

# Spatial Interpolants

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# Problem

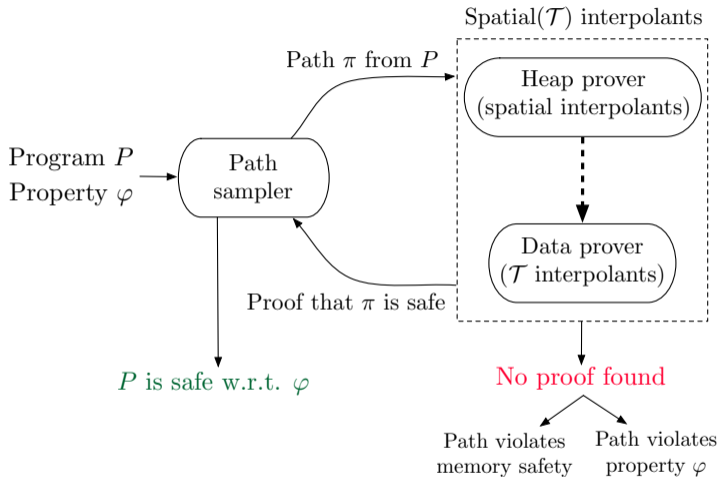
Combined heap and data reasoning for automatic verification

Examples:

- ▶ scalar constraints on heap-resident data
- ▶ traversing linked structures by size
- ▶ storing array indices in linked data-structures
- ▶ manual reference counting

```
1: int i = nondet();
   node* x = null;
2: while (i != 0)
   node* tmp = malloc(node);
   tmp->N = x;
   tmp->D = i;
   x = tmp;
   i--;
3: while (x != null)
4:   assert(x->D >= 0);
   x = x->N;
```

## SPLINTER from 10,000 feet



- ▶ No heap: specializes to IMPACT (McMillan's *lazy abstraction with interpolants*)
- ▶ No data: specializes to new path-based separation logic analysis

# Motivation for Path Sampling

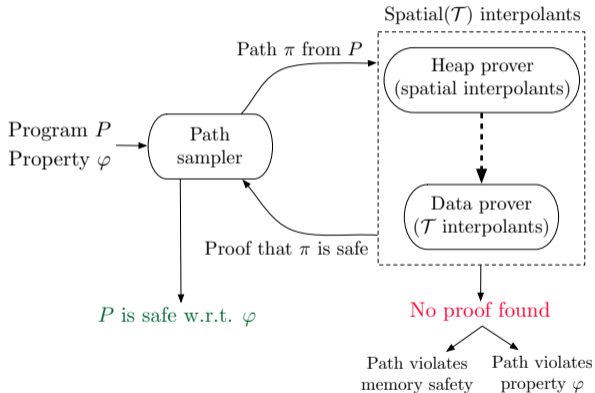
*Path sampling* enables

- ▶ *Path-based refinement*
  - ▶ *progress* guarantee by tightly correlating program exploration with refinement
  - ▶ *precision* guarantee by avoiding lossy join and widening operations
  - ▶ produces *counter-examples* for violated properties
  - ▶ no false alarms (diverges instead, as usual)
- ▶ *Property-direction*
  - ▶ don't try to compute strongest invariant possible
  - ▶ compute one *just strong enough* to prove property holds
  - ▶ key enabler for scalable precise reasoning in “rich” program logics

Main impediment

- ▶ (*infinitely-*) *many* paths may be analyzed before finding proof

# Path Sampling

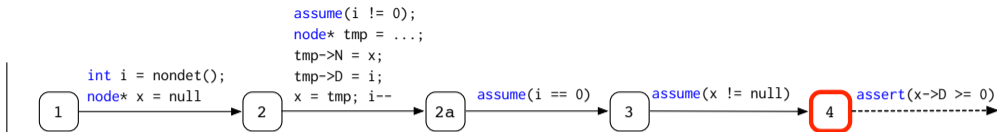


► Follows IMPACT

► Optimizations exist, *but basically*:

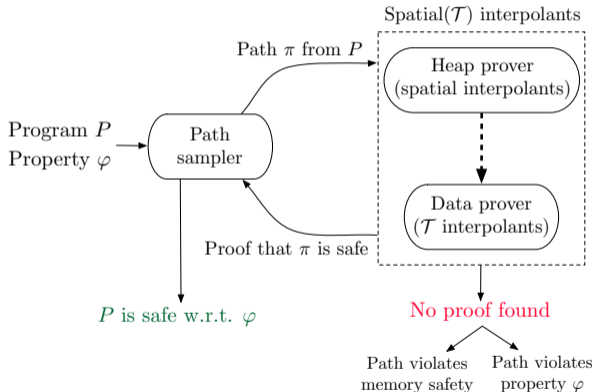
- Maintain set of paths and their proofs
- At each step, choose an arbitrary path
  - finite path through control-flow graph
  - from program entry to an assertion
  - not already proved

# Path Sampling: Example



```
1: int i = nondet();
   node* x = null;
2: while (i != 0)
   node* tmp = malloc(node);
   tmp->N = x;
   tmp->D = i;
   x = tmp;
   i--;
3: while (x != null)
4:  assert(x->D >= 0);
   x = x->N;
```

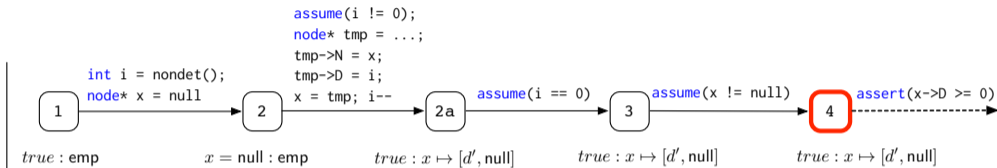
# Spatial Interpolation



- ▶ Construct Hoare-style memory safety proof for path
- ▶ Call annotations *spatial path interpolants*
  - ▶ logical strength *between* strongest postconditions and weakest preconditions
  - ▶ *do not* impose other conditions of Craig interpolants
- ▶ Two-phase computation
  1. symbolically execute path *forward* to compute strongest data-free postconditions
  2. relax proof via *backward* under-approximation of weakest preconditions
    - ▶ heuristic
    - ▶ guided by strongest postconditions along path

# Strongest Postconditions: Example

Symbolic Heaps  
(strongest post)

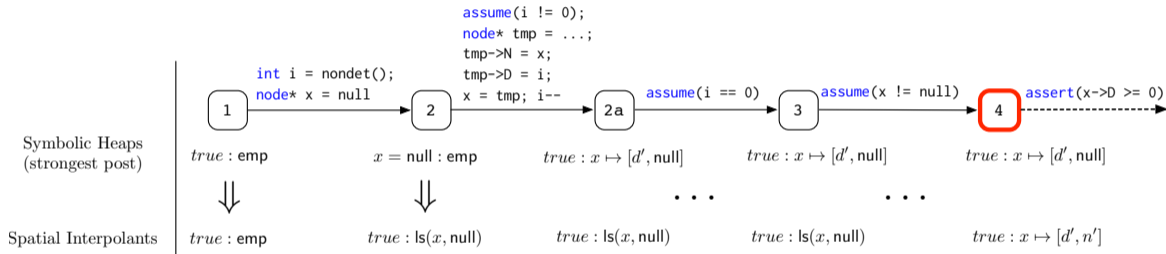


$$\text{exec}(x \rightarrow N_i := E, (\exists X. \Pi : \Sigma * z \mapsto [\vec{d}, \vec{n}])) = (\exists X. \Pi : \Sigma * x \mapsto [\vec{d}, \vec{n}[E/n_i]])$$

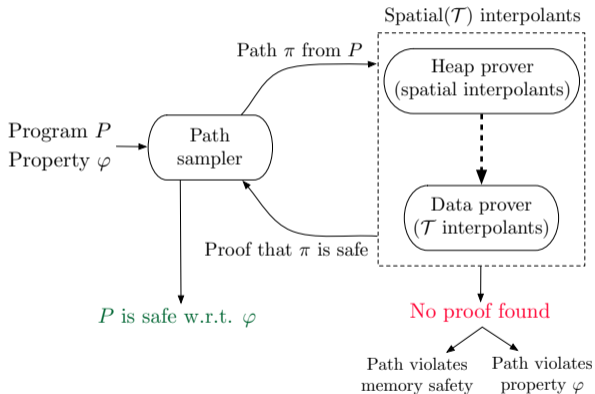
where  $i \leq |\vec{n}|$  and  $\Pi : \Sigma * z \mapsto [\vec{d}, \vec{n}] \vdash x = z$



# Spatial Interpolation: Example

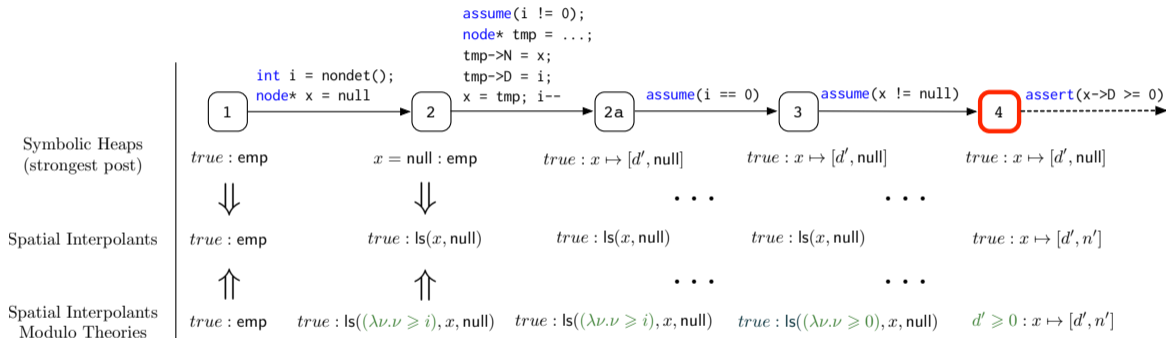


# Spatial Interpolation Modulo Theories



- ▶ Strengthen memory safety proof of path
  - ▶ add data constraints
  - ▶ prove path satisfies safety property
- ▶ Generate system of Horn clause constraints
  - ▶ encode data manipulation along path, and its memory safety proof
  - ▶ solve using existing techniques
  - ▶ solution determines *refinement* (strengthening) of memory safety proof

# Spatial Interpolation Modulo Theories: Example



# Spatial Interpolants

- ▶ Bounded from *below* by strongest memory safety proof
- ▶ Bounded from *above* (implicitly) by weakest memory safety proof
- ▶ Without *upper* bound
  - ▶ Interpolant/invariant computable using forward transformer and widening
  - ▶ Risks widening too aggressively
    - ▶ so analyses widen conservatively at the price of computing unnecessarily strong proofs
  - ▶ Upper bound captures information needed to prove future execution
- ▶ Without *lower* bound
  - ▶ Interpolant/invariant computable using backward transformer (and lower widening)
  - ▶ Backward transformers in shape analysis explode
    - ▶ due to issues such as not knowing the aliasing relationship in the pre-state
  - ▶ Lower bound captures such information, containing the explosion
- ▶ Price of both bounds is operating over *full paths* from entry to error
- ▶ Heuristics for weakening at each point along the path have information about
  - ▶ *one* execution's *past and future* when analyzing full paths
  - ▶ *many past* executions in a forwards iterative analysis via join or widening

# Bounded Abduction

## Definition (Bounded abduction)

A solution to the *bounded abduction problem*  $L \vdash (\exists X. M * [ ]) \vdash R$

is a formula  $A$  such that

$$L \models (\exists X. M * A) \models R .$$

Compared to bi-abduction

- ▶ *Bounded abduction* solution: 1 formula constrained from above and below
- ▶ *Bi-abduction* solution: 2 formulas, one constrained from above and one from below
- ▶ Bounded abduction: fixed lower and upper bounds give considerable guidance to solvers
- ▶ Bi-abduction: bounds are part of the solution

# Solving Bounded Abduction

$$L \vdash (\exists X. M * [ ]) \vdash R$$

Sound but incomplete algorithm

1. Find a *coloring* of  $L$ 
  - ▶ each heaplet in  $L$  is either red or blue
  - ▶ red heaplets satisfy  $M$ , blue heaplets are left over
  - ▶ computed by recursion on proof of  $L \vdash (\exists X. M * \text{true})$
2. Find a *colored strengthening*  $\Pi : [M']^r * [A]^b$  of  $R$ 
  - ▶ entails  $R$
  - ▶ is colored such that
    - ▶ red heaplets correspond to red heaplets of  $L$
    - ▶ blue heaplets correspond to blue heaplets of  $L$
  - ▶ computed by recursion on proof of  $L \vdash R$  using coloring of  $L$
3. Check  $\Pi' : M * A \models R$ , where  $\Pi'$  is the strongest pure formula implied by  $L$ 
  - ▶ necessary because  $M$  may be weaker than  $M'$
  - ▶ if entailment check fails, then algorithm fails
  - ▶ if entailment check succeeds, then  $\Pi'' : A$  is a solution
    - ▶  $\Pi''$  is all equalities and disequalities used in proof of  $\Pi' : M * A \models R$

# Bounded Abduction: Example

## Example

$$\underbrace{x \mapsto [a, y] * y \mapsto [b, \text{null}] \vdash \text{ls}(x, y) * []}_{L} \vdash \underbrace{(\exists z. x \mapsto [a, z] * \text{ls}(y, \text{null}))}_{R}$$

1. Color  $L$ :  $[x \mapsto [a, y]]^r * [y \mapsto [b, \text{null}]]^b$  using proof of  $L \vdash \text{ls}(x, y) * \text{true}$
2. Color  $R$ :  $(\exists z. [x \mapsto [a, z]]^r * [\text{ls}(y, \text{null})]^b)$  using proof of  $L \vdash R$
3. Prove

$$\underbrace{x \neq \text{null} \wedge y \neq \text{null} \wedge x \neq y}_{\text{strongest pure consequence of } L} : \text{ls}(x, y) * \text{ls}(y, \text{null}) \models R$$

This proof succeeds, and uses pure assertion  $x \neq y$ .

4. Return solution  $x \neq y : \text{ls}(y, \text{null})$

## Computing Spatial Interpolants

Given command  $c$  and Sep formulas  $S$  and  $I'$  such that  $\text{exec}(c, S) \vdash I'$

Compute a Sep formula  $\text{itp}(S, c, I')$  such that  $S \models I$  and  $\{I\} c \{I'\}$  is valid

$$\text{itp}(S, x \rightarrow N_i := E, I') = (\exists \vec{a}, \vec{z}. A * x \mapsto [\vec{a}, \vec{z}])$$

where  $A$  satisfies

$$\text{exec}(c, S) \vdash (\exists \vec{a}, \vec{z}. x \mapsto [\vec{a}, \vec{z}[E/z_i]] * [A]) \vdash I'$$

### Example

Suppose  $S = t \mapsto [4, y, \text{null}] * x \mapsto [2, \text{null}, \text{null}]$        $c = t \rightarrow N_0 := x$        $I' = \text{bt}(t)$

Compute  $\text{exec}(c, S) = t \mapsto [4, x, \text{null}] * x \mapsto [2, \text{null}, \text{null}]$

Solve  $\text{exec}(c, S) \vdash (\exists a, z_1. t \mapsto [a, x, z_1] * [ ]) \vdash I'$

One solution is  $\text{bt}(x) * \text{bt}(z_1)$ , yielding

$$\text{itp}(S, c, I') = (\exists a, z_0, z_1. t \mapsto [a, z_0, z_1] * \text{bt}(z_1) * \text{bt}(x))$$



# Spatial Interpolation Modulo Theories

Given proof  $\zeta$  of  $\{true : emp\} \pi \{true : true\}$ , and a postcondition  $\phi$

Transform  $\zeta$  into proof of  $\{true : emp\} \pi \{\phi : true\}$

1. Traverse  $\zeta$  and build
  - ▶ *refined* proof  $\zeta'$  where refinements may contain 2nd-order variables
  - ▶ constraint system  $\mathcal{C}$  which encodes logical dependencies between 2nd-order variables
2. Solve  $\mathcal{C}$ 
  - ▶ for an assignment of data formulas to 2nd-order variables that satisfies all constraints
3. If successful, instantiate 2nd-order variables in  $\zeta'$ 
  - ▶ yields valid proof of  $\{true : emp\} \pi \{\phi : true\}$

Sound and Complete (per path, when heap-feasible)

# Spatial Interpolation Modulo Theories: Example

Refined memory safety proof  $\zeta'$

```
{R0(i) : true}
i = nondet(); x = null
{R1(i) : ls((λa.Rls1(ν, i)), x, null)}
assume(i != 0); ...; i-;
{R2(i) : ls((λa.Rls2(ν, i)), x, null)}
assume(i == 0)
{R3(i) : ls((λa.Rls3(ν, i)), x, null)}
assume(x != null)
{ (∃d', y. R4(i, d') : x ↦ [d', y]) }
```

Constraint system  $\mathcal{C}$

```
R0(i') ← true
R1(i') ← R0(i)
R2(i') ← R1(i) ∧ i ≠ 0 ∧ i' = i + 1
R3(i) ← R2(i) ∧ i = 0
R4(i, d') ← R3(i) ∧ Rls3(d', i)
Rls2(ν, i') ← R1(i) ∧ Rls1(ν, i) ∧ i ≠ 0 ∧ i' = i + 1
Rls2(ν, i') ← R1(i) ∧ ν = i ∧ i ≠ 0 ∧ i' = i + 1
Rls3(ν, i) ← R2(i) ∧ Rls2(ν, i) ∧ i = 0
d' ≥ 0 ← R4(i, d')
```

Solution  $\sigma$

```
R0(i) : true
R1(i) : true
R2(i) : true
R3(i) : true
R4(i, d') : d' ≥ 0
Rls1(ν, i) : ν ≥ i
Rls2(ν, i) : ν ≥ i
Rls3(ν, i) : ν ≥ 0
```

Symbolic Heaps (strongest post)	$true : \text{emp}$	$x = \text{null} : \text{emp}$	$true : x \mapsto [d', \text{null}]$	$true : x \mapsto [d', \text{null}]$	$true : x \mapsto [d', \text{null}]$
	⇓	⇓	...	...	
Spatial Interpolants	$true : \text{emp}$	$true : \text{ls}(x, \text{null})$	$true : \text{ls}(x, \text{null})$	$true : \text{ls}(x, \text{null})$	$true : x \mapsto [d', n']$
	⇑	⇑	...	...	
Spatial Interpolants Modulo Theories	$true : \text{emp}$	$true : \text{ls}((\lambda\nu.\nu \geq i), x, \text{null})$	$true : \text{ls}((\lambda\nu.\nu \geq i), x, \text{null})$	$true : \text{ls}((\lambda\nu.\nu \geq 0), x, \text{null})$	$d' \geq 0 : x \mapsto [d', n']$

## Conclusions & Challenges

- ▶ SPLINTER is IMHO an important step in precise and generic automatic heap/data analyses
- ▶ Novel heap analysis, that specializes to a leading technique for numerical and control-sensitive property verification
  
- ▶ Not the last word on interface between spatial interpolation and bounded abduction
- ▶ Unclear if the spatial then data phasing can be relaxed
- ▶ Want better understanding of currently enumerative heuristic for spatial interpolation of assumptions
- ▶ Want better under-approximation of classical conjunction in separation logic
  - ▶ or generalize everything to handle it natively
- ▶ Want to revise “real” separation logic provers to generate data constraints