

VTSA summer school 2015

Exploiting SMT for Verification of Infinite-State Systems

2. Interpolation in SMT and in Verification

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Introduction

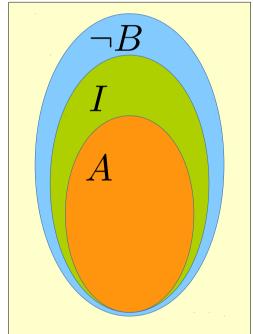
Interpolants in Formal Verification

Computing interpolants in SMT



- (Craig) Interpolant for an ordered pair (*A*, *B*) of formulae s.t. $A \land B \models_T \bot$ (or: $A \models_T \neg B$) is a formula *I* s.t.
 - $\blacksquare A \models_T I$
 - $\blacksquare I \land B \models_T \bot (I \models_T \neg B)$
 - All the uninterpreted (in T) symbols of I are shared between A and B
- Why are interpolants useful?
 - Overapproximation of A relative to B
 - Overapprox. of $\exists_{\{x \notin B\}} \vec{x}.A$







Several important applications in formal verification:

- Approximate image computation for model checking of infinite-state systems
- Predicate discovery for Counterexample-Guided Abstraction Refinement
- Approximation of transition relation for infinite-state systems
- An alternative to (lazy) predicate abstraction for program verification
- Automatic generation of loop invariants



Introduction

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Background



Symbolic transition systems

- State variables X
- Initial states formula I(X)
- Transition relation formula T(X, X')
- A state σ is an assignment to the state vars $\bigwedge_{x_i \in X} x_i = v_i$
- A path of the system S is a sequence of states $\sigma_0, \ldots, \sigma_k$ such that $\sigma_0 \models I$ and $\sigma_i, \sigma'_{i+1} \models T$
- A k-step (symbolic) unrolling of S is a formula

 $I(X^0) \wedge \bigwedge_{i=0}^{k-1} T(X^i, X^{i+1})$

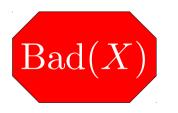
- Encodes all possible paths of length up to k
- A state property is a formula P over X
 - \blacksquare Encodes all the states $\,\sigma\,$ such that $\,\sigma\models P$



• Compute all states reachable from σ in one transition: $Img(\sigma(X)) := \exists X.\sigma(X) \land T(X, X')[X/X']$

Prove that a set of states Bad(X) is not reachable:

$$R(X) := I(X)$$



 $\mathrm{Img}(R(X))$



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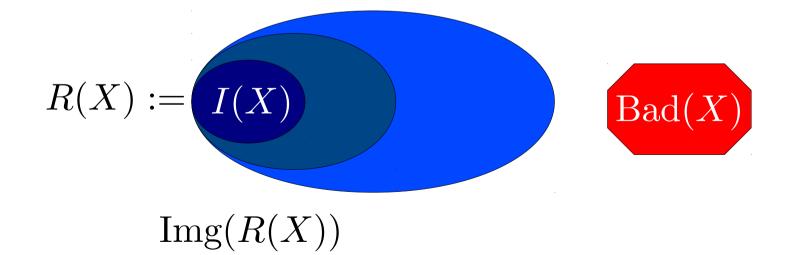
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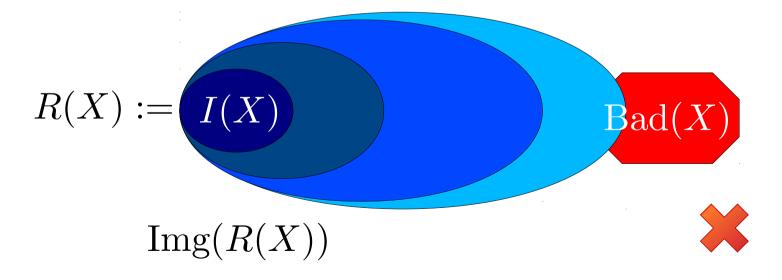


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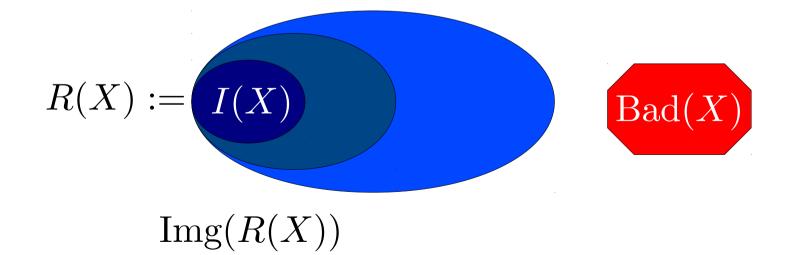


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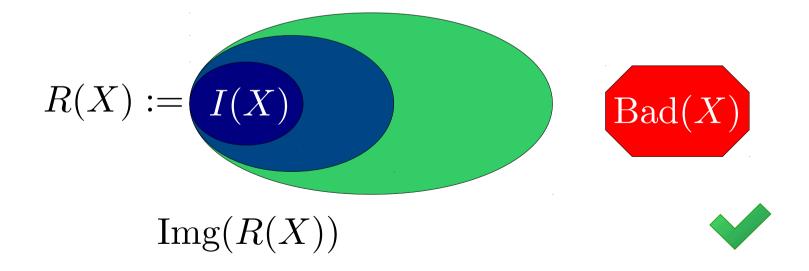


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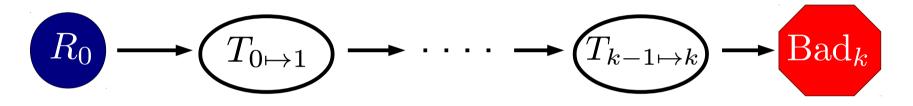


- Image computation requires quantifier elimination, which is typically very expensive (both in theory and in practice)
- Interpolation-based algorithm (McMillan CAV'03): use interpolants to overapproximate image computation
 - much more efficient than the previous algorithm
 - interpolation is often much cheaper than quantifier elimination
 - abstraction (overapproximation) accelerates convergence
 - termination is still guaranteed for finite-state systems



• Set R(X) := I(X)

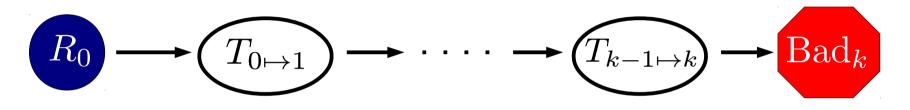
• Check satisfiability of $R_0 \wedge \bigwedge_{i=0}^{k-1} T_i \wedge \operatorname{Bad}_k$





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If SAT:

• If $R \equiv I$, return **REACHABLE**

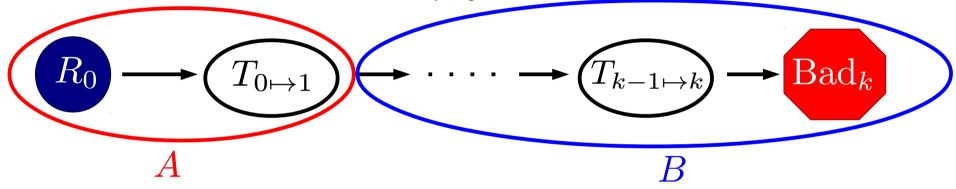
the unrolling hits Bad

else, increase k and repeat



• Set R(X) := I(X)

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If UNSAT:

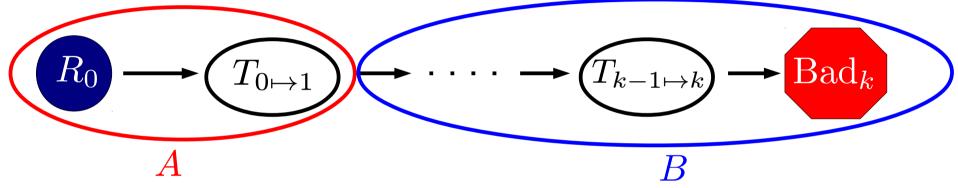
• Set $\varphi(X) := \text{Interpolant}(A, B)[X'/X]$

 φ is an abstraction of the forward image guided by the property



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If UNSAT:

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If $\varphi \models R$, return UNREACHABLE fixpoint found
 else, set $R(X) := R(X) \lor \varphi(X)$ and continue

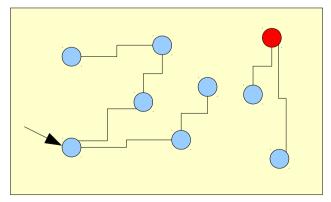
- Given a Transition System S := (I, T) and predicates \mathbb{P}
 - Abstract initial states

$$\widehat{I(X)}_{\mathbb{P}} := \exists X. (I(X) \land \bigwedge_{p \in \mathbb{P}} (x_p \leftrightarrow p(X))[p(X)/x_p]$$

Abstract forward image

$$\widehat{\mathrm{Img}}(\varphi(X))_{\mathbb{P}} := \exists X, X', \vec{x_p}. (\varphi(X) \wedge T(X, X') \wedge \bigwedge_{p \in \mathbb{P}} (x_p \leftrightarrow p(X) \wedge x'_p \leftrightarrow p(X')) [p(X)/x'_p]$$

- Standard technique applied in many verification tools
 - In conjunction with counterexample-guided refinement (CEGAR)



Extract new predicates from spurious counterexamples and compute a more precise abstraction

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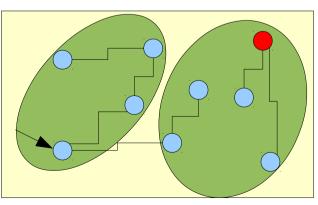
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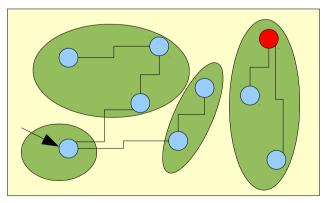
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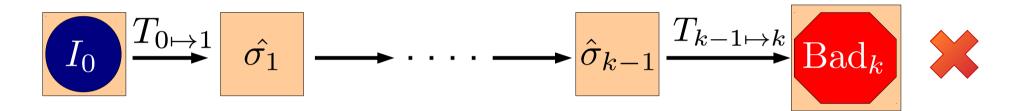
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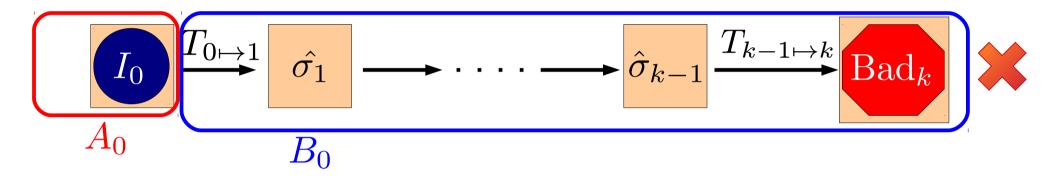
- An abstract cex path $\hat{\sigma_0},\ldots,\hat{\sigma_k}$ (wrt. $\mathbb P$) might be spurious
 - Because abstraction is overapproximating



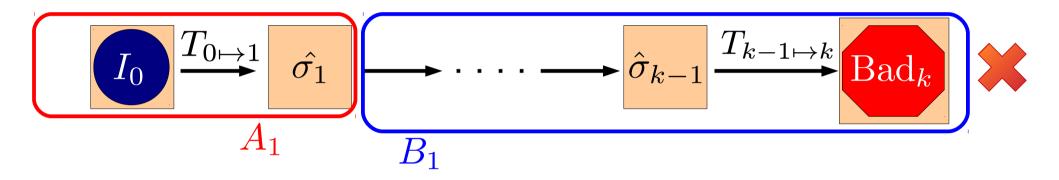
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$$\overbrace{I_0} \xrightarrow{T_{0 \mapsto 1}} \widehat{\sigma_1} \longrightarrow \cdots \longrightarrow \widehat{\sigma_{k-1}} \xrightarrow{T_{k-1 \mapsto k}} \operatorname{Bad}_k \quad ($$

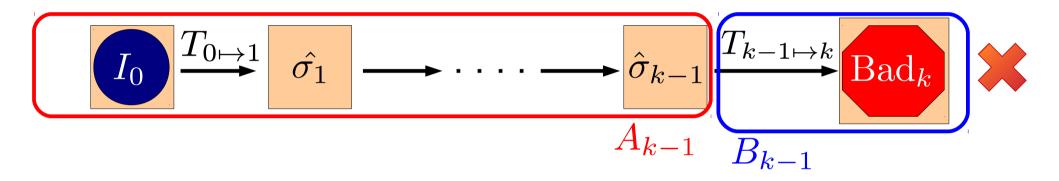
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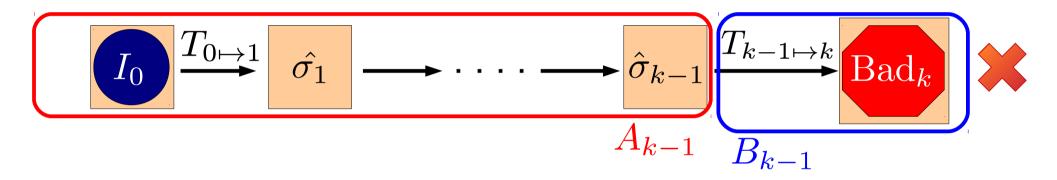
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- Compute a sequence of interpolants $\varphi_0, \dots, \varphi_{k-1}$ such that $T_{i \mapsto i+1} \land \varphi_i \models \varphi_{i+1}$ for all $i \in [0, k-1)$
- Let P_{new} be the set of all the predicates in *φ*₀,...,*φ*_{k-1}
 Set P' := P ∪ P_{new}

Theorem: $\hat{\sigma_0}, \ldots, \hat{\sigma_k}$ is not an abstract cex path wrt. \mathbb{P}'

Proof sketch



- φ_i is an overapproximation of the states reachable in *i* steps, compatible with the abstract trace $\hat{\sigma}_0, \ldots, \hat{\sigma}_i$
- φ_i is also incompatible with the rest of the abstract trace $\hat{\sigma}_{i+1}, \ldots, \hat{\sigma}_k$ (since it is an interpolant)
- By the requirement that T_{i→i+1} ∧ φ_i ⊨ φ_{i+1} it follows that Img(φ_i) ⊨ φ_{i+1}
 Therefore, Img(..., Img(φ₀)) ⊨ φ_{k-1} and Img(φ_{k-1}) ⊨ ⊥ (since the trace is spurious)
- Since we add all the atomic predicates of $\varphi_0, \ldots, \varphi_{k-1}$ to \mathbb{P}' and the abstraction is precise wrt. \mathbb{P}' , then

$$\widehat{\mathrm{Img}}(\underbrace{\ldots}_{k-1}\widehat{\mathrm{Img}}(\varphi_0)_{\mathbb{P}'})_{\mathbb{P}'} \models \bot$$



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- Interpolants for Boolean CNF formulae (A, B) can be computed from resolution refutations in linear time
- Traverse the resolution proof, annotating each node with a partial interpolant /
 - The partial interpolant for the root node (the empty clause) is the computed interpolant



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- McMillan's annotation rules (others exist):
 - For each leaf node (input clause) C in the proof:
 - If $C \in A$, set $I := \bigvee \{ l \in C \mid \operatorname{var}(l) \in B \}$
 - Otherwise ($C \in \boldsymbol{B}$), set $I := \top$

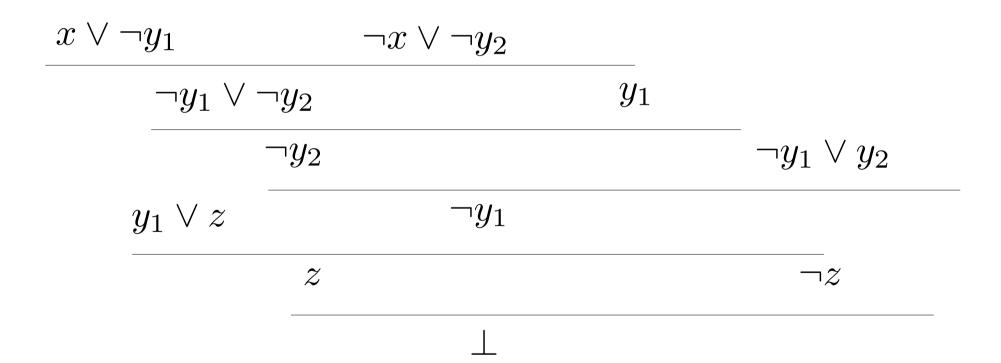


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 - Otherwise ($C \in \mathbf{B}$), set $I := \top$
 - For each inner node (resolution) with parents $\varphi \lor l$ and $\psi \lor \neg l$ and annotations I_1 and I_2
 - If $var(l) \in B$, set $I := I_1 \wedge I_2$; otherwise, set $I := I_1 \vee I_2$

Example



$$A := (x \lor \neg y_1) \land (\neg x \lor \neg y_2) \land y_1$$
$$B := (\neg y_1 \lor y_2) \land (y_1 \lor z) \land \neg z$$



Example



$$A := (x \lor \neg y_1) \land (\neg x \lor \neg y_2) \land y_1$$
$$B := (\neg y_1 \lor y_2) \land (y_1 \lor z) \land \neg z$$



By induction on the structure of the resolution refutation

• Lemma: for each annotated node C[I], we have 1) $A \models I \lor \bigvee \{l \in C \mid var(l) \notin B\}$ 2) $B \land I \models \lor \bigvee \{l \in C \mid var(l) \in B\}$ 3) *I* contains only variables that occur in both *A* and *B*

- Then as a corollary, for the root $\perp [I]$, I is an interpolant
- The lemma trivially holds for leaf nodes (check)



Resolution step with parents $(\varphi \lor l)$ $[I_1]$ and $(\psi \lor \neg l)$ $[I_2]$ Case $var(l) \in B$

1) By ind. hyp $A \models I_1 \lor \bigvee \{ p \in \varphi \mid \operatorname{var}(p) \notin B \}$ and $A \models I_2 \lor \bigvee \{ p \in \psi \mid \operatorname{var}(p) \notin B \}$

Therefore $A \models (I_1 \land I_2) \lor \bigvee \{ p \in \varphi \land \psi \mid \operatorname{var}(p) \notin B \}$

2) By inductive hypotesis $B \wedge I_1 \models \bigvee \{p \in \varphi \lor l \mid \operatorname{var}(p) \in B\}$ which means $B \models \neg I_1 \lor \bigvee \{p \in \varphi \lor l \mid \operatorname{var}(p) \in B\}$ Similarly, $B \models \neg I_2 \lor \bigvee \{p \in \psi \lor \neg l \mid \operatorname{var}(p) \in B\}$ By resolution on $\operatorname{var}(l)$, then

 $\mathbf{B} \models \neg I_1 \lor \neg I_2 \lor \bigvee \{ p \in \varphi \lor \psi \mid \operatorname{var}(p) \in \mathbf{B} \}$

3) Trivial by the inductive hypothesis



Resolution step with parents $(\varphi \lor l)$ $[I_1]$ and $(\psi \lor \neg l)$ $[I_2]$ Case $var(l) \notin B$

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By resolution on var(l), then

 $\mathbf{A} \models (I_1 \lor I_2) \lor \bigvee \{ p \in \varphi \lor \psi \mid \operatorname{var}(p) \notin \mathbf{B} \}$

2) By ind. hyp $B \models \neg I_1 \lor \bigvee \{ p \in \varphi \mid \operatorname{var}(p) \in B \}$ and $B \models \neg I_2 \lor \bigvee \{ p \in \psi \mid \operatorname{var}(p) \in B \}$

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Resolution refutations in SMT:

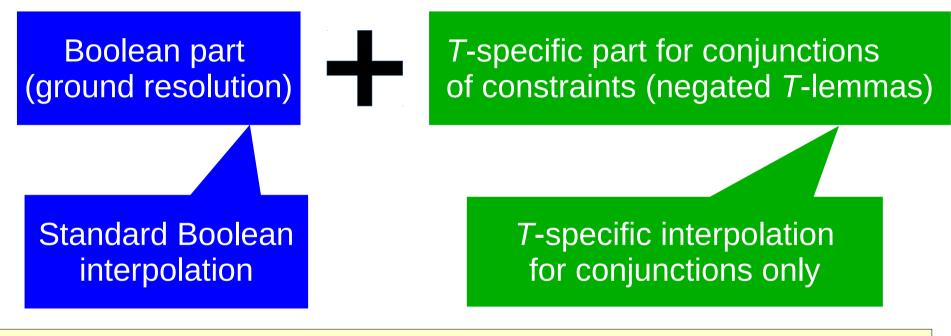
Boolean part (ground resolution)

T-specific part for conjunctions of constraints (negated *T*-lemmas)

Interpolants in SMT



Resolution refutations in SMT:

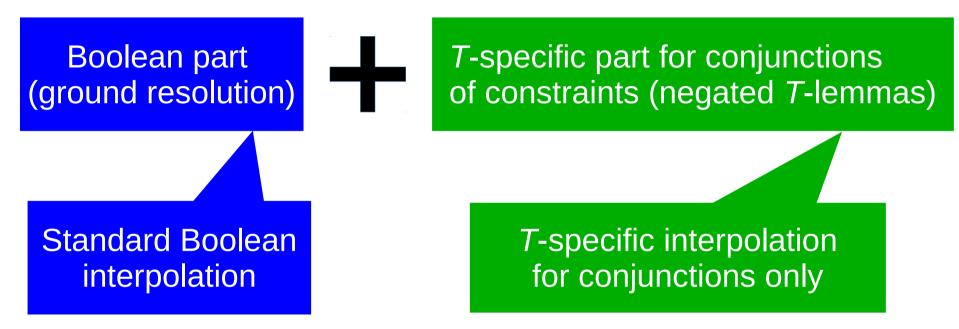


Theory interpolation only for sets of *T*-literals

Interpolants in SMT



Resolution refutations in SMT:



Theory interpolation only for sets of *T*-literals

Annotation for a T-lemma C:

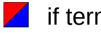
 $I := T \text{-interpolant}(\bigwedge \{l \in \neg C \mid \operatorname{var}(l) \notin B\},\$

 $\bigwedge \{l \in \neg C \mid \operatorname{var}(l) \in B\})$



Interpolants from coloured congruence graphs

- Nodes with colours:
- if term occurs in A if term occurs in B



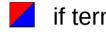
if term is shared

- Edges with colours of the nodes they connect
 - Uncolorable edge: connects nodes of two different colours
- Always possible to obtain a coloured graph
 - (by introducing new nodes)



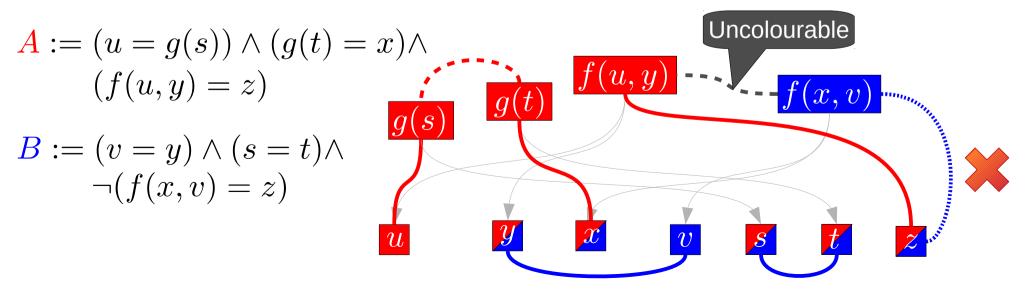
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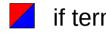
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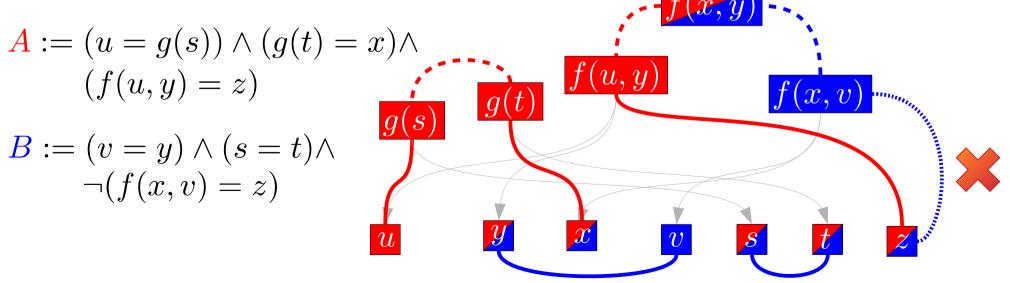
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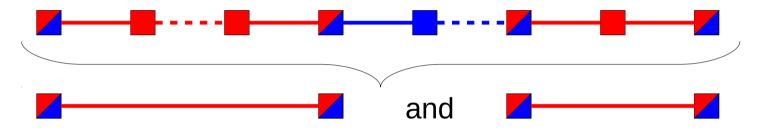
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Interpolation algorithm (sketch)



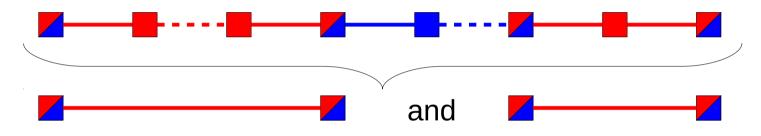
- Start from disequality edge _____
- Compute summaries for A-paths with shared endpoints



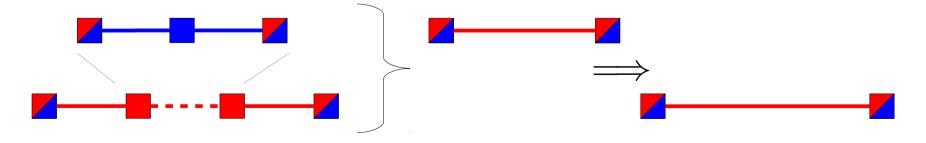
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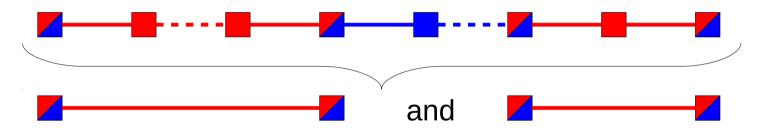
- If an A-summary involves a congruence edge, compute summaries recursively on function arguments
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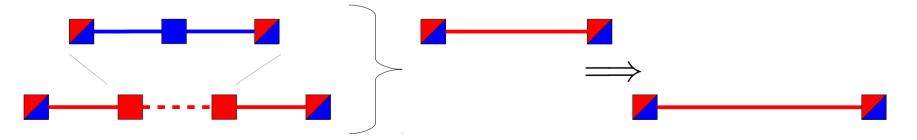
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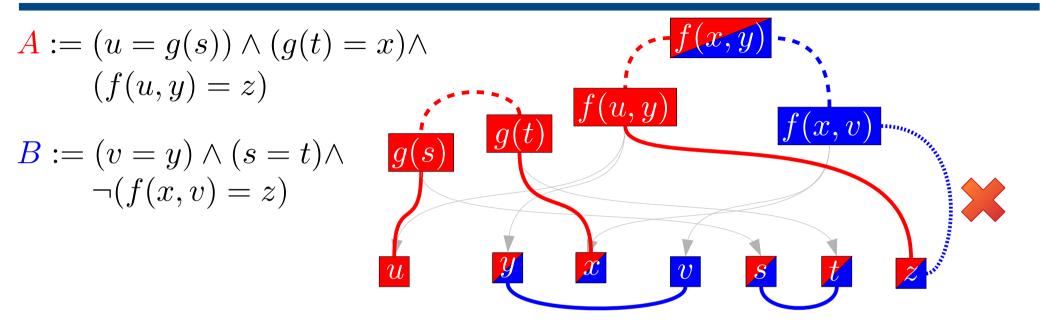


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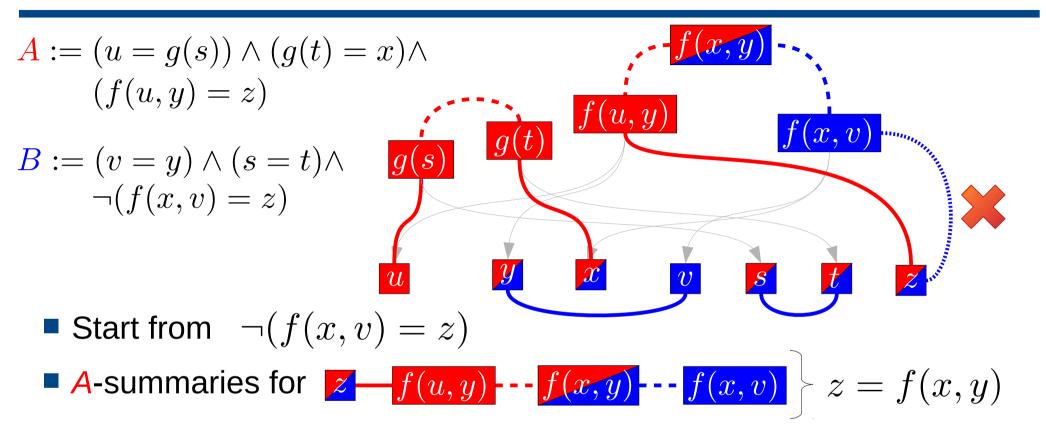


(Several cases to consider)

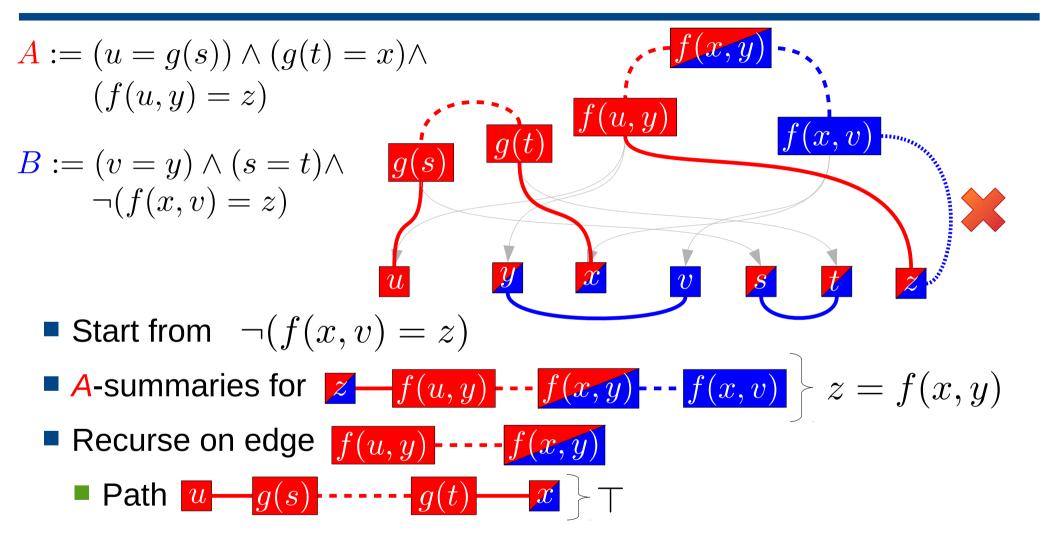




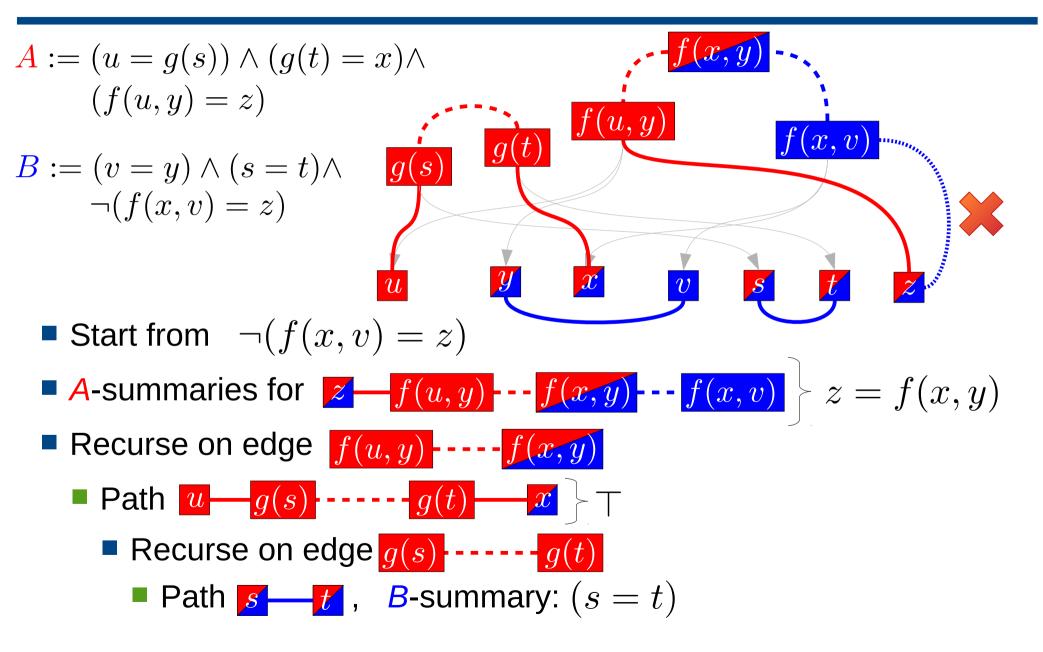




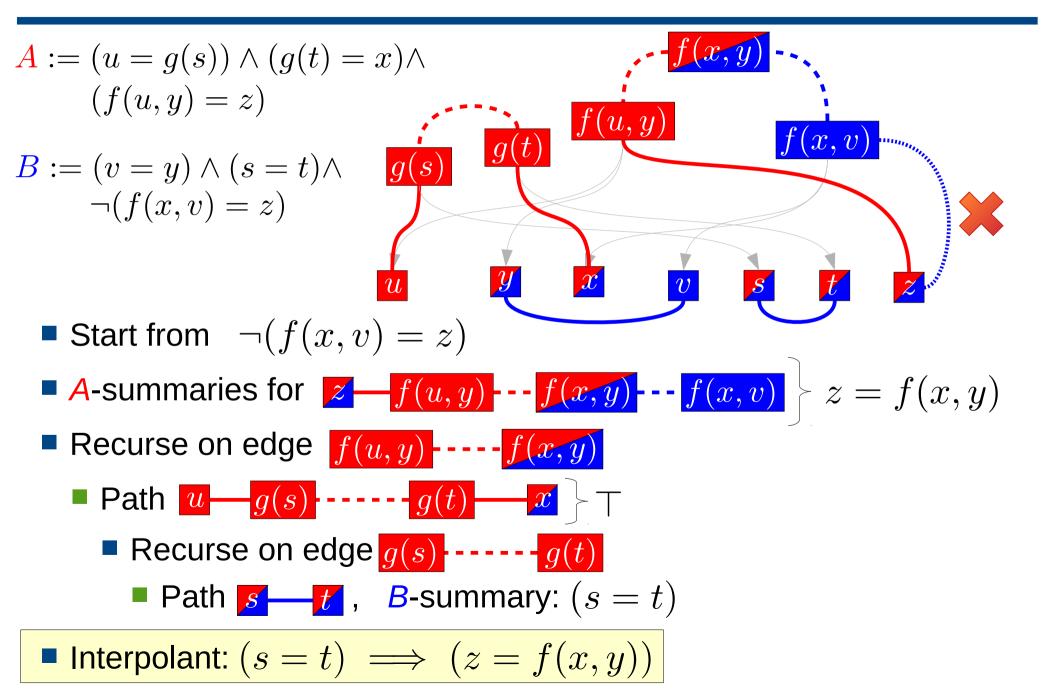














- Interpolants from proofs of unsatisfiability of a system of inequalities $\sum_i a_i x_i \leq c$
- Proof of unsatisfiability: linear combination of inequalities with positive coefficients to derive a contradiction ($0 \le c$ with c < 0)
- Interpolant obtained out of the proof by combining inequalities from A (using the same coefficients)
- Proof of unsatisfiability generated from the Simplex



 $\begin{array}{ccc} s_3 & \mapsto & 0 \\ s_4 & \mapsto & 0 \end{array}$

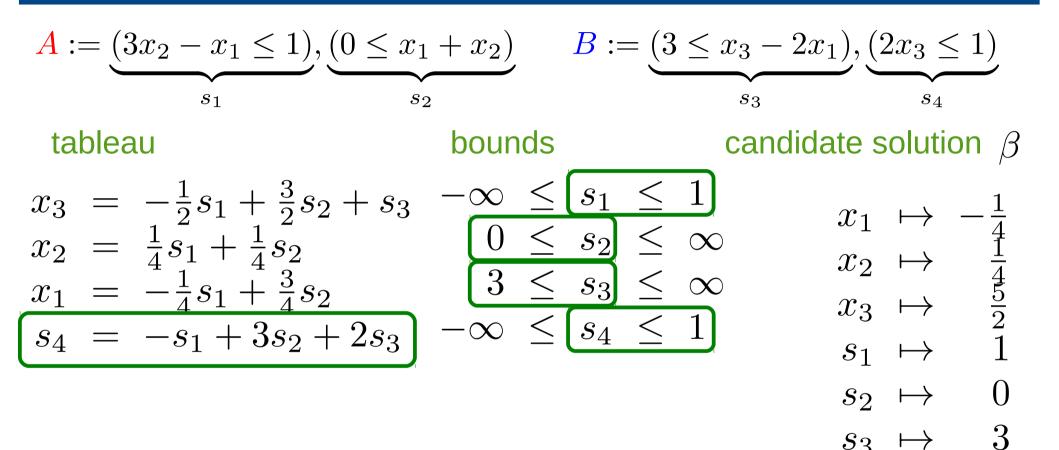
$\mathbf{A} := \underbrace{(3x_2 - x_1 \le 1)}, \underbrace{(0)}$	$:=\underbrace{(3\leq x_3-2x_1)},\underbrace{(2x_3\leq x_3)},\underbrace{(2x_3\leq x_3)},\underbrace{(2x_3\geq x_3)},$	$\leq 1)$	
s_1	s_2	s_3 s_4	
tableau	bounds	candidate solut	ion eta
$s_1 = 3x_2 - x_1$ $s_2 = x_1 + x_2$ $s_3 = x_3 - 2x_1$ $s_4 = 2x_3$	$egin{array}{cccc} -\infty &\leq s_1 \ 0 &\leq s_2 \ 3 &\leq s_3 \ -\infty &\leq s_4 \end{array}$	$egin{array}{cccccccccccccccccccccccccccccccccccc$	0 0 0



 $s_3 \mapsto$

 $s_4 \mapsto$

5



No suitable variable for pivoting! Conflict



$A := \underbrace{(3x_2 - x_1 \le 1)}, \underbrace{(0 \le x)}_{x_1 \le 1}, \underbrace{(0 \le x)}_{x_2 \ge 1}, \underbrace{(0 \ge x)}_{x_2 \ge 1}, \underbrace{(0 \ge x)}_{x_2 \ge 1}, \underbrace{(0 \ge x)}_{x_2 $	$(1+x_2)$ $B := (3 \le$	$(2x_3 - 2x_1), (2x_3 - 2x_1)$	$c_3 \leq 1)$
s_1 s	\sim $^{2}2$	\mathbf{v} s_3	\mathbf{v}_{4}
tableau	bounds	candidate sol	lution β
$x_{3} = -\frac{1}{2}s_{1} + \frac{3}{2}s_{2} + s_{3}$ $x_{2} = \frac{1}{4}s_{1} + \frac{1}{4}s_{2}$ $x_{1} = -\frac{1}{4}s_{1} + \frac{3}{4}s_{2}$ $s_{4} = -s_{1} + 3s_{2} + 2s_{3}$ Proof:	$\begin{array}{c cccc} 0 & \leq & s_2 \\ \hline 3 & \leq & s_3 \\ \hline \end{array} & \leq & \circ \end{array}$		$ \begin{array}{c} \rightarrow & \frac{1}{4} \\ \rightarrow & \frac{5}{2} \\ \rightarrow & 1 \end{array} $
$ \frac{1 \cdot (2x_3 \le 1) 1 \cdot (3x_2 - x_1)}{(2x_3 + 3x_2 - x_1 \le 2)} $		$(-2x_1)$	



$A := \underbrace{(3x_2 - x_1 \le 1)}, \underbrace{(0 \le 1)}_{x_1 x_2 x_2 x_2 x_3 x_3 x_3 x_3 x_3 x_3 x_3 x_3 x_3 x_3$	$\leq x_1 + x_2) \qquad \mathbf{B} := (3 \leq \mathbf{A})$	$(x_3 - 2x_1), (2x_3 \le 1)$
s_1	s_2	s_3 s_4
tableau	bounds	candidate solution β
$x_{3} = -\frac{1}{2}s_{1} + \frac{3}{2}s_{2} + x_{2} = \frac{1}{4}s_{1} + \frac{1}{4}s_{2} + \frac{1}{4}s_{2} + \frac{1}{4}s_{1} + \frac{3}{4}s_{2} + \frac{1}{4}s_{1} + \frac{1}{4}s_{2} + \frac{1}{4}s_{2} + \frac{1}{4}s_{1} + \frac{1}{4}s_{2} + \frac{1}{4}s_{2} + \frac{1}{4}s_{1} + \frac{1}{4}s_{2} + $	$\begin{array}{c ccc} 0 & \leq & s_2 \\ \hline 3 & \leq & s_3 \\ \hline \end{array} & \leq & \circ \end{array}$	
Interpolant:		$s_2 \mapsto 0$
$ \begin{array}{c c} $	$3 \cdot (0 \le x_1 + x_2)$	$\begin{array}{cccc} s_3 \mapsto & 3 \\ s_4 \mapsto & 5 \end{array}$



Constraints of the form $\sum_{i} c_{i} x_{i} + c \bowtie 0, \qquad \bowtie \in \{\leq, =\}$

In general, no quantifier-free interpolation for LIA

Example: A := (y - 2x = 0) B := (y - 2z - 1 = 0)The only interpolant is: $\exists w.(y = 2w)$

Solution: extend the signature to include modular equations (divisibility predicates)

$$(t+c =_d 0) \equiv \exists w.(t+c = d \cdot w), \ d \in \mathbb{Z}^{>0}$$

The interpolant now becomes: $(y =_2 0)$

SMT(LIA) with modular equations



- Modular equations can be eliminated via preprocessing:
 - Replace every atom $a := (t + c =_d 0)$ with a fresh Boolean variable p_a

Add the 4 clauses

$$p_a
ightarrow (t + c - dw_1 = 0)$$

 $\neg p_a
ightarrow (t + c - dw_1 - w_2 = 0)$
 $(-w_2 + 1 \le 0)$
 $(w_2 - d + 1 \le 0)$

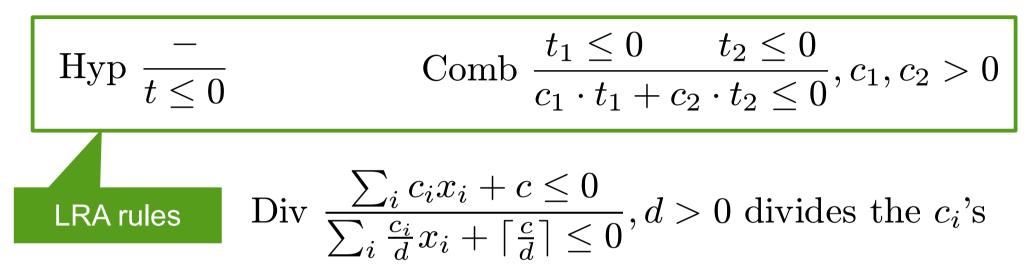
where w_1, w_2 are fresh integer variables



Hyp
$$\frac{-}{t \le 0}$$
 Comb $\frac{t_1 \le 0}{c_1 \cdot t_1 + c_2 \cdot t_2 \le 0}, c_1, c_2 > 0$

Div
$$\frac{\sum_{i} c_{i} x_{i} + c \leq 0}{\sum_{i} \frac{c_{i}}{d} x_{i} + \lceil \frac{c}{d} \rceil \leq 0}, d > 0$$
 divides the c_{i} 's







Hyp
$$\frac{-}{t \le 0}$$
 Comb $\frac{t_1 \le 0}{c_1 \cdot t_1 + c_2 \cdot t_2 \le 0}, c_1, c_2 > 0$

Strenghten
$$\frac{\sum_{i} c_{i} x_{i} + c \leq 0}{\sum_{i} c_{i} x_{i} + d \cdot \lceil \frac{c}{d} \rceil \leq 0}, d > 0 \text{ divides the } c_{i}\text{'s}$$



Hyp
$$\frac{-}{t \le 0}$$
 Comb $\frac{t_1 \le 0}{c_1 \cdot t_1 + c_2 \cdot t_2 \le 0}, c_1, c_2 > 0$

Strenghten
$$\frac{\sum_{i} c_{i} x_{i} + c \leq 0}{\sum_{i} c_{i} x_{i} + d \cdot \lceil \frac{c}{d} \rceil \leq 0}, d > 0$$
 divides the c_{i} 's

Interpolation by annotating proof rules

- Annotation: a set of pairs $\{\langle t_i \leq 0, \bigwedge_j (t_{ij} = 0) \rangle\}_i$
- When \perp is derived, then

 $I := \bigvee_i (t_i \leq 0 \land \bigwedge_j \text{ExistElim}(x_i \notin B).(t_{ij} = 0))$ is the computed interpolant



Annotations for Hyp and Comb from McMillan (same as LRA)

$$\begin{aligned} \text{Hyp} & \frac{-}{t \leq 0 \left[\left\{\left\langle t \leq 0, \top \right\rangle\right\}\right]} t' = \begin{cases} t & \text{if } t \leq 0 \in A \\ 0 & \text{if } t \leq 0 \in B \end{cases} \\ \text{Comb} & \frac{t_1 \leq 0 \left[I_1\right] & t_2 \leq 0 \left[I_2\right]}{c_1 \cdot t_1 + c_2 \cdot t_2 \leq 0 \left[I\right]} \\ I &:= \left\{\left\langle c_1 t'_i + c_2 t'_j \leq 0, E_i \wedge E_j \right\rangle \mid \left\langle t'_i, E_i \right\rangle \in I_1, \left\langle t'_j, E_j \right\rangle \in I_2 \right\} \end{aligned}$$

k-Strengthen rule of [Brillout et al. IJCAR'10]

Str.
$$\frac{\sum_{i} c_{i} x_{i} + c \leq 0 \left[\left\{ \langle t \leq 0, \top \rangle \right\} \right]}{\sum_{i} c_{i} x_{i} + d \cdot \left\lceil \frac{c}{d} \right\rceil \leq 0 \left[I \right]}, d > 0 \text{ divides the } c_{i} \text{'s}$$
$$I := \left\{ \langle (t + n \leq 0), \ (t + n = 0) \rangle \mid 0 \leq n < d \cdot \left\lceil \frac{c}{d} \right\rceil - c \right\} \cup \left\{ \langle (t + d \cdot \left\lceil \frac{c}{d} \right\rceil - c \leq 0), \top \rangle \right\}$$



Annotations for Hyp and Comb from McMillan (same as LRA)

Hyp
$$\frac{-}{t \leq 0 \left[\left\{\left\langle 0 \leq 0, \top \right\rangle\right\}\right]} t' = \begin{cases} t & \text{if } t \leq 0 \in A \\ 0 & \text{if } t \leq 0 \in B \end{cases}$$

Comb
$$\frac{t_1 \leq 0 \left[I_1\right] \quad t_2 \leq 0 \left[I_2\right]}{c_1 \cdot t_1 + c_2 \cdot t_2 \leq 0 \left[I\right]}$$

 $I := \{ \langle c_1 t'_i + c_2 t'_j \leq 0, E_i \wedge E_j \rangle \mid \langle t'_i, E_i \rangle \in I_1, \langle t'_j, E_j \rangle \in I_2 \}$

k-Strengthen rule of [Brillout et al. IJCAR'10]

Str.
$$\frac{\sum_{i} c_{i} x_{i} + c \leq 0 \left[\left\{ \langle t \leq 0, \top \rangle \right\} \right]}{\sum_{i} c_{i} x_{i} + d \cdot \left\lceil \frac{c}{d} \right\rceil \leq 0 \left[I \right]}, d > 0 \text{ divides the } c_{i} \text{'s}$$
$$I := \left\{ \langle (t + n \leq 0), \ (t + n = 0) \rangle \mid 0 \leq n < d \cdot \left\lceil \frac{c}{d} \right\rceil - c \right\} \cup \left\{ \langle (t + d \cdot \left\lceil \frac{c}{d} \right\rceil - c \leq 0), \top \rangle \right\}$$





$$A := \begin{cases} -y - 4x - 1 \le 0\\ y + 4x \le 0 \end{cases} \qquad B := \begin{cases} -y - 4z + 1 \le 0\\ y + 4z - 2 \le 0 \end{cases}$$

 $y + 4x \le 0 \qquad -y - 4z + 1 \le 0$

 $4x - 4z + 1 \le 0$

 $-y - 4x - 1 \le 0$ $y + 4z - 2 \le 0$

 $4x - 4z + 1 + 3 \le 0$

 $-4x + 4z - 3 \le 0$

$$(1 \le 0) \equiv \bot$$



$$\mathbf{A} := \begin{cases} -y - 4x - 1 \le 0\\ y + 4x \le 0 \end{cases} \qquad \mathbf{B} := \begin{cases} -y - 4z + 1 \le 0\\ y + 4z - 2 \le 0 \end{cases}$$

 $y+4x \leq 0 \qquad -y-4z+1 \leq 0 \\ [\{\langle y+4x \leq 0, \top \rangle\}] \ [\{\langle 0 \leq 0, \top \rangle\}]$

$$\begin{array}{ll} 4x - 4z + 1 \leq 0 \\ [\{\langle y + 4x \leq 0, \top \rangle\}] \end{array} & \begin{array}{ll} -y - 4x - 1 \leq 0 & y + 4z - 2 \leq 0 \\ [\{\langle -y - 4x - 1 \leq 0, \top \rangle\}] & [\{\langle 0 \leq 0, \top \rangle\}] \end{array}$$

 $\begin{array}{ll} 4x - 4z + 1 + \mathbf{3} \leq 0 & -4x + 4z - 3 \leq 0 \\ [\{\langle y + 4x + n \leq 0, y + 4x + n = 0 \rangle \mid & [\{\langle -y - 4x - 1 \leq 0, \top \rangle\}] \\ 0 \leq n < 3\} \cup \{\langle y + 4x + 2 \leq 0, \top \rangle\}] \end{array}$

$$(1 \leq 0) \equiv \bot$$
$$[\{\langle n-1 \leq 0, y+4x+n=0 \rangle \mid 0 \leq n < 3\} \cup \{\langle 2-1 \leq 0, \top \rangle\}]$$



$$\mathbf{A} := \begin{cases} -y - 4x - 1 \le 0\\ y + 4x \le 0 \end{cases} \qquad \mathbf{B} := \begin{cases} -y - 4z + 1 \le 0\\ y + 4z - 2 \le 0 \end{cases}$$

 $y+4x \leq 0 \qquad -y-4z+1 \leq 0 \\ [\{\langle y+4x \leq 0, \top \rangle\}] \ [\{\langle 0 \leq 0, \top \rangle\}]$

$$\begin{array}{l} 4x - 4z + 1 \leq 0 \\ [\{\langle y + 4x \leq 0, \top \rangle\}] \end{array} \qquad \begin{array}{l} -y - 4x - 1 \leq 0 \quad y + 4z - 2 \leq 0 \\ [\{\langle -y - 4x - 1 \leq 0, \top \rangle\}] \quad [\{\langle 0 \leq 0, \top \rangle\}] \end{array}$$

 $\begin{array}{ll} 4x - 4z + 1 + \mathbf{3} \leq 0 & -4x + 4z - 3 \leq 0 \\ [\{\langle y + 4x + n \leq 0, y + 4x + n = 0 \rangle \mid & [\{\langle -y - 4x - 1 \leq 0, \top \rangle\}] \\ 0 \leq n < 3\} \cup \{\langle y + 4x + 2 \leq 0, \top \rangle\}] \end{array}$

$$(1 \le 0) \equiv \bot$$

[{ $\langle n-1 \le 0, y+4x+n=0 \rangle \mid 0 \le n < 3$ } \cup { $\langle 2-1 \le 0, \top \rangle$ }]
Interpolant: $(y =_4 0) \lor (y+1 =_4 0)$



- Interpolation of Strengthen creates potentially very big disjunctions
 - Linear in the strengthening factor k := d \[\[\frac{c}{d} \] \] c
 Can be exponential in the size of the proof

Example:

$$A := \begin{cases} -y - 4x - 1 \le 0 \\ y + 4x \le 0 \end{cases}$$
 $B := \begin{cases} -y - 4z + 1 \le 0 \\ y + 4z - 2 \le 0 \end{cases}$
Interpolant: $(y =_4 0) \lor (y + 1 =_4 0)$



- Interpolation of Strengthen creates potentially very big disjunctions
 - Linear in the strengthening factor k := d \[\[\frac{c}{d} \] \] c
 Can be exponential in the size of the proof

$$\begin{array}{l} \text{Example:} \\ A := \left\{ \begin{array}{l} -y - 2nx - n + 1 \leq 0 \\ y + 2nx \leq 0 \end{array} \right. B := \left\{ \begin{array}{l} -y - 2nz + 1 \leq 0 \\ y + 2nz - n \leq 0 \end{array} \right. \\ \text{Interpolant:} (y =_{2n} 0) \lor (y + 1 =_{2n} 0) \lor \ldots \lor (y =_{2n} n - 1) \end{array} \right. \end{array}$$



- Interpolation of Strengthen creates potentially very big disjunctions
 - Linear in the strengthening factor k := d \[\[\frac{c}{d} \] \] c
 Can be exponential in the size of the proof

$$\begin{array}{l} \text{Example:} \\ A := \left\{ \begin{array}{l} -y - 2nx - n + 1 \leq 0 \\ y + 2nx \leq 0 \end{array} \right. B := \left\{ \begin{array}{l} -y - 2nz + 1 \leq 0 \\ y + 2nz - n \leq 0 \end{array} \right. \\ \text{Interpolant:} (y =_{2n} 0) \lor (y + 1 =_{2n} 0) \lor \ldots \lor (y =_{2n} n - 1) \end{array} \right. \end{array}$$

The problem are AB-mixed cuts:

Strengthen
$$\frac{\sum_{x_i \notin B} c_i x_i + \sum_{y_j \notin A} c_j y_j + c \leq 0}{\sum_{x_i \notin B} c_i x_i + \sum_{y_j \notin A} c_j y_j + d \cdot \lceil \frac{c}{d} \rceil \leq 0}$$



- Idea: use a different extension of the signature of LIA, and extend also its domain
 - Introduce the ceiling function $\lceil \cdot \rceil$ [Pudlák '97]
 - Allow non-variable terms to be non-integers (e.g. $\frac{x}{2}$)
- Much simpler interpolation procedure
 - Proof annotations are single inequalities $(t \le 0)$



- Idea: use a different extension of the signature of LIA, and extend also its domain
 - Introduce the ceiling function $\lceil \cdot \rceil$ [Pudlák '97]
 - Allow non-variable terms to be non-integers (e.g. $\frac{x}{2}$)
- Much simpler interpolation procedure
 - Proof annotations are single inequalities $(t \le 0)$

$$\begin{aligned} \text{Hyp} & \frac{-}{t \leq 0 \ [t' \leq 0]} & \text{Comb} \ \frac{t_1 \leq 0 \ [t'_1 \leq 0]}{c_1 \cdot t_1 + c_2 \cdot t_2 \leq 0 \ [c_1 \cdot t'_1 + c_2 \cdot t'_2 \leq 0]} \\ \\ \text{Div} & \frac{\sum_{y_j \notin B} a_j y_j + \sum_{z_k \notin A} b_k z_k + \sum_{x_i \in A \cap B} c_i x_i + c}{\left[\sum_{y_j \notin B} a_j y_j + \sum_{x_i \in A \cap B} c'_i x_i + t'\right]} \\ \\ \frac{\sum_{y_j \notin B} \frac{a_j}{d} y_j + \sum_{z_k \in B} \frac{b_k}{d} z_k + \sum_{x_i \in A \cap B} \frac{c_i}{d} x_i + \lceil \frac{c}{d} \rceil}{\left[\sum_{y_j \notin B} \frac{a_j}{d} y_j + \lceil \frac{\sum_{x_i \in A \cap B} c'_i x_i + t'}{d} \rceil\right]} d > 0 \text{ divides } a_j, b_k, c_i \end{aligned}$$



No blowup of interpolants wrt. the size of the proofs

$$A := \begin{cases} -y - 2nx - n + 1 \le 0\\ y + 2nx \le 0 \end{cases} \qquad B := \begin{cases} -y - 2nz + 1 \le 0\\ y + 2nz - n \le 0 \end{cases}$$

 $y + 2nx \le 0 \quad -y - 2nz + 1 \le 0$

$$2nx - 2nz + 1 \le 0$$

 $-y - 2nx - n + 1 \le 0 \quad y + 2nz - n \le 0$

 $2n \cdot (x - z + 1 \le 0) \qquad \qquad -2nx + 2nz - 2n + 1 \le 0$

$$(1 \le 0) \equiv \bot$$



No blowup of interpolants wrt. the size of the proofs

$$A := \begin{cases} -y - 2nx - n + 1 \le 0 \\ y + 2nx \le 0 \end{cases} \quad B := \begin{cases} -y - 2nz + 1 \le 0 \\ y + 2nz - n \le 0 \end{cases}$$
$$y + 2nx \le 0 \quad -y - 2nz + 1 \le 0 \\ y + 2nx \le 0 \end{bmatrix} \quad [0 \le 0]$$
$$2nx - 2nz + 1 \le 0 \\ [y + 2nx \le 0] \\ 2n \cdot (x - z + 1 \le 0) \\ [x + \lceil \frac{y}{2n} \rceil \le 0] \end{cases} \quad \begin{array}{c} -y - 2nx - n + 1 \le 0 \\ [-y - 2nx - n + 1 \le 0] \\ -2nx + 2nz - 2n + 1 \le 0 \\ [-y - 2nx - n + 1 \le 0] \\ [-y - 2nx - n + 1 \le 0] \end{cases}$$

$$(1 \le 0) \equiv \bot$$
$$[2n \lceil \frac{y}{2n} \rceil - y - n + 1 \le 0]$$



No blowup of interpolants wrt. the size of the proofs

$$A := \begin{cases} -y - 2nx - n + 1 \le 0 \\ y + 2nx \le 0 \end{cases} \quad B := \begin{cases} -y - 2nz + 1 \le 0 \\ y + 2nz - n \le 0 \end{cases}$$
$$y + 2nx \le 0 \quad -y - 2nz + 1 \le 0 \\ y + 2nx \le 0 \end{bmatrix} \quad [0 \le 0]$$
$$2nx - 2nz + 1 \le 0 \\ [y + 2nx \le 0] \\ 2n \cdot (x - z + 1 \le 0) \\ [x + \lceil \frac{y}{2n} \rceil \le 0] \end{cases} \quad \begin{array}{r} -y - 2nx - n + 1 \le 0 \\ -y - 2nx - n + 1 \le 0 \\ [-y - 2nx - n + 1 \le 0] \\ -2nx + 2nz - 2n + 1 \le 0 \\ [-y - 2nx - n + 1 \le 0] \\ -2nx + 2nz - 2n + 1 \le 0 \\ [-y - 2nx - n + 1 \le 0] \\ -2nx + 2nz - 2n + 1 \le 0 \\ [-y - 2nx - n + 1 \le 0] \\ -2nx + 2nz - 2n + 1 \le 0 \\ [-y - 2nx - n + 1 \le 0] \\ \end{array}$$

$$(1 \le 0) \equiv \bot$$

Interpolant: $[2n \lceil \frac{y}{2n} \rceil - y - n + 1 \le 0]$



- Like modular equations, also ceilings can be eliminated via preprocessing
 - Replace every term $\lceil t \rceil$ with a fresh integer variable $x_{\lceil t \rceil}$
 - Add the 2 unit clauses (encoding the meaning of ceiling: $\lceil t \rceil 1 < t \leq \lceil t \rceil$)

$$\begin{aligned} &(l \cdot x_{\lceil t \rceil} - l \cdot t + l \leq 0) \\ &(l \cdot t - l \cdot x_{\lceil t \rceil} \leq 0) \end{aligned}$$

where l is the least common multiple of the denominators of the coefficients in t



- Interpolation for bit-vectors is hard
 - Only some limited work done so far
- Most efficient solvers use eager encoding into SAT, which is efficient but not good for interpolation
 - Easy in principle, but not very useful interpolants
- Try to exploit lazy bit-blasting to incorporate BV into DPLL(T)



- Interpolation via bit-blasting is easy...
 - From A_{BV} and B_{BV} generate A_{Bool} and B_{Bool} Each var x of width n encoded with n Boolean vars $b_1^x \dots b_n^x$
 - Generate a Boolean interpolant I_{BOOL} for (A_{BOOL}, B_{BOOL})
 - Replace every variable b_i^x in IBool with the bit-selection x[i] and every Boolean connective with the corresponding bit-wise connective: $\land \mapsto \&, \lor \lor \mid, \neg \mapsto \sim$

...but quite impractical

- Generates "ugly" interpolants
- Word-level structure of the original problem completely lost
 - How to apply word-level simplifications?



$$\begin{split} \mathbf{A} \stackrel{\text{def}}{=} & (\mathbf{a}_{[8]} * b_{[8]} = 15_{[8]}) \land (\mathbf{a}_{[8]} = 3_{[8]}) \\ \mathbf{B} \stackrel{\text{def}}{=} & \neg (b_{[8]} \%_u \mathbf{c}_{[8]} = 1_{[8]}) \land (\mathbf{c}_{[8]} = 2_{[8]}) \end{split}$$

A word-level interpolant is:

$$I \stackrel{\text{\tiny def}}{=} (b_{[8]} * 3_{[8]} = 15_{[8]})$$

...but with bit-blasting we get:

 $I' \stackrel{\text{\tiny def}}{=} (b_{[8]}[0] = 1_{[1]}) \land ((b_{[8]}[0]\& \sim ((((((\sim b_{[8]}[7]\& \sim b_{[8]}[6])\& \sim b_{[8]}[6])\& \sim b_{[8]}[5])\& \sim b_{[8]}[4])\& \sim b_{[8]}[3])\& b_{[8]}[2])\& \sim b_{[8]}[1])) = 0_{[1]})$

Alternative: lazy bit-blasting and DPLL(T) ES STREEDED ->C

- Exploit <u>lazy bit-blasting</u>
 - Bit-blast only BV-atoms, not the whole formula
 - Boolean skeleton of the formula handled by the "main" DPLL, like in DPLL(T)
 - Conjunctions of BV-atoms handled (via bit-blasting) by a "sub"-DPLL (DPLL-BV) that acts as a BV-solver

Standard Boolean Interpolation



BV-specific Interpolation for *conjunctions of constraints*

Interpolation for BV constraints



A layered approach

- Apply in sequence a chain of procedures of increasing generality and cost
 - Interpolation in EUF
 - Interpolation via equality inlining
 - Interpolation via Linear Integer Arithmetic encoding
 - Interpolation via bit-blasting



- Treat all the BV-operators as uninterpreted functions
- Exploit cheap, efficient algorithms for solving and interpolating modulo EUF
 - Possible because we avoid bit-blasting upront!

Example:
$$A \stackrel{\text{def}}{=} (x_{1[32]} = 3_{[32]}) \land (x_{3[32]} = x_{1[32]} \cdot x_{2[32]})$$

 $B \stackrel{\text{def}}{=} (x_{4[32]} = x_{2[32]}) \land (x_{5[32]} = 3_{[32]} \cdot x_{4[32]}) \land$
 $\neg (x_{3[32]} = x_{5[32]})$
 $I_{\mathsf{UF}} \stackrel{\text{def}}{=} x_3 = f^{\cdot}(f^3, x_2)$
 $I_{\mathsf{BV}} \stackrel{\text{def}}{=} x_{3[32]} = 3_{[32]} \cdot x_{2[32]}$



- Interpolation via quantifier elimination: given (A, B), an interpolant can be computed by eliminating quantifiers from $\exists_{x \notin B} A$ or from $\exists_{x \notin A} \neg B$
- In general, this can be very expensive for BV
 - Might require bit-blasting and can cause blow-up of the formula
- Cheap case: non-common variables occurring in "definitional" equalities

Example: $(x=e)\wedge \varphi$ and x does not occur in e, then

$$\exists_x ((x=e) \land \varphi) \Longrightarrow \varphi[x \mapsto e]$$



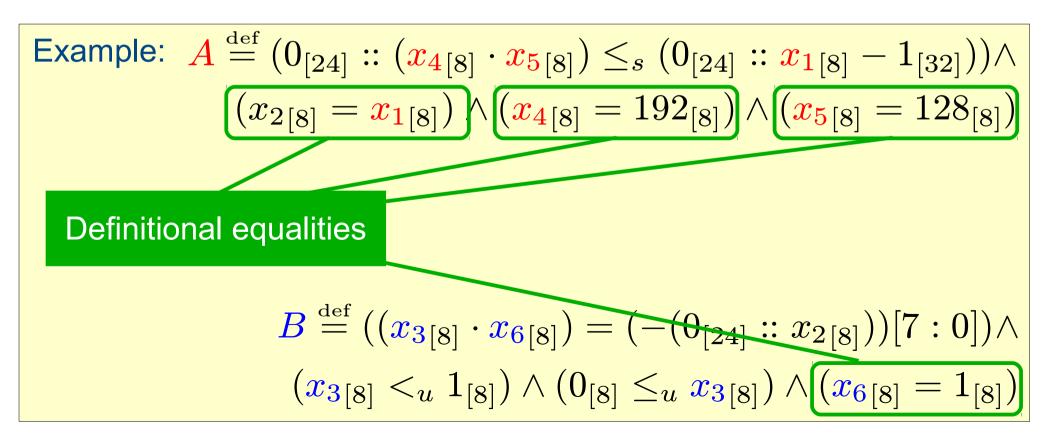
- Inline definitional equalities until either all all non-common variables are removed, or a fixpoint is reached
 - Try both from \underline{A} and $\neg \underline{B}$
 - If one of them succeeds, we have an interpolant

Example:
$$A \stackrel{\text{def}}{=} (0_{[24]} :: (x_{4[8]} \cdot x_{5[8]}) \leq_s (0_{[24]} :: x_{1[8]} - 1_{[32]})) \land (x_{2[8]} = x_{1[8]}) \land (x_{4[8]} = 192_{[8]}) \land (x_{5[8]} = 128_{[8]})$$

$$B \stackrel{\text{def}}{=} \left(\left(x_{3[8]} \cdot x_{6[8]} \right) = \left(-\left(0_{[24]} :: x_{2[8]} \right) \right) [7:0] \right) \land \\ \left(x_{3[8]} <_{u} 1_{[8]} \right) \land \left(0_{[8]} \leq_{u} x_{3[8]} \right) \land \left(x_{6[8]} = 1_{[8]} \right)$$



- Inline definitional equalities until either all all non-common variables are removed, or a fixpoint is reached
 - Try both from A and $\neg B$
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Example:
$$A \stackrel{\text{def}}{=} (0_{[24]} :: (x_{4[8]} \cdot x_{5[8]}) \leq_s (0_{[24]} :: x_{1[8]} - 1_{[32]})) \land$$

 $(x_{2[8]} = x_{1[8]}) \land (x_{4[8]} = 192_{[8]}) \land (x_{5[8]} = 128_{[8]})$
 $B \stackrel{\text{def}}{=} ((x_{3[8]} \cdot x_{6[8]}) = (-(0_{[24]} :: x_{2[8]}))[7:0]) \land$
 $(x_{3[8]} <_u 1_{[8]}) \land (0_{[8]} \leq_u x_{3[8]}) \land (x_{6[8]} = 1_{[8]})$



- Inline definitional equalities until either all all non-common variables are removed, or a fixpoint is reached
 - Try both from A and $\neg B$
 - If one of them succeeds, we have an interpolant

Example: $A \stackrel{\text{def}}{=} (0_{[24]} :: (x_{4[8]} \cdot x_{5[8]}) \leq_s (0_{[24]} :: x_{2[8]} - 1_{[32]})) \land (x_{4[8]} = 192_{[8]}) \land (x_{5[8]} = 128_{[8]})$

 $B \stackrel{\text{\tiny def}}{=} \left(\left(x_{3[8]} \cdot x_{6[8]} \right) = \left(-\left(0_{[24]} :: x_{2[8]} \right) \right) [7:0] \right) \land \\ \left(x_{3[8]} <_{u} 1_{[8]} \right) \land \left(0_{[8]} \leq_{u} x_{3[8]} \right) \land \left(x_{6[8]} = 1_{[8]} \right)$



- Inline definitional equalities until either all all non-common variables are removed, or a fixpoint is reached
 - Try both from A and $\neg B$
 - If one of them succeeds, we have an interpolant

Example:
$$A \stackrel{\text{def}}{=} (0_{[24]} :: (x_{4[8]} \cdot x_{5[8]}) \leq_s (0_{[24]} :: x_{2[8]} - 1_{[32]})) \land \land \land (x_{4[8]} = 192_{[8]}) \land (x_{5[8]} = 128_{[8]})$$

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- Inline definitional equalities until either all all non-common variables are removed, or a fixpoint is reached
 - Try both from A and $\neg B$
 - If one of them succeeds, we have an interpolant

Example:
$$A \stackrel{\text{def}}{=} (0_{[24]} :: (192_{[8]} \cdot 128_{[8]}) \leq_s (0_{[24]} :: x_{2[8]} - 1_{[32]}))$$

 $\land \qquad \land$
 $B \stackrel{\text{def}}{=} ((x_{3[8]} \cdot x_{6[8]}) = (-(0_{[24]} :: x_{2[8]}))[7:0]) \land$
 $(x_{3[8]} <_u 1_{[8]}) \land (0_{[8]} \leq_u x_{3[8]}) \land (x_{6[8]} = 1_{[8]})$



- Inline definitional equalities until either all all non-common variables are removed, or a fixpoint is reached
 - Try both from \underline{A} and $\neg \underline{B}$
 - If one of them succeeds, we have an interpolant

Example:
$$A \stackrel{\text{def}}{=} (0_{[24]} :: (192_{[8]} \cdot 128_{[8]}) \leq_s (0_{[24]} :: x_{2[8]} - 1_{[32]}))$$

 $\land \qquad \land$
 $I \stackrel{\text{def}}{=} (0_{32} \leq_s (0_{24} :: x_{2[8]} - 1_{[32]}))$
 $B \stackrel{\text{def}}{=} ((x_{3[8]} \cdot x_{6[8]}) = (-(0_{[24]} :: x_{2[8]}))[7:0]) \land$
 $(x_{3[8]} <_u 1_{[8]}) \land (0_{[8]} \leq_u x_{3[8]}) \land (x_{6[8]} = 1_{[8]}))$



- Simple idea (in principle):
 - Encode a set of BV-constraints into an SMT(LIA)-formula
 - Generate a LIA-interpolant using existing algorithms
 - Map back to a BV-interpolant

- However, several problems to solve:
 - Efficiency
 - More importantly, soundness



- Use well-known encodings from BV to SMT(LIA)
 - Encode each BV term $t_{[n]}$ as an integer variable x_t and the constraints $(0 \le x_t) \land (x_t \le 2^n 1)$
 - Encode each BV operation as a LIA-formula.

$$\begin{aligned} & \underset{[i-j+1]}{\overset{\text{def}}{=}} t_{1[n]}[i:j] \implies (x_t = m) \land (x_{t_1} = 2^{i+1}h + 2^jm + l) \land \\ & l \in [0, 2^i) \land m \in [0, 2^{i-j+1}) \land h \in [0, 2^{n-i-1}) \end{aligned} \\ & t_{[n]} \overset{\text{def}}{=} t_{1[n]} + t_{2[n]} \implies (x_t = x_{t_1} + x_{t_2} - 2^n\sigma) \land (0 \le \sigma \le 1) \end{aligned} \\ & t_{[n]} \overset{\text{def}}{=} t_{1[n]} \cdot k \qquad \Longrightarrow (x_t = k \cdot x_{t_1} - 2^n\sigma) \land (0 \le \sigma \le k) \end{aligned}$$



- "Invert" the LIA encoding to get a BV interpolant
- Unsound in general
 - Issues due to overflow and (un)signedness of operations
- Our (very simple) solution: <u>check the interpolants</u>
 - Given a candidate interpolant \hat{I} , use our SMT(BV) solver to check the unsatisfiability of $(A \land \neg \hat{I}) \lor (B \land \hat{I})$
 - If successful, then \hat{I} is an interpolant



$$\begin{split} A \stackrel{\text{def}}{=} & (y_{1[8]} = y_{5[4]} :: y_{5[4]}) \land (y_{1[8]} = y_{2[8]}) \land (y_{5[4]} = 1_{[4]}) \\ B \stackrel{\text{def}}{=} & \neg (y_{4[8]} + 1_{[8]} \leq_u y_{2[8]}) \land (y_{4[8]} = 1_{[8]}) \end{split}$$

Encoding into LIA:

$$A_{\text{LIA}} \stackrel{\text{def}}{=} (x_{y_2} = 16x_{y_5} + x_{y_5}) \land (x_{y_1} = x_{y_2}) \land (x_{y_5} = 1) \land (x_{y_1} \in [0, 2^8)) \land (x_{y_2} \in [0, 2^8)) \land (x_{y_5} \in [0, 2^4))$$

$$B_{\text{LIA}} \stackrel{\text{def}}{=} \neg (x_{y_4+1} \le x_{y_2}) \land (x_{y_4+1} = x_{y_4} + 1 - 2^8 \sigma) \land$$
$$(x_{y_4} = 1) \land$$
$$(x_{y_4+1} \in [0, 2^8)) \land (x_{y_4} \in [0, 2^8)) \land (0 \le \sigma \le 1)$$



$$\begin{split} A &\stackrel{\text{def}}{=} (y_{1[8]} = y_{5[4]} :: y_{5[4]}) \land (y_{1[8]} = y_{2[8]}) \land (y_{5[4]} = 1_{[4]}) \\ B &\stackrel{\text{def}}{=} \neg (y_{4[8]} + 1_{[8]} \leq_u y_{2[8]}) \land (y_{4[8]} = 1_{[8]}) \end{split}$$

LIA-Interpolant:

$$I_{\mathrm{LIA}} \stackrel{\mathrm{\tiny def}}{=} (17 \le x_{y_2})$$

BV-interpolant:

$$I \stackrel{\text{\tiny def}}{=} (17_{[8]} \leq_u y_{2[8]})$$





$$A \stackrel{\text{\tiny def}}{=} (y_{2[8]} = 81_{[8]}) \land (y_{3[8]} = 0_{[8]}) \land (y_{4[8]} = y_{2[8]})$$

 $B \stackrel{\text{\tiny def}}{=} (y_{13[16]} = 0_{[8]} :: y_{4[8]}) \land (255_{[16]} \leq_u y_{13[16]} + (0_{[8]} :: y_{3[8]}))$

Encoding into LIA:

$$A_{\text{LIA}} \stackrel{\text{def}}{=} (x_{y_2} = 81) \land (x_{y_3} = 0) \land (x_{y_4} = x_{y_2}) \land (x_{y_2} \in [0, 2^8)) \land (x_{y_3} \in [0, 2^8)) \land (x_{y_4} \in [0, 2^8))$$

$$B_{\text{LIA}} \stackrel{\text{def}}{=} (x_{y_{13}} = 2^8 \cdot 0 + x_{y_4}) \land (255 \le x_{y_{13} + (0::y_3)}) \land (x_{y_{13} + (0::y_3)} = x_{y_{13}} + 2^8 \cdot 0 + x_{y_3} - 2^{16}\sigma) \land (x_{y_{13}} \in [0, 2^{16})) \land (x_{y_{13} + (0::y_3)} \in [0, 2^{16})) \land (0 \le \sigma \le 1)$$



$$A \stackrel{\text{\tiny def}}{=} (y_{2[8]} = 81_{[8]}) \land (y_{3[8]} = 0_{[8]}) \land (y_{4[8]} = y_{2[8]})$$

 $B \stackrel{\text{\tiny def}}{=} (y_{13[16]} = 0_{[8]} :: y_{4[8]}) \land (255_{[16]} \leq_u y_{13[16]} + (0_{[8]} :: y_{3[8]}))$

LIA-interpolant:

$$I_{\text{LIA}} \stackrel{\text{\tiny def}}{=} (x_{y_3} + x_{y_4} \le 81)$$

BV-interpolant:

$$\hat{I} \stackrel{\text{\tiny def}}{=} (y_{3[8]} + y_{4[8]} \le_u 81_{[8]})$$





$$A \stackrel{\text{\tiny def}}{=} (y_{2[8]} = 81_{[8]}) \land (y_{3[8]} = 0_{[8]}) \land (y_{4[8]} = y_{2[8]})$$

 $\mathbf{B} \stackrel{\text{\tiny def}}{=} (\mathbf{y_{13}}_{[16]} = \mathbf{0}_{[8]} :: y_{4[8]}) \land (255_{[16]} \leq_u \mathbf{y_{13}}_{[16]} + (\mathbf{0}_{[8]} :: y_{3[8]}))$

LIA-interpolant:

$$I_{\text{LIA}} \stackrel{\text{\tiny def}}{=} (x_{y_3} + x_{y_4} \le 81)$$

Addition might overflow in BV!

BV-interpolant:

$$\hat{I} \stackrel{\text{\tiny def}}{=} (y_{3[8]} + y_{4[8]} \not)_u 81_{[8]})$$





$$A \stackrel{\text{\tiny def}}{=} (y_{2[8]} = 81_{[8]}) \land (y_{3[8]} = 0_{[8]}) \land (y_{4[8]} = y_{2[8]})$$

 $\mathbf{B} \stackrel{\text{\tiny def}}{=} (\mathbf{y_{13}}_{[16]} = \mathbf{0}_{[8]} :: y_{4[8]}) \land (255_{[16]} \leq_u \mathbf{y_{13}}_{[16]} + (\mathbf{0}_{[8]} :: y_{3[8]}))$

LIA-interpolant:

$$I_{\text{LIA}} \stackrel{\text{def}}{=} (x_{y_3} + x_{y_4} \le 81)$$

Addition might overflow in BV!

BV-interpolant:

A correct interpolant would be $I \stackrel{\text{def}}{=} (0_{[1]} :: y_{3[8]} + 0_{[1]} :: y_{4[8]} \leq_u 81_{[9]})$





$$\begin{split} & A \stackrel{\text{def}}{=} \neg (y_{4[8]} + 1_{[8]} \leq_{u} y_{3[8]}) \land (y_{2[8]} = y_{4[8]} + 1_{[8]}) \\ & B \stackrel{\text{def}}{=} (y_{2[8]} + 1_{[8]} \leq_{u} y_{3[8]}) \land (y_{7[8]} = 3_{[8]}) \land (y_{7[8]} = y_{2[8]} + 1_{[8]}) \end{split}$$

Encoding into LIA:

$$A_{\text{LIA}} \stackrel{\text{def}}{=} \neg (x_{y_4+1} \le x_{y_3}) \land (x_{y_2} = x_{y_4+1}) \land \\ (x_{y_4+1} = x_{y_4} + 1 - 2^8 \sigma_1) \land \\ (x_{y_2} \in [0, 2^8)) \land (x_{y_3} \in [0, 2^8)) \land (x_{y_4} \in [0, 2^8)) \land \\ (x_{y_4+1} \in [0, 2^8)) \land (0 \le \sigma_1 \le 1)$$

$$B_{\text{LIA}} \stackrel{\text{def}}{=} (x_{y_2+1} \le x_{y_3}) \land (x_{y_7} = 3) \land (x_{y_7} = x_{y_2+1}) \land$$
$$(x_{y_2+1} = x_{y_2} + 1 - 2^8 \sigma_2) \land$$
$$(x_{y_7} \in [0, 2^8)) \land (x_{y_2+1} \in [0, 2^8)) \land (0 \le \sigma_2 \le 1)$$



$$\begin{split} & A \stackrel{\text{def}}{=} \neg (y_{4[8]} + 1_{[8]} \leq_{u} y_{3[8]}) \land (y_{2[8]} = y_{4[8]} + 1_{[8]}) \\ & B \stackrel{\text{def}}{=} (y_{2[8]} + 1_{[8]} \leq_{u} y_{3[8]}) \land (y_{7[8]} = 3_{[8]}) \land (y_{7[8]} = y_{2[8]} + 1_{[8]}) \end{split}$$

LIA-interpolant:

$$I_{\text{LIA}} \stackrel{\text{def}}{=} \left(-255 \le x_{y_2} - x_{y_3} + 256 \lfloor -1\frac{x_{y_2}}{256} \rfloor\right)$$

BV-interpolant:(after fixing overflows) $\hat{I'} \stackrel{\text{def}}{=} (65281_{[16]} \leq_u (0_{[8]} :: y_{2[8]}) - (0_{[8]} :: y_{3[8]}) + 256_{[16]} \cdot (65535_{[16]} \cdot (0_{[8]} :: y_{2[8]}) / u 256_{[16]}))$



$$\begin{split} & A \stackrel{\text{def}}{=} \neg (\mathbf{y}_{4[8]} + \mathbf{1}_{[8]} \leq_{u} y_{3[8]}) \land (y_{2[8]} = \mathbf{y}_{4[8]} + \mathbf{1}_{[8]}) \\ & B \stackrel{\text{def}}{=} (y_{2[8]} + \mathbf{1}_{[8]} \leq_{u} y_{3[8]}) \land (\mathbf{y}_{7[8]} = \mathbf{3}_{[8]}) \land (\mathbf{y}_{7[8]} = y_{2[8]} + \mathbf{1}_{[8]}) \end{split}$$

LIA-interpolant:

$$I_{\rm LIA} \stackrel{\rm def}{=} \left(-255 \le x_{y_2} - x_{y_3} + 256 \lfloor -1\frac{x_{y_2}}{256} \rfloor\right)$$

 BV-interpolant:
 (after fixing overflows)

 $\hat{I'} \stackrel{\text{def}}{=} (65281_{[16]} \leq_u 0_{[8]} :: y_{2[8]}) - (0_{[8]} :: y_{3[8]}) + 256_{[16]} \cdot (65535_{[16]} \cdot (0_{[8]} :: y_{2[8]}) / u 256_{[16]}))$

 In this case, the problem is also the sign
 Still Wrong!



$$\begin{split} & A \stackrel{\text{def}}{=} \neg (y_{4[8]} + 1_{[8]} \leq_{u} y_{3[8]}) \land (y_{2[8]} = y_{4[8]} + 1_{[8]}) \\ & B \stackrel{\text{def}}{=} (y_{2[8]} + 1_{[8]} \leq_{u} y_{3[8]}) \land (y_{7[8]} = 3_{[8]}) \land (y_{7[8]} = y_{2[8]} + 1_{[8]}) \end{split}$$

LIA-interpolant:

$$I_{\text{LIA}} \stackrel{\text{\tiny def}}{=} (-255 \le x_{y_2} - x_{y_3} + 256\lfloor -1\frac{x_{y_2}}{256} \rfloor)$$

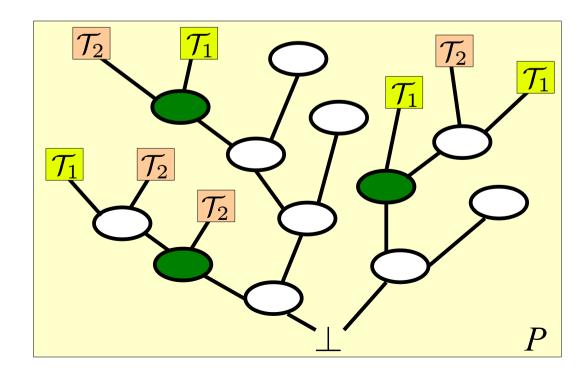
BV-interpolant:

 $I \stackrel{\text{def}}{=} (65281_{[16]} \leq_s (0_{[8]} :: y_{2[8]}) - (0_{[8]} :: y_{3[8]}) + 256_{[16]} \cdot (65535_{[16]} \cdot (0_{[8]} :: y_{2[8]}) / u \, 256_{[16]}))$

Correct interpolant



- Delayed Theory Combination (DTC): use the DPLL engine to perform theory combination
 - Independent \mathcal{T}_i -solvers, that interact only with DPLL
 - How: Boolean search space augmented with interface equalities
 - Equalities between variables shared by the two theories
- Combination of theories encoded directly in the proof of unsatisfiability P
 - \mathcal{T}_i -lemmas for the individual theories
 - P contains interface equalities



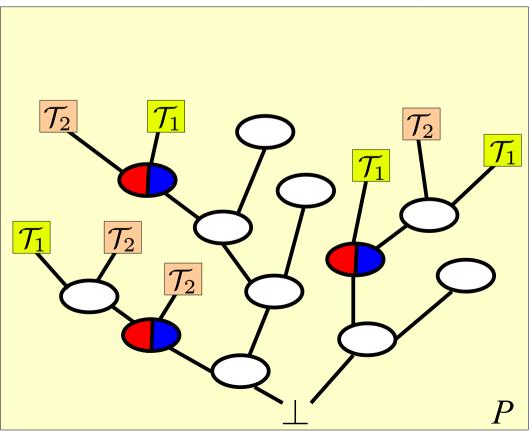


Problem for interpolation:

- Some interface equalities (x = y) are AB-mixed: $x \notin B, y \notin A$
- Interpolation procedures don't work with AB-mixed terms
- Solution: Split AB-mixed equalities occurring in *P*, and fix the proof

• How: Split each \mathcal{T} -lemma $\eta \lor (\mathbf{x} = \mathbf{y})$ into $(\eta \lor (\mathbf{x} = t)) \land$ $\eta \lor (t = \mathbf{y})$ with $t \in A \cap B$ using available algorithms

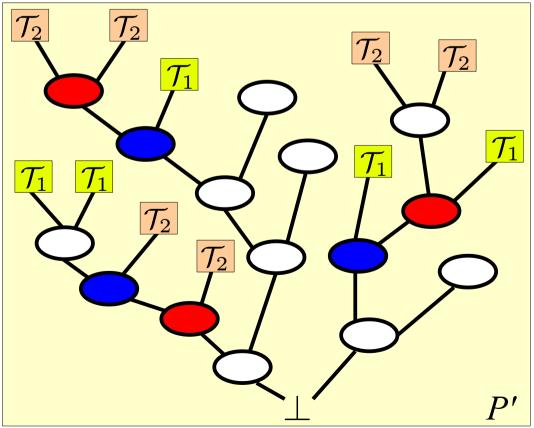
- *T_i*'s must be equalityinterpolating and convex
- Propagate the changes throughout P





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 - *T_i*'s must be equalityinterpolating and convex
 - Propagate the changes throughout P



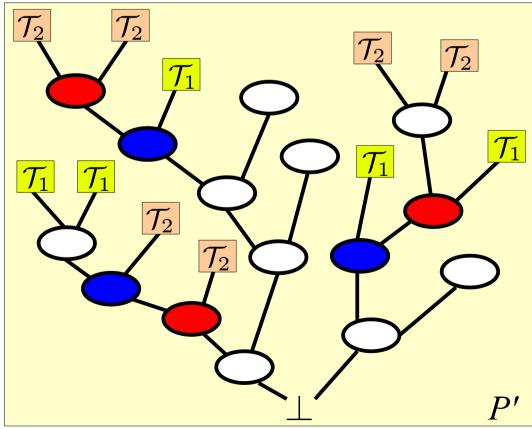


- Problem for interpolation:
 - Some interface equalities (x = y) are AB-mixed: $x \notin B, y \notin A$
 - Interpolation procedures don't work with AB-mixed terms
- Solution: Split AB-mixed equalities occurring in *P*, and fix the proof

How: Split each *T*-lemma

Problem: splitting can cause exponential blow-up in *P*

Solution: control the kind of proofs generated by DPLL, so that the splitting can be performed **efficiently** (ie-local proofs)



Interpolation in combined theories



- After splitting AB-mixed equalities, we can compute an interpolant as usual
 - Nothing special needed for theory combination!
 - Because theory combination is encoded in the proof, we can reuse the Boolean interpolation algorithm
- Features:
 - No need of ad-hoc interpolant combination procedures
 - Exploit state-of-the-art SMT solvers, based on (variants of) DTC
 - Split only when necessary



$$A := (a_1 = f(x_1)) \land (z - x_1 = 1) \land (a_1 + z = 0)$$
$$B := (a_2 = f(x_2)) \land (z - x_2 = 1) \land (a_2 + z = 1)$$



$$A := (a_1 = f(x_1)) \land (z - x_1 = 1) \land (a_1 + z = 0)$$
$$B := (a_2 = f(x_2)) \land (z - x_2 = 1) \land (a_2 + z = 1)$$

T-lemmas:

$$C_{1} \equiv (x_{1} = x_{2}) \lor \neg (z - x_{1} = 1) \lor \\ \neg (z - x_{2} = 1)$$

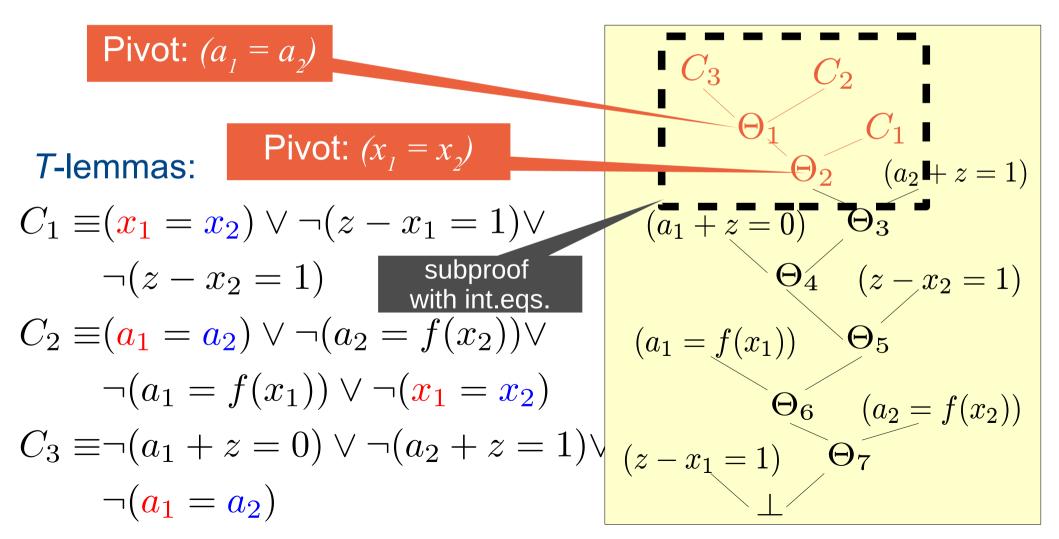
$$C_{2} \equiv (a_{1} = a_{2}) \lor \neg (a_{2} = f(x_{2})) \lor \\ \neg (a_{1} = f(x_{1})) \lor \neg (x_{1} = x_{2})$$

$$C_{3} \equiv \neg (a_{1} + z = 0) \lor \neg (a_{2} + z = 1) \lor \\ \neg (a_{1} = a_{2})$$

$$C_{3} \qquad C_{2} \\ \Theta_{1} \qquad C_{1} \\ \Theta_{2} \qquad (a_{2} + z = 1) \\ (a_{1} + z = 0) \qquad \Theta_{3} \\ \Theta_{4} \qquad (z - x_{2} = 1) \\ (a_{1} = f(x_{1})) \qquad \Theta_{5} \\ \Theta_{6} \qquad (a_{2} = f(x_{2})) \\ (z - x_{1} = 1) \qquad \Theta_{7} \\ \bot$$



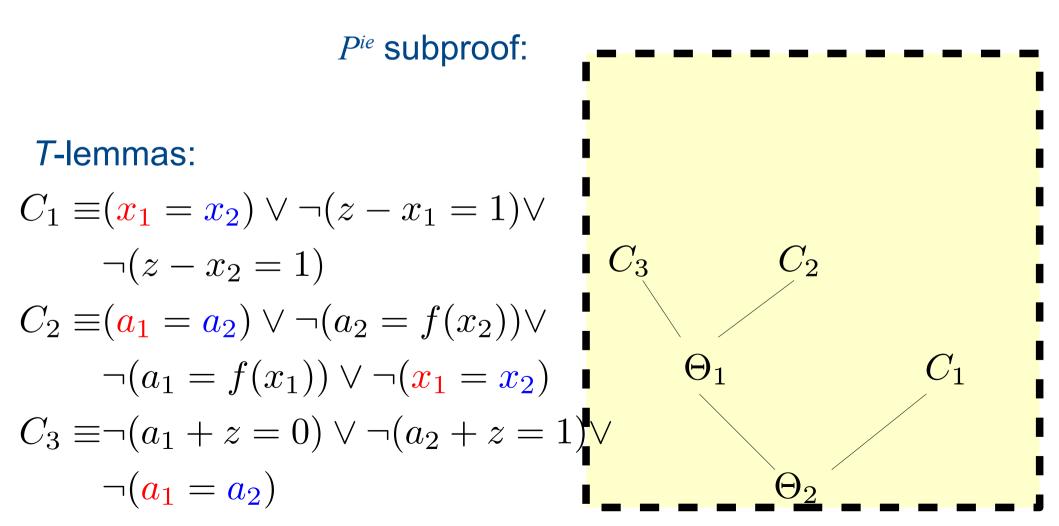
$$A := (a_1 = f(x_1)) \land (z - x_1 = 1) \land (a_1 + z = 0)$$
$$B := (a_2 = f(x_2)) \land (z - x_2 = 1) \land (a_2 + z = 1)$$





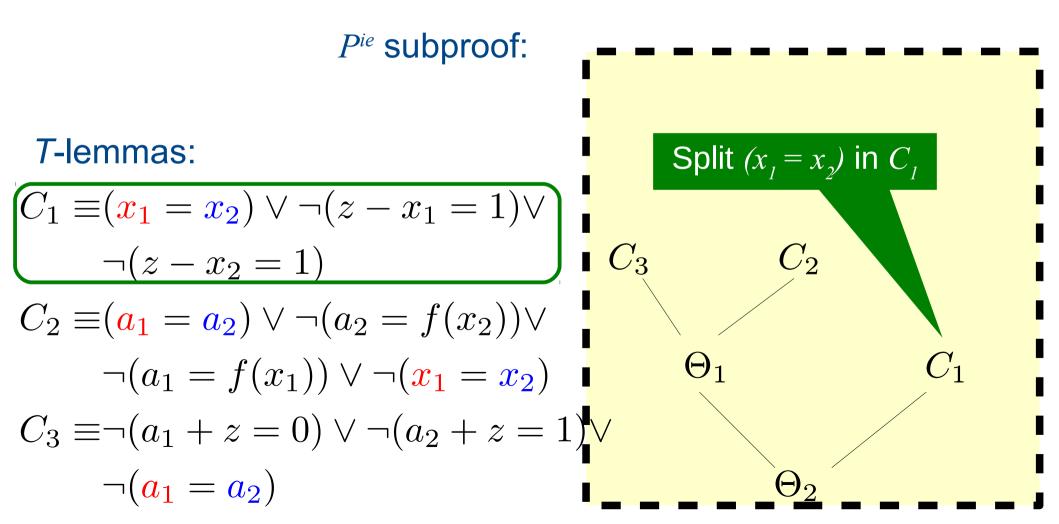
$$A := (a_1 = f(x_1)) \land (z - x_1 = 1) \land (a_1 + z = 0)$$

$$B := (a_2 = f(x_2)) \land (z - x_2 = 1) \land (a_2 + z = 1)$$





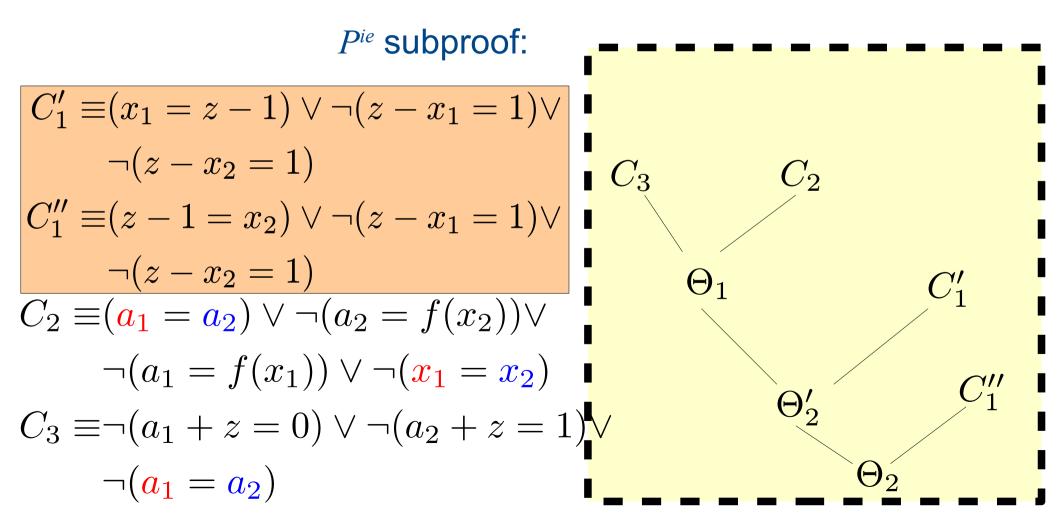
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$$A := (a_1 = f(x_1)) \land (z - x_1 = 1) \land (a_1 + z = 0)$$

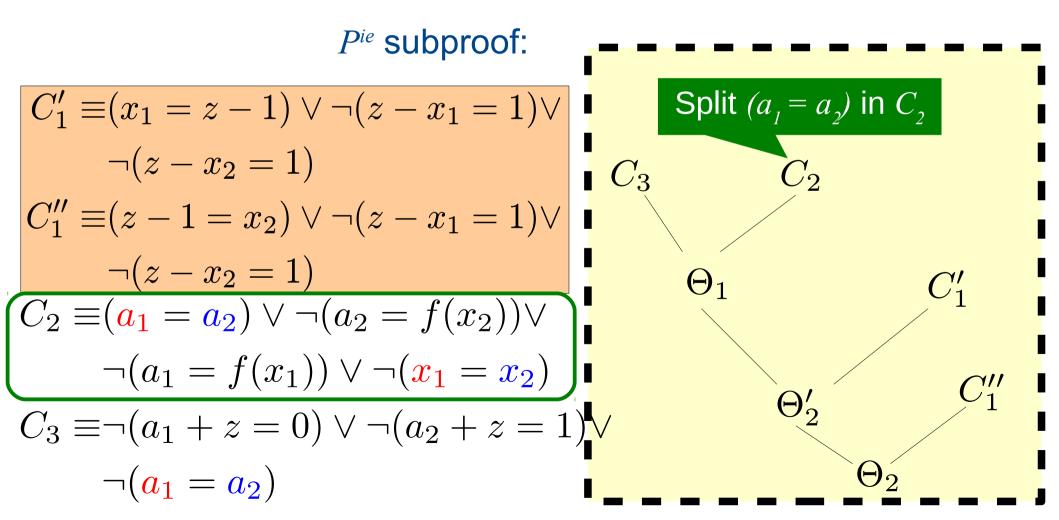
$$B := (a_2 = f(x_2)) \land (z - x_2 = 1) \land (a_2 + z = 1)$$





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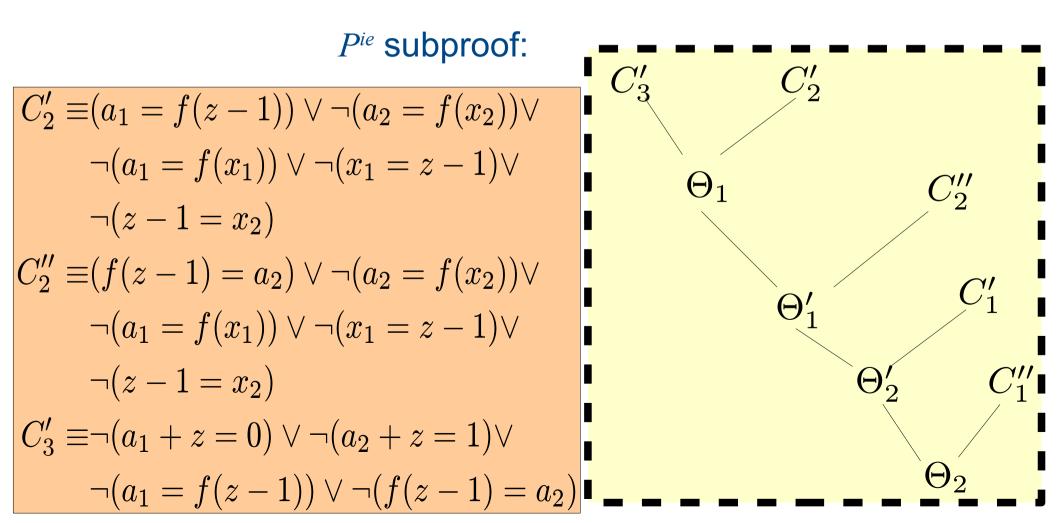
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$$A := (a_1 = f(x_1)) \land (z - x_1 = 1) \land (a_1 + z = 0)$$

$$B := (a_2 = f(x_2)) \land (z - x_2 = 1) \land (a_2 + z = 1)$$





- [Christ, Hoenicke and Nutz, TACAS 2013]
- Interpolants with AB-mixed literals without proof rewriting
 - Replace AB-mixed terms $(s \le t)$ with $(s \le x) \land (x \le t)$ in leaves, where x is a fresh purification variable
 - Eliminate the purification variable when resolving on $(s \leq t)$

 $\frac{C_1 \vee (\mathbf{s} \leq \mathbf{t}) \ [I_1(x)]}{C_1 \vee C_2 \ [I_3]} \qquad C_2 \vee \neg (\mathbf{s} \leq \mathbf{t}) \ [I_2(x)]$

- Advantages:
 - no need of proof rewriting
 - handles also for non-convex theories
- Drawbacks:
 - need T-specific interpolation rules for resolution steps
 - more complex interpolation system



- An ordered sequence of formulae F_1, \ldots, F_n such that $\bigwedge_i F_i \models \bot$
- We want a sequence of interpolants I_1, \ldots, I_{n-1} such that
 - I_k is an interpolant for $(\bigwedge_{i=1}^k F_i, \bigwedge_{j=k+1}^n F_j)$

•
$$F_k \wedge I_{k-1} \models I_k$$
 for all $k \in [2, n-1]$

- Needed in various applications (e.g. abstraction refinement)
 How to compute them?
 - In general, if we compute arbitrary binary interpolants for $(\bigwedge_{i=1}^{k} F_i, \bigwedge_{j=k+1}^{n} F_j)$, the second condition will not hold



- Compute I_1 as an interpolant of $(F_1, \bigwedge_{j=2}^n F_j)$ Compute I_k as an interpolant of $(I_{k-1} \land F_k, \bigwedge_{j=k+1}^n F_j)$
- Claim: I_k is an interpolant for $(\bigwedge_{i=1}^k F_i, \bigwedge_{j=k+1}^n F_j)$ Proof (sketch):
 - By ind.hyp. I_{k-1} is an interpolant for $(\bigwedge_{i=1}^{k-1} F_i, \bigwedge_{j=k}^n F_j)$ so $\bigwedge_{i=1}^{k-1} F_i \models I_{k-1}$ and $I_{k-1} \wedge F_k \wedge \bigwedge_{j=k+1}^n F_j \models \bot$
- Advantages:
 - simple to implement
 - can use any off-the-shelf binary interpolation
- Drawback: requires n-1 SMT calls



- Compute an SMT proof of unsatisfiablity *P* for $\bigwedge_{i=1}^{n} F_i$
- Compute each $I_k := \text{Interpolant}(\bigwedge_{i=1}^k F_i, \bigwedge_{j=k+1}^n F_j)$ from the same proof *P*
- Theorem: $F_k \wedge I_{k-1} \models I_k$



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- Compute each $I_k := \text{Interpolant}(\bigwedge_{i=1}^k F_i, \bigwedge_{j=k+1}^n F_j)$ from the same proof *P*
- Theorem: $F_k \wedge I_{k-1} \models I_k$
- Proof (sketch) case n=3:
 - Let *C* be a node of *P* with partial interpolants *I*' and *I*'' for the partitionings $(F_1, F_2 \land F_3)$ and $(F_1 \land F_2, F_3)$ resp. Then we can prove, by induction on the structure of *P*, that:

$$I' \wedge F_2 \models I'' \vee \bigvee \{l \in C \mid \operatorname{var}(l) \notin F_3\}$$

- The theorem then follows as a corollary
- Works also for DTC-rewritten proofs



DISCLAIMER: this is **very** incomplete. Apologies to missing authors/works

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Thank You