities are propagated.
- Model-based Theory Combination
 - Optimistic approach.
 - Use a candidate model M_i for one of the theories T_i and propagate all equalities implied by the candidate model, hedging that other theories will agree.

if $M_i \models {\mathcal T}_i \cup \Gamma_i \cup \{u=v\}$ then propagate u=v .

- If not, use backtracking to fix the model.
- It is cheaper to enumerate equalities that are true in a particular model than the equalities implied by all models.

$$x = f(\mathbf{y-1}), f(x) \neq f(y), 0 \leq x \leq 1, 0 \leq y \leq 1$$

$$x = f(z), f(x) \neq f(y), 0 \le x \le 1, 0 \le y \le 1, z = y - 1$$

${\cal T}_E$			${\cal T}_A$	
Literals	Eq. Classes	Model	Literals	Model
x = f(z)	$\{x, f(z)\}$	$E(x) = *_1$	$0 \le x \le 1$	A(x) = 0
$f(x) \neq f(y)$	$\{y\}$	$E(y) = *_2$	$0 \le y \le 1$	A(y) = 0
	$\{z\}$	$E(z) = *_3$	z = y - 1	A(z) = -1
	$\{f(x)\}$	$E(f) = \{ *_1 \mapsto *_4, $		
	$\{f(y)\}$	$*_2 \mapsto *_5,$		
		$*_3 \mapsto *_1,$		
		$\textit{else}\mapsto *_6\}$		

${\cal T}_E$		${\cal T}_A$		
Literals	Eq. Classes	Model	Literals	Model
x = f(z)	$\{x, y, f(z)\}$	$E(x) = *_1$	$0 \le x \le 1$	A(x) = 0
$f(x) \neq f(y)$	$\{z\}$	$E(y) = *_1$	$0 \le y \le 1$	A(y) = 0
x = y	$\{f(x), f(y)\}$	$E(z) = *_2$	z = y - 1	A(z) = -1
		$E(f) = \{ *_1 \mapsto *_3, $	x = y	
		$*_2 \mapsto *_1,$		
		$\textit{else}\mapsto *_4\}$		

Unsatisfiable

${\cal T}_E$			${\cal T}_A$	
Literals	Eq. Classes	Model	Literals	Model
x = f(z)	$\{x, f(z)\}$	$E(x) = *_1$	$0 \le x \le 1$	A(x) = 0
$f(x) \neq f(y)$	$\{y\}$	$E(y) = *_2$	$0 \le y \le 1$	A(y) = 0
$x \neq y$	$\{z\}$	$E(z) = *_3$	z = y - 1	A(z) = -1
	$\{f(x)\}$	$E(f) = \{ *_1 \mapsto *_4, $	$x \neq y$	
	$\{f(y)\}$	$*_2 \mapsto *_5,$		
		$*_3 \mapsto *_1,$		
		$\textit{else}\mapsto *_6\}$		

Backtrack, and assert $x \neq y$. \mathcal{T}_A model need to be fixed.

${\cal T}_E$			${\cal T}_A$	
Literals	Eq. Classes	Model	Literals	Model
x = f(z)	$\{x, f(z)\}$	$E(x) = *_1$	$0 \le x \le 1$	A(x) = 0
$f(x) \neq f(y)$	$\{y\}$	$E(y) = *_2$	$0 \le y \le 1$	A(y) = 1
$x \neq y$	$\{z\}$	$E(z) = *_3$	z = y - 1	A(z) = 0
	$\{f(x)\}$	$E(f) = \{ *_1 \mapsto *_4, $	$x \neq y$	
	$\{f(y)\}$	$*_2 \mapsto *_5,$		
		$*_3 \mapsto *_1,$		
		$\textit{else}\mapsto *_6\}$		

Assume x = z

${\cal T}_E$		${\cal T}_A$		
Literals	Eq. Classes	Model	Literals	Model
x = f(z)	$\{x, z,$	$E(x) = *_1$	$0 \le x \le 1$	A(x) = 0
$f(x) \neq f(y)$	$f(x), f(z)\}$	$E(y) = *_2$	$0 \le y \le 1$	A(y) = 1
$x \neq y$	$\{y\}$	$E(z) = *_{1}$	z = y - 1	A(z) = 0
x = z	$\{f(y)\}$	$E(f) = \{ *_1 \mapsto *_1, $	$x \neq y$	
		$*_2 \mapsto *_3,$	x = z	
		$\textit{else}\mapsto *_4\}$		

Satisfiable
Model based theory combination: Example

${\cal T}_E$			${\cal T}_A$	
Literals	Eq. Classes	Model	Literals	Model
x = f(z)	$\{x, z,$	$E(x) = *_1$	$0 \le x \le 1$	A(x) = 0
$f(x) \neq f(y)$	f(x), f(z)	$E(y) = *_2$	$0 \le y \le 1$	A(y) = 1
$x \neq y$	$\{y\}$	$E(z) = *_1$	z = y - 1	A(z) = 0
x = z	$\{f(y)\}$	$E(f) = \{ *_1 \mapsto *_1, $	$x \neq y$	
		$*_2 \mapsto *_3,$	x = z	
		$\textit{else}\mapsto *_4\}$		

Let h be the bijection between |E| and |A|.

$$h = \{*_1 \mapsto 0, *_2 \mapsto 1, *_3 \mapsto -1, *_4 \mapsto 2, \ldots\}$$

Model based theory combination: Example

${\cal T}_E$		${\cal T}_A$		
Literals	Model	Literals	Model	
x = f(z)	$E(x) = *_1$	$0 \le x \le 1$	A(x) = 0	
$f(x) \neq f(y)$	$E(y) = *_2$	$0 \le y \le 1$	A(y) = 1	
$x \neq y$	$E(z) = *_1$	z = y - 1	A(z) = 0	
x = z	$E(f) = \{ *_1 \mapsto *_1, $	$x \neq y$	$A(f) = \{0 \mapsto 0$	
	$*_2 \mapsto *_3,$	x = z	$1\mapsto -1$	
	$\textit{else}\mapsto *_4\}$		$\mathit{else}\mapsto 2\}$	

Extending A using h.

$$h = \{ *_1 \mapsto 0, *_2 \mapsto 1, *_3 \mapsto -1, *_4 \mapsto 2, \ldots \}$$

Model mutation

• Sometimes M(x) = M(y) by accident.

$$\bigwedge_{i=1}^{N} f(x_i) \ge 0 \ \land \ x_i \ge 0$$

Model mutation: diversify the current model.

Roadmap

- Logic Background
- Modern SAT Solvers
- DPLL with Theory Solvers
- Theory Combination
- Equality
- Arithmetic
- Applications

Reflexivity x = x

Symmetry $x = y \Rightarrow y = x$

Transitivity $x = y, y = z \Rightarrow x = z$

Congruence

$$x_1 = y_1, \dots, x_n = y_n \Rightarrow f(x_1, \dots, x_n) = f(y_1, \dots, y_n)$$

$$f(f(a)) = a, \ b = f(a), \ \neg f(f(f(a))) = b$$

$$f(f(a)) = a, \ b = f(a), \ \neg f(f(f(a))) = b$$

congruence $\rightsquigarrow f(f(f(a))) = f(a)$

$$f(f(a)) = a, \ b = f(a), \ \neg f(f(f(a))) = b,$$

$$f(f(f(a))) = f(a)$$

symmetry $\rightsquigarrow f(a) = b$

SAT/SMT – p.31/57

$$f(f(a)) = a, \ b = f(a), \ \neg f(f(f(a))) = b,$$

$$f(f(f(a))) = f(a), \ f(a) = b$$

transitivity $\rightsquigarrow f(f(f(a))) = b$

$$f(f(a)) = a, \ b = f(a), \ \neg f(f(f(a))) = b,$$

$$f(f(f(a))) = f(a), \ f(a) = b, \ f(f(f(a))) = b$$

unsatisfiable

- A conjunction of equalities is trivially satisfiable.
- Example: f(x) = y, x = y, g(x) = z, f(y) = f(z)

- A conjunction of equalities is trivially satisfiable.
- Example: f(x) = y, x = y, g(x) = z, f(y) = f(z)
- Model:
 - $|M| = \{*_1\}$
 - $M(x) = M(y) = M(z) = *_1$
 - $M(f)(*_1) = *_1$
 - $M(g)(*_1) = *_1$

Variable equality

- Assume the problem has not function symbols.
- Use union-find data structure to represent equalities.
- The state consists of a find structure F that maintains equivalence classes and a set of disequalities D.
- Initially, F(x) = x for each variable x.
- $F^*(x)$ is the root of the equivalence class containing x:

$$F^*(x) = \begin{cases} x, & \text{ if } F(x) = x \\ F^*(F(x)) & \text{ otherwise} \end{cases}$$

• Let sz(F, x) denote the size of the equivalence class containing x.

Variable equality: union

An equality x = y is processed by merging distinct equivalence classes using the *union* operation:

$$\begin{array}{ll} \textit{union}(F,x,y) & = & \left\{ \begin{array}{ll} F[x':=y'], & \textit{sz}(F,x) < \textit{sz}(F,y) \\ & F[y':=x'], & \textit{otherwise} \end{array} \right. \\ & \text{where } x' \equiv F^*(x) \not\equiv F^*(y) \equiv y' \end{array} \right. \end{array}$$

Optimization: path compression, update F when executing F*(x).
F[x := F*(x)]

Processing equalities

The entire inference system consists of operations for adding equalities, disequalities, and dectecting unsatisfiability.

$$\begin{aligned} & \textit{addeq}(x = y, F, D) \quad := \quad \langle F, D \rangle, \text{ if } F^*(x) \equiv F^*(y) \\ & \textit{addeq}(x = y, F, D) \quad := \quad \begin{cases} & \textit{unsat}, & \textit{if } F'^*(u) \equiv F'^*(v) \text{ for some} \\ & u \neq v \in D \\ & \langle F', D \rangle, & \textit{otherwise} \\ & \textit{where } F^*(x) \not\equiv F^*(y) \\ & F' = \textit{union}(F, x, y) \end{aligned}$$

$$\begin{aligned} \text{addneq}(x \neq y, F, D) &:= \text{ unsat, if } F^*(x) \equiv F^*(y) \\ \text{addneq}(x \neq y, F, D) &:= \langle F, D \rangle, \text{ if} \\ F^*(x) = F^*(u), F^*(y) = F^*(v), \\ \text{ for } u \neq v \in D \text{ or } v \neq u \in D \end{aligned}$$

 $addneq(x \neq y, F, D) := \langle F, D \cup \{x \neq y\} \rangle$, otherwise

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_2, x_3 \mapsto x_3, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{\}$$

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_2, x_3 \mapsto x_3, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{\}$$

Merge equivalence classes of x_1 and x_2 .

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_3, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{\}$$

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_3, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{\}$$

Merge equivalence classes of x_1 and x_3 .

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_1, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{\}$$

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_1, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{\}$$

Skip equality

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_1, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{\}$$

Add disequality

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_1, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{x_2 \neq x_4\}$$

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_1, x_4 \mapsto x_4, x_5 \mapsto x_5\}$$
$$D = \{x_2 \neq x_4\}$$

Merge equivalence classes of x_4 and x_5 .

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_1, x_4 \mapsto x_4, x_5 \mapsto x_4\}$$
$$D = \{x_2 \neq x_4\}$$

$$x_1 = x_2, x_1 = x_3, x_2 = x_3, x_2 \neq x_4, x_4 = x_5$$

$$F = \{x_1 \mapsto x_1, x_2 \mapsto x_1, x_3 \mapsto x_1, x_4 \mapsto x_4, x_5 \mapsto x_4\}$$
$$D = \{x_2 \neq x_4\}$$

Model M: $|M| = \{*_1, *_2\}$ $M(x_1), M(x_2), M(x_3) = *_1$ $M(x_4), M(x_5) = *_2$

Equality with offsets

- Many terms are equal modulo a numeric offset (e.g., x = y + 1).
- If these are placed in separate equivalence classes, then the equality reasoning on these terms must invoke the arithmetic module.
- We can modify the *find* data structure so that F(x) returns y + c, and similarly F*(x).
- Example: $x_1 \neq x_2 + c$ if $F^*(x_1) = y + c_1$ and $F^*(x_2) = y + c_2$, where $c \neq c_1 c_2$.

Retracting assertions

- Checkpointing the find data structure can be expensive.
- A disequality can be retracted by just deleting it from D.
- Retracting equality assertions is more difficult, the history of the merge operations have to be maintained.
- On retraction, the find values have to be restored.

Congruence Closure

- Equivalence is extended to *congruence* with the rule that for each n-ary function f, $f(s_1, \ldots, s_n) = f(t_1, \ldots, t_n)$ if $s_i = t_i$ for each $1 \le 1 \le n$.
- New index: π(t) is the set of parents of the equivalence class rooted by t (aka use-list).
- Example:

 $\begin{cases} f(f(a)), \ g(a), \ a, \ g(b) \end{cases} F = \{ b \mapsto a, g(a) \mapsto g(b), \dots \} \\ \pi(a) &= \{ f(a), \ g(a), \ g(b) \} \\ \pi(f(a)) &= \{ f(f(a)) \} \\ \pi(g(a)) &= \emptyset \\ \pi(f(f(a))) &= \emptyset \\ \end{cases}$

Congruence Closure (cont.)

- As with equivalence, the *find* roots s' = F^{*}(s) and t' = F^{*}(t) are merged. The use lists π(s') and π(t') are also merged.
- How to merge use-lists?
 - 1. Use-lists are circular lists:
 - Constant time merge and unmerge.
 - 2. Use-lists are vectors:
 - Linear time merge: copy $\pi(s')$ to $\pi(t')$.
 - Constant time unmerge: shrink the vector.
 - 3. Do not merge: to traverse the set of parents, traverse the equivalence class.
- Any pair p₁ in π(s') and p₂ in π(t') that are congruent in F is added to a queue of equalities to be merged.

Congruence Closure (cont.)

- Any pair p₁ in π(s') and p₂ in π(t') that are congruent in F is added to a queue of equalities to be merged.
 - Naïve solution: for each p_i of π(s') traverse π(t') looking for a congruence p_j.
 - Efficient solution: congruence table.
 - Hashtable of ground terms.
 - Hash of $f(t_1, \ldots, t_n)$ is based on f, $F^*(t_1), \ldots, F^*(t_n)$
 - $f(s_1, \dots, s_n) = f(t_1, \dots, t_n)$ if $F^*(s_1) = F^*(t_1), \dots, F^*(s_n) = F^*(t_n)$
 - The operation F[x' := y'] affects the hashcode of π(x'), before executing it remove terms in π(x') from the table, and reinsert them back after.
 - Detect new congruences during reinsertion.

$$f(g(a)) = c, c \neq f(g(b)), a = b$$

$$F = \{a \mapsto a, b \mapsto b, c \mapsto c, g(a) \mapsto g(a), g(b) \mapsto g(b) \\ f(g(a)) \mapsto f(g(a)), f(g(b)) \mapsto f(g(b))\}$$

$$D = \{\}$$

$$\pi(a) = \{g(a)\}$$

$$\pi(b) = \{g(b)\}$$

$$\pi(g(a)) = \{f(g(a))\}\$$

$$\pi(g(b)) = \{f(g(b))\}\$$

$$f(g(a)) = c, c \neq f(g(b)), a = b$$

$$F = \{a \mapsto a, b \mapsto b, c \mapsto c, g(a) \mapsto g(a), g(b) \mapsto g(b) \\ f(g(a)) \mapsto f(g(a)), f(g(b)) \mapsto f(g(b))\}$$

$$D = \{\}$$

$$\pi(a) = \{g(a)\}$$

$$\pi(b) = \{g(b)\}\$$

$$\pi(g(a)) = \{f(g(a))\}\$$

$$\pi(g(b)) = \{f(g(b))\}\$$

Merge equivalence classes of f(g(a)) and c.

$$f(g(a)) = c, c \neq f(g(b)), a = b$$

$$F = \{a \mapsto a, b \mapsto b, c \mapsto c, g(a) \mapsto g(a), g(b) \mapsto g(b)$$
$$f(g(a)) \mapsto c, f(g(b)) \mapsto f(g(b))\}$$
$$D = \{\}$$

$$\pi(a) = \{g(a)\} \\ \pi(b) = \{g(b)\} \\ \pi(g(a)) = \{f(g(a))\} \\ \pi(g(b)) = \{f(g(b))\}$$

$$f(g(a)) = c, c \neq f(g(b)), a = b$$

$$F = \{a \mapsto a, b \mapsto b, c \mapsto c, g(a) \mapsto g(a), g(b) \mapsto g(b)$$
$$f(g(a)) \mapsto c, f(g(b)) \mapsto f(g(b))\}$$
$$D = \{\}$$
$$\pi(a) = \{g(a)\}$$
$$\pi(b) = \{g(b)\}$$

 $\pi(g(a)) = \{f(g(a))\}\$ $\pi(g(b)) = \{f(g(b))\}\$

Add disequality
$$f(g(a)) = c, c \neq f(g(b)), a = b$$

$$F = \{a \mapsto a, b \mapsto b, c \mapsto c, g(a) \mapsto g(a), g(b) \mapsto g(b) \\ f(g(a)) \mapsto c, f(g(b)) \mapsto f(g(b))\}$$

$$D = \{c \neq f(g(b))\}$$

$$\pi(a) = \{g(a)\}$$

$$\pi(b) = \{g(b)\}$$

$$\pi(g(a)) = \{f(g(a))\}$$

$$\pi(g(b)) = \{f(g(b))\}$$

$$f(g(a)) = c, c \neq f(g(b)), a = b$$

$$F = \{a \mapsto a, b \mapsto b, c \mapsto c, g(a) \mapsto g(a), g(b) \mapsto g(b) \\ f(g(a)) \mapsto c, f(g(b)) \mapsto f(g(b))\} \\ D = \{c \neq f(g(b))\} \\ \pi(a) = \{g(a)\} \\ \pi(b) = \{g(b)\} \\ \pi(g(a)) = \{f(g(a))\} \\ \pi(g(b)) = \{f(g(b))\}$$

Merge equivalence classes of a and b.

$$f(g(a)) = c, c \neq f(g(b)), a = b, g(a) = g(b)$$

$$F = \{a \mapsto a, b \mapsto a, c \mapsto c, g(a) \mapsto g(a), g(b) \mapsto g(b) \\ f(g(a)) \mapsto c, f(g(b)) \mapsto f(g(b))\}$$

$$D = \{c \neq f(g(b))\}$$

$$\pi(a) = \{g(a), \mathbf{g}(\mathbf{b})\}$$

$$\pi(b) = \{g(b)\}$$

$$\pi(g(a)) = \{f(g(a))\}$$

$$\pi(g(b)) = \{f(g(b))\}$$

$$f(g(a)) = c, c \neq f(g(b)), a = b, g(a) = g(b)$$

$$F = \{a \mapsto a, b \mapsto a, c \mapsto c, g(a) \mapsto g(a), g(b) \mapsto g(b) \\ f(g(a)) \mapsto c, f(g(b)) \mapsto f(g(b))\}$$

$$D = \{c \neq f(g(b))\}$$

$$\pi(a) = \{g(a), g(b)\}$$

$$\pi(b) = \{g(b)\}$$

 $\pi(g(a)) = \{f(g(a))\}\$ $\pi(g(b)) = \{f(g(b))\}\$

Merge equivalence classes of g(a) and g(b).

$$f(g(a)) = c, c \neq f(g(b)), a = b, g(a) = g(b), f(g(a)) = f(g(b))$$

$$F = \{a \mapsto a, b \mapsto a, c \mapsto c, g(a) \mapsto g(b), g(b) \mapsto g(b) \\ f(g(a)) \mapsto c, f(g(b)) \mapsto f(g(b))\}$$

$$D = \{c \neq f(g(b))\}$$

$$\pi(a) = \{g(a), g(b)\}$$

$$\pi(b) = \{g(b)\}$$

$$\pi(g(a)) = \{f(g(a))\}$$

$$\pi(g(b)) = \{f(g(b)), f(g(a))\}$$

$$f(g(a)) = c, c \neq f(g(b)), a = b, g(a) = g(b), f(g(a)) = f(g(b))$$

$$F = \{a \mapsto a, \ b \mapsto a, \ c \mapsto c, \ g(a) \mapsto g(b), \ g(b) \mapsto g(b) \\ f(g(a)) \mapsto c, \ f(g(b)) \mapsto f(g(b))\}$$

$$D = \{c \neq f(g(b))\}$$

$$\pi(a) = \{g(a), g(b)\}$$

$$\pi(b) = \{g(b)\}$$

$$\pi(g(a)) = \{f(g(a))\}$$

$$\pi(g(b)) = \{f(g(b)), f(g(a))\}$$

Merge equivalence classes of f(g(a)) and $f(g(b)) \rightsquigarrow$ unsat.

Example: Satisfiable Version

$$f(g(a)) = c, a \neq f(g(b)), a = b, g(a) = g(b), f(g(a)) = f(g(b))$$

$$F = \{a \mapsto a, b \mapsto a, c \mapsto c, g(a) \mapsto g(b), g(b) \mapsto g(b)$$
$$f(g(a)) \mapsto c, f(g(b)) \mapsto c\}$$
$$D = \{a \neq f(g(b))\}$$

Example: Satisfiable Version

$$f(g(a)) = c, a \neq f(g(b)), a = b, g(a) = g(b), f(g(a)) = f(g(b))$$

$$F = \{a \mapsto a, b \mapsto a, c \mapsto c, g(a) \mapsto g(b), g(b) \mapsto g(b)$$
$$f(g(a)) \mapsto c, f(g(b)) \mapsto c\}$$
$$D = \{a \neq f(g(b))\}$$

Model:
$$|M| = \{*_1, *_2, *_3\}$$
 One value for each eq. class root.
 $M(a) = M(b) = *_1$
 $M(c) = *_2$
 $M(g) = \{*_1 \mapsto *_3, \text{else} \mapsto *_?\}$ *? can be any value.
 $M(f) = \{*_3 \mapsto *_2, \text{else} \mapsto *_?\}$

Example: Satisfiable Version

$$f(g(a)) = c, a \neq f(g(b)), a = b, g(a) = g(b), f(g(a)) = f(g(b))$$

$$F = \{a \mapsto a, b \mapsto a, c \mapsto c, g(a) \mapsto g(b), g(b) \mapsto g(b)$$
$$f(g(a)) \mapsto c, f(g(b)) \mapsto c\}$$
$$D = \{a \neq f(g(b))\}$$

Model:
$$|M| = \{*_1, *_2, *_3\}$$
 One value for each eq. class root.
 $M(a) = M(b) = *_1$
 $M(c) = *_2$
 $M(g) = \{*_1 \mapsto *_3, \text{else} \mapsto *_?\}$ *? can be any value.
 $M(f) = \{*_3 \mapsto *_2, \text{else} \mapsto *_?\}$

Equality: T-Justifications

- A T-Justification for F is a set of literals S such that $S \models_T F$.
- S is a non-redudant if there is no $S' \subset S$ such that $S' \models_T F$.
- Non-redundant T-Justifications for variable equalities is easy: shortest-path between two variables.
- With uninterpreted functions the problem is more difficult:
- Example:

 $f_1(x_1) = x_1 = x_2 = f_1(x_{n+1}),$..., $f_n(x_1) = x_n = x_{n+1} = f_n(x_{n+1}),$ $g(f_1(x_1), \dots, f_n(x_1)) \neq g(f_1(x_{n+1}), \dots, f_n(x_{n+1}))$

Roadmap

- Logic Background
- Modern SAT Solvers
- DPLL with Theory Solvers
- Theory Combination
- Equality
- Arithmetic
- Applications

- Algorithms:
 - Graph based for difference logic ($x \le y k$).
 - Fourier-Motzkin elimination.

 $t_1 \leq ax, \ bx \leq t_2 \Rightarrow bt_1 \leq at_2$

- Standard Simplex.
- Standard Simplex based solvers:
 - Standard Form: Ax = b and $x \ge 0$.
 - Incremental: add/remove equations (i.e., rows).
 - Slow backtracking.
 - No theory propagation.

Fast Linear Arithmetic

- Simplex General Form.
- Algorithm based on the Dual Simplex.
- Non-redundant T-Justifications.
- Efficient Backtracking.
- Efficient T-Propagate.
- Support for strict inequalities (t > 0).
- Presimplification step.
- Integer problems: Gomory cuts, Branch & Bound, GCD test.

General Form

- General Form: Ax = 0 and $l_j \le x_j \le u_j$
- Example:

$$x \ge 0, (x + y \le 2 \lor x + 2y \ge 6), (x + y = 2 \lor x + 2y > 4)$$

$$\rightsquigarrow$$
$$s_1 = x + y, s_2 = x + 2y,$$
$$y \ge 0, (x + y \le 2) \lor x \ge 2) \lor x \ge 4)$$

- $x \ge 0, (s_1 \le 2 \lor s_2 \ge 6), (s_1 = 2 \lor s_2 > 4)$
- Only bounds (e.g., $s_1 \leq 2$) are asserted during the search.
- Unconstrained variables can be eliminated before the beginning of the search.

Model + Equations + Bounds

- An assignment (model) is a mapping from variables to values.
- We maintain an assignment that satisfies all equations and bounds.
- The assignment of non dependent variables implies the assignment of dependent variables.
- Equations + Bounds can be used to derive new bounds.
- Example: $x = y z, y \le 2, z \ge 3 \rightsquigarrow x \le -1.$
- The new bound may be inconsistent with the already known bounds.
- Example: $x \leq -1, x \geq 0$.

- The method described only handles non-strict inequalities (e.g., $x \le 2$).
- For integer problems, strict inequalities can be converted into non-strict inequalities. x < 1 → x ≤ 0.</p>
- For rational/real problems, strict inequalities can be converted into non-strict inequalities using a small δ. x < 1 → x ≤ 1 − δ.</p>
- We do not compute a δ , we treat it symbolically.
- δ is an infinitesimal parameter: $(c, k) = c + k\delta$

Initial state

$$s \ge 1, x \ge 0$$
$$(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$$



• Asserting $s \ge 1$

$$s \ge 1, x \ge 0$$
$$(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$$



• Asserting $s \ge 1$ assignment does not satisfy new bound.

$s \ge 1, x \ge 0$



• Asserting $s \ge 1$ pivot s and x (s is a dependent variable).



• Asserting $s \ge 1$ pivot s and x (s is a dependent variable).



• Asserting $s \ge 1$ pivot s and x (s is a dependent variable).



• Asserting $s \ge 1$ update assignment.



• Asserting $s \ge 1$ update dependent variables assignment.

$s \ge 1, x \ge 0$



• Asserting $x \ge 0$



• Asserting $x \ge 0$ assignment satisfies new bound.

$s \ge 1, x \ge 0$



• Case split $\neg y \leq 1$

$$s \ge 1, x \ge 0$$
$$(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$$



• Case split $\neg y \leq 1$ assignment does not satisfies new bound.

$s \ge 1, x \ge 0$



• Case split $\neg y \leq 1$ update assignment.

Model	Equations	Bounds
M(x) = 1	x = s - y	$s \geq 1$
$M(y) = 1 + \delta$	u = s + y	$x \geq 0$
M(s) = 1	v = s - 2y	y > 1
M(u) = 1		
M(v) = 1		

• Case split $\neg y \leq 1$ update dependent variables assignment.

 $s \ge 1, x \ge 0$

Model			Equations			Bounds			
M(x)	=	$-\delta$	x	—	s-y		S	\geq	1
M(y)	=	$1 + \delta$	u	—	s + y		x	\geq	0
M(s)	=	1	v	—	s-2y		y	>	1
M(u)	—	$2+\delta$							
M(v)	=	$-1-2\delta$							

Bound violation

Model			Equations			Bounds			
M(x)	=	$-\delta$	x	—	s-y		S	\geq	1
M(y)	=	$1 + \delta$	u	=	s + y		x	\geq	0
M(s)	=	1	v	=	s-2y		y	>	1
M(u)	=	$2+\delta$							
M(v)	= -	$-1-2\delta$							

• Bound violation pivot x and s (x is a dependent variables).

Model			Equations			Bounds			
M(x)	=	$-\delta$	x	—	s - y		S	\geq	1
M(y)	=	$1 + \delta$	u	=	s + y		x	\geq	0
M(s)	=	1	v	=	s-2y		y	>	1
M(u)	—	$2+\delta$							
M(v)	—	$-1-2\delta$							

• Bound violation pivot x and s (x is a dependent variables).

Model			Equations			Bounds			
M(x)	=	$-\delta$	s	—	x + y		S	\geq	1
M(y)	=	$1 + \delta$	u	—	s + y		x	\geq	0
M(s)	=	1	v	—	s-2y		y	>	1
M(u)	=	$2+\delta$							
M(v)	= -	$-1-2\delta$							

• Bound violation pivot x and s (x is a dependent variables).

Model			Equations			Bounds			
M(x)	=	$-\delta$	s	=	x + y		S	\geq	1
M(y)	=	$1 + \delta$	u	=	x + 2y		x	\geq	0
M(s)	=	1	v	=	x - y		y	>	1
M(u)	—	$2+\delta$							
M(v)	= -	$-1-2\delta$							

Bound violation update assignment.

Model			Equations			Bounds			
M(x)	—	0	s	=	x + y		S	\geq	1
M(y)	=	$1 + \delta$	u	=	x + 2y		x	\geq	0
M(s)	=	1	v	=	x - y		y	>	1
M(u)	=	$2+\delta$							
M(v)	= -	$-1-2\delta$							

Bound violation update dependent variables assignment.

 $s \ge 1, x \ge 0$ (y \le 1 \le v \ge 2), (v \le -2 \le v \ge 2), (v \le -2 \le v \ge 0), (v \le -2 \le u \le -1)

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	y > 1
$M(u) = 2 + 2\delta$		
$M(v) = -1 - \delta$		
• Theory propagation
$$x \ge 0, y > 1 \rightsquigarrow u > 2$$

 $s \ge 1, x \ge 0$
 $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	y > 1
$M(u) = 2 + 2\delta$		
$M(v) = -1 - \delta$		

• Theory propagation $u > 2 \rightsquigarrow \neg u \leq -1$ $s \geq 1, x \geq 0$ $(y \leq 1 \lor v \geq 2), (v \leq -2 \lor v \geq 0), (v \leq -2 \lor u \leq -1)$

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	y > 1
$M(u) = 2 + 2\delta$		u > 2
$M(v) = -1 - \delta$		

▶ Boolean propagation $\neg y \leq 1 \rightsquigarrow v \geq 2$ $s \geq 1, x \geq 0$ $(y \leq 1 \lor v \geq 2), (v \leq -2 \lor v \geq 0), (v \leq -2 \lor u \leq -1)$

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	y > 1
$M(u) = 2 + 2\delta$		u > 2
$M(v) = -1 - \delta$		

• Theory propagation $v \ge 2 \rightsquigarrow \neg v \le -2$ $s \ge 1, x \ge 0$ $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	y > 1
$M(u) = 2 + 2\delta$		u > 2
$M(v) = -1 - \delta$		

Conflict empty clause

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	y > 1
$M(u) = 2 + 2\delta$		u > 2
$M(v) = -1 - \delta$		

Backtracking

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	
$M(u) = 2 + 2\delta$		
$M(v) = -1 - \delta$		

• Asserting $y \leq 1$

$$s \ge 1, x \ge 0$$
$$(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$$

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	
$M(u) = 2 + 2\delta$		
$M(v) = -1 - \delta$		

• Asserting $y \leq 1$ assignment does not satisfy new bound.

 $s \ge 1, x \ge 0$

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
$M(y) = 1 + \delta$	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	$y~\leq~1$
$M(u) = 2 + 2\delta$		
$M(v) = -1 - \delta$		

• Asserting $y \leq 1$ update assignment.

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
M(y) = 1	u = x + 2y	$x \geq 0$
$M(s) = 1 + \delta$	v = x - y	$y \leq 1$
$M(u) = 2 + 2\delta$		
$M(v) = -1 - \delta$		

• Asserting $y \leq 1$ update dependent variables assignment.

Model	Equations	Bounds
M(x) = 0	s = x + y	$s \geq 1$
M(y) = 1	u = x + 2y	$x \geq 0$
M(s) = 1	v = x - y	$y \leq 1$
M(u) = 2		
M(v) = -1		

Theory propagation
$$s \ge 1, y \le 1 \rightsquigarrow v \ge -1$$

 $s \ge 1, x \ge 0$
 $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model		Equations	Bounds
M(x) =	0 <i>x</i>	= s - y	$s \geq 1$
M(y) =	1 <i>u</i>	= s + y	$x \geq 0$
M(s) =	1 <i>v</i>	= s - 2y	$y \leq 1$
M(u) =	2		
M(v) = -	-1		

Theory propagation
$$v \ge -1 \rightsquigarrow \neg v \le -2$$

 $s \ge 1, x \ge 0$
 $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model		Equations	Boun	ds
M(x) =	0 <i>x</i>	= s - y	$s \geq$	1
M(y) =	$1 \qquad u$	= s + y	$x \geq$	0
M(s) =	1 v	= s - 2y	$y \leq$	1
M(u) =	2		$v \geq$	-1
$M(v) = \cdot$	-1			

▶ Boolean propagation $\neg v \leq -2 \rightsquigarrow v \geq 0$ $s \geq 1, x \geq 0$ $(y \leq 1 \lor v \geq 2), (v \leq -2 \lor v \geq 0), (v \leq -2 \lor u \leq -1)$

Model	Equations	Bounds
M(x) = 0	x = s - y	$s \geq 1$
M(y) = 1	u = s + y	$x \geq 0$
M(s) = 1	v = s - 2y	$y \leq 1$
M(u) = 2		$v \geq -1$
M(v) = -1		

Bound violation assignment does not satisfy new bound.

$s \ge 1, x \ge 0$



• Bound violation pivot u and s (u is a dependent variable).

$s \ge 1, x \ge 0$



• Bound violation pivot u and s (u is a dependent variable).

$s \ge 1, x \ge 0$



• Bound violation pivot u and s (u is a dependent variable).

$s \ge 1, x \ge 0$

Model	Equations	Bounds
M(x) = 0	x = v + y	$s \geq 1$
M(y) = 1	u = v + 3y	$x \geq 0$
M(s) = 1	s = v + 2y	$y \leq 1$
M(u) = 2		$v \geq 0$
M(v) = -1		

▶ Bound violation update assignment. $s \ge 1, x \ge 0$ $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model	Equations	Bounds
M(x) = 0	x = v + y	$s \geq 1$
M(y) = 1	u = v + 3y	$x \geq 0$
M(s) = 1	s = v + 2y	$y \leq 1$
M(u) = 2		$v \geq 0$
M(v) = 0		

▶ Bound violation update dependent variables assignment. $s \ge 1, x \ge 0$ $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model	Equations	Bounds
M(x) = 1	x = v + y	$s \geq 1$
M(y) = 1	u = v + 3y	$x \geq 0$
M(s) = 2	s = v + 2y	$y \leq 1$
M(u) = 3		$v \geq 0$
M(v) = 0		

▶ Boolean propagation $\neg v \leq -2 \rightsquigarrow u \leq -1$ $s \geq 1, x \geq 0$ $(y \leq 1 \lor v \geq 2), (v \leq -2 \lor v \geq 0), (v \leq -2 \lor u \leq -1)$

Model	Equations	Bounds
M(x) = 1	x = v + y	$s \geq 1$
M(y) = 1	u = v + 3y	$x \geq 0$
M(s) = 2	s = v + 2y	$y \leq 1$
M(u) = 3		$v \geq 0$
M(v) = 0		

Bound violation assignment does not satisfy new bound.

$s \ge 1, x \ge 0$

Model	Equations	Bounds
M(x) = 1	x = v + y	$s \geq 1$
M(y) = 1	u = v + 3y	$x \geq 0$
M(s) = 2	s = v + 2y	$y \leq 1$
M(u) = 3		$v \geq 0$
M(v) = 0		$u \leq -1$

• Bound violation pivot u and y (u is a dependent variable).

$s \ge 1, x \ge 0$

Model	Equations	Bounds
M(x) = 1	x = v + y	$s \geq 1$
M(y) = 1	u = v + 3y	$x \geq 0$
M(s) = 2	s = v + 2y	$y \leq 1$
M(u) = 3		$v \geq 0$
M(v) = 0		$u \leq -1$

• Bound violation pivot u and y (u is a dependent variable).

$s \ge 1, x \ge 0$



• Bound violation pivot u and y (u is a dependent variable).

$s \ge 1, x \ge 0$



Bound violation update assignment. $s \ge 1, x \ge 0$ $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model	Equations	Bounds
M(x) = 1	$x = \frac{1}{3}u + \frac{2}{3}v$	$s \geq 1$
M(y) = 1	$y = \frac{1}{3}u - \frac{1}{3}v$	$x \geq 0$
M(s) = 2	$s = \frac{2}{3}u + \frac{1}{3}v$	$y \leq 1$
M(u) = -1		$v \geq 0$
M(v) = 0		$u \leq -1$

▶ Bound violation update dependent variables assignment. $s \ge 1, x \ge 0$ $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model		Equ	ations	E	Bound	ls
M(x) = -	$-\frac{1}{3}$ x	=	$\frac{1}{3}u + \frac{2}{3}v$	S	\geq	1
M(y) = -	$-\frac{1}{3}$ y	=	$\frac{1}{3}u - \frac{1}{3}v$	x	\geq	0
M(s) = -	$-\frac{2}{3}$ S	=	$\frac{2}{3}u + \frac{1}{3}v$	y	\leq	1
$M(u) = \cdot$	-1			v	\geq	0
M(v) =	0			u	\leq	-1

Bound violations

Model		Equa	ations	B	Sound	ls
M(x) =	$-\frac{1}{3}$ x	—	$\frac{1}{3}u + \frac{2}{3}v$	S	\geq	1
M(y) =	$-\frac{1}{3}$ y	=	$\frac{1}{3}u - \frac{1}{3}v$	x	\geq	0
M(s) =	$-\frac{2}{3}$ s	=	$\frac{2}{3}u + \frac{1}{3}v$	y	\leq	1
M(u) =	-1			v	\geq	0
M(v) =	0			U	\leq	-1

• Bound violations pivot s and v (s is a dependent variable).

$s \ge 1, x \ge 0$



• Bound violations pivot s and v (s is a dependent variable).

$s \ge 1, x \ge 0$



• Bound violations pivot s and v (s is a dependent variable).

$s \ge 1, x \ge 0$

Model	Equations	Bounds
$M(x) = -\frac{1}{3}$	x = 2s - u	$s \geq 1$
$M(y) = -\frac{1}{3}$	y = -s + u	$x \geq 0$
$M(s) = -\frac{2}{3}$	v = 3s - 2u	$y \leq 1$
M(u) = -1		$v \geq 0$
M(v) = 0		$u \leq -1$

Bound violations update assignment.

Model	Equations	Bounds
$M(x) = -\frac{1}{3}$	x = 2s - u	$s \geq 1$
$M(y) = -\frac{1}{3}$	y = -s + u	$x \geq 0$
M(s) = 1	v = 3s - 2u	$y \leq 1$
M(u) = -1		$v \geq 0$
M(v) = 0		$u \leq -1$

• Bound violations update dependent variables assignment. $s \ge 1, x \ge 0$ $(y \le 1 \lor v \ge 2), (v \le -2 \lor v \ge 0), (v \le -2 \lor u \le -1)$

Model	Equations	Bounds
M(x) = 3	x = 2s - u	$s \geq 1$
M(y) = -2	y = -s + u	$x \geq 0$
M(s) = 1	v = 3s - 2u	$y \leq 1$
M(u) = -1		$v \geq 0$
M(v) = 5		$u \leq -1$

Found satisfying assignment

Model	Equations	Bounds
M(x) = 3	x = 2s - u	$s \geq 1$
M(y) = -2	y = -s + u	$x \geq 0$
M(s) = 1	v = 3s - 2u	$y \leq 1$
M(u) = -1		$v \geq 0$
M(v) = 5		$u \leq -1$

Opportunistic equality propagation

- Efficient (and incomplete) methods for propagating equalities.
- Notation
 - A variable x_i is fixed iff $l_i = u_i$.
 - A linear polynomial $\sum_{x_j \in \mathcal{V}} a_{ij} x_j$ is fixed iff x_j is fixed or $a_{ij} = 0$.
 - Given a linear polynomial $P = \sum_{x_j \in \mathcal{V}} a_{ij} x_j$, and a model M: M(P) denotes $\sum_{x_j \in \mathcal{V}} a_{ij} M(x_j)$.

Opportunistic equality propagation

Equality propagation in arithmetic:

FixedEq

$$l_i \le x_i \le u_i, \ l_j \le x_j \le u_j \Longrightarrow \ x_i = x_j \ \text{if} \ l_i = u_i = l_j = u_j$$

EqRow

$$x_i = x_j + P \implies x_i = x_j$$
 if P is fixed, and $M(P) = 0$

EqOffsetRows

$$\begin{aligned} x_i &= x_k + P_1 \\ x_j &= x_k + P_2 \end{aligned} \implies x_i = x_j \text{ if } \begin{cases} P_1 \text{ and } P_2 \text{ are fixed, and} \\ M(P_1) &= M(P_2) \end{cases} \end{aligned}$$

EqRows

$$\begin{aligned} x_i &= P + P_1 \\ x_j &= P + P_2 \end{aligned} \implies x_i = x_j \text{ if } \begin{cases} P_1 \text{ and } P_2 \text{ are fixed, and} \\ M(P_1) &= M(P_2) \end{cases} \end{aligned}$$

Opportunistic theory/equality propagation

- These rules can miss some implied equalities.
- Example: z = w is detected, but x = y is not because w is not a fixed variable.

x = y + w + sz = w + s $0 \leq z$ $w \leq 0$ $0 \leq s \leq 0$

Remark: bound propagation can be used imply the bound 0 ≤ w, making w a fixed variable.
Linear Integer Arithmetic

- GCD test
- Gomory Cuts
- Branch and Bound

Beyond Linear Arithmetic

- Gröbner Basis
- Cylindric Algebraic Decomposition