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Abstract Interpretation

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Abstract Interpretation

□ **Abstract Interpretation: a solid mathematical foundation for reasoning about static program analyses**

■ Is the analysis **sound**?

(Does it safely approximate the actual program behavior?)

■ Is it as **precise** as possible for the currently used analysis lattice?

If not, **where** can precision **losses** arise?

Which precision losses can be **avoided** (without sacrificing soundness)?

→ **Require: a precise definition of the semantics of the PL and precise definitions of the analysis abstractions in terms of the semantics**



Agenda

- **Collecting semantics**
- Abstraction and concretization
- Soundness
- Optimality
- Completeness



Program Semantics as Constraint Systems

- Concrete state: program variables to integers

$$\text{ConcreteStates} = \text{Vars} \hookrightarrow \mathbb{Z}$$

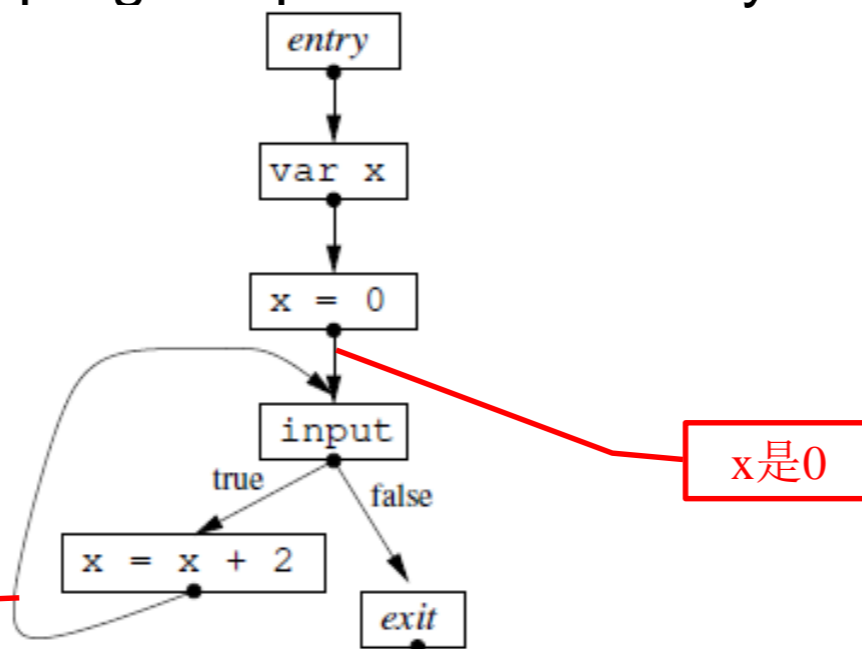
- Constraint variable for each CFG node v

- $\{v\} \subseteq \text{ConcreteStates}$ *reachable states collecting semantics*

- Denote the state at the program point immediately after v

```
var x;  
x = 0;  
while (input) {  
    x = x + 2;  
}
```

状态不唯一
x是正偶数





The Semantics of Expressions

□ Concrete execution → Abstract execution

$$ceval: ConcreteStates \times Exp \rightarrow \mathcal{P}(\mathbb{Z})$$

- Evaluate E relative to a concrete state ρ , and result in a set of possible integer values

$$ceval(\rho, X) = \{\rho(X)\}$$

$$ceval(\rho, I) = \{I\}$$

$$ceval(\rho, \text{input}) = \mathbb{Z}$$

$$ceval(\rho, E_1 \text{ op } E_2) = \{v_1 \text{ op } v_2 \mid v_1 \in ceval(\rho, E_1) \wedge v_2 \in ceval(\rho, E_2)\}$$

□ Overload $ceval$ to work on sets of concrete states

$$ceval: \mathcal{P}(ConcreteStates) \times Exp \rightarrow \mathcal{P}(\mathbb{Z})$$

$$ceval(R, E) = \bigcup_{\rho \in R} ceval(\rho, E)$$



- Possible successors of a CFG node relative to a concrete state
→ work on a set of concrete states

$$csucc: \mathcal{P}(ConcreteStates) \times Nodes \rightarrow \mathcal{P}(Nodes):$$

$$csucc(R, v) = \bigcup_{\rho \in R} csucc(\rho, v)$$

$$CJOIN(v) =$$

$$\{\rho \in ConcreteStates \mid \exists w \in Nodes: \rho \in \{w\} \wedge v \in csucc(\rho, w)\}$$



A flow-**insensitive semantics** analysis that tracks function values:

□ Assignment

$$\llbracket X=E \rrbracket = \{ \rho[X \mapsto z] \mid \rho \in CJOIN(v) \wedge z \in ceval(\rho, E) \}$$

□ Variable declaration

$$\llbracket \text{var } X_1, \dots, X_n \rrbracket = \\ \{ \rho[X_1 \mapsto z_1, \dots, X_n \mapsto z_n] \mid \rho \in CJOIN(v) \wedge z_1 \in \mathbb{Z} \wedge \dots \wedge z_n \in \mathbb{Z} \}$$

□ Initial state $\llbracket \text{entry} \rrbracket = \{ \square \}$

□ Others $\llbracket v \rrbracket = CJOIN(v)$



- A program with n CFG nodes, v_1, \dots, v_n

$$\{v_1\} = cf_{v_1}(\{v_1\}, \dots, \{v_n\})$$

$$\{v_2\} = cf_{v_2}(\{v_1\}, \dots, \{v_n\})$$

⋮

$$\{v_n\} = cf_{v_n}(\{v_1\}, \dots, \{v_n\})$$

- Combine n functions into one

$$cf: (\mathcal{P}(\text{ConcreteStates}))^n \rightarrow (\mathcal{P}(\text{ConcreteStates}))^n$$

$$cf(x_1, \dots, x_n) = (cf_{v_1}(x_1, \dots, x_n), \dots, cf_{v_n}(x_1, \dots, x_n))$$

$$x = cf(x)$$

$$\{P\} = lfp(cf)$$

the semantics of P
lfp: least fixed point



Example

```
var x;  
x = 0;  
while (input) {  
    x = x + 2;  
}
```

	solution 1	solution 2
{ <i>entry</i> }	{ \square }	{ \square }
{ <i>var x</i> }	{ $[x \mapsto z] \mid z \in \mathbb{Z}$ }	{ $[x \mapsto z] \mid z \in \mathbb{Z}$ }
{ <i>x = 0</i> }	{ $[x \mapsto 0]$ }	{ $[x \mapsto 0]$ }
{ <i>input</i> }	{ $[x \mapsto z] \mid z \in \{0, 2, 4, \dots\}$ }	{ $[x \mapsto z] \mid z \in \mathbb{Z}$ }
{ <i>x = x + 2</i> }	{ $[x \mapsto z] \mid z \in \{2, 4, \dots\}$ }	{ $[x \mapsto z] \mid z \in \mathbb{Z}$ }
{ <i>exit</i> }	{ $[x \mapsto z] \mid z \in \{0, 2, 4, \dots\}$ }	{ $[x \mapsto z] \mid z \in \mathbb{Z}$ }

the least solution



Tarski's Fixed Point Theorem for Continuous Functions

□ $f: L_1 \rightarrow L_2$ is continuous if

$$f(\bigsqcup A) = \bigsqcup_{a \in A} f(a) \quad \text{for every } A \subseteq L$$

□ If f is continuous

$$\text{fix}(f) = \bigsqcup_{i \geq 0} f^i(\perp)$$

(even when L has infinite height!)

□ cf is continuous



```
var a,b,c;  
a = 42;  
b = 87;  
if (input) {  
    c = a + b;  
} else {  
    c = a - b;  
}
```

the semantics of P

$$\llbracket b = 87 \rrbracket = \{[a \mapsto 42, b \mapsto 87, c \mapsto z] \mid z \in \mathbb{Z}\}$$

$$\llbracket c = a - b \rrbracket = \{[a \mapsto 42, b \mapsto 87, c \mapsto -45]\}$$

$$\llbracket \text{exit} \rrbracket = \{[a \mapsto 42, b \mapsto 87, c \mapsto 129], [a \mapsto 42, b \mapsto 87, c \mapsto -45]\}$$

the analysis result of P

$$\llbracket b = 87 \rrbracket = [a \mapsto +, b \mapsto +, c \mapsto \top]$$

$$\llbracket c = a - b \rrbracket = [a \mapsto +, b \mapsto +, c \mapsto \top]$$

$$\llbracket \text{exit} \rrbracket = [a \mapsto +, b \mapsto +, c \mapsto \top]$$



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□ Abstract functions

$\alpha_a: \mathcal{P}(\mathbb{Z}) \rightarrow \text{Sign}$ Sets of Concrete value \rightarrow sets of abstract value

$\alpha_b: \mathcal{P}(\text{ConcreteStates}) \rightarrow \text{States}$ $\text{ConcreteStates} = \text{Vars} \rightarrow \mathbb{Z}$

$\alpha_c: (\mathcal{P}(\text{ConcreteStates}))^n \rightarrow \text{States}^n$ $\text{State} = \text{Vars} \rightarrow \text{Sign}$

n-tuple of sets of concrete states \rightarrow those of abstract states

$$\alpha_a(D) = \begin{cases} \perp & \text{if } D \text{ is empty} \\ + & \text{if } D \text{ is nonempty and contains only positive integers} \\ - & \text{if } D \text{ is nonempty and contains only negative integers} \\ \emptyset & \text{if } D \text{ is nonempty and contains only the integer 0} \\ \top & \text{otherwise} \end{cases}$$

for any $D \in \mathcal{P}(\mathbb{Z})$

$$\alpha_b(R) = \sigma \text{ where } \sigma(X) = \alpha_a(\{\rho(X) \mid \rho \in R\}) \\ \text{for any } R \subseteq \text{ConcreteStates} \text{ and } X \in \text{Vars}$$

$$\alpha_c(R_1, \dots, R_n) = (\alpha_b(R_1), \dots, \alpha_b(R_n)) \\ \text{for any } R_1, \dots, R_n \subseteq \text{ConcreteStates}$$



□ Concretization functions

$$\gamma_a: \text{Sign} \rightarrow \mathcal{P}(\mathbb{Z})$$

$$\gamma_b: \text{States} \rightarrow \mathcal{P}(\text{ConcreteStates})$$

$$\gamma_c: \text{States}^n \rightarrow (\mathcal{P}(\text{ConcreteStates}))^n$$

$$\gamma_a(s) = \begin{cases} \emptyset & \text{if } s = \perp \\ \{1, 2, 3, \dots\} & \text{if } s = + \\ \{-1, -2, -3, \dots\} & \text{if } s = - \\ \{0\} & \text{if } s = \mathbf{0} \\ \mathbb{Z} & \text{if } s = \top \end{cases}$$

for any $s \in \text{Sign}$

$$\gamma_b(\sigma) = \{\rho \in \text{ConcreteStates} \mid \rho(X) \in \gamma_a(\sigma(X)) \text{ for all } X \in \text{Vars}\}$$

for any $\sigma \in \text{States}$

$$\gamma_c(\sigma_1, \dots, \sigma_n) = (\gamma_b(\sigma_1), \dots, \gamma_b(\sigma_n))$$

for any $(\sigma_1, \dots, \sigma_n) \in \text{States}^n$

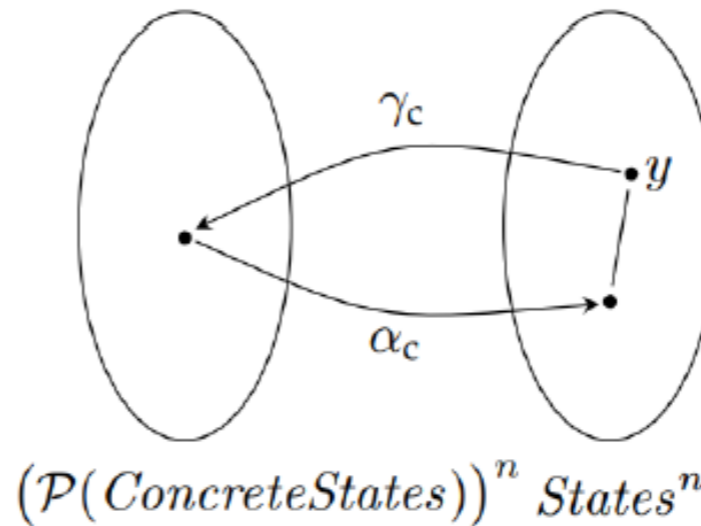
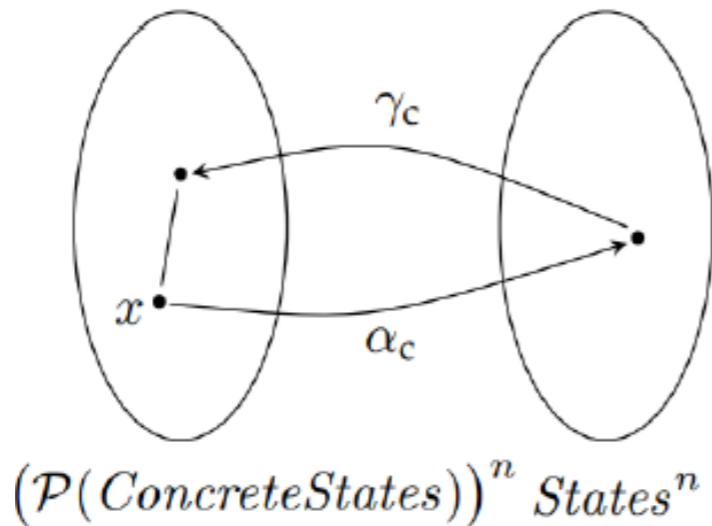


□ Galois Theory 伽罗瓦理论

■ 建立域论和群论之间的联系：自同构群（Galois 群）

The pair of monotone functions, α and γ , is called a *Galois connection* if

$$\begin{aligned} \gamma \circ \alpha \text{ is extensive} & \quad x \sqsubseteq \gamma(\alpha(x)) \text{ for all } x \in L_1 \\ \alpha \circ \gamma \text{ is reductive} & \quad \alpha(\gamma(y)) \sqsubseteq y \text{ for all } y \in L_2 \end{aligned}$$



all three pairs of abstraction and concretization functions (α_a, γ_a) , (α_b, γ_b) , and (α_c, γ_c) from the sign analysis example are Galois connections



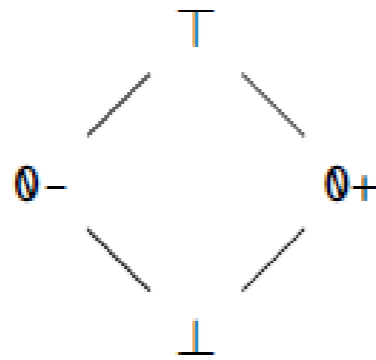
- The concretization function uniquely determines the abstraction function and vice versa:

$$\gamma(y) = \bigsqcup_{x \in L_1 \text{ where } \alpha(x) \sqsubseteq y} x$$

$$\alpha(x) = \bigsqcap_{y \in L_2 \text{ where } x \sqsubseteq \gamma(y)} y$$



- For this lattice, given the “obvious” concretization function, is there an **abstraction** function such that the concretization function and the abstraction function form a Galois connection? **[Not exist]**



How should we define $\alpha_a(\{0\})$?
Exercise 11.22

the concretization function

$$\gamma_a(s) = \begin{cases} \emptyset & \text{if } s = \perp \\ \{0, 1, 2, 3, \dots\} & \text{if } s = 0_+ \\ \{0, -1, -2, -3, \dots\} & \text{if } s = 0_- \\ \mathbb{Z} & \text{if } s = \top \end{cases}$$



□ Assume $\alpha_a(\{0\}) = \mathbf{0}_-$ and $\alpha_a(\{0,1\}) = \mathbf{0}_+$

```
x = input > 3;
if (!x) {
  output x;
}
```

Flow-**insensitive** sign analysis

$$\mathbf{x} \rightarrow \mathbf{0}_+$$

$$\mathbf{x} \rightarrow \mathbf{0}_+$$

```
x = input > 3;
if (!x) {
  output x;
}
```

Flow-**sensitive** sign analysis

$$\mathbf{x} \rightarrow \mathbf{0}_+$$

$$\text{assert}(!x): \llbracket v \rrbracket = \begin{cases} \sigma[\mathbf{x} \mapsto \alpha_a(\{0\})] & \text{if } 0 \in \gamma_a(\sigma(\mathbf{x})) \\ \perp & \text{otherwise} \end{cases}$$

where $\sigma = \text{JOIN}(v)$

$$\mathbf{x} \rightarrow \mathbf{0}_-$$



□ For ordinary Sign lattice

$$\beta: \mathbb{Z} \rightarrow \text{Sign}$$

Representation Function

$$\beta(d) = \begin{cases} + & \text{if } d > 0 \\ - & \text{if } d < 0 \\ \emptyset & \text{if } d = 0 \end{cases}$$

Abstraction Function

$$\alpha_a(D) = \bigsqcup \{\beta(d) \mid d \in D\}$$



Agenda

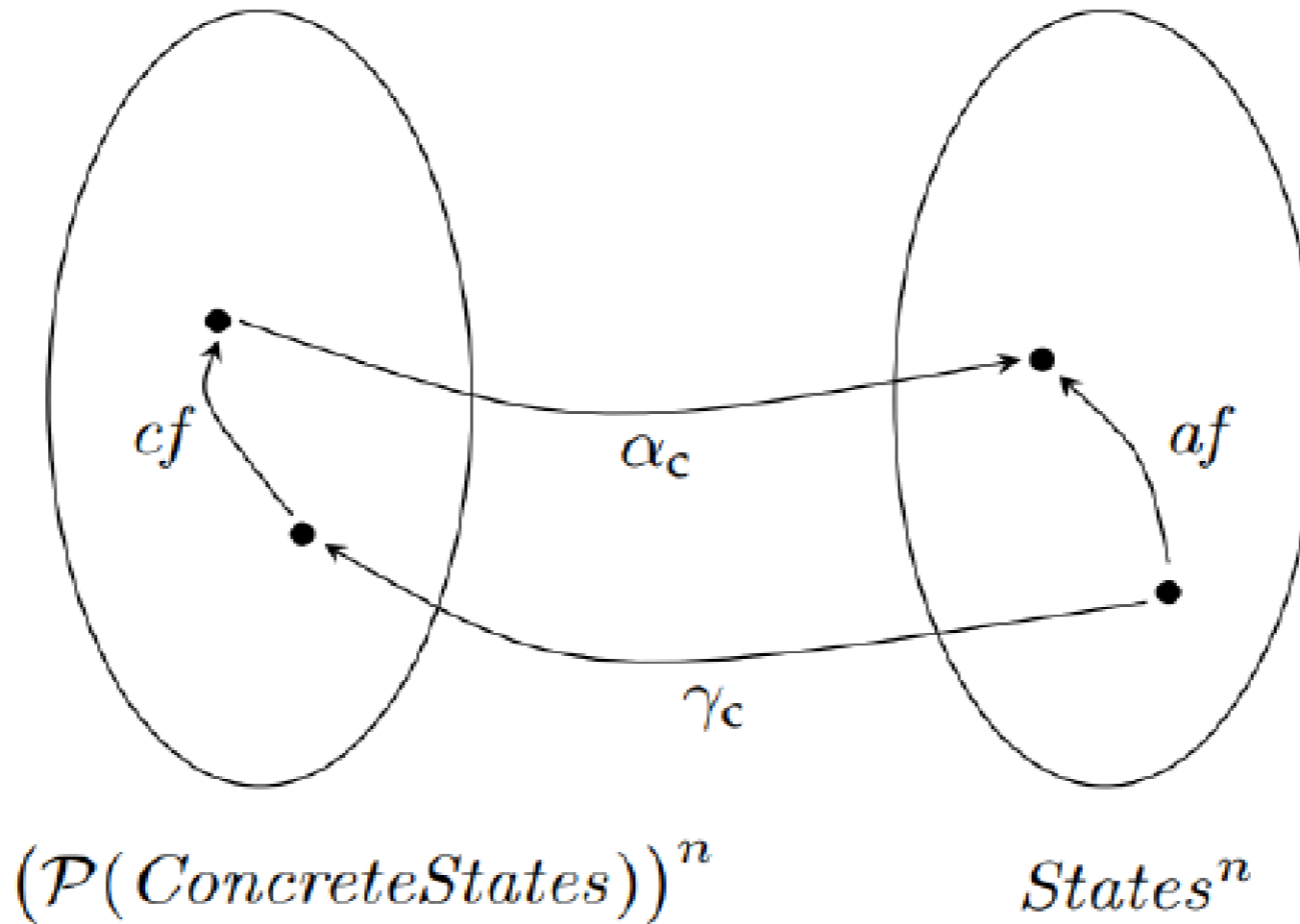
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Optimal Approximations

af is an *optimal* approximation of cf if

$$af = \alpha \circ cf \circ \gamma$$





Optimal Approximations in Sign Analysis?

$\hat{*}$ is optimal:

$$s_1 \hat{*} s_2 = \alpha_a(\gamma_a(s_1) \cdot \gamma_a(s_2))$$

eval is not optimal:

$$\sigma(\mathbf{x}) = \top$$

$$\text{eval}(\sigma, \mathbf{x} - \mathbf{x}) = \top$$

$$\alpha_b(\text{ceval}(\gamma_b(\sigma), \mathbf{x} - \mathbf{x})) = \mathbf{0}$$

Even if we could make *eval* optimal, the analysis result is not always optimal:

```
x = input;
```

```
y = x;
```

```
z = x - y;
```



We need

- Static analysis (the analysis lattices and constraint rules)
- Language semantics (suitable collecting semantics)
- Abstract/concretization functions that specify the meaning of the elements in the analysis lattice in terms of the semantic lattice

... and then

- If each constituent of the analysis is a sound abstraction of its semantic counterpart, then the analysis is sound
- If an abstraction is optimal, then it is as precise as possible, relative to the choice of analysis lattice
- If the analysis is sound and complete, then the analysis result is as precise as possible, relative to the choice of analysis lattice



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Thanks