

# Partitioning abstractions

MPRI — Cours 2.6 “Interprétation abstraite :  
application à la vérification et à l’analyse statique”

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# Towards disjunctive abstractions

## Extending the expressiveness of abstract domains

- **disjunctions** are **often needed**...
- ... but **potentially costly**

In this lecture, we will discuss:

- **precision issues** that motivate the use of abstract domains able to **express disjunctions**
- **several ways** to **express disjunctions** using **abstract domain combinators**
  - ▶ disjunctive completion
  - ▶ cardinal power
  - ▶ state partitioning
  - ▶ trace partitioning

# Domain combinators (or combiners)

## General combination of abstract domains

- takes one or more abstract domains as **inputs**
- produces a **new abstract domain**

Input and output abstract domains are **characterized by an “interface”**: concrete domain, abstraction relation, abstract elements and operators

## Advantages:

- **general definition**, formalized and proved once
- can be **implemented** in a separate way, e.g., in ML:
  - ▶ abstract domain: **module**

```
module D = (struct ... end: Interface)
```
  - ▶ abstract domain combinator: **functor**

```
module C = functor (D: Interface) ->
  (struct ... end: Interface)
```

# Example: product abstraction

## Set notations: Assumptions:

- $\mathbb{V}$ : values
- $\mathbb{X}$ : variables
- $\mathbb{M}$ : stores  
 $\mathbb{M} = \mathbb{X} \rightarrow \mathbb{V}$
- concrete domain  $(\mathcal{P}(\mathbb{M}), \subseteq)$  with  $\mathbb{M} = \mathbb{X} \rightarrow \mathbb{V}$
- we assume an abstract domain  $\mathbb{D}^\#$  that provides
  - ▶ **concretization function**  $\gamma : \mathbb{D}^\# \rightarrow \mathcal{P}(\mathbb{M})$
  - ▶ **element  $\perp$  with empty concretization**  $\gamma(\perp) = \emptyset$

## Product combinator (implemented as a functor)

Given abstract domains  $(\mathbb{D}_0^\#, \gamma_0, \perp_0)$  and  $(\mathbb{D}_1^\#, \gamma_1, \perp_1)$ , the **product abstraction** is  $(\mathbb{D}_x^\#, \gamma_x, \perp_x)$  where:

- $\mathbb{D}_x^\# = \mathbb{D}_0^\# \times \mathbb{D}_1^\#$
- $\gamma_x(x_0^\#, x_1^\#) = \gamma_0(x_0^\#) \cap \gamma_1(x_1^\#)$
- $\perp_x = (\perp_0, \perp_1)$

**This amounts to expressing conjunctions of elements of  $\mathbb{D}_0^\#$  and  $\mathbb{D}_1^\#$**

## Example: product abstraction, coalescent product

The product abstraction is not very precise and **needs a reduction**:

$$\forall x_0^\# \in \mathbb{D}_0^\#, x_1^\# \in \mathbb{D}_1^\#, \gamma_x(\perp_0, x_1^\#) = \gamma_x(x_0^\#, \perp_1) = \emptyset = \gamma_x(\perp_x)$$

### Coalescent product

Given abstract domains  $(\mathbb{D}_0^\#, \gamma_0, \perp_0)$  and  $(\mathbb{D}_1^\#, \gamma_1, \perp_1)$ , the **coalescent product abstraction** is  $(\mathbb{D}_x^\#, \gamma_x, \perp_x)$  where:

- $\mathbb{D}_x^\# = \{\perp_x\} \uplus \{(x_0^\#, x_1^\#) \in \mathbb{D}_0^\# \times \mathbb{D}_1^\# \mid x_0^\# \neq \perp_0 \wedge x_1^\# \neq \perp_1\}$
- $\gamma_x(\perp_x) = \emptyset, \gamma_x(x_0^\#, x_1^\#) = \gamma_0(x_0^\#) \cap \gamma_1(x_1^\#)$

In many cases, this is **not enough to achieve reduction**:

- let  $\mathbb{D}_0^\#$  be the interval abstraction,  $\mathbb{D}_1^\#$  be the congruences abstraction
- $\gamma_x(\{x \in [3, 4]\}, \{x \equiv 0 \pmod{5}\}) = \emptyset$

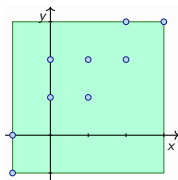
- how to define abstract domain combinators to **add disjunctions** ?

# Outline

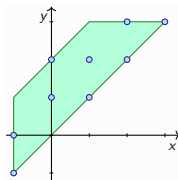
- 1 Introduction
- 2 Imprecisions in convex abstractions**
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# Convex abstractions

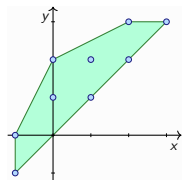
Many numerical abstractions describe convex sets of points



interval domain

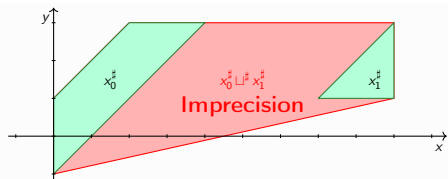


octagon domain



polyhedra domain

Imprecisions inherent in the **convexity**, and when computing **abstract join** (over-approximation of concrete union):



Such imprecisions may impact analysis results

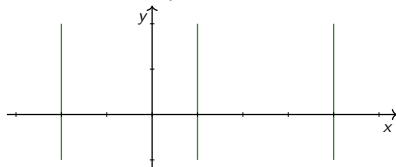
# Non convex abstractions

We consider abstractions of  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$

## Congruences:

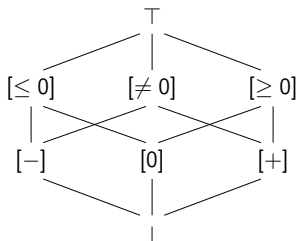
- $\mathbb{D}^\# = \mathbb{Z} \times \mathbb{N}$
- $\gamma(n, k) = \{n + k \cdot p \mid p \in \mathbb{Z}\}$
- $-2 \in \gamma(1, 2)$  and  $1 \in \gamma(1, 2)$   
but  $0 \notin \gamma(1, 2)$

Non relational product two variables



## Signs:

- $0 \notin \gamma([\neq 0])$  so  $[\neq 0]$  describes a non convex set
- other abstract elements describe convex sets





# Example 1: verification problem

```

bool b0, b1;
int x, y;      (uninitialized)
b0 = x ≥ 0;
b1 = x ≤ 0;
if(b0 && b1){
    y = 0;
} else {
  ①   y = 100/x;
}

```

- if  $\neg b_0$ , then  $x < 0$
- if  $\neg b_1$ , then  $x > 0$
- if either  $b_0$  or  $b_1$  is false, then  $x \neq 0$
- thus, if point ① is reached the division is safe

## How to verify the division operation ?

- Non relational abstraction (e.g., intervals), at point ①:

$$\begin{cases} b_0 = \text{FALSE} \vee b_1 = \text{FALSE} \\ x : \top \end{cases}$$

- Signs, congruences do not help:  
in the concrete,  $x$  may take any value but 0

## Example 1: program annotated with local invariants

```

bool b0, b1;
int x, y;      (uninitialized)
b0 = x ≥ 0;
      (b0 ∧ x ≥ 0) ∨ (¬b0 ∧ x < 0)
b1 = x ≤ 0;
      (b0 ∧ b1 ∧ x = 0) ∨ (b0 ∧ ¬b1 ∧ x > 0) ∨ (¬b0 ∧ b1 ∧ x < 0)
if(b0 && b1){
      (b0 ∧ b1 ∧ x = 0)
  y = 0;
      (b0 ∧ b1 ∧ x = 0 ∧ y = 0)
} else {
      (b0 ∧ ¬b1 ∧ x > 0) ∨ (¬b0 ∧ b1 ∧ x < 0)
  y = 100/x;
      (b0 ∧ ¬b1 ∧ x > 0) ∨ (¬b0 ∧ b1 ∧ x < 0)
}

```

The obvious way to successfully analyzing this program consists in **adding symbolic disjunctions** to our abstract domain

## Example 2: verification problem

```

int x ∈ ℤ;
int s;
int y;
if(x ≥ 0){
    s = 1;
} else {
    s = -1;
}
① y = x/s;
② assert(y ≥ 0);

```

- s is either 1 or -1
- thus, the division at ① should not fail
- moreover s has the same sign as x
- thus, the value stored in y should always be positive at ②

- How to verify the division operation ?
- In the concrete, s is **always non null**:  
convex abstractions **cannot** establish this; **congruences** can
- Moreover, s has always the **same sign** as x  
expressing this would require a non trivial numerical abstraction

## Example 2: program annotated with local invariants

```

int x ∈ ℤ;
int s;
int y;
if(x ≥ 0){
    (x ≥ 0)
    s = 1;
    (x ≥ 0 ∧ s = 1)
} else {
    (x < 0)
    s = -1;
    (x < 0 ∧ s = -1)
}
(x ≥ 0 ∧ s = 1) ∨ (x < 0 ∧ s = -1)
① y = x/s;
(x ≥ 0 ∧ s = 1 ∧ y ≥ 0) ∨ (x < 0 ∧ s = -1 ∧ y > 0)
② assert(y ≥ 0);

```

Again, the obvious solution consists in  
**adding disjunctions** to our abstract domain

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# Distributive abstract domain

## Principle:

- 1 consider concrete domain  $(\mathbb{D}, \sqsubseteq)$ , with lower upper bound operator  $\sqcup$
- 2 assume an abstract domain  $(\mathbb{D}^\#, \sqsubseteq^\#)$  with concretization  $\gamma : \mathbb{D}^\# \rightarrow \mathbb{D}$
- 3 build a domain containing **all the disjunctions** of elements of  $\mathbb{D}^\#$

## Definition: distributive abstract domain

Abstract domain  $(\mathbb{D}^\#, \sqsubseteq^\#)$  with concretization function  $\gamma : \mathbb{D}^\# \rightarrow \mathbb{D}$  is **distributive** (or **complete for disjunction**) if and only if:

$$\forall \mathcal{E} \subseteq \mathbb{D}^\#, \exists x^\# \in \mathbb{D}^\#, \gamma(x^\#) = \bigsqcup_{y^\# \in \mathcal{E}} \gamma(y^\#)$$

## Examples:

- the lattice  $\{\perp, < 0, = 0, > 0, \leq 0, \neq 0, \geq 0, \top\}$  is distributive
- the lattice of intervals is not distributive:  
there is no interval with concretization  $\gamma([0, 10]) \cup \gamma([12, 20])$

# Definition

## Definition: disjunctive completion

The **disjunctive completion** of abstract domain  $(\mathbb{D}^\#, \sqsubseteq^\#)$  with concretization function  $\gamma : \mathbb{D}^\# \rightarrow \mathbb{D}$  is the **smallest abstract domain**  $(\mathbb{D}_{\text{disj}}^\#, \sqsubseteq_{\text{disj}}^\#)$  with concretization function  $\gamma_{\text{disj}} : \mathbb{D}_{\text{disj}}^\# \rightarrow \mathbb{D}$  such that:

- $\mathbb{D}^\# \subseteq \mathbb{D}_{\text{disj}}^\#$
- $\forall x^\# \in \mathbb{D}^\#, \gamma_{\text{disj}}(x^\#) = \gamma(x^\#)$
- $(\mathbb{D}_{\text{disj}}^\#, \sqsubseteq_{\text{disj}}^\#)$  with concretization  $\gamma_{\text{disj}}$  is distributive

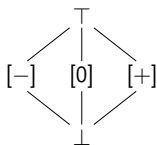
## Building a disjunctive completion domain:

- start with  $\mathbb{D}_{\text{disj}}^\# = \mathbb{D}^\#$
- for all set  $\mathcal{E} \subseteq \mathbb{D}^\#$  such that there is no  $x^\# \in \mathbb{D}^\#$ , such that  $\gamma(x^\#) = \bigsqcup_{y^\# \in \mathcal{E}} \gamma(y^\#)$ , add  $\lfloor \sqcup \mathcal{E} \rfloor$  to  $\mathbb{D}_{\text{disj}}^\#$ , and extend  $\gamma_{\text{disj}}$  by

$$\gamma_{\text{disj}}(\lfloor \sqcup \mathcal{E} \rfloor) = \bigsqcup_{y^\# \in \mathcal{E}} \gamma(y^\#)$$

# Example 1: completion of signs

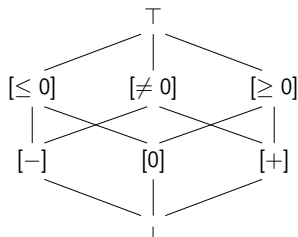
We consider **concrete lattice**  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$ , with  $\sqsubseteq = \subseteq$  and  $(\mathbb{D}^\#, \sqsubseteq^\#)$  defined by:



$$\gamma: \begin{array}{ll} \perp & \mapsto \emptyset \\ [-] & \mapsto \{k \in \mathbb{Z} \mid k < 0\} \\ [0] & \mapsto \{k \in \mathbb{Z} \mid k = 0\} \\ [+] & \mapsto \{k \in \mathbb{Z} \mid k > 0\} \\ \top & \mapsto \mathbb{Z} \end{array}$$

Then, the disjunctive completion is defined by adding elements corresponding to:

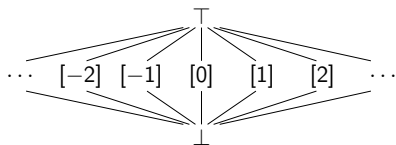
- $\{[-], [0]\}$
- $\{[-], [+]\}$
- $\{[0], [+]\}$





## Example 2: completion of constants

We consider **concrete lattice**  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$ , with  $\sqsubseteq = \subseteq$   
and  $(\mathbb{D}^\#, \sqsubseteq^\#)$  defined by:



$$\gamma : \begin{array}{ll} \perp & \mapsto \emptyset \\ \{k\} & \mapsto \{k\} \\ \top & \mapsto \mathbb{Z} \end{array}$$

Then, the disjunctive completion coincides with **the power-set**:

- $\mathbb{D}_{\text{disj}}^\# \equiv \mathcal{P}(\mathbb{Z})$
- $\gamma_{\text{disj}}$  is the **identity function !**
- this lattice contains **infinite sets which are not representable**

## Example 3: completion of intervals

We consider concrete lattice  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$ , with  $\sqsubseteq = \subseteq$  and let  $(\mathbb{D}^\#, \sqsubseteq^\#)$  the domain of intervals

- $\mathbb{D}^\# = \{\perp, \top\} \uplus \{[a, b] \mid a \leq b\}$
- $\gamma([a, b]) = \{x \in \mathbb{Z} \mid a \leq x \leq b\}$

Then, the disjunctive completion is the set of **unions of intervals** :

- $\mathbb{D}_{\text{disj}}^\#$  collects all the families of disjoint intervals
- this lattice contains **infinite sets which are not representable**
- as expressive as the completion of constants, but more efficient representation

The disjunctive completion of  $(\mathbb{D}^\#)^n$  is **not equivalent** to  $(\mathbb{D}_{\text{disj}}^\#)^n$

- which is more expressive ?
- show it on an example !

## Example 3: completion of intervals and verification

We use the disjunctive completion of  $(\mathbb{D}^\#)^3$ .

The invariants below can be expressed in the disjunctive completion:

```

int x  $\in$   $\mathbb{Z}$ ;
int s;
int y;
if(x  $\geq$  0){
    (x  $\geq$  0)
    s = 1;
    (x  $\geq$  0  $\wedge$  s = 1)
} else {
    (x < 0)
    s = -1;
    (x < 0  $\wedge$  s = -1)
}
(x  $\geq$  0  $\wedge$  s = 1)  $\vee$  (x < 0  $\wedge$  s = -1)
y = x/s;
(x  $\geq$  0  $\wedge$  s = 1  $\wedge$  y  $\geq$  0)  $\vee$  (x < 0  $\wedge$  s = -1  $\wedge$  y > 0)
assert(y  $\geq$  0);
  
```

# Static analysis with disjunctive completion

**Transfer functions** for the computation of **abstract post-conditions**:

- we assume a concrete post-condition operation (assignment, guard...)  $post : \mathbb{D} \rightarrow \mathbb{D}$ , and an abstract  $post^\sharp : \mathbb{D}^\sharp \rightarrow \mathbb{D}^\sharp$  such that:

$$post \circ \gamma \sqsubseteq \gamma \circ post^\sharp$$

- then, we can simply use, **for the disjunctive completion domain**:

$$post_{\text{disj}}^\sharp([\sqcup \mathcal{E}]) = \begin{cases} y^\sharp & \text{if } \gamma(y^\sharp) = \sqcup \{post^\sharp(x^\sharp) \mid x^\sharp \in \mathcal{E}\} \\ [\sqcup \{post^\sharp(x^\sharp) \mid x^\sharp \in \mathcal{E}\}] & \text{otherwise} \end{cases}$$

**Abstract join:**

- disjunctive completion provides **an exact join** (exercise !)

**Inclusion check:** **exercise !**

# Limitations of disjunctive completion

- **Combinatorial explosion:**

- ▶ if  $\mathbb{D}^\sharp$  is infinite,  $\mathbb{D}_{\text{disj}}^\sharp$  may have elements that **cannot be represented** e.g., completion of constants or intervals
- ▶ even when  $\mathbb{D}^\sharp$  is finite,  $\mathbb{D}_{\text{disj}}^\sharp$  may be **huge**  
in the worst case, if  $\mathbb{D}^\sharp$  has  $n$  elements,  $\mathbb{D}_{\text{disj}}^\sharp$  may have  $2^n$  elements

- **Many elements useless in practice:**

disjunctive completion of intervals: may express any set of integers...

- **No general definition of a widening operator**

- ▶ most common approach to achieve that:  **$k$ -limiting**  
bound the numbers of disjuncts  
i.e., the size of the sets added to the base domain
- ▶ **remaining issue:** the join operator should “select” which disjoints to merge

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# Principle

## Observation

Disjuncts **that are required for static analysis**  
can usually be **characterized** by some **semantic property**

## Examples:

- **sign** of a variable
- **value** of a **boolean** variable
- **execution path**, e.g., side of a condition that was visited

**Solution:** perform a kind of **indexing** of disjuncts

- use an abstraction to **describe labels**  
e.g., sign of a variable, value of a boolean, or trace property...
- apply the abstraction that needs be completed on the images

## Disjuncts indexing: example

```

int x ∈ ℤ;
int s;
int y;
if(x ≥ 0){
    (x ≥ 0)
    s = 1;
    (x ≥ 0 ∧ s = 1)
} else {
    (x < 0)
    s = -1;
    (x < 0 ∧ s = -1)
}
(x ≥ 0 ∧ s = 1) ∨ (x < 0 ∧ s = -1)
y = x/s;
(x ≥ 0 ∧ s = 1 ∧ y ≥ 0) ∨ (x < 0 ∧ s = -1 ∧ y > 0)
assert(y ≥ 0);

```

- natural “indexing”: **sign of x**
- but we could also rely on the **sign of s**



# Cardinal power abstraction

We assume  $(\mathbb{D}, \sqsubseteq) = (\mathcal{P}(\mathcal{E}), \subseteq)$ , and two abstractions  $(\mathbb{D}_0^\#, \sqsubseteq_0^\#)$ ,  $(\mathbb{D}_1^\#, \sqsubseteq_1^\#)$  given by their concretization functions:

$$\gamma_0 : \mathbb{D}_0^\# \longrightarrow \mathbb{D} \quad \gamma_1 : \mathbb{D}_1^\# \longrightarrow \mathbb{D}$$

## Definition

We let the **cardinal power abstract domain** be defined by:

- $\mathbb{D}_{\text{cp}}^\# = \mathbb{D}_0^\# \xrightarrow{\mathcal{M}} \mathbb{D}_1^\#$  be the set of monotone functions from  $\mathbb{D}_0^\#$  into  $\mathbb{D}_1^\#$
- $\sqsubseteq_{\text{cp}}^\#$  be the pointwise extension of  $\sqsubseteq_1^\#$
- $\gamma_{\text{cp}}$  is defined by:

$$\begin{aligned} \gamma_{\text{cp}} : \mathbb{D}_{\text{cp}}^\# &\longrightarrow \mathbb{D} \\ X^\# &\longmapsto \{y \in \mathcal{E} \mid \forall z^\# \in \mathbb{D}_0^\#, y \in \gamma_0(z^\#) \implies y \in \gamma_1(X^\#(z^\#))\} \end{aligned}$$

We sometimes denote it by  $\mathbb{D}_0^\# \rightrightarrows \mathbb{D}_1^\#$ ,  $\gamma_{\mathbb{D}_0^\# \rightrightarrows \mathbb{D}_1^\#}$  to make it more explicit.

# Use of cardinal power abstractions

**Intuition:** cardinal power expresses properties of the form

$$\left\{ \begin{array}{l} p_0 \implies p'_0 \\ \wedge p_1 \implies p'_1 \\ \vdots \quad \vdots \quad \vdots \quad \vdots \\ \wedge p_n \implies p'_n \end{array} \right.$$

**Two independent choices:**

- 1  $\mathbb{D}_0^\sharp$ : **set of partitions** (the “labels”)
- 2  $\mathbb{D}_1^\sharp$ : **abstraction of sets of states**, e.g., a numerical abstraction

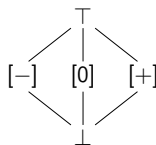
**Application**  $(x \geq 0 \wedge s = 1 \wedge y \geq 0) \vee (x < 0 \wedge s = -1 \wedge y > 0)$

- $\mathbb{D}_0^\sharp$ : sign of  $s$
- $\mathbb{D}_1^\sharp$ : other constraints

## Another example, with a single variable

## Assumptions:

- concrete lattice  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$ , with  $\sqsubseteq = \subseteq$
- $(\mathbb{D}_0^\#, \sqsubseteq_0^\#)$  be the **lattice of signs**  
(strict values only)
- $(\mathbb{D}_1^\#, \sqsubseteq_1^\#)$  be the **lattice of intervals**



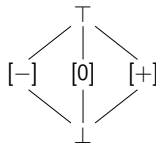
## Example abstract values:

- $[0, 8]$  is expressed by:
 
$$\left\{ \begin{array}{l} \perp \mapsto \perp_1 \\ [-] \mapsto \perp_1 \\ [0] \mapsto [0, 0] \\ [+] \mapsto [1, 8] \\ \top \mapsto [0, 8] \end{array} \right.$$

- $[-10, -3] \uplus [7, 10]$  is expressed by:
 
$$\left\{ \begin{array}{l} \perp \mapsto \perp_1 \\ [-] \mapsto [-10, -3] \\ [0] \mapsto \perp_1 \\ [+] \mapsto [7, 10] \\ \top \mapsto [-10, 10] \end{array} \right.$$

## Example reduction (1): relation between the two domains

- concrete lattice  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$ , with  $\sqsubseteq = \subseteq$
- $(\mathbb{D}_0^\#, \sqsubseteq_0^\#)$  be the **lattice of signs**
- $(\mathbb{D}_1^\#, \sqsubseteq_1^\#)$  be the **lattice of intervals**



We let:

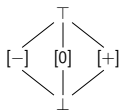
$$X^\# = \begin{cases} \perp & \mapsto \perp_1 \\ [-] & \mapsto [1, 8] \\ [0] & \mapsto [1, 8] \\ [+] & \mapsto \perp_1 \\ \top & \mapsto [1, 8] \end{cases} \quad Y^\# = \begin{cases} \perp & \mapsto \perp_1 \\ [-] & \mapsto [2, 45] \\ [0] & \mapsto [-5, -2] \\ [+] & \mapsto [-5, -2] \\ \top & \mapsto \top_1 \end{cases} \quad Z^\# = \begin{cases} \perp & \mapsto \perp_1 \\ [-] & \mapsto \perp_1 \\ [0] & \mapsto \perp_1 \\ [+] & \mapsto \perp_1 \\ \top & \mapsto \perp_1 \end{cases}$$

Then,

$$\gamma_{\text{cp}}(X^\#) = \gamma_{\text{cp}}(Y^\#) = \gamma_{\text{cp}}(Z^\#) = \emptyset$$

## Example reduction (2): tightening disjunctions

- concrete lattice  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$ , with  $\sqsubseteq = \subseteq$
- $(\mathbb{D}_0^\#, \sqsubseteq_0^\#)$  be the **lattice of signs**
- $(\mathbb{D}_1^\#, \sqsubseteq_1^\#)$  be the **lattice of intervals**



$$\text{We let: } X^\# = \begin{cases} \perp & \mapsto \perp_1 \\ [-] & \mapsto [-5, -1] \\ [0] & \mapsto [0, 0] \\ [+] & \mapsto [1, 5] \\ \top & \mapsto [-10, 10] \end{cases} \quad Y^\# = \begin{cases} \perp & \mapsto \perp_1 \\ [-] & \mapsto [-5, -1] \\ [0] & \mapsto [0, 0] \\ [+] & \mapsto [1, 5] \\ \top & \mapsto [-5, 5] \end{cases}$$

- Then,  $\gamma_{\text{cp}}(X^\#) = \gamma_{\text{cp}}(Y^\#)$
- $\gamma_0([-]) \cup \gamma_0([0]) \cup \gamma([+]) = \gamma(\top)$   
but

$$\gamma_0(X^\#([-])) \cup \gamma_0(X^\#([0])) \cup \gamma(X^\#([+])) \subset \gamma(X^\#(\top))$$

In fact, **we can improve the image of  $\top$  into  $[-5, 5]$**

## Reduction, and improving precision in the cardinal power

In general, **the cardinal power construction requires reduction**

Strengthening using both sides of  $\Rightarrow$

Tightening of  $y_0^\# \mapsto y_1^\#$  when:

- $\exists z_1^\# \neq y_1^\#, \gamma(y_1^\#) \cap \gamma(y_0^\#) \subseteq \gamma(z_1^\#)$
- in the example,  $z_1^\# = \perp_1 \dots$

Strengthening of one mapping using other mappings

Tightening of mapping  $(\sqcup\{z^\# \mid z^\# \in \mathcal{E}\}) \mapsto x_1^\#$  when:

- $\bigcup\{\gamma_0(z^\#) \mid z^\# \in \mathcal{E}\} = \gamma_0(\sqcup\{z^\# \mid z^\# \in \mathcal{E}\})$
- $\exists y^\#, \bigcup\{\gamma_1(X^\#(z^\#)) \mid z^\# \in \mathcal{E}\} \subseteq \gamma_1(y^\#) \subset \gamma_1(X^\#(\sqcup\{z^\# \mid z^\# \in \mathcal{E}\}))$
- in the example, we use a set of elements that cover  $\top \dots$

# Representation of the cardinal power

## Basic ML representation:

- using **functions**, i.e. type  $cp = d_0 \rightarrow d_1$   
 $\Rightarrow$  obviously a bad choice, as it makes it hard to operate in the  $\mathbb{D}_0^\sharp$  side
- using **some kind of dictionaries** type  $cp = (d_0, d_1) \text{ map}$   $\Rightarrow$   
 better, but not straightforward...

## The latter is not a very efficient representation:

- if  $\mathbb{D}_0^\sharp$  has  $N$  elements, then an abstract value in  $\mathbb{D}_{cp}^\sharp$  requires  $N$   
**elements of  $\mathbb{D}_1^\sharp$**
- if  $\mathbb{D}_0^\sharp$  is infinite, and  $\mathbb{D}_1^\sharp$  is non trivial, then  $\mathbb{D}_{cp}^\sharp$  **has elements that cannot be represented**
- the 1st reduction shows it is **unnecessary to represent bindings for all elements of  $\mathbb{D}_0^\sharp$**   
**example:** this is the case of  $\perp_0$

# More compact representation of the cardinal power

## Principle:

- use a **dictionary data-type** (most likely functional arrays)
- **avoid representing information attached to redundant elements**

## Compact representation

Reduced cardinal power of  $\mathbb{D}_0^\sharp$  and  $\mathbb{D}_1^\sharp$  can be represented by considering only a subset  $\mathcal{C} \subseteq \mathbb{D}_0^\sharp$  where

$$\forall x^\sharp \in \mathbb{D}_0^\sharp, \exists \mathcal{E} \subseteq \mathcal{C}, \gamma_0(x^\sharp) = \cup\{\gamma_0(y^\sharp) \mid y^\sharp \in \mathcal{E}\}$$

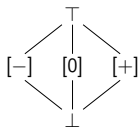
In particular:

- if possible,  $\mathcal{C}$  should be **minimal**
- in any case,  $\perp_0 \notin \mathcal{C}$



## Example: compact cardinal power over signs

- concrete lattice  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$ , with  $\sqsubseteq = \subseteq$
- $(\mathbb{D}_0^\sharp, \sqsubseteq_0^\sharp)$  be the **lattice of signs**
- $(\mathbb{D}_1^\sharp, \sqsubseteq_1^\sharp)$  be the **lattice of intervals**



### Observations

- $\perp$  does not need be considered (obvious right hand side:  $\perp_1$ )
- $\gamma_0([< 0]) \cup \gamma_0([= 0]) \cup \gamma_0([> 0]) = \gamma(T)$  thus  $T$  does not need be considered

Thus, we let  $\mathcal{C} = \{[-], [0], [+]\}$

- $[0, 8]$  is expressed by: 
$$\begin{cases} [-] \mapsto \perp_1 \\ [0] \mapsto [0, 0] \\ [+] \mapsto [1, 8] \end{cases}$$
- $[-10, -3] \uplus [7, 10]$  is expressed by: 
$$\begin{cases} [-] \mapsto [-10, -3] \\ [0] \mapsto \perp_1 \\ [+] \mapsto [7, 10] \end{cases}$$

# Lattice operations

## Infimum:

- we assume that  $\perp_1$  is the infimum of  $\mathbb{D}_1^\sharp$
- then,  $\perp_{\text{cp}} = \lambda(z^\sharp \in \mathbb{D}_0^\sharp) \cdot \perp_1$  is the **infimum** of  $\mathbb{D}_{\text{cp}}^\sharp$

## Ordering:

- we let  $\sqsubseteq_{\text{cp}}^\sharp$  denote the **pointwise ordering**:

$$X_0^\sharp \sqsubseteq_{\text{cp}}^\sharp X_1^\sharp \stackrel{\text{def}}{\iff} \forall z^\sharp \in \mathbb{D}_0^\sharp, X_0^\sharp(z^\sharp) \sqsubseteq_1^\sharp X_1^\sharp(z^\sharp)$$

- then,  $X_0^\sharp \sqsubseteq_{\text{cp}}^\sharp X_1^\sharp \implies \gamma_{\text{cp}}(X_0^\sharp) \subseteq \gamma_{\text{cp}}(X_1^\sharp)$

## Join operation:

- we assume that  $\sqcup_1$  is a sound upper bound operator in  $\mathbb{D}_1^\sharp$
- then,  $\sqcup_{\text{cp}}$  defined below is a **sound upper bound operator** in  $\mathbb{D}_{\text{cp}}^\sharp$ :

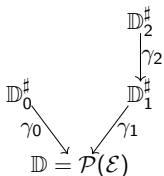
$$X_0^\sharp \sqcup_{\text{cp}} X_1^\sharp \stackrel{\text{def}}{::=} \lambda(z^\sharp \in \mathbb{D}_0^\sharp) \cdot (X_0^\sharp(z^\sharp) \sqcup_1 X_1^\sharp(z^\sharp))$$

- the same construction applies to widening, if  $\mathbb{D}_0^\sharp$  is finite

# Composition with another abstraction

We assume three abstractions

- $(\mathbb{D}_0^\#, \sqsubseteq_0^\#)$ , with concretization  $\gamma_0 : \mathbb{D}_0^\# \rightarrow \mathbb{D}$
- $(\mathbb{D}_1^\#, \sqsubseteq_1^\#)$ , with concretization  $\gamma_1 : \mathbb{D}_1^\# \rightarrow \mathbb{D}$
- $(\mathbb{D}_2^\#, \sqsubseteq_2^\#)$ , with concretization  $\gamma_2 : \mathbb{D}_2^\# \rightarrow \mathbb{D}_1^\#$



Cardinal power abstract domains  $\mathbb{D}_0^\# \rightrightarrows \mathbb{D}_1^\#$  and  $\mathbb{D}_0^\# \rightrightarrows \mathbb{D}_2^\#$  can be bound by an **abstraction relation** defined by concretization function  $\gamma$ :

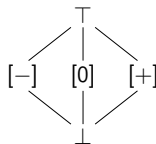
$$\begin{aligned} \gamma : (\mathbb{D}_0^\# \rightrightarrows \mathbb{D}_2^\#) &\longrightarrow (\mathbb{D}_0^\# \rightrightarrows \mathbb{D}_1^\#) \\ X^\# &\longmapsto \lambda(z^\# \in \mathbb{D}_0^\#). \gamma(X^\#(z^\#)) \end{aligned}$$

**Applications:**

- start with  $\mathbb{D}_1^\#$  as the **identity abstraction**
- **compose several** cardinal power abstractions (or partitioning abstractions)

## Composition with another abstraction

- concrete lattice  $\mathbb{D} = \mathcal{P}(\mathbb{Z})$ , with  $\sqsubseteq = \subseteq$
- $(\mathbb{D}_0^\sharp, \sqsubseteq_0^\sharp)$  be the **lattice of signs**
- $(\mathbb{D}_1^\sharp, \sqsubseteq_1^\sharp)$  be the **identity abstraction**  
 $\mathbb{D}_1^\sharp = \mathcal{P}(\mathbb{Z})$ ,  $\gamma_1 = \text{Id}$
- $(\mathbb{D}_2^\sharp, \sqsubseteq_2^\sharp)$  be the **lattice of intervals**



Then,  $[-10, -3] \uplus [7, 10]$  is **abstracted in two steps**:

- in  $\mathbb{D}_0^\sharp \Rightarrow \mathbb{D}_1^\sharp$ ,  $\begin{cases} [-] \mapsto [-10, -3] \\ [0] \mapsto \emptyset \\ [+] \mapsto [7, 10] \end{cases}$
- in  $\mathbb{D}_0^\sharp \Rightarrow \mathbb{D}_2^\sharp$ ,  $\begin{cases} [-] \mapsto [-10, -3] \\ [0] \mapsto \perp_1 \\ [+] \mapsto [7, 10] \end{cases}$

# Outline

- 1 Introduction
- 2 Imprecisions in convex abstractions
- 3 Disjunctive completion
- 4 Cardinal power and partitioning abstractions
- 5 State partitioning**
  - Definition and examples
  - Control states partitioning and iteration techniques
  - Abstract interpretation with boolean partitioning
- 6 Trace partitioning
- 7 Conclusion

# Definition

We consider **concrete domain**  $\mathbb{D} = \mathcal{P}(\mathbb{S})$  where

- $\mathbb{S} = \mathbb{L} \times \mathbb{M}$  where  $\mathbb{L}$  denotes the set of control states
- $\mathbb{M} = \mathbb{X} \rightarrow \mathbb{V}$

## State partitioning

A **state partitioning** abstraction is defined as the cardinal power of two abstractions  $(\mathbb{D}_0^\#, \sqsubseteq_0^\#, \gamma_0)$  and  $(\mathbb{D}_1^\#, \sqsubseteq_1^\#, \gamma_1)$  of sets of states:

- $(\mathbb{D}_0^\#, \sqsubseteq_0^\#, \gamma_0)$  defines the **partitions**
- $(\mathbb{D}_1^\#, \sqsubseteq_1^\#, \gamma_1)$  defines the **abstraction of each element of partitions**

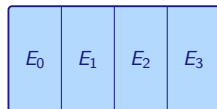
## Typical instances:

- either  $\mathbb{D}_1^\# = \mathcal{P}(\mathbb{S})$ , ordered with the inclusion
- or an abstraction of sets of memory states: numerical abstraction can be obtained by composing another abstraction on top of  $(\mathcal{P}(\mathbb{S}), \subseteq)$

## Use of a partition: intuition

We fix a partition  $\mathcal{U}$  of  $\mathcal{P}(\mathbb{S})$ :

- 1  $\forall E, E' \in \mathcal{U}, E \neq E' \implies E \cap E' = \emptyset$
- 2  $\mathbb{S} = \bigcup \mathcal{U}$



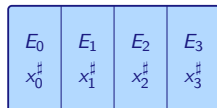
We can apply the **cardinal power construction**:

### State partitioning abstraction

We let  $\mathbb{D}_0^\# = \mathcal{U} \cup \{\perp, \top\}$  and  $\gamma_0 : E \mapsto E$ . Thus,  $\mathbb{D}_{\text{cp}}^\# = \mathcal{U} \rightarrow \mathbb{D}_1^\#$  and:

$$\begin{aligned} \gamma_{\text{cp}} : \mathbb{D}_{\text{cp}}^\# &\longrightarrow \mathbb{D} \\ X^\# &\longmapsto \{s \in \mathbb{S} \mid \forall E \in \mathcal{U}, s \in E \implies s \in \gamma_0(X^\#(E))\} \end{aligned}$$

- each  $E \in \mathcal{U}$  is attached to a piece of information in  $\mathbb{D}_1^\#$
- exercise: what happens if use only a **covering**, i.e., if we drop property 1 ?
- we will often focus on  $\mathcal{U}$  and drop  $\perp, \top$



# Application 1: flow sensitive abstraction

**Principle:** abstract separately the states at distinct control states

This is **what we have been often doing already**, without formalizing it for instance, using the **the interval abstract domain**:

$\ell_0$ : // assume $x \geq 0$	$\ell_0 \mapsto x : \top \wedge y : \top$
$\ell_1$ : <b>if</b> ( $x < 10$ ) {	$\ell_1 \mapsto x : [0, +\infty[ \wedge y : \top$
$\ell_2$ : $y = x - 2$ ;	$\ell_2 \mapsto x : [0, 9] \wedge y : \top$
$\ell_3$ : } <b>else</b> {	$\ell_3 \mapsto x : [0, 9] \wedge y : [-2, 7]$
$\ell_4$ : $y = 2 - x$ ;	$\ell_4 \mapsto x : [10, +\infty[ \wedge y : \top$
$\ell_5$ : }	$\ell_5 \mapsto x : [10, +\infty[ \wedge y : ] - \infty, -8]$
$\ell_6$ : ...	$\ell_6 \mapsto x : [0, +\infty[ \wedge y : ] - \infty, 7]$



## Application 1: flow sensitive abstraction

**Principle:** abstract separately the states at distinct control states

### Flow sensitive abstraction

We apply the cardinal power based partitioning abstraction with:

- $\mathcal{U} = \mathbb{L}$
- $\gamma_0 : \ell \mapsto \{\ell\} \times \mathbb{M}$

It is induced by partition  $\{\{\ell\} \times \mathbb{M} \mid \ell \in \mathbb{L}\}$

Then, if  $X^\sharp$  is an element of the reduced cardinal power,

$$\begin{aligned} \gamma_{\text{cp}}(X^\sharp) &= \{s \in \mathbb{S} \mid \forall x \in \mathbb{D}_0^\sharp, s \in \gamma_0(x) \implies s \in \gamma_1(X^\sharp(x))\} \\ &= \{(l, m) \in \mathbb{S} \mid m \in \gamma_1(X^\sharp(l))\} \end{aligned}$$

- after this abstraction step,  $\mathbb{D}_1^\sharp$  only needs to represent sets of memory states (numeric abstractions...)
- this abstraction step is *very common* as part of the design of abstract interpreters

## Application 1: flow insensitive abstraction

Flow sensitive abstraction is **sometimes too costly**:

- e.g., **ultra fast pointer analyses** (a few seconds for 1 MLOC) for compilation and program transformation
- **context insensitive** abstraction simply **collapses all control states**

### Flow insensitive abstraction

We apply the cardinal power based partitioning abstraction with:

- $\mathbb{D}_0^\# = \{\cdot\}$
- $\gamma_0 : \cdot \mapsto \mathbb{S}$
- $\mathbb{D}_1^\# = \mathcal{P}(M)$
- $\gamma_1 : M \mapsto \{(\ell, m) \mid \ell \in \mathbb{L}, m \in M\}$

It is induced by a trivial partition of  $\mathcal{P}(\mathbb{S})$

## Application 1: flow insensitive abstraction

We compare with **flow sensitive abstraction**:

$\ell_0$ : // assume $x \geq 0$	$\ell_0 \mapsto x : \top \wedge y : \top$
$\ell_1$ : <b>if</b> ( $x < 10$ ) {	$\ell_1 \mapsto x : [0, +\infty[ \wedge y : \top$
$\ell_2$ : $y = x - 2$ ;	$\ell_2 \mapsto x : [0, 9] \wedge y : \top$
$\ell_3$ : } <b>else</b> {	$\ell_3 \mapsto x : [0, 9] \wedge y : [-2, 7]$
$\ell_4$ : $y = 2 - x$ ;	$\ell_4 \mapsto x : [10, +\infty[ \wedge y : \top$
$\ell_5$ : }	$\ell_5 \mapsto x : [10, +\infty[ \wedge y : ] - \infty, -8]$
$\ell_6$ : ...	$\ell_6 \mapsto x : [0, +\infty[ \wedge y : ] - \infty, 7]$

- the **best global information** is  $x : \top \wedge y : \top$  (**very imprecise**)
- even if we exclude the point before the assume, we get  $x : [0, +\infty[ \wedge y : \top$  (still **very imprecise**)

For a few specific applications flow insensitive is ok

In **most cases** (e.g., numeric properties), flow sensitive is absolutely needed

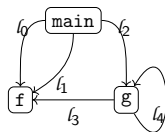
## Application 2: context sensitive abstraction

We consider programs **with procedures**

### Example:

```

void main(){...  $l_0$  : f(); ...  $l_1$  : f(); ...  $l_2$  : g() ...}
void f(){...}
void g(){if(...){ $l_3$  : f()}else{ $l_4$  : g()}}
```



- assumption: **flow sensitive abstraction** used inside each function
- we need to also describe the **call stack state**

### Call string

Thus,  $\mathbb{S} = \mathbb{K} \times \mathbb{L} \times \mathbb{M}$ , where  $\mathbb{K}$  is the set of **call strings**

$\kappa \in \mathbb{K}$	calling contexts
$\kappa ::= \epsilon$	empty call stack
$\quad   (f, l) \cdot \kappa$	call to $f$ from stack $\kappa$ at point $l$

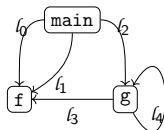
## Application 2: context sensitive abstraction, $\infty$ -CFA

### Fully context sensitive abstraction ( $\infty$ -CFA)

- $\mathbb{D}_0^\# = \mathbb{K} \times \mathbb{L}$
- $\gamma_0 : (\kappa, \ell) \mapsto \{(\kappa, \ell, m) \mid m \in \mathbb{M}\}$

```

void main(){...  $\ell_0$  : f(); ...  $\ell_1$  : f(); ...  $\ell_2$  : g() ...}
void f(){...}
void g(){if(...){ $\ell_3$  : f()}else{ $\ell_4$  : g()}}
  
```



### Abstract contexts in function f:

$(\ell_0, f) \cdot \epsilon$ ,  $(\ell_1, f) \cdot \epsilon$ ,  $(\ell_4, f) \cdot (\ell_2, g) \cdot \epsilon$ ,  
 $(\ell_4, f) \cdot (\ell_3, g) \cdot (\ell_2, g) \cdot \epsilon$ ,  $(\ell_4, f) \cdot (\ell_3, g) \cdot (\ell_3, g) \cdot (\ell_2, g) \cdot \epsilon$ , ...

- one invariant per calling context, **very precise** (used, e.g., in Astrée)
- **infinite in presence of recursion** (i.e., not practical in this case)

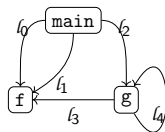
## Application 2: context sensitive abstraction, 0-CFA

### Non context sensitive abstraction (0-CFA)

- $\mathbb{D}_0^\# = \mathbb{L}$
- $\gamma_0 : \ell \mapsto \{(\kappa, \ell, m) \mid \kappa \in \mathbb{K}, m \in \mathbb{M}\}$

```

void main(){...  $\ell_0$  : f(); ...  $\ell_1$  : f(); ...  $\ell_2$  : g() ...}
void f(){...}
void g(){if(...){ $\ell_3$  : f()}else{ $\ell_4$  : g()}}
  
```



**Abstract contexts** in **function**  $f$  are of the form  $(?, f) \cdot \dots$ ,

- 0-CFA merges **all** calling contexts to a same procedure, **very coarse** abstraction
- but is **usually quite efficient to compute**

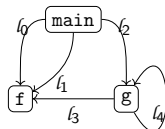
## Application 2: context sensitive abstraction, $k$ -CFA

### Partially context sensitive abstraction ( $k$ -CFA)

- $\mathbb{D}_0^\# = \{\kappa \in \mathbb{K} \mid \mathbf{length}(\kappa) \leq k\} \times \mathbb{L}$
- $\gamma_0 : (\kappa, \ell) \mapsto \{(\kappa \cdot \kappa', \ell, m) \mid \kappa' \in \mathbb{K}, m \in \mathbb{M}\}$

```

void main(){...  $\ell_0 : f()$ ; ...  $\ell_1 : f()$ ; ...  $\ell_2 : g()$  ...}
void f(){...}
void g(){if(...){ $\ell_3 : f()$ }else{ $\ell_4 : g()$ }}
  
```



### Abstract contexts in function $f$ , in 2-CFA:

$$(\ell_0, f) \cdot \epsilon, (\ell_1, f) \cdot \epsilon, (\ell_4, f) \cdot (\ell_3, g) \cdot (? , g) \cdot \dots, (\ell_4, f) \cdot (\ell_4, g) \cdot (? , g) \cdot \dots$$

- usually **intermediate** level of precision and efficiency
- can be applied to programs with **recursive procedures**

## Application 3: partitioning by a boolean condition

- so far, we only used abstractions of the context to partition
- we now consider abstractions of memory states properties

### Function guided memory states partitioning

We let:

- $\mathbb{D}_0^\# = A$  where  $A$  finite set is a finite set of values / properties
- $\phi : \mathbb{M} \rightarrow A$  maps each store to its property
- $\gamma_0$  is of the form  $(a \in A) \mapsto \{(\ell, m) \in \mathbb{S} \mid \phi(m) = a\}$

**Common choice** for  $A$ : **the set of boolean values**  $\mathbb{B}$

(or another finite set of values —convenient for enum types!)

**Many choices** for function  $\phi$  are possible:

- **value** of one or several variables (boolean or scalar)
- **sign** of a variable
- ...



## Application 3: partitioning by a boolean condition

We assume:

- $\mathbb{X} = \mathbb{X}_{\text{bool}} \uplus \mathbb{X}_{\text{int}}$ , where  $\mathbb{X}_{\text{bool}}$  (resp.,  $\mathbb{X}_{\text{int}}$ ) collects **boolean** (resp., **integer**) variables
- $\mathbb{X}_{\text{bool}} = \{\mathbf{b}_0, \dots, \mathbf{b}_{k-1}\}$
- $\mathbb{X}_{\text{int}} = \{\mathbf{x}_0, \dots, \mathbf{x}_{l-1}\}$

Thus,  $\mathbb{M} = \mathbb{X} \rightarrow \mathbb{V} \equiv (\mathbb{X}_{\text{bool}} \rightarrow \mathbb{V}_{\text{bool}}) \times (\mathbb{X}_{\text{int}} \rightarrow \mathbb{V}_{\text{int}}) \equiv \mathbb{V}_{\text{bool}}^k \times \mathbb{V}_{\text{int}}^l$

### Boolean partitioning abstract domain

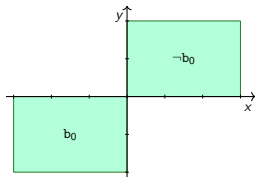
We apply the cardinal power abstraction, with a domain of partitions defined by a function, with:

- $A = \mathbb{B}^k$
- $\phi(m) = (m(\mathbf{b}_0), \dots, m(\mathbf{b}_{k-1}))$
- we let  $(\mathbb{D}_1^\sharp, \sqsubseteq_1^\sharp, \gamma_1)$  be any **numerical abstract domain** for  $\mathcal{P}(\mathbb{V}_{\text{int}}^l)$

## Application 3: example

With  $\mathbb{X}_{\text{bool}} = \{b_0, b_1\}$ ,  $\mathbb{X}_{\text{int}} = \{x, y\}$ , we can express:

$$\left\{ \begin{array}{ll} b_0 \wedge b_1 & \implies x_0 \in [-3, 0] \wedge y \in [-2, 0] \\ b_0 \wedge \neg b_1 & \implies x_0 \in [-3, 0] \wedge y \in [-2, 0] \\ \neg b_0 \wedge b_1 & \implies x_0 \in [0, 3] \wedge y \in [0, 3] \\ \neg b_0 \wedge \neg b_1 & \implies x_0 \in [0, 3] \wedge y \in [0, 3] \end{array} \right.$$



- this abstract value expresses a **relation** between  $b_0$  and  $x, y$  (which induces a relation between  $x$  and  $y$ )
- **alternative**: partition with respect to only **some** variables e.g., here  $b_0$  only as  $b_1$  is irrelevant
- **typical representation** of abstract values: based on some kind of decision trees (variants of BDDs)

## Application 3: example

- Left side abstraction **shown in blue**: boolean partitioning for  $b_0, b_1$
- Right side abstraction **shown in green**: interval abstraction
- We omit the cases of the form  $P \implies \perp \dots$

```

bool b0, b1;
int x, y;      (uninitialized)
b0 = x ≥ 0;
    (b0  $\implies$  x ≥ 0) ∧ (¬b0  $\implies$  x < 0)
b1 = x ≤ 0;
    (b0 ∧ b1  $\implies$  x = 0) ∧ (b0 ∧ ¬b1  $\implies$  x > 0) ∧ (¬b0 ∧ b1  $\implies$  x < 0)
if(b0 && b1){
    (b0 ∧ b1  $\implies$  x = 0)
    y = 0;
    (b0 ∧ b1  $\implies$  x = 0 ∧ y = 0)
} else{
    (b0 ∧ ¬b1  $\implies$  x > 0) ∧ (¬b0 ∧ b1  $\implies$  x < 0)
    y = 100/x;
    (b0 ∧ ¬b1  $\implies$  x > 0 ∧ y ≥ 0) ∧ (¬b0 ∧ b1  $\implies$  x < 0 ∧ y ≤ 0)
}
  
```

## Application 3: partitioning by the sign of a variable

We now consider a **semantic property**: the **sign of a variable**

We assume:

- $\mathbb{X} = \mathbb{X}_{\text{int}}$ , i.e., all variables have **integer** type
- $\mathbb{X}_{\text{int}} = \{x_0, \dots, x_{l-1}\}$

Thus,  $\mathbb{M} = \mathbb{X} \rightarrow \mathbb{V} \equiv \mathbb{V}_{\text{int}}^l$

### Sign partitioning abstract domain

We apply the cardinal power abstraction, with a domain of partition defined by a function, with:

- $A = \{[< 0], [= 0], [> 0]\}$
- $\phi(m) = \begin{cases} [< 0] & \text{if } x_0 < 0 \\ [= 0] & \text{if } x_0 = 0 \\ [> 0] & \text{if } x_0 > 0 \end{cases}$
- $(\mathbb{D}_1^\#, \sqsubseteq_1^\#, \gamma_1)$  an abstraction of  $\mathcal{P}(\mathbb{V}_{\text{int}}^{l-1})$  (no need to abstract  $x_0$  twice)

## Application 3: example

- Sign abstraction fixing partitions **shown in blue**
- Right side abstraction **shown in green**: interval abstraction
- We omit the cases of the form  $P \implies \perp \dots$

```

int x ∈ ℤ;
int s;
int y;
if(x ≥ 0){
    (x < 0 ⇒ ⊥) ∧ (x = 0 ⇒ ⊤) ∧ (x > 0 ⇒ ⊤)
    s = 1;
    (x < 0 ⇒ ⊥) ∧ (x = 0 ⇒ s = 1) ∧ (x > 0 ⇒ s = 1)
} else {
    (x < 0 ⇒ ⊤) ∧ (x = 0 ⇒ ⊥) ∧ (x > 0 ⇒ ⊥)
    s = -1;
    (x < 0 ⇒ s = -1) ∧ (x = 0 ⇒ ⊥) ∧ (x > 0 ⇒ ⊥)
}
(x < 0 ⇒ s = -1) ∧ (x = 0 ⇒ s = 1) ∧ (x > 0 ⇒ s = 1)
① y = x/s;
(x < 0 ⇒ s = -1 ∧ y > 0) ∧ (x = 0 ⇒ s = 1 ∧ y = 0) ∧ (x > 0 ⇒ s = 1 ∧ y > 0)
② assert(y ≥ 0);

```

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- 7 Conclusion

# Computation of abstract semantics and partitioning

- we first consider **partitioning by control states** and let  $\mathbb{L} = \{\ell_0, \dots, \ell_s\}$
- we rely on the two steps partitioning abstraction i.e., to be **composed** with an abstraction of  $\mathcal{P}(\mathbb{M})$
- the techniques shown below **extend to other forms of partitioning**

The first abstraction defines to a **Galois connection**:

$$(\mathcal{P}(\mathbb{L} \times \mathbb{M}), \subseteq) \begin{array}{c} \xleftarrow{\gamma_{\text{part}}} \\ \xrightarrow{\alpha_{\text{part}}} \end{array} (\mathbb{D}_{\text{part}}^{\#}, \dot{\subseteq})$$

where  $\mathbb{D}_{\text{part}}^{\#} = \mathbb{L} \rightarrow \mathcal{P}(\mathbb{M})$  and:

$$\begin{array}{lcl} \alpha_{\text{part}} : & \mathcal{P}(\mathbb{L} \times \mathbb{M}) & \longrightarrow \mathbb{D}_{\text{part}}^{\#} \\ & \mathcal{E} & \longmapsto \lambda(\ell \in \mathbb{L}) \cdot \{m \in \mathbb{M} \mid (\ell, m) \in \mathcal{E}\} \\ \gamma_{\text{part}} : & \mathbb{D}_{\text{part}}^{\#} & \longrightarrow \mathcal{P}(\mathbb{L} \times \mathbb{M}) \\ & \mathbb{X}^{\#} & \longmapsto \{(\ell, m) \in \mathbb{S} \mid m \in \mathbb{X}^{\#}(\ell)\} \end{array}$$

We first study the **“computational form”** of this semantics (fixpoint)

## Fixpoint form of a partitioned semantics

- we consider a **transition system**  $\mathcal{S} = (\mathbb{S}, \rightarrow, \mathbb{S}_I)$  where  $\mathbb{S}_I = \{\ell_0\} \times \mathbb{M}$
- the **reachable states** are computed as  $\llbracket \mathcal{S} \rrbracket_{\mathcal{R}} = \mathbf{lfp}_{\mathbb{S}_I} F_{\mathcal{R}}$  where

$$F_{\mathcal{R}} : \mathcal{P}(\mathbb{S}) \longrightarrow \mathcal{P}(\mathbb{S})$$

$$X \longmapsto \{s \in \mathbb{S} \mid \exists s' \in X, s' \rightarrow s\}$$

### Semantic function over the partitioned system

We let  $F_{\text{part}}$  be defined over  $\mathbb{D}_{\text{part}}^{\#}$  by:

$$F_{\text{part}} : \mathbb{D}_{\text{part}}^{\#} \longrightarrow \mathbb{D}_{\text{part}}^{\#}$$

$$X^{\#} \longmapsto \lambda(\ell \in \mathbb{L}). \{m \in \delta_{\ell, \ell'}(m') \mid \ell' \in \mathbb{L}, m' \in X^{\#}(\ell')\}$$

$$\{m \in \mathbb{M} \mid \exists \ell' \in \mathbb{L}, \exists m' \in X^{\#}(\ell'),$$

where  $\delta_{\ell, \ell'}(m') = \{m' \in \mathbb{M} \mid (\ell, m) \rightarrow (\ell', m')\}$ .

Then  $F_{\text{part}} \circ \alpha_{\text{part}} = \alpha_{\text{part}} \circ F$  and  $\alpha_{\text{part}}(\llbracket \mathcal{S} \rrbracket_{\mathcal{R}}) = \mathbf{lfp}_{\alpha_{\text{part}}(\mathbb{S}_I)} F_{\text{part}}$



## Concrete equations form of a partitioned semantics

We look for **an equivalent set of abstract equations** following the intuition:

- at  $l_0$ , we observe any memory state (start **from any memory state**)
- at  $l \neq l_0$ , we observe states reached from a predecessor of  $l$ , **in a single step**

### Set of concrete semantic equations

We define the **set of concrete semantic equations** by:

$$\begin{cases} \mathcal{M}_0 &= \mathbb{M} \\ \mathcal{M}_1 &= \bigcup_i \{m \in \mathbb{M} \mid \exists m' \in \mathcal{M}_i, (l_i, m') \rightarrow (l_1, m)\} \\ &\vdots \\ \mathcal{M}_s &= \bigcup_i \{m \in \mathbb{M} \mid \exists m' \in \mathcal{M}_i, (l_i, m') \rightarrow (l_s, m)\} \end{cases}$$

where variables  $\mathcal{M}_0, \dots, \mathcal{M}_s$  range over set of memory states, i.e., we look for solutions where  $\mathcal{M}_i \subseteq \mathbb{M}$

# Concrete equations form of a partitioned semantics

In the following, we note:

$$F_j : (\mathcal{M}_1, \dots, \mathcal{M}_s) \mapsto \bigcup_i \{m \in \mathbb{M} \mid \exists m' \in \mathcal{E}_i, (l_i, m') \rightarrow (l_j, m)\}$$

## Computational form of the concrete semantics

$\alpha_{\text{part}}(\llbracket \mathcal{S} \rrbracket_{\mathcal{R}})$  is the least solution of the system

$$\begin{cases} \mathcal{M}_0 & = & \mathbb{M} \\ \mathcal{M}_1 & = & F_1(\mathcal{M}_1, \dots, \mathcal{M}_s) \\ & \vdots & \\ \mathcal{M}_s & = & F_s(\mathcal{M}_1, \dots, \mathcal{M}_s) \end{cases}$$

- **proof:** the system defines exactly the fixpoints of  $F_{\text{part}}$
- iterating  $F_{\text{part}}$  boils down to running the system from  $(\mathbb{M}, \emptyset, \dots, \emptyset)$
- we cannot “implement” this (convergence issue !), but we can do the same in the abstract

# Abstract equations form of a partitioned semantics

We can now move on **to the abstract level**:

- $\mathcal{M}_i^\sharp$  denotes an element of  $\mathbb{D}_1^\sharp$
- **abstract functions**  $F_i^\sharp : (\mathbb{D}_1^\sharp)^s \rightarrow \mathbb{D}_1^\sharp$  over-approximate the concrete functions  $F_i$

## Abstract equations

A solution of the system

$$\left\{ \begin{array}{l} \mathcal{M}_0^\sharp \sqsupseteq \top \\ \mathcal{M}_1^\sharp \sqsupseteq F_1^\sharp(\mathcal{M}_1^\sharp, \dots, \mathcal{M}_s^\sharp) \\ \vdots \\ \mathcal{M}_s^\sharp \sqsupseteq F_s^\sharp(\mathcal{M}_1^\sharp, \dots, \mathcal{M}_s^\sharp) \end{array} \right.$$

over-approximates  $\alpha_{\text{part}}(\llbracket \mathcal{S} \rrbracket_{\mathcal{R}})$

# Partitioned systems and fixpoint computation

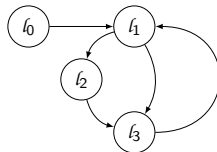
For now, we overlook the convergence issue, and focus on **the computation of the iterates**

**Typical properties of real transition systems:**

- in practice  $F_i$  depends **only on a few of its arguments**  
i.e.,  $\mathcal{E}_k$  depends only on the predecessors of  $l_k$  in the control flow graph of the program under consideration
- also,  $F_i$  is  **$\emptyset$ -strict**:  
if there is no predecessor, there is no transition...

**Example** of a simple system of abstract equations:

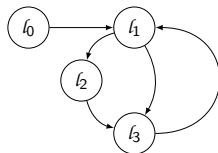
$$\left\{ \begin{array}{l} \mathcal{M}_0 = \mathbb{M} \\ \mathcal{M}_1 = F_1(\mathcal{M}_0) \cup G_1(\mathcal{M}_3) \\ \mathcal{M}_2 = F_2(\mathcal{M}_1) \\ \mathcal{M}_3 = F_3(\mathcal{M}_1) \cup G_3(\mathcal{M}_2) \end{array} \right.$$



# Example concrete iteration

## System:

$$\begin{cases} \mathcal{M}_0 &= \mathbb{M} \\ \mathcal{M}_1 &= F_1(\mathcal{M}_0) \cup G_1(\mathcal{M}_3) \\ \mathcal{M}_2 &= F_2(\mathcal{M}_1) \\ \mathcal{M}_3 &= F_3(\mathcal{M}_1) \cup G_3(\mathcal{M}_2) \end{cases}$$



## Iterates for $(\mathcal{M}_0, \mathcal{M}_1, \mathcal{M}_2, \mathcal{M}_3)$ :

rank	iterate
0	$(\mathbb{M}, \emptyset, \emptyset, \emptyset)$
1	$(\mathbb{M}, F_1(\mathbb{M}), \emptyset, \emptyset)$
2	$(\mathbb{M}, F_1(\mathbb{M}), F_2 \circ F_1(\mathbb{M}), F_3 \circ F_1(\mathbb{M}))$
3	$(\mathbb{M}, F_1(\mathbb{M}) \cup G_1 \circ F_3 \circ F_1(\mathbb{M}), F_2 \circ F_1(\mathbb{M}), F_3 \circ F_1(\mathbb{M}) \cup G_3 \circ F_2 \circ F_1(\mathbb{M}))$
4	...

- we highlight in red the “computations” that are done
- there is a lot of **un-necessary recomputation**

# Example computation of abstract iterates

Using an abstraction of sets of memory states:

$$\begin{cases} \mathcal{M}_0 &= \mathbb{M} \\ \mathcal{M}_1 &= F_1(\mathcal{M}_0) \cup G_1(\mathcal{M}_3) \\ \mathcal{M}_2 &= F_2(\mathcal{M}_1) \\ \mathcal{M}_3 &= F_3(\mathcal{M}_1) \cup G_3(\mathcal{M}_2) \end{cases}$$

- we assume **abstract domain**  $\mathbb{M}^\sharp$ , over-approximating  $\mathcal{P}(\mathbb{M})$
- we assume **abstract transfer functions**  $F_i^\sharp : \mathbb{M}^\sharp \rightarrow \mathbb{M}^\sharp$  over-approximates  $F_i$

Abstract iterates:

rank	iterate
0	$(\mathbb{M}, \emptyset, \emptyset, \emptyset)$
1	$(\mathbb{M}, F_1^\sharp(\mathbb{M}), \emptyset, \emptyset)$
2	$(\mathbb{M}, F_1^\sharp(\mathbb{M}), F_2^\sharp \circ F_1^\sharp(\mathbb{M}), F_3^\sharp \circ F_1^\sharp(\mathbb{M}))$
3	$(\mathbb{M}, F_1^\sharp(\mathbb{M}) \sqcup G_1^\sharp \circ F_3^\sharp \circ F_1^\sharp(\mathbb{M}), F_2^\sharp \circ F_1^\sharp(\mathbb{M}), F_3^\sharp \circ F_1^\sharp(\mathbb{M}) \sqcup G_3^\sharp \circ F_2^\sharp \circ F_1^\sharp(\mathbb{M}))$
4	...

The **same issue** occurs: **recomputation of abstract iterates**

# Chaotic iterations: principle

## Fairness

Let  $K$  be a finite set. A sequence  $(k_n)_{n \in \mathbb{N}}$  of elements of  $K$  is **fair** if and only if, for all  $k \in K$ , the set  $\{n \in \mathbb{N} \mid k_n = k\}$  is infinite.

- alternate definition:

$$\forall k \in K, \forall n_0 \in \mathbb{N}, \exists n \in \mathbb{N}, n > n_0 \wedge k_n = k$$

- i.e., all elements of  $K$  is encountered infinitely often

## Chaotic iterations

A **chaotic sequence of iterates** is a sequence of abstract iterates  $(X_n^\#)_{n \in \mathbb{N}}$  in  $\mathbb{D}_{\text{part}}^\#$  such that there exists a sequence  $(k_n)_{n \in \mathbb{N}}$  of elements of  $\{1, \dots, s\}$  (we disregard component 0, which is  $\top$ ) such that:

$$X_{n+1}^\# = \lambda(l_i \in \mathbb{L}). \begin{cases} X_n^\#(l_i) & \text{if } i \neq k_n \\ X_n^\#(l_i) \sqcup F^\#(X_n^\#(l_1), \dots, X_n^\#(l_s)) & \text{if } i = k_n \end{cases}$$

# Chaotic iterations: soundness

## Soundness

Assuming the abstract lattice satisfies the ascending chain condition, any sequence of chaotic iterates computes the abstract fixpoint:

$$\lim (X_n^\sharp)_{n \in \mathbb{N}} = \alpha_{\text{part}}(\llbracket \mathcal{S} \rrbracket_{\mathcal{R}})$$

- **proof:** exercise
- **benefit:** **no more useless recomputation**
- **back to the example,** where **recomputed components** are **in red:**

rank	iterate
0	$(\mathbb{M}, \emptyset, \emptyset, \emptyset)$
1	$(\mathbb{M}, F_1^\sharp(\mathbb{M}), \emptyset, \emptyset)$
2	$(\mathbb{M}, F_1^\sharp(\mathbb{M}), F_2^\sharp \circ F_1^\sharp(\mathbb{M}), \emptyset)$
3	$(\mathbb{M}, F_1^\sharp(\mathbb{M}), F_2^\sharp \circ F_1^\sharp(\mathbb{M}), F_3^\sharp \circ F_1^\sharp(\mathbb{M}))$
4	$(\mathbb{M}, F_1^\sharp(\mathbb{M}), F_2^\sharp \circ F_1^\sharp(\mathbb{M}), F_3^\sharp \circ F_1^\sharp(\mathbb{M}) \sqcup G_3^\sharp \circ F_2^\sharp \circ F_1^\sharp(\mathbb{M}))$
5	...



# Chaotic iterations: work-list algorithm

## Work-list algorithms

Principle:

- 1 maintain a **queue of partitions to update**
- 2 initialize the queue with the **entry label** of the program and the local invariant at that point at  $\top$
- 3 for each iterate, **update the first partition in the queue** (after removing it), and add to the queue all its successors *unless* the updated invariant is equal to the former one
- 4 **terminate** when the queue is **empty**

This algorithm implements a **chaotic iteration** strategy, thus it is sound

- **benefit: no more useless recomputation**
- implemented in **many static analyzers**

# Work-list algorithm

**Pseudo code implementation**, where  $\delta_{\ell, \ell'}^\#$  denotes the transfer function from  $\ell$  to  $\ell'$ , that over-approximates  $\delta_{\ell, \ell'}$ :

```

to_propagate ← {initial states}
 $\mathcal{E}_{\text{initial}}^\# \leftarrow \top$ 
while(to_propagate  $\neq \emptyset$ ){
  pick  $\ell \in$  to_propagate
  to_propagate = to_propagate  $\setminus \{\ell\}$ 
  for( $\ell'$  successor of  $\ell$  in the CFG){
     $y^\# \leftarrow \delta_{\ell, \ell'}^\#(\mathcal{E}_\ell^\#)$ 
    if( $\neg(y^\# \sqsubseteq^\# \mathcal{E}_{\ell'}^\#)$ ){
       $\mathcal{E}_{\ell'}^\# = \mathcal{E}_{\ell'}^\# \sqcup^\# y^\#$ 
      to_propagate = to_propagate  $\cup \{\ell'\}$ 
    }
  }
}

```

# Selection of a set of widening points for a partitioned system

## Assumptions:

- abstract domain  $\mathbb{D}_{\text{num}}^{\sharp}$ , with concretization  $\gamma_{\text{num}} : \mathbb{D}_{\text{num}}^{\sharp} \rightarrow \mathcal{P}(\mathbb{M})$ ,  
**does not satisfy ascending chain condition**
- $\mathbb{D}_{\text{num}}^{\sharp}$  provides **widening** operator  $\nabla$

How to adapt the chaotic iteration strategy, i.e. **guarantee termination and soundness** ?

## Enforcing termination of chaotic iterates

Let  $K_{\nabla} \subseteq \{1, \dots, s\}$  such that each cycle in the control flow graph of the program contains at least one point in  $K_{\nabla}$ ; we define the chaotic abstract iterates with widening as follows:

$$X_{n+1}^{\sharp} = \lambda(l_i \in \mathbb{L}). \begin{cases} X_n^{\sharp}(l_i) & \text{if } i \neq k_n \\ X_n^{\sharp}(l_i) \sqcup F^{\sharp}(X_n^{\sharp}(l_1), \dots, X_n^{\sharp}(l_s)) & \text{if } i = k_n \wedge l_i \notin K_{\nabla} \\ X_n^{\sharp}(l_i) \nabla F^{\sharp}(X_n^{\sharp}(l_1), \dots, X_n^{\sharp}(l_s)) & \text{if } i = k_n \wedge l_i \in K_{\nabla} \end{cases}$$

# Selection of a set of widening points for a partitioned system

## Soundness and termination

Under the assumption of a fair iteration strategy, sequence  $(X_n^\#)_{n \in \mathbb{N}}$  terminates and computes a sound abstract post-fixpoint:

$$\exists n_0 \in \mathbb{N}, \begin{cases} \forall n \geq n_0, X_{n_0}^\# = X_n^\# \\ \llbracket S \rrbracket_{\mathcal{R}} \subseteq \gamma_{\text{part}}(X_{n_0}) \end{cases}$$

**Proof:** exercise

**Algorithm for iteration with widening:** exercise

# Outline

- 1 Introduction
- 2 Imprecisions in convex abstractions
- 3 Disjunctive completion
- 4 Cardinal power and partitioning abstractions
- 5 State partitioning**
  - Definition and examples
  - Control states partitioning and iteration techniques
  - Abstract interpretation with boolean partitioning
- 6 Trace partitioning
- 7 Conclusion

# Computation of abstract semantics and partitioning

We now **compose two forms of partitioning**

- by **control states** (as previously), using a chaotic iteration strategy
- by **the values of the boolean variables**

Thus, **the abstract domain is of the form**

$$\mathbb{L} \longrightarrow (\mathbb{V}_{\text{bool}}^k \longrightarrow \mathbb{D}_1^\sharp)$$

- we could do a partitioning by  $\mathbb{L} \times \mathbb{V}_{\text{bool}}^k$
- yet, it is not practical, as transitions from “boolean states” are not known before the analysis
- **data types** skeleton:

```

type abs1 = ... (* abstract elements of  $\mathbb{D}_1^\sharp$  *)
type abs_state = ... (*
    boolean trees with elements of type abs1 at the leaves *)
type abs_cp = (labels, abs_state) Map.t
  
```

# Abstract operations

## Transfer functions:

we seek, for all pair  $\ell, \ell' \in \mathbb{L}$  for an approximation  $\delta_{\ell, \ell'}^\sharp$  of

$$\begin{aligned} \delta_{\ell, \ell'} : \mathbb{M} &\longrightarrow \mathcal{P}(\mathbb{M}) \\ m &\longmapsto \{m' \in \mathbb{M} \mid (\ell, m) \rightarrow (\ell', m')\} \end{aligned}$$

This includes:

- **assignment to scalar**, e.g.,  $x = 1 - x$ ;
- **assignment to boolean**, e.g.,  $b_0 = x \leq 7$
- **scalar test**, e.g., **if**( $x \geq 8$ ) ...
- **boolean test**, e.g., **if**( $\neg b_1$ ) ...

**Lattice operations:** inclusion check, join, widening

## Transfer functions: assignment to scalar

**Assignment**  $\ell_0 : x = e$ ;  $\ell_1$  affecting **only integer variables** (i.e.,  $e$  depends only on  $x_0, \dots, x_I$ ):

- **example**:  $x = 1 - x$ ;
- **concrete transition**  $\delta_{\ell_0, \ell_1}$  defined by

$$\delta_{\ell_0, \ell_1}(m) = \{m[x \leftarrow \llbracket e \rrbracket(m)]\}$$

- the values of the boolean variables are unchanged  
thus the partitions are preserved (**pointwise** transfer function):

$$\text{assign}_{\text{cp}}(x, e, X^\#) = \lambda(z^\# \in \mathbb{D}_0^\#) \cdot \text{assign}_1(x, e, X^\#(z^\#))$$

### Soundness

If  $\text{assign}_1$  is sound, so is  $\text{assign}_{\text{cp}}$ , in the sense that:

$$\forall X^\# \in \mathbb{D}_{\text{cp}}^\#, \forall m \in \gamma_{\text{cp}}(X^\#), m[x \leftarrow \llbracket e \rrbracket(m)] \in \gamma_{\text{cp}}(\text{assign}_{\text{cp}}(x, e, X^\#))$$



# Transfer functions: assignment to scalar, example

- **abstract precondition:**

$$\left\{ \begin{array}{l} \mathbf{b} \Rightarrow \mathbf{x} \geq 0 \\ \wedge \quad \neg \mathbf{b} \Rightarrow \mathbf{x} \leq 0 \end{array} \right\}$$

- **statement:**

$$\mathbf{x} = 1 - \mathbf{x};$$

- **abstract post-condition:**

$$\begin{aligned} \mathit{assign}_{\mathbf{cp}} \left( \mathbf{x}, 1 - \mathbf{x}, \left\{ \begin{array}{l} \mathbf{b} \Rightarrow \mathbf{x} \geq 0 \\ \wedge \quad \neg \mathbf{b} \Rightarrow \mathbf{x} \leq 0 \end{array} \right\} \right) \\ = \left\{ \begin{array}{l} \mathbf{b} \Rightarrow \mathbf{x} \leq 1 \\ \wedge \quad \neg \mathbf{b} \Rightarrow \mathbf{x} \geq 1 \end{array} \right\} \end{aligned}$$

## Transfer functions: scalar test

**Condition test**  $\ell_0 : \text{if}(c)\{\ell_1 : \dots\}$  affecting **only scalar variables** (i.e.,  $c$  depends only on  $x_0, \dots, x_l$ ):

- **example:**  $\text{if}(x \geq 8) \dots$
- **concrete transition**  $\delta_{\ell_0, \ell_1}$  defined by

$$\delta_{\ell_0, \ell_1}(m) = \begin{cases} \{m\} & \text{if } \llbracket c \rrbracket(m) = \text{TRUE} \\ \emptyset & \text{if } \llbracket c \rrbracket(m) = \text{FALSE} \end{cases}$$

- the partitions are preserved, thus we get a **pointwise** transfer function:

$$\text{test}_{\text{cp}}(c, X^\sharp) = \lambda(z^\sharp \in \mathbb{D}_0^\sharp) \cdot \text{test}_1(c, X^\sharp(z^\sharp))$$

### Soundness

If  $\text{test}_1$  is sound, so is  $\text{test}_{\text{cp}}$ , in the sense that:

$$\forall X^\sharp \in \mathbb{D}_{\text{cp}}^\sharp, \forall m \in \gamma_{\text{cp}}(X^\sharp), \llbracket c \rrbracket(m) = \text{TRUE} \implies m \in \gamma_{\text{cp}}(\text{test}_{\text{cp}}(x, e, X^\sharp))$$

# Transfer functions: scalar test, example

- **abstract pre-condition:**

$$\left\{ \begin{array}{l} \mathbf{b} \Rightarrow \mathbf{x} \geq 0 \\ \wedge \neg \mathbf{b} \Rightarrow \mathbf{x} \leq 0 \end{array} \right\}$$

- **statement:**

**if**( $\mathbf{x} \geq 8$ )...

- **abstract post-condition:**

$$\text{test}_{\mathbf{cp}} \left( \mathbf{x} \geq 8, \left\{ \begin{array}{l} \mathbf{b} \Rightarrow \mathbf{x} \geq 0 \\ \wedge \neg \mathbf{b} \Rightarrow \mathbf{x} \leq 0 \end{array} \right\} \right) = \left\{ \begin{array}{l} \mathbf{b} \Rightarrow \mathbf{x} \geq 8 \\ \wedge \neg \mathbf{b} \Rightarrow \perp \end{array} \right\}$$

## Transfer functions: boolean condition test

**Condition test**  $\ell_0 : \text{if}(c)\{\ell_1 : \dots\}$  affecting **only boolean variables** (i.e.,  $c$  depends only on  $b_0, \dots, b_k$ ):

- **example:**  $\text{if}(\neg b_1) \dots$
- then, we simply need to filter the boolean partitions **satisfying**  $c$ :

$$\text{test}_{\text{cp}}(c, X^\#) = \lambda(z^\# \in \mathbb{D}_0^\#) . \begin{cases} X^\#(z^\#) & \text{if } \text{test}_0(c, X^\#(z^\#)) \neq \perp_0 \\ \perp_1 & \text{otherwise} \end{cases}$$

### Soundness

If  $\text{test}_0$  is sound, so is  $\text{test}_{\text{cp}}$ , in the sense that:

$$\forall X^\# \in \mathbb{D}_{\text{cp}}^\#, \forall m \in \gamma_{\text{cp}}(X^\#), \llbracket c \rrbracket(m) = \text{TRUE} \implies m \in \gamma_{\text{cp}}(\text{test}_{\text{cp}}(x, e, X^\#))$$

# Transfer functions: boolean condition test, example

- abstract pre-condition:**

$$\left\{ \begin{array}{l} b_0 \wedge b_1 \Rightarrow 15 \leq x \\ \wedge \quad b_0 \wedge \neg b_1 \Rightarrow 9 \leq x \leq 14 \\ \wedge \quad \neg b_0 \wedge b_1 \Rightarrow 6 \leq x \leq 8 \\ \wedge \quad \neg b_0 \wedge \neg b_1 \Rightarrow x \leq 5 \end{array} \right\}$$
- statement:**  $\text{if}(\neg b_1) \dots$
- abstract post-condition:**

$$\begin{aligned} \text{test}_{\text{cp}} \left( \neg b_1, \left\{ \begin{array}{l} b_0 \wedge b_1 \Rightarrow 15 \leq x \\ \wedge \quad b_0 \wedge \neg b_1 \Rightarrow 9 \leq x \leq 14 \\ \wedge \quad \neg b_0 \wedge b_1 \Rightarrow 6 \leq x \leq 8 \\ \wedge \quad \neg b_0 \wedge \neg b_1 \Rightarrow x \leq 5 \end{array} \right\} \right) \\ = \left\{ \begin{array}{l} b_0 \wedge b_1 \Rightarrow \perp_1 \\ \wedge \quad b_0 \wedge \neg b_1 \Rightarrow 9 \leq x \leq 14 \\ \wedge \quad \neg b_0 \wedge b_1 \Rightarrow \perp_1 \\ \wedge \quad \neg b_0 \wedge \neg b_1 \Rightarrow x \leq 5 \end{array} \right\} \end{aligned}$$

## Transfer functions: assignment to boolean

**Assignment**  $l_0 : b = e; l_1$  to a **boolean variable**, where the right hand side contains **only integer variables** (i.e.,  $e$  depends only on  $x_0, \dots, x_I$ ):

- **example:**  $b_0 = x \leq 7$

### Algorithm:

- let  $z^\# \in \mathbb{D}_0^\#$ , and let us assume that  $z^\#(b) = \text{TRUE}$ :  
then,  $\text{assign}_{\text{cp}}(b, e[x_0, \dots, x_I], X^\#)(z^\#)$  should account for all states where  $b$  becomes true, other boolean variables remaining unchanged:

$$\text{assign}_{\text{cp}}(b, e, X^\#)(z^\#) = \begin{cases} \text{test}_1(e, X^\#(z^\#)) \\ \sqcup_1 \text{test}_1(e, X^\#(z^\#[b \leftarrow \text{FALSE}])) \end{cases}$$

- when  $z^\# \in \mathbb{D}_0^\#$ , and  $z^\#(b) = \text{FALSE}$ : similar computation

**The partitions get modified** (this is a **costly step**, involving join)

## Transfer functions: assignment to boolean, example

- **abstract pre-condition:** 
$$\left\{ \begin{array}{ll} b_0 \wedge b_1 & \Rightarrow 15 \leq x \\ \wedge b_0 \wedge \neg b_1 & \Rightarrow 9 \leq x \leq 14 \\ \wedge \neg b_0 \wedge b_1 & \Rightarrow 6 \leq x \leq 8 \\ \wedge \neg b_0 \wedge \neg b_1 & \Rightarrow x \leq 5 \end{array} \right\}$$
- **statement:**  $b_0 = x \leq 7$
- **abstract post-condition:**

$$\begin{aligned} \text{assign}_{\text{cp}} \left( b_0, x \leq 7, \left\{ \begin{array}{ll} b_0 \wedge b_1 & \Rightarrow 15 \leq x \\ \wedge b_0 \wedge \neg b_1 & \Rightarrow 9 \leq x \leq 14 \\ \wedge \neg b_0 \wedge b_1 & \Rightarrow 6 \leq x \leq 8 \\ \wedge \neg b_0 \wedge \neg b_1 & \Rightarrow x \leq 5 \end{array} \right\} \right) \\ = \left\{ \begin{array}{ll} b_0 \wedge b_1 & \Rightarrow 6 \leq x \leq 7 \\ \wedge b_0 \wedge \neg b_1 & \Rightarrow x \leq 5 \\ \wedge \neg b_0 \wedge b_1 & \Rightarrow 8 \leq x \\ \wedge \neg b_0 \wedge \neg b_1 & \Rightarrow 9 \leq x \leq 14 \end{array} \right\} \end{aligned}$$

**The partitions get modified** (this is a **costly step**, involving join)

## Choice of boolean partitions

Boolean partitioning allows to express relations between boolean and scalar variables

**But these relations are expensive to maintain:**

- 1 partitioning with respect to  $N$  boolean variables translates into a  $2^N$  **space cost factor**
- 2 after assignments, partitions need be recomputed (**use of join**)

Packing addresses the first issue

- select groups of variables for which relations would be **useful**
- can be based on **syntactic** or **semantic** criteria

Whatever the packs, the transfer functions will produce a sound result (but possibly not the most precise one)

**How to alleviate the second issue ?**



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# Definition of trace partitioning

## Principle

We start from a **trace semantics** and rely on an **abstraction of execution history for partitioning**

- **concrete domain**:  $\mathbb{D} = \mathcal{P}(\mathbb{S}^*)$
- **left side abstraction**  $\gamma_0 : \mathbb{D}_0^\# \rightarrow \mathbb{D}$ : a **trace abstraction to be defined precisely later**
- **right side abstraction**, as a **composition** of two abstractions:
  - ▶ the **final state abstraction** defined by  $(\mathbb{D}_1^\#, \sqsubseteq_1^\#) = (\mathcal{P}(\mathbb{S}), \subseteq)$  and:
 
$$\gamma_1 : M \mapsto \{ \langle s_0, \dots, s_k, (\ell, m) \rangle \mid m \in M, \ell \in \mathbb{L}, s_0, \dots, s_k \in \mathbb{S} \}$$
  - ▶ a **store abstraction** applied to the traces final memory state
 
$$\gamma_2 : \mathbb{D}_2^\# \rightarrow \mathbb{D}_1^\#$$

## Trace partitioning

**Cardinal power abstraction** defined by abstractions  $\gamma_0$  and  $\gamma_1 \circ \gamma_2$

# Application 1: partitioning by control states

## Flow sensitive abstraction

- We let  $\mathbb{D}_0^\# = \mathbb{L} \cup \{\top\}$
- Concretization is defined by:

$$\begin{aligned} \gamma_0 : \mathbb{D}_0^\# &\longrightarrow \mathcal{P}(\mathbb{S}^*) \\ \ell &\longmapsto \mathbb{S}^* \cdot (\{\ell\} \times \mathbb{M}) \end{aligned}$$

This produces the same flow sensitive abstraction as with state partitioning; in the following we always compose context sensitive abstraction with other abstractions...

## Trace partitioning is more general than state partitioning

It can also express

- **context-sensitivity**, **partial context sensitivity**
- partitioning guided by a **boolean condition**...

## Application 2: partitioning guided by a condition

We consider a program with a **conditional statement**:

```

l0 : if(c){
l1 :   ...
l2 : }else{
l3 :   ...
l4 : }
l5 : ...

```

### Domain of partitions

The partitions are defined by  $\mathbb{D}_0^\# = \{\tau_{\text{if:t}}, \tau_{\text{if:f}}, \top\}$  and:

$$\begin{aligned}
 \gamma_0 : \quad \tau_{\text{if:t}} &\longmapsto \{ \langle (l_0, m), (l_1, m'), \dots \rangle \mid m \in \mathbb{M}, m' \in \mathbb{M} \} \\
 \tau_{\text{if:f}} &\longmapsto \{ \langle (l_0, m), (l_3, m'), \dots \rangle \mid m \in \mathbb{M}, m' \in \mathbb{M} \} \\
 \top &\longmapsto \mathbb{S}^*
 \end{aligned}$$

**Application:** discriminate the executions depending on the branch they visited

## Application 2: partitioning guided by a condition

This partitioning **resolves the second example**:

```

int x ∈ ℤ;
int s;
int y;
if(x ≥ 0){
    τif:t ⇒ (0 ≤ x) ∧ τif:f ⇒ ⊥
    s = 1;
    τif:t ⇒ (0 ≤ x ∧ s = 1) ∧ τif:f ⇒ ⊥
} else {
    τif:f ⇒ (x < 0) ∧ τif:t ⇒ ⊥
    s = -1;
    τif:f ⇒ (x < 0 ∧ s = -1) ∧ τif:t ⇒ ⊥
}
y = x/s;

```

$$\begin{cases} \tau_{\text{if:t}} \Rightarrow (0 \leq x \wedge s = 1) \\ \wedge \tau_{\text{if:f}} \Rightarrow (x < 0 \wedge s = -1) \end{cases}$$

$$\begin{cases} \tau_{\text{if:t}} \Rightarrow (0 \leq x \wedge s = 1 \wedge 0 \leq y) \\ \wedge \tau_{\text{if:f}} \Rightarrow (x < 0 \wedge s = -1 \wedge 0 < y) \end{cases}$$

## Application 3: partitioning guided by a loop

We consider a program with a **conditional statement**:

```

 $l_0$  : while(c){
 $l_1$  :     ...
 $l_2$  : }
 $l_3$  : ...

```

### Domain of partitions

For a given  $k \in \mathbb{N}$ , the partitions are defined by

$\mathbb{D}_0^\# = \{\tau_{\text{loop}:0}, \tau_{\text{loop}:1}, \dots, \tau_{\text{loop}:k}, \top\}$  and:

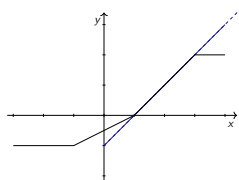
$$\begin{array}{ll} \gamma_0 : \tau_{\text{loop}:i} & \longmapsto \text{traces that visit } l_1 \text{ } i \text{ times} \\ & \top & \longmapsto \mathbb{S}^* \end{array}$$

**Application:** discriminate executions depending on the number of iterations in a loop

# Application 3: partitioning guided by a loop

## An interpolation function:

$$y = \begin{cases} -1 & \text{if } x \leq -1 \\ -\frac{1}{2} + \frac{x}{2} & \text{if } x \in [-1, 1] \\ -1 + x & \text{if } x \in [1, 3] \\ 2 & \text{if } 3 \leq x \end{cases}$$



## Typical implementation:

- use tables of coefficients and loops to search for the range of  $x$

```

int i = 0;
while(i < 4 && x > t_x[i + 1]){
    i ++;
}

```

$$\begin{cases} \tau_{\text{loop:0}} \Rightarrow x \leq -1 \\ \tau_{\text{loop:1}} \Rightarrow -1 \leq x \leq 1 \\ \tau_{\text{loop:2}} \Rightarrow 1 \leq x \leq 3 \\ \tau_{\text{loop:3}} \Rightarrow 3 \leq x \end{cases}$$

$$y = t_c[i] \times (x - t_x[i]) + t_y[i]$$

## Application 4: partitioning guided by the value of a variable

We consider a program with an integer **variable**  $x$ , and a **program point**  $\ell$ :

```
int x; ...;  $\ell$  : ...
```

### Domain of partitions: partitioning by the value of a variable

For a given  $\mathcal{E} \subseteq \mathbb{V}_{\text{int}}$  finite set of integer values, the partitions are defined by  $\mathbb{D}_0^\# = \{\tau_{\text{val}:i} \mid i \in \mathcal{E}\} \uplus \{\top\}$  and:

$$\begin{aligned} \gamma_0 : \quad \tau_{\text{val}:k} &\longmapsto \{ \langle \dots, (\ell, m), \dots \rangle \mid m(x) = k \} \\ \top &\longmapsto \mathbb{S}^* \end{aligned}$$

### Domain of partitions: partitioning by the property of a variable

For a given abstraction  $\gamma : (V^\#, \sqsubseteq^\#) \rightarrow (\mathcal{P}(\mathbb{V}_{\text{int}}), \subseteq)$ , the partitions are defined by  $\mathbb{D}_0^\# = \{\tau_{\text{var}:v^\#} \mid v^\# \in V^\#\}$  and:

$$\gamma_0 : \quad \tau_{\text{val}:v^\#} \longmapsto \{ \langle \dots, (\ell, m), \dots \rangle \mid m(x) \in \tau_{\text{var}:v^\#} \}$$



## Application 4: partitioning guided by the value of a variable

- Left side abstraction shown in blue: **sign of x at entry**
  - Right side abstraction shown in green:
- non relational abstraction (we omit the information about x)
- **Same precision** and **similar results** as boolean partitioning, but **very different abstraction**, fewer partitions, no re-partitioning

```

bool b0, b1;
int x, y;      (uninitialized)
① (x < 0@① ⇒ T) ∧ (x = 0@① ⇒ T) ∧ (x > 0@① ⇒ T)
b0 = x ≥ 0;
   (x < 0@① ⇒ ¬b0) ∧ (x = 0@① ⇒ b0) ∧ (x > 0@① ⇒ b0)
b1 = x ≤ 0;
   (x < 0@① ⇒ ¬b0 ∧ b1) ∧ (x = 0@① ⇒ b0 ∧ b1) ∧ (x > 0@① ⇒ b0 ∧ ¬b1)
if(b0 && b1){
   (x < 0@① ⇒ ⊥) ∧ (x = 0@① ⇒ b0 ∧ b1) ∧ (x > 0@① ⇒ ⊥)
   y = 0;
   (x < 0@① ⇒ ⊥) ∧ (x = 0@① ⇒ b0 ∧ b1 ∧ y = 0) ∧ (x > 0@① ⇒ ⊥)
} else {
   (x < 0@① ⇒ ¬b0 ∧ b1) ∧ (x = 0@① ⇒ ⊥) ∧ (x > 0@① ⇒ b0 ∧ ¬b1)
   y = 100/x;
   (x < 0@① ⇒ ¬b0 ∧ b1 ∧ y ≤ 0) ∧ (x = 0@① ⇒ ⊥) ∧ (x > 0@① ⇒ b0 ∧ ¬b1 ∧ y ≥ 0)
}

```

# Trace partitioning induced by a refined transition system

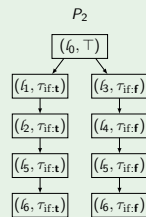
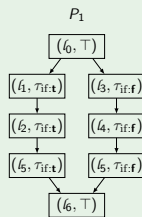
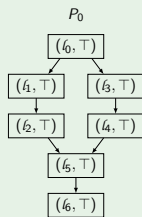
We consider the **partitions induced by a condition**:

- $P_0$ : the analysis may *never* merge traces from both branches
- $P_1$ : the analysis may merge them *right after the condition* (this amounts to doing no partitioning at all)
- $P_2$ : the analysis may merge them *at some point*

```

l0  if(x < 0){
l1    s = -1;
l2  } else {
l3    s = 1;
l4  }
l5  y = x/s;
l6  ...

```



**Intuition:** we can view this form of trace partitioning as **the use of a refined control flow graph**

# Trace partitioning induced by a refined transition system

We now **formalize this intuition**:

- we **augment** control states **with partitioning tokens**:  $\mathbb{L}' = \mathbb{L} \times \mathbb{D}_0^\sharp$   
and let  $\mathbb{S}' = \mathbb{L}' \times \mathbb{M}$
- let  $\rightarrow' \subseteq \mathbb{S}' \times \mathbb{S}'$  be an **extended transition relation**

## Partition of a transition system

System  $\mathcal{S}' = (\mathbb{S}', \rightarrow', \mathbb{S}'_I)$  is a **partition** of transition system  $\mathcal{S} = (\mathbb{S}, \rightarrow, \mathbb{S}_I)$  (and note  $\mathcal{S}' \prec \mathcal{S}$ ) if and only if

- $\forall (\ell, m) \in \mathbb{S}_I, \exists \tau \in \mathbb{D}_0^\sharp, ((\ell, \tau), m) \in \mathbb{S}'_I$
- $\forall (\ell, m), (\ell', m') \in \mathbb{S}, \forall \tau \in \mathbb{D}_0^\sharp,$   
 $(\ell, m) \rightarrow (\ell', m') \implies \exists \tau' \in \mathbb{D}_0^\sharp, ((\ell, \tau), m) \rightarrow ((\ell', \tau'), m')$

Then:

$$\forall \langle (\ell_0, m_0), \dots, (\ell_n, m_n) \rangle \in \llbracket \mathcal{S} \rrbracket_{\mathcal{R}},$$

$$\exists \tau_0, \dots, \tau_n \in \mathbb{D}_0^\sharp, \langle ((\ell_0, \tau_0), m_0), \dots, ((\ell_n, \tau_n), m_n) \rangle \in \llbracket \mathcal{S}' \rrbracket_{\mathcal{R}},$$

# Trace partitioning induced by a refined transition system

## Assumptions:

- **refined control system**  $(S', \rightarrow', S'_I) \prec (S, \rightarrow, S_I)$
- **erasure function**:  $\Psi : (S')^* \rightarrow S^*$  removes the tokens

## Definition of a trace partitioning

The abstraction defining partitions is defined by:

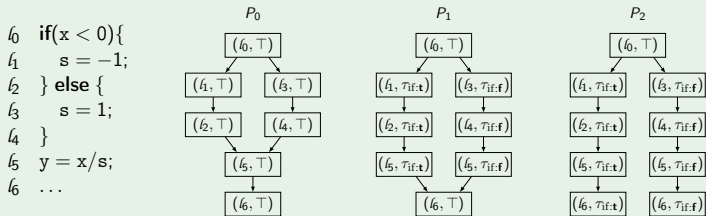
$$\begin{aligned} \gamma_0 : \mathbb{D}_0^\# &\longrightarrow \mathcal{P}(S^*) \\ \tau &\longmapsto \{\sigma \in S^* \mid \exists \sigma' = \langle \dots, ((\ell, \tau), m) \rangle \in (S')^*, \Psi(\sigma') = \sigma \} \end{aligned}$$

Not all instances of trace partitionings can be expressed that way but **many interesting instances can**:

- **control states** and **call stack** partitioning
- partitioning guided by **conditions** and **loops**
- partitioning **guided by the value of a variable**

## Trace partitioning induced by a refined transition system

Example of the **partitioning guided by a condition**:



- each system induces a partitioning, with different merging points:

$$P_1 \prec P_0 \qquad P_2 \prec P_1$$

- these systems induce **hierarchy** of refining control structures

$$P_2 \prec P_1 \prec P_0$$

- this approach **also applies to**:
  - partitioning **induced by a loop**
  - partitioning **induced by the value of a variable at a given point...**

## Transfer functions: example

<code>int x ∈ ℤ;</code>	
<code>int s;</code>	
<code>int y;</code>	
<code>if(x ≥ 0){</code>	
$\tau_{\text{if:t}} \Rightarrow (0 \leq x) \wedge \tau_{\text{if:f}} \Rightarrow \perp$	partition creation: $\tau_{\text{if:t}}$
<code>s = 1;</code>	
$\tau_{\text{if:t}} \Rightarrow (0 \leq x \wedge s = 1) \wedge \tau_{\text{if:f}} \Rightarrow \perp$	no modification of partitions
<code>} else {</code>	
$\tau_{\text{if:f}} \Rightarrow (x < 0) \wedge \tau_{\text{if:t}} \Rightarrow \perp$	partition creation: $\tau_{\text{if:f}}$
<code>s = -1;</code>	
$\tau_{\text{if:f}} \Rightarrow (x < 0 \wedge s = -1) \wedge \tau_{\text{if:t}} \Rightarrow \perp$	no modification of partitions
<code>}</code>	
$\left\{ \begin{array}{l} \tau_{\text{if:t}} \Rightarrow (0 \leq x \wedge s = 1) \\ \wedge \tau_{\text{if:f}} \Rightarrow (x < 0 \wedge s = -1) \end{array} \right.$	no modification of partitions
<code>y = x/s;</code>	
$\left\{ \begin{array}{l} \tau_{\text{if:t}} \Rightarrow (0 \leq x \wedge s = 1 \wedge 0 \leq y) \\ \wedge \tau_{\text{if:f}} \Rightarrow (x < 0 \wedge s = -1 \wedge 0 < y) \end{array} \right.$	no modification of partitions
<code>...</code>	
$\_ \Rightarrow s \in [-1, 1] \wedge 0 \leq y$	fusion of partitions

**Partitions are rarely modified, and only *some* (branching) points**

# Transfer functions: partition creation

## Analysis of an if statement, with partitioning

$l_0$ :	<b>if</b> (c){	$\delta_{l_0, l_1}^\#(X^\#)$	$= [\tau_{\text{if:t}} \mapsto \text{test}(c, \sqcup X^\#(l_0)(\tau)), \top \mapsto \perp]$
$l_1$ :	...	$\delta_{l_0, l_3}^\#(X^\#)$	$= [\tau_{\text{if:t}} \mapsto \text{test}(\neg c, \sqcup_\tau X^\#(l_0)(\tau)), \top \mapsto \perp]$
$l_2$ :	<b>}else</b> {	$\delta_{l_2, l_5}^\#(X^\#)$	$= X^\#$
$l_3$ :	...	$\delta_{l_4, l_5}^\#(X^\#)$	$= X^\#$
$l_4$ :	}		
$l_5$ :	...		

### Observations:

- in the body of the condition: either  $\tau_{\text{if:t}}$  or  $\tau_{\text{if:f}}$   
i.e., **no partition modification there**
- effect at point  $l_5$ : **both  $\tau_{\text{if:t}}$  and  $\tau_{\text{if:f}}$  exist**
- **partitions are modified only at the condition point**, that is only by  $\delta_{l_0, l_1}^\#(X^\#)$  and  $\delta_{l_0, l_2}^\#(X^\#)$

# Transfer functions: partition fusion

When **partitions are not useful anymore, they can be merged**

$$\delta_{\ell_0, \ell_1}^\#(X^\#) = [\_ \mapsto \sqcup_\tau X^\#(\ell_0)(\tau)]$$

Remarks:

- at this point, all partitions are **effectively collapsed** into just one set
- **example**: fusion of the partition of a condition when not useful
- **choice of fusion point**:
  - ▶ **precision**: merge point should not occur as long as partitions are useful
  - ▶ **efficiency**: merge point should occur as early as partitions are not needed anymore



# Choice of partitions

## How are the partitions chosen ?

### Static partitioning

- a fixed partitioning abstraction  $\mathbb{D}_0^\#, \gamma_0$  is **fixed before the analysis**
- usually  $\mathbb{D}_0^\#, \gamma_0$  are chosen by a pre-analysis
- static partitioning is rather easy to formalize and implement
- but it might be limiting, when the choice of partitions is hard

### Dynamic partitioning

- the partitioning abstraction  $\mathbb{D}_0^\#, \gamma_0$  is **not fixed before the analysis**
- instead, it is **computed as part of the analysis**
- i.e., the analysis uses on a lattice of partitioning abstractions  $\mathcal{D}^\#$  and computes  $(\mathbb{D}_0^\#, \gamma_0)$  as an element of this lattice

# Outline

- 1 Introduction
- 2 Imprecisions in convex abstractions
- 3 Disjunctive completion
- 4 Cardinal power and partitioning abstractions
- 5 State partitioning
- 6 Trace partitioning
- 7 Conclusion**

# Adding disjunctions in static analyses

- **Disjunctive completion**: **brutally adds disjunctions**  
**too expensive** in practice
- **Cardinal power abstraction** expresses collections of implications between abstract facts in **two abstract domains**

$$(P_0 \implies Q_0) \wedge \dots \wedge (P_n \implies Q_n)$$

**State partitioning** and **trace partitioning** are particular cases of cardinal power abstraction

- **State partitioning** is **easier** to use when the criteria for partitioning can be easily expressed at the state level
- **Trace partitioning** is **more expressive** in general  
it can also allow the use of **simpler partitioning criteria**, with less “re-partitioning”

# Assignment: paper reading

## Refining static analyses by trace-partitioning using control flow

Maria Handjieva and Stanislas Tzolovski,  
Static Analysis Symposium, 1998,

[http://link.springer.com/chapter/10.1007/3-540-49727-7\\_12](http://link.springer.com/chapter/10.1007/3-540-49727-7_12)

## Abstract interpretation by dynamic partitioning,

François Bourdoncle,  
Journal of Functional Programming, 2(4) 407–423, 1992.

Extended report available at:

<http://www.hpl.hp.com/techreports/Compaq-DEC/PRL-RR-18.pdf>