Shape analysis based on separation logic MPRI — Cours "Interprétation abstraite : application à la vérification et à l'analyse statique"

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Overview of the lecture

How to reason about memory properties

Last lecture:

- analyses specific to several kinds of structures
- concrete and abstract memory models
- an introduction to shape analysis with TVLA

Today:

- a logic to describe properties of memory states
- abstract domain
- static analysis algorithms
- combination with numerical domains
- widening operators...

Outline

- 1 An introduction to separation logic
- 2 A shape abstract domain relying on separation
- Combination with a numerical domain
- 4 Standard static analysis algorithms
- 5 Inference of inductive definitions / call-stack summarization
- Conclusion

Our model

Environment + Heap

- Addresses are values: $\mathbb{V}_{addr} \subseteq \mathbb{V}$
- Environments $e \in \mathbb{E}$ map variables into their addresses
- Heaps ($h \in \mathbb{H}$) map addresses into values

$$\begin{array}{ll} \mathbb{E} & = & \mathbb{X} \to \mathbb{V}_{\mathrm{addr}} \\ \mathbb{H} & = & \mathbb{V}_{\mathrm{addr}} \to \mathbb{V} \end{array}$$

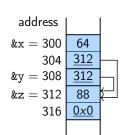
- h is actually only a partial function
- Memory states:

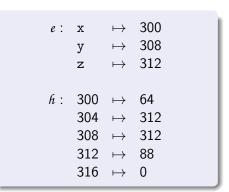
$$\mathbb{M} = \mathbb{E} \times \mathbb{H}$$

Example of a concrete memory state (variables)

- x and z are two list elements containing values 64 and 88, and where the former points to the latter
- y stores a pointer to z

Memory layout (pointer values underlined)

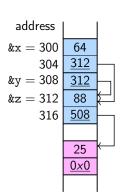




Example of a concrete memory state (variables + heap)

- same configuration
- + z points to a heap allocated list element (in purple)

Memory layout



| e : | х у z | | 300 308 312 |
|-----|--------------------------|----------------------|-------------------------|
| h : | 304 308 312 316 | \mapsto \mapsto | 312 312 88 508 |
| | 508 512 | $\mapsto \\ \mapsto$ | 0 |

Weak update problems

```
\begin{split} x &\in [-10, -5]; \ y \in [5, 10] \\ &\text{int} \star \ p; \\ &\text{if}(?) \\ & p = \&x; \\ &\text{else} \\ & p = \&y; \\ &\star p = 0; \end{split}
```

- ullet What is the final range for x ?
- What is the final range for y?

- After the if statement, p may contain any address in {&x, &y}
- Thus, the assignment must consider all cases, in a conservative way
- ullet Thus, x may receive a new value (0) or keep its old value
- Conclusion: $x \in [-10, 0], y \in [0, 10]$

Weak updates

Any imprecision in the analysis may lead to weak updates...

Separation logic principle: avoid weak updates

How to deal with weak updates?

Avoid them !

- Always materialize exactly the cell that needs be modified
- Can be very costly to achieve, and not always feasible
- Notion of property that holds over a memory region
- Use a special separating conjunction operator *
- Local reasoning: powerful principle, which allows to consider only part of the program memory
- Separation logic has been used in many contexts, including manual verification, static analysis, etc...

Separation logic

- Logic made of a set of formulas
- inference rules...

Pure formulas

Set of pure formulas, similar to first order logics

Denote numerical properties among the values

Heap formulas (syntax on the next slide)

- Set of formulas to describe concrete heaps
- Concretization relation of the form $(e, h) \in \gamma(F)$

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Heap formulas

Main connectors

Each formula describes a heap region

Denotations: the usual stuff...

- $\gamma(\mathsf{emp}) = \emptyset$; $\gamma(\mathsf{true}) = \mathbb{M}$
- $(e, h) \in \gamma(F' \wedge F'')$ if and only if $(e, h) \in \gamma(F')$ and $(e, h) \in \gamma(F'')$

Separating conjunction: next slide...

The separating conjunction

Single cells

$$(e, h) \in \gamma(I \mapsto v)$$
 if and only if $h = [\llbracket I \rrbracket (e, h) \mapsto v]$

Merge of concrete stores

Let
$$h_0, h_1 \in (\mathbb{V}_{\mathrm{addr}} \to \mathbb{V})$$
, such that $\mathsf{dom}(h_0) \cap \mathsf{dom}(h_1) = \emptyset$.

Then, we let $h_0 \otimes h_1$ be defined by:

$$h_0 \circledast h_1 : \operatorname{dom}(h_0) \cup \operatorname{dom}(h_1) \longrightarrow \mathbb{V}$$

$$x \in \operatorname{dom}(h_0) \longmapsto h_0(x)$$

$$x \in \operatorname{dom}(h_1) \longmapsto h_1(x)$$

Concretization of separating conjunction

- Logical formulas denote sets of heaps; concretization γ
- Binary logical connector on formulas * defined by:

$$\gamma(F_0 * F_1) = \{(e, h_0 \circledast h_1) \mid (e, h_0) \in \gamma(F_0) \land (e, h_1) \in \gamma(F_1)\}$$

Separating conjunction vs non separating conjunction

- Classical conjunction: properties for the same memory region
- Separating conjunction: properties for disjoint memory regions

$a \mapsto \&b \land b \mapsto \&a$

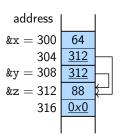
- the same heap verifies $a \mapsto \&b$ and $b \mapsto \&a$
- there can be only one cell
- thus *a* = *b*

$$a \mapsto \&b * b \mapsto \&a$$

- two separate sub-heaps respectively satisfy a → &b and b → &a
- thus $a \neq b$
- Separating conjunction and non-separating conjunction have very different properties
- Both express very different properties
 e.g., no ambiguity on weak / strong updates

An example

Concrete memory layout (pointer values underlined)



 $\begin{array}{ccc} 312 & \mapsto & 88 \\ 316 & \mapsto & 0 \end{array}$

A formula that abstracts away the addresses:

$$x \mapsto \langle 64, \& z \rangle * y \mapsto \& z * z \mapsto \langle 88, 0 \rangle$$

Separating and non separating conjunction

- There are two conjunction operators ∧ and *
- How to relate them ?

Separating conjunction vs normal conjunction

$$\frac{(e, h_0) \in \gamma(F_0) \qquad (e, h_1) \in \gamma(F_1)}{(e, h_0 \circledast h_1) \in \gamma(F_0 * F_1)} \qquad \frac{(e, h) \in \gamma(F_0) \qquad (e, h) \in \gamma(F_1)}{(e, h) \in \gamma(F_0 \land F_1)}$$

 Reminiscent of Linear Logic [Girard87]: resource aware / non resource aware conjunction operators

Programs with pointers: syntax

Syntax extension: quite a few additional constructions

```
1 ::= I-valules
                        (x \in X)
                        pointer dereference
        1 \cdot f
                        field read
e ::= expressions
                        "address of" operator
s ::= statements
        x = malloc(c) allocation of c bytes
        free(x) deallocation of the block pointed to by x
```

We do not consider pointer arithmetics here

Programs with pointers: semantics

Case of I-values:

Case of expressions:

$$[[1]](e, heap) = h([[1]](e, heap))$$
$$[[\&1]](e, heap) = [[1]](e, heap)$$

Case of statements:

- memory allocation x = malloc(c): $(e, h) \rightarrow (e, h')$ where $h' = h[e(\mathbf{x}) \leftarrow k] \uplus \{k \mapsto v_k, k+1 \mapsto v_{k+1}, \dots, k+c-1 \mapsto v_{k+c-1}\}$ and $k, \ldots, k+c-1$ are fresh in h
- memory deallocation free(x): $(e, h) \rightarrow (e, h')$ where k = e(x) and $h = h' \uplus \{k \mapsto v_k, k+1 \mapsto v_{k+1}, \dots, k+c-1 \mapsto v_{k+c-1}\}$

Shape analysis based on separation logic

Separating logic triple

Program proofs based on triples

• Notation: $\{F\}p\{F'\}$ if and only if:

$$\forall s, s' \in \mathbb{S}, \ s \in \gamma(F) \land s' \in \llbracket p \rrbracket(s) \Longrightarrow s' \in \gamma(F')$$

Hoare triples

• Application: formalize proofs of programs

A few rules (straightforward proofs):

$$\frac{F_0 \Longrightarrow F_0' \qquad \{F_0'\} p\{F_1'\} \qquad F_1' \Longrightarrow F_0'}{\{F_0\} p\{F_1\}} \ \ \textit{consequence}$$

$$\overline{\{x\mapsto?\}x:=e\{x\mapsto e\}} \ \ \textit{mutation}$$

$$\overline{\{x\mapsto?*F\}x:=e\{x\mapsto e*F\}}$$
 mutation – 2

(we assume that e does not allocate memory)

The frame rule

What about the resemblance between rules "mutation" and "mutation-2"?

Theorem: the frame rule

$$\frac{\{F_0\}s\{F_1\}}{\{F_0*F\}s\{F_1*F\}} \text{ frame}$$

- Proof by induction on the rules (see biblio for a more complete set of rules)
- Rules are proved by case analysis on the program syntax

We can reason locally about programs

Application of the frame rule

Let us consider the program below:

```
int i;

int \star x;

int \star y; {i \mapsto? * x \mapsto? * y \mapsto?}

x = &i; {i \mapsto? * x \mapsto &i * y \mapsto?}

y = &i; {i \mapsto? * x \mapsto &i * y \mapsto &i}

\star x = 42; {i \mapsto 42 * x \mapsto &i * y \mapsto &i}
```

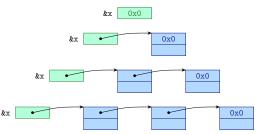
- Each step impacts a disjoint memory region
- This case is easy
 See biblio for more complex applications
 (verification of the Deutsch-Shorr-Waite algorithm)

Summarization and inductive definitions

What do we still miss?

So far, formulas denote **fixed sets of cells**Thus, no summarization of unbounded regions...

Example all lists pointed to by x, such as:



 How to precisely abstract these stores with one formula i.e., no infinite disjunction?

Inductive definitions in separation logic

List definition

$$\begin{array}{ll} \alpha \cdot \mathbf{list} &:= & \alpha = \mathbf{0} \, \wedge \, \mathbf{emp} \\ & \vee & \alpha \neq \mathbf{0} \, \wedge \, \alpha \cdot \mathbf{next} \mapsto \gamma * \alpha \cdot \mathbf{data} \mapsto \beta * \gamma \cdot \mathbf{list} \end{array}$$

Formula abstracting our set of structures:

$$\&x \mapsto \alpha * \alpha \cdot \mathsf{list}$$

- Summarization: this formula is finite and describe infinitely many heaps
- Concretization: next slide...

Practical implementation in verification/analysis tools

- Verification: hand-written definitions
- Analysis: either built-in or user-supplied, or partly inferred

Concretization by unfolding

Intuitive semantics of inductive predicates

- Inductive predicates can be **unfolded**, by unrolling their definitions Syntactic unfolding is noted $\stackrel{\mathcal{U}}{\longrightarrow}$
- A formula F with inductive predicates describes all stores described by all formulas F' such that $F \xrightarrow{\mathcal{U}} F'$

Example:

• Let us start with $x \mapsto \alpha_0 * \alpha_0 \cdot \mathbf{list}$; we can unfold it as follows: $\&x \mapsto \alpha_0 * \alpha_0 \cdot \mathbf{list}$

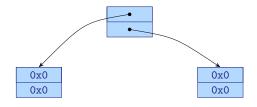
$$\begin{array}{ll} \stackrel{\mathcal{U}}{\longrightarrow} & \&\mathtt{x} \mapsto \alpha_0 * \alpha_0 \cdot \mathtt{next} \mapsto \alpha_1 * \alpha_0 \cdot \mathtt{data} \mapsto \beta_1 * \alpha_1 \cdot \mathsf{list} \\ \stackrel{\mathcal{U}}{\longrightarrow} & \&\mathtt{x} \mapsto \alpha_0 * \alpha_0 \cdot \mathtt{next} \mapsto \alpha_1 * \alpha_0 \cdot \mathtt{data} \mapsto \beta_1 * \mathsf{emp} \wedge \alpha_1 = \mathbf{0x0} \end{array}$$

• We get the concrete state below:



Example: tree

• Example:



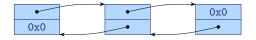
Inductive definition

• Two recursive calls instead of one:

$$\begin{array}{ll} \alpha \cdot \mathsf{tree} &:= & \alpha = \mathsf{0} \, \wedge \, \mathsf{emp} \\ & \vee & \alpha \neq \mathsf{0} \, \wedge \, \alpha \cdot \mathsf{left} \mapsto \beta * \alpha \cdot \mathsf{right} \mapsto \gamma \\ & * \beta \cdot \mathsf{tree} * \gamma \cdot \mathsf{tree} \end{array}$$

Example: doubly linked list

• Example:



Inductive definition

• We need to propagate the prev pointer as an additional parameter:

$$\begin{array}{lll} \alpha \cdot \mathbf{dII}(p) & := & \alpha = 0 \, \wedge \, \mathbf{emp} \\ & \vee & \alpha \neq 0 \, \wedge \, \alpha \cdot \mathtt{next} \mapsto \gamma * \alpha \cdot \mathtt{prev} \mapsto p * \gamma \cdot \mathbf{dII}(\alpha) \end{array}$$

Example: sortedness

Example: sorted list



Inductive definition

- Each element should be greater than the previous one
- The first element simply needs be greater than $-\infty...$
- We need to propagate the lower bound, using a scalar parameter

$$\begin{array}{ll} \alpha \cdot \mathsf{Isort}_{\mathrm{aux}}(\mathit{n}) & := & \alpha = \mathsf{0} \, \land \, \mathsf{emp} \\ & \lor & \alpha \neq \mathsf{0} \, \land \, \beta \leq \mathit{n} \, \land \, \alpha \cdot \mathsf{next} \mapsto \gamma \\ & & \ast \, \alpha \cdot \mathsf{data} \mapsto \beta \ast \gamma \cdot \mathsf{Isort}_{\mathrm{aux}}(\beta) \end{array}$$

$$\alpha \cdot \mathsf{Isort}() := \alpha \cdot \mathsf{Isort}_{\mathrm{aux}}(-\infty)$$

A new overview of the remaining part of the lecture

How to apply separation logic to static analysis and design abstract interpretation algorithms based on it ?

In this lecture, we will:

- choose a small but expressive set of separation logic formulas
- define wide families of abstract domains
- study algorithms for local concretization (equivalent to focus) and global abstraction (widening...)

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- 3 Combination with a numerical domain
- 4) Standard static analysis algorithms
- 5 Inference of inductive definitions / call-stack summarization
- 6 Conclusion

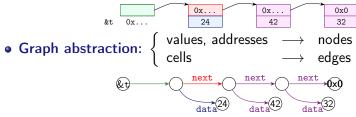
Choice of a set of formulas

Our set of predicates

- An abstract value is a separating conjunction of terms
- Each term describes
 - either a contiguous region
 - or a summarized region, described by an inductive defintion
- Abstract elements have a straightforward interpretation as a shape graph
- Unless necessary, we omit environments (concretization into sets of heaps)

Abstraction into separating shape graphs

Memory splitting into regions



Region summarization:



- abstraction parameterized by a set of inductive definitions
- Defines a concretization relation
- Let us formalize this...

Contiguous regions

Shape graphs

- Edges: denote memory regions
- Nodes: denote values, i.e. addresses or cell contents

Points-to edge, denote contiguous memory regions

- Separation logic formula: $\alpha \cdot f \mapsto \beta$
- Abstract and concrete views:





Concretization:

$$\gamma_{\mathrm{S}}(\alpha \cdot \mathbf{f} \mapsto \beta) = \{([\nu(\alpha) + \mathsf{offset}(\mathbf{f}) \mapsto \nu(\beta)], \nu) \mid \nu : \{\alpha, \beta, \ldots\} \to \mathbb{N}\}$$

 \triangleright ν : bridge between memory and values

Separation

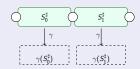
- A graph = a set of edges
- Denotes the separating conjunction of the edges

Empty graph emp

 $\gamma_{\mathrm{S}}(\mathsf{emp}) = \{(\emptyset, \nu) \mid \nu : \mathsf{nodes} \to \mathbb{V}\} \text{ i.e., empty store}$

Separating conjunction

$$\gamma_{\mathrm{S}}(S_{0}^{\sharp} * S_{1}^{\sharp}) \ = \ \{(\mathit{h}_{0} \circledast \mathit{h}_{1}, \nu) \mid (\mathit{h}_{0}, \nu) \in \gamma_{\mathrm{S}}(S_{0}^{\sharp}) \land (\mathit{h}_{1}, \nu) \in \gamma_{\mathrm{S}}(S_{1}^{\sharp})\}$$



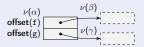
Separation example

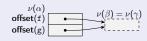
Field splitting model

- Separation impacts edges / fields, not pointers
- Shape graph



accounts for both abstract states below:





In other words, separation

- asserts addresses are distinct
- says nothing about contents

Inductive edges

List definition

$$\begin{array}{ll} \alpha \cdot \mathsf{list} & ::= & (\mathsf{emp}, \alpha = \mathsf{0}) \\ & | & (\alpha \cdot \mathsf{next} \mapsto \beta_0 * \alpha \cdot \mathsf{data} \mapsto \beta_1 * \beta_0 \cdot \mathsf{list}, \alpha \neq \mathsf{0}) \end{array}$$

where emp denotes the empty heap

Concretization as a least fixpoint

Given an inductive def ι

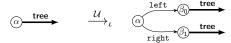
$$\gamma_{S}(\alpha \cdot \iota) = \bigcup \left\{ \gamma_{S}(F) \mid \alpha \cdot \iota \xrightarrow{\mathcal{U}} F \right\}$$

• Alternate approach:

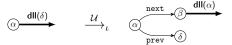
index inductive applications with induction depth allows to reason on length of structures

Inductive structures IV: a few instances

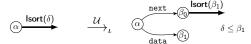
More complex shapes: trees



Relations among pointers: doubly-linked lists

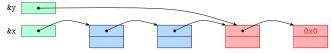


Relations between pointers and numerical: sorted lists



Inductive segments

A frequent pattern:



Could be expressed directly as an inductive with a parameter:

$$\begin{array}{ll} \alpha \cdot \mathsf{list_endp}(\pi) & ::= & (\mathsf{emp}, \alpha = \pi) \\ & | & (\alpha \cdot \mathsf{next} \mapsto \beta_0 * \alpha \cdot \mathsf{data} \mapsto \beta_1 \\ & * \beta_0 \cdot \mathsf{list_endp}(\pi), \alpha \neq 0) \end{array}$$

 This definition would derive from list
 Thus, we make segments part of the fundamental predicates of the domain



Multi-segments: possible, but harder for analysis

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Example

How to express both shape and numerical properties ?

List of even elements:



Sorted list:



- Many other examples:
 - ► list of open filed descriptors
 - tries
 - balanced trees: red-black, AVL...
- Note: inductive definitions also talk about data

A first approach to domain combination

Basis

 \bullet Graphs form a shape domain \mathbb{D}_S^\sharp abstract stores together with a physical mapping of nodes

$$\gamma_{\mathrm{S}}: \mathbb{D}^{\sharp}_{\mathrm{S}}
ightarrow \mathcal{P}((\mathbb{D}^{\sharp}_{\mathrm{S}}
ightarrow \mathbb{M}) imes (\mathsf{nodes}
ightarrow \mathbb{V}))$$

• Numerical values are taken in a numerical domain $\mathbb{D}_{\mathrm{num}}^{\sharp}$ abstracts physical mapping of nodes

$$\gamma_{ ext{num}}: \mathbb{D}^{\sharp}_{ ext{num}} o \mathcal{P}((\mathsf{nodes} o \mathbb{V}))$$

Concretization of the combined domain [CR]

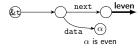
$$\gamma(S^{\sharp}, N^{\sharp}) = \{ \sigma \in \mathbb{M} \mid \exists \nu \in \gamma_{\text{num}}(N^{\sharp}), \ (\sigma, \nu) \in \gamma_{\text{S}}(S^{\sharp}) \}$$

Quite similar to a reduced product

Combination by reduced product

Reduced product

- Product abstraction: $\mathbb{D}^{\sharp} = \mathbb{D}_{0}^{\sharp} \times \mathbb{D}_{1}^{\sharp}$ $\gamma(x_{0}, x_{1}) = \gamma(x_{0}) \cap \gamma(x_{1})$
- Reduction: \mathbb{D}_r^{\sharp} is the quotient of \mathbb{D}^{\sharp} by the equivalence relation \equiv defined by $(x_0, x_1) \equiv (x_0', x_1') \iff \gamma(x_0, x_1) = \gamma(x_0', x_1')$
- Domain operations (join, transfer functions) are pairwise (are usually composed with reduction)
- Why not to use a product of the shape domain with a numerical domain?
- How to compare / join the following two elements ?



and



Towards a more adapted combination operator

Why does this fail here?

- The set of nodes / symbolic variables is not fixed
- Variables represented in the numerical domain depend on the shape abstraction
- ⇒ Thus the product is **not** symmetric

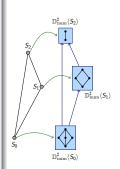
Intuitions

- ullet Graphs form a shape domain \mathbb{D}_{S}^{\sharp}
- ullet For each graph $S^\sharp\in\mathbb{D}_{\mathrm{S}}^\sharp$, we have a numerical lattice $\mathbb{D}_{\mathrm{num}(S^\sharp)}^\sharp$
 - example: if graph S^{\sharp} contains nodes $\alpha_0, \alpha_1, \alpha_2, \mathbb{D}^{\sharp}_{\mathbf{num}\langle S^{\sharp}\rangle}$ should abstract $\{\alpha_0, \alpha_1, \alpha_2\} \to \mathbb{V}$
- An abstract value is a pair (S^{\sharp}, N^{\sharp}) , such that $N^{\sharp} \in \mathbb{D}^{\sharp}_{\mathbf{num}(N^{\sharp})}$

Cofibered domain

Definition [AV]

- Basis: abstract domain $(\mathbb{D}_0^{\sharp},\sqsubseteq_0^{\sharp})$, with concretization $\gamma_0:\mathbb{D}_0^{\sharp}\to\mathbb{D}$
- Function: $\phi: \mathbb{D}_0^\sharp \to \mathcal{D}_1$, where each element of \mathcal{D}_1 is an abstract domain $(\mathbb{D}_1^\sharp, \sqsubseteq_1^\sharp)$, with a concretization $\gamma_{\mathbb{D}_1^\sharp}: \mathbb{D}_1^\sharp \to \mathbb{D}$
- Lift functions: $\forall x^{\sharp}, y^{\sharp} \in \mathbb{D}_{0}^{\sharp}$, such that $x^{\sharp} \sqsubseteq_{0}^{\sharp} y^{\sharp}$, there exists a function $\Pi_{x^{\sharp}, y^{\sharp}} : \phi(x^{\sharp}) \to \phi(y^{\sharp})$, that is monotone for $\gamma_{x^{\sharp}}$ and $\gamma_{y^{\sharp}}$
- Domain: \mathbb{D}^{\sharp} is the set of pairs $(x_0^{\sharp}, x_1^{\sharp})$ where $x_1^{\sharp} \in \phi(x_0^{\sharp})$



- Generic product, where the second lattice depends on the first
- Provides a generic scheme for widening, comparison

Domain operations

Lift functions allow to switch domain when needed

Comparison of $(x_0^{\sharp}, x_1^{\sharp})$ and $(y_0^{\sharp}, y_1^{\sharp})$

- First, compare x_0^{\sharp} and y_0^{\sharp} in \mathbb{D}_0^{\sharp}
- 2 If $x_0^{\sharp} \sqsubseteq_0^{\sharp} y_0^{\sharp}$, compare $\Pi_{x_0^{\sharp}, y_0^{\sharp}}(x_1^{\sharp})$ and y_1^{\sharp}

Widening of $(x_0^{\sharp}, x_1^{\sharp})$ and $(y_0^{\sharp}, y_1^{\sharp})$

- f 0 First, compute the widening in the basis $z_0^\sharp = x_0^\sharp riangle y_0^\sharp$
- Then move to $\phi(z_0^\sharp)$, by computing $x_2^\sharp = \Pi_{x_0^\sharp, z_0^\sharp}(x_1^\sharp)$ and $y_2^\sharp = \Pi_{v_0^\sharp, z_0^\sharp}(y_1^\sharp)$
- **3** Last widen in $\phi(z_0^{\sharp})$: $z_1^{\sharp} = x_2^{\sharp} \nabla_{z_0^{\sharp}} y_2^{\sharp}$

 $(x_0^{\sharp}, x_1^{\sharp}) \nabla (y_0^{\sharp}, y_1^{\sharp}) = (z_0^{\sharp}, z_1^{\sharp})$

Domain operations

Transfer functions, e.g., assignment

- Require memory location be materialized in the graph
 - i.e., the graph may have to be modified
 - the numerical component should be updated with lift functions
- Require update in the graph and the numerical domain
 - i.e., the numerical component should be kept coherent with the graph

Proofs of soundness of transfer functions rely on:

- the soundness of the lift functions
- the soundness of both domain transfer functions

Outline

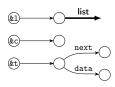
- An introduction to separation logic
- A shape abstract domain relying on separation
- 3 Combination with a numerical domain
- 4 Standard static analysis algorithms
 - Overview of the analysis
 - Post-conditions and unfolding
 - Folding: widening and inclusion checking
- 5 Inference of inductive definitions / call-stack summarization
- 6 Conclusion

Static analysis overview

A list insertion function:

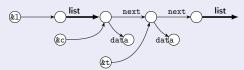
```
\label{list to a list} \begin{split} & \text{list} \star 1 \text{ assumed to point to a list} \\ & \text{list} \star t \text{ assumed to point to a list element} \\ & \text{list} \star c = 1; \\ & \text{while}(c \mathrel{!=} \texttt{NULL} \&\& c \rightarrow \texttt{next} \mathrel{!=} \texttt{NULL} \&\& (\ldots)) \{ \\ & c = c \rightarrow \texttt{next}; \\ & t \rightarrow \texttt{next} = c \rightarrow \texttt{next}; \\ & c \rightarrow \texttt{next} = t; \end{split}
```

- list inductive structure def.
- Abstract precondition:



Result of the (interprocedural) analysis

• Over-approximations of reachable concrete states e.g., after the insertion:



Transfer functions

Abstract interpreter design

- Follows the semantics of the language under consideration
- The abstract domain should provide sound transfer functions

Transfer functions

- Assignment: $x \to f = y \to g$ or $x \to f = e_{arith}$
- Test: analysis of conditions (if, while)
- Variable creation and removal
- Memory management: malloc, free

Should be sound i.e., not forget any concrete behavior

Abstract operators

- Join and widening: over-approximation
- Inclusion checking: check stabilization of abstract iterates

 Xavier Rival (INRIA) Shape analysis based on separation logic Dec. 17th, 2014

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Abstract operations

Denotational style abstract interpreter

- Concrete denotational semantics $\llbracket p \rrbracket : s \mapsto \mathcal{P}(s')$
- Abstract semantics $[p]^{\sharp}(S) = S'$, computed by the analysis:

$$s \in \gamma(\mathsf{S}) \Longrightarrow \llbracket p
rbracket(s) \subseteq \gamma(\llbracket p
rbracket^\sharp(\mathsf{S}))$$

Analysis by induction on the syntax using domain operators

The algorithms underlying the transfer functions

Unfolding: cases analysis on summaries

Abstract postconditions, on "exact" regions, e.g. insertion

• Widening: builds summaries and ensures termination

Outline

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Analysis of an assignment in the graph domain

Steps for analyzing $x = y \rightarrow next$ (local reasoning)

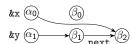
- **1** Evaluate I-value x into points-to edge $\alpha \mapsto \beta$
- 2 Evaluate r-value y -> next into node β'
- **3** Replace points-to edge $\alpha \mapsto \beta$ with points-to edge $\alpha \mapsto \beta'$

With pre-condition:

$$\&x @_0 \longrightarrow @_0$$

$$\&y @_1 \longrightarrow @_1 \xrightarrow{next} @_2$$

- Step 1 produces $\alpha_0 \mapsto \beta_0$
- Step 2 produces β_2
- End result:



With pre-condition:

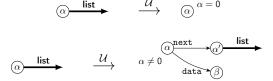
&x
$$\textcircled{00}$$
 $\textcircled{00}$ $\textcircled{00}$ $\textcircled{00}$ $\textcircled{00}$ $\textcircled{00}$ $\textcircled{00}$

- Step 1 produces $\alpha_0 \mapsto \beta_0$
- Step 2 fails
- Abstract state too abstract
- We need to refine it

Unfolding as a local case analysis

Unfolding principle

- Case analysis, based on the inductive definition
- Generates symbolic disjunctions analysis performed in a disjunction domain
- Example, for lists:



• Numeric predicates: approximated in the numerical domain

Soundness: by definition of the concretization of inductive structures

$$\gamma_{\mathrm{S}}(S^{\sharp}) \subseteq \bigcup \{\gamma_{\mathrm{S}}(S_{0}^{\sharp}) \mid S^{\sharp} \stackrel{\mathcal{U}}{\longrightarrow} S_{0}^{\sharp}\}$$

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Analysis of an assignment, with unfolding

Principle

- We have $\gamma_S(\alpha \cdot \iota) = \bigcup \{\gamma_S(S^{\sharp}) \mid \alpha \cdot \iota \xrightarrow{\mathcal{U}} S^{\sharp}\}$
- ullet Replace $lpha \cdot \iota$ with a finite number of disjuncts and continue

Disjunct 1:

- Step 1 produces $\alpha_0 \mapsto \beta_0$
- Step 2 fails:

Null pointer dereference!

Disjunct 2:



- Step 1 produces $\alpha_0 \mapsto \beta_0$
- Step 2 produces β_2
- End result:



Unfolding and degenerated cases

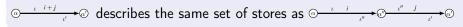
$$\begin{array}{l} \textbf{assume}(1 \text{ points to a dll}) \\ c=1; \\ \textcircled{0 while}(c \neq \texttt{NULL \&\& condition}) \\ c=c-\texttt{next}; \\ \textcircled{2 if}(c \neq 0 \&\& c-\texttt{prev} \neq 0) \\ c=c-\texttt{prev} \rightarrow \texttt{prev}; \end{array}$$

• at ①:
$$\textcircled{\tiny 000} \xrightarrow{\text{dll}(\delta_1)}$$
• at ②: $\textcircled{\tiny 000} \xrightarrow{\text{dll}(\delta_0)} \textcircled{\tiny 000} \xrightarrow{\text{dll}(\delta_1)} \textcircled{\tiny 000}$
 \Rightarrow non trivial unfolding

• Materialization of c -> prev:



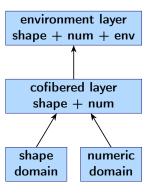
Segment splitting lemma: basis for segment unfolding

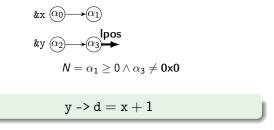


Materialization of c -> prev -> prev:

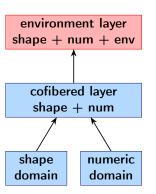


Implementation issue: discover which inductive edge to unfold





Abstract post-condition?

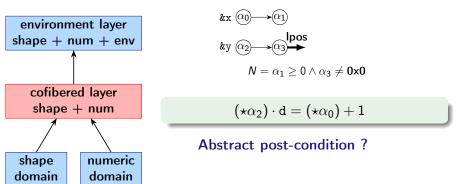


&x
$$(\alpha_0)$$
 (α_1)
&y (α_2) (α_3)
 $N = \alpha_1 \ge 0 \land \alpha_3 \ne 0 \times 0$
 $y \rightarrow d = x + 1 \implies (\star \alpha_2) \cdot d = (\star \alpha_0) + 1$

Abstract post-condition?

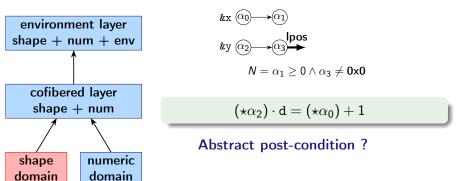
Stage 1: environment resolution

• replaces x with $\star e^{\sharp}(x)$



Stage 2: propagate into the shape + numerics domain

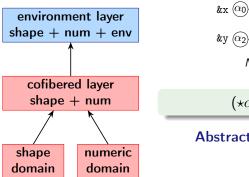
only symbolic nodes appear



Stage 3: resolve cells in the shape graph abstract domain

- $\star \alpha_0$ evaluates to α_1 ; $\star \alpha_2$ evaluates to α_3
- $(\star \alpha_2) \cdot d$ fails to evaluate: no points-to out of α_3

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&x
$$\alpha_0$$
 α_1 α_4

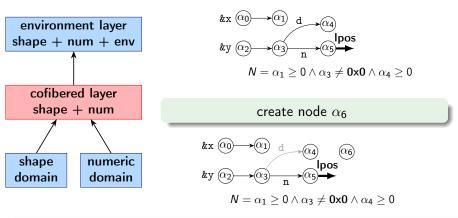
&y α_2 α_3 α_4

N = $\alpha_1 \ge 0 \land \alpha_3 \ne 0$ x0 $\land \alpha_4 \ge 0$
 $(\star \alpha_2) \cdot d = (\star \alpha_0) + 1$

Abstract post-condition?

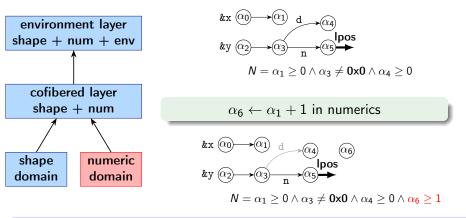
Stage 4: unfolding (several steps, skipped here)

- locally materialize $\alpha_3 \cdot lpos$; update keys / relations in the numerics
- I-value $\alpha_3 \cdot d$ now evaluates into edge $\alpha_3 \cdot d \mapsto \alpha_4$



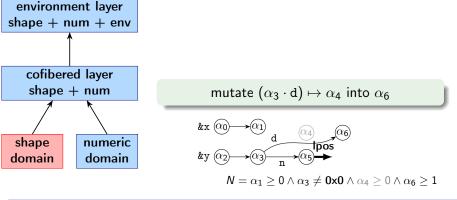
Stage 5: create a new node

• new node α_6 denotes a new value will store the new value



Stage 6: perform numeric assignment

 numeric assignment completely ignores pointer structures to the new node



Stage 7: perform the update in the graph

- classic strong update in a pointer aware domain
- ullet symbolic node $lpha_4$ becomes redundant and can be removed

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Need for a folding operation

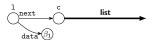
Back to the list traversal example...

```
\label{eq:continuity} \begin{array}{l} \textbf{assume}(1 \text{ points to a list}) \\ \textbf{c} = \textbf{1}; \\ \textbf{while}(\textbf{c} \neq \texttt{NULL}) \{ \\ \textbf{c} = \textbf{c} \rightarrow \texttt{next}; \\ \} \end{array}
```

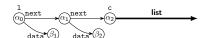
- First iterates in the loop:
 - at iteration 0 (before entering the loop):



at iteration 1:



at iteration 2:



- How to guarantee termination of the analysis?
- How to introduce segment edges / perform abstraction ?

Widening

- The lattice of shape abstract values has infinite height
- Thus iteration sequences may not terminate

Definition of a widening operator ∇

Over-approximates join:

$$\left\{ \begin{array}{l} X^{\sharp} & \subseteq & \gamma(X^{\sharp} \triangledown Y^{\sharp}) \\ Y^{\sharp} & \subseteq & \gamma(X^{\sharp} \triangledown Y^{\sharp}) \end{array} \right.$$

• Enforces termination: for all sequence $(X_n^{\sharp})_{n \in \mathbb{N}}$, the sequence $(Y_n^{\sharp})_{n \in \mathbb{N}}$ defined below is ultimately stationary

$$\left\{ \begin{array}{ccc} Y_0^{\sharp} & = & X_0^{\sharp} \\ \forall n \in \mathbb{N}, & Y_{n+1}^{\sharp} & = & Y_n^{\sharp} \nabla X_{n+1}^{\sharp} \end{array} \right.$$

Canonicalization

Upper closure operator

 $\rho: \mathbb{D}^{\sharp} \longrightarrow \mathbb{D}^{\sharp}_{\operatorname{can}} \subseteq \mathbb{D}^{\sharp}$ is an **upper closure operator** (uco) iff it is monotone, extensive and idempotent.

Canonicalization

- Disjunctive completion: $\mathbb{D}^{\sharp}_{\vee}$ = finite disjunctions over \mathbb{D}^{\sharp}
- Canonicalization operator ρ_{\vee} defined by $\rho_{\vee}: \mathbb{D}^{\sharp}_{\vee} \longrightarrow \mathbb{D}^{\sharp}_{\operatorname{can}^{\vee}}$ and $\rho_{\vee}(X^{\sharp}) = \{\rho(x^{\sharp}) \mid x^{\sharp} \in X^{\sharp}\}$ where ρ is an uco and $\mathbb{D}^{\sharp}_{\operatorname{can}}$ has finite height
- Usually more simple to compute
- Canonicalization is used in many shape analysis tools:
 TVLA, most separation logic based analysis tools
- However less powerful than widening: does not exploit history of computation

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Per region weakening

The weakening principles shown in the following apply both in canonicalization and widening approaches

We can apply the local reasoning principle to weakening

- inclusion test (comparison)
- canonicalization
- join / widening

Application: inclusion test

- Operator \sqsubseteq^{\sharp} should satisfy $X^{\sharp} \sqsubseteq^{\sharp} Y^{\sharp} \Longrightarrow \gamma(X^{\sharp}) \subseteq \gamma(Y^{\sharp})$
- If $S_0^{\sharp} \sqsubseteq^{\sharp} S_{0,\text{weak}}^{\sharp}$ and $S_1^{\sharp} \sqsubseteq^{\sharp} S_{1,\text{weak}}^{\sharp}$







Inductive weakening

Weakening identity

- X[‡]□[‡]X[‡]...
- Trivial, but useful, when a graph region appears in both widening arguments

Weakening unfolded region

- If $S_0^{\sharp} \xrightarrow{\mathcal{U}} S_1^{\sharp}$, $\gamma_{\mathrm{S}}(S_1^{\sharp}) \subseteq \gamma_{\mathrm{S}}(S_0^{\sharp})$
- Soundness follows the the soundness of unfolding
- Application to a simple example:



Comparison operator in the shape domain

Algorithm structure

Based on separation and local reasoning:

$$\gamma_{\mathrm{S}}(S_0^\sharp) \subseteq \gamma_{\mathrm{S}}(S_1^\sharp) \Longrightarrow \gamma_{\mathrm{S}}(S_0^\sharp * S^\sharp) \subseteq \gamma_{\mathrm{S}}(S_1^\sharp * S^\sharp)$$

- Algorithm:
 - applies local rules and "consumes" graph regions proved weaker
 - keeps discovering new rule applications
- Structural rules such as:
 - segment splitting:

$$S^{\sharp} \sqsubseteq^{\sharp} \textcircled{\alpha} \xrightarrow{\iota} \implies S^{\sharp} * \textcircled{\beta} \xrightarrow{\iota} \textcircled{\alpha} \sqsubseteq^{\sharp} \textcircled{\beta} \xrightarrow{\iota}$$

Correctness:

$$S_0^{\sharp} \sqsubseteq^{\sharp} S_1^{\sharp} \implies \gamma_{\mathrm{S}}(S_0^{\sharp}) \subseteq \gamma_{\mathrm{S}}(S_1^{\sharp})$$

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Comparison operator in the combined domain

We need to tackle the fact nodes names may differ (cofibered domain)



Instrumented comparison in the shape domain

- Comparison $S_0^{\sharp} \sqsubseteq^{\sharp} S_1^{\sharp}$: rules should compute a physical mapping $\Psi : \mathbf{nodes}_1 \longrightarrow \mathbf{nodes}_0$
- Soundness condition: $(\sigma, \nu) \in \gamma_S(S_0^\sharp) \Longrightarrow (\sigma, \nu \circ \Psi) \in \gamma_S(S_0^\sharp)$

Comparison in the cofibered domain

- Lift function for numerical abstract values: $\Pi_{S_0^{\sharp},S_1^{\sharp}}(N_0^{\sharp}) = N_0^{\sharp} \circ \Psi$
- Thus, we simply need to compare $N_0^{\sharp} \circ \Psi$ and N_1^{\sharp}

Xavier Rival (INRIA) Shape analysis based on separation logic

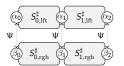
Join operator

- Similar iterative scheme, based on local rules
- But needs to reason locally on two graphs in the same time: each rule maps two regions into a common over-approximation

Graph partitioning and mapping

- Inputs: $S_0^{\sharp}, S_1^{\sharp}$
- Performed by a function Ψ : nodes₀ × nodes₁ \rightarrow nodes₁
- Ψ is computed at the same time as the join

If
$$\forall i \in \{0,1\}, \ \forall s \in \{\text{lft,rgh}\}, \ S_{i,s}^{\sharp} \sqsubseteq^{\sharp} S_{s}^{\sharp},$$







Segment introduction

Rule



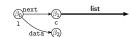
$$\mathsf{then} \quad \left\{ \begin{array}{l} S_{\mathrm{lft}}^{\sharp} \triangledown S_{\mathrm{rgh}}^{\sharp} \ = \ \textcircled{\tiny{0}} \qquad \qquad \stackrel{\iota}{\longleftarrow} \ \textcircled{\tiny{0}} \qquad \qquad \\ (\alpha,\beta_0) \overset{\Psi}{\longleftrightarrow} \gamma_0 \\ (\alpha,\beta_1) \overset{\Psi}{\longleftrightarrow} \gamma_1 \end{array} \right.$$

Application to list traversal, at the end of iteration 1:

• before iteration 0:



• end of iteration 0:



join, before iteration 1:

 $\begin{cases}
\Psi(\alpha_0, \beta_0) = \gamma_0 \\
\Psi(\alpha_0, \beta_1) = \gamma_1
\end{cases}$

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Shape analysis based on separation logic

Dec. 17th, 2014

Segment extension

Rule

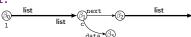


Application to list traversal, at the end of iteration 1:

previous invariant before iteration 1:

$$\begin{array}{c|c}
 & \text{list} & \text{list} \\
\hline
1 & \text{list} & c
\end{array}$$

• end of iteration 1:



join, before iteration 1:

$$\begin{cases}
\Psi(\alpha_0, \beta_0) = \gamma_0 \\
\Psi(\alpha_1, \beta_2) = \gamma_1
\end{cases}$$

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Rewrite system properties

- Comparison, canonicalization and widening algorithms can be considered rewriting systems over tuples of graphs
- Each step applies a rule / computation step

Termination

- The systems are terminating
- This ensures comparison, canonicalization, widening are computable

Non confluence!

- The results depends on the order of application of the rules
- Implementation requires the choice of an adequate strategy

Properties

Inclusion checking is sound

If
$$S_0^{\sharp} \sqsubseteq^{\sharp} S_1^{\sharp}$$
, then $\gamma(S_0^{\sharp}) \subseteq \gamma(S_1^{\sharp})$

Canonicalization is sound

$$\gamma(S^{\sharp}) \subseteq \gamma(
ho_{\mathsf{can}}(S^{\sharp}))$$

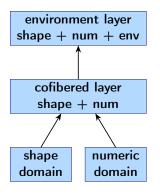
Widening is sound and terminating

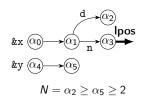
$$\gamma(S_0^{\sharp}) \subseteq \gamma(S_0^{\sharp} \triangledown S_1^{\sharp})$$
$$\gamma(S_1^{\sharp}) \subseteq \gamma(S_0^{\sharp} \triangledown S_1^{\sharp})$$

∇ ensures termination of abstract iterates

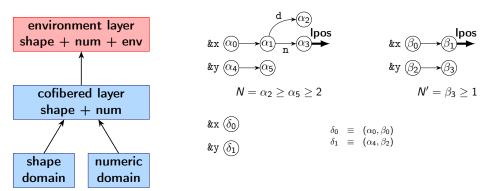
- Soundness of local reasoning and of local rules
- Termination of widening: ∇ can introduce only segments, and may not introduce infintely many of them

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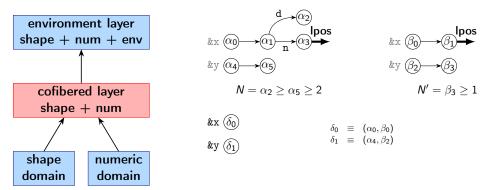






Stage 1: abstract environment

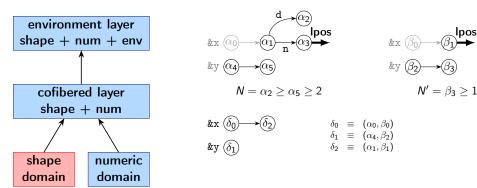
• compute new abstract environment and initial node relation e.g., α_0 , β_0 both denote &x



Stage 2: join in the "cofibered" layer

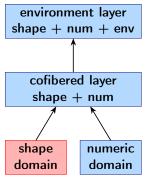
operations to perform:

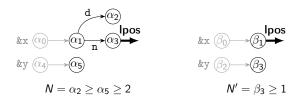
- 1 compute the join in the graph
- convert value abstractions, and join the resulting lattice



Stage 2: graph join

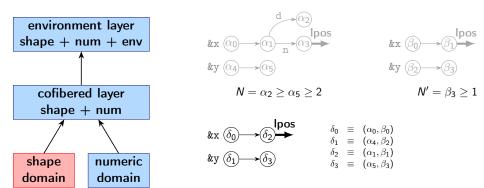
- apply local join rules
 ex: points-to matching, weakening to inductive...
- incremental algorithm





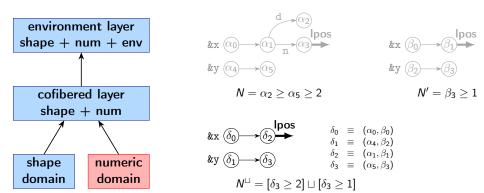
Stage 2: graph join

- apply local join rules
 ex: points-to matching, weakening to inductive...
- incremental algorithm



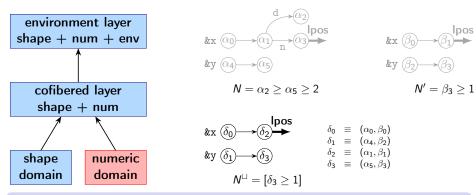
Stage 2: graph join

- apply local join rules
 ex: points-to matching, weakening to inductive...
- incremental algorithm



Stage 3: conversion function application in numerics

- remove nodes that were abstracted away
- rename other nodes



Stage 4: join in the numeric domain

• apply ⊔ for regular join, ∇ for a widening

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Interprocedural analysis

- Analysis of programs that consist in several functions (or procedures)
- Difficulty: how to cope with multiple (possibly recursive) calls

Relational approach

- analyze each function once
- compute function summaries abstraction of input-output relations
- analysis of a function call: instantiate the function summary (hard)

Inlining approach

- inline functions at function calls
- just an extension of intraprocedural analysis

- In this section, we study the inlining approach for recursion
- Side result: a widening for inductive definitions

Approaches to interprocedural analysis

"relational" approach

"inlining" approach

analyze each definition abstracts $\mathcal{P}(\bar{\mathbb{S}} \to \bar{\mathbb{S}})$

- + modularity
- + reuse of invariants
- deals with state relations
 - complex higher order iteration strategy

challenge: frame problem

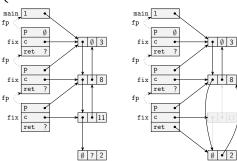
analyze each call abstracts $\mathcal{P}(\mathbb{S})$

- not modular
- re-analysis in ≠ contexts
 - + deals with states
- + straightforward iteration

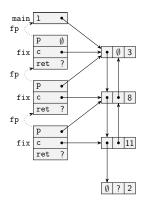
challenge: unbounded calls

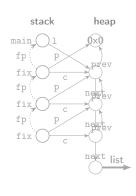
Challenges in interprocedural analysis

turns a linked list into a doubly linked list removes some elements

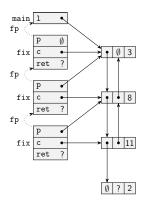


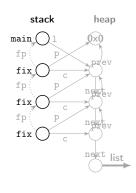
- Heap is unbounded, needs abstraction (shape analysis)
- But stack may also grow unbounded, needs abstraction
- Complex relations between both stack and heap



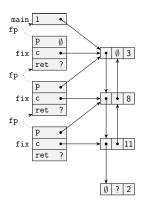


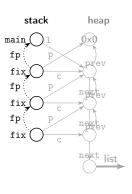
- Concrete assembly call stack modelled in a separating shape graph together with the heap
 - one node per activation record address



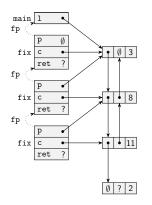


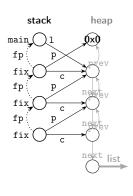
- Concrete assembly call stack modelled in a separating shape graph together with the heap
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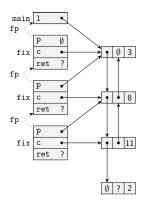


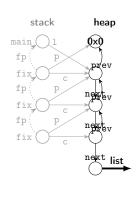
- Concrete assembly call stack modelled in a separating shape graph together with the heap
 - one node per activation record address
 - explicit edges for frame pointers





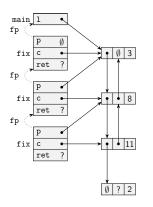
- Concrete assembly call stack modelled in a separating shape graph together with the heap
 - one node per activation record address
 - explicit edges for frame pointers
 - ▶ local variables turn into activation record fields

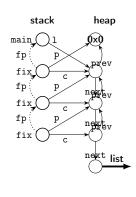




- Concrete assembly call stack modelled in a separating shape graph together with the heap
 - one node per activation record address
 - explicit edges for frame pointers
 - ▶ local variables turn into activation record fields

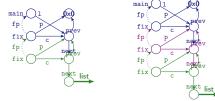
Xavier Rival (INRIA)





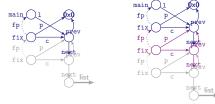
- Concrete assembly call stack modelled in a separating shape graph together with the heap
 - one node per activation record address
 - explicit edges for frame pointers
 - ► local variables turn into activation record fields

Second and third iterates: a repeating pattern



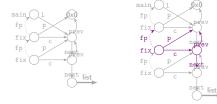
• Computing an inductive rule for summarization: subtraction

Second and third iterates: a repeating pattern



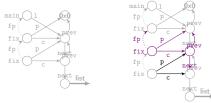
- Computing an inductive rule for summarization: subtraction
 - subtract top-most activation record

Second and third iterates: a repeating pattern



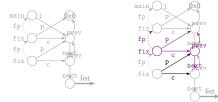
- Computing an inductive rule for summarization: subtraction
 - subtract top-most activation record
 - subtract common stack region

Second and third iterates: a repeating pattern



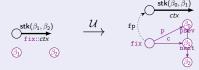
- Computing an inductive rule for summarization: subtraction
 - subtract top-most activation record
 - subtract common stack region
 - ▶ gather relations with next activation records: additional parameters
 - collect numerical constraints

Second and third iterates: a repeating pattern



• Computing an inductive rule for summarization: subtraction

Inferred inductive rule



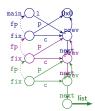
Inference of a call-stack summary: widening iterates

Fixpoint at function entry:

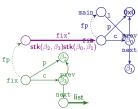
first iterate:



second iterate:



widened iterate:



Fixpoint reached

- Fixpoint upon function return:
 - function return involves unfolding of stack summaries
 - ▶ simpler widening sequence: no rule to infer

Widening over inductive definitions

Domain structure

An abstract value should comprise:

- a set S of unfolding rules for the stack inductive
- a shape graph G
- a numeric abstract value N

Shape graph G is defined in a lattice specified by S, thus, this is an instance of the **cofibered abstraction**

- Lift functions are trivial:
 - adding rules simply weakens shape graphs
 - i.e., no need to change them syntactically, their concretization just gets weaker
- Termination in the lattice of rules:
 limiting of the number of rules that can be generated to some given bound

Outline

- An introduction to separation logic
- 2 A shape abstract domain relying on separation
- 3 Combination with a numerical domain
- 4 Standard static analysis algorithms
- 5 Inference of inductive definitions / call-stack summarization
- 6 Conclusion

Abstraction choices

Many families of symbolic abstractions including TVLA and separation logic based approaches

Variants: region logic, ownership, fractional permissions

Common ingredients

- Splitting of the heap in regions
 - TVLA: covering, via embedding
 - ► Separation logic: partitioning, enforced at the concrete level
- Use of induction in order to summarize large regions
- More limited pointer analyses: no inductives, no summarization...

Algorithms

Rather different process, compared to numerical domains

From abstract to concrete (locally)

- Unfold abstract properties in a local maner
- Allows quasi-exact analysis of usual operations (assignment, condition test...)

From concrete to abstract (globally)

- Guarantees termination
- Allows to infer pieces of code build complex structures
- Intuition:
 - static analysis involves post-fixpoint computations (over program traces)
 - widening produces a fixpoint over memory cells

Open problems

Many opportunities for research:

- Improving expressiveness
 e.g., sharing in data-structures
 - new abstractions
 - combining several abstractions into more powerful ones
- Improving scalability
 - shape analyses remain expensive analyses, with few "cheap" and useful abstractions
 - cut down the cost of complex algorithms
 - ▶ isolate smaller families of predicates
- Applications, beyond software safety:
 - security
 - verification of functional properties

Internships

Several topics possible, soon to be announced on the lecture webpage:

Internal reduction operator

- inductive definitions are very expressive thus tricky to reason about
- design of an internal reduction operator on abstract elements with inductive definitions

Modular inter-procedural analysis

- a relation between pre-conditions and post-conditions can be formalized in a single shape graph
- design of an abstract domain for relations between states

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