

Partitioning abstractions

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

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Oct, 29th. 2014

Towards disjunctive abstractions

- **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

Notations

for sets:

- \mathbb{M} : stores
- \mathbb{V} : values
- \mathbb{X} : variables

Assumptions:

- concrete domain $(\mathcal{P}(\mathbb{M}), \subseteq)$ with $\mathbb{M} = \mathbb{X} \rightarrow \mathbb{V}$
- we require an abstract domain $\mathbb{D}^\#$ to provide
 - ▶ a **concretization function** $\gamma : \mathbb{D}^\# \rightarrow \mathcal{P}(\mathbb{M})$
 - ▶ an **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 **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
- $\gamma_x(\{x \in [3, 4]\}, \{x \equiv 0 \pmod{5}\}) = \emptyset$

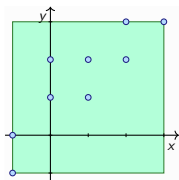
- how to define abstract domain combinators to **add disjunctions** ?

Outline

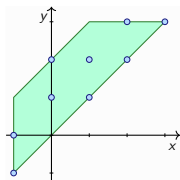
- 1 Introduction
- 2 Imprecisions in convex abstractions**
- 3 Disjunctive completion
- 4 Cardinal power and partitioning abstractions
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Convex abstractions

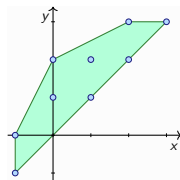
Many numerical abstractions describe convex sets of points



interval domain

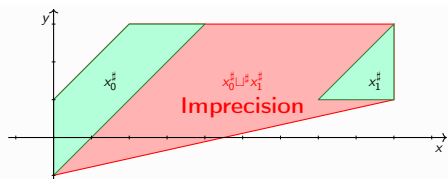


octagon domain



polyhedra domain

Imprecisions inherent in the **convexity**, and when computing **abstract join**:



Such imprecisions may impact analysis results

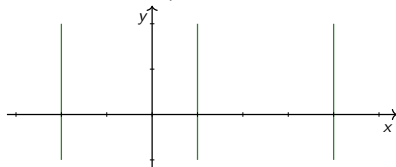
Non convex abstractions

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

Congruences:

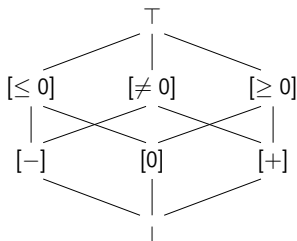
- $\mathbb{D}^\# = \mathbb{Z} \times \mathbb{N}$
- $\gamma(n, k) = \{n + k \cdot p \mid p \in \mathbb{Z}\}$
- $-2, 1 \in \gamma(1, 2)$
but $0 \notin \gamma(1, 2)$

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:

$$\left\{ \begin{array}{l} b_0 = \text{FALSE} \\ b_1 = \text{FALSE} \\ x : \top \end{array} \right.$$

- 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)
}

```

We need to **add 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 fairly complex 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);

```

We need to **add 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 start with 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 γ_{disj} is distributive

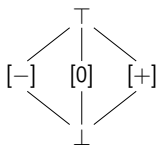
Building a disjunctive completion domain:

- 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 γ_{disj} by

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

Example 1: completion of signs

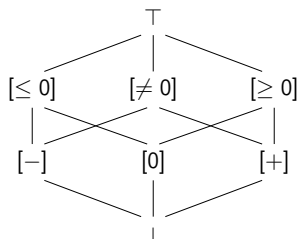
We consider **concrete lattice** $\mathbb{D} = \mathcal{P}(\mathbb{Z})$, with $\sqsubseteq = \subseteq$ and $(\mathbb{D}^\#, \sqsubseteq^\#)$ defined by:



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

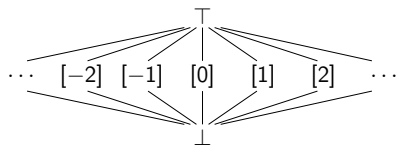
Then, the disjunctive completion is defined by adding elements corresponding to:

- $\{[< 0], [= 0]\}$
- $\{[< 0], [> 0]\}$
- $\{[= 0], [> 0]\}$



Example 2: completion of constants

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



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

Then, the disjunctive completion is the power-set:

- $\mathbb{D}_{\text{disj}}^\# \equiv \mathcal{P}(\mathbb{Z})$
- γ_{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**

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:

- e.g. to compute **abstract post-conditions** (assignment, guard...):
given concrete $\tau : \mathbb{D} \rightarrow \mathbb{D}$, we assume $\tau^\# : \mathbb{D}^\# \rightarrow \mathbb{D}^\#$ such that:

$$\tau \circ \gamma \sqsubseteq \gamma \circ \tau^\#$$

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

$$\tau_{\mathbf{disj}}^\#([\sqcup \mathcal{E}]) = \sqcup[\{\tau^\#(x^\#) \mid x^\# \in \mathcal{E}\}]$$

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**
- ▶ 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

issue: the join operator should “select” which disjoints to merge

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Principle

- disjuncts **that are needed for static analysis** can usually be **characterized** by some property

for instance:

- ▶ **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

Definition

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

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

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

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

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

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

Use of cardinal power abstractions

Intuition: we can express 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

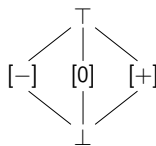
Application $(x \geq 0 \wedge s = 1 \wedge y \geq 0) \vee (x < 0 \wedge s = -1 \wedge y > 0)$

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

Another example, with a single variable

We consider:

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



A few example abstract values:

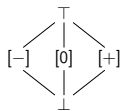
- $[0, 8]$ is expressed by:

$$\begin{cases} \perp_0 & \mapsto \perp_1 \\ [< 0] & \mapsto \perp_1 \\ [= 0] & \mapsto [0, 0] \\ [> 0] & \mapsto [1, 8] \\ \top_0 & \mapsto [0, 8] \end{cases}$$
- $[-10, -3] \uplus [7, 10]$ is expressed by:

$$\begin{cases} \perp_0 & \mapsto \perp_1 \\ [< 0] & \mapsto [-10, -3] \\ [= 0] & \mapsto \perp_1 \\ [> 0] & \mapsto [7, 10] \\ \top_0 & \mapsto [-10, 10] \end{cases}$$

Reduction (1): 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**



$$\text{We let: } X^\# = \begin{cases} \perp_0 & \mapsto \perp_1 \\ [< 0] & \mapsto [-5, -1] \\ [= 0] & \mapsto [0, 0] \\ [> 0] & \mapsto [1, 5] \\ \top_0 & \mapsto [-10, 10] \end{cases} \quad Y^\# = \begin{cases} \perp_0 & \mapsto \perp_1 \\ [< 0] & \mapsto [-5, -1] \\ [= 0] & \mapsto [0, 0] \\ [> 0] & \mapsto [1, 5] \\ \top_0 & \mapsto [-5, 5] \end{cases}$$

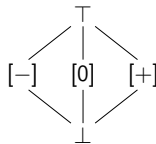
- Then, $\gamma_{\text{cp}}(X^\#) = \gamma_{\text{cp}}(Y^\#)$
- $\gamma_0([< 0]) \cup \gamma_0([= 0]) \cup \gamma_0([> 0]) = \gamma(\top_0)$
but $\gamma_0(X^\#([< 0])) \cup \gamma_0(X^\#([= 0])) \cup \gamma_0(X^\#([> 0])) \subsetneq \gamma(X^\#(\top_0))$

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

- $\bigcup\{\gamma_0(z^\#) \mid z^\# \in \mathcal{E}\} = \gamma_0(\sqcup\{z^\# \mid z^\# \in \mathcal{E}\})$
- $\exists y^\#, \bigcup\{\gamma_1(X^\#(z^\#)) \mid z^\# \in \mathcal{E}\} \subseteq \gamma_1(y^\#) \subset \gamma_1(X^\#(\sqcup\{z^\# \mid z^\# \in \mathcal{E}\}))$

Reduction (2): 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:

$$X^\# = \begin{cases} \perp_0 & \mapsto \perp_1 \\ [< 0] & \mapsto [1, 8] \\ [= 0] & \mapsto [1, 8] \\ [> 0] & \mapsto \perp_1 \\ \top_0 & \mapsto [1, 8] \end{cases} \quad Y^\# = \begin{cases} \perp_0 & \mapsto \perp_1 \\ [< 0] & \mapsto [2, 45] \\ [= 0] & \mapsto [-5, -2] \\ [> 0] & \mapsto [-5, -2] \\ \top_0 & \mapsto \top_1 \end{cases} \quad Z^\# = \begin{cases} \perp_0 & \mapsto \perp_1 \\ [< 0] & \mapsto \perp_1 \\ [= 0] & \mapsto \perp_1 \\ [> 0] & \mapsto \perp_1 \\ \top_0 & \mapsto \perp_1 \end{cases}$$

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

Relation between $\mathbb{D}_0^\#$ elements and $\mathbb{D}_1^\#$ elements

Binding $y_0^\# \mapsto y_1^\#$ can be improved if $\exists z_1^\# \neq y_1^\#, \gamma(y_1^\#) \cap \gamma(y_0^\#) \subseteq \gamma(z_1^\#)$

Representation of the cardinal power

Basic ML representation:

```

type cp = d0 -> d1      not convenient to operate on d0
type cp = (d0,d1) map   maps or functional arrays
  
```

This is not a very efficient representation:

- if \mathbb{D}_0^\sharp has N elements, then an abstract value in $\mathbb{D}_{\text{cp}}^\sharp$ requires N elements of \mathbb{D}_1^\sharp
- if \mathbb{D}_0^\sharp is infinite, and \mathbb{D}_1^\sharp is non trivial, then $\mathbb{D}_{\text{cp}}^\sharp$ 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:

- keep the **same data-type** (most likely functional arrays)
- **avoid representing information attached to redundant elements**

Compact representation

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

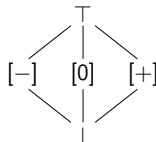
$$\forall x^\# \in \mathbb{D}_0^\#, \exists \mathcal{E} \subseteq \mathcal{C}, \gamma_0(x^\#) = \cup\{\gamma_0(y^\#) \mid y^\# \in \mathcal{E}\}$$

In particular:

- \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**



We remark that:

- \perp_0 does not need be considered
- $\gamma_0([< 0]) \cup \gamma_0([= 0]) \cup \gamma([> 0]) = \gamma(T_0)$ thus T_0 does not need be considered

Thus, we let $\mathcal{C} = \{[< 0], [= 0], [> 0]\}$; then:

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

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_{cp} defined below is a **sound upper bound operator** in $\mathbb{D}_{\text{cp}}^\sharp$:

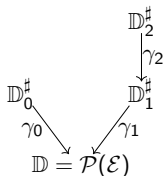
$$X_0^\sharp \sqcup_{\text{cp}} X_1^\sharp \stackrel{\text{def}}{::=} \lambda(z^\sharp \in \mathbb{D}_0^\sharp) \cdot (X_0^\sharp(z^\sharp) \sqcup_1 X_1^\sharp(z^\sharp))$$

- 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 γ :

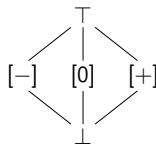
$$\begin{aligned} \gamma : (\mathbb{D}_0^\# \rightrightarrows \mathbb{D}_2^\#) &\longrightarrow (\mathbb{D}_0^\# \rightrightarrows \mathbb{D}_1^\#) \\ X^\# &\longmapsto \lambda(z^\# \in \mathbb{D}_0^\#) \cdot \gamma(X^\#(z^\#)) \end{aligned}$$

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

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

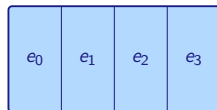
A **state partitioning** abstraction is defined as the cardinal power of two abstractions $(\mathbb{D}_0^\#, \sqsubseteq_0^\#, \gamma_0)$ and $(\mathbb{D}_1^\#, \sqsubseteq_1^\#, \gamma_1)$ of sets of states:

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

Instantiation with a partition

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

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



We can apply **cardinal power construction**:

State partitioning abstraction

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

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

- each $e \in \mathcal{E}$ 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 ?



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**:

| | |
|----------------------------------|--|
| l_0 : // assume $x \geq 0$ | $l_0 \mapsto x : \top \wedge y : \top$ |
| l_1 : if ($x < 10$) { | $l_1 \mapsto x : [0, +\infty[\wedge y : \top$ |
| l_2 : $y = x - 2$; | $l_2 \mapsto x : [0, 9] \wedge y : \top$ |
| l_3 : } else { | $l_3 \mapsto x : [0, 9] \wedge y : [-2, 7]$ |
| l_4 : $y = 2 - x$; | $l_4 \mapsto x : [10, +\infty[\wedge y : \top$ |
| l_5 : } | $l_5 \mapsto x : [10, +\infty[\wedge y :] - \infty, -2]$ |
| l_6 : ... | $l_6 \mapsto x : [0, +\infty[\wedge y :] - \infty, 7]$ |

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:

- $\mathbb{D}_0^\# = \mathbb{L}$
- $\gamma_0 : \ell \mapsto \{\ell\} \times \mathbb{M}$

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

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

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

- after this abstraction step, $\mathbb{D}_1^\#$ 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

- representing one set of memory states per program point may be costly for some applications (e.g., compilation)
- **context insensitive** abstraction simply **forgets about control states**

Flow sensitive 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 $\mathcal{P}(\mathbb{S})$

- used for some ultra-fast pointer analyses
(very quick analyses used for, e.g., compiler optimization)
- otherwise, usually **too coarse**

Application 1: flow insensitive abstraction

We compare with **flow sensitive abstraction**:

| | |
|-------------------------------------|---|
| ℓ_0 : // assume $x \geq 0$ | $\ell_0 \mapsto x : \top \wedge y : \top$ |
| ℓ_1 : if ($x < 10$) { | $\ell_1 \mapsto x : [0, +\infty[\wedge y : \top$ |
| ℓ_2 : $y = x - 2$; | $\ell_2 \mapsto x : [0, 9] \wedge y : \top$ |
| ℓ_3 : } else { | $\ell_3 \mapsto x : [0, 9] \wedge y : [-2, 7]$ |
| ℓ_4 : $y = 2 - x$; | $\ell_4 \mapsto x : [10, +\infty[\wedge y : \top$ |
| ℓ_5 : } | $\ell_5 \mapsto x : [10, +\infty[\wedge y :] - \infty, -2]$ |
| ℓ_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 programs), flow sensitive is absolutely needed

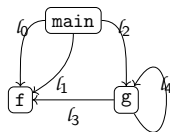
Application 2: context sensitive abstraction

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 |
| $\mid (f, l) \cdot \kappa$ | call to f from stack κ at point l |

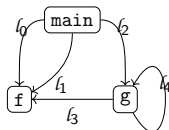
Application 2: context sensitive abstraction, ∞ -CFAFully context sensitive abstraction (∞ -CFA)

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

```

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

```

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)

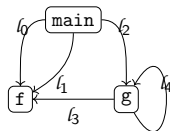
Application 2: context sensitive abstraction, 0-CFA

Non context sensitive abstraction (0-CFA)

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

```

void main(){... l0 : f(); ... l1 : f(); ... l2 : g() ...}
void f(){...}
void g(){if(...){l3 : f()}else{l4 : g()}}
  
```



Contexts in function f :

$(?, f) \cdot \dots,$

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

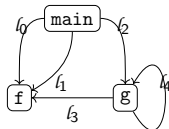
Application 2: context sensitive abstraction, k -CFA

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()}}
  
```



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:

- $\mathbb{D}_0^\# = \mathcal{P}(A)$ for some set A , and $\phi : \mathbb{M} \rightarrow A$
- γ_0 be of the form $(x^\# \in \mathbb{D}_0^\#) \mapsto \{(\ell, m) \in \mathbb{S} \mid \phi(m) \in x^\#\}$

Common choice for A : **the set of boolean values** \mathbb{B}
(or a variation of this)

Many choices for function ϕ 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}_{bool} (resp., \mathbb{X}_{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

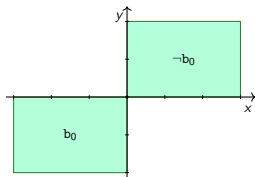
We apply the cardinal power abstraction, with a domain of partition defined by a function, with:

- $A = \mathbb{B}^k$
- $\phi(m) = (m(\mathbf{b}_0), \dots, m(\mathbf{b}_{k-1}))$
- $(\mathbb{D}_1^\#, \sqsubseteq_1^\#, \gamma_1)$ an abstraction of $\mathcal{P}(\mathbb{V}_{\text{int}}^l)$

Application 3: example

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

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



- 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
- **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

```

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 ∧ y ≥ 0) ∧ (¬b0 ∧ b1 ⇒ x < 0 ∧ y ≤ 0)
}

```

Application 3: partitioning by 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

- Abstraction fixing partitions **shown in blue**
- Right side abstraction **shown in green**: interval abstraction

```

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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Computation of abstract semantics and partitioning

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

This abstraction corresponds 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:

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

Fixpoint form of a partitioned semantics

- We consider a transition system $\mathcal{S} = (\mathbb{S}, \rightarrow, \mathbb{S}_I)$
- The reachable states are computed as $\llbracket \mathcal{S} \rrbracket_{\mathcal{R}} = \mathbf{lfp}_{\mathbb{S}_I} F$ where

$$\begin{aligned}
 F : \mathcal{P}(\mathbb{S}) &\longrightarrow \mathcal{P}(\mathbb{S}) \\
 X &\longmapsto \{s \in \mathbb{S} \mid \exists s' \in X, s' \rightarrow s\}
 \end{aligned}$$

Semantic function over the partitioned system

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

$$\begin{aligned}
 F_{\text{part}} : \mathbb{D}_{\text{part}}^{\#} &\longrightarrow \mathbb{D}_{\text{part}}^{\#} \\
 X^{\#} &\longmapsto \lambda(\ell \in \mathbb{L}). \{m \in \mathbb{M} \mid \exists \ell' \in \mathbb{L}, \exists m' \in X^{\#}(\ell'), \\
 &\quad (\ell', m') \rightarrow (\ell, m)\}
 \end{aligned}$$

Then $F_{\text{part}} \circ \alpha_{\text{part}} = \alpha_{\text{part}} \circ F$, and

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

Abstract equations form of a partitioned semantics

- we look for a set of equivalent abstract equations
- let us consider the system of semantic equations over sets of states $\mathcal{E}_1, \dots, \mathcal{E}_s \in \mathcal{P}(\mathbb{M})$:

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

If we let $F_i : (\mathcal{E}_1, \dots, \mathcal{E}_s) \mapsto \bigcup_i \{m \in \mathbb{M} \mid \exists m' \in \mathcal{E}_i, (l_i, m') \rightarrow (l_i, m)\}$, then, we can prove that:

$$\alpha_{\text{part}}(\llbracket \mathcal{S} \rrbracket_{\mathcal{R}}) \text{ is the least solution of the system } \begin{cases} \mathcal{E}_1 &= F_1(\mathcal{E}_1, \dots, \mathcal{E}_s) \\ &\vdots \\ \mathcal{E}_s &= F_s(\mathcal{E}_1, \dots, \mathcal{E}_s) \end{cases}$$

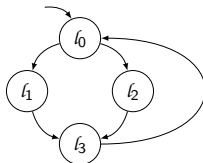
Partitioned systems and fixpoint computation

How to compute an abstract invariant for a partitioned system described by a set of abstract equations ?

(for now, we assume no convergence issue, i.e., that the abstract lattice is of finite height)

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

$$\begin{cases} \mathcal{E}_0 &= \mathcal{I} \cup F_0(\mathcal{E}_3) \\ \mathcal{E}_1 &= F_1(\mathcal{E}_0) \\ \mathcal{E}_2 &= F_2(\mathcal{E}_0) \\ \mathcal{E}_3 &= F_3(\mathcal{E}_1, \mathcal{E}_2) \end{cases}$$



where $\alpha_{\text{part}}(\mathbb{S}_I) = (\mathbb{S}_I, \perp, \perp, \perp)$ (i.e., init states are at point l_0)

Partitioned systems and fixpoint computation

Following the fixpoint transfer, we obtain the following abstract iterates $(\mathcal{E}_n^\sharp)_{n \in \mathbb{N}}$:

$$\begin{aligned}
 \mathcal{E}_0^\sharp &= (\mathbb{I}, & \perp, & \perp, & \perp) \\
 \mathcal{E}_1^\sharp &= (\mathbb{I}, & F_1^\sharp(\mathbb{I}), & F_2^\sharp(\mathbb{I}), & \perp) \\
 \mathcal{E}_2^\sharp &= (\mathbb{I}, & F_1^\sharp(\mathbb{I}), & F_2^\sharp(\mathbb{I}), & F_3^\sharp(F_1^\sharp(\mathbb{I}), F_2^\sharp(\mathbb{I}))) \\
 \mathcal{E}_3^\sharp &= (\mathbb{I} \sqcup F_0^\sharp(F_3^\sharp(F_1^\sharp(\mathbb{I}), F_2^\sharp(\mathbb{I}))), & F_1^\sharp(\mathbb{I}), & F_2^\sharp(\mathbb{I}), & F_3^\sharp(F_1^\sharp(\mathbb{I}), F_2^\sharp(\mathbb{I})))
 \end{aligned}$$

- Each iteration causes **the recomputation of all components**
- Though, each iterate differs from the previous one **in only a few components**

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.

- Other 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 $\{1, \dots, s\}$ such that:

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

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

- **Applications:** we can recompute only what is necessary
- **Back to the example,** where only the **recomputed components** are colored:

$$\begin{aligned}
 \mathcal{E}_0^\sharp &= (\mathbb{I}, && \perp, && \perp, && \perp) \\
 \mathcal{E}_1^\sharp &= (\mathbb{I}, && F_1^\sharp(\mathbb{I}), && \perp, && \perp) \\
 \mathcal{E}_2^\sharp &= (\mathbb{I}, && F_1^\sharp(\mathbb{I}), && F_2^\sharp(\mathbb{I}), && \perp) \\
 \mathcal{E}_3^\sharp &= (\mathbb{I}, && F_1^\sharp(\mathbb{I}), && F_2^\sharp(\mathbb{I}), && F_3^\sharp(F_1^\sharp(\mathbb{I}), F_2^\sharp(\mathbb{I}))) \\
 \mathcal{E}_4^\sharp &= (\mathbb{I} \sqcup F_0^\sharp(F_3^\sharp(F_1^\sharp(\mathbb{I}), F_2^\sharp(\mathbb{I}))), && F_1^\sharp(\mathbb{I}), && F_2^\sharp(\mathbb{I}), && F_3^\sharp(F_1^\sharp(\mathbb{I}), F_2^\sharp(\mathbb{I})))
 \end{aligned}$$

Chaotic iterations: work-list algorithm

Work-list algorithms

Principle:

- maintain a **queue of partitions to update**
- initialize the queue with the **entry label** of the program and the local invariant at that point at $\alpha_{\text{num}}(\mathbb{S}_I)$
- 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
- **terminate** when the queue is **empty**

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

- **Application**: only partitions that need be updated are recomputed
- **Implemented in many static analyzers**

Work-list algorithm

Pseudo code implementation, with $\delta_{\ell, \ell'}^\#$ denoting the transfer function from ℓ to ℓ' :

```

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

- We compose an abstraction $\mathbb{D}_{\text{num}}^\sharp$, with concretization $\gamma_{\text{num}} : \mathbb{D}_{\text{num}}^\sharp \rightarrow \mathcal{P}(\mathbb{M})$, that may not satisfy ascending chain condition
- We assume $\mathbb{D}_{\text{num}}^\sharp$ provides widening operator ∇

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_∇ ; 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

$$\mathbb{L} \longrightarrow (\mathbb{V}_{\text{bool}}^k \longrightarrow \mathbb{D}_0^\#)$$

- 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 abs0 = ... (* abstract elements of  $\mathbb{D}_0^\#$  *)
type abs_state = ... (*
    boolean trees with elements of type abs0 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

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

- that includes
 - ▶ **scalar assignment**, e.g., $x = 1 - x$;
 - ▶ **scalar test**, e.g., $\mathbf{if}(x \geq 8) \dots$
 - ▶ **boolean test**, e.g., $\mathbf{if}(\neg b_1) \dots$
 - ▶ **mixed assignment**, e.g., $b_0 = x \leq 7$

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

Transfer functions: scalar assignment

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** δ_{ℓ_0, ℓ_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):

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

Soundness

If assign_1 is sound, so is $\mathit{assign}_{\rightarrow}$, in the sense that:

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

Transfer functions: scalar assignment, 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:**

$$\begin{aligned} & \mathit{assign}_{\rightarrow} \left(\mathbf{x}, 1 - \mathbf{x}, \left\{ \begin{array}{l} \mathbf{b} \Rightarrow \mathbf{x} \geq 0 \\ \wedge \quad \neg \mathbf{b} \Rightarrow \mathbf{x} \leq 0 \end{array} \right\} \right) \\ &= \left\{ \begin{array}{l} \mathbf{b} \Rightarrow \mathbf{x} \geq 8 \\ \wedge \quad \neg \mathbf{b} \Rightarrow \top \end{array} \right\} \end{aligned}$$

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** δ_{ℓ_0, ℓ_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:

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

Soundness

If test_1 is sound, so is $\text{test}_{\rightarrow}$, in the sense that:

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

Transfer functions: scalar test, example

- **abstract pre-condition:**

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

- **statement:**

if($x \geq 8$)...

- **abstract post-condition:**

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

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 :

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

Soundness

If test_0 is sound, so is $\text{test}_{\rightarrow}$, in the sense that:

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

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:**

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

Transfer functions: mixed assignment

Assignment $\ell_0 : b = e; \ell_1$ 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:** $b_0 = x \leq 7$
- let $z^\# \in \mathbb{D}_0^\#$, such that $z^\#(b) = \text{TRUE}$
 $\text{assign}_{\rightarrow}(b, e[x_0, \dots, x_I], X^\#)(z^\#)$ should account for all states where b becomes true, other boolean variables remaining unchanged:

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

- same computation for cases where $z^\#(b) = \text{FALSE}$

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

Transfer functions: mixed assignment, example

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

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

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**
- These relations are **expensive**:
 - ① Partitioning with respect to N boolean variables translates into a 2^N space cost factor
 - ② After assignments, partitions need be recomputed
- **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 ?

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

Definition of trace partitioning

Assumptions: we start from a **trace semantics** and use **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**
- **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:

$$\begin{array}{l} \gamma_1 : \mathbb{D}_1^\# \longrightarrow \mathcal{P}(\mathbb{S}^*) \\ M \longmapsto \{ \langle s_0, \dots, s_k, (\ell, m) \rangle \mid m \in M, \ell \in \mathbb{L}, s_0, \dots, s_k \in \mathbb{S} \} \end{array}$$
 - ▶ 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 the above, and by an abstraction of sets of traces $\gamma_0 : \mathbb{D}_0^\# \rightarrow \mathcal{P}(\mathbb{S}^*)$

Application 1: partitioning by control states

Flow sensitive abstraction

- We let $\mathbb{D}_0^\# = \mathbb{L}$
- 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 $\mathbb{D}_0^\# = \{\text{if}_t, \text{if}_f, \top\}$ and:

$$\begin{aligned}
 \gamma_0 : \text{if}_t &\longmapsto \{ \langle (l_0, m), (l_1, m'), \dots \rangle \mid m \in \mathbb{M}, m' \in \mathbb{M} \} \\
 \text{if}_f &\longmapsto \{ \langle (l_0, m), (l_3, m'), \dots \rangle \mid m \in \mathbb{M}, m' \in \mathbb{M} \} \\
 \top &\longmapsto \mathbb{S}^*
 \end{aligned}$$

Application: discriminate the executions depending on the branch they visited

Application 2: partitioning guided by a condition

This partitioning **resolves the second example** (we do not represent \top when it gives no information):

```

int x ∈ ℤ;
int s;
int y;
if(x ≥ 0){
    ift ⇒ (0 ≤ x) ∧ iff ⇒ ⊥
    s = 1;
    ift ⇒ (0 ≤ x ∧ s = 1) ∧ iff ⇒ ⊥
} else {
    iff ⇒ (x < 0) ∧ ift ⇒ ⊥
    s = -1;
    iff ⇒ (x < 0 ∧ s = -1) ∧ ift ⇒ ⊥
}

```

$$\left\{ \begin{array}{l} \text{if}_t \Rightarrow (0 \leq x \wedge s = 1) \\ \wedge \text{if}_f \Rightarrow (x < 0 \wedge s = -1) \end{array} \right.$$

```

y = x/s;

```

$$\left\{ \begin{array}{l} \text{if}_t \Rightarrow (0 \leq x \wedge s = 1 \wedge 0 \leq y) \\ \wedge \text{if}_f \Rightarrow (x < 0 \wedge s = -1 \wedge 0 < y) \end{array} \right.$$

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

$\mathbb{D}_0^\# = \{\text{loop}_0, \text{loop}_1, \dots, \text{loop}_k, \top\}$ and:

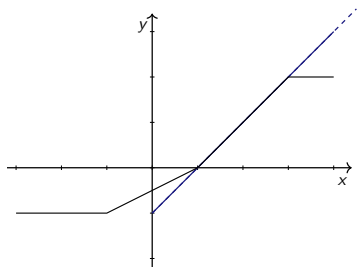
$$\begin{array}{ll} \gamma_0 : \text{loop}_i & \mapsto \text{traces that visit } l_1 \text{ } i \text{ times} \\ \top & \mapsto \mathbb{S}^* \end{array}$$

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 ++;
}
{
    loop_0  =>  x <= -1
    loop_1  => -1 <= x <= 1
    loop_2  =>  1 <= x <= 3
    loop_3  =>  3 <= x
}
y = t_c[i] * (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** ℓ :

```
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 $\mathbb{D}_0^\# = \{\text{val}_i \mid i \in \mathcal{E}\} \uplus \{\top\}$ and:

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

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 $\mathbb{D}_0^\# = \{\text{var}_{v^\#} \mid v^\# \in V^\#\}$ and:

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

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 ⇒ ⊤) ∧ (x = 0@① ⇒ ⊤) ∧ (x > 0@① ⇒ ⊤)
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

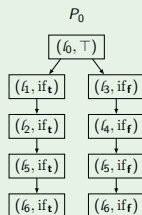
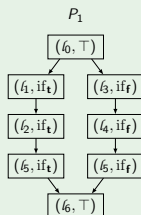
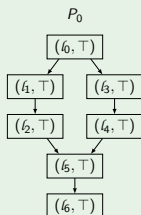
Let us consider the **partitions induced by a condition**:

- we may *never* merge traces from both branches
- we may merge them *right after the condition* (this amounts to doing no partitioning at all)
- we may merge them *at some point*

```

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

```



Thus, 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

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

Then:

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

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

Trace partitioning induced by a refined transition system

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

Definition of a trace partitioning

The abstraction defining partitions is defined by:

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

- not all instances of trace partitionings can be expressed that way
- ... but many interesting instances can

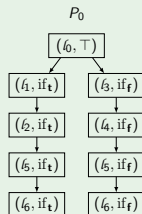
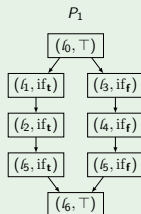
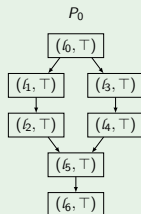
Trace partitioning induced by a refined transition system

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

```

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

```



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

$$P_1 \prec P_0$$

$$P_2 \prec P_1$$

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

$$P_2 \prec P_1$$

- this approach **also applies to**:

- ▶ partitioning **induced by a loop**
- ▶ partitioning **induced by the value of a variable at a given point...**

Abstract interpretation of a partitioned transition system

- let $\mathcal{S} = (\mathbb{S}, \rightarrow, \mathbb{S}_I)$, and a refining system $\mathcal{S}' = (\mathbb{S}', \rightarrow', \mathbb{S}'_I)$, with $\mathbb{S} = \mathbb{L} \times \mathbb{M}$, $\mathbb{S}' = (\mathbb{L} \times \mathbb{D}_0^\#) \times \mathbb{M}$
- transfer functions of \mathcal{S}' :**
 $\delta_{\ell, \ell'} : (\mathbb{D}_0^\# \rightarrow \mathbb{D}_1^\#) \longrightarrow (\mathbb{D}_0^\# \rightarrow \mathbb{D}_1^\#)$ over-approximating \rightarrow'

Partition irrelevant transfer function

ℓ, ℓ' induces a **partition irrelevant transfer function** if and only if:

$$\forall \text{tok}, \text{tok}' \in \mathbb{D}_0^\#, \forall m, m' \in \mathbb{M}, \\ ((\ell, \text{tok}), m) \rightarrow' ((\ell', \text{tok}'), m') \implies \text{tok} = \text{tok}'$$

- partition irrelevant transfer functions: **pointwise operators of $\mathbb{D}_1^\#$** for our examples of partitioning: this is the **most common case**
- other transfer functions**: usually for partition **creation** or **fusion** or **simple composition** of a creation / fusion + partition irrelevant t.f.

Transfer functions: example

| | |
|--|-------------------------------------|
| <code>int x ∈ ℤ;</code> | |
| <code>int s;</code> | |
| <code>int y;</code> | |
| <code>if(x ≥ 0){</code> | |
| <code>if_t ⇒ (0 ≤ x) ∧ if_f ⇒ ⊥</code> | partition creation: if _t |
| <code>s = 1;</code> | |
| <code>if_t ⇒ (0 ≤ x ∧ s = 1) ∧ if_f ⇒ ⊥</code> | no modification of partitions |
| <code>} else {</code> | |
| <code>if_f ⇒ (x < 0) ∧ if_t ⇒ ⊥</code> | partition creation: if _f |
| <code>s = -1;</code> | |
| <code>if_f ⇒ (x < 0 ∧ s = -1) ∧ if_t ⇒ ⊥</code> | no modification of partitions |
| <code>}</code> | |
| <code>{</code> | |
| <code>if_t ⇒ (0 ≤ x ∧ s = 1)</code> | |
| <code>∧ if_f ⇒ (x < 0 ∧ s = -1)</code> | no modification of partitions |
| <code>y = x/s;</code> | |
| <code>{</code> | |
| <code>if_t ⇒ (0 ≤ x ∧ s = 1 ∧ 0 ≤ y)</code> | |
| <code>∧ if_f ⇒ (x < 0 ∧ s = -1 ∧ 0 < y)</code> | no modification of partitions |
| <code>...</code> | |
| <code>_ ⇒ s ∈ [-1, 1] ∧ 0 ≤ y</code> | fusion of partitions |

In general, partitions are rarely modified (only *some* branching points)

Transfer functions: partition creation

Analysis of an if statement, with partitioning

| | | |
|---------|---------------------|---|
| l_0 : | if (c) { | |
| l_1 : | ... | $\delta_{l_0, l_1}^\#(X^\#) = [\text{if}_t \mapsto \text{test}(c, \sqcup X^\#(l_0)(t)), \top \mapsto \perp]$ |
| l_2 : | } else { | $\delta_{l_0, l_3}^\#(X^\#) = [\text{if}_t \mapsto \text{test}(\neg c, \sqcup_t X^\#(l_0)(t)), \top \mapsto \perp]$ |
| l_3 : | ... | $\delta_{l_2, l_5}^\#(X^\#) = X^\#$ |
| l_4 : | } | $\delta_{l_4, l_5}^\#(X^\#) = X^\#$ |
| l_5 : | ... | |

- in the body of the condition: either if_t or if_f
- effect at point l_5 : **both if_t and if_f exist**

Transfer functions: partition fusion

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

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

- 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** is **too expensive** in practice
- The **cardinal power abstraction** expresses collections of implications between abstract facts in **two abstract domains**
- **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

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>