Abstract Interpretation (an introduction)

COST Action IC0701 - 2nd Action Training School

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Static program analysis

The goals of static program analysis

- to prove properties about the run-time behaviour of a program
- ▶ in a fully automatic way
- without actually executing this program

Applications

- code optimisation
- error detection (array out of bound access, null pointers)
- proof support (invariant extraction)

Abstract Interpretation

[Cousot & Cousot 75, 76, 77, 79, 80, 81, 82, 84, 85, 86, 87, 88, 89, 90, 91, 92, 93, 94, 95, 96, 97, 98, 99, 00, 01, 02, 03, 04, 05, 06, 07,08,09,10,11,...]



Patrick Cousot



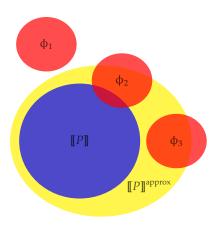
Radhia Cousot

A theory which unifies a large variety of static analysis

- formalises the approximated analyse of programs
- allows to compare relative precision of analyses
- facilitates the conception of sophisticated analyses

See http://www.di.ens.fr/~cousot/

Static analysis computes approximations²



 $\llbracket P \rrbracket$:

ightharpoonup P is safe w.r.t. φ¹ and the analyser proves it

$$\llbracket P \rrbracket \cap \varphi_1 = \emptyset$$
 $\llbracket P \rrbracket^{\operatorname{approx}} \cap \varphi_1 = \emptyset$

P is unsafe w.r.t. φ₂ and the analyser warns about it

$$\llbracket P \rrbracket \cap \varphi_2 \neq \emptyset \qquad \llbracket P \rrbracket^{\text{approx}} \cap \varphi_2 \neq \emptyset$$

 but P is safe w.r.t. φ₃ and the analyser can't prove it (this is called a *false alarm*)

$$\llbracket P \rrbracket \cap \varphi_3 = \emptyset \qquad \llbracket P \rrbracket^{\text{approx}} \cap \varphi_3 \neq \emptyset$$

concrete semantics (e.g. set of reachable states)

 ϕ_1, ϕ_2, ϕ_3 : erroneous/dangerous set of states

 $[\![P]\!]^{approx}$: analyser result (here over-approximation)

(not computable) (computable)

(computable)

Abstract interpretation executes programs on state properties instead of states.

- A state property is a subset in $\mathcal{P}(\mathbb{Z}^2)$ of (\mathbf{x}, \mathbf{y}) values.
- When a point is reached for a second time we make an union with the previous property.
- We "execute" the program until stability
 - It may take an infinite number of steps...
 - But the limit always exists (explained later)

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```
x = 0; y = 0;
       \{(0,0)\}
while (x<6) {
  if (?) {
       \{(0,0)\}
    y = y+2;
  };
  x = x+1;
```

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while (x<6) {
  if (?) {
       \{(0,0),(1,0),(1,2)\}
    v = v+2:
       \{(0,2),(1,2),(1,4)\}
  };
       \{(0,0),(0,2),(1,0),(1,2),(1,4)
  x = x+1:
       \{(1,0),(1,2)\}
```

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  };
       \{(0,0),(0,2),(1,0),(1,2),(1,4)\}
  x = x+1;
       \{(1,0),(1,2),(2,0),(2,2),(2,4)\}
```

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x = 0; y = 0;
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while (x<6) {
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    y = y+2;
        \{(0,2),(1,2),(1,4),\dots\}
  };
        \{(0,0),(0,2),(1,0),(1,2),(1,4),\dots\}
  x = x+1:
        \{(1,0),(1,2),(2,0),(2,2),(2,4),\ldots\}
        \{(6,0),(6,2),(6,4),(6,6),\ldots\}
```

Abstract interpretation executes programs on state properties instead of states.

Approximation

► The set of manipulated properties may be restricted to ensure computability of the semantics. Example: sign of variables

$$P ::= x C 0 \land y C 0$$

$$C ::= \langle | \leq | = | \rangle | \geqslant$$

```
x = 0; y = 0;
        \mathbf{x} = 0 \land \mathbf{v} = 0
while (x<6) {
  if (?) {
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x = 0; y = 0;
x = 0 \land y = 0
while (x<6) {
   if (?) {
      x = 0 \land y = 0
      y = y+2;
      x = 0 \land y > 0 over-approximation!
};

x = x+1;
}
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x = 0; y = 0;

x = 0 \wedge y = 0

while (x<6) {

if (?) {

x = 0 \wedge y = 0

y = y+2;

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};

x = 0 \wedge y \geqslant 0

x = x+1;
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         x \ge 0 \land y \ge 0
  x = x+1:
         x > 0 \land y \ge 0
         x \ge 0 \land v \ge 0
```

An other example : the interval analysis

For each point k and each numeric variable x, we infer an interval in which x must belong to.

Example: insertion sort, array access verification

```
assert(T.length=100); i=1;
                                           \{i \in [1, 100]\}
while (i<T.length) {</pre>
                                           \{i \in [1,99]\}
     p = T[i]; j = i-1;
                                           \{i \in [1,99], j \in [-1,98]\}
     while (0 \le j \text{ and } T[j] > p) {
                                           \{i \in [1,99], i \in [0,98]\}
           T[j]=T[j+1]; j = j-1;
                                           \{i \in [1,99], i \in [-1,97]\}
     };
                                           \{i \in [1,99], i \in [-1,98]\}
     T[i+1]=p; i = i+1;
                                           \{i \in [2, 100], j = [-1, 98]\}
};
                                           \{i = 100\}
```

An other example : the polyhedral analysis

For each point k and we infer invariant linear equality and inequality relationships among variables.

Example: insertion sort, array access verification

```
assert(T.length>=1); i=1;
                                            \{1 \le i \le T.length\}
while i<T.length {</pre>
                                            \{1 \le i \le T.length - 1\}
     p = T[i]; j = i-1;
                                            \{1 \leq i \leq T.length - 1 \land -1 \leq i \leq i - 1\}
     while 0<=j and T[j]>p {
                                            \{1 \leq i \leq T.length - 1 \land 0 \leq i \leq i - 1\}
           T[j]=T[j+1]; j = j-1;
                                            \{1 \leq i \leq T.length - 1 \land -1 \leq i \leq i - 2\}
     };
                                            \{1 \le i \le T.length - 1 \land -1 \le i \le i - 1\}
     T[i+1]=p; i = i+1;
                                            \{2 \leq i \leq T.length + 1 \land -1 \leq i \leq i - 2\}
};
                                            \{i = T.length\}
```

This lecture

- Introduction
- 2 Intermediate representation: syntax and semantics
- Collecting semantics
- ¶ Just put some [♯]...
- Building a generic abstract interpreter
- 6 Numeric abstraction by intervals
- Widening/Narrowing
- Polyhedral abstract interpretation
- Readings

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A flowchart representation of program

The standard model of program in static analysis is *control flow graph*. The graph model used here :

- ▶ the nodes are program point $k \in \mathbb{P}$,
- the edges are labeled with basic instructions

(*Exp* and *Test* to be defined in the next slide)

- formally a cfg is a couple (k_{init}, S) with
 - $k_{\text{init}} \in \mathbb{P}$: the entry point,
 - $S \subseteq \mathbb{P} \times Instr \times \mathbb{P}$ the set of edges.

Remark: data-flow analyses are generally based on other versions of control flow graph (nodes are put in instructions).

Expression and test language for today

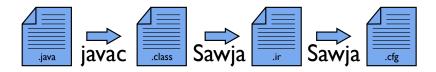
In OCaml syntax

We will restrict our study to a simple numeric subset of Java expressions

```
type binop =
  | Add | Sub | Mult
type expr =
  | Const of int
  | Var of var
  | Binop of binop * expr * expr
type comp = Eq | Neq | Le | Lt
type instr =
   Nop
                           (* x := ? *)
  | Forget of var
 | Assign of var * expr (* x := e *)
  | Assert of expr * comp * expr (* e_1 cmp e_2 *)
```

From Java to CFG

http://sawja.inria.fr/



See

- ▶ D. Demange, T. Jensen, and D. Pichardie. *A provably correct stackless intermediate representation for Java bytecode*. In Proc. of APLAS 2010.
- L. Hubert, N. Barré, F. Besson, D. Demange, T. Jensen, V. Monfort, D. Pichardie, and T. Turpin. *Sawja: Static Analysis Workshop for Java*. In Proc. of FoVeOOS 2010.

Demo

Semantics

Semantic domains

$$\begin{array}{ccc} \textit{Env} & \stackrel{\text{def}}{=} & \mathbb{V} \to \mathbb{Z} \\ \textit{State} & \stackrel{\text{def}}{=} & \mathbb{P} \times \textit{Env} \end{array}$$

Semantics of expressions (standard then omitted)

$$A[e] \rho \in \mathbb{Z}, e \in Exp, \rho \in Env$$

Semantics of tests (standard then omitted)

$$\mathbb{B}[t] \rho \in \mathbb{B}, \quad t \in Test, \ \rho \in Env$$

Small-step semantics of cfg

We first define the semantics of instructions : $\stackrel{i}{\rightarrow} \subseteq Env \times Env$

Then a small-step relation $\rightarrow_{\mathit{cfg}} \subseteq \mathit{State} \times \mathit{State}$ for a $\mathit{cfg} = (k_{\mathsf{init}}, S)$

$$\frac{(k_1,i,k_2) \in S \qquad \rho_1 \xrightarrow{i} \rho_2}{(k_1,\rho_1) \xrightarrow{}_{cf_X} (k_2,\rho_2)}$$

Reachable states for control flow graphs

$$\llbracket cfg \rrbracket = \{ (k, \rho) \mid \exists \rho_0 \in Env, \ (k_{init}, \rho_0) \rightarrow_{cfg}^* (k, \rho) \}$$
 where $cfg = (k_{init}, S)$

Starting from an other semantics?

Remark: for the purpose of the talk, we directly start with a *cfg*-semantics. We could have started from a more conventionnal operational semantics. See

- Patrick Cousot, MIT Course 16.399: Abstract Interpretation, http://www.mit.edu/~cousot/
- ▶ David Cachera and David Pichardie. *A certified denotational abstract interpreter*. In Proc. of ITP-10, 2010.

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Collecting Semantics

We will consider a collecting semantics that give us the set of reachable states $\llbracket p \rrbracket_k^{\text{col}}$ at each program points k.

$$\forall k \in \mathbb{P}, \ [\![p]\!]_k^{\text{col}} = \{ \ \rho \mid (k, \rho) \in [\![p]\!] \}$$

Theorem

 $\llbracket p \rrbracket^{col}$ may be characterized as the least fixpoint of the following equation system.

$$\forall k \in labels(p), \ X_k = X_k^{init} \cup \bigcup_{(k',i,k) \in p} \llbracket i \rrbracket (X_{k'})$$

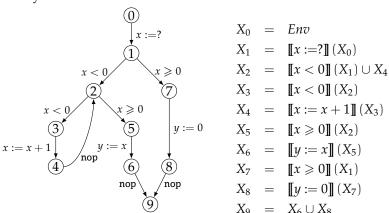
with
$$X_k^{\text{init}} = \begin{cases} Env & \text{if } k = k_{\text{init}} \\ \emptyset & \text{otherwise} \end{cases}$$
 and

$$\forall i \in \mathit{Instr}, \ \forall X \subseteq \mathit{Env}, \ \llbracket i \rrbracket \left(X \right) = \left\{ \ \rho_2 \ | \ \exists \rho_1 \in X, \ \rho_1 \xrightarrow{i} \rho_2 \ \right\} = \mathsf{post} \left[\xrightarrow{i} \left(X \right) \right]$$



Example

For the following program, $[\![P]\!]^{\mathrm{col}}$ is the least solution of the following equation system :



Fixpoint Lattice Theory

Theorem (Knaster-Tarski)

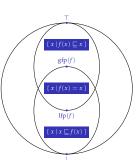
In a complete lattice $(A, \sqsubseteq, \bigsqcup)$, for all monotone functions $f \in A \to A$, the least fixpoint lfp(f) of f exists and is $\bigcap \{x \in A \mid f(x) \sqsubseteq x\}$.

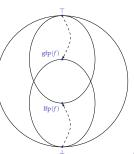
Theorem (Kleene fixpoint theorem)

In a complete lattice $(A, \sqsubseteq, \bigsqcup)$, for all continuous function $f \in A \to A$, the least fixpoint lfp(f) of f is equal to $| \{f^n(\bot) \mid n \in \mathbb{N} \}$.

Theorem

Let (A, \sqsubseteq) a poset that verifies the ascending chain condition and f a monotone function. The sequence $\bot, f(\bot), \ldots, f^n(\bot), \ldots$ eventually stabilises. Its limit is the least fixpoint of f.





Collecting semantics and exact analysis

The $(X_k)_{i=1..N}$ are hence specified as the least solution of a fixpoint equation system

$$X_k = F_k(X_1, X_2, \dots, X_N)$$
 , $k \in labels(p)$

or, equivalently $\vec{X} = \vec{F}(\vec{X})$.

Exact analysis:

- ▶ Thanks to Knaster-Tarski, the least solution exists (complete lattice, F_k are monotone functions),
- ▶ Kleen fixpoint theorem (F_k are continuous functions) says it is the limit of

$$X_k^0 = \emptyset$$
 , $X_k^{n+1} = F_k(X_1^n, X_2^n, \dots, X_N^n)$

Uncomputable problem:

- ightharpoonup Representing the X_k may be hard (infinite sets)
- ► The limit may not be reachable in a finite number of steps



Approximate analysis

Exact analysis:

```
Least solution of X = F(X) in the complete lattice (\mathcal{P}(Env)^N, \subseteq, \cup, \cap) or limit of X^0 = \bot, X^{n+1} = F(X^n)
```

Approximate analysis:

- ▶ Static approximation : we replace the concrete lattice $(\mathcal{P}(Env), \subseteq, \cup, \cap)$ by an abstract lattice $(L^{\sharp}, \sqsubseteq^{\sharp}, \sqcup^{\sharp}, \sqcap^{\sharp})$
 - whose elements can be (efficiently) represented in computers,
 - ▶ in which we know how to compute \sqcup^{\ddagger} , \sqcap^{\ddagger} , \sqsubseteq^{\ddagger} , . . .

and we "transpose" the equation X = F(X) of $\mathfrak{P}(Env)^N$ into $(L^{\sharp})^N$.

▶ Dynamic approximation: when L^{\sharp} does not verifies the ascending chain condition, the iterative computation may not terminate in a finite number of steps (or sometimes too slowly). In this case, we can only approximate the limit (see widening/narrowing).

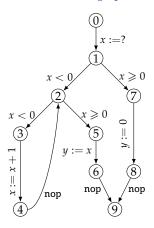
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Just put some #...

From $\mathcal{P}(Env)$ to Env^{\sharp}

control flow graph



collecting semantics

$$X_0 = Env$$

$$X_1 = [x :=?](X_0)$$
 $X_1^{\sharp} = [x :=?]^{\sharp}(X_0^{\sharp})$

$$X_2 = [x < 0](X_1) \cup X_4$$

 $X_3 = [x < 0](X_2)$

$$X_4 = [x := x + 1](X_3)$$

$$X_5 = [x \geqslant 0](X_2)$$

$$X_5 = [x \geqslant 0](X_2)$$

$$X_6 = [y := x](X_5)$$

$$X_7 = [x \ge 0](X_1)$$

$$X_8 = [y := 0](X_7)$$

$$X_8 = [y := 0](X_7)$$

$$X_9 = X_6 \cup X_8$$

abstract semantics

$$=$$
 \top_{Env}^{p}

$$= [x :=?]^{\sharp} (X_0^{\sharp})$$

$$X_{2}^{\sharp} = [x < 0]^{\sharp} (X_{1}^{\sharp}) \sqcup^{\sharp} X_{4}^{\sharp}$$

$$X_3^{\sharp} = [x < 0]^{\sharp} (X_2^{\sharp})$$

$$X_4^{\sharp} = [x := x + 1]^{\sharp} (X_3^{\sharp})$$

$$X_5^{\sharp} = [x \geqslant 0]^{\sharp} (X_2^{\sharp})$$

$$X_6^{\sharp} = \llbracket y := x \rrbracket^{\sharp} (X_5^{\sharp})$$

$$X_6 = [y - x] (X_5)$$

$$X_7^{\sharp} = [x \geqslant 0]^{\sharp} (X_1^{\sharp})$$

$$X_8^{\sharp} = [[y := 0]]^{\sharp} (X_7^{\sharp})$$

$$X_9^{\sharp} = X_6^{\sharp} \sqcup^{\sharp} X_8^{\sharp}$$

Abstract semantics: the ingredients

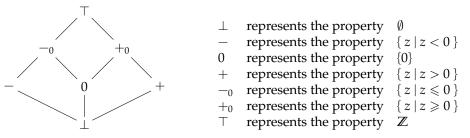
- ► A lattice structure $(Env^{\sharp}, \sqsubseteq_{Env}^{\sharp}, \sqcup_{Env}^{\sharp}, \sqcap_{Env}^{\sharp}, \perp_{Env}^{\sharp}, \top_{Env}^{\sharp})$
 - ▶ $\sqsubseteq_{Env}^{\sharp}$ is an approximation of \subseteq
 - ► \sqcup_{Env}^{\sharp} is an approximation of \cup
 - ► \sqcap_{Env}^{\sharp} is an approximation of \cap
 - \perp_{Env}^{\sharp} is an approximation of \emptyset
 - ▶ \top_{Env}^{\sharp} is an approximation of Env
- ▶ For all $x \in \mathbb{V}$,

$$\llbracket x :=? \rrbracket^{\sharp} \in Env^{\sharp} \to Env^{\sharp}$$
 an approximation of $\llbracket x :=? \rrbracket$

- ▶ For all $x \in \mathbb{V}$, $e \in Exp$,
 - $[\![x:=e]\!]^\sharp \in Env^\sharp \to Env^\sharp$ an approximation of $[\![x:=e]\!]$
- ▶ For all $t \in Test$,
 - $[\![t]\!]^{\sharp} \in Env^{\sharp} \to Env^{\sharp}$ an approximation of $[\![t]\!]$
- ▶ A concretisation $\gamma \in Env^{\sharp} \to \mathcal{P}(Env)$ that explains which property $\gamma(x^{\sharp}) \in \mathcal{P}(Env)$ is represented by each abstract element $x^{\sharp} \in Env^{\sharp}$.



An abstraction by signs



 $Env^{\sharp} \stackrel{\text{def}}{=} \mathbb{V} \to Sign$: a sign is associated to each variable.

An abstraction by signs: example

with

$$\begin{aligned} succ^{\sharp}(\bot) &= &\bot\\ succ^{\sharp}(-) &= &-_0\\ succ^{\sharp}(0) &= ≻^{\sharp}(+) = succ^{\sharp}(+_0) = +\\ succ^{\sharp}(-_0) &= ≻^{\sharp}(\top) = \top \end{aligned}$$

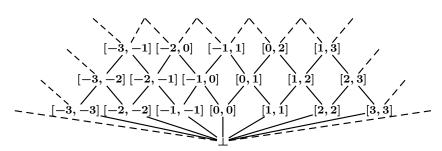


Abstraction by intervals

$$Int \stackrel{\text{def}}{=} \{ [a,b] \mid a,b \in \overline{\mathbb{Z}}, \ a \leqslant b \} \cup \{\bot\}$$

with
$$\overline{\mathbb{Z}} = \mathbb{Z} \cup \{-\infty, +\infty\}$$
.

 \perp represents \emptyset and [a,b] the property $\{z \mid a \leqslant z \leqslant b\}$.



 $Env^{\sharp} \stackrel{\text{def}}{=} \mathbb{V} \to Int$: an interval is associated to each variable.



Abstraction by intervals : example

$$\begin{array}{llll} X_0^{\sharp} & = & \top_{Env}^{\sharp} & X_0^{\sharp} & = & [x:[-\infty,+\infty]; \ y:[-\infty,+\infty]] \\ X_1^{\sharp} & = & [x:=?]^{\sharp} \ (X_0^{\sharp}) & X_1^{\sharp} & = & X_0^{\sharp} [x\mapsto [-\infty,+\infty]] \\ X_2^{\sharp} & = & [x<0]^{\sharp} \ (X_1^{\sharp}) \sqcup^{\sharp} \ X_4^{\sharp} & X_2^{\sharp} & = & X_1^{\sharp} [x\mapsto X_1^{\sharp}(x)\sqcap^{\sharp} [-\infty,-1]] \sqcup^{\sharp} \ X_4^{\sharp} \\ X_3^{\sharp} & = & [x<0]^{\sharp} \ (X_2^{\sharp}) & X_3^{\sharp} & = & X_2^{\sharp} [x\mapsto X_2^{\sharp}(x)\sqcap^{\sharp} [-\infty,-1]] \\ X_4^{\sharp} & = & [x:=x+1]^{\sharp} \ (X_3^{\sharp}) & X_4^{\sharp} & = & X_3^{\sharp} [x\mapsto \operatorname{succ}^{\sharp}(X_3^{\sharp}(x))] \\ X_5^{\sharp} & = & [x\geqslant 0]^{\sharp} \ (X_2^{\sharp}) & X_5^{\sharp} & = & X_2^{\sharp} [x\mapsto X_2^{\sharp}(x)\sqcap^{\sharp} [0,+\infty]] \\ X_6^{\sharp} & = & [y:=x]^{\sharp} \ (X_5^{\sharp}) & X_6^{\sharp} & = & X_5^{\sharp} [y\mapsto X_5^{\sharp}(x)] \\ X_7^{\sharp} & = & [x\geqslant 0]^{\sharp} \ (X_1^{\sharp}) & X_7^{\sharp} & = & X_1^{\sharp} [x\mapsto X_1^{\sharp}(x)\sqcap^{\sharp} [0,+\infty]] \\ X_8^{\sharp} & = & [y:=0]^{\sharp} \ (X_7^{\sharp}) & X_8^{\sharp} & = & X_7^{\sharp} [y\mapsto [0,0]] \\ X_9^{\sharp} & = & X_6^{\sharp} \sqcup^{\sharp} X_8^{\sharp} & X_9^{\sharp} & = & X_6^{\sharp} \sqcup^{\sharp} X_8^{\sharp} \end{array}$$

with

$$\operatorname{succ}^{\sharp}(\bot) = \bot$$

 $\operatorname{succ}^{\sharp}([a,b]) = [a+1,b+1]$

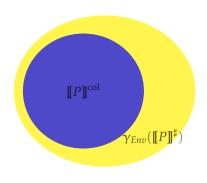


Demo

Outline

- Introduction
- 2 Intermediate representation : syntax and semantics
- 3 Collecting semantics
- 4 Just put some #....
- **5** Building a generic abstract interpreter
- Mumeric abstraction by intervals
- Widening/Narrowing
- Polyhedral abstract interpretation
- Readings

Soundness criterion



Given an environment concretisation function $\gamma_{Env} \in Env^{\sharp} \to \mathcal{P}(Env)$, we want to compute an abstract semantics $\llbracket P \rrbracket^{\sharp} \in \mathbb{P} \to Env^{\sharp}$ that is a conservative approximation of $\llbracket P \rrbracket^{\operatorname{col}}$.

$$\forall k \in \mathbb{P}, \llbracket P \rrbracket^{\text{col}}(k) \subseteq \gamma(\llbracket P \rrbracket^{\sharp}(k))$$

This leads to a sound over-approximation of $\llbracket P \rrbracket$ since $\llbracket P \rrbracket$ and $\llbracket P \rrbracket^{\operatorname{col}}$ are equivalents.

$$[\![P]\!] = \{ (k, \rho) \mid \rho \in [\![p]\!]^{\text{col}}(k) \}$$

Function approximation

When some computations in the concrete world are uncomputable or too costly, the abstract world can be used to execute a simplified version of these computations.

the abstract computation must always give a conservative answer w.r.t. the concrete computation

Let $f \in \mathcal{A} \to \mathcal{A}$ in the concrete world and $f^{\sharp} \in \mathcal{A}^{\sharp} \to \mathcal{A}^{\sharp}$ which correctly approximates each concrete computation.

$$\begin{array}{ccc}
\mathcal{A} & \stackrel{f}{\longrightarrow} & \mathcal{A} \\
\uparrow^{\gamma} & & \uparrow^{\gamma} \\
\mathcal{A}^{\sharp} & \stackrel{f^{\sharp}}{\longrightarrow} & \mathcal{A}^{\sharp}
\end{array}$$

Correctness criterion : $f \circ \gamma \sqsubseteq \gamma \circ f^{\sharp}$

Fixpoint transfert

Theorem

Given a monotone concretisation between two complete lattices $\left(\mathcal{A}^{\sharp},\sqsubseteq^{\sharp},\bigsqcup^{\sharp},\sqcap^{\sharp}\right) \to (\mathcal{A},\sqsubseteq,\bigsqcup,\sqcap)$, a function $f^{\sharp} \in \mathcal{A}^{\sharp} \to \mathcal{A}^{\sharp}$ and a monotone function $f \in \mathcal{A} \to \mathcal{A}$ which verify $f \circ \gamma \sqsubseteq \gamma \circ f^{\sharp}$, we have

$$\mathrm{lfp}(f) \sqsubseteq \gamma(\mathrm{lfp}(f^{\sharp}))$$

It means it is generally sound to mimic fixpoint computation in the abstract.

Environment abstraction: sufficient elements

Thanks to the previous theorem, it is sufficient to design an abstraction domain Env^{\sharp} with a correct approximation $[\![i]\!]^{\sharp}$ of $[\![i]\!]$ for all instructions i.

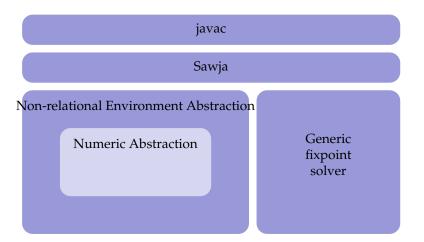
$$\forall \rho^{\sharp} \in \mathit{Env}^{\sharp}, \ \llbracket i \rrbracket \left(\gamma_{\mathit{Env}}(\rho^{\sharp}) \right) \subseteq \gamma_{\mathit{Env}}(\llbracket i \rrbracket^{\sharp} \left(\rho^{\sharp} \right) \right)$$

And $\llbracket P \rrbracket^{\sharp}$ is defined as the least fixpoint of the system :

$$\forall k \in labels(P), \ X_k^{\sharp} = X_k^{\sharp init} \sqcup^{\sharp} \bigsqcup_{(k',i,k) \in P} \llbracket i \rrbracket^{\sharp} (X_{k'}^{\sharp})$$

with
$$X_k^{\sharp init} = \left\{ \begin{array}{ll} \top_{Env} & \text{ if } k = k_{init} \\ \emptyset & \text{ otherwise} \end{array} \right.$$

A Generic Abstract Interpreter



Non-relational environment abstraction

We start with the description of a non-relational abstraction: each variable is abstracted independently.

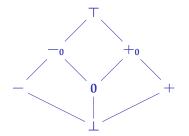
$$Env^{\sharp} \stackrel{\text{def}}{=} \mathbb{V} \to \text{Num}^{\sharp}$$

$$\forall \rho_{1}^{\sharp}, \rho_{2}^{\sharp} \in Env^{\sharp}, \ \rho_{1}^{\sharp} \sqsubseteq_{Env}^{\sharp} \rho_{2}^{\sharp} \stackrel{\text{def}}{=} \forall x \in \mathbb{V}, \ \rho_{1}^{\sharp}(x) \sqsubseteq_{\text{Num}}^{\sharp} \rho_{2}^{\sharp}(x)$$

$$\forall \rho^{\sharp} \in Env^{\sharp}, \ \gamma_{Env}(\rho^{\sharp}) \stackrel{\text{def}}{=} \left\{ \ \rho \mid \forall x \in \mathbb{V}, \ \rho(x) \in \gamma_{\text{Num}}(\rho^{\sharp}(x)) \ \right\}$$

See the of the lecture for a relational abstraction.

Sign abstraction



```
\begin{array}{lll} \gamma_{Num}(\bot) & = & \emptyset \\ \gamma_{Num}(-) & = & \{z \mid z < 0\} \\ \gamma_{Num}(0) & = & \{0\} \\ \gamma_{Num}(+) & = & \{z \mid z > 0\} \\ \gamma_{Num}(-_0) & = & \{z \mid z \le 0\} \\ \gamma_{Num}(+_0) & = & \{z \mid z \ge 0\} \\ \gamma_{Num}(\top) & = & \mathbb{Z} \end{array}
```

We will use this abstract domain as runnign example but you should keep in mind this is just an example among other numerical abstract domains.

Construction of [x := ?]

$$\llbracket x := ? \rrbracket^{\sharp} (\rho^{\sharp}) = \rho^{\sharp} [x \mapsto \top_{\text{Num}}], \ \forall \rho^{\sharp} \in Env^{\sharp}$$

with $\top_{Num} \in Num^{\sharp}$ such that $\mathbb{Z} \subseteq \gamma_{Num}(\top_{Num})$.

Construction of $[x := e]^{\sharp}$

$$\llbracket x := e \rrbracket^{\sharp} \left(\rho^{\sharp} \right) = \rho^{\sharp} \left[x \mapsto \mathcal{A} \llbracket e \rrbracket^{\sharp} \left(\rho^{\sharp} \right) \right], \ \forall \rho^{\sharp} \in Env^{\sharp}$$

with

$$\forall e \in \text{Expr}, \ \mathcal{A}\llbracket e \rrbracket^{\sharp} \in Env^{\sharp} \to \text{Num}^{\sharp}$$

a (forward) abstract evaluation of expressions

$$\mathcal{A}[[n]]^{\sharp} (\rho^{\sharp}) = \operatorname{const}^{\sharp}(n)$$

$$\mathcal{A}[[x]]^{\sharp} (\rho^{\sharp}) = \rho^{\sharp}(x)$$

$$\mathcal{A}[[e_{1} o e_{2}]]^{\sharp} (\rho^{\sharp}) = o^{\sharp} (\mathcal{A}[[e_{1}]]^{\sharp} (\rho^{\sharp}), \mathcal{A}[[e_{2}]]^{\sharp} (\rho^{\sharp}))$$

Required operators on the numeric abstraction

▶ $const^{\sharp} \in Num \rightarrow Num^{\sharp}$ computes an approximation of constants

$$\forall n \in \mathbb{Z}, \{n\} \subseteq \gamma_{\text{Num}}(\text{const}^{\sharp}(n))$$

▶ $T_{Num} \in Num^{\sharp}$ approximates any numeric value

$$\mathbb{Z} \subseteq \gamma_{Num}(\top_{Num})$$

• $o^{\sharp} \in \text{Num}^{\sharp} \times \text{Num}^{\sharp} \to \text{Num}^{\sharp}$ is a correct approximation of the arithmetic operators $o \in \{+, -, \times\}$

$$\forall n_{1}^{\sharp}, n_{2}^{\sharp} \in \text{Num}^{\sharp}, \\ \{ n_{1} \circ n_{2} \mid n_{1} \in \gamma_{\text{Num}}(n_{1}^{\sharp}), \ n_{2} \in \gamma_{\text{Num}}(n_{2}^{\sharp}) \} \subseteq \gamma_{\text{Num}}(o^{\sharp}(n_{1}^{\sharp}, n_{2}^{\sharp}))$$

$$const^{\sharp}(n) = \left\{$$

+#	1	_	+	0	-0	+0	Т
_							
+							
0							
- 0							
+0							
Т							

_#	1	_	+	0	-0	+0	Т	× [‡]	1	_	+	0	- 0	$+_0$	Т
								T							
_								_							
+								+							
0								0							
— 0								— 0							
+0								$+_0$							
T								T							

$$\operatorname{const}^{\sharp}(n) = \left\{ \begin{array}{ll} + & \text{if } n > 0 \\ 0 & \text{if } n = 0 \\ - & \text{if } n < 0 \end{array} \right.$$

+#	1	_	+	0	-0	+0	Т
_							
+							
0							
— 0							
$+_0$							
T							

_#	1	_	+	0	0	$+_0$	Т	\times^{\sharp}	Τ.	_	+	0	0	$+_0$	T
								T							
_								_							
+								+							
0								0							
— 0								-0							
+0								+0							
T								T							

$$\operatorname{const}^{\sharp}(n) = \left\{ \begin{array}{ll} + & \text{if } n > 0 \\ 0 & \text{if } n = 0 \\ - & \text{if } n < 0 \end{array} \right.$$

$+^{\sharp}$	1	_	+	0	-0	$+_0$	Т
	Т	Т	T	1	1	1	Τ
_	Τ	_	Т	_	_	Т	Т
+	1	Т	+	+	T	+	T
0	T	_	+	0	-0	+0	Т
— 0	Τ	_	Т	-0	-0	Т	Т
+0	Т	Т	+	+0	T	+0	Т
T	T	Т	Т	T	T	T	Т

_#	\perp	_	+	0	0	$+_0$	Т	\times^{\sharp}	Τ.	_	+	0	0	$+_0$	T
上															
_								_							
+								+							
0								0							
-0								0							
$+_0$								$+_0$							
T								T							

$$\operatorname{const}^{\sharp}(n) = \left\{ \begin{array}{ll} + & \text{if } n > 0 \\ 0 & \text{if } n = 0 \\ - & \text{if } n < 0 \end{array} \right.$$

+#	1	_	+	0	-0	+0	Т
上	Τ	Т	1	1	1	1	T
_	Τ	_	Т	_	_	Т	Т
+	1	Т	+	+	T	+	T
0	Τ	_	+	0	-0	+0	Т
— 0	1	_	T	-0	-0	T	Τ
+0	Т	Т	+	+0	T	+0	Т
T	T	Т	Т	T	T	T	Т

_#	\perp	_	+	0	0	+0	Т	\times^{\sharp}	Τ.	_	+	0	0	$+_0$	Т
上	\perp	Τ	Τ	1	1	1	1								
_	\perp	Т	_	_	T	_	T	_							
+	\perp	+	Т	+	+	Т	T	+							
0	\perp	+	_	0	+0	-0	Т	0							
— 0	\perp	T	_	-0	T	-0	T	-0							
+0	\perp	+	T	+0	+0	Т	T	+0							
T	\perp	Т	Т	T	T	Т	Т	T							

$$\mathrm{const}^{\sharp}(n) = \left\{ \begin{array}{ll} + & \mathrm{if}\ n > 0 \\ 0 & \mathrm{if}\ n = 0 \\ - & \mathrm{if}\ n < 0 \end{array} \right.$$

$+^{\sharp}$	1	_	+	0	-0	$+_0$	Т
	Т	Т	T	1	1	1	Τ
_	Τ	_	Т	_	_	Т	Т
+	1	Т	+	+	T	+	T
0	T	_	+	0	-0	+0	Т
— 0	Τ	_	Т	-0	-0	Т	Т
+0	Т	Т	+	+0	T	+0	Т
T	T	Т	Т	T	T	T	Т

#	1	_	+	0	- 0	$+_0$	T	\times^{\sharp}	1	_	+	0	- 0	$+_0$	T
	\perp	1	Τ	1	1	1	\perp		Т			T	1		1
_	\perp	Т	_	_	Т	_	\top	_	1	+	+	0	+0	-0	T
+	\perp	+	Т	+	+	T	T	+	T	_	+	0	-0	+0	T
0	\perp	+	_	0	+0	-0	\vdash	0	Т	0	0	0	0	0	0
— 0	\perp	Т	_	-0	Т	-0	\top	-0	1	+0	-0	0	+0	-0	T
$+_0$	T	+	Т	+0	+0	T	T	+0	T	-0	+0	0	-0	+0	T
T	\perp	Т	T	T	T	T	\vdash	T	Т	T	T	0	T	T	\top

Construction of $[t]^{\sharp}$

More difficult, because ideally such a refinement should be possible...

$$[x \mapsto +; y \mapsto -_0] \xrightarrow{\llbracket (0-y)-x>0 \rrbracket^{\sharp}} [x \mapsto +; y \mapsto -]$$

Construction of $[t]^{\sharp}$

$$\begin{bmatrix} e_1 & c & e_2 \end{bmatrix}^{\sharp} (\rho^{\sharp}) = \left(\begin{bmatrix} e_1 \end{bmatrix} \downarrow_{\text{expr}}^{\sharp} (\rho^{\sharp}, n_1^{\sharp}) \sqcap_{Env}^{\sharp} \begin{bmatrix} e_2 \end{bmatrix} \downarrow_{\text{expr}}^{\sharp} (\rho^{\sharp}, n_2^{\sharp}) \right) \\
\text{with } (n_1^{\sharp}, n_2^{\sharp}) = \begin{bmatrix} c \end{bmatrix} \downarrow_{\text{comp}}^{\sharp} \left(\mathcal{A} \llbracket e_1 \rrbracket^{\sharp} (\rho^{\sharp}), \mathcal{A} \llbracket e_2 \rrbracket^{\sharp} (\rho^{\sharp}) \right)$$

- ▶ $[\![c]\!]\downarrow_{comp}^{\sharp} \in Num^{\sharp} \times Num^{\sharp} \rightarrow Num^{\sharp} \times Num^{\sharp}$ computes a refinement of two numeric abstract values, knowing that they verify condition c
- ▶ $\llbracket e \rrbracket \downarrow_{\exp r}^{\sharp} \in Env^{\sharp} \times \operatorname{Num}^{\sharp} \to Env^{\sharp} : \llbracket e \rrbracket \downarrow_{\exp r}^{\sharp} (\rho^{\sharp}, n^{\sharp})$ computes a refinement of the abstract environment ρ^{\sharp} , knowing that the expression e evaluates into a value that is approximated by n^{\sharp} in this environment.

$$\llbracket = \rrbracket \downarrow^{\sharp} (x^{\sharp}, y^{\sharp}) =$$

[[≠]]↓ [♯]	1	_	+	0	0	+0	Т
_							
+							
0							
-0							
$+_0$							
T							

$$\llbracket = \rrbracket \downarrow^{\sharp} (x^{\sharp}, y^{\sharp}) = (x^{\sharp} \sqcap^{\sharp} y^{\sharp}, x^{\sharp} \sqcap^{\sharp} y^{\sharp})$$

[[≠]]↓ [♯]	Τ.	_	+	0	_ ₀	$+_0$	Т
_							
+							
0							
-0							
$+_0$							
T							

$$\llbracket = \rrbracket \downarrow^{\sharp} (x^{\sharp}, y^{\sharp}) = (x^{\sharp} \sqcap^{\sharp} y^{\sharp}, x^{\sharp} \sqcap^{\sharp} y^{\sharp})$$

[[≠]]↓ [♯]		_	+	0	0	+0	Т
上	(\perp, \perp)	(\bot,\bot)	(\bot, \bot)	(\bot, \bot)	(\bot, \bot)	(\bot,\bot)	(\bot,\bot)
_	(\bot, \bot)	(-,-)	(-,+)	(-,0)	(-,0)	$(-,+_0)$	(−,⊤)
+	(\bot, \bot)	(+,-)	(+,+)	(+,0)	(+,0)	$(+,+_0)$	(+,⊤)
0	(\bot, \bot)	(0, -)	(0, +)	(\bot, \bot)	(0, -)	(0,+)	(0, ⊤)
-0	(\bot, \bot)	(-0,-)	(-0,+)	(-,0)	(0,0)	(-0,+0)	$(-0,\top)$
$+_{0}$	(\perp, \perp)	$(+_0, -)$	$(+_0,+)$	(+,0)	$(+_0,0)$	$(+_0, +_0)$	$(+_0, \top)$
T	(\bot, \bot)	(⊤,−)	(⊤,+)	$(\top,0)$	$(\top,0)$	$(\top, +_0)$	(\top, \top)

$[\![<]\!]\downarrow^\sharp$	Τ.	_	+	0	-0	$+_0$	Т
\perp							
_							
+							
0							
-0							
$+_0$						-	
T							

[[≤]]↓ [♯]	 _	+	0	-0	$+_0$	Т
_						
+						
0						
-0						
$+_0$						
T						

$\llbracket < \rrbracket \downarrow^\sharp$	1	_	+	0	-0	+0	Т
	(\bot, \bot)	(\bot,\bot)	(\perp, \perp)	(\bot,\bot)	(\bot,\bot)	(\bot, \bot)	(\bot,\bot)
_	(\bot, \bot)	(-,-)	(-,+)	(-,0)	(-,0)	$(-,+_0)$	(−,⊤)
+	(\bot, \bot)	(\bot,\bot)	(+,+)	(\bot, \bot)	(\bot,\bot)	(+,+)	(+,+)
0	(\bot, \bot)	(\bot,\bot)	(0, +)	(\bot,\bot)	(\bot,\bot)	(0, +)	(0, +)
-0	(\bot, \bot)	(-0,-)	(-0,+)	(-0,0)	(0,0)	(-0,+0)	$(0,\top)$
+0	(\bot, \bot)	(\bot,\bot)	$(+_0,+)$	(\bot, \bot)	(\bot,\bot)	$(+_0, +_0)$	$(+_0,+)$
T	(\bot, \bot)	(-,-)	(T,+)	(-,0)	(-,0)	$(\top, +_0)$	(\top, \top)

$\llbracket \leqslant \rrbracket \downarrow^\sharp$	 _	+	0	-0	+0	Т
_						
+						
0						
-0						
+0						
T						

$\llbracket < \rrbracket \downarrow^\sharp$		_	+	0	-0	+0	Т
	(\bot, \bot)	(\bot,\bot)	(\bot, \bot)	(\bot, \bot)	(\bot,\bot)	(\bot, \bot)	(\bot, \bot)
					(-,0)		
+	(\perp, \perp)	(\bot,\bot)	(+,+)	(\bot, \bot)	(\bot,\bot)	(+,+)	(+,+)
0	(\perp, \perp)	(\bot,\bot)	(0,+)	(\bot, \bot)	(\bot,\bot)	(0, +)	(0, +)
-0	(\perp, \perp)	(-0,-)	(-0,+)	(-0,0)	(0,0)	(-0,+0)	$(0,\top)$
+0	(\perp, \perp)	(\bot,\bot)	$(+_0,+)$	(\bot, \bot)	(\bot,\bot)	$(+_0, +_0)$	$(+_0,+)$
T	(\perp, \perp)	(-,-)	(⊤,+)	(-,0)	(-,0)	$(\top, +_0)$	(\top, \top)

$\llbracket \leqslant \rrbracket \downarrow^\sharp$		_	+	0	-0	+0	Т
	(\bot, \bot)	(\bot,\bot)	(\bot,\bot)	(\bot, \bot)	(\bot,\bot)	(\bot,\bot)	(\bot, \bot)
_	(\bot, \bot)	(-,-)	(-,+)	(-,0)	(-,0)	$(-,+_0)$	$(-,\top)$
+	(\perp, \perp)	(\bot,\bot)	(+,+)	(\bot, \bot)	(\bot, \bot)	(+,+)	(+,+)
0	(\bot, \bot)	(\bot,\bot)	(0,+)	(\bot, \bot)	(\bot, \bot)	$(0,+_0)$	$(0,+_0)$
-0	(\bot, \bot)	(-0, -)	(-0,+)	(-0,0)	(0,0)	(-0,+0)	$(0,\top)$
+0	(\bot, \bot)	(\bot,\bot)	$(+_0,+)$	(0,0)	(0,0)	$(+_0,+_0)$	$(+_0,+_0)$
T	(\bot, \bot)	(-,-)	(⊤,+)	(-0,0)	(0,0)	$(\top, +_0)$	(\top, \top)

Required operators on the numeric abstraction

$$\left\{ (n_1, n_2) \mid n_1 \in \gamma_{\text{Num}}(n_1^{\sharp}), n_2 \in \gamma_{\text{Num}}(n_2^{\sharp}), n_1 c n_2 \right\} \\
\subseteq \gamma_{\text{Num}}(m_1^{\sharp}) \times \gamma_{\text{Num}}(m_2^{\sharp}) \\
\text{with } (m_1^{\sharp}, m_2^{\sharp}) = \llbracket c \rrbracket \downarrow_{\text{comp}}^{\sharp} (n_1^{\sharp}, n_2^{\sharp})$$

Required operators on the numeric abstraction

$$\llbracket \sigma \rrbracket \! \! \downarrow_{op}^\sharp \in Num^\sharp \times Num^\sharp \times Num^\sharp \to Num^\sharp \times Num^\sharp$$

 $[\![o]\!]\downarrow_{\mathrm{op}}^{\sharp}(n^{\sharp},n_{1}^{\sharp},n_{2}^{\sharp})$ computes a refinement of two numeric values n_{1}^{\sharp} and n_{2}^{\sharp} knowing that the result of the binary operation o is approximated by n^{\sharp} on their concretisations.

$$\forall n^{\sharp}, n_{1}^{\sharp}, n_{2}^{\sharp} \in \text{Num}^{\sharp},$$

$$\left\{ \begin{array}{l} (n_{1}, n_{2}) \mid n_{1} \in \gamma_{\text{Num}}(n_{1}^{\sharp}), \ n_{2} \in \gamma_{\text{Num}}(n_{2}^{\sharp}), \ (n_{1} \circ n_{2}) \in \gamma_{\text{Num}}(n^{\sharp}) \end{array} \right.$$

$$\subseteq \gamma_{\text{Num}}(m_{1}^{\sharp}) \times \gamma_{\text{Num}}(m_{2}^{\sharp})$$

$$\text{with } (m_{1}^{\sharp}, m_{2}^{\sharp}) = \llbracket o \rrbracket \downarrow_{\text{op}}^{\sharp} (n^{\sharp}, n_{1}^{\sharp}, n_{2}^{\sharp})$$

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+						
0						
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$+_0$						
T						

. . .

$\llbracket \times \rrbracket \downarrow^{\sharp} (0, \cdot, \cdot)$	 _	+	0	0	+0	Т
_						
+						
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$+_0$						
T						

$\llbracket + \rrbracket \downarrow^{\sharp} (+, \cdot, \cdot)$		_	+	0	- 0	+0	Т
	(\bot, \bot)	(\bot,\bot)	(\bot,\bot)	(\bot,\bot)	(\bot,\bot)	(\bot, \bot)	(\bot,\bot)
_	(\bot,\bot)	(\bot,\bot)	(-,+)	(\bot,\bot)	(\bot,\bot)	(-,+)	(-,+)
+	(\bot,\bot)	(+,-)	(+,+)	(+,0)	(+,0)	$(+,+_0)$	(+,⊤)
0	(\bot, \bot)	(\bot,\bot)	(0,+)	(\bot,\bot)	(\bot,\bot)	(0,+)	(0,+)
-0	(\bot,\bot)	(\bot,\bot)	(-0,+)	(\bot,\bot)	(\bot,\bot)	(-0,+)	(-0,+)
+0	(\bot,\bot)	(+,-)	$(+_0,+)$	(+,0)	(+,0)	$(+_0, +_0)$	$(+_0,\top)$
T	(\bot,\bot)	(+,-)	(T,+)	(+,0)	(+,0)	$(\top, +_0)$	(T,T)

. . .

$\llbracket \times \rrbracket \downarrow^{\sharp} (0, \cdot, \cdot)$	1	_	+	0	0	+0	Т
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+							
0							
- 0							
$+_0$							
Т							

[[+]]↓ [♯] (+,·,·)		_	+	0	0	+0	Т
	(\bot,\bot)	(\bot,\bot)	(\bot, \bot)	(\bot,\bot)	(\bot, \bot)	(\bot, \bot)	(\bot,\bot)
_	(\bot,\bot)	(\bot,\bot)	(-,+)	(\bot,\bot)	(\bot, \bot)	(-,+)	(-,+)
+	(\bot,\bot)	(+,-)	(+,+)	(+,0)	(+,0)	$(+,+_0)$	(+,⊤)
0	(\bot,\bot)	(\bot,\bot)	(0,+)	(\bot,\bot)	(\bot,\bot)	(0, +)	(0,+)
- 0	(\bot,\bot)	(\bot,\bot)	(0,+)	(\bot,\bot)	(\bot, \bot)	(0,+)	(-0,+)
$+_0$	(\bot,\bot)	(+,-)	$(+_0,+)$	(+,0)	(+,0)	$(+_0, +_0)$	$(+_0, \top)$
Т	(\bot,\bot)	(+,-)	$(\top,+)$	(+,0)	(+,0)	$(\top, +_0)$	(\top,\top)

. . .

$\llbracket \times \rrbracket \downarrow^{\sharp} (0, \cdot, \cdot)$		_	+	0	0	+0	Т
	(\bot,\bot)						
_	(\bot,\bot)	(\bot,\bot)	(\bot,\bot)	(-,0)	(-,0)	(-,0)	(-,0)
+	(\bot,\bot)	(\bot,\bot)	(\bot,\bot)	(+,0)	(+,0)	(+,0)	(+,0)
0	(\bot,\bot)	(0,)	(0,+)	(0,0)	(0,0)	$(0,+_0)$	(0, ⊤)
-0	(\bot,\bot)	(0, -)	(0,+)	(-0,0)	(-0,-0)	(-0,+0)	$(0, \top)$
$+_0$	(\bot,\bot)	(0, -)	(0,+)	$(+_0,0)$	$(+_0,0)$	$(+_0,+_0)$	$(+_0, \top)$
Т	(\bot,\bot)	(0, -)	(0,+)	(⊤,0)	(⊤,0)	(⊤,0)	(\top, \top)

Ocaml code...

```
module type NumAbstraction =
sia
 module L : Lattice
 val backTest : comp -> L.t -> L.t * L.t
 val semOp : op -> L.t -> L.t -> L.t
 val back_semOp : op -> L.t -> L.t -> L.t * L.t
  val const : int -> L.t
 val top : L.t
 val to_string : string -> L.t -> string
end
module EnvNotRelational = functor (AN:NumAbstraction) ->
 (struct ... end : EnvAbstraction)
```

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- Introduction
- Intermediate representation : syntax and semantics
- Collecting semantics
- Just put some [‡]...
- Building a generic abstract interpreter
- 6 Numeric abstraction by intervals
- Widening/Narrowing
- Polyhedral abstract interpretation
- Readings

Abstraction by intervals

Int
$$\stackrel{\text{def}}{=} \{ [a, b] \mid a, b \in \overline{\mathbb{Z}}, a \leqslant b \} \cup \{\bot\} \text{ with } \overline{\mathbb{Z}} = \mathbb{Z} \cup \{-\infty, +\infty\}$$

Lattice:

$$\begin{array}{c|c} I \in \operatorname{Int} & c \leqslant a & b \leqslant d & a,b,c,d \in \overline{\mathbb{Z}} \\ \hline \bot \sqsubseteq_{\operatorname{Int}} I & & [a,b] \sqsubseteq_{\operatorname{Int}} [c,d] \\ \\ I \sqcup_{\operatorname{Int}} \bot & \stackrel{\operatorname{def}}{=} & I, \ \forall I \in \operatorname{Int} \\ \bot \sqcup_{\operatorname{Int}} I & \stackrel{\operatorname{def}}{=} & I, \ \forall I \in \operatorname{Int} \\ [a,b] \sqcup_{\operatorname{Int}} [c,d] & \stackrel{\operatorname{def}}{=} & [\min(a,c),\max(b,d)] \\ \\ I \sqcap_{\operatorname{Int}} \bot & \stackrel{\operatorname{def}}{=} & \bot, \ \forall I \in \operatorname{Int} \\ \bot \sqcap_{\operatorname{Int}} I & \stackrel{\operatorname{def}}{=} & \bot, \ \forall I \in \operatorname{Int} \\ [a,b] \sqcap_{\operatorname{Int}} [c,d] & \stackrel{\operatorname{def}}{=} & \bot, \ \forall I \in \operatorname{Int} \\ [a,b] \sqcap_{\operatorname{Int}} [c,d] & \stackrel{\operatorname{def}}{=} & \bot, \ \forall I \in \operatorname{Int} \\ \end{array}$$

with $\rho_{Int} \in (\overline{\mathbb{Z}} \times \overline{\mathbb{Z}}) \to Int$ defined by

$$\rho_{\text{Int}}(a,b) = \begin{cases} [a,b] & \text{if } a \leq b, \\ \bot & \text{otherwise} \end{cases}$$

$$\begin{array}{ccc} \bot_{Int} & \stackrel{\scriptscriptstyle def}{=} & \bot \\ \top_{Int} & \stackrel{\scriptscriptstyle def}{=} & [-\infty, +\infty] \end{array}$$

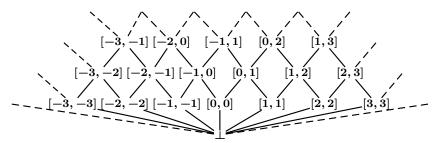
$$\begin{array}{ccc} \gamma_{\mathrm{Int}}(\bot) & \stackrel{\mathrm{def}}{=} & \emptyset \\ \gamma_{\mathrm{Int}}([a,b]) & \stackrel{\mathrm{def}}{=} & \{\, z \in \mathbb{Z} \, | \, a \leqslant z \text{ and } z \leqslant b \, \} \end{array}$$

All the other operators are *stricts*: they return \bot if one of their arguments is \bot .

$$+^{\sharp} ([a,b],[c,d]) = [a+c,b+d]$$
 $-^{\sharp} ([a,b],[c,d]) = [a-d,b-c]$
 $\times^{\sharp} ([a,b],[c,d]) = [\min(ac,ad,bc,bd),\max(ac,ad,bc,bd)]$



Convergence problem



Such a lattice does not satisfy the ascending chain condition.

Example of infinite increasing chain:

$$\perp \sqsubset [0,0] \sqsubset [0,1] \sqsubset \cdots \sqsubset [0,n] \sqsubset \cdots$$

Solution: dynamic approximation

• we extrapolate the limit thanks to a widening operator ∇

$$\bot \sqsubset [0,0] \sqsubset [0,1] \sqsubset [0,2] \sqsubset [0,+\infty] = [0,2] \nabla [0,3]$$

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Fixpoint approximation

Lemma

Let $(A, \sqsubseteq, \sqcup, \sqcap)$ a complete lattice and f a monotone operator on A. If a is a post-fixpoint of f (i.e. $f(a) \sqsubseteq a$), then $lfp(f) \sqsubseteq a$.

We may want to compute an over-approximation of lfp(f) in the following cases :

- ▶ The lattice does not satisfies the ascending chain condition, the iteration $\bot, f(\bot), \ldots, f^n(\bot), \ldots$ may never terminates.
- ► The ascending chain condition is satisfied but the iteration chain is too long to allow an efficient computation.
- ▶ Id the underlying lattice is not complete, the limits of the ascending iterations do not necessarily belongs to the abstraction domain.

Widening

Idea: the standard iteration is of the form

$$x^{0} = \bot, x^{n+1} = F(x^{n}) = x^{n} \sqcup F(x^{n})$$

We will replace it by something of the form

$$y^0 = \bot, y^{n+1} = y^n \nabla F(y^n)$$

such that

- (i) (y^n) is increasing,
- (ii) $x^n \sqsubseteq y^n$, for all n,
- (iii) and (y^n) stabilizes after a finite number of steps.

But we also want a ∇ operator that is independent of F.

Widening: definition

A widening is an operator $\nabla : L \times L \to L$ such that

- ▶ $\forall x, x' \in L, x \sqcup x' \sqsubseteq x \nabla x'$ (implies (i) & (ii))
- ▶ If $x^0 \sqsubseteq x^1 \sqsubseteq \dots$ is an increasing chain, then the increasing chain $y^0 = x^0, y^{n+1} = y^n \nabla x^{n+1}$ stabilizes after a finite number of steps (implies (iii)).

Usage : we replace
$$x^0 = \bot, x^{n+1} = F(x^n)$$

by $y^0 = \bot, y^{n+1} = y^n \nabla F(y^n)$



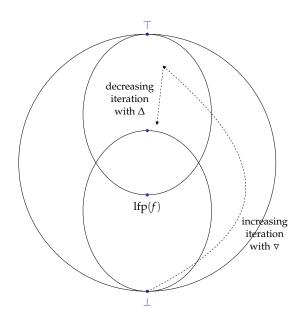
Widening: theorem

Theorem

Let L a complete lattice, $F: L \to L$ a monotone function and $\nabla: L \times L \to L$ a widening operator. The chain $y^0 = \bot, y^{n+1} = y^n \nabla F(y^n)$ stabilizes after a finite number of steps towards a post-fixpoint y of F.

Corollary : $lfp(F) \sqsubseteq y$.

Scheme



Example: widening on intervals

Idea : as soon as a bound is not stable, we extrapolate it by $+\infty$ (or $-\infty$). After such an extrapolation, the bound can't move any more.

Definition:

$$\begin{array}{lll} [a,b] \nabla_{\mathrm{Int}} [a',b'] & = & [& \mathrm{if} \ a' < a \ \mathrm{then} \ -\infty \ \mathrm{else} \ a, \\ & & \mathrm{if} \ b' > b \ \mathrm{then} \ +\infty \ \mathrm{else} \ b \,] \\ \bot \nabla_{\mathrm{Int}} [a',b'] & = & [a',b'] \\ I \ \nabla_{\mathrm{Int}} \ \bot & = & I \\ \end{array}$$

Examples:

$$[-3,4]\nabla_{\text{Int}}[-3,2] = [-3,4]$$

 $[-3,4]\nabla_{\text{Int}}[-3,5] = [-3,+\infty]$



Example

$$x := 100;$$
while $0 < x$ {
 $x := x - 1;$
}
$$0$$
 $x := 100$

$$0 < x$$

$$0 > x$$

$$2$$

$$3$$

$$X_1 = [100, 100] \sqcup_{\text{Int}} (X_2 - ^{\sharp} [1, 1])$$
 $X_2 = [1, +\infty] \sqcap_{\text{Int}} X_1$
 $X_3 = [-\infty, 0] \sqcap_{\text{Int}} X_1$

Example: without widening

$$X_1 = [100, 100] \sqcup_{\text{Int}} (X_2 - ^{\sharp} [1, 1])$$

 $X_2 = [1, +\infty] \sqcap_{\text{Int}} X_1$
 $X_3 = [-\infty, 0] \sqcap_{\text{Int}} X_1$

$$\begin{array}{ll} X_1^0 = \bot & X_1^{n+1} = [100,100] \sqcup_{\mathrm{Int}} \left(X_2^n - ^{\sharp} [1,1] \right) \\ X_2^0 = \bot & X_2^{n+1} = [1,+\infty] \sqcap_{\mathrm{Int}} X_1^{n+1} \\ X_3^0 = \bot & X_3^{n+1} = [-\infty,0] \sqcap_{\mathrm{Int}} X_1^{n+1} \end{array}$$

Example: without widening

$$X_1 = [100, 100] \sqcup_{\text{Int}} (X_2 - ^{\sharp} [1, 1])$$

 $X_2 = [1, +\infty] \sqcap_{\text{Int}} X_1$
 $X_3 = [-\infty, 0] \sqcap_{\text{Int}} X_1$

$$\begin{array}{ll} X_1^0 = \bot & X_1^{n+1} = [100,100] \sqcup_{\mathrm{Int}} \left(X_2^n - ^{\sharp} [1,1] \right) \\ X_2^0 = \bot & X_2^{n+1} = [1,+\infty] \sqcap_{\mathrm{Int}} X_1^{n+1} \\ X_3^0 = \bot & X_3^{n+1} = [-\infty,0] \sqcap_{\mathrm{Int}} X_1^{n+1} \end{array}$$

	X_1	1	[100, 100]	[99, 100]	[98, 100]	[97, 100]	 [1, 100]	[0, 100]
İ	X_2	_	[100, 100]	[99, 100]	[98, 100]	[97, 100]	 [1, 100]	[1, 100]
ı	X_3	_	\perp	\perp	\perp	\perp	 \perp	[0, 0]

Example: with widening at each nodes of the cfg

$$X_1 = [100, 100] \sqcup_{\text{Int}} (X_2 - ^{\sharp} [1, 1])$$

 $X_2 = [1, +\infty] \sqcap_{\text{Int}} X_1$
 $X_3 = [-\infty, 0] \sqcap_{\text{Int}} X_1$

$$\begin{array}{ll} X_1^0 = \bot & X_1^{n+1} = X_1^n \triangledown_{\mathrm{Int}} \left([100, 100] \sqcup_{\mathrm{Int}} \left(X_2^n - ^{\sharp} [1, 1] \right) \right) \\ X_2^0 = \bot & X_2^{n+1} = X_2^n \triangledown_{\mathrm{Int}} \left([1, + \infty] \sqcap_{\mathrm{Int}} X_1^{n+1} \right) \\ X_3^0 = \bot & X_3^{n+1} = X_3^n \triangledown_{\mathrm{Int}} \left([-\infty, 0] \sqcap_{\mathrm{Int}} X_1^{n+1} \right) \end{array}$$

$$egin{array}{|c|c|c|c|}\hline X_1 & \bot & & & \\ X_2 & \bot & & & \\ X_3 & \bot & & & \\ \hline \end{array}$$

Example: with widening at each nodes of the cfg

$$X_1 = [100, 100] \sqcup_{\text{Int}} (X_2 - [1, 1])$$

 $X_2 = [1, +\infty] \sqcap_{\text{Int}} X_1$
 $X_3 = [-\infty, 0] \sqcap_{\text{Int}} X_1$

$$\begin{array}{ll} X_1^0 = \bot & X_1^{n+1} = X_1^n \triangledown_{\mathrm{Int}} \left([100, 100] \sqcup_{\mathrm{Int}} \left(X_2^n - ^{\sharp} [1, 1] \right) \right) \\ X_2^0 = \bot & X_2^{n+1} = X_2^n \triangledown_{\mathrm{Int}} \left([1, + \infty] \sqcap_{\mathrm{Int}} X_1^{n+1} \right) \\ X_3^0 = \bot & X_3^{n+1} = X_3^n \triangledown_{\mathrm{Int}} \left([-\infty, 0] \sqcap_{\mathrm{Int}} X_1^{n+1} \right) \end{array}$$

Improving fixpoint approximation

Idea: iterating a little more may help...

Theorem

Let $(A, \sqsubseteq, \sqcup, \sqcap)$ a complete lattice, f a monotone operator on A and a a post-fixpoint of f. The chain $(x_n)_n$ defined by $\begin{cases} x_0 = a \\ x_{k+1} = f(x_k) \end{cases}$ admits for limit $(\sqcup \{x_n\})$ the greatest fixpoint of f lower than a (written $gfp_a(f)$). In particular, $lfp(f) \sqsubseteq \sqcup \{x_n\}$. Each intermediate step is a correct approximation:

$$\forall k$$
, $lfp(f) \sqsubseteq gfp_a(f) \sqsubseteq x_k \sqsubseteq a$

Narrowing: definition

A *narrowing* is an operator $\Delta : L \times L \rightarrow L$ such that

- $\forall x, x' \in L, x' \sqsubseteq x \Delta x' \sqsubseteq x$
- ► If $x^0 \supseteq x^1 \supseteq \dots$ is a decreasing chain, then the increasing chain $y^0 = x^0, y^{n+1} = y^n \Delta x^{n+1}$ stabilizes after a finite number of steps.



Narrowing: decreasing iteration

Theorem

If Δ is a narrowing operator on a poset (A, \sqsubseteq) , if f is a monotone operator on A and a is a post-fixpoint of f then the chain $(x_n)_n$ defined by $\begin{cases} x_0 &= a \\ x_{k+1} &= x_k \Delta f(x_k) \end{cases}$ stabilizes after a finite number of steps on a post-fixpoint of f lower than a.



Narrowing on intervals

```
[a,b]\Delta_{\mathrm{Int}}[c,d]=[\mathrm{if}\ a=-\infty\ \mathrm{then}\ c\ \mathrm{else}\ a\ ;\ \mathrm{if}\ b=+\infty\ \mathrm{then}\ d\ \mathrm{else}\ b]
I\ \Delta_{\mathrm{Int}}\ \bot\ =\ \bot
\bot\ \Delta_{\mathrm{Int}}\ I\ =\ \bot
```

Intuition: we only improve infinite bounds.

In practice: a few standard iterations already improve a lot the result that has been obtained after widening...

▶ Assignments by constants and conditional guards make the decreasing iterations efficient: they *filter* the (too big) approximations computed by the widening

Example: with narrowing at each nodes of the cfg

$$X_1 = [100, 100] \sqcup_{\text{Int}} (X_2 - ^{\sharp} [1, 1])$$

 $X_2 = [1, +\infty] \sqcap_{\text{Int}} X_1$
 $X_3 = [-\infty, 0] \sqcap_{\text{Int}} X_1$

$$\begin{array}{ll} X_1^0 = [-\infty, 100] & X_1^{n+1} = X_1^n \Delta_{\mathrm{Int}} \left([100, 100] \sqcup_{\mathrm{Int}} \left(X_2^n - ^{\sharp} [1, 1] \right) \right) \\ X_2^0 = [-\infty, 100] & X_2^{n+1} = X_2^n \Delta_{\mathrm{Int}} \left([1, +\infty] \sqcap_{\mathrm{Int}} X_1^{n+1} \right) \\ X_3^0 = [-\infty, 0] & X_3^{n+1} = X_3^n \Delta_{\mathrm{Int}} \left([-\infty, 0] \sqcap_{\mathrm{Int}} X_1^{n+1} \right) \end{array}$$

$$egin{array}{|c|c|c|c|} \hline X_1 & [-\infty, 100] \\ X_2 & [-\infty, 100] \\ X_3 & [-\infty, 0] \\ \hline \end{array}$$



Example: with narrowing at each nodes of the cfg

$$X_1 = [100, 100] \sqcup_{\text{Int}} (X_2 - ^{\sharp} [1, 1])$$

 $X_2 = [1, +\infty] \sqcap_{\text{Int}} X_1$
 $X_3 = [-\infty, 0] \sqcap_{\text{Int}} X_1$

$$\begin{array}{ll} X_1^0 = [-\infty, 100] & X_1^{n+1} = X_1^n \Delta_{\mathrm{Int}} \left([100, 100] \sqcup_{\mathrm{Int}} \left(X_2^n - ^{\sharp} [1, 1] \right) \right) \\ X_2^0 = [-\infty, 100] & X_2^{n+1} = X_2^n \Delta_{\mathrm{Int}} \left([1, +\infty] \sqcap_{\mathrm{Int}} X_1^{n+1} \right) \\ X_3^0 = [-\infty, 0] & X_3^{n+1} = X_3^n \Delta_{\mathrm{Int}} \left([-\infty, 0] \sqcap_{\mathrm{Int}} X_1^{n+1} \right) \end{array}$$

The particular case of an equation system

Consider a system

$$\begin{cases} x_1 &= f_1(x_1, \dots, x_n) \\ \vdots \\ x_n &= f_n(x_1, \dots, x_n) \end{cases}$$

with f_1, \ldots, f_n monotones.

Standard iteration:

$$\begin{aligned}
 x_1^{i+1} &= f_1(x_1^i, \dots, x_n^i) \\
 x_2^{i+1} &= f_2(x_1^i, \dots, x_n^i) \\
 &\vdots \\
 x_n^{i+1} &= f_n(x_1^i, \dots, x_n^i)
 \end{aligned}$$

Standard iteration with widening:

$$\begin{array}{rcl} x_1^{i+1} & = & x_1^i \nabla f_1(x_1^i, \dots, x_n^i) \\ x_2^{i+1} & = & x_2^i \nabla f_2(x_1^i, \dots, x_n^i) \\ & \vdots \\ x_n^{i+1} & = & x_n^i \nabla f_n(x_1^i, \dots, x_n^i) \end{array}$$



The particular case of an equation system

$$\begin{cases} x_1 &= f_1(x_1, \dots, x_n) \\ \vdots \\ x_n &= f_n(x_1, \dots, x_n) \end{cases}$$

It is sufficient (and generally more precise) to use \triangledown for a selection of index W such that each dependence cycle in the system goes through at least one point in W.

$$\forall k = 1..n, \ x_k^{i+1} = \ x_k^i \nabla f_k(x_1^i, \dots, x_n^i) \quad \text{if } k \in W$$
$$f_k(x_1^i, \dots, x_n^i) \quad \text{otherwise}$$

Chaotic iteration: at each step, we use only one equation, without forgetting one for ever.



Contrary, to what happen in a standard dataflow framework (with monotone functions and ascending chain condition), the iteration strategy may affect a lot the precision of the result. See F. Bourdoncle, *Efficient Chaotic Iteration Strategies with Widenings*, 1993.

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Polyhedral abstract interpretation

Automatic discovery of linear restraints among variables of a program. P. Cousot and N. Halbwachs. POPL'78.



Patrick Cousot



Nicolas Halbwachs

Polyhedral analysis seeks to discover invariant linear equality and inequality relationships among the variables of an imperative program.

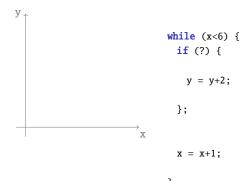
Convex polyhedra

A convex polyhedron can be defined algebraically as the set of solutions of a system of linear inequalities.

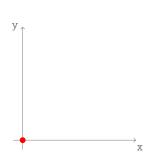
Geometrically, it can be defined as a finite intersection of half-spaces.



$$x = 0; y = 0;$$



State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .



```
{x = 0 ∧ y = 0}

while (x<6) {

if (?) {

{x = 0 ∧ y = 0}

y = y+2;

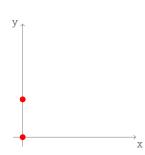
};
```

x = 0; y = 0;

$$x = x+1;$$

4 🗗 ►

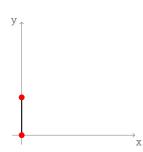
State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .



At junction points, we over-approximates union by a convex union.

```
x = 0; y = 0;
         \{x = 0 \land y = 0\}
while (x<6) {</pre>
  if (?) {
         \{ \mathbf{x} = 0 \land \mathbf{v} = 0 \}
     y = y+2;
         \{x = 0 \land v = 2\}
  };
         \{x = 0 \land y = 0\} \uplus \{x = 0 \land y = 2\}
  x = x+1;
}
```

State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .

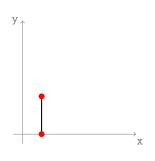


At junction points, we over-approximates union by a convex union.

```
x = 0; y = 0;
          \{x = 0 \land y = 0\}
while (x<6) {</pre>
  if (?) {
          \{ \mathbf{x} = 0 \land \mathbf{v} = 0 \}
     y = y+2;
         \{x = 0 \land y = 2\}
  };
          \{\mathbf{x} = 0 \land 0 \leq \mathbf{y} \leq 2\}
  x = x+1;
}
```

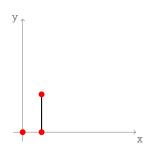
State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .

x = 0; y = 0;



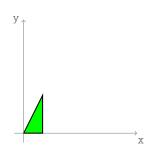
State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .

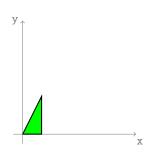
x = 0; y = 0;



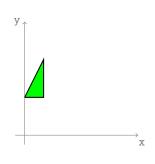
State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .

x = 0; y = 0;

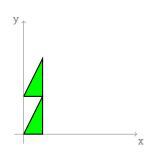




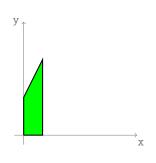
```
x = 0; y = 0;
           \{x \leqslant 1 \land 0 \leqslant y \leqslant 2x\}
while (x<6) {
  if (?) {
           \{x \le 1 \land 0 \le y \le 2x\}
      y = y+2;
          \{x = 0 \land y = 2\}
  };
          \{\mathbf{x} = 0 \land 0 \leq \mathbf{y} \leq 2\}
   x = x+1;
          \{\mathbf{x} = 1 \land 0 \leqslant \mathbf{y} \leqslant 2\}
```



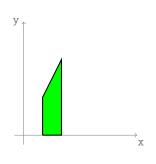
```
x = 0; y = 0;
          \{x \le 1 \land 0 \le y \le 2x\}
while (x<6) {
  if (?) {
          \{x \le 1 \land 0 \le y \le 2x\}
      y = y+2;
          \{x \le 1 \land 2 \le y \le 2x + 2\}
  };
          \{\mathbf{x} = 0 \land 0 \leqslant \mathbf{y} \leqslant 2\}
   x = x+1;
          \{\mathbf{x} = 1 \land 0 \leqslant \mathbf{y} \leqslant 2\}
```



```
x = 0; y = 0;
           \{x \le 1 \land 0 \le y \le 2x\}
while (x<6) {</pre>
   if (?) {
           \{x \le 1 \land 0 \le y \le 2x\}
      y = y+2;
           \{x \leqslant 1 \land 2 \leqslant y \leqslant 2x + 2\}
   };
           \{x \leqslant 1 \land 0 \leqslant y \leqslant 2x\}
                                    \oplus \{x \leq 1 \land 2 \leq y \leq 2x + 2\}
   x = x+1;
           \{\mathbf{x} = 1 \land 0 \leqslant \mathbf{y} \leqslant 2\}
```

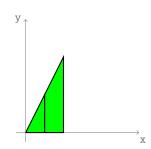


```
x = 0; y = 0;
           \{x \le 1 \land 0 \le y \le 2x\}
while (x<6) {</pre>
   if (?) {
            \{x \le 1 \land 0 \le y \le 2x\}
      y = y+2;
           \{x \leqslant 1 \land 2 \leqslant y \leqslant 2x + 2\}
   };
           \{0 \le \mathbf{x} \le 1 \land 0 \le \mathbf{y} \le 2\mathbf{x} + 2\}
   x = x+1;
           \{\mathbf{x} = 1 \land 0 \leqslant \mathbf{y} \leqslant 2\}
```



```
x = 0; y = 0;
            \{x \le 1 \land 0 \le y \le 2x\}
while (x<6) {</pre>
   if (?) {
            \{x \le 1 \land 0 \le y \le 2x\}
      y = y+2;
            \{x \leqslant 1 \land 2 \leqslant y \leqslant 2x + 2\}
   };
            \{0 \le \mathbf{x} \le 1 \land 0 \le \mathbf{y} \le 2\mathbf{x} + 2\}
   x = x+1;
            \{1 \leqslant \mathbf{x} \leqslant 2 \land 0 \leqslant \mathbf{y} \leqslant 2\mathbf{x}\}\
}
```

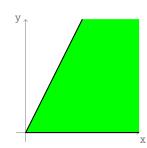
State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .



At loop headers, we use heuristics (widening) to ensure finite convergence.

```
x = 0; y = 0;
           \{x \le 1 \land 0 \le y \le 2x\}
                                     \nabla \{x \leq 2 \land 0 \leq y \leq 2x\}
while (x<6) {
   if (?) {
            \{x \le 1 \land 0 \le y \le 2x\}
      v = y+2;
           \{x \le 1 \land 2 \le y \le 2x + 2\}
   };
            \{0 \le \mathbf{x} \le 1 \land 0 \le \mathbf{y} \le 2\mathbf{x} + 2\}
   x = x+1;
           \{1 \leqslant \mathbf{x} \leqslant 2 \land 0 \leqslant \mathbf{y} \leqslant 2\mathbf{x}\}\
```

State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .



At loop headers, we use heuristics (widening) to ensure finite convergence.

```
x = 0; y = 0;
           \{0 \le y \le 2x\}
while (x<6) {</pre>
   if (?) {
            \{x \le 1 \land 0 \le y \le 2x\}
      y = y+2;
           \{x \le 1 \land 2 \le y \le 2x + 2\}
   };
            \{0 \le \mathbf{x} \le 1 \land 0 \le \mathbf{y} \le 2\mathbf{x} + 2\}
   x = x+1;
           \{1 \leqslant \mathbf{x} \leqslant 2 \land 0 \leqslant \mathbf{y} \leqslant 2\mathbf{x}\}\
```

State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .

```
x = 0; y = 0;
          \{0 \le y \le 2x\}
while (x<6) {
  if (?) {
          \{0 \le y \le 2x \land x \le 5\}
     y = y+2;
          \{2 \leqslant y \leqslant 2x + 2 \land x \leqslant 5\}
  };
          \{0 \le \mathbf{v} \le 2\mathbf{x} + 2 \land 0 \le x \le 5\}
   x = x+1;
          \{0 \le y \le 2x \land 1 \le x \le 6\}
          \{0 \le \mathbf{v} \le 2\mathbf{x} \land 6 \le x\}
```

By propagation we obtain a post-fixpoint

State properties are over-approximated by convex polyhedra in \mathbb{Q}^2 .

By propagation we obtain a post-fixpoint which is enhanced by downward iteration.

```
x = 0; y = 0;
           \{0 \le y \le 2x \land x \le 6\}
while (x<6) {
   if (?) {
           \{0 \le \mathbf{v} \le 2\mathbf{x} \land x \le 5\}
      y = y+2;
           \{2 \le \mathbf{v} \le 2\mathbf{x} + 2 \land x \le 5\}
   };
           \{0 \le \mathbf{v} \le 2\mathbf{x} + 2 \land 0 \le x \le 5\}
   x = x+1:
           \{0 \le y \le 2x \land 1 \le x \le 6\}
           \{0 \le \mathbf{v} \le 2\mathbf{x} \land 6 = x\}
```

A more complex example.

The analysis accepts to replace some constants by parameters.

```
x = 0; y = A;
        {A \le y \le 2x + A \land x \le N}
while (x<N) {
  if (?) {
        \{A \le v \le 2x + A \land x \le N - 1\}
    y = y+2;
        \{A+2 \le y \le 2x+A+2 \land x \le N-1\}
  };
        \{A \le y \le 2x + A + 2 \land 0 \le x \le N - 1\}
  x = x+1;
        \{A \le v \le 2x + A \land 1 \le x \le N\}
        {A \leq y \leq 2x + A \land N = x}
```

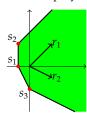
The four polyhedra operations

- ▶ \uplus ∈ $\mathbb{P}_n \times \mathbb{P}_n \to \mathbb{P}_n$: convex union
 - over-approximates the concrete union at junction points
- $ightharpoonup \cap \in \mathbb{P}_n \times \mathbb{P}_n \to \mathbb{P}_n$: intersection
 - over-approximates the concrete intersection after a conditional intruction
- ▶ $\llbracket \mathbf{x} := e \rrbracket \in \mathbb{P}_n \to \mathbb{P}_n$: affine transformation
 - over-approximates the assignment of a variable by a linear expression
- ▶ ∇ ∈ $\mathbb{P}_n \times \mathbb{P}_n \to \mathbb{P}_n$: widening
 - ensures (and accelerates)
 convergence of (post-)fixpoint
 iteration
 - includes heuristics to infer loop invariants

```
x = 0: v = 0:
        P_0 = [v := 0] [x := 0] (\mathbb{O}^2) \nabla P_A
while (x<6) {</pre>
  if (?) {
        P_1 = P_0 \cap \{x < 6\}
    y = y+2;
        P_2 = [[y := y + 2]](P_1)
  };
        P_3 = P_1 \uplus P_2
  x = x+1:
        P_4 = [x := x + 1] (P_3)
        P_5 = P_0 \cap \{x \ge 6\}
```

Library for manipulating polyhedra

- ▶ Parma Polyhedra Library ³ (PPL), NewPolka : complex C/C++ libraries
- They rely on the Double Description Method
 - polyhedra are managed using two representations in parallel



by set of inequalities

$$P = \left\{ (x,y) \in \mathbb{Q}^2 \middle| \begin{array}{c} x \geqslant -1 \\ x - y \geqslant -3 \\ 2x + y \geqslant -2 \\ x + 2y \geqslant -4 \end{array} \right\}$$

by set of generators

$$P = \left\{ \begin{array}{l} \lambda_{1}s_{1} + \lambda_{2}s_{2} + \lambda_{3}s_{3} + \mu_{1}r_{1} + \mu_{2}r_{2} \in \mathbb{Q}^{2} \middle| \begin{array}{l} \lambda_{1}, \lambda_{2}, \lambda_{3}, \mu_{1}, \mu_{2} \in \mathbb{R}^{+} \\ \lambda_{1} + \lambda_{2} + \lambda_{3} = 1 \end{array} \right\}$$

 operations efficiency strongly depends on the chosen representations, so they keep both

^{3.} Previous tutorial on polyhedra partially comes from http://www.cs.unipr.it/ pp1/

Outline

- Introduction
- 2 Intermediate representation: syntax and semantics
- Collecting semantics
- 4 Just put some #...
- 5 Building a generic abstract interpreter
- 6 Numeric abstraction by intervals
- Widening/Narrowing
- Polyhedral abstract interpretation
- Readings

References (1)

A few articles

a short formal introduction

P. Cousot and R. Cousot. Basic Concepts of Abstract Interpretation. http://www.di.ens.fr/~cousot/COUSOTpapers/WCC04.shtml

technical but very complete (the logic programming part is optional) :

P. Cousot and R. Cousot. Abstract Interpretation and Application to Logic Programs. http://www.di.ens.fr/~cousot/COUSOTpapers/JLP92.shtml

 a nice application of abstract interpretation theory to verify airbus flight commands

P. Cousot, R. Cousot, J. Feret, L. Mauborgne, A. Miné, D. Monniaux, and X. Rival. The ASTRÉE Analyser.

http://www.di.ens.fr/~cousot/COUSOTpapers/ESOP05.shtml

References (2)

On the web:

- informal presentation of AI with nice pictures
- http://www.di.ens.fr/~cousot/AI/IntroAbsInt.html
- a short abstract of various works around AI http://www.di.ens.fr/~cousot/AI/
- very complete lecture notes

http://web.mit.edu/afs/athena.mit.edu/course/16/16.399/www/

