

Use of SMT Solvers in Verification

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Overview

Part 1

Use of Craig Interpolants in Fault Localization

Part 2

Computing Craig Interpolants

Part 1

Use of Craig Interpolants in Fault Localization:

Error Invariants [FM'12]

joint work with

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Faulty Shell Sort

Program

- takes a sequence of integers as input
- returns the sorted sequence.

On the input sequence 11, 14 the program returns 0, 11 instead of 11,14.

```
1  #include <stdio.h>
2  #include <stdlib.h>
3
4  static void shell_sort(int a[], int size)
5  {
6      int i, j;
7      int h = 1;
8      do {
9          h = h * 3 + 1;
10     } while (h <= size);
11     do {
12         h /= 3;
13         for (i = h; i < size; i++) {
14             int v = a[i];
15             for (j = i; j >= h && a[j - h] > v; j -= h)
16                 a[j] = a[j-h];
17             if (i != i)
18                 a[j] = v;
19         }
20     } while (h != 1);
21 }
22
23 int main(int argc, char *argv[])
24 {
25     int i = 0;
26     int *a = NULL;
27
28     a = (int *)malloc((argc-1) * sizeof(int));
29     for (i = 0; i < argc - 1; i++)
30         a[i] = atoi(argv[i + 1]);
31
32     shell_sort(a, argc);
33
34     for (i = 0; i < argc - 1; i++)
35         printf("%d", a[i]);
36     printf("\n");
37
38     free(a);
39     return 0;
40 }
```

Error Trace

```
0 int i,j, a[];
1 int size=3;
2 int h=1;
3 h = h*3+1;
4 assume !(h<=size);
5 h/=3;
6 i=h;
7 assume (i<size);
8 v=a[i];
9 j=i;
10 assume !(j>=h && a[j-h]>v);
11 i++;
12 assume (i<size);
13 v=a[i];
```

```
14 j=i;
15 assume (j>=h && a[j-h]>v);
16 a[j]=a[j-h];
17 j-=h;
18 assume (j>=h && a[j-h]>v);
19 a[j]=a[j-h];
20 j-=h;
21 assume !(j>=h && a[j-h]>v);
22 assume (i!=j);
23 a[j]=v;
24 i++;
25 assume !(i<size);
26 assume (h==1);
27 assert a[0] == 11 && a[1] == 14;
```

Error Trace

Input Values

Control-Flow Path

Expected Outcome

Can be obtained, e.g.,

- from a static analysis tool;
- from a failing test case;
- during debugging.

The Fault Localization Problem

Error traces

- can become very long (thousands of statements);
- contains many statements and program variables that are irrelevant for understanding the error;
- provide no explicit information about the program state.

Fault Localization:

- Identify the relevant statements and variables.
- Provide an explanation for the error that incorporates the state of the program.

Error Trace

```
0 int i,j, a[];
1 int size=3;
2 int h=1;
3 h = h*3+1;
4 assume !(h<=size);
5 h/=3;
6 i=h;
7 assume (i<size);
8 v=a[i];
9 j=i;
10 assume !(j>=h && a[j-h]>v);
11 i++;
12 assume (i<size);
13 v=a[i];
```

```
14 j=i;
15 assume (j>=h && a[j-h]>v);
16 a[j]=a[j-h];
17 j-=h;
18 assume (j>=h && a[j-h]>v);
19 a[j]=a[j-h];
20 j-=h;
21 assume !(j>=h && a[j-h]>v);
22 assume (i!=j);
23 a[j]=v;
24 i++;
25 assume !(i<size);
26 assume (h==1);
27 assert a[0] == 11 && a[1] == 14;
```


Error Invariants

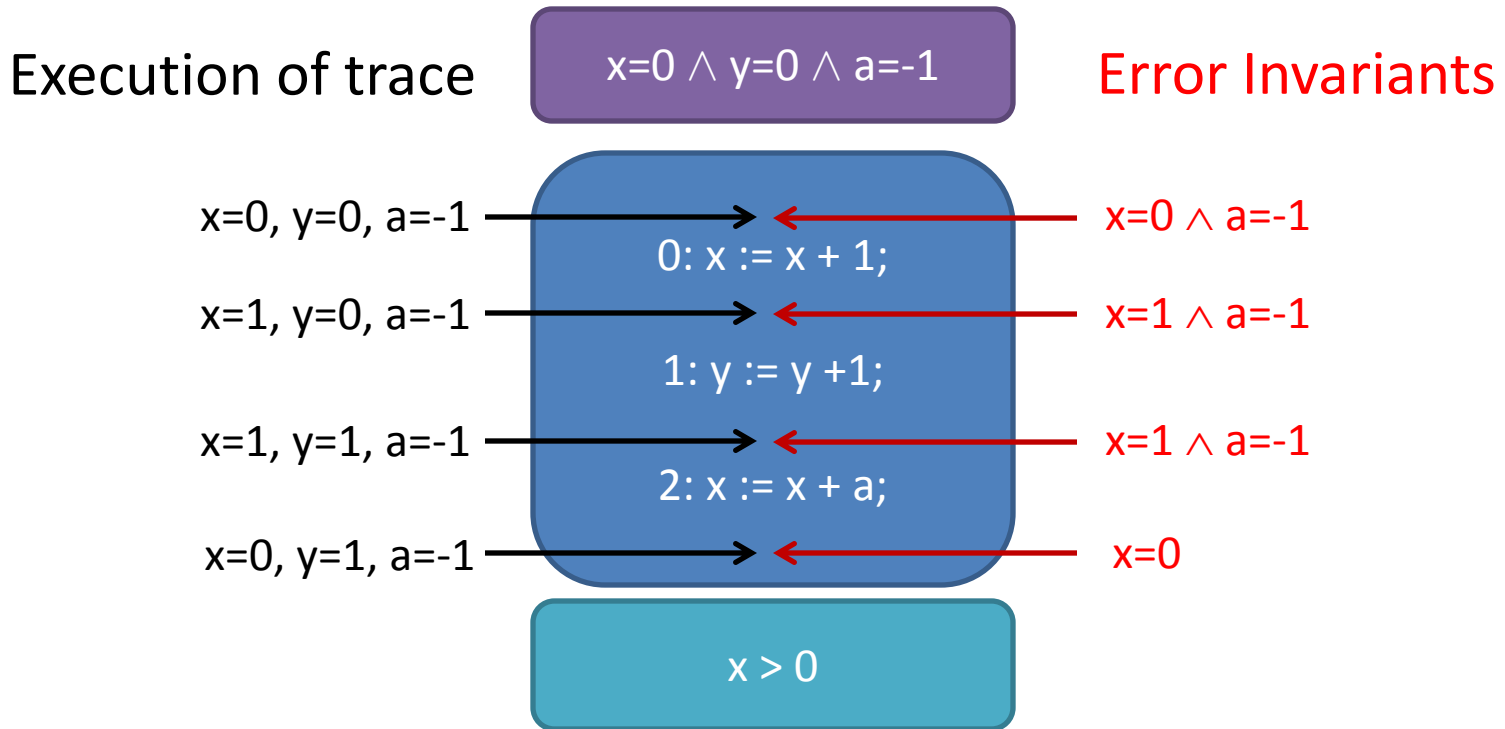
An **error invariant** I for a position i in an error trace τ is a formula over program variables s.t.

- all states reachable by executing the prefix of τ up to position i satisfy I
- all executions of the suffix of τ that start from i in a state that satisfies I , still lead to the error.

I is called an **inductive error invariant** for positions $i < j$ if I is an error invariant for both i and j .

Error Invariants

Example



Information provided
by the error invariants

- Statement $y := y + 1$ is irrelevant
- Variable y is irrelevant
- Variable a is irrelevant after position 2

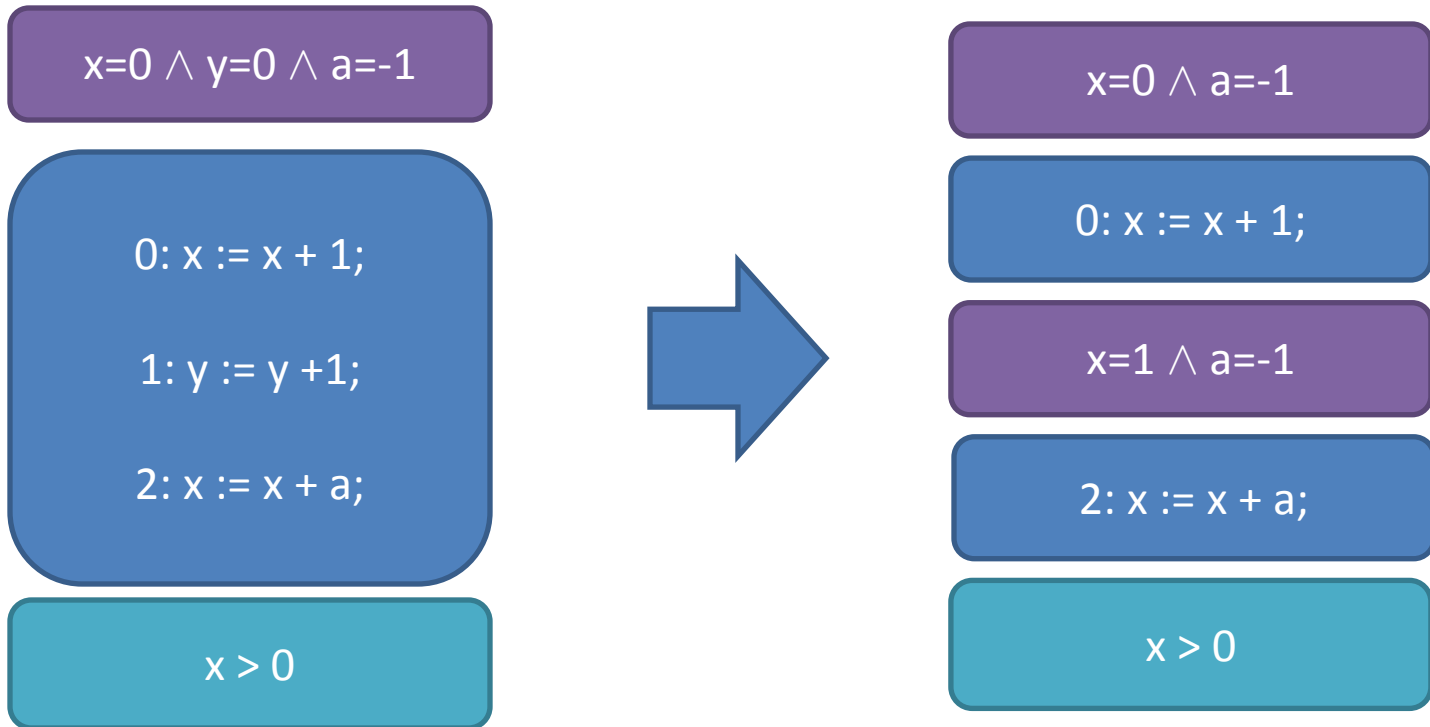
Abstract Error Trace

Abstract error trace consists only of

- relevant statements and
- error invariants that hold before and after these statements.

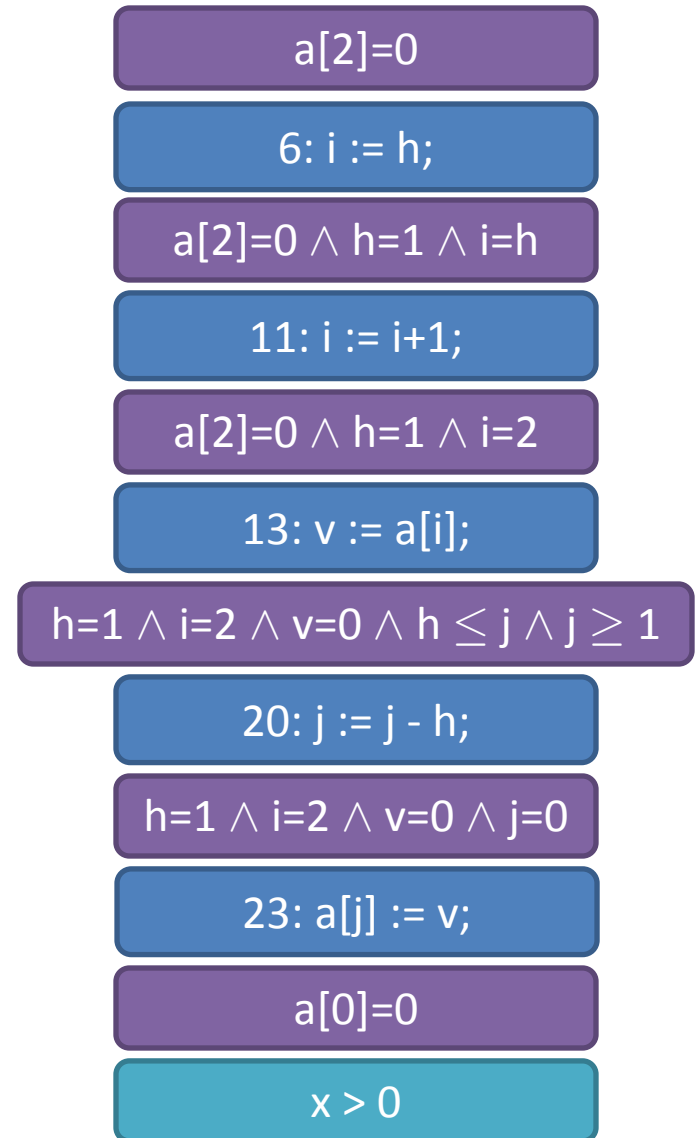
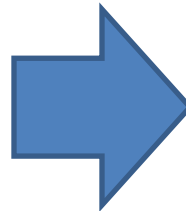
Abstract Error Trace

Example



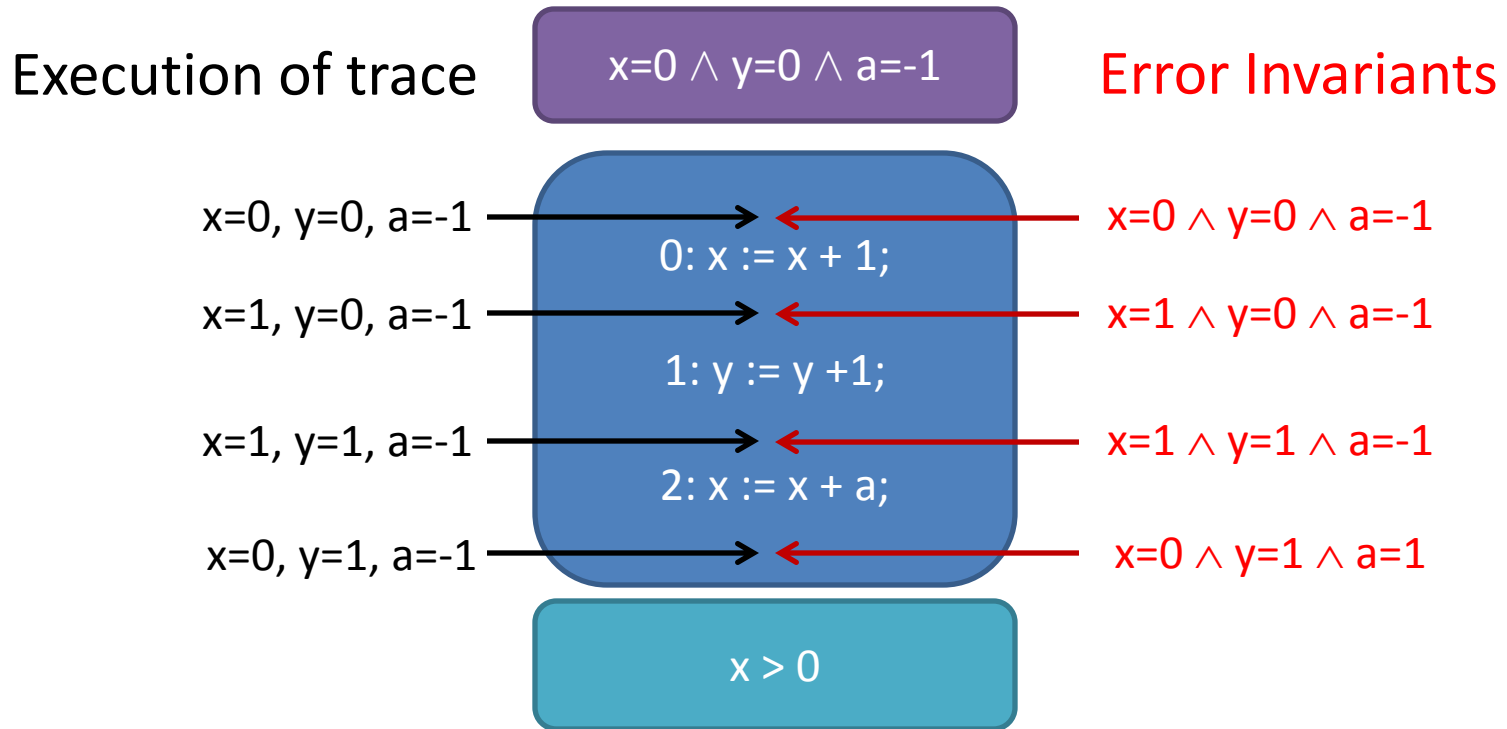
Abstract Error Trace for Faulty Shell Sort

```
0 int i,j, a[];
1 int size=3;
2 int h=1;
3 h = h*3+1;
4 assume !(h<=size);
5 h/=3;
6 i=h;
7 assume (i<size);
8 v=a[i];
9 j=i;
10 assume !(j>=h && a[j-h]>v);
11 i++;
12 assume (i<size);
13 v=a[i];
14 j=i;
15 assume (j>=h && a[j-h]>v);
16 a[j]=a[j-h];
17 j-=h;
18 assume (j>=h && a[j-h]>v);
19 a[j]=a[j-h];
20 j-=h;
21 assume !(j>=h && a[j-h]>v);
22 assume (i!=j);
23 a[j]=v;
24 i++;
25 assume !(i<size);
26 assume (h==1);
27 assert a[0] == 11 && a[1] == 14;
```



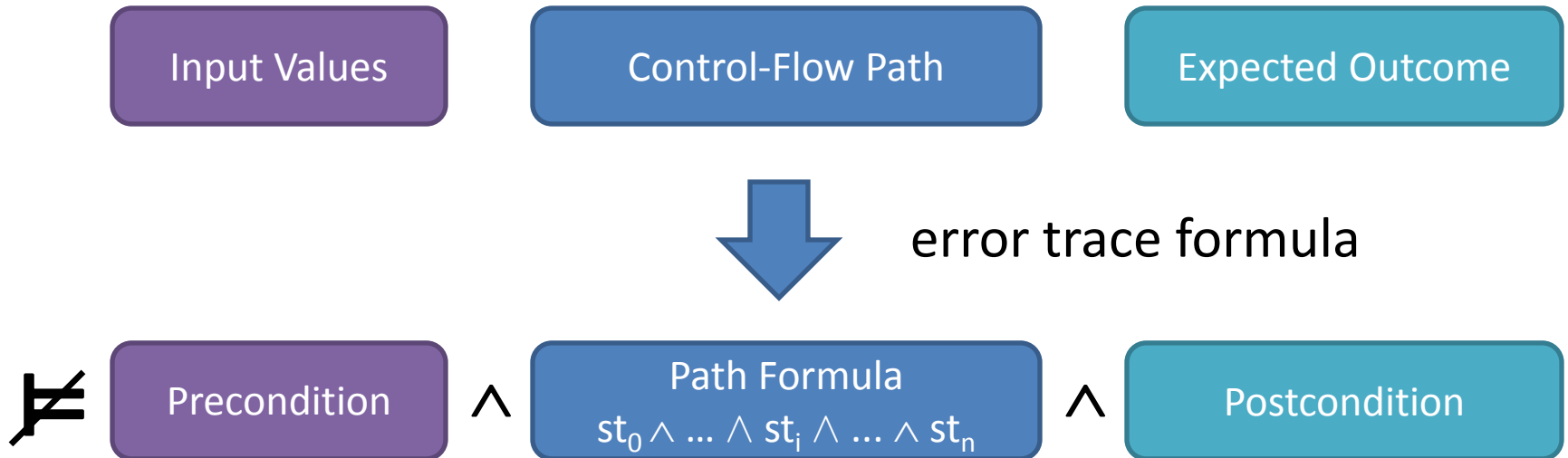
How can we compute error invariants?

Error invariants are not unique



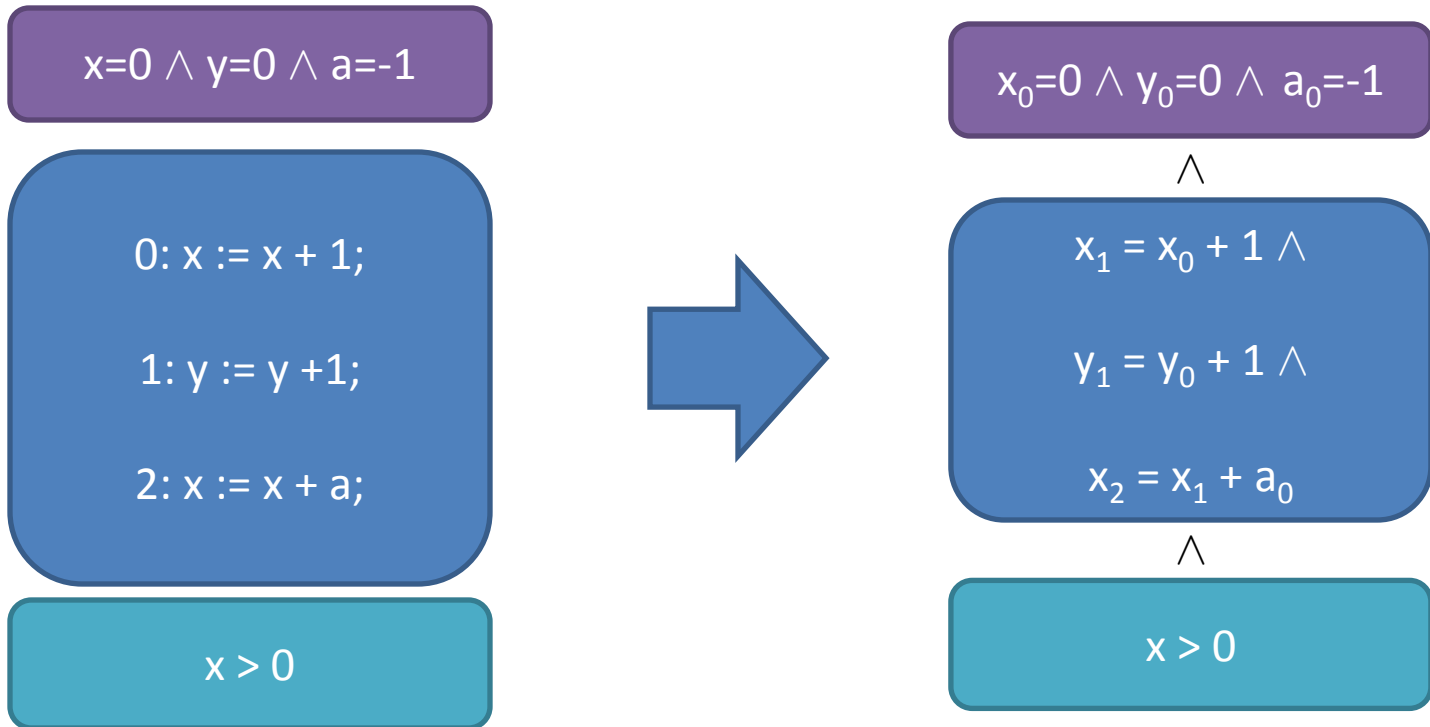
We are interested in **inductive** error invariants!

Checking Error Invariants

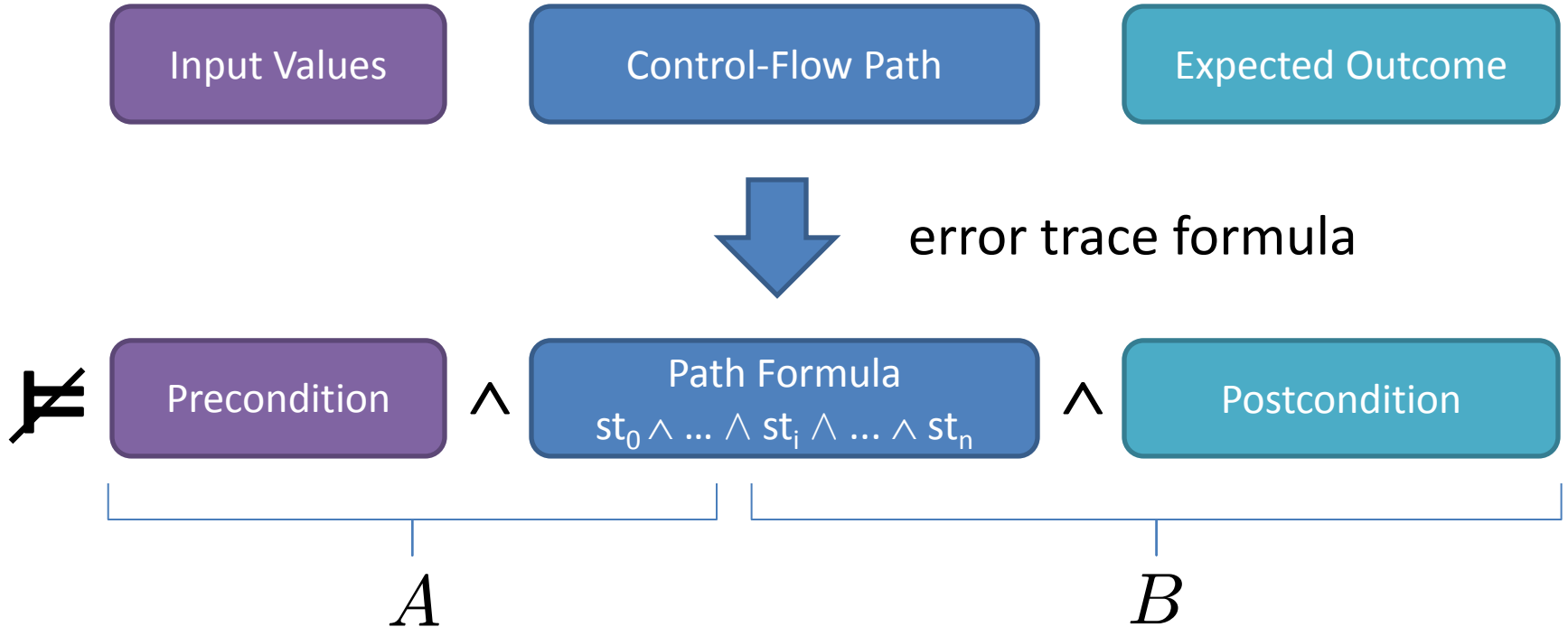


Error Trace Formula

Example



Checking Error Invariants



I is an error invariant for position i iff

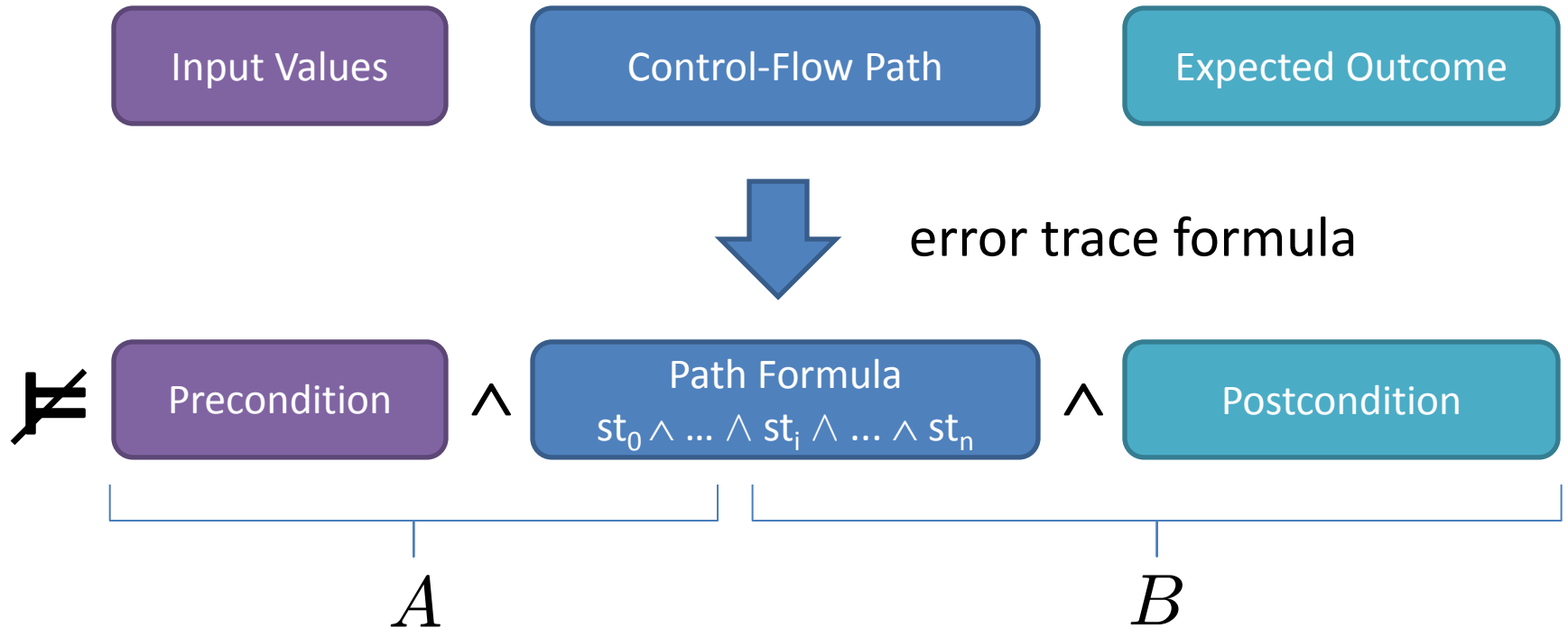
$$A \models I \quad \text{and} \quad I \wedge B \models \perp$$

Craig Interpolants

Given formulas A, B whose conjunction is unsatisfiable, a **Craig interpolant** for (A, B) is a formula I such that

- $A \models I$
- $I \wedge B \models \perp$
- $\text{fv}(I) \subseteq \text{fv}(A) \wedge \text{fv}(B)$

Craig Interpolants are Error Invariants



Craig interpolant for $A \wedge B$ is an error invariant for position i

\Rightarrow use Craig interpolation to compute candidates
for inductive error invariants.

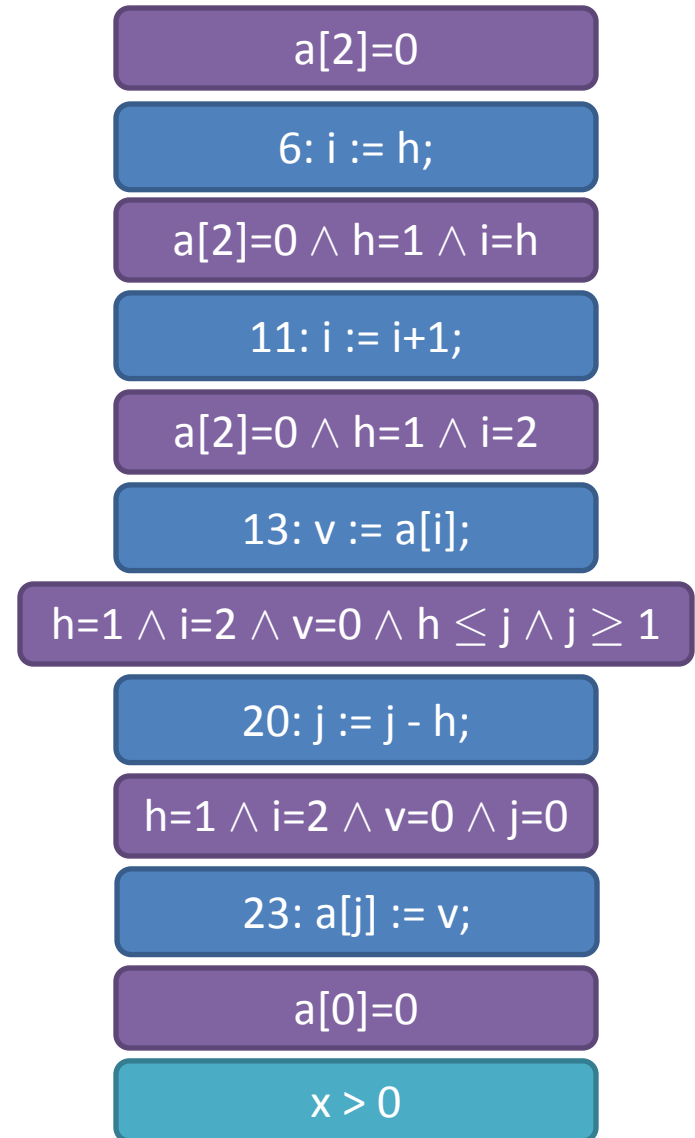
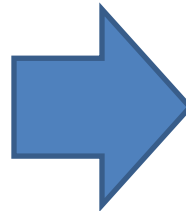
Computing Abstract Error Traces

Basic Algorithm:

1. Compute the error trace formula from the error trace.
2. Compute a Craig interpolant I_i for each position i in the error trace.
3. Compute the error invariant matrix:
 - for each I_i and j , check whether I_i is an error invariant for j .
4. Choose minimal covering of error trace with inductive error invariants.
5. Output abstract error trace.

Abstract Error Trace for Faulty Shell Sort

```
0 int i,j, a[];
1 int size=3;
2 int h=1;
3 h = h*3+1;
4 assume !(h<=size);
5 h/=3;
6 i=h;
7 assume (i<size);
8 v=a[i];
9 j=i;
10 assume !(j>=h && a[j-h]>v);
11 i++;
12 assume (i<size);
13 v=a[i];
14 j=i;
15 assume (j>=h && a[j-h]>v);
16 a[j]=a[j-h];
17 j-=h;
18 assume (j>=h && a[j-h]>v);
19 a[j]=a[j-h];
20 j-=h;
21 assume !(j>=h && a[j-h]>v);
22 assume (i!=j);
23 a[j]=v;
24 i++;
25 assume !(i<size);
26 assume (h==1);
27 assert a[0] == 11 && a[1] == 14;
```



Some Related Approaches

- Bug-Assist [Jose, Majumdar, '11]
- Whodunit? [Wang, Yang, Ivancic, Gupta, '06]
- Delta debugging [Cleve, Zeller, '05]
- Distance metrics [Groce, Kroening, Lerda, '04]

Summary (Part 1)

Error invariants:

- new approach to fault localization
- enables computation of concise error explanations
- underlying work horse: Craig interpolation

Part 2

Computing Craig Interpolants: Hierarchical Interpolation Procedures

joint work with Nishant Totla (IIT Bombay, India)

Computing Craig Interpolants

Craig interpolants

- have many applications in Formal Methods
- can be automatically computed from proofs of unsatisfiability
- typically interested in ground interpolants
- many standard theories of SMT solvers admit ground interpolation
 - linear arithmetic
 - free function symbols with equality
 - ... (more tomorrow)

Challenges in Interpolation

- Formulas generated in program verification are often in theories that are not directly supported by SMT solvers.
- Instead these theories are encoded using axioms and instantiation heuristics.
- Sometimes these heuristics are complete: (hierarchical decision procedure, SMT modulo theories).

How can we compute ground interpolants for such theory extensions (hierarchical interpolation)?

Challenges in Interpolation

- We are not interested in arbitrary interpolants but only in inductive ones.
- Different proofs of unsatisfiability produce different interpolants.
- Finding *good* interpolants requires a lot of heuristics in the interpolation procedure.
- This is considered a black art.

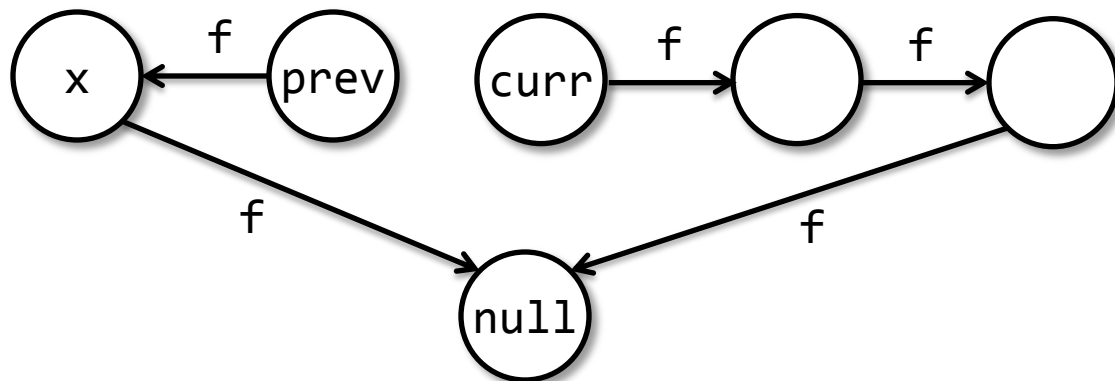
How can we decouple these heuristics from the actual interpolation procedure?

Example: List Reversal

```
assume x  $\xrightarrow{f}$  null;  
prev := null;  
curr := x;  
while curr  $\neq$  null do  
  succ := curr.f;  
  curr.f := prev;  
  prev := curr;  
  curr := succ;  
end  
x := prev;  
assert x  $\xrightarrow{f}$  null;
```

Safe inductive invariant:

$prev \xrightarrow{f} null \wedge$
 $curr \xrightarrow{f} null \wedge$
 $disjoint(f, curr, prev)$



Theory of **L**inked **L**ists with **R**eachability

(Variation of [Lahiri, Qadeer, POPL'08])

- $x \xrightarrow{f/u} y$ constrained reachability
- $x.f$ field access: `select(f,x)`
- $f [x := y]$ field update: `update(f, x, y)`

$x \xrightarrow{f/u} y$ means y is reachable from x via f without going through u (but $y = u$ is allowed)

$x \xrightarrow{f} y$ stands for $x \xrightarrow{f/y} y$

Axioms of TLLR

- Refl: $x \xrightarrow{f/u} x$
- Step: $x \xrightarrow{f/u} x.f \vee x=u$
- Linear: $x \xrightarrow{f} y \Rightarrow x \xrightarrow{f/y} u \vee x \xrightarrow{f/u} y$
- ...
- ReadWrite1: $x.(f[x := y]) = y$
- ReadWrite2: $x = y \vee y.(f[x := z]) = y.f$
- ReachWrite: $x \xrightarrow{f[u := v]/w} y \Leftrightarrow$

$$x \xrightarrow{f/w} y \wedge x \xrightarrow{f/u} y \vee$$

$$x \xrightarrow{f/w} u \wedge v \xrightarrow{f/u} y \wedge v \xrightarrow{f/w} y \wedge u \neq w$$

Example Proof

$$a \xrightarrow{f} b \wedge b.f = c \wedge \neg a \xrightarrow{f} c$$

1. $a \xrightarrow{f} b$
2. $b.f = c$
3. $\neg a \xrightarrow{f} c$
4. $b \xrightarrow{f} b.f$ (Instantiation of Step)
5. $a \xrightarrow{f} b \wedge b \xrightarrow{f} c \Rightarrow a \xrightarrow{f} c$ (Inst. of Trans)
6. $b \xrightarrow{f} c$ (Rewriting of 4 with 2)
7. \perp (Resolution of 5 with 1,2,6)

Local Theory Extensions

[Sofronie-Stokkermans '05]

Given

- a first-order signature Σ_0
- a (universal) first-order theory T_0 over Σ_0 (**base theory**)
- an extended signature $\Sigma_1 = \Sigma_0 \cup \{f_1, \dots, f_n\}$
- a (universal) **theory extension** $T_1 = T_0 \cup K$

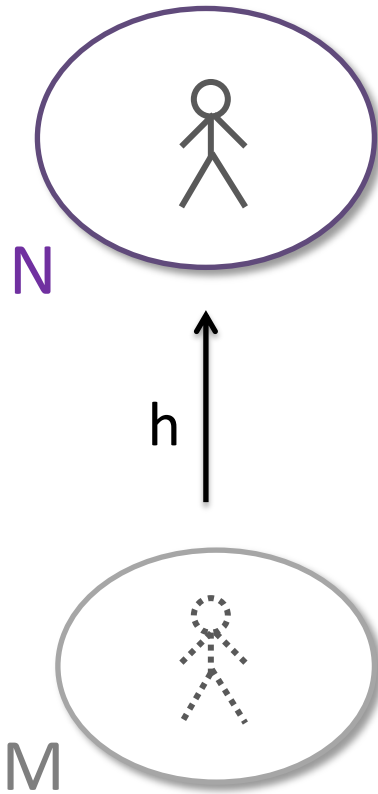
T_1 is called **local** if for all ground Σ_1 -formulas G

$$T_1 \wedge G \models \perp \text{ iff } K[G] \wedge T_0 \wedge G \models \perp$$

Local extensions of decidable base theories are decidable.

Detecting Locality

[Sofronie-Stokkermans '05]



M: weak partial model of the theory

N: total model of the theory

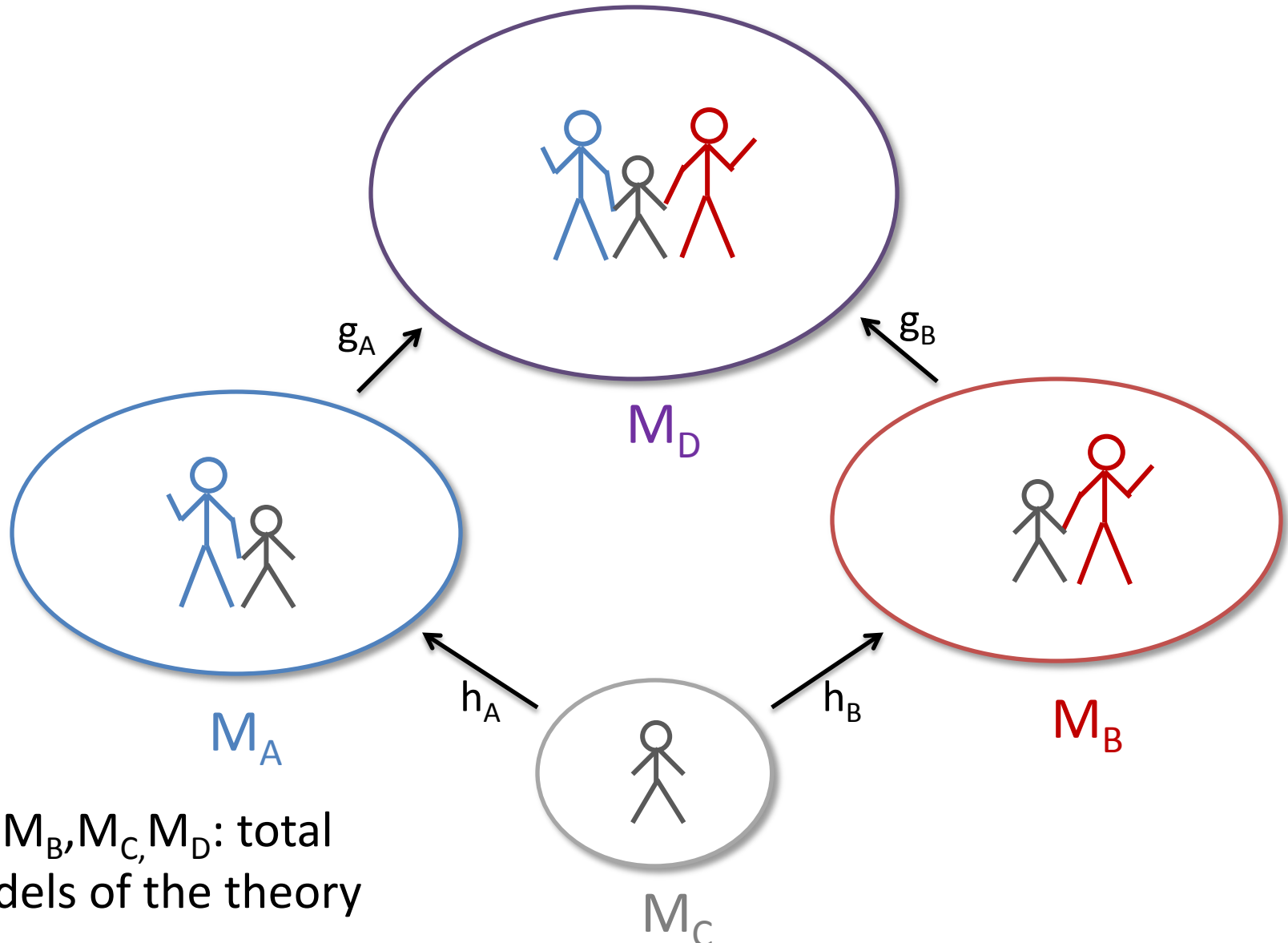
h: homomorphic embedding of M into N

Idea of Hierarchical Interpolation

Reduce interpolation problem $A \wedge B$ in theory extension $T_1 = K \cup T_0$ to interpolation problem in the base theory T_0 :

- In order to find an T_1 -interpolant I for $A \wedge B$
- find a T_0 -interpolant I for $A_0 \wedge B_0$ where $A_0 = K[A] \wedge A$ and $B_0 = K[B] \wedge B$
- This is complete whenever for all $A \wedge B$
 $T_1 \wedge A \wedge B \models \perp$ iff $T_0 \wedge A_0 \wedge B_0 \models \perp$

Amalgamation Property

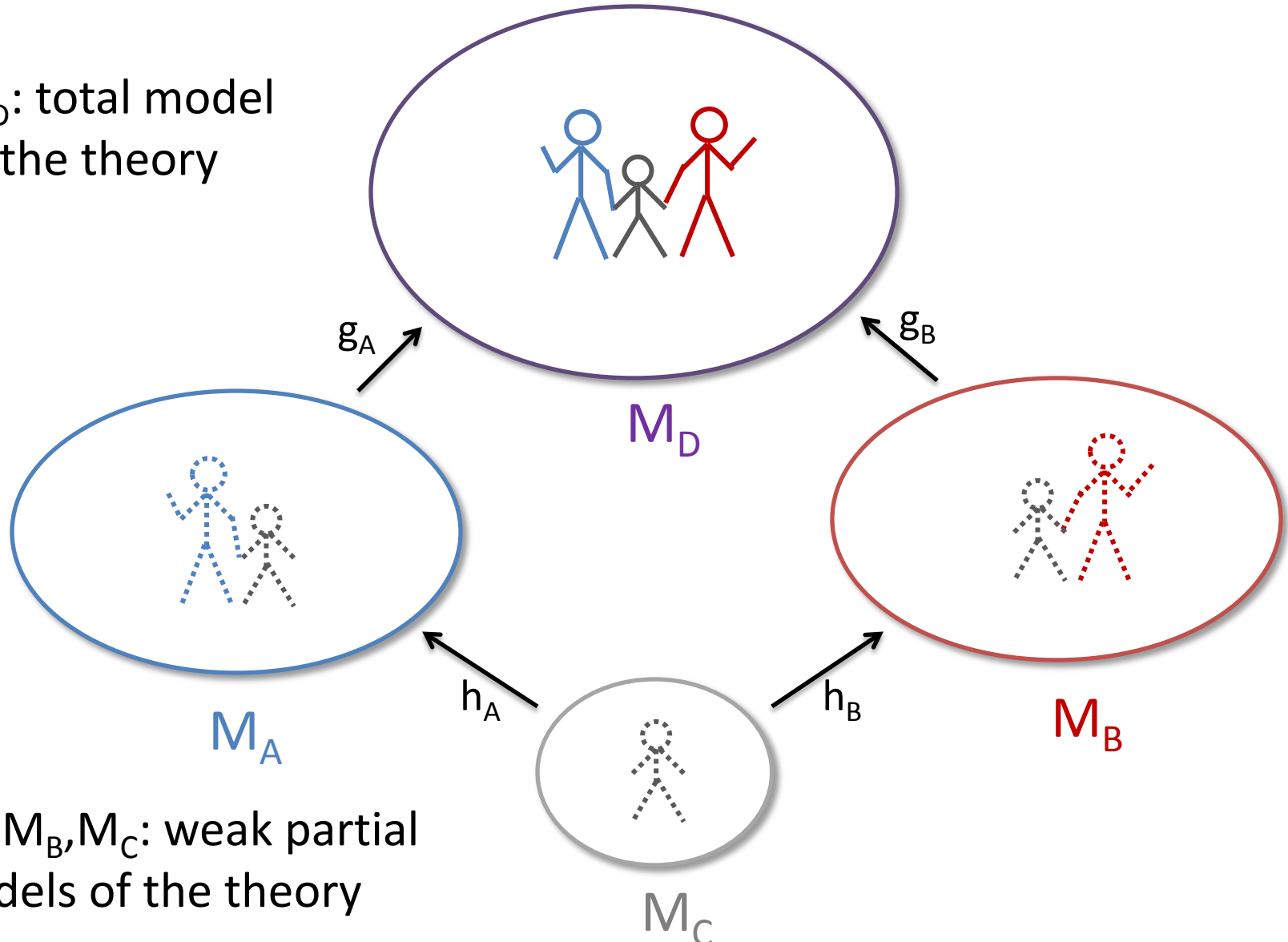


Amalgamation Property

- A theory T admits ground interpolation iff T has the amalgamation property [Bacsich '75]
- Amalgamation does not tell us how to compute ground interpolants.
- Also, this property is too weak for our purposes.

Weak Amalgamation Property

M_D : total model
of the theory



M_A, M_B, M_C : weak partial
models of the theory

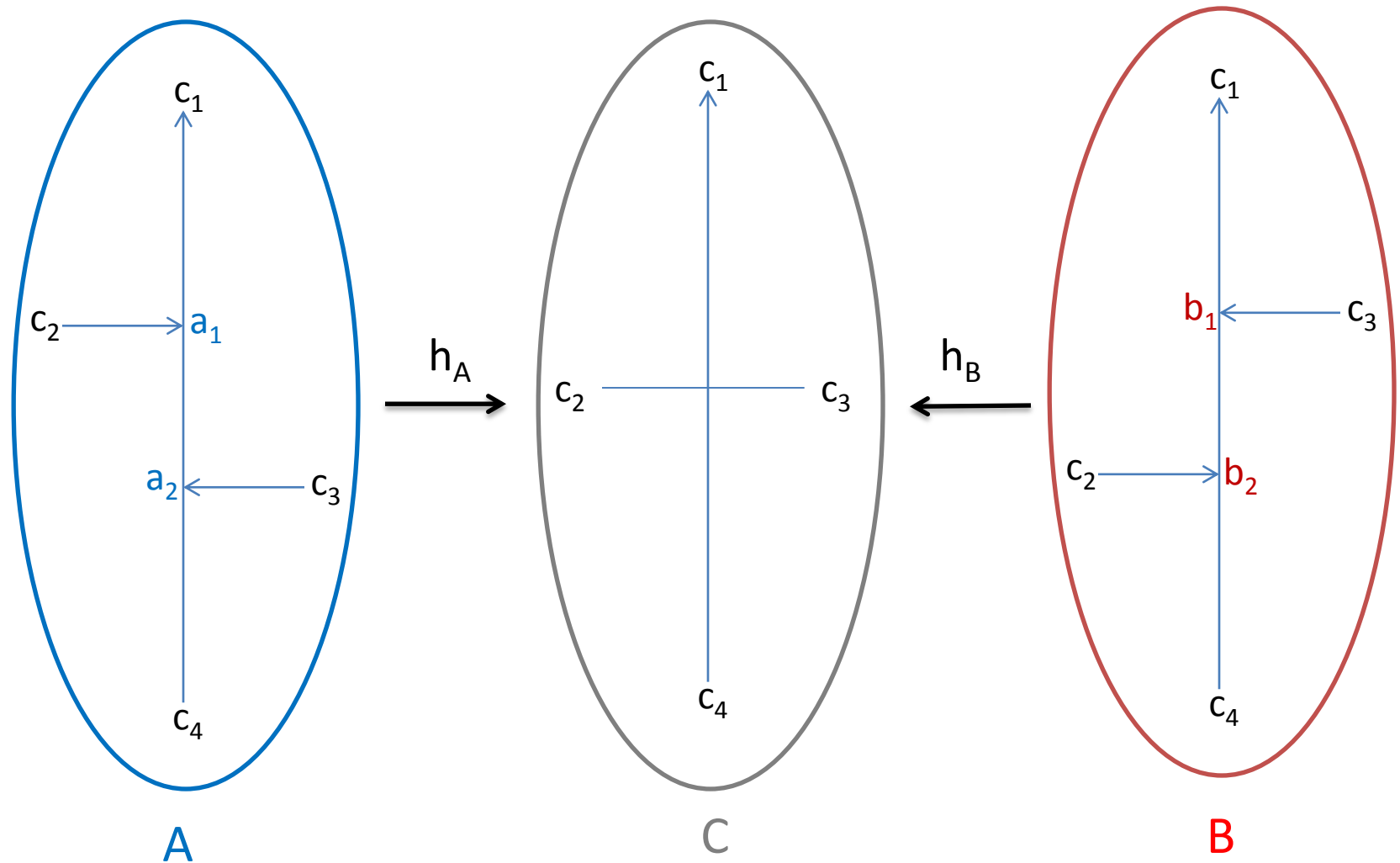
Hierarchical Interpolation via Weak Amalgamation

Main Result:

If $T_1 = T_0 \cup K$ has the weak amalgamation property and T_0 admits effective ground interpolation, then T_1 admits effective ground interpolation.

Generic technique to obtain new interpolation procedures from existing ones.

TLLR does not have weak amalgamation



Making TLLR complete

Add two additional functions:

- $\text{join}(f,x,y)$ node where f-paths from x and y join (if they do so)
- $\text{diff}(f,g)$ node on which fields f and g differ if $f \neq g$

$$x \xrightarrow{f} z \wedge y \xrightarrow{f} z \Rightarrow \text{join}(f,x,y) \xrightarrow{f} z$$

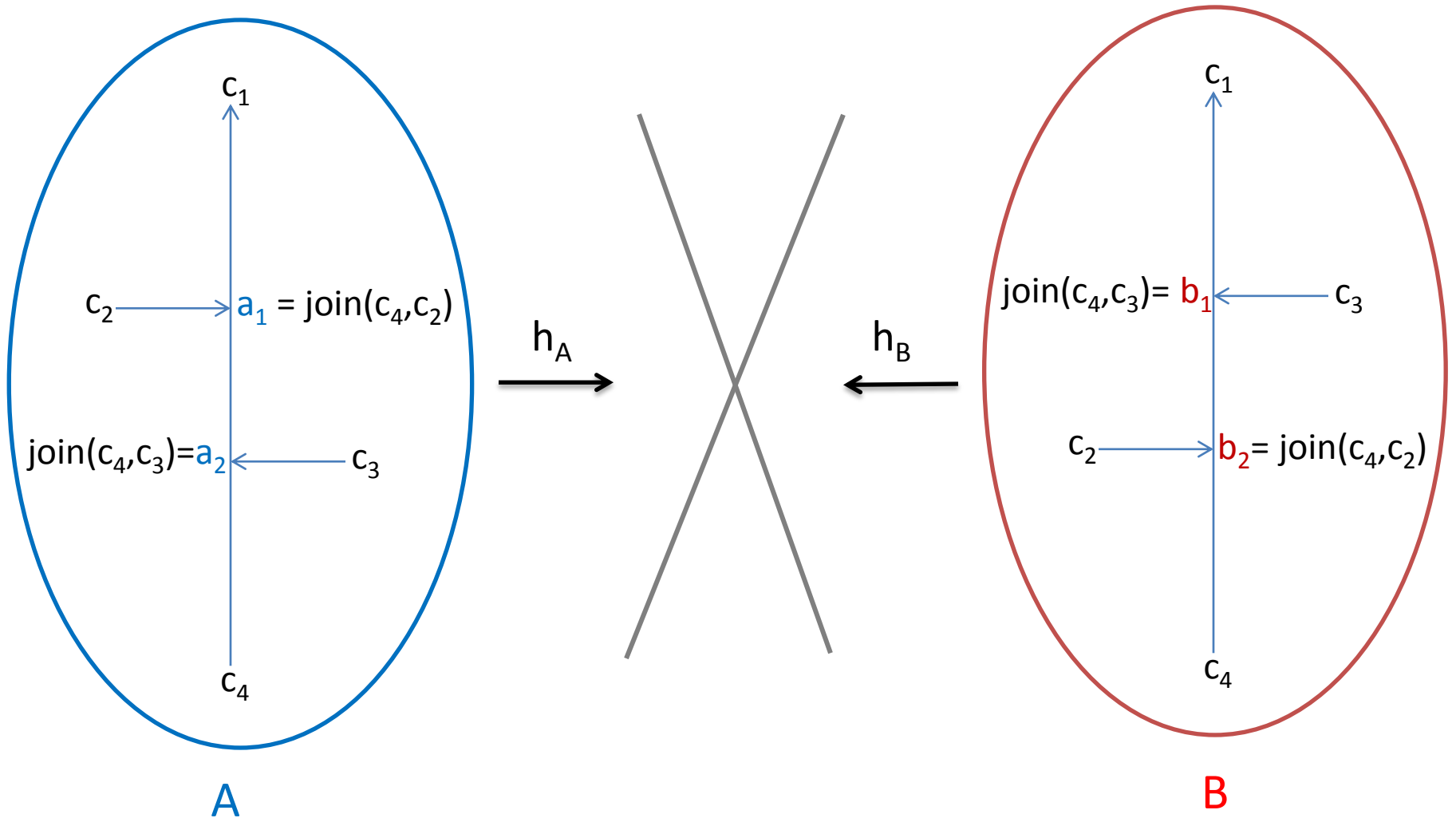
$$x \xrightarrow{f} z \wedge y \xrightarrow{f} z \Rightarrow x \xrightarrow{f} \text{join}(f,x,y)$$

$$x \xrightarrow{f} z \wedge y \xrightarrow{f} z \Rightarrow y \xrightarrow{f} \text{join}(f,x,y)$$

$$\text{diff}(f,g).f = \text{diff}(f,g).g \Rightarrow f=g$$

$$\text{disjoint}(f,x,y) \equiv x \xrightarrow{f} \text{join}(f,x,y) \wedge y \xrightarrow{f} \text{join}(f,x,y) \Rightarrow \text{join}(f,x,y)=\text{null}$$

TLLR + join/diff has weak amalgamation



Example: List Reversal

```
assume x  $\xrightarrow{f}$  null;  
prev := null;  
curr := x;  
while curr  $\neq$  null do  
  succ := curr.f;  
  curr.f := prev;  
  prev := curr;  
  curr := succ;  
end  
x := prev;  
assert x  $\xrightarrow{f}$  null;
```

Safe inductive invariant:

$\text{prev} \xrightarrow{f} \text{null} \wedge$
 $\text{curr} \xrightarrow{f} \text{null} \wedge$
 $\text{disjoint}(f, \text{curr}, \text{prev})$

Computed interpolant for
2 loop unrollings:

$(\text{curr} = \text{null} \wedge \text{prev.f.f} = \text{null}) \vee$
 $(\text{curr} \neq \text{null} \wedge \text{prev.f} \neq \text{curr} \wedge$
 $\text{prev.f} \neq \text{curr.f} \wedge \text{curr.f} \neq \text{prev} \wedge$
 $\text{prev.f.f} = \text{null})$

Enumerating Partial Models

Given: theory extension T with weak amalgamation.

Input: A, B : ground formulas with $T \wedge A \wedge B \models \perp$

Output: I : T_1 -interpolant for (A, B)

$I := \perp$

while \exists partial model M of $T \wedge A \wedge \neg I$ **do**

$I := I \vee \text{interpolate}(T, M, B)$

end

return I

Combining Interpolation and Abstraction

Given: theory extension T with weak amalgamation.

Input: A, B : ground formulas with $T \wedge A \wedge B \models \perp$

Output: I : T_1 -interpolant for (A, B)

$I := \perp$

while \exists partial model M of $T_0 \cup K[A] \wedge A \wedge \neg I$ **do**

if $T \wedge \alpha(M) \wedge B \models \perp$ **then**

 // $M \models \alpha(M)$

$I := I \vee \text{interpolate}(T, \alpha(M), B)$

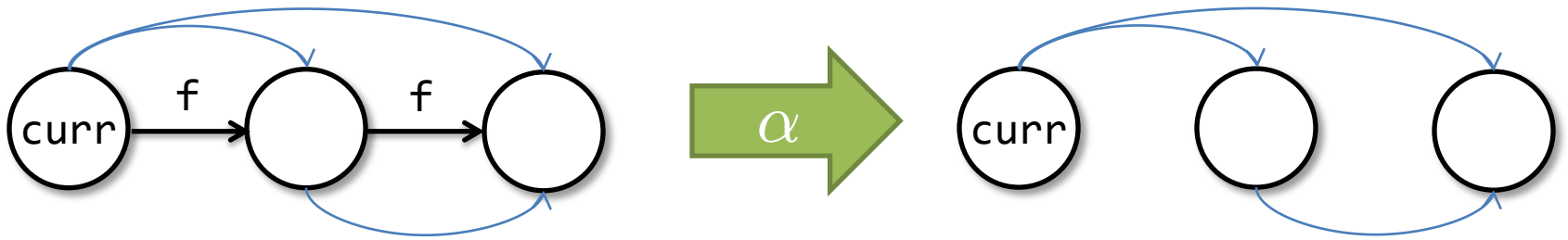
else

$I := I \vee \text{interpolate}(T, M, B)$

end

return I

List Abstraction



Abstract from the length of the list.

Example: List Reversal

```
assume x  $\xrightarrow{f}$  null;  
prev := null;  
curr := x;  
while curr  $\neq$  null do  
  succ := curr.f;  
  curr.f := prev;  
  prev := curr;  
  curr := succ;  
end  
x := prev;  
assert x  $\xrightarrow{f}$  null;
```

Safe inductive invariant:

prev \xrightarrow{f} null \wedge
curr \xrightarrow{f} null \wedge
disjoint(f, curr, prev)

Computed interpolant for
2 loop unrollings:

prev $\xrightarrow{f/curr}$ null \wedge
join(f, prev, curr) = null

Related Work

- Sofronie-Stokkermans '06: Interpolation in local theory extensions
- Rybalchenko, Sofronie-Stokkermans '07: Constraint Solving for Interpolation
- Bruttomesso, Ghilardi, Ranise '12: Strong amalgamation for interpolation in theory combinations
- Bruttomesso, Ghilardi, Ranise '11: Ground interpolation for arrays
- McMillan '08: Quantified interpolation

Summary (Part 2)

- A new generic technique to obtain new interpolation procedures from existing ones
 - depends only on a model theoretic notion (weak amalgamation)
 - interpolation procedure for base theory can be treated as a black box
 - allows easy implementation of domain-specific heuristics
- Many theories of practical interest have weak amalgamation
 - arrays with extensionality
 - linked lists with reachability
 - imperative trees with reachability [CADE'11]
 - ...