

Combination of Abstract Domains

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

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Overview of the lecture

- **Construction of abstract semantics**
a **step-by-step process** from basic abstractions
 - ▶ **numerical abstractions**
 - ▶ **conjunctions** of abstract properties: product
 - ▶ **disjunctions** of abstract properties: disjunctive completion, partitioning
- **Decomposing abstraction** has **many advantages**:
 - ▶ **modular** design of static analyzers:
split into several different abstractions
 - ▶ **flexibility** of the resulting tools:
better scalability, extensibility to broader analysis setups
- Also, we will get a **better understanding** of abstract domain properties: **reduction**

An example

How to verify the following program ?

```
int i = 0;           // integer variable
bool b;             // boolean variable
while(i < 10){
  i = i + 2;
  b = brand();
  if(b){
    break;
  }
}
assert(b  $\vee$  i == 10); // assertion to prove
```

- We want to do an abstract interpretation of the code
- First, we need to **construct an abstract domain**

Hoare proof and choice of an abstract domain

```

int i = 0;
{i = 0}
bool b;
{i = 0}
while(i < 10){
{0 ≤ i ≤ 8 ∧ i ≡ 0(2)}
  i = i + 2;
{2 ≤ i ≤ 10 ∧ i ≡ 0(2)}
  b = brand();
{2 ≤ i ≤ 10 ∧ i ≡ 0(2)}
  if(b){
{2 ≤ i ≤ 10 ∧ i ≡ 0(2) ∧ b = TRUE}
    break;
  }
}
{2 ≤ i ≤ 10 ∧ i ≡ 0(2) ∧ b = FALSE}
}
{b = TRUE ∨ i = 10}
assert(b ∨ i == 10);

```

Abstract interpretation

Which abstract domain ?

We need:

- interval constraints
 - congruences constraints
 - conjunctions
 - disjunctions
- This lecture shows **how to build such a domain** using combinations of basic abstract domains

A first (de)composition: function composition

Flashback: composition of Galois connections

Let $(\mathbb{D}_0, \sqsubseteq_0)$, $(\mathbb{D}_1, \sqsubseteq_1)$ and $(\mathbb{D}_2, \sqsubseteq_2)$ be three abstract domains, and let us assume the Galois connections below are defined:

$$(\mathbb{D}_0, \sqsubseteq_0) \xleftrightarrow[\alpha_{01}]{\gamma_{10}} (\mathbb{D}_1, \sqsubseteq_1) \quad (\mathbb{D}_1, \sqsubseteq_1) \xleftrightarrow[\alpha_{12}]{\gamma_{21}} (\mathbb{D}_2, \sqsubseteq_2)$$

Then, we have a third Galois connection

$$(\mathbb{D}_0, \sqsubseteq_0) \xleftrightarrow[\alpha_{12} \circ \alpha_{01}]{\gamma_{10} \circ \gamma_{21}} (\mathbb{D}_2, \sqsubseteq_2)$$

We can generalize this principle:

Composition of concretization functions

If $\gamma_{21} : \mathbb{D}_2 \rightarrow \mathbb{D}_1$ (resp., $\gamma_{10} : \mathbb{D}_1 \rightarrow \mathbb{D}_0$) describe concretization functions from $(\mathbb{D}_2, \sqsubseteq_2)$ to $(\mathbb{D}_1, \sqsubseteq_1)$ (resp., from $(\mathbb{D}_1, \sqsubseteq_1)$ to $(\mathbb{D}_0, \sqsubseteq_0)$), then $\gamma_{20} = \gamma_{10} \circ \gamma_{21}$ describes a concretization from $(\mathbb{D}_2, \sqsubseteq_2)$ to $(\mathbb{D}_0, \sqsubseteq_0)$

Decomposition of abstract domains

We inspect the predicates needed in the Hoare proof:

- **One invariant per control point:**
 - ▶ already seen informally in previous lectures
 - ▶ different control states need be abstracted separately
 - ▶ **partitioning** abstraction
- $\{0 \leq i \leq 8 \wedge i \equiv 0(2)\}$:
 - ▶ **conjunction** of an **interval** constraint and of a **congruence** constraint
 - ▶ expressible in a **product** of abstractions
- $\{b = \text{TRUE} \vee i = 10\}$:
 - ▶ **disjunction** of constraints
 - ▶ several ways to express this:
state partitioning, trace partitioning

Notations and definitions: concrete level

Concrete states

Concrete states are of the form $\mathcal{S} = \mathbb{L} \times \mathbb{M}$

- \mathbb{L} is the set of *labels* or *control states*
- \mathbb{M} is the set of *memory states*

Moreover, $\mathbb{M} = \mathbb{X} \rightarrow \mathbb{V}$, where:

- \mathbb{X} is the *set of variables*
- \mathbb{V} is the *set of values*

We will use several concrete semantics during this lecture:

- **finite traces semantics** $\llbracket \mathcal{S} \rrbracket^* \in \mathcal{P}(\mathcal{S}^*)$
- **reachable states semantics** $\llbracket \mathcal{S} \rrbracket_{\mathcal{R}} \in \mathcal{P}(\mathcal{S})$

Notations and definitions: abstract level

We shall use abstract-domains to over-approximate sets of **concrete values**, sets of **states**, sets of **traces**

Abstract domain definitions

An abstract domain will comprise a set of abstract values $\mathbb{D}^\#$ and:

- a concretization function γ and optionnally an abstraction α
 - an abstract order $\sqsubseteq^\#$, an abstract infimum \perp
 - an abstract upper bound $\sqcup^\#$, and a widening operator ∇
 - abstract transfer functions $f^\#, g^\#, \dots$ associated to common concrete operations
- These allow defining static analyses computing abstract least-fixpoints or abstract post-fixpoints

When we build composite abstract domains from basic ones, we will assume / ensure such elements

Outline

- 1 Introduction
- 2 Abstraction of partitioned systems**
- 3 Product of abstractions
- 4 Reduction and application to reduced product
- 5 Reduced cardinal power abstraction
- 6 State partitioning, trace partitioning
- 7 Concluding remarks

Partitioning of an abstraction

Partitioning abstraction

Given set E and partition \mathfrak{P} of E , we let the **partitioning abstraction** over E be defined by:

$$\begin{array}{lll} \alpha_{\text{part}} : & \mathcal{P}(E) & \longrightarrow (\mathfrak{P} \rightarrow \mathcal{P}(E)) \\ & X & \longmapsto \lambda(p \in \mathfrak{P}) \cdot (p \cap X) \\ \gamma_{\text{part}} : & (\mathfrak{P} \rightarrow \mathcal{P}(E)) & \longrightarrow \mathcal{P}(E) \\ & \Phi & \longmapsto \bigcup_{p \in \mathfrak{P}} \Phi(p) \end{array}$$

It indeed forms a Galois connection:

$$(\mathcal{P}(E), \subseteq) \begin{array}{c} \xleftarrow{\gamma_{\text{part}}} \\ \xrightarrow{\alpha_{\text{part}}} \end{array} (\mathfrak{P} \rightarrow \mathcal{P}(E), \dot{\subseteq})$$

Proof: $\alpha_{\text{part}}(X) \dot{\subseteq} \Phi \iff X \subseteq \gamma_{\text{part}}(\Phi)$

Example: control state partitioning

How to abstract separately memory states associated to different control states ?

Control state partitioning

We apply the partitioning abstraction with:

- $E = \mathbb{S}$
- $\mathfrak{P} = \{ \{(l, m) \mid m \in \mathbb{M}\} \mid l \in \mathbb{L} \}$

We note that $\mathfrak{P} \equiv \mathbb{L}$ and that, for all $l \in \mathbb{L}$, $\{(l, m) \mid m \in \mathbb{M}\} \equiv \mathbb{M}$, therefore, the partitioning abstraction is:

$$\begin{array}{lll}
 \alpha_{\text{part}} : & \mathcal{P}(E) & \longrightarrow (\mathbb{L} \rightarrow \mathcal{P}(E)) \\
 & X & \longmapsto \lambda(l \in \mathbb{L}) \cdot \{m \in \mathbb{M} \mid (l, m) \in X\} \\
 \gamma_{\text{part}} : & (\mathbb{L} \rightarrow \mathcal{P}(E)) & \longrightarrow \mathcal{P}(E) \\
 & \Phi & \longmapsto \bigcup_{l \in \mathbb{L}} \{(l, m) \mid m \in \Phi(l)\}
 \end{array}$$

Example: control state partitioning

We can compose this abstraction with any other abstraction over memory states:

Abstraction over a partitioned system

Let $(\mathbb{D}_{\text{num}}^{\#}, \sqsubseteq_{\text{num}}^{\#})$ be an abstraction of $(\mathcal{P}(\mathbb{M}), \subseteq)$, with a Galois connection $(\mathcal{P}(\mathbb{M}), \subseteq) \xleftrightarrow[\alpha_{\text{num}}]{\gamma_{\text{num}}} (\mathbb{D}_{\text{num}}^{\#}, \sqsubseteq_{\text{num}}^{\#})$.

Then, we define the abstract domain $(\mathbb{D}_{\text{part}}^{\#}, \sqsubseteq_{\text{part}}^{\#}) = (\mathbb{L} \rightarrow \mathbb{D}_{\text{num}}^{\#}, \sqsubseteq_{\text{num}}^{\#})$, with the abstraction and concretization defined by:

$$\begin{array}{ll}
 \dot{\alpha}_{\text{num}} \circ \alpha_{\text{part}} : \mathcal{P}(\mathcal{S}) & \longrightarrow (\mathbb{L} \rightarrow \mathbb{D}_{\text{num}}^{\#}) \\
 \mathcal{S} & \longmapsto \lambda(l \in \mathbb{L}). \alpha_{\text{num}}(\{m \in \mathbb{M} \mid (l, m) \in \mathcal{S}\}) \\
 \gamma_{\text{part}} \circ \dot{\gamma}_{\text{num}} : (\mathbb{L} \rightarrow \mathbb{D}_{\text{num}}^{\#}) & \longrightarrow \mathcal{P}(\mathcal{S}) \\
 \Phi & \longmapsto \{(l, m) \mid \exists l \in \mathbb{L}, m \in \gamma_{\text{num}}(\Phi(l))\}
 \end{array}$$

- Case with only a γ_{num} (no α_{num}): similar definitions

Example: context sensitive abstraction

We consider a language with procedures (set of procedures \mathbb{P})

Semantics with procedures

The set of states is of the form $\mathcal{S} = \mathbb{K} \times \mathbb{L} \times \mathbb{M}$, where \mathbb{K} is the set of contexts defined by:

$$\begin{array}{ll}
 k \in \mathbb{K} & ::= \epsilon & \text{empty call stack} \\
 & | f \cdot k & \text{call to } f \text{ from stack } k
 \end{array}$$

Context sensitive abstraction

$$\mathfrak{P} = \{ \{ (k, l, m) \mid m \in \mathbb{M} \} \mid k \in \mathbb{K}, l \in \mathbb{L} \}$$

- one invariant per calling context
- infinite if recursion

Context insensitive abstraction

$$\mathfrak{P} = \{ \{ (f \cdot k, l, m) \mid m \in \mathbb{M}, k \in \mathbb{K} \} \mid f \in \mathbb{P}, l \in \mathbb{L} \}$$

- merges different calling contexts to a same procedure
- coarser abstraction

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 $\mathbb{D}_{\text{part}}^{\#} = \mathfrak{P} \rightarrow \mathcal{P}(\mathbb{S})$ by:

$$\begin{aligned}
 F_{\text{part}} : \mathbb{D}_{\text{part}}^{\#} &\longrightarrow \mathbb{D}_{\text{part}}^{\#} \\
 \Phi &\longmapsto \lambda(p \in \mathfrak{P}) \cdot \{s \in p \mid \exists p' \in \mathfrak{P}, \exists s' \in \Phi(p'), s' \rightarrow s\}
 \end{aligned}$$

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
- We consider the case of a system partitioned by control states
 $\mathbb{L} = \{l_1, \dots, l_s\}$
- Let us consider the system of semantic equations over sets of states
 $\mathcal{E}_1, \dots, \mathcal{E}_s \in \mathcal{P}(\mathbb{M})$:

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

So, 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:

$\alpha_{\text{part}}(\llbracket \mathcal{S} \rrbracket_{\mathcal{R}})$ 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 being analyzed
- **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 $\alpha_{\text{part}}(\mathcal{S}_{\mathcal{I}}) = (\mathcal{S}_{\mathcal{I}}, \perp, \perp, \perp)$ (i.e., init states are at point l_0)

Partitioned systems and fixpoint computation

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

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

- 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:

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

Chaotic iterations: worklist algorithm

Worklist 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 $\alpha_{\text{num}}(\mathcal{S}_{\mathcal{I}})$
- 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**

Selection of a set of widening points for a partitioned system

- We do not assume anymore that $\mathbb{D}_{\text{num}}^\#$ satisfies the ascending chain condition
- We assume $\mathbb{D}_{\text{num}}^\#$ provides widening operator ∇

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

Enforcing termination of chaotic iterates

Let $K \subseteq \{1, \dots, s\}$ 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:

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

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:

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

Proof: exercise

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Product abstraction

Definition

Let $(\mathbb{D}_0^\#, \sqsubseteq_0^\#)$ and $(\mathbb{D}_1^\#, \sqsubseteq_1^\#)$ be two abstract domains:

$$(\mathbb{D}, \sqsubseteq) \xleftarrow[\alpha_0]{\gamma_0} (\mathbb{D}_0^\#, \sqsubseteq_0^\#) \quad \text{and} \quad (\mathbb{D}, \sqsubseteq) \xleftarrow[\alpha_1]{\gamma_1} (\mathbb{D}_1^\#, \sqsubseteq_1^\#)$$

The product abstract domain $(\mathbb{D}_\times^\#, \sqsubseteq_\times^\#)$ is defined by:

- $\mathbb{D}_\times^\# = \mathbb{D}_0^\# \times \mathbb{D}_1^\#$
- $(x_0, x_1) \sqsubseteq_\times^\# (y_0, y_1) \iff x_0 \sqsubseteq_0^\# y_0 \wedge x_1 \sqsubseteq_1^\# y_1$

The product abstraction is defined by:

$$(\mathbb{D}, \sqsubseteq) \xleftarrow[\alpha_\times]{\gamma_\times} (\mathbb{D}_\times^\#, \sqsubseteq_\times^\#) \quad \text{where}$$

$$\alpha_\times : \mathbb{D} \longrightarrow \mathbb{D}_\times^\# \qquad \gamma_\times : \mathbb{D}_\times^\# \longrightarrow \mathbb{D}$$

$$a \longmapsto (\alpha_0(a), \alpha_1(a)) \qquad (x_0, x_1) \longmapsto \gamma_0(x_0) \cap \gamma_1(x_1)$$

Product abstraction

Proof, following the usual principle:

$$\begin{aligned}
 \alpha(a) \sqsubseteq_{\times}^{\#} (x_0, x_1) &\iff (\alpha_0(a), \alpha_1(a)) \sqsubseteq_{\times}^{\#} (x_0, x_1) \\
 &\iff \alpha_0(a) \sqsubseteq_0^{\#} x_0 \wedge \alpha_1(a) \sqsubseteq_1^{\#} x_1 \\
 &\iff a \subseteq \gamma_0(x_0) \wedge a \subseteq \gamma_1(x_1) \\
 &\iff a \subseteq \gamma_0(x_0) \cap \gamma_1(x_1) \\
 &\iff a \subseteq \gamma_{\times}(x_0, x_1)
 \end{aligned}$$

Conjunctions of abstract properties

Elements of the product abstract domain stand for conjunctions of abstract properties of $\mathbb{D}_0^{\#}$ and of $\mathbb{D}_1^{\#}$.

Example: conjunctions of constraints

Assumptions:

- \mathbb{D} is $\mathcal{P}(\mathbb{Z})$ and \subseteq the set inclusion
- \mathbb{D}_0^\sharp is $\mathbb{Z} \cup \{-\infty, +\infty\}$, \sqsubseteq_0^\sharp is \leq and $\alpha_0(E) = \inf E$
- \mathbb{D}_1^\sharp is $\mathbb{Z} \cup \{-\infty, +\infty\}$, \sqsubseteq_1^\sharp is \leq and $\alpha_1(E) = \sup E$

Product abstraction:

- Then:

$$\begin{aligned} \alpha_x(\mathbb{Z}) &= (-\infty, +\infty) & \alpha_x(\{0, 2, 4, 6, 8\}) &= (0, 8) \\ \alpha_x(\emptyset) &= (+\infty, -\infty) & \alpha_x(\{1, 2, 3\}) &= (1, 3) \end{aligned}$$

- Moreover:

$$\gamma_x(x_0, x_1) = \{x \in \mathbb{Z} \mid x_0 \leq x \wedge x \leq x_1\}$$

Therefore \mathbb{D}_x^\sharp is the **interval abstraction**, where an interval is viewed as a conjunction of two constraints

Example: intervals and congruences

Assumptions:

- \mathbb{D} is $\mathcal{P}(\mathbb{Z})$ and \subseteq the set inclusion
- \mathbb{D}_0^\sharp is the interval abstract domain (an abstract values is either \perp or a pair of elements of $\mathbb{Z} \cup \{-\infty, +\infty\}$)
- \mathbb{D}_1^\sharp is the congruences abstract domain:
 - ▶ abstract values are either \perp , or of the form $\langle a, b \rangle$ with $0 \leq a < b$ or $b = 0$
 - ▶ $\gamma_1(\perp) = \emptyset$ and $\gamma_1(\langle a, b \rangle) = \{a + k \cdot b \mid k \in \mathbb{Z}\}$

Product abstraction:

- Then:

$$\begin{array}{ll} \alpha_\times(\emptyset) &= (\perp, \perp) & \alpha_\times(\{1, 3, \dots\}) &= ([1, +\infty[, \langle 1, 2 \rangle) \\ \alpha_\times(\mathbb{Z}) &= (] - \infty, +\infty[, \langle 0, 1 \rangle) & \alpha_\times(\{1, 3, 7\}) &= ([1, 7], \langle 1, 2 \rangle) \end{array}$$

- Moreover:

$$\begin{array}{ll} \gamma_\times([1, 7], \langle 1, 2 \rangle) &= \{1, 3, 5, 7\} & \gamma_\times([0, 10], \langle 3, 6 \rangle) &= \{3, 9\} \\ \gamma_\times([1, 8], \langle 1, 2 \rangle) &= \{1, 3, 5, 7\} & \gamma_\times([0, +\infty[, \langle 3, 6 \rangle) &= \{3, 9, \dots\} \end{array}$$

Operations in the product domain

- Least element:** if \perp_0 (resp., \perp_1) is the least element of \mathbb{D}_0^\sharp (resp. of \mathbb{D}_1^\sharp), then $\perp_\times = (\perp_0, \perp_1)$ is the least element of \mathbb{D}_\times^\sharp
- Upper bound:** if \sqcup_0 (resp., \sqcup_1) is a sound upper bound operator on \mathbb{D}_0^\sharp (resp., \mathbb{D}_1^\sharp), then \sqcup_\times defined by $(x_0, x_1) \sqcup_\times (y_0, y_1) = (x_0 \sqcup_0 y_0, x_1 \sqcup_1 y_1)$ is a sound upper bound operator on \mathbb{D}_\times^\sharp
- Widening:** if \sqcup_0 (resp. \sqcup_1) is a widening on \mathbb{D}_0^\sharp (resp. \mathbb{D}_1^\sharp), then \sqcup_\times defined by $(x_0, x_1) \sqcup_\times (y_0, y_1) = (x_0 \sqcup_0 y_0, x_1 \sqcup_1 y_1)$ is a widening on \mathbb{D}_\times^\sharp

Proofs: exercise!

Operations in the product domain

• Transfer functions:

We assume that:

- ▶ $f : \mathbb{D} \rightarrow \mathbb{D}$ is a concrete transfer function (e.g., describing the effect of a test or of an assignment)
- ▶ $f_0^\# : \mathbb{D}_0^\# \rightarrow \mathbb{D}_0^\#$ is a sound transfer function with respect to f , that is such that $f \circ \gamma_0 \subseteq \gamma_0 \circ f_0^\#$
- ▶ $f_1^\# : \mathbb{D}_1^\# \rightarrow \mathbb{D}_1^\#$ achieves the same condition in $\mathbb{D}_1^\#$

Then, we let $f_x^\#$ be defined by:

$$\begin{aligned} f_x^\# : \mathbb{D}_x^\# &\longrightarrow \mathbb{D}_x^\# \\ (x_0, x_1) &\longmapsto (f_0^\#(x_0), f_1^\#(x_1)) \end{aligned}$$

Then $f_x^\#$ is sound with respect to f

Transfer functions in the product abstraction

We consider **the interval abstraction** as a **product of constraints**

- \mathbb{D} is $\mathcal{P}(\mathbb{Z})$ and \subseteq the set inclusion
- \mathbb{D}_0^\sharp is $\mathbb{Z} \cup \{-\infty, +\infty\}$, \sqsubseteq_0^\sharp is \leq and $\alpha_0(E) = \inf E$
- \mathbb{D}_1^\sharp is $\mathbb{Z} \cup \{-\infty, +\infty\}$, \sqsubseteq_1^\sharp is \leq and $\alpha_1(E) = \sup E$

We consider the concrete function $f : x \mapsto -x$

- The lower bound before gives no information on the lower bound after:
 $f_0^\sharp : x_0 \mapsto -\infty$
- The same goes for the upper bounds: $f_1^\sharp : x_1 \mapsto +\infty$
- Hence, $f_x^\sharp(x_0, x_1) =]-\infty, +\infty[= \top$
- Though, we would like the more precise: $(x_0, x_1) \mapsto (-x_1, -x_0)$

- Decomposed transfer function may lose precision
- Decomposing the interval abstract domain in a product abstraction does not make sense for the computation of transfer functions

Transfer functions in the product abstraction

We now consider the product of intervals and congruences, with transfer functions:

- \mathbb{D} is $\mathcal{P}(\mathbb{Z})$ and \subseteq the set inclusion
 - **Test:** $f(t, \mathcal{E}) = \{z \in \mathbb{Z} \mid \llbracket t \rrbracket(v \mapsto z) = \text{TRUE}\}$ returns the values that satisfy condition t on variable v
 - **Random add:** $g(\mathcal{E}) = \{x + k \mid x \in \mathcal{E} \wedge -1 \leq k \leq 1\}$
-
- $x^\# ::= ([0, 10], \langle 0, 2 \rangle)$
 - $y^\# ::= p_x^\#(v = 5, x^\#) = ([5, 5], \perp)$
 - $\gamma_x(y^\#) = \emptyset$
 - why not $y^\# = (\perp, \perp)$ then ?
 - $x^\# ::= ([0, 10], \langle 0, 2 \rangle)$
 - $y^\# ::= p_x^\#(v \leq 5, x^\#) = ([0, 5], \langle 0, 2 \rangle)$
 - $z^\# ::= p_x^\#(v \geq 5, y^\#) = ([5, 5], \langle 0, 2 \rangle)$
 - $\gamma_x(z^\#) = \emptyset$
 - why not $z^\# = (\perp, \perp)$ then ?

Improving transfer functions

We consider the program:

```

assume( $x \in [0, 10]$ , even);
if( $x \leq 5$ ){
    if( $x \geq 5$ ){
         $x + \text{rand}([-1, 1])$ ;
        assert(FALSE);
    }
}

```

- analysis, from state $x^\# ::= ([0, 10], \langle 0, 2 \rangle)$
- $y^\# ::= p_x^\#(v \leq 5, x^\#) = ([0, 5], \langle 0, 2 \rangle)$
- $z^\# ::= p_x^\#(v \geq 5, y^\#) = ([5, 5], \langle 0, 2 \rangle)$
- $v^\# ::= g^\#(z^\#) = ([4, 6], \langle 0, 1 \rangle)$

Then, we notice that:

- In the concrete, the body of the second **if** is **unreachable**
- In the abstract, $\gamma_x(v^\#) = \{4, 5, 6\} \neq \emptyset$
- The product abstraction misses the fact that:

$$x = 5 \wedge x \equiv 0 \pmod{2} \implies x \in \emptyset$$

Limitations of product abstraction

- It does not allow information be sent from one domain to the other
- This is the source of a **loss of precision** in the analysis

How to overcome this ?

Outline

- 1 Introduction
- 2 Abstraction of partitioned systems
- 3 Product of abstractions
- 4 Reduction and application to reduced product**
- 5 Reduced cardinal power abstraction
- 6 State partitioning, trace partitioning
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Injective concretization

We consider the loss of information in the interval + congruences example:

- $\gamma_{\times}([5, 5], \langle 0, 2 \rangle) = \emptyset = \gamma_{\times}(\perp, \perp)$
- $\mathfrak{g}([5, 5], \langle 0, 2 \rangle) = ([4, 6], \langle 0, 1 \rangle)$
- $\mathfrak{g}(\perp, \perp) = (\perp, \perp)$, which means that (\perp, \perp) is **much more useful** for the rest of the analysis than $([5, 5], \langle 0, 2 \rangle)$
- converting $([5, 5], \langle 0, 2 \rangle)$ into (\perp, \perp) amounts to applying the mathematical result:

$$x = 5 \wedge x \equiv 0 \pmod{2} \implies x \in \emptyset$$

- Some product elements are semantically “equivalent” for computing other transfer functions, proving semantic assertions...
- Some semantically equivalent product elements are “better”
Computing those “better” elements is **reduction**

Galois surjection (or Galois insertion)

Definition

Let us consider an abstraction defined by a Galois connection

$$(\mathbb{D}, \subseteq) \xleftrightarrow[\alpha]{\gamma} (\mathbb{D}^\#, \subseteq^\#)$$

Then, the following properties are equivalent:

- α is surjective (onto)
- γ is injective (into)
- $\alpha \circ \gamma = \lambda(x \in \mathbb{D}^\#) \cdot x$

When they hold, the Galois connection is said to be a **Galois insertion**

Intuition:

- there is no pair of distinct abstract elements with the same meaning
- less chance of losing precision by taking the “wrong” abstraction of concrete property x

Galois surjection (or Galois insertion)

Proof:

- Let us assume α surjective, i.e. $\forall y \in \mathbb{D}^\sharp, \exists x \in \mathbb{D}, \alpha(x) = y$.
If $\gamma(x) = \gamma(y)$,
 - ▶ as α is surjective, there exist $x', y' \in \mathbb{D}$, such that $\alpha(x') = x$ and $\alpha(y') = y$
 - ▶ thus, $\gamma(\alpha(x')) = \gamma(\alpha(y'))$, which implies $x' \subseteq \gamma(\alpha(y'))$, and thus $\alpha(x') \sqsubseteq^\sharp \alpha(y')$ ($\alpha \circ \gamma \circ \alpha = \alpha$)
 - ▶ similarly $\alpha(y') \sqsubseteq^\sharp \alpha(x')$, thus $x = y$
- Let us assume γ is injective:
Let $y \in \mathbb{D}^\sharp$; as $\gamma \circ \alpha \circ \gamma = \gamma$, we get that $\gamma \circ \alpha \circ \gamma(y) = \gamma(y)$, thus $\alpha \circ \gamma(y) = y$
- Let us assume that $\alpha \circ \gamma$ is the identity, and let $y \in \mathbb{D}^\sharp$. Then, $\alpha \circ \gamma(y) = y$, which means there exists $x \in \mathbb{D}$ such that $\alpha(x) = y$. Thus α is surjective.

Reduction of an abstraction

Quotient abstract domain

Let us consider an abstraction defined by a Galois connection

$$(\mathbb{D}, \sqsubseteq) \xrightleftharpoons[\alpha]{\gamma} (\mathbb{D}^\#, \sqsubseteq^\#)$$

We let \equiv be the equivalence relation over $\mathbb{D}^\#$ defined by:

$$\forall x, y \in \mathbb{D}^\#, x \equiv y \iff \gamma(x) = \gamma(y)$$

We define the **quotient abstract domain** $(\mathbb{D}_{\equiv}^\#, \sqsubseteq_{\equiv}^\#)$ by:

- $\mathbb{D}_{\equiv}^\#$ is the set of equivalence classes of $\mathbb{D}^\#$ for \equiv
- $\bar{x} \sqsubseteq_{\equiv}^\# \bar{y} \iff x \sqsubseteq^\# y$

Proof:

- \equiv is an equivalence relation, so the quotient is well-defined
- well-definedness of $\sqsubseteq_{\equiv}^\#$: exercise

Reduction of an abstraction

Reduced abstraction (sing the same notations)

The reduced abstraction is defined by the Galois connection

$$(\mathbb{D}, \subseteq) \begin{matrix} \xleftarrow{\gamma_{\equiv}} \\ \xrightarrow{\alpha_{\equiv}} \end{matrix} (\mathbb{D}_{\equiv}^{\#}, \subseteq_{\equiv}^{\#})$$

where

$$\begin{array}{lcl} \alpha_{\equiv} : \mathbb{D} & \longrightarrow & \mathbb{D}_{\equiv}^{\#} \\ x & \longmapsto & \alpha(x) \end{array} \quad \begin{array}{lcl} \gamma_{\equiv} : \mathbb{D}_{\equiv}^{\#} & \longrightarrow & \mathbb{D} \\ \bar{x} & \longmapsto & \gamma(x) \end{array}$$

The above Galois connection is a Galois insertion.

Proof:

- well-definedness of γ , Galois insertion property: exercises

Notes:

- the construction works even with no α
- representation of abstract element: **use representants** of equivalence classes, i.e. elements of $\mathbb{D}_{\equiv}^{\#}$ are **selected** elements of $\mathbb{D}^{\#}$

Reduction operator

Definition (using the same notations)

A reduction operator over \mathbb{D}^\sharp is an operator ρ_{\equiv} such that:

- $\forall x \in \mathbb{D}^\sharp, \gamma(\rho_{\equiv}(x)) = \gamma(x)$;
- $\forall x, y \in \mathbb{D}^\sharp, \gamma(x) = \gamma(y) \implies \rho_{\equiv}(x) = \rho_{\equiv}(y)$

Such an operator allows to construct the quotient abstraction, using elements of \mathbb{D}^\sharp to represent equivalence classes, thanks to the following definitions:

- $\mathbb{D}_{\equiv}^\sharp = \mathbb{D}^\sharp$;
- $\alpha_{\equiv}(x) = \rho_{\equiv}(\alpha(x))$
- $\gamma_{\equiv}(x) = \gamma(x)$

Note:

- the construction works even with no α

Example: reduction of intervals as a product

We still use:

- \mathbb{D} is $\mathcal{P}(\mathbb{Z})$ and \subseteq the set inclusion
- $\mathbb{D}_0^\#$ is $\mathbb{Z} \cup \{-\infty, +\infty\}$, $\sqsubseteq_0^\#$ is \leq and $\alpha_0(E) = \sup E$
- $\mathbb{D}_1^\#$ is $\mathbb{Z} \cup \{-\infty, +\infty\}$, $\sqsubseteq_1^\#$ is \leq and $\alpha_1(E) = \inf E$

We write $\perp = (+\infty, -\infty)$, and we let:

$$\rho_{\equiv} : \mathbb{D}_x^\# \longrightarrow \mathbb{D}_x^\#$$

$$(x, y) \longmapsto \begin{cases} (x, y) & \text{if } x \leq y \\ \perp & \text{if } x > y \end{cases}$$

- ρ_{\equiv} defines a reduction operator over $\mathbb{D}_x^\#$
- this does not solve the issue of the transfer function for $x \mapsto -x$

Proof: exercise

Example: reduction of interval + congruences

We still use:

- \mathbb{D} is $\mathcal{P}(\mathbb{Z})$ and \subseteq the set inclusion
- $\mathbb{D}_0^\#$ is the interval abstract domain (an abstract values is either \perp or a pair of elements of $\mathbb{Z} \cup \{-\infty, +\infty\}$)
- $\mathbb{D}_1^\#$ is the congruences abstract domain:
 - ▶ abstract values are \perp , or of the form $\langle a, b \rangle$ with $0 \leq a < b$ or $b = 0$
 - ▶ $\gamma_1(\perp) = \emptyset$ and $\gamma_1(\langle a, b \rangle) = \{a + k \cdot b \mid k \in \mathbb{Z}\}$

Exercise: define ρ_{\equiv}

- 1 reduce to (\perp, \perp) when the concretization is empty:
 $\rho_{\equiv}([1, 4], \langle 0, 5 \rangle) = (\perp, \perp)$
- 2 reduce interval bounds to match the congruence constraint
 $\rho_{\equiv}([0, 10], \langle 3, 6 \rangle) = ([3, 9], \langle 3, 6 \rangle)$
- 3 build a congruence constraint when there is none and the interval contains only one value
 $\rho_{\equiv}([5, 5], \langle 0, 1 \rangle) = ([5, 5], \langle 5, 0 \rangle)$

This solves the imprecision in the example

Example: reduction of non relational abstractions

Assumptions:

- $\mathbb{D} = \mathcal{P}(\mathbb{X} \rightarrow \mathbb{V})$, and \subseteq is the inclusion order
- $\mathbb{D}^\# = \mathbb{X} \rightarrow \mathcal{P}(\mathbb{V})$, and $\sqsubseteq^\#$ is the pointwise inclusion
- α, γ define the non relational abstraction, by

$$\begin{aligned}\alpha(\mathcal{E}) &= \lambda(x \in \mathbb{X}) \cdot \{\phi(x) \mid \phi \in \mathcal{E}\} \\ \gamma(\phi^\#) &= \{\phi : \mathbb{X} \rightarrow \mathbb{V} \mid \forall x \in \mathbb{X}, \phi(x) \in \phi^\#(x)\}\end{aligned}$$

Then, for all $x \in \mathbb{X}$, if $\phi^\# \in \mathbb{D}^\#$ is such that $\phi^\#(x) = \emptyset$, then $\gamma(\phi^\#) = \emptyset$

- we let $\perp = \lambda(x \in \mathbb{X}) \cdot \emptyset$
- the reduction operator ρ_{\equiv} is defined by (Proof: exercise):

$$\begin{aligned}\rho_{\equiv} : \mathbb{D}^\# &\longrightarrow \mathbb{D}^\# \\ \phi^\# &\longmapsto \begin{cases} \phi^\# & \text{if } \forall x \in \mathbb{X}, \phi^\#(x) \neq \emptyset \\ \perp & \text{if } \exists x \in \mathbb{X}, \phi^\#(x) = \emptyset \end{cases}\end{aligned}$$

Thus, we can view non relational abstraction as **a reduced product over $|\mathbb{X}|$ instances of $(\mathcal{P}(\mathbb{V}), \subseteq)$**

Operations in the reduced domain

We define abstract operations on $\mathbb{D}_{\equiv}^{\sharp}$ from operations on \mathbb{D}^{\sharp} :

- **Least element:** if \perp is the least element of \mathbb{D}^{\sharp} , then $\rho_{\equiv}(\perp)$ is the least element of $\mathbb{D}_{\equiv}^{\sharp}$;
- **Upper bound:** if \sqcup is a sound upper bound operator on \mathbb{D}^{\sharp} then \sqcup_{\equiv} defined by $x \sqcup_{\equiv} y = \rho_{\equiv}(x \sqcup y)$ is a sound upper bound operator on $\mathbb{D}_{\equiv}^{\sharp}$;
- **Transfer functions:**

We assume that:

- ▶ $f : \mathbb{D} \rightarrow \mathbb{D}$ is a concrete transfer function (e.g., describing the effect of a test or of an assignment)
- ▶ $f^{\sharp} : \mathbb{D}^{\sharp} \rightarrow \mathbb{D}^{\sharp}$ is a sound transfer function with respect to f , that is such that $f \circ \gamma \subseteq \gamma \circ f^{\sharp}$

Then, f_{\equiv}^{\sharp} defined below is sound with respect to f :

$$\begin{array}{rcl}
 f_{\equiv}^{\sharp} : & \mathbb{D}_{\equiv}^{\sharp} & \longrightarrow & \mathbb{D}_{\equiv}^{\sharp} \\
 & x & \longmapsto & \rho_{\equiv}(f^{\sharp}(x))
 \end{array}$$

Caveat 1: widening

This construction does not work for widening

- Termination condition of ∇ on \mathbb{D}^\sharp :
for all sequence $(x_n^\sharp)_{n \in \mathbb{N}}$, the sequence $(y_n^\sharp)_{n \in \mathbb{N}}$ defined below is ultimately stationary:

$$y_0^\sharp = x_0^\sharp \quad \forall n \in \mathbb{N}, y_{n+1}^\sharp = y_n^\sharp \nabla x_{n+1}^\sharp$$

- Applying ρ_{\equiv} to the widening output would boil down to:

$$y_0^\sharp = \rho_{\equiv}(x_0^\sharp) \quad \forall n \in \mathbb{N}, y_{n+1}^\sharp = \rho_{\equiv}(y_n^\sharp \nabla x_{n+1}^\sharp)$$

Thus the termination condition of ∇ **does not apply here**

Solution

- Simply use ∇ on \mathbb{D}^\sharp
- Apply reduction in the body of loops (whenever we like)

Caveat 2: reduction cost

The optimal reduction function may be computationally very costly

Approximate reduction function

An **approximate reduction operator** is an operator $\rho_{\equiv} : \mathbb{D}^{\#} \rightarrow \mathbb{D}^{\#}$ which preserves concretization:

$$\forall x^{\#} \in \mathbb{D}^{\#}, \gamma(\rho_{\equiv}(x^{\#})) = \gamma(x^{\#})$$

We can require additional conditions such as:

- idempotence: $\forall x^{\#} \in \mathbb{D}^{\#}, \rho_{\equiv} \circ \rho_{\equiv}(x^{\#}) = \rho_{\equiv}(x^{\#})$
- contraction: $\forall x^{\#} \in \mathbb{D}^{\#}, \rho_{\equiv}(x^{\#}) \sqsubseteq^{\#} x^{\#}$

In all cases, **we may not obtain the reduced abstraction**

Reduced product abstraction

Definition

The reduced product abstraction is obtained by applying the reduction to the product abstraction

- **Examples:** as seen previously
 - ▶ intervals as products of constraints
 - ▶ intervals and congruences
 - ▶ non relational abstraction
- Abstract operators and transfer functions are defined by composition with reduction
- In many cases, only a partial reduction can be applied i.e., an approximation of reduced product is used

Reduced product: implementation

The modularity of the abstraction

- The whole point of reduced product is to keep the domain implementations separate
- The reduction operator should reflect that

To achieve this, we typically use a separate constraint language:

Reduced product interface

- \mathcal{C} is a set of constraints with a concretization function $\gamma_{\mathcal{C}} : \mathcal{C} \rightarrow \mathbb{D}$
- $\text{read}_i : \mathbb{D}_i^{\#} \rightarrow \mathcal{C}$, such that $\gamma_i(x_i^{\#}) \subseteq \gamma(\text{read}_i(x_i^{\#}))$
- $\text{constr}_i : \mathbb{D}_i^{\#} \times \mathcal{C} \rightarrow \mathbb{D}_i^{\#}$ such that $\gamma_i(x_i^{\#}) \cap \gamma_{\mathcal{C}}(c) \subseteq \gamma_i(\text{constr}_i(x_i^{\#}, c))$

Then, a simple reduction is: $\rho_{\equiv}(x_0^{\#}, x_1^{\#}) = (x_0^{\#}, \text{constr}_1(x_1^{\#}, \text{read}_0(x_0^{\#})))$

- **Example**, non relational abstraction: **read** = “is empty”
- Already demonstrated in the previous lecture

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Example

We consider the program and the basic abstractions below [CC'79]:

```

int x = 100;
bool b = TRUE;
while(b){
    x = x - 1;
    b = x > 0;
}

```

Property to establish:
 $x = 0$ at the end

Basic abstractions:

- possible values for b:
 $\{\emptyset, \{\mathcal{T}\}, \{\mathcal{F}\}, \{\mathcal{T}, \mathcal{F}\}\}$
- sign abstraction of x:
 $(\perp, = 0, < 0, > 0, \neq 0, \geq 0, \leq 0)$

Properties:

loop head

$$b \implies x > 0$$

loop end

$$\begin{cases} b \implies x > 0 \\ \neg b \implies x = 0 \end{cases}$$

Cardinal power abstraction

Definition

We assume $\mathbb{D} = \mathcal{P}(\mathcal{E})$, and that two abstractions are given by their concretization functions:

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

We let:

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

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

Then γ_{\rightarrow} defines a **cardinal power abstraction**

Example

Back to the example:

- \mathbb{D}_0^\sharp : abstraction of the values of b ;
- \mathbb{D}_1^\sharp : sign abstraction of the values of x ;
- the properties needed to establish the condition on the exit states are all expressible in the cardinal power abstraction

Intuition:

- cardinal power allows to express properties of the form $\bigwedge_{i \in I} (A_i \Rightarrow B_i)$
- exercise: prove that partitioning is a cardinal power abstraction

Reduction

- In general, the cardinal power is not a reduced abstraction (γ_{\rightarrow} not injective)
- Reduced cardinal power is obtained by composing the reduction construction

Application: control state partitioning abstraction

Assumptions:

- $\mathbb{D} = \mathcal{P}(\mathbb{S})$ where $\mathbb{S} = \mathbb{L} \times \mathbb{M}$
- $\mathbb{D}_0^\# = \mathbb{L} \uplus \{\perp, \top\}$
- $\mathbb{D}_1^\# = \mathcal{P}(\mathbb{M})$, ordered with the inclusion

Then, if Φ is an element of the reduced cardinal power,

- By reduction, $\Phi(\perp) = \emptyset$ and $\Phi(\top) = \bigcup\{\Phi(I) \mid I \in \mathbb{L}\}$
- Moreover:

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

- Thus is the control state partitioning abstraction
- This property also holds for partitioning abstraction in general

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Disjunctions in static analysis

Unusual computation of the absolute value:

```

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

```

- Interval abstraction:
 - ▶ after the **if**, $s \in [-1, 1]$
 - ▶ **possible division by 0**
- Same with polyhedra, octagons (convex abstractions)
- Interval + congruences would work

What if we want to use intervals only ?
Disjunctions are needed

Disjunctive completion

Definition

We consider an abstraction defined by a concretization function

$$\gamma : (\mathbb{D}^\#, \sqsubseteq^\#) \longrightarrow (\mathbb{D}, \subseteq).$$

The disjunctive completion abstraction is defined by:

- $\mathbb{D}_\vee^\# = \mathcal{P}(\mathbb{D}^\#)$
- $\sqsubseteq_\vee^\#$ is defined by:

$$\mathcal{E}^\# \sqsubseteq_\vee^\# \mathcal{F}^\# \iff \forall e^\# \in \mathcal{E}^\#, \exists f^\# \in \mathcal{F}^\#, e^\# \sqsubseteq^\# f^\#$$

- $\forall \mathcal{E}^\# \in \mathbb{D}, \gamma_\vee(\mathcal{E}^\#) = \bigcup \{ \gamma(e^\#) \mid e^\# \in \mathcal{E}^\# \}$
- $\forall x \in \mathbb{D}, \alpha_\vee(x) = \{ e^\# \in \mathbb{D}^\# \mid x \subseteq \gamma(e^\#) \}$

These define a Galois connection $(\mathbb{D}, \subseteq) \xleftarrow{\gamma_\vee} (\mathbb{D}_\vee^\#, \sqsubseteq_\vee^\#) \xrightarrow{\alpha_\vee}$

- **Proof:** exercise

State partitioning

- **Disjunctive completion** has **several severe limitations**:
 - ▶ analyses may manipulate huge abstract states
 - ▶ no obvious widening: has to be defined on a per case basis
it may be non trivial to define one
 - ▶ this abstraction ignores properties of the system to analyze
- **Partitioning allows to express disjunctions too**

Flashback: partitioning abstraction

Given set E and a partition \mathfrak{P} of E , we let the **partitioning abstraction** over E be defined by:

$$\begin{array}{lcl} \gamma_{\text{part}} : & (\mathfrak{P} \rightarrow \mathcal{P}(E)) & \longrightarrow \mathcal{P}(E) \\ & \Phi & \longmapsto \bigcup_{p \in \mathfrak{P}} \Phi(p) \end{array}$$

- **Advantages**:
 - ▶ the size of disjunctions is bounded by \mathfrak{P}
 - ▶ the choice of \mathfrak{P} can exploit problem properties

State partitioning based on values

Back to our example, **we design a cardinal power abstraction:**

```

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

- \mathbb{D}_0^\sharp : interval of x
- \mathbb{D}_1^\sharp : intervals for all variables
- Property at the end of the **if**:

$$\begin{cases} x \in [0, +\infty[& \Rightarrow \mathbf{s} = 1 \wedge \dots \\ x \in]-\infty, -1] & \Rightarrow \mathbf{s} = -1 \wedge \dots \end{cases}$$

- Some of the issues of disjunctive completion remain:
in particular, no obvious widening...
- Representing the full cardinal power is too costly:
limit the number of partitions

Transfer functions

```

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

```

- At ①:

$$\begin{cases} x \in [0, +\infty[& \Rightarrow s = -1 \wedge \dots \\ x \in]-\infty, -1] & \Rightarrow s = 1 \wedge \dots \end{cases}$$

- At ②:

$$\begin{cases} x \in [1, +\infty[& \Rightarrow s = 1 \wedge \dots \\ x \in]-\infty, 0] & \Rightarrow s = -1 \wedge \dots \end{cases}$$

Most abstract transfer functions may modify **both sides of the cardinal power**:

- The assignment to x modifies the abstraction in the left hand side of the cardinal power
- Thus partitions need to be recomputed: costly operation

Trace partitioning abstraction: example

Alternate way to look at the example:

```

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

```

At ①:

- if the execution went through the TRUE branch of the if:

$$x \in [0, +\infty[\wedge s = 1 \wedge$$

- if the execution went through the FALSE branch of the if:

$$x \in] - \infty, -1] \wedge s = -1 \wedge$$

- This abstraction should be formalized as an abstraction of traces, not states

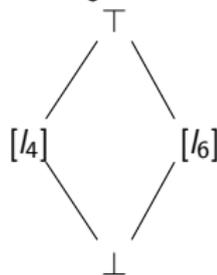
Trace partitioning abstraction: formalization

```

l0 : int x ∈ ℤ;
l1 : int s;
l2 : int y;
l3 : if(x ≥ 0){
l4 :     s = 1;
l5 : } else {
l6 :     s = -1;
l7 : }
l8 : y = x/s;

```

- Trace domain \mathbb{D}_0^\sharp :



- Concretization γ_0 :

$$\begin{aligned} \gamma_0 : [l_4] &\mapsto \{ \langle \dots, (l_4, m), \dots \rangle \in \mathcal{S}^* \} \\ [l_6] &\mapsto \{ \langle \dots, (l_6, m), \dots \rangle \in \mathcal{S}^* \} \end{aligned}$$

- Right hand side abstraction:

$(\mathcal{P}(\mathcal{S}), \subseteq)$, with abstraction defined by
 $(\alpha_{\mathcal{R}}, \gamma_{\mathcal{R}})$

Trace partitioning abstraction: definition

Definition: static trace partitioning

Let $(\mathbb{D}_0^\#, \sqsubseteq_0^\#)$ be a *finite* abstraction of sets of traces, defined by a Galois connection:

$$(\mathcal{P}(S^*), \subseteq) \begin{matrix} \xleftarrow{\gamma_0} \\ \xrightarrow{\alpha_0} \end{matrix} (\mathbb{D}_0^\#, \sqsubseteq_0^\#)$$

It defines a **static trace partitioning abstraction** by reduced cardinal power over the reachability abstraction.

There are many ways to instantiate $\mathbb{D}_0^\#$:

Trace partitioning criteria

- control flow based criteria:
 - ▶ branch taken in a **if** statement
 - ▶ number of times a **while** body was executed
- value of some variable at a given point
- conjunctions of such criteria

Trace partitioning transfer functions

We assume \mathbb{D}_0^\sharp is finite (case \mathbb{D}_0 is infinite: dynamic partitioning, see later)

Static partitioning composed with state abstraction

By composing a state abstraction $(\mathcal{P}(\mathbb{S}), \subseteq) \xrightleftharpoons[\alpha_1]{\gamma_1} (\mathbb{D}_1^\sharp, \sqsubseteq_1^\sharp)$, and applying the same reduced cardinal power abstraction, we get a new instance of the static trace partitioning abstraction

- **Least element:** $\lambda(x^\sharp \in \mathbb{D}_0^\sharp) \cdot \perp_1$
- **Upper bound:** $\phi^\sharp \sqcup \psi^\sharp ::= \lambda(x^\sharp \in \mathbb{D}_0^\sharp) \cdot (\phi^\sharp(x^\sharp) \sqcup_1 \psi^\sharp(x^\sharp))$
- **Widening operator:** similar definition
- **Transfer functions with no partition change:** We assume that:
 - ▶ $f : \mathbb{D} \rightarrow \mathbb{D}$ is a concrete transfer function (e.g., describing the effect of a test or of an assignment)
 - ▶ $f_1^\sharp : \mathbb{D}_1^\sharp \rightarrow \mathbb{D}_1^\sharp$ is a sound transfer function with respect to f , that is such that $f \circ \gamma \subseteq \gamma \circ f_1^\sharp$

Then, $\lambda(x^\sharp \in \mathbb{D}_0^\sharp) \cdot f_1^\sharp$ is sound with respect to f

Transfer functions in the trace partitioning domain

Control history based partitioning:

Abstract partition matching

A sound abstract partition matching is a family of relations $(\rightarrow_{l,l'}^{\sharp})_{l,l' \in \mathbb{L}}$ where $\rightarrow_{l,l'}^{\sharp} \subseteq (\mathbb{D}_0^{\sharp})^2$, such that:

$$\left. \begin{array}{l} \langle (l_0, m_0), \dots, (l_n, m_n) \rangle \in \gamma_0(x^{\sharp}) \\ \wedge x^{\sharp} \rightarrow_{l_n, l_{n+1}}^{\sharp} y^{\sharp} \end{array} \right\} \Rightarrow \langle (l_0, m_0), \dots, (l_{n+1}, m_{n+1}) \rangle \in \gamma_0(y^{\sharp})$$

Analysis of a transition

Given a sound abstract partition matching $\rightarrow_{l,l'}$, and sound transfer function $f_{l,l'} : \mathbb{D}_1^{\sharp} \rightarrow \mathbb{D}_1^{\sharp}$ in the underlying domain, the transfer function below in the trace partitioning domain is sound:

$$\phi^{\sharp} \longmapsto \lambda(x^{\sharp} \in \mathbb{D}_0^{\sharp}) \cdot \sqcup_1 \{f_{l,l'}(\phi^{\sharp}(y^{\sharp})) \mid y^{\sharp} \rightarrow_{l,l'} x^{\sharp}\}$$

Creation and fusion of trace partitions

- **Proof** of soundness: exercise
- **Typical choice for the abstract partition matching:**
 - ▶ at most points, the partitions are unchanged
i.e., $\rightarrow_{I,I'}$ is the identity relation
 - ▶ at points where partitions should be merged, it reflects creation of partitions or fusion of partitions
- **Other partitioning criteria:** should provide similar operations on partitions

Dynamic partitioning

Principle:

- the domain of partitions depends on the context
- can be applied to state partitioning, trace partitioning...
 - ▶ in trace partitioning, this corresponds to cases where $\mathbb{D}_0^\#$ is infinite
 - ▶ indeed, only a finite number of partitions can be represented at any point in the analysis; this set is dynamic (i.e., also determined as a result of the analysis)

Formalization: cofibered abstract domain [AV], [MR'05]

Outline

- 1 Introduction
- 2 Abstraction of partitioned systems
- 3 Product of abstractions
- 4 Reduction and application to reduced product
- 5 Reduced cardinal power abstraction
- 6 State partitioning, trace partitioning
- 7 Concluding remarks**

Main points of the lecture

There exists many techniques to combine abstract domains into more interesting ones

- **Product, reduced product:**
conjunctions of abstract properties
- **Partitioning, disjunctive completion:**
disjunctions of abstract properties
- The list is not exhaustive

Advantages

- **Modular** design of static analyzers
- A same construction **may be used in many contexts**

Bibliography: abstract domain combination

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- [MR'05]: Trace partitioning in abstract interpretation static analyzers. Laurent Mauborgne and Xavier Rival. In ESOP, 2005.