

# Construction of abstract domains for heterogeneous memory properties

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# Memory abstraction difficulties

## Memory abstraction

- **Abstraction** of  $\mathcal{P}(\text{Environment} \times \text{Stack} \times \text{Heap})$
- **Operations to analyze**: C statements including **pointer arithmetic**, **memory management**...
- **Operations to verify**: **memory safety**, **structure preservation**...

## Difficulties:

- 1 **Complex concrete semantics**: memory states, pointers...
- 2 **Huge variety of properties to abstract**:
  - ▶ **linked structures**: lists, trees...
  - ▶ **arrays, buffers, strings**...
  - ▶ structures involving **relations** (sorted lists, balanced trees) or **sharing**
  - ▶ **combination** of numerical and memory properties
- 3 **Expensive analysis algorithms** to infer such abstractions

## Existing approaches

### Many existing analyses for specific structures:

- **Inductive structures** (lists, trees, ASTs...):  
TVLA, separation logic based static analyses...
- **Complex aggregates** (unions and structs):  
precise **abstractions of blocks**, account for alignment and sizes
- **Arrays**:  
**smashing** based abstractions,  
**partitioning** based abstractions (array split into several regions)
- **Buffers** and **strings**:  
abstractions of **size** and **positions of zeros**

**But abstracting stores with heterogeneous structures remains hard**

# Main ideas of the MemCAD project

Towards a library of abstract domains

## Common abstract domain interfaces:

- Concretization to the **same domain**
- **Similar** transfer functions / lattice operations

## A few basic abstract domains (implemented as **ML modules**):

- **Lightweight** abstraction of **fixed size blocks**
- **Parametric** abstraction for **inductive structures...**

## Combination operations (implemented as **ML functors**):

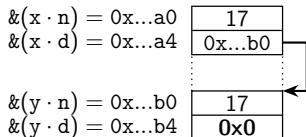
- **Product** with a **value** abstraction (e.g., numeric)
- **Separating product** of memory abstraction
- **Reduced, non separating product** of memory abstraction...

- Combinations of abstract domains can be tuned for precision/efficiency
- Abstract domains are not analyzer specific

# A generic memory abstraction step by step

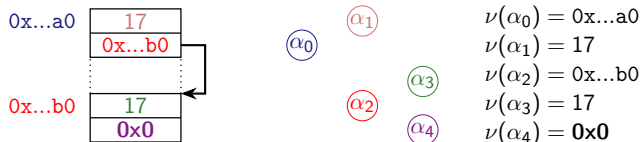
- Concrete memory states

- ▶ very **low level** description
- ▶ pointers, numeric values:  
**raw sequences of bits**



## A generic memory abstraction step by step

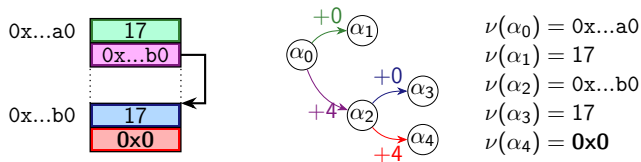
- Concrete memory states
- Abstraction of values into symbolic variables (nodes)



- ▶ characterized by **valuation**  $\nu$
- ▶  $\nu$  maps **symbolic variables** into **concrete addresses**

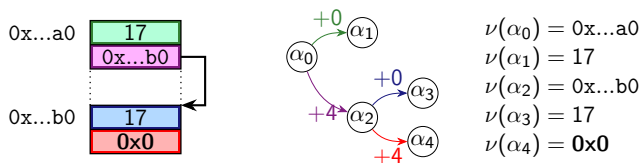
# A generic memory abstraction step by step

- Concrete memory states
- Abstraction of values into symbolic variables / nodes
- Abstraction of regions into points-to edges



## A generic memory abstraction step by step

- Concrete memory states
- Abstraction of values into symbolic variables / nodes
- Abstraction of regions into points-to edges



- Shape graph concretization

$$\gamma(G) = \{(h, \nu) \mid \dots\}$$

valuation  $\nu$  plays an important role in the combined domain

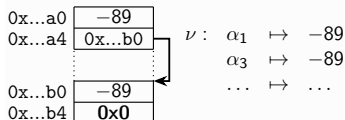
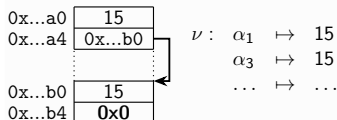
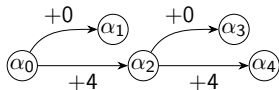


## Adding value information (here, numeric)

- Concrete numeric values appear in the valuation: **abstract  $\nu$  !**

**Example:** all lists of length 2, with equal data fields

- Memory abstraction:**



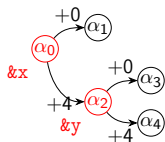
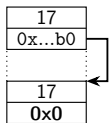
- Abstraction of valuations:**  $\nu(\alpha_1) = \nu(\alpha_3)$ , (**constraint  $\alpha_1 = \alpha_3$** )
  - combined abstraction: **product of shape and numeric** over nodes
  - $\gamma(\mathbf{G}, \mathbf{N}) = \{(h, \nu) \mid (h, \nu) \in \gamma(\mathbf{G}) \wedge \nu \in \gamma(\mathbf{N})\}$

## Adding environment abstraction

- Addresses can be read in the valuations

$$\begin{aligned} \&x &= \&(x \cdot n) = 0x\dots a0 \\ & & \&(x \cdot d) = 0x\dots a4 \end{aligned}$$

$$\begin{aligned} \&y &= \&(y \cdot n) = 0x\dots b0 \\ & & \&(y \cdot d) = 0x\dots b4 \end{aligned}$$



$$\begin{aligned} \nu : \alpha_0 &\mapsto 0x\dots a0 \\ &\alpha_2 \mapsto 0x\dots b0 \\ &\dots \mapsto \dots \end{aligned}$$

$$\begin{aligned} \mathbf{E} : x &\mapsto \alpha_0 \quad (\overset{\nu}{\mapsto} 0x\dots a0) \\ &y \mapsto \alpha_2 \quad (\overset{\nu}{\mapsto} 0x\dots b0) \end{aligned}$$

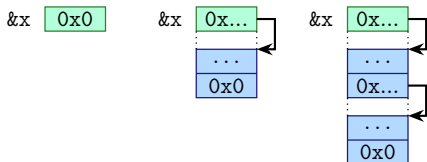
## Abstraction of environments:

- mapping  $\mathbf{E}$  from variables to symbolic nodes
- $\gamma(\mathbf{E}, \mathbf{G}, \mathbf{N}) = \{(e, h, \nu) \mid (h, \nu) \in \gamma(\mathbf{G}, \mathbf{N}) \wedge e = \nu \circ \mathbf{E}\}$

## Adding summarization

## Example:

set of all lists of any length



## List inductive predicate

$$\alpha \cdot \text{list} :=$$

$$(\text{emp} \wedge \alpha = \mathbf{0x0})$$

$$\vee (\alpha \cdot \mathbf{d} \mapsto \beta_0 * \alpha \cdot \mathbf{n} \mapsto \beta_1$$

$$* \beta_1 \cdot \text{list} \wedge \alpha \neq \mathbf{0x0})$$

well-founded predicate

## Inductive summary predicates

Concretization based on **unfolding** and **least-fixpoint**:

- $\rightsquigarrow$  replaces **an  $\alpha \cdot \text{list}$  predicate** with **one of its premises**
- $\gamma(\mathbf{G}, \mathbf{N}) = \bigcup \{ \gamma(\mathbf{G}', \mathbf{N}') \mid (\mathbf{G}, \mathbf{N}) \rightsquigarrow (\mathbf{G}', \mathbf{N}') \}$   
(all inductive predicates are required to be well-founded)

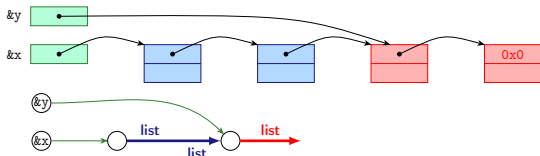
# Summarization of incomplete structures

## Structure fragments

- **Common pattern:** structure **traversal** using a **cursor**
- **Segment** abstract predicate: incomplete structure

- Example with **lists**:

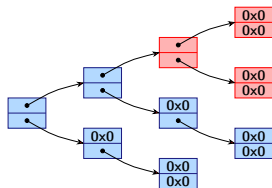
can be abstracted by:



- Works also for **trees**:



abstracts

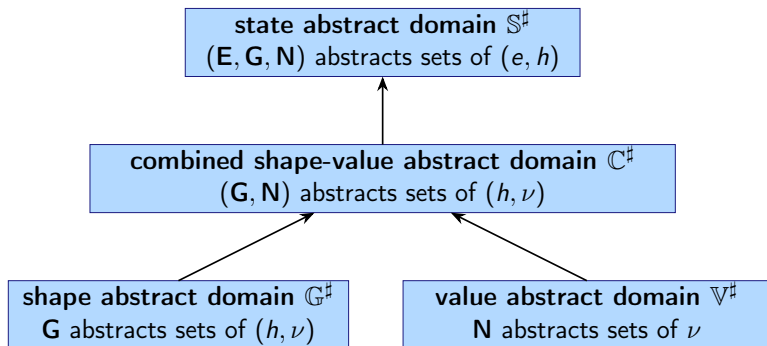


# Overall abstract domain structure

## Modular structure

- Each layer accounts for one **aspect of the concrete states**
- Each layer boils down to a **module or functor in ML**

Makes the implementation **a lot simpler**



# Outline

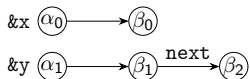
- 1 Memory abstraction
- 2 A few memory abstractions, with value information
- 3 Abstract operations**
- 4 Instantiations
- 5 Separating product of abstractions
- 6 Reduced product of abstractions
- 7 Abstraction of arrays and hierarchical abstraction
- 8 Project status

# Analysis of an assignment in the graph domain

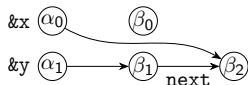
## Steps for analyzing $x = y \rightarrow \text{next}$ (local reasoning)

- 1 Evaluate **l-value**  $x$  into **points-to edge**  $\alpha \mapsto \beta$
- 2 Evaluate **r-value**  $y \rightarrow \text{next}$  into **node**  $\beta'$
- 3 Replace points-to edge  $\alpha \mapsto \beta$  with **points-to edge**  $\alpha \mapsto \beta'$

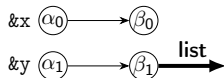
### With pre-condition:



- Step 1 produces  $\alpha_0 \mapsto \beta_0$
- Step 2 produces  $\beta_2$
- End result:



### With pre-condition:



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

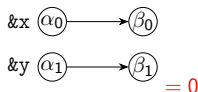
- Abstract state **too abstract**
- We need to **refine it**

# Analysis of an assignment, with unfolding

## Principle

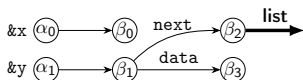
- We have  $\gamma_G(\alpha \cdot \iota) = \bigcup \{ \gamma_G(\mathbf{G}) \mid \alpha \cdot \iota \rightsquigarrow_{\text{unfold}} \mathbf{G} \}$
- Replace  $\alpha \cdot \iota$  with a finite number of disjuncts and continue

### Disjunct 1:



- Step 1 produces  $\alpha_0 \mapsto \beta_0$
- **Step 2 fails:**  
**Null pointer dereference !**

### Disjunct 2:

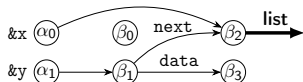


- Step 1 produces  $\alpha_0 \mapsto \beta_0$
- Step 2 produces  $\beta_2$
- **End result:**

### Case of a correct code...

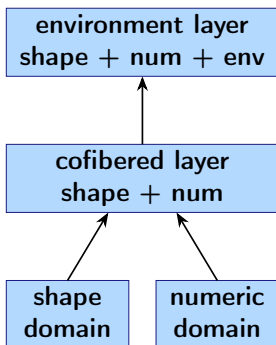
Would be ruled out by a **condition**

$y \neq 0$  i.e.,  $\beta_1 \neq 0$  in numerics





## Analysis of an assignment in the combined domain

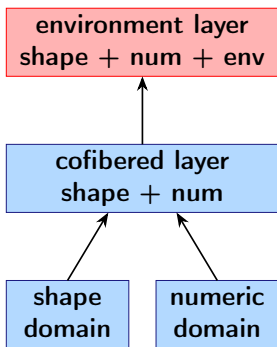

 $\&x \quad \alpha_0 \rightarrow \alpha_1$ 
 $\&y \quad \alpha_2 \rightarrow \alpha_3 \xrightarrow{\text{lpos}}$ 

$$N = \alpha_1 \geq 0 \wedge \alpha_3 \neq 0 \times 0$$

 $y \rightarrow d = x + 1$ 

Abstract post-condition ?

## Analysis of an assignment in the combined domain



$$\&x \quad \alpha_0 \rightarrow \alpha_1$$

$$\&y \quad \alpha_2 \rightarrow \alpha_3 \xrightarrow{\text{lpos}}$$

$$\mathbf{N} = \alpha_1 \geq 0 \wedge \alpha_3 \neq \mathbf{0x0}$$

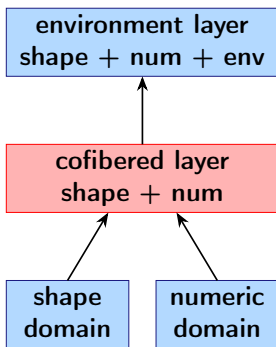
$$y \rightarrow d = x + 1 \quad \Rightarrow \quad (\star\alpha_2) \cdot d = (\star\alpha_0) + 1$$

Abstract post-condition ?

### Stage 1: environment resolution

- replaces  $x$  with  $\star\mathbf{E}(x)$

## Analysis of an assignment in the combined domain



$$\&x \quad \alpha_0 \rightarrow \alpha_1$$

$$\&y \quad \alpha_2 \rightarrow \alpha_3 \xrightarrow{\text{lpos}}$$

$$N = \alpha_1 \geq 0 \wedge \alpha_3 \neq 0x0$$

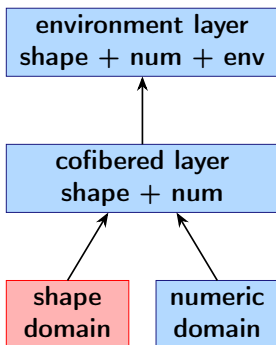
$$(*\alpha_2) \cdot d = (*\alpha_0) + 1$$

Abstract post-condition ?

Stage 2: propagate into the shape + numerics domain

- only symbolic nodes appear

## Analysis of an assignment in the combined domain



$$\&x \quad \alpha_0 \rightarrow \alpha_1$$

$$\&y \quad \alpha_2 \rightarrow \alpha_3 \xrightarrow{\text{lpos}}$$

$$\mathbf{N} = \alpha_1 \geq 0 \wedge \alpha_3 \neq \mathbf{0x0}$$

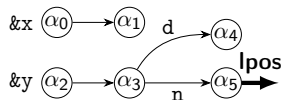
$$(\star\alpha_2) \cdot \mathbf{d} = (\star\alpha_0) + 1$$

Abstract post-condition ?

Stage 3: resolve cells in the shape graph abstract domain

- $\star\alpha_0$  evaluates to  $\alpha_1$ ;  $\star\alpha_2$  evaluates to  $\alpha_3$
- $(\star\alpha_2) \cdot \mathbf{d}$  fails to evaluate: no points-to out of  $\alpha_3$

## Analysis of an assignment in the combined domain



$$N = \alpha_1 \geq 0 \wedge \alpha_3 \neq 0 \times 0 \wedge \alpha_4 \geq 0$$

$$(*\alpha_2) \cdot d = (*\alpha_0) + 1$$

Abstract post-condition ?

environment layer  
shape + num + env

cofibered layer  
shape + num

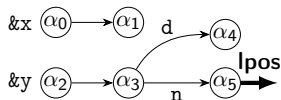
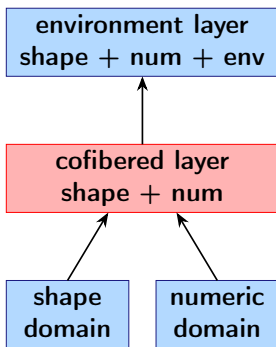
shape  
domain

numeric  
domain

Stage 4: unfolding (several steps, skipped here)

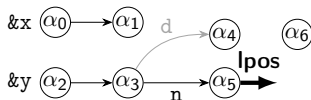
- locally materialize  $\alpha_3 \cdot \mathbf{lpos}$ ; update keys / relations in the numerics
- l-value  $(*\alpha_2) \cdot d$  **now evaluates into edge**  $\alpha_3 \rightarrow d \mapsto \alpha_4$

## Analysis of an assignment in the combined domain



$$N = \alpha_1 \geq 0 \wedge \alpha_3 \neq \mathbf{0x0} \wedge \alpha_4 \geq 0$$

create node  $\alpha_6$

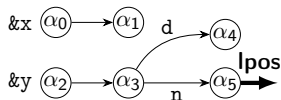
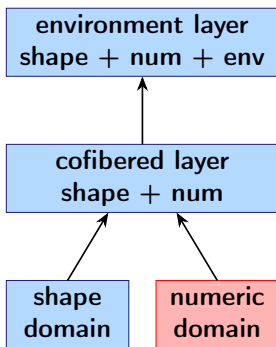


$$N = \alpha_1 \geq 0 \wedge \alpha_3 \neq \mathbf{0x0} \wedge \alpha_4 \geq 0$$

## Stage 5: create a new node

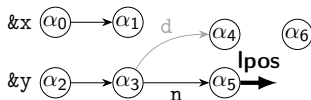
- new node  $\alpha_6$  denotes a new value  
will store the new value

## Analysis of an assignment in the combined domain



$$\mathbf{N} = \alpha_1 \geq 0 \wedge \alpha_3 \neq \mathbf{0x0} \wedge \alpha_4 \geq 0$$

$$\alpha_6 \leftarrow \alpha_1 + 1 \text{ in numerics}$$

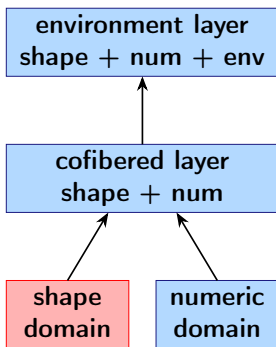


$$\mathbf{N} = \alpha_1 \geq 0 \wedge \alpha_3 \neq \mathbf{0x0} \wedge \alpha_4 \geq 0 \wedge \alpha_6 \geq 1$$

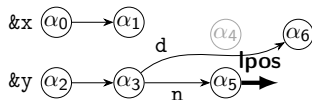
## Stage 6: perform numeric assignment

- numeric assignment **completely ignores pointer structures** to the new node

## Analysis of an assignment in the combined domain



mutate  $(\alpha_3 \cdot d) \mapsto \alpha_4$  into  $\alpha_6$



$$\mathbf{N} = \alpha_1 \geq 0 \wedge \alpha_3 \neq \mathbf{0x0} \wedge \alpha_4 \geq 0 \wedge \alpha_6 \geq 1$$

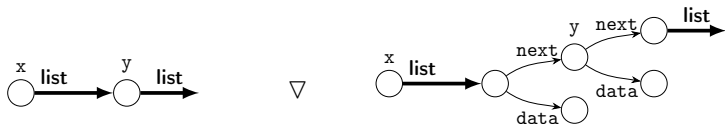
## Stage 7: perform the update in the graph

- classic **strong update** in a pointer aware domain
- symbolic node  $\alpha_4$  becomes redundant and can be removed

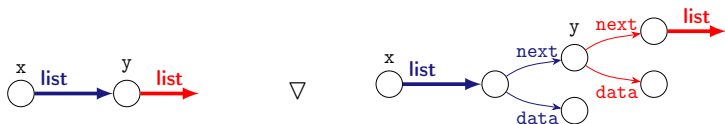


## Widening in the graph domain

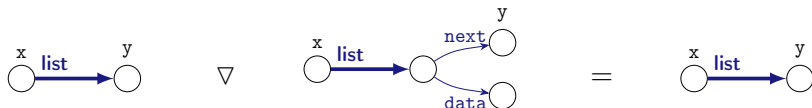
- Takes **two abstract values** as inputs



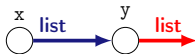
- **Region matching** (non unique choice: use of **strategies**)



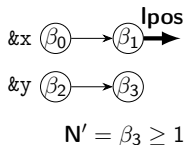
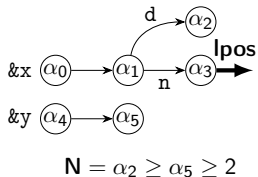
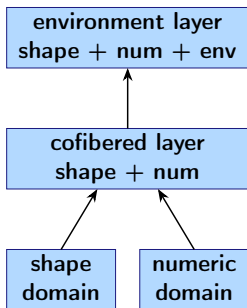
- **Semantic rules for per region weakening**



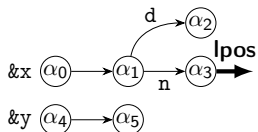
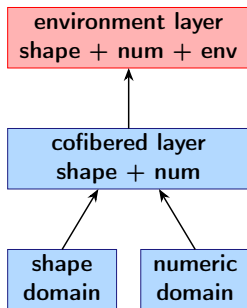
- **Widening:**



## Widening / join in the combined domain



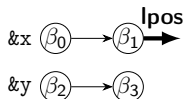
## Widening / join in the combined domain



$$N = \alpha_2 \geq \alpha_5 \geq 2$$

$$\&x \ \delta_0$$

$$\&y \ \delta_1$$



$$N' = \beta_3 \geq 1$$

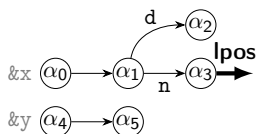
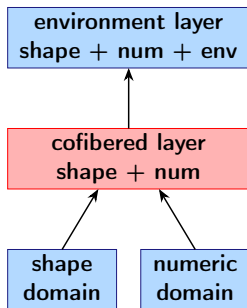
$$\delta_0 \equiv (\alpha_0, \beta_0)$$

$$\delta_1 \equiv (\alpha_4, \beta_2)$$

## Stage 1: abstract environment

- compute new abstract environment and initial node relation  
e.g.,  $\alpha_0, \beta_0$  both denote  $\&x$

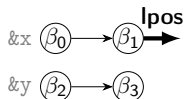
## Widening / join in the combined domain



$$N = \alpha_2 \geq \alpha_5 \geq 2$$

$$\&x \ \delta_0$$

$$\&y \ \delta_1$$



$$N' = \beta_3 \geq 1$$

$$\delta_0 \equiv (\alpha_0, \beta_0)$$

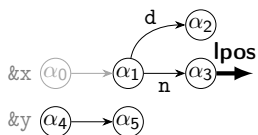
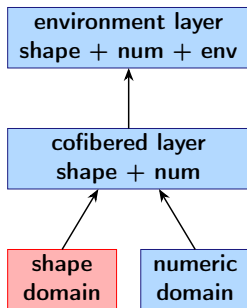
$$\delta_1 \equiv (\alpha_4, \beta_2)$$

## Stage 2: join in the “cofibered” layer

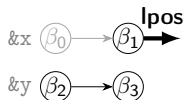
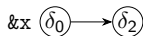
operations to perform:

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

## Widening / join in the combined domain



$$N = \alpha_2 \geq \alpha_5 \geq 2$$



$$N' = \beta_3 \geq 1$$

$$\delta_0 \equiv (\alpha_0, \beta_0)$$

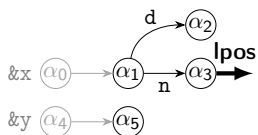
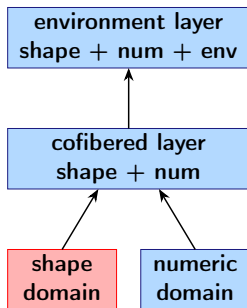
$$\delta_1 \equiv (\alpha_4, \beta_2)$$

$$\delta_2 \equiv (\alpha_1, \beta_1)$$

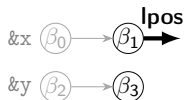
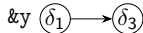
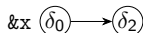
## Stage 2: graph join

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

## Widening / join in the combined domain



$$N = \alpha_2 \geq \alpha_5 \geq 2$$



$$N' = \beta_3 \geq 1$$

$$\delta_0 \equiv (\alpha_0, \beta_0)$$

$$\delta_1 \equiv (\alpha_4, \beta_2)$$

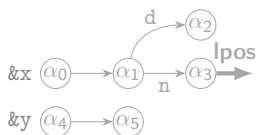
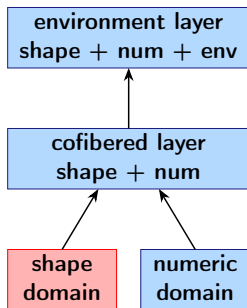
$$\delta_2 \equiv (\alpha_1, \beta_1)$$

$$\delta_3 \equiv (\alpha_5, \beta_3)$$

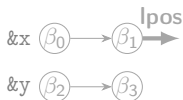
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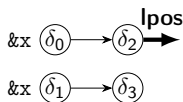
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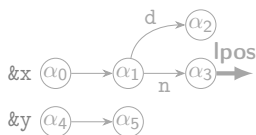
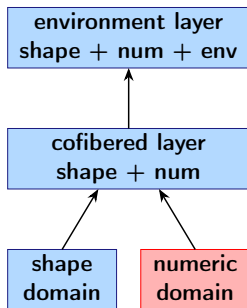


$$\begin{aligned} \delta_0 &\equiv (\alpha_0, \beta_0) \\ \delta_1 &\equiv (\alpha_4, \beta_2) \\ \delta_2 &\equiv (\alpha_1, \beta_1) \\ \delta_3 &\equiv (\alpha_5, \beta_3) \end{aligned}$$

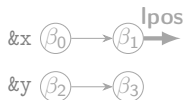
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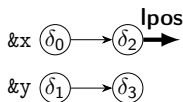
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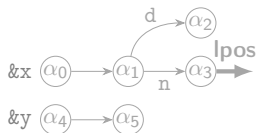
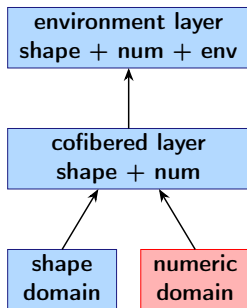
$$N^\sqcup = [\delta_3 \geq 2] \sqcup [\delta_3 \geq 1]$$

## Stage 3: conversion function application in numerics

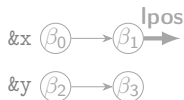
- remove nodes that were abstracted away
- rename other nodes



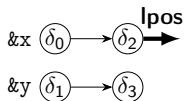
## Widening / join in the combined domain



$$\mathbf{N} = \alpha_2 \geq \alpha_5 \geq 2$$



$$\mathbf{N}' = \beta_3 \geq 1$$



$$\mathbf{N}^{\sqcup} = [\delta_3 \geq 1]$$

$$\begin{aligned} \delta_0 &\equiv (\alpha_0, \beta_0) \\ \delta_1 &\equiv (\alpha_4, \beta_2) \\ \delta_2 &\equiv (\alpha_1, \beta_1) \\ \delta_3 &\equiv (\alpha_5, \beta_3) \end{aligned}$$

## Stage 4: join in the numeric domain

- apply  $\sqcup$  for regular join,  $\nabla$  for a widening

# Memory abstract domain interfaces

- **Symbolic variables** represent **concrete values**  $\mathbf{Val}^\sharp$
- They appear in  $\mathbf{C}^\sharp$ , in  $\mathbf{S}^\sharp$ , in  $\mathbf{G}^\sharp$ , and in  $\mathbf{V}^\sharp$

**Concretizations** all involve **valuations** (values of symbolic variables)

$$\text{for } \mathbf{G}^\sharp : \quad \gamma_{\mathbf{G}} : \mathbf{G}^\sharp \rightarrow \mathcal{P}((\mathbf{Val}^\sharp \rightarrow \mathbf{Val}) \times \mathbf{Mem})$$

$$\text{for } \mathbf{V}^\sharp : \quad \gamma_{\mathbf{V}} : \mathbf{V}^\sharp \rightarrow \mathcal{P}(\mathbf{Val}^\sharp \rightarrow \mathbf{Val})$$

$$\text{for } \mathbf{S}^\sharp : \quad \gamma_{\mathbf{C}} : \mathbf{C}^\sharp \rightarrow \mathcal{P}((\mathbf{Val}^\sharp \rightarrow \mathbf{Val}) \times \mathbf{Mem})$$

**Abstract operations** in  $\mathbf{G}^\sharp$  (simplified excerpt):

$$\text{unfold} : \quad \mathbf{G}^\sharp \times \mathbf{Val}^\sharp \rightarrow \mathcal{P}_{\text{fin}}(\mathbf{G}^\sharp)$$

$$\text{eval-lvalue} : \quad \mathbf{G}^\sharp \times \mathbf{Lvalue}[\mathbf{Val}^\sharp] \rightarrow \mathbf{Val}^\sharp$$

$$\text{eval-expr} : \quad \mathbf{G}^\sharp \times \mathbf{Expr}[\mathbf{Val}^\sharp] \rightarrow \mathbf{Val}^\sharp$$

$$\text{read} : \quad \mathbf{G}^\sharp \times \mathbf{Val}^\sharp \times \text{offset} \times \text{size} \rightarrow \mathbf{G}^\sharp \times \mathbf{Val}^\sharp$$

$$\text{write} : \quad \mathbf{G}^\sharp \times \mathbf{Val}^\sharp \times \text{offset} \times \text{size} \rightarrow \mathbf{G}^\sharp$$

# Outline

- 1 Memory abstraction
- 2 A few memory abstractions, with value information
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- 4 Instantiations**
- 5 Separating product of abstractions
- 6 Reduced product of abstractions
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Instantiations of  $G^\#$ 

A general  $G^\#$  definition; needs to be instantiated

- It is **parameterized** by a **set of inductive definitions**
- Inductive definitions could be **built-in** in the implementation, pre-defined by the user, inferred...

**Many possible instantiations**, result in **different implementations**:

- $G^\#_{\langle \iota_0, \dots, \iota_k \rangle}$ , **parameterized by a set of inductive definitions**:  
**inputs a finite set of inductive predicates, very expressive**
- $G^\#_{\text{flat}}$ , **no inductive definitions** [Miné, LCTES'06]
- $G^\#_{\text{list}}$ , **a single, built-in definition** (e.g., list)

# Instantiations of $G^\#$ : impact on scalability

**Data-structures** to represent abstract values:

- $G^\#_{\langle \iota_0, \dots, \iota_k \rangle}$ : based on **graphs**
- $G^\#_{\text{flat}}$ , **no inductive definitions**: based on **functional arrays** (maps)
- $G^\#_{\text{list}}$ , **a single, built-in definition**: also graphs

**Algorithms**, e.g. for **join**:

- $G^\#_{\langle \iota_0, \dots, \iota_k \rangle}$ :  
needs to traverse the graphs (almost never physically equal)
- $G^\#_{\text{flat}}$ , **no inductive definitions**:  
can be done in  $\mathcal{O}(1)$  when physically equal
- $G^\#_{\text{list}}$ : graph join, as in  $G^\#_{\langle \iota_0, \dots, \iota_k \rangle}$ ,  
but much simpler: **no inductive inference in folding**

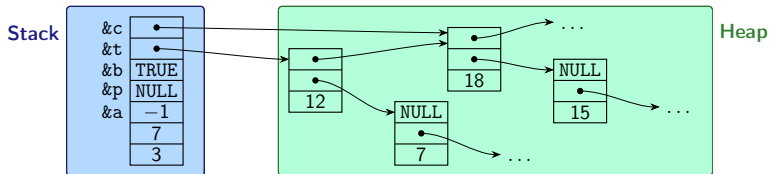
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# Applying different abstractions to different regions

- Stores often contain (large) **trivial** regions (e.g., local variables)
- More complex structures may be **limited to precise areas** (heap)

**Expensive abstraction should not be applied uniformly**

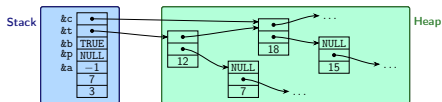


## Different local abstractions on different regions

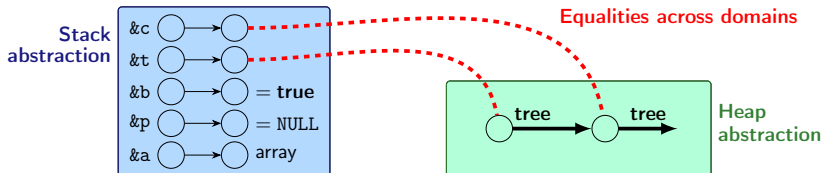
- Separating conjunction is applied **at the abstract domain level**
- Lightweight abstraction for region **Stack**:  $G_{\text{flat}}^{\#}$
- More precise abstraction for region **Heap**:  $G_{\langle \text{tree} \rangle}^{\#}$

## The separating product [SAS'14]

## A concrete state:



## Abstract state:



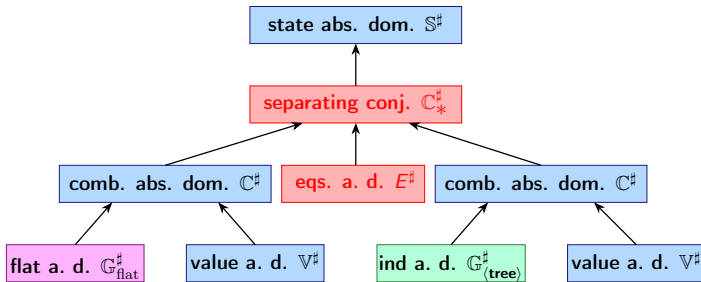
## Abstract domain and concretization

- $\mathbb{C}_*^\# = \mathbb{C}_1^\# \times \mathbb{C}_2^\# \times E^\#$  where  $E^\#$  abstracts equalities over  $\text{Val}^\# \times \text{Val}^\#$
- $\gamma_*(\mathbf{G}_0, \mathbf{G}_1, R) = \{(h_0 \uplus h_1, \nu) \mid \forall i, (h_i, \nu) \in \gamma_{\mathbb{C}}(\mathbf{G}_i) \wedge \nu \in \gamma_{\equiv}(R)\}$



## Use of the separating product

## Resulting abstract domain structure:



## A set of ML modules:

- Lightweight  $G_{\text{flat}}^{\#}$
- Expressive  $G_{\langle \text{tree} \rangle}^{\#}$ , applied locally
- **Combination functor + equalities**

## Alternate composition order:

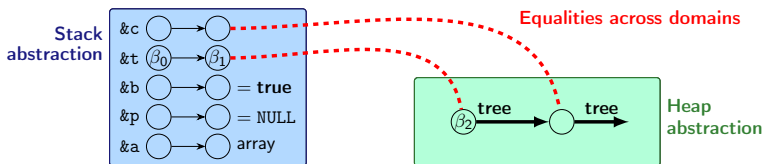
- Product with  $V^{\#}$  above  $C^{\#}_*$
- More expressive
- More expensive

# Computation of transfer functions

## Extension of abstract operations

- L-values may need to **evaluate over several sub-domains**
- **Equalities** also have to be used, in order to evaluate paths

**Example:** read  $t \rightarrow 1$  in



- 1  $\&t$  evaluates into  $\beta_0$  in  $G_{\text{flat}}^\#$
- 2  $\star(\&t)$  evaluates into  $\beta_1$ , also in  $G_{\text{flat}}^\#$
- 3 But then  $t \rightarrow 1$  needs to evaluate inside  $G_{\langle \text{tree} \rangle}^\#$   
To do this, we need to use **equalities**:  $\beta_1 = \beta_2$

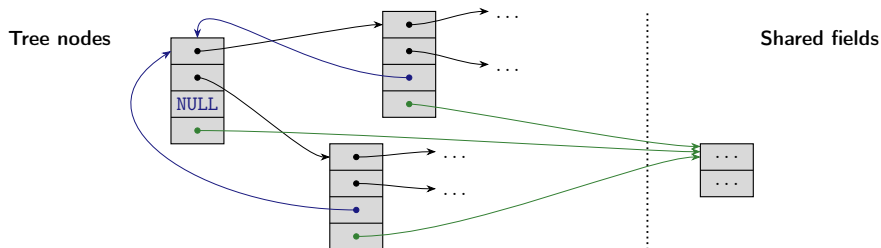
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# Conjoining heap abstractions

## An overlaid data-structure:

- **Tree** with parent pointers
- All nodes have a **shared pointer to a common record**



## A global conjunction of properties

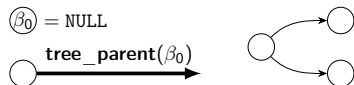
- Calls for a **reduced product** of abstract domains
- **Issue:** conjunction in memory abstractions can be expensive

# The reduced product of memory abstract domains

[VMCAI'13]

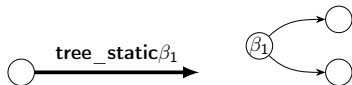
## Domain $\mathbb{C}_0^\#$ :

- Inductive **tree\_parent**: tracks **parent pointers**
- $\mathbb{C}_0^\#$  is  $\mathbb{G}_{\langle \text{tree\_parent} \rangle}^\#$



## Domain $\mathbb{C}_1^\#$ :

- Inductive **tree\_static**: tracks **pointers to static fields**
- $\mathbb{C}_1^\#$  is  $\mathbb{G}_{\langle \text{tree\_static} \rangle}^\#$

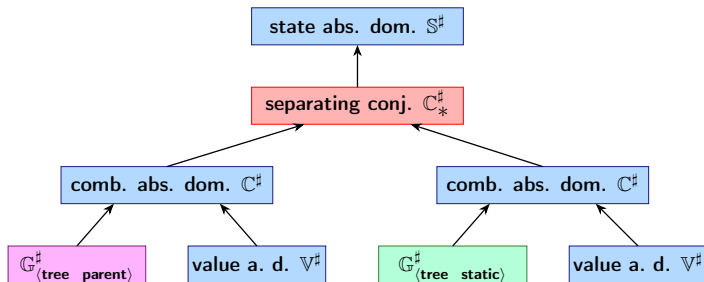


### Abstract domain and concretization

- $\mathbb{C}_\wedge^\# = \mathbb{C}_1^\# \times \mathbb{C}_2^\#$
- $\gamma_\wedge(\mathbf{G}_0, \mathbf{G}_1) = \gamma_{\mathbb{C}}(\mathbf{G}_0) \cap \gamma_{\mathbb{C}}(\mathbf{G}_1)$

# Use of the reduced product

## Resulting abstract domain structure:



- **Reduction** at assignment / tests: exchange of **reachability** predicates, inferred from inductive definitions
- **Run-time**: about twice as long as the run-time of a single analysis
- **Gain**: **separation of concerns** in the inductive predicate specification

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# Array abstraction

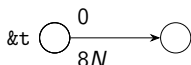
The shape abstract domain can abstract arrays in several ways

Example **concrete state**:

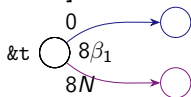


**Abstractions:**

- 1 A points-to edge may represent a **block of any size**:  $\beta_0 \cdot [0, 8[ \rightarrow \beta_1$   
Then,  $\forall^\#$  views  $\beta_1$  an  $8N$  bytes long block,  
can let a **dedicate array abstract domain** deal with it...



- 2 Offsets can be **expressions**, as in, e.g.,  $\beta_0 \cdot [8\beta_1, 8\beta_1 + 4[ \rightarrow \beta_2$   
as in [Cousot, Cousot, Logozzo'11]





# A Process Table in Minix 1.1

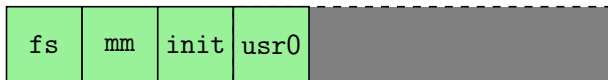
- Minix 1.1 is a **tiny operating system**
- It features **services** like **memory management**
- Memory management stores a **process table**, with one record per process



- Records of system processes are **permanent**
  - ▶ `mm` : memory management process
  - ▶ `fs` : file system process
  - ▶ `init` : the ancestor of all user processes

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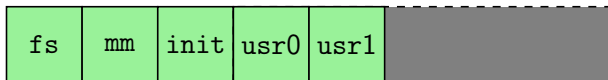
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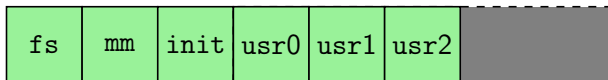
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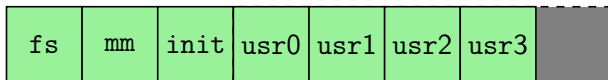
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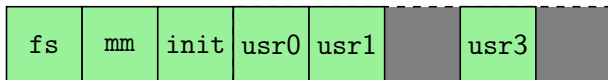
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# A Process Table in Minix 1.1

- Minix 1.1 is a **tiny operating system**
- It features **services** like **memory management**
- Memory management stores a **process table**, with one record per process



- Records of system processes are **permanent**
  - ▶ `mm` : memory management process
  - ▶ `fs` : file system process
  - ▶ `init` : the ancestor of all user processes
- User processes `usr $i$` , **dynamically created** or **removed**

**Cells with similar properties are not contiguous**

## A Process Table in Minix 1.1: Data-Type

```

struct{
    int flag;
    /* valid is not 0 */
    int parent;
    /* the index of parent */
}mproc[24]

```

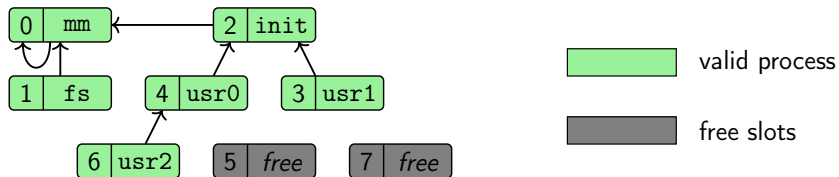
	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## A Process Table in Minix 1.1: Correctness Invariant

## Structural correctness property MMTPC

Field `parent` of any valid process descriptor should be the index of a valid process descriptor

- i.e., **no dangling processes**
- should be **established after `mm_init`**
- should be **preserved by system calls**





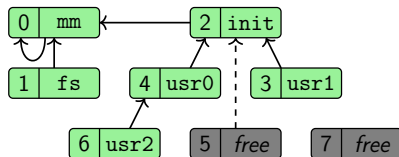
# A Code Segment from Fork

{Pre-condition: MMTPC}

```
int fork(int seed)
{
  ...
  for(i = 0; i < 24; i++){
    if(mproc[i].flag == 0)
      break;
  }
  mproc[i].flag = mproc[seed].flag;
  mproc[i].parent = seed;
  return 1;
}
```

{Post-condition: MMTPC}

- ① searches for a free slot
- ② allocates a record
- ③ update parent

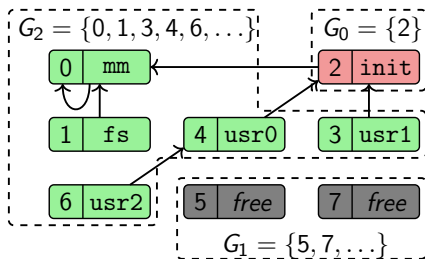


We want to verify MMTPC automatically  
 Thus, we need contiguous partitions

Towards an Abstraction of `mproc`

All cells in array `mproc` are partitioned into **three disjoint groups**:

- **Group  $G_0$** : `init` process
- **Group  $G_1$** : free slots in array `mproc`
- **Group  $G_2$** : all valid process descriptors except `init`



**Key distinguishing factor with common array abstractions:  
groups may be non-contiguous**

## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## Abstract memory state:

## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
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## Abstract memory state:



Partitions and field properties will be computed automatically

## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## Abstract memory state:



$\text{Index}_i$  : the summary of indexes in group  $i$

Partitions and field properties will be computed automatically

## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## Abstract memory state:



$\text{Size}_i$  : the number of elements in group  $i$

Partitions and field properties will be computed automatically

## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## Abstract memory state:



$\text{flag}_i$  : the summary of values on field `flag` in group  $i$

Partitions and field properties will be computed automatically

## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## Abstract memory state:



$\text{parent}_i$  : the summary of values on field parent in group  $i$

Partitions and field properties will be computed automatically

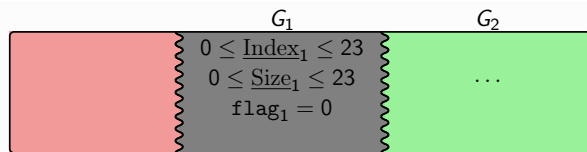


## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## Abstract memory state:



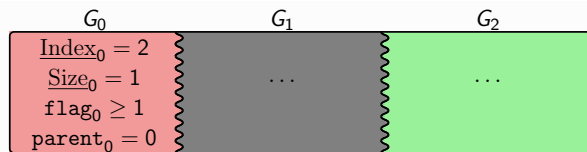
Partitions and field properties will be computed automatically

## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## Abstract memory state:



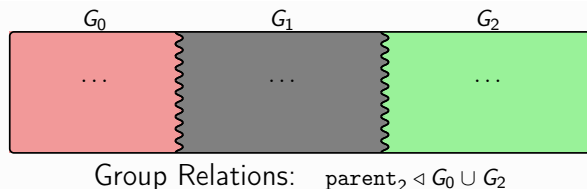
Partitions and field properties will be computed automatically

## Towards an Abstraction of mproc

## A concrete memory state:

	[0]	[1]	[2]	[3]	[4]	[5]	[6]	[7]
flag	1	1	1	1	1	0	1	0
parent	0	0	0	2	2	?	4	?

## Abstract memory state:



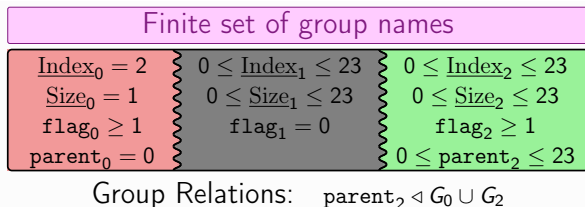
Partitions and field properties will be computed automatically

## Abstract States

$G_0$	$G_1$	$G_2$
$\underline{\text{Index}}_0 = 2$ $\underline{\text{Size}}_0 = 1$ $\text{flag}_0 \geq 1$ $\text{parent}_0 = 0$	$0 \leq \underline{\text{Index}}_1 \leq 23$ $0 \leq \underline{\text{Size}}_1 \leq 23$ $\text{flag}_1 = 0$	$0 \leq \underline{\text{Index}}_2 \leq 23$ $0 \leq \underline{\text{Size}}_2 \leq 23$ $\text{flag}_2 \geq 1$ $0 \leq \text{parent}_2 \leq 23$
Group Relations: $\text{parent}_2 \triangleleft G_0 \cup G_2$		

- **Concretization into sets of concrete states:** in the paper

# Abstract States



An abstract state comprises:

- A finite set of groups

- **Concretization into sets of concrete states:** in the paper

# Abstract States

Finite set of group names

Numerical constraints

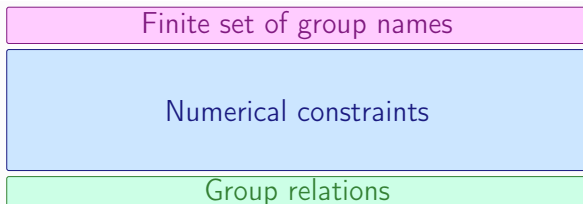
Group Relations:  $\text{parent}_2 \triangleleft G_0 \cup G_2$

An abstract state comprises:

- A finite set of groups
- Numerical constraints over:
  - ▶ base type variables
  - ▶ group sizes and cell indexes
  - ▶ group fields summary variables

- Concretization into sets of concrete states: in the paper

# Abstract States



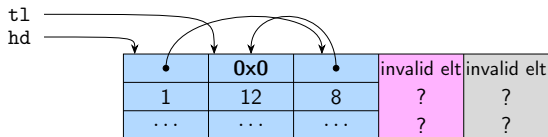
An abstract state comprises:

- A **finite set of groups**
  - **Numerical constraints** over:
    - ▶ **base type variables**
    - ▶ group **sizes** and **cell indexes**
    - ▶ group **fields summary variables**
  - **Group relations**: basic membership constraints
- **Concretization into sets of concrete states**: in the paper

# Hierarchical abstraction [APLAS'12]

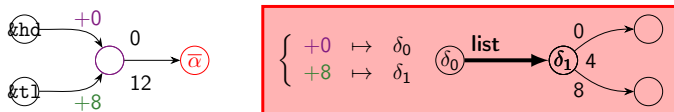
A **common pattern** in **embedded** software:

- No use of `malloc`, manual management of a **static zone** instead
- One particular implementation: **large array** of structures, allocated **from the beginning**, fully erased from time to time



This makes the analysis **very challenging**:  
both array and dynamic structures abstractions will fail...

**Hierarchical abstraction**: **attaches a (shape) predicate to a node**





# Outline

- 1 Memory abstraction
- 2 A few memory abstractions, with value information
- 3 Abstract operations
- 4 Instantiations
- 5 Separating product of abstractions
- 6 Reduced product of abstractions
- 7 Abstraction of arrays and hierarchical abstraction
- 8 Project status**

# Results and future works

## Abstract domains

- Inductive-parameterized  $\mathbb{G}_{\langle \dots \rangle}^{\#}$
- List specific  $\mathbb{G}_{\text{flat}}^{\#}$
- Flat  $\mathbb{G}_{\text{list}}^{\#}$
- Array, non contiguous partitions
- Numeric (Apron)

## Combination operations

- Separating product
- Reduced product
- Hierarchical functor

## Ongoing and future works:

- **Additional domains** (arrays, strings)
- **Improvement of the analyzer / tuning of the operators**
- **Packaging of the domain** for integration in other tools, e.g. Astrée