

# *Pointer Analysis and Separation Logic*

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Thèse pour l'obtention du Doctorat de l'École Polytechnique

1er décembre 2007

# Outline

## 1. Introduction

(a) **General**

(b) Introduction to separation logic

## 2. Contents of the thesis

(a) Results

(b) Adding fixpoints to separation logic

(c) Pointer analysis: an abstract language to translate separation logic formulae

## 3. Comparisons

## 4. Conclusions

# Analysing programs

## Why ?

- **Safety:** Programs are used in spaceships, public transportations, powerplants, banking,...
- **Money:** Debugging (finding errors) is a big part of the effort of programming

## How ?

- The industry usually does **testing** (trying to run the program in various situations) but this is **not safe**:
  - one cannot test a program on an infinity of values to cover all behaviours;
  - one cannot run a program forever before insuring it behaves properly forever.
- **Formal methods** try to address the problem by providing mathematically **sound techniques** that guarantee a full coverage of all program behaviours.

## Requirements for our analyses

- are always **safe**: if we say no error, there are indeed no error possible, we cover all possible behaviours of the programs;
- can be **unprecise**;  
From undecidability theorems: for any analyzer, there always exist programs for which it will answer “I don’t know” (or not terminates).
- always **terminates**;
- are **automatised**, we do not want to make proofs by hand.

# The methodology: Abstraction

Example, the program

$$x := y + 3; z := 3/x;$$

runs to a division error if  $x = 0$  that is if  $y = -3$ .

We can not try all integer for  $y$  to find this  $-3$ .

So we build an **abstract domain**, for example the *sign domain* and we get that:

if $y$ is	then $x$ is	and the result is
$> 0$	$> 0$	no ERROR
$= 0$	$> 0$	no ERROR
$< 0$	DONT_KNOW	DONT_KNOW
DONT_KNOW	DONT_KNOW	DONT_KNOW

If the answers are too imprecise, we **refine** our abstract domain, which means create or use a domain for which the answer is more costly to get but more precise.

## Using logics to analyse programs

In the history of program analysis, people have often used [Hoare logics](#) as *abstract domains*.

Take a short program:

$$x := 3; y := x;$$

You can run it starting with  $x$  and  $y$  equal to 0

$$[x \mapsto 0 \mid y \mapsto 0] \xrightarrow{x:=3;} [x \mapsto 3 \mid y \mapsto 0] \xrightarrow{y:=x;} [x \mapsto 3 \mid y \mapsto 3]$$

or you could also have

$$[x \mapsto 5 \mid y \mapsto 2] \xrightarrow{x:=3;} [x \mapsto 3 \mid y \mapsto 2] \xrightarrow{y:=x;} [x \mapsto 3 \mid y \mapsto 3]$$

So people started to use logic to characterise the state before and after running a program:

$$\{\text{true}\} x:=3; y:=x; \{x = 3 \wedge y = x\}$$

## Automatisation and Search for precision:

We wrote

$$\{\text{true}\} x:=3; y:=x; \{x = 3 \wedge y = x\}$$

- we could also have written

$$\{x = 5\} x:=3; y:=x; \{x = 3 \wedge y = x\}$$

**Weakest precondition:**  $\{ ? \} C \{ Q \}$

for a formula  $Q$ , and a program  $C$  what is the least restrictive formula  $P$  such that  $\{P\}C\{Q\}$  is correct ?

- we could also have written

$$\{\text{true}\} x:=3; y:=x; \{x = 3\}$$

**Strongest postcondition:**  $\{P\}C\{ ? \}$

for a formula,  $P$ , and a program,  $C$ , what is the most precise formula  $Q$  such that  $\{P\}C\{Q\}$  is correct ?

# Pointer programs

The choice of the *abstract domain* is driven by the kind of program to analyse and the kind of property we want to prove.

We focused on programs using pointers, for what is called **pointer analysis**: check dereferencing errors, aliases, ...

Example of a pointer program with a bug, where  $(a \hookrightarrow b, c)$  asserts that  $a$  points to a cons cell whose head value is  $b$  and tail value is  $c$ :

$$\begin{array}{l} \uparrow \{ \exists z_1, z_2. (\text{nil} \hookrightarrow z_1, z_2) \equiv \mathbf{FALSE} \} \\ \quad \mathbf{x} := \text{nil}; \\ \quad \{ \exists z_1, z_2. (x \hookrightarrow z_1, z_2) \} \\ \quad \quad \mathbf{z} := \mathbf{x}; \\ \quad \{ \exists z_1, z_2. (z \hookrightarrow z_1, z_2) \} \\ \quad \quad \mathbf{y} := \mathbf{z} \cdot \mathbf{1}; \\ \quad \quad \{ \mathbf{TRUE} \} \end{array}$$



## Pointers analyses: Shape/alias analyses

- **Shape analyses**: the analysis builds a graph where
  - the nodes represent locations in the heap
  - the edges represent fields between locations

The analysis usually does approximation (represent more or less nodes/fields than what is in the heap) and computes some more informations about the nodes or edges of the graph.

- **Alias analyses**: a point-to analysis which computes sets of variables

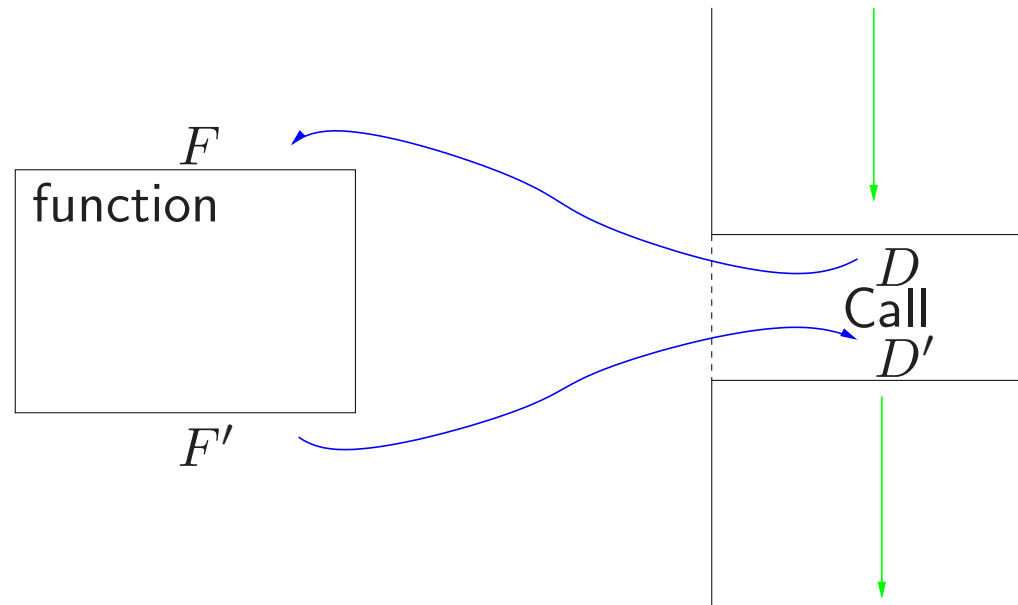
There have been and there are still tons of work on pointers: TVLA [POPL'99, SAS'00], Smallfoot, SpaceInvader [FMCO'05, SAS'07], Magill [SPACE'06], Whaley Rinard, Vivien Rinard [PLDI'01], Salcianu, Yang [ESOP'05], Rival [SAS'07], Andersen, Steensgaard, Heintze, Tzolovski, Foster Aiken [SAS'00], Ryder Landi, Emilianov, Deutsch, Jonkers, Møller, Reddy, ...

# Separation logic: a logic for pointer analysis

Separation logic allows easy descriptions of memory states, e.g.

- $x$  points to a list of [1;2;3]  
 $\exists x_2, x_3. (x \hookrightarrow 1, x_2) * (x_2 \hookrightarrow 2, x_3) * (x_3 \hookrightarrow 3, \text{nil})$
- $x$  and  $y$  are aliased pointers  
 $x = y \wedge \exists x_1, x_2. (x \hookrightarrow x_1, x_2)$
- Partitioning:  $x$  and  $y$  belong to two disjoint parts of the heap which have no pointers from one to the other...

- We wanted to use separation logic as an **interface** language for **modular analysis**



$F, F'$ : sep. logic formulae;  $D, D'$  other analysis's domain elements.

So we wanted to characterise programs with pre- and post-conditions in sep. logic, and translate formulae into and from other domains. For this last point, we created an intermediate language into which we translate separation logic formulae.

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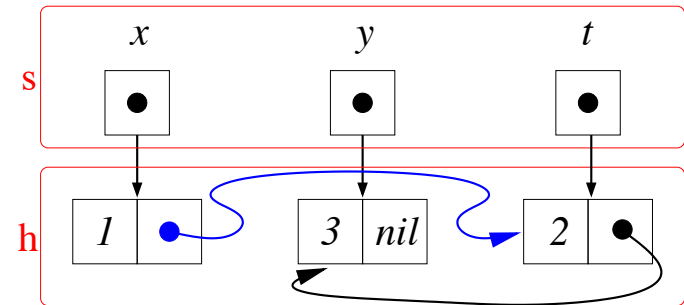
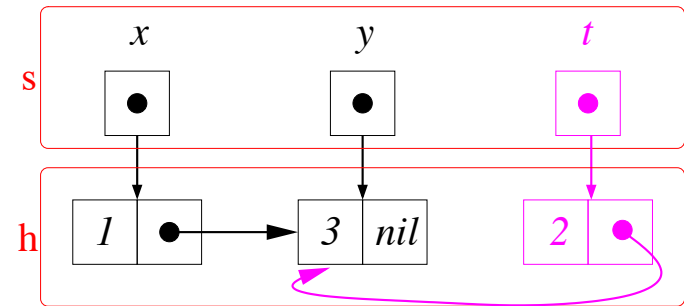
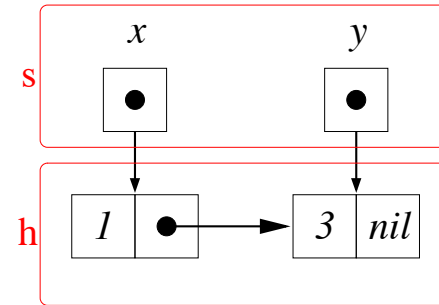
## Example for a piece of code inserting a cell in a linked list

$$\{(x \mapsto 1, y) * (y \mapsto 3, \text{nil})\}$$

$$t := \text{cons}(2, y);$$

$$\{(x \mapsto 1, y) * (y \mapsto 3, \text{nil}) * (t \mapsto 2, y)\}$$

$$x \cdot 2 := t;$$

$$\{(x \mapsto 1, t) * (y \mapsto 3, \text{nil}) * (t \mapsto 2, y)\}$$


# Local reasoning

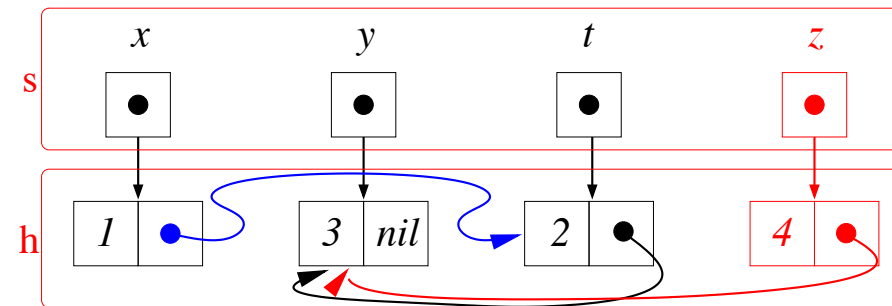
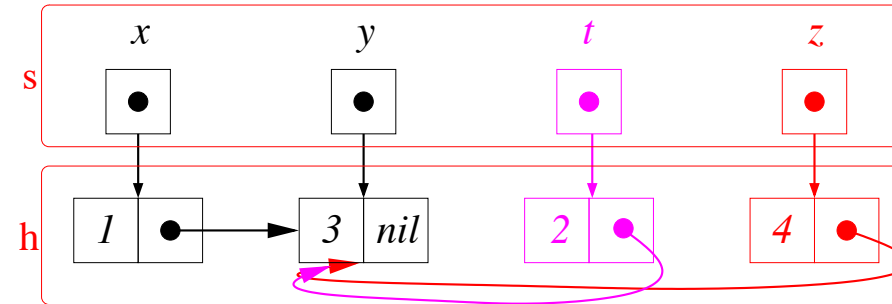
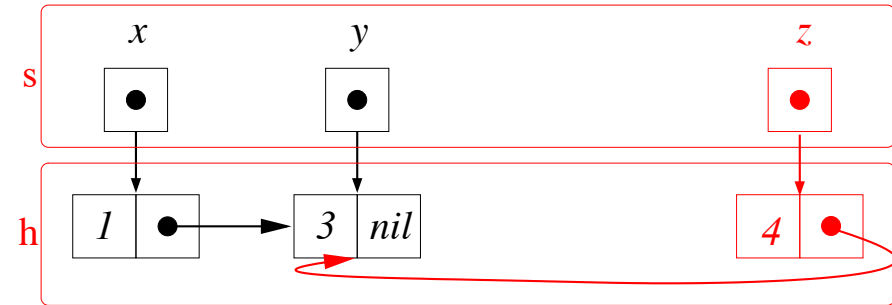
$$\left\{ \begin{array}{l} (x \mapsto 1, y) * (y \mapsto 3, \text{nil}) \\ *(z \mapsto 4, y) \end{array} \right\}$$

$t := \text{cons}(2, y);$

$$\left\{ \begin{array}{l} (x \mapsto 1, y) * (y \mapsto 3, \text{nil}) * (t \mapsto 2, y) \\ *(z \mapsto 4, y) \end{array} \right\}$$

$x \cdot 2 := t;$

$$\left\{ \begin{array}{l} (x \mapsto 1, t) * (t \mapsto 2, y) * (y \mapsto 3, \text{nil}) \\ *(z \mapsto 4, y) \end{array} \right\}$$





# Separation logic

<i>Classical</i> connectives			
$E = E'$			false
$P \Rightarrow Q$			$\exists x.P$
<i>Spatial</i> connectives			
<b>emp</b>	Empty heap		$E \mapsto E_1, E_2$ Points to
$P * Q$	Spatial conj.		$P \multimap Q$ Spatial imp.

Expressions can be:

$x \mid n \mid \mathbf{nil} \mid \mathit{True} \mid \mathit{False} \mid E_1 \text{ op } E_2$

## Domain of interpretation: *State*

We have a set of variables  $Var$ .

$$\begin{array}{llll} Val & = & Int \cup Bool \cup Atoms \cup Loc & Values \\ S & = & Var \rightarrow Val & Stacks \\ H & = & Loc \rightarrow Val \times Val & Heaps \\ State & = & S \times H & \end{array}$$

## Semantics of $*$

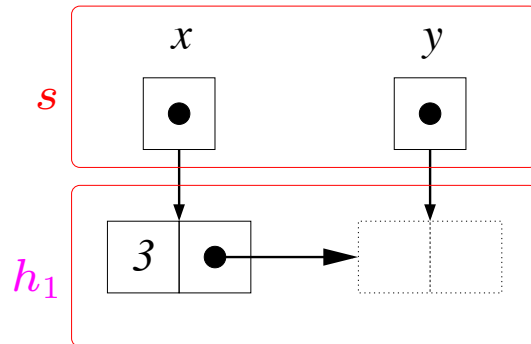
$$\llbracket P * Q \rrbracket_\rho = \left\{ s, h_0 \cdot h_1 \mid \begin{array}{l} \bullet \text{ dom}(h_0) \cap \text{dom}(h_1) = \emptyset \\ \bullet s, h_0 \in \llbracket P \rrbracket_\rho \\ \bullet s, h_1 \in \llbracket Q \rrbracket_\rho \end{array} \right\}$$

# Examples of formulae

Ex. 1

$$s = [x \rightarrow l_1, y \rightarrow l_2]$$

$$h_1 = [l_1 \rightarrow \langle 3, l_2 \rangle]$$

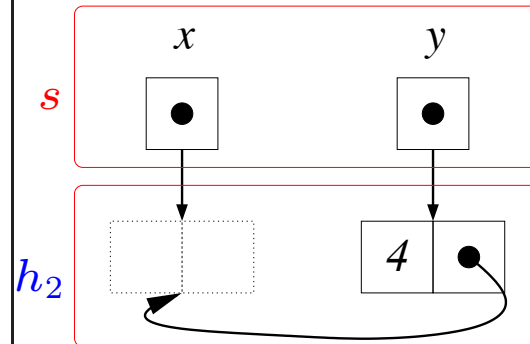


$$\models (x \mapsto 3, y)$$

Ex. 2

$$s = [x \rightarrow l_1, y \rightarrow l_2]$$

$$h_2 = [l_2 \rightarrow \langle 4, l_1 \rangle]$$

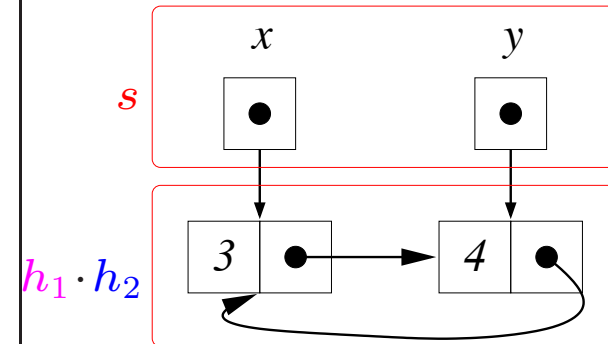


$$\models (y \mapsto 4, x)$$

Ex. 3

$$s = [x \rightarrow l_1, y \rightarrow l_2]$$

$$h_1 \cdot h_2 = \left[ \begin{array}{l} l_1 \rightarrow \langle 3, l_2 \rangle, \\ l_2 \rightarrow \langle 4, l_1 \rangle \end{array} \right]$$



$$\models (x \mapsto 3, y) * (y \mapsto 4, x)$$

$$\not\models (x \mapsto 3, y) \wedge (y \mapsto 4, x)$$

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## Result 1

- ◆ When starting the work, recursive formulae could not be expressed within separation logic, and moreover pre-conditions ( $wlp$ ) and post-conditions ( $sp$ ) for `while` loops could not be expressed
- We have extended separation logic such that we can express **recursive formulae**, and use them to instantiate existing triples rules and new ones.
- We have a backward ( $wlp$ ) and forward ( $sp$ ) analysis with their soundness proofs for any formula and any command, in particular for **while-loops**.
- We have proved **various properties** of the extended logic.

## Result 2

We have built an intermediate language such that:

- it is similar to the existing shape/alias analysis domains to allow translations of our intermediate language from and to those existing domains
- it comes with a concrete semantics in term of sets of states which is the same domain as for the formulae's semantics
- we translated the separation logic formulae into our intermediate language and proved sound those translations
- it is a partially reduced product of different subdomains so that we can cheaply tune the precision depending on the needs (for example, the language is parametrised by a numerical domain which can be ignored if we do not care about numericals)

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# The logic: $BI^{\mu\nu}$

<i>Classical connectives</i>			
$E = E'$			false
$P \Rightarrow Q$			$\exists x.P$
<i>Spatial connectives</i>			
emp	Empty heap		$E \mapsto E_1, E_2$ Points to
$P * Q$	Spatial conj.		$P \multimap Q$ Spatial imp.
<i>Fixpoints connectives</i>			
$X_v$	Variable for formulae		$P[E/x]$ Postponed substitution
$\nu X_v.P$	Greatest fixpoint		$\mu X_v.P$ Least fixpoint

$Var_v = \{X_v, Y_v, \dots\}$  infinite set of variables of formulae



## Fixpoint connectives semantics

$\rho$  is an environment mapping formula variables to sets of State

$$\llbracket X_v \rrbracket_\rho = \rho(X_v) \text{ if } X_v \in \text{dom}(\rho)$$

$$\llbracket \mu X_v. P \rrbracket_\rho = \text{lfp}_{\emptyset}^{\subseteq} \lambda Y. \llbracket P \rrbracket_{[\rho | X_v \rightarrow Y]}$$

$$\llbracket \nu X_v. P \rrbracket_\rho = \text{gfp}_{\emptyset}^{\subseteq} \lambda Y. \llbracket P \rrbracket_{[\rho | X_v \rightarrow Y]}$$

$$\llbracket P[E/x] \rrbracket_\rho = \left\{ s, h \mid \begin{array}{l} \llbracket E \rrbracket^s \text{ exists and} \\ [s \mid x \rightarrow \llbracket E \rrbracket^s], h \in \llbracket P \rrbracket_\rho \end{array} \right\}$$

$\llbracket \cdot \rrbracket$  may be undefined: if the formula is not closed for variables of formulae (e.g.  $\llbracket X_v \rrbracket_{\emptyset}$ ) or if the fixpoint does not exist (e.g.  $\llbracket \mu X_v. \neg X_v \rrbracket_\rho$ )

## [ / ] is not { / }

$\llbracket \text{true}[y/x] \rrbracket = \{s, h \mid \llbracket y \rrbracket^s \text{ exists} \}$  Postponed substitution connective

$\llbracket \text{true}\{y/x\} \rrbracket = \llbracket \text{true} \rrbracket = \text{State}$  Capture-avoiding substitution

$\{\text{true}\}x := y\{\text{true}\}$  is false since the command will be stuck from a state that has no value on its stack for  $y$

but  $\{is(y)\}x := y\{\text{true}\}$  is true

so  $\{P\{y/x\}\}x := y\{P\}$  is unsound

but  $\{P[y/x]\}x := y\{P\}$  is sound

With  $is(E) \triangleq (E = E)$ , since  $\llbracket E = E \rrbracket_\rho = \{s, h \mid \llbracket E \rrbracket^s \text{ has a value}\}$

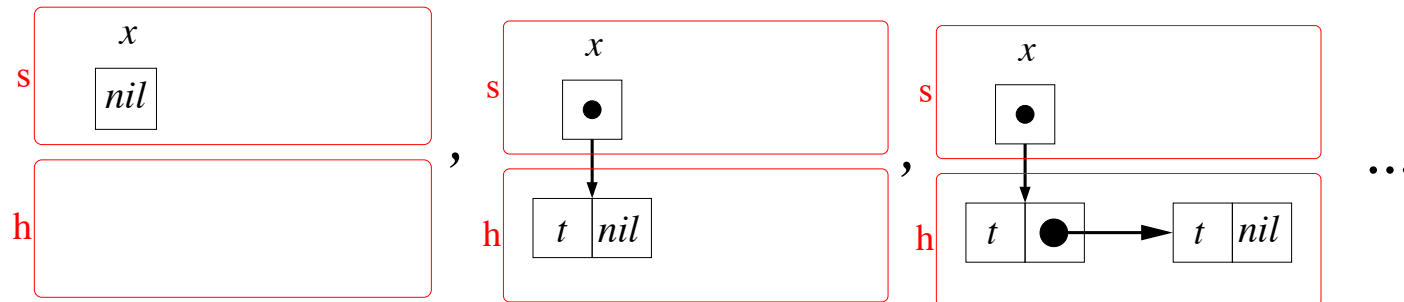
## List formula

“x points to a finite non-cyclic list of True”

$$\mathbf{nclist}(x) \triangleq \mu X_v. ((x = \mathbf{nil}) \vee \exists x_2. (x \mapsto \mathbf{true}, x_2 * X_v[x_2/x]))$$

Notice the combination of fixpoint and postponed substitution to write recursive definitions

“ $\mathbf{nclist}(x) = (x = \mathbf{nil}) \vee \exists x_2. (x \mapsto \mathbf{true}, x_2 * \mathbf{nclist}(x_2))$ ”



(to define finite or infinite lists replace  $\mu$  by  $\nu$ )

## Tree formula

“x points to a tree of True”

$$\text{tree}(x) \triangleq \mu X_v. \quad (x = \text{nil}) \vee \exists x_l, x_r, x'. \\ ((x \mapsto \text{true}, x') * (x' \mapsto x_l, x_r) * X_v[x_l/x] * X_v[x_r/x])$$

## Unfolding theorems

As usual, the following theorems hold

$$\mu X_v. P \equiv P\{\mu X_v. P/X_v\}$$

$$\nu X_v. P \equiv P\{\nu X_v. P/X_v\}$$

We have proved some other theorems like variable renaming, variable substitution, equivalence of  $\mu$  and  $\nu$  using  $\neg$ , simplifications of  $[ / ]$  by equivalent formulae.

## Backward Analysis: $wlp$

$wlp$  : weakest liberal precondition, such that

$$\{wlp(P, C)\}C\{P\} \text{ true}$$

$wlp$  is expressed and proved sound for any  $P$  and any  $C$

- ◆  $\{P[E/x]\}x := E\{P\}$
- ◆  $\{\nu X_v. ((E = \text{true} \wedge wlp(X_v, C)) \vee (E = \text{false} \wedge P))\}$   
 $\text{while } E \text{ do } C \{P\}$

## Forward analysis: $sp$

$sp$  : strongest postcondition, such that

$$\llbracket sp(P, C) \rrbracket_{\emptyset} = \{m' \mid \exists m \in \llbracket P \rrbracket_{\emptyset}. C, m \rightsquigarrow^* m'\}$$

$sp$  are expressed and proved sound for all  $P$  and all  $C$

- ◆  $sp(P, x := E) = \exists x'. P[x'/x] \wedge x = E\{x'/x\}$  with  $x' \notin FV(E, P)$
- ◆  $sp(P, \text{while } E \text{ do } C) =$   
 $(E = \text{false}) \wedge (\mu X_v. sp(X_v \wedge E = \text{true}, C) \vee P)$

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## Elements of the domain are tuples

Elements are 7-tuples  $(sg, hu, ho, sn, sn^\infty, t, d)$

- $sg \in SG$  A kind of shape graph
- $hu \in \mathcal{P}(TVar)$  Under approximation of heap domain
- $ho \in \mathcal{P}(TVar) \uplus \text{full}$  Over approximation of heap domain
- $sn \in \mathcal{P}(TVar)$  Set of finite summary nodes
- $sn^\infty \in \mathcal{P}(TVar)$  Set of infinite summary nodes
- $t \in TB$  Tabular expressing inclusions on the concrete values represented
- $d \in \mathcal{D}$  Numerical domain

# Simple abstract values (*Nilt*, *Truet*,...) and disjunction

Formulae	$x = \text{nil}$
Semantics	$\{s, h \mid s(x) = \text{nil}\}, \dots$
Translation	$\left( \boxed{x} \rightarrow \text{Nilt}, \text{---}, \text{---}, \text{---}, \text{---}, \text{---} \right)$
Formulae	$(x = \text{nil} \vee x = \text{true})$
Translation	$\left( \boxed{x} \rightarrow \begin{matrix} \text{Nilt} \\ \text{Truet} \end{matrix}, \text{---}, \text{---}, \text{---}, \text{---}, \text{---} \right)$

# Aliasing and Conjunction

We want cheap translation of  $\wedge : T(A \wedge B) \triangleq T'(T'(\top, A), B)$

Formula	$x = y$
Constraints	refine the information for one variable while also refining the information of the second one in a cheap way
Adds	infinite set of auxiliary variables $TVar$ $VAR \triangleq Var \uplus TVar$
Translation	
Formula	$x = y \wedge x = \text{nil}$
Translation	


# Quantifier

Formula	$x = y \wedge x = \mathbf{nil}$
Translation	<p>Diagram illustrating the translation of the formula <math>x = y \wedge x = \mathbf{nil}</math>. It shows two variables, <math>x</math> and <math>y</math>, both pointing to a node <math>\alpha</math>. Node <math>\alpha</math> points to a node labeled <math>Nilt</math>. The diagram is enclosed in large parentheses.</p>
Formula	$(\exists x. x = y \wedge x = \mathbf{nil}) \equiv (y = \mathbf{nil})$
Translation	<p>Diagram illustrating the translation of the formula <math>(\exists x. x = y \wedge x = \mathbf{nil}) \equiv (y = \mathbf{nil})</math>. It shows a single variable <math>y</math> pointing to a node <math>\alpha</math>. Node <math>\alpha</math> points to a node labeled <math>Nilt</math>. The diagram is enclosed in large parentheses.</p>

# Numericals

Formula	$(x < y + 3)$
Translation	<p style="text-align: center;"><math>d \in \mathcal{D}</math> encodes that <math>\alpha &lt; \beta + 3</math></p>

# Dangling pointers

Formula	“x is a location not allocated” $\text{isdangling}(x) \equiv \text{isloc}(x) \wedge \neg \text{isinheap}(x)$
Semantics	$\{s, h \mid s(x) \in \text{Loc} \wedge s(x) \notin \text{dom}(h)\}$
Translation	

where

$$\begin{aligned} \text{isint}(x) &\equiv \exists n. n = x + 1 \\ \text{isloc}(x) &\equiv \neg(x = \text{nil}) \wedge \neg(x = \text{true}) \wedge \neg(x = \text{false}) \wedge \neg(\text{isint}(x)) \\ \text{isinheap}(x) &\equiv \exists x_1, x_2. (x \hookrightarrow x_1, x_2) \end{aligned}$$

## emp, approximation of the heap

Formula	emp
Semantics	$\{s, h \mid \text{dom}(h) = \emptyset\}$
Subdomains	$HU \triangleq \mathcal{P}(TVar)$ $HO \triangleq \mathcal{P}(TVar) \uplus \text{full}$
Translation	$(\top, \emptyset, \emptyset, -, -, -, -)$

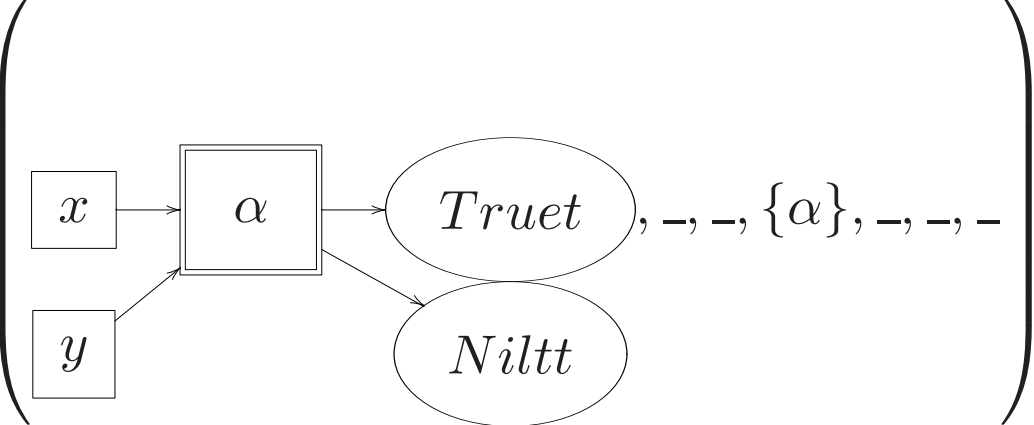


# Heap locations

Formula	$(x \mapsto \text{true}, \text{nil})$
Semantics	$\{s, h \mid [s(x) \rightarrow \langle \text{True}, \text{nil} \rangle] = h\}$
Translation	
Formula	$(x \hookrightarrow \text{true}, \text{nil})$
Semantics	$\{s, h \mid [s(x) \rightarrow \langle \text{True}, \text{nil} \rangle] \subseteq h\}$
Translation	

## Summary nodes

Variables represent at most one value. To allow approximation we introduce summary nodes which can represent several values.

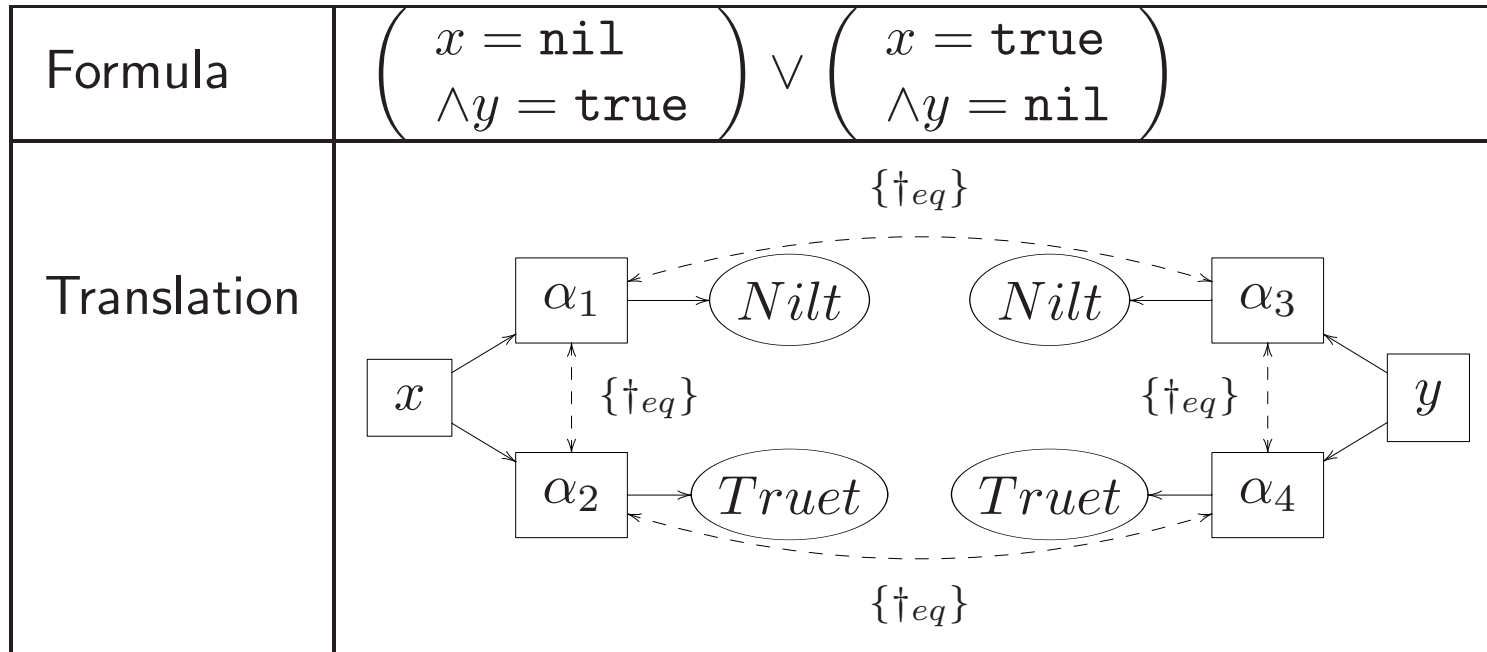
Formula	approx. of $(x = \text{true} \wedge y = \text{nil})$ by $(x = \text{true} \vee x = \text{nil})$ $\wedge (y = \text{true} \vee y = \text{nil})$
Translation	 <p>The diagram illustrates the translation of variables <math>x</math> and <math>y</math> into a summary node <math>\alpha</math>. Both <math>x</math> and <math>y</math> (represented as boxes) have arrows pointing to <math>\alpha</math> (a double-bordered box). From <math>\alpha</math>, arrows point to two summary nodes: <math>Truet</math> and <math>Niltt</math> (represented as ovals). The entire diagram is enclosed in large parentheses.</p>

## Finite acyclic list of *True* starting from $x$

Formula	$\mu X_v. \left( \begin{array}{l} (x = \text{nil}) \vee \exists x_2. \\ x \hookrightarrow (\text{true}, x_2) * X_v[x_2/x] \end{array} \right)$
Translation	

$\emptyset$  is the set of infinite summary nodes, for infinite list  $\mu$  would be replaced by  $\nu$  and  $\emptyset$  by  $\{\alpha\}$ .

# Tabular to increase precision of union



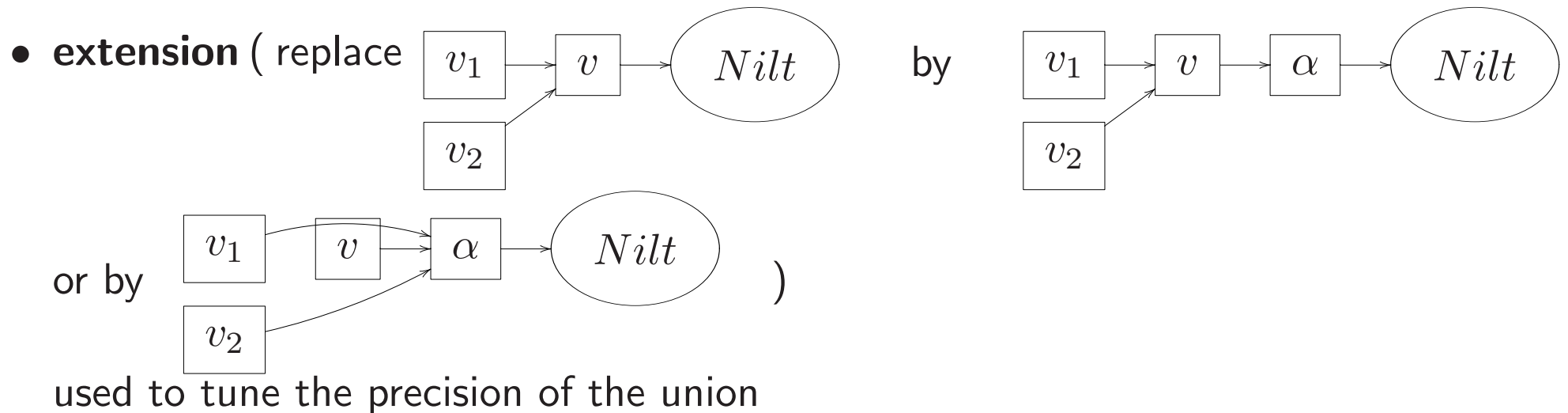
The dashed arrows are drawn to represent the tabular:

	$\alpha_1$	$\alpha_2$	$\alpha_3$	$\alpha_4$
$\alpha_1$	$\{\dagger_{eq}, =_{eq}\}$	$\{\dagger_{eq}\}$	$\{\dagger_{eq}\}$	$\top_{eq}$
$\alpha_2$		$\{\dagger_{eq}, =_{eq}\}$	$\top_{eq}$	$\{\dagger_{eq}\}$
$\alpha_3$			$\{\dagger_{eq}, =_{eq}\}$	$\{\dagger_{eq}\}$
$\alpha_4$				$\{\dagger_{eq}, =_{eq}\}$

# Operations

We have **proved soundness** of the operations we use, in particular:

- **union, intersection**



- **merging** (replace  $[v_1 \rightarrow S_1 \mid v_2 \rightarrow S_2]$  by  $[v_2 \rightarrow (S_1 \cup S_2)]$ )  
used with the **widening**

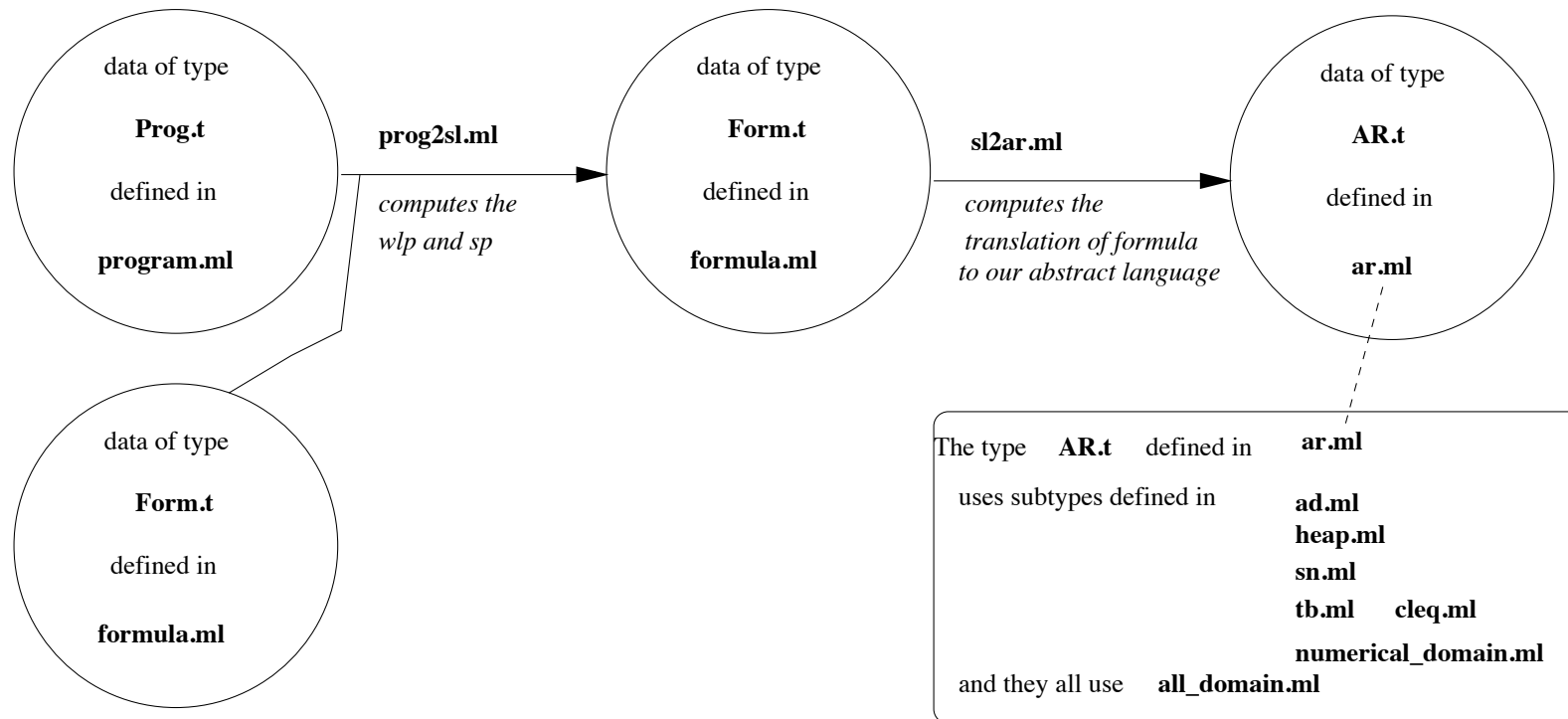
- **translations** from formulae to the abstract language

# Comparisons

- the  $\bullet \begin{matrix} 1 \\ \rightsquigarrow \\ 2 \end{matrix}$  represent nodes in the usual shape graphs
- summary nodes, as for other shape graphs, seems to give more possibilities than predicate abstraction (with each time a specific predicate for list, etc...) but the framework of predicate abstraction and their algorithm/heuristics (like folding/unfolding) could probably also be use on our graphs
- a lonely outgoing edge can be seen as a “must” arrow (or valued 1), several outgoing edges from a variable can be seen as a “may” arrow (or valued 1/2, but it is a bit more precise because we know that one of them should exist), and an edge to  $\emptyset$  can be seen as a “must not” arrow (or valued 0)
- we deal with numerical information (not many works do, for example *Magill & al.* also do)
- we have a formal semantic of our domain, the semantics of auxiliary variables are formally defined and formally used in the proofs. We don't have to check for equalities of variables
- we directly have in the domain the “Dangling” information which is suitable for cleaning checking

# Prototype

We have build a prototype implementation:



Implement the computation of pre- and post-conditions in the extended separation logic and the translation of the formulae into the abstract language.

# Conclusions

- ✓ We added fixpoints to separation logic, which provides a way to express recursive formulae and while-loops pre- and post-conditions.
- ✓ We proved useful properties about the extended logic
- ✓ We gave a precise semantics of the abstract language for separation formulae in terms of sets of states. We gave a semantics to auxiliary variables and did not leave this as an implementation design question
- ✓ We designed the abstract language as a partially reduced product of subdomains. We combined the domain's heap analysis with a numerical domain which could be chosen from existing ones (e.g. polyhedra, octogons)
- ✓ We designed a novel tabular data structure which allows extra precision by using a graph of sets instead of sets of graphs
- ✓ We expressed and proved the translation of separation logic formulae into our abstract language and implemented it in a prototype.



## Future work

- finish the prototype implementation, profiling studies with standard example programs, experiment various strategies to build summary nodes
- improve the abstract language with labels indicating where we do overapproximation
- add sugar structures to do abstract language like lists (or a system to add those structures) and functions to use the mechanisms of folding/unfolding those structures when needed
- adding labels to \*1 and \*2 to allow to have several families of uncycling edges instead of one
- it could be fun to design a program analysis directly in our abstract language
- actually do interface the abstract language with some existing pointer analysis

End