

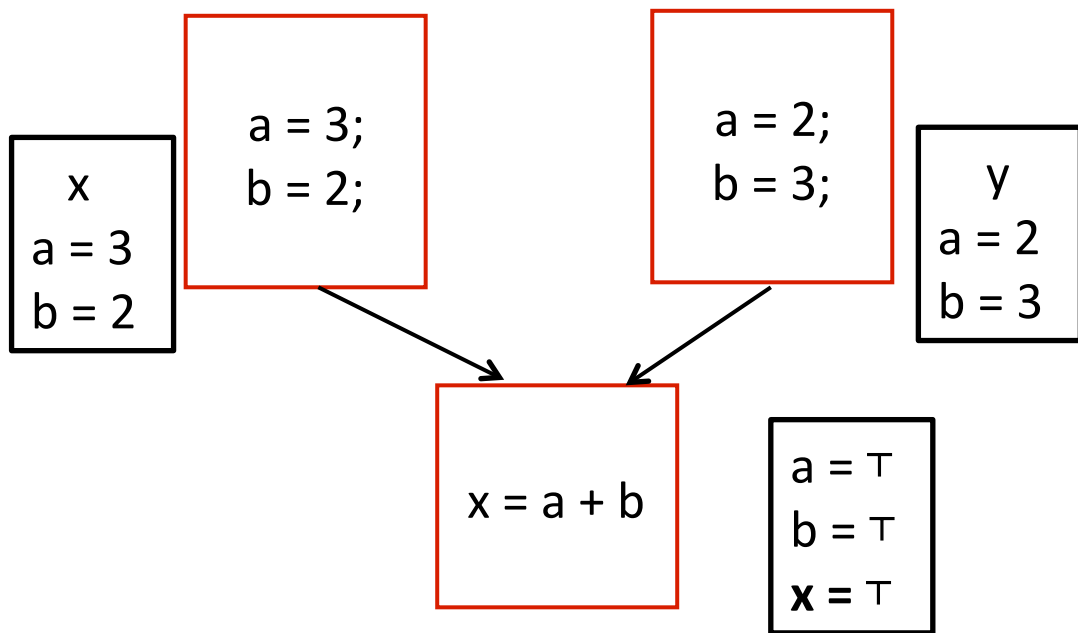
Abstract Interpretation

CSE 501

Spring 15

Distributivity of Frameworks

- (G, L, F, M) is distributive iff
 $f(x \wedge y) = f(x) \wedge f(y)$



$$\begin{aligned}
 f(x \wedge y) &= f(\{3, 2, ?\} \wedge \{2, 3, ?\}) \\
 &= f(\{\top, \top, ?\}) \\
 &= \{\top, \top, \top\}
 \end{aligned}$$

$$\begin{aligned}
 f(x) \wedge f(y) &= f(\{3, 2, ?\}) \wedge f(\{2, 3, ?\}) \\
 &= \{3, 2, 5\} \wedge \{2, 3, 5\} \\
 &= \{\top, \top, 5\}
 \end{aligned}$$

Ordering of evaluation matters!

Maximal Fixed Point (MFP) Solution

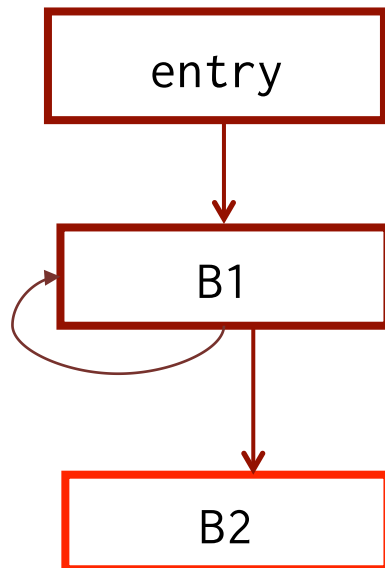
- Fact: the iterative solution to dataflow equations is the most precise
- Intuition:
 - Start with the top element at each program point
 - Refine during each iteration to satisfy all dataflow equations
 - Final result will be closest to the top
- Hence for any solution FP of dataflow equations:
 $FP \leq MFP$

Meet Over Paths (MOP) Solution

- Another approach to solve the dataflow equations:
 - Enumerate each path $p_k = [\text{entry}, n_1, n_2, \dots, n_k]$
 - Define $IN[p_k] = f_{n_{k-1}}(\dots (f_{n_1}(f_{n_0}(d_0))))$, where d_0 is the flow element for entry
 - Compute final solution as
$$IN[n] = U \{ IN[p] \ . \ p \text{ is a path from entry to } p \}$$

MFP and MOP

- Fact: $MFP \leq MOP$
- Why not compute MOP in practice?



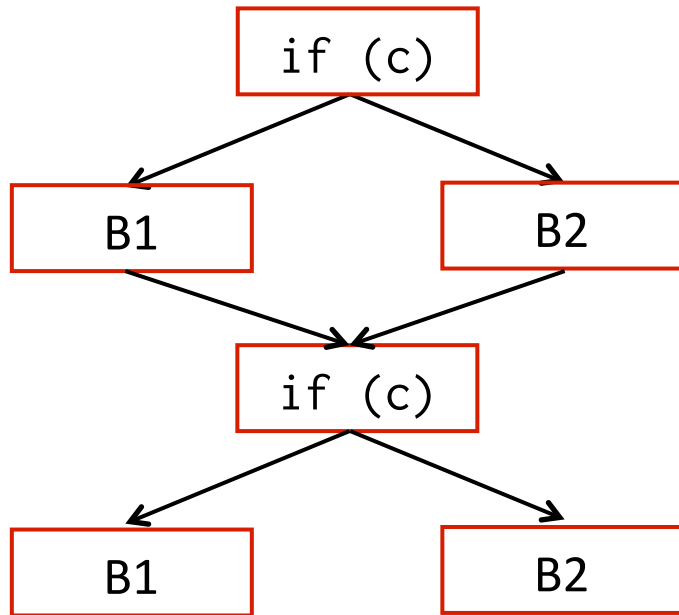
How many paths can reach B2?

MFP and MOP

- Fact: For transfer functions that are distributive, then MFP = MOP
- Recall: $f(x \wedge y) = f(x) \wedge f(y)$
- Hence $f(x_1) \wedge f(x_2) \wedge f(x_3) \dots = f(\wedge x_i)$
- We can compute MOP using iterative algorithm!

Can we do even better?

- Fact: MFP, MOP are conservative



- Some paths are not possible
- IDEAL = solution that takes into account of feasible paths
- $FP \leq MFP \leq MOP \leq IDEAL$
- Great!
 - but this is undecidable 😞

Summary

- Dataflow framework = (G, L, F, M)
- Possible solutions: FP, MFP, MOP, IDEAL
 - $FP \leq MFP \leq MOP \leq IDEAL$
- In practice, compilers compute MFP using the iterative algorithm

ABSTRACT INTERPRETATION : A UNIFIED LATTICE MODEL FOR STATIC ANALYSIS
OF PROGRAMS BY CONSTRUCTION OR APPROXIMATION OF FIXPOINTS

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[Abstract Interpretation: A Unified Lattice Model for Static ...](#)

dl.acm.org/ft_gateway.cfm?id... Association for Computing Machinery ▾

by P Cousot - 1977 - Cited by 5755 - [Related articles](#)

Where it all started...

- Inspirations from
 - Dataflow analysis
 - Denotational semantics
- Enthusiastically embraced by the community
 - At least the functional community . . .
 - At least the first half of the paper . . .

AI by Example

with slides from Prof. Alex Aiken

A Tiny Language

- Language with only integers and multiplication

$$e = i \mid e * e$$

$\mu : \text{Exp} \rightarrow \text{Int}$ \longleftarrow Denotation / meaning function

$$\mu(i) = i$$

$$\mu(e * e) = \mu(e) \times \mu(e)$$

- Goal: define a semantics to compute the sign of all expressions without actually carrying out the computation

An Abstraction

- Define an abstract semantics that computes only the sign of the result.

$\sigma: \text{Exp} \rightarrow \{+, -, 0\}$

+ if $i > 0$

$\sigma(i) = 0$ if $i = 0$

- if $i < 0$

<u>x</u>	+	0	-
+	+	0	-
0	0	0	0
-	-	0	+

$\sigma(e * e) = \sigma(e) \underline{x} \sigma(e)$

Soundness

- We can show that this abstraction is correct in that it correctly predicts the sign of an expression.
- Proof is by structural induction on e .

$$\mu(e) > 0 \Leftrightarrow \sigma(e) = +$$

$$\mu(e) = 0 \Leftrightarrow \sigma(e) = 0$$

$$\mu(e) < 0 \Leftrightarrow \sigma(e) = -$$

Another View of Soundness

- The soundness proof is clunky
- Instead, directly associate each abstract value with the set of concrete values it represents.

$$\gamma : \{+, 0, -\} \rightarrow 2^{\text{Int}}$$

$$\gamma(+)=\{i \mid i > 0\}$$

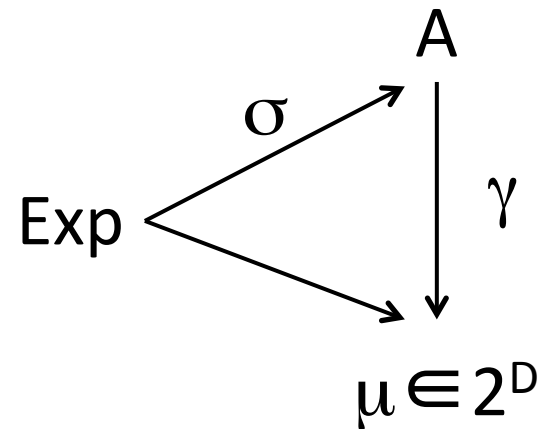
$$\gamma(0)=\{0\}$$

$$\gamma(-)=\{i \mid i < 0\}$$

Another View of Soundness

- The concretization function γ
 - Mapping from abstract values to (sets of) concrete values
- Let
 - \mathbf{D} be the concrete domain
 - \mathbf{A} be the abstract domain

$$\mu(e) \in \gamma(\sigma(e))$$



Abstract Interpretation

- This is an abstract interpretation
 - Computation in an abstract domain
 - In this case $\{+,0,-\}$.
- The abstract semantics is sound
 - approximates the standard semantics.
- The concretization function establishes the connection between the two domains.

Adding -

- Extend our language with unary -

$$\mu(-e) = -\mu(e)$$

$$\sigma(-e) = _ \sigma(e)$$

-	+	0	-
	-	0	+

Adding +

- Adding addition is not so easy.
- The abstract values are not closed under addition.

$$\mu(e_1 + e_2) = \mu(e_1) + \mu(e_2)$$

$$\sigma(e_1 + e_2) = \sigma(e_1) \pm \sigma(e_2)$$

<u>±</u>	+	0	-
+	+	+	?
0	+	0	-
-	?	-	-

Solution

- We need another abstract value to represent a result that can be any integer
- Finding a domain closed under all the abstract operations is often a key design problem
- Recall: defining lattice for dataflow analysis

$\gamma(T) = \text{all integers}$

<u>±</u>	+	0	-	T
+	+	+	T	T
0	+	0	-	T
-	T	-	-	T
T	T	T	T	T

Extending Other Operations

- We also need to extend the other abstract operations to work with T.

<u>X</u>	+	0	-	T
+	+	0	-	T
0	0	0	0	0
-	-	0	+	T
T	T	0	T	T

<u>-</u>	+	0	-	T
-	-	0	+	T

Examples

- Abstract computation doesn't lose information:

$$\mu((5 * 5) + 6) = 31$$

$$\sigma((5 * 5) + 6) = (+ \underline{x} +) \underline{\pm} + = +$$

- Sometimes it does:

$$\mu((1 + 2) + -3) = 0$$

$$\sigma((1 + 2) + -3) = (+ \underline{\pm} +) \underline{\pm} (\underline{-}+) = T$$

Adding / (Integer Division)

- Adding / is straightforward except for the case of division by 0.
- If we divide each integer in a set by 0, what set of integers results?
 - The empty set.

$$\gamma(\perp) = \emptyset$$

\backslash	+	0	-	\top	\perp
+	+	0	-	\top	\perp
0	0	0	0	0	\perp
-	-	0	+	\top	\perp
\top	\top	0	\top	\top	\perp
\perp	\perp	\perp	\perp	\perp	\perp

Adding / (Integer Division)

- As before we need to extend the other abstract operations.
- In this case, every entry involving bottom is bottom
 - all operations are strict in bottom

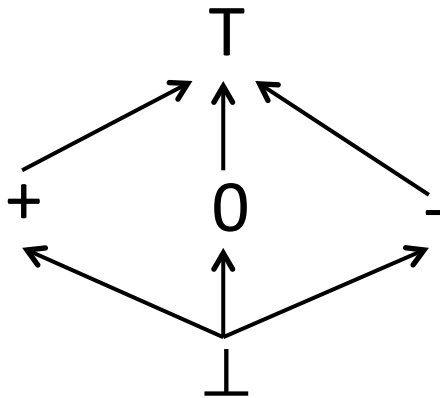
$$\perp \underline{+} X = \perp$$

$$X \underline{\times} \perp = \perp$$

$$\underline{-} \perp = \perp$$

The Abstract Domain

- Our abstract domain forms a **complete lattice**.
 - A partial order $x \leq y \Leftrightarrow \gamma(x) \subseteq \gamma(y)$
- Every finite subset has a least upper bound (lub, \sqcup) and greatest lower bound (glb, \sqcap).
- We write **A** for an abstract domain
 - a set of values + an ordering



The Abstraction Function

- The abstraction function maps concrete values to abstract values.
 - The dual of concretization.
 - The smallest value of **A** that is the abstraction of a set of concrete values.

$$\alpha: 2^{\text{Int}} \rightarrow A$$

$$\alpha(S) = \text{lub}(\{- | i < 0 \wedge i \in S\}, \{0 | 0 \in S\}, \{+ | i > 0 \wedge i \in S\})$$

An Aside: Galois Connection

- (L, α, γ, M) is a Galois connection between complete lattices (L, \leq) and (M, \leq) iff:
 - $\alpha: L \rightarrow M$ and $\gamma: M \rightarrow L$ are monotone functions

Furthermore:

- $\text{id} \leq \gamma \circ \alpha$
- $\text{id} \leq \alpha \circ \gamma$

- The function $\alpha \circ \gamma$ is called a **Galois insertion**

A General Definition

- An abstract interpretation consists of
 - An abstract domain **A** and concrete domain **D**
 - Concretization and abstraction functions forming a Galois insertion.
 - A (sound) abstract semantic function.

In our example:

$$\forall x \in 2^D . x \subseteq \gamma(\alpha(x))$$

$$\forall a \in \mathbf{A} . a = \alpha(\gamma(x))$$

or

$$\text{id} \leq \gamma \circ \alpha$$

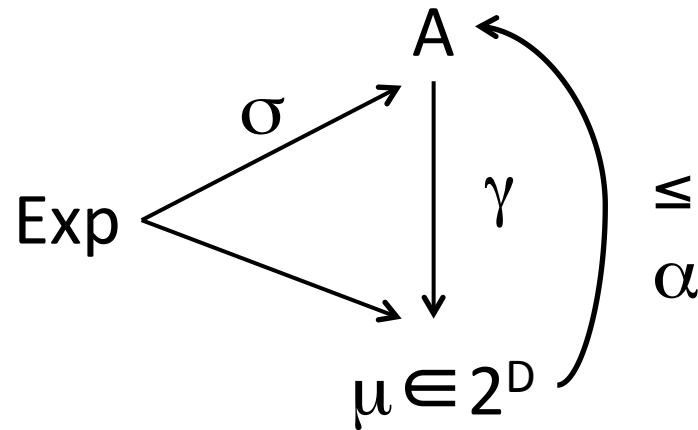
$$\text{id} = \alpha \circ \gamma$$

Galois Insertions

- The abstract domain can be thought of as dividing the concrete domain into non-disjoint subsets
- The abstraction function maps a subset of the domain to the smallest containing abstract value.

Pictorially

- In correct abstract interpretations, we expect the following diagram to commute.



General Conditions for Correctness

- Three conditions guarantee correctness in general:
 - α and γ form a Galois insertion
 - $\text{id} \leq \gamma \circ \alpha$, $\text{id} = \alpha \circ \gamma$
 - α and γ are monotonic
 - $x \leq y \Rightarrow \alpha(x) \leq \alpha(y)$
 - Abstract operations op are locally correct:
 - $\gamma(\underline{\text{op}}(s_1, \dots, s_n)) \subseteq \text{op}(\gamma(s_1), \dots, \gamma(s_n))$

Generic Correctness Proof

- Proof by induction on the structure of e :
 $\mu(e) \in \gamma(\sigma(e))$

$$\mu(e_1 \text{ op } e_2)$$

$$= \mu(e_1) \text{ op } \mu(e_2) \quad [\text{definition of } \mu]$$

$$\in \gamma(\sigma(e_1)) \text{ op } \gamma(\sigma(e_2)) \quad [\text{induction}]$$

$$\subseteq \gamma(\sigma(e_1) \text{ op } \sigma(e_2)) \quad [\text{local correctness}]$$

$$= \gamma(\sigma(e_1 \text{ op } e_2)) \quad [\text{definition of } \sigma]$$

Another Notion of Correctness

- We can define correctness using abstraction instead of concretization.

$$\mu(e) \in \gamma(\sigma(e)) \Leftrightarrow \alpha(\{\mu(e)\}) \leq \sigma(e)$$

Proof for \Rightarrow direction:

$$\mu(e) \in \gamma(\sigma(e))$$

$$\alpha(\{\mu(e)\}) \leq \alpha(\gamma(\sigma(e)))$$

[monotonicity]

$$\alpha(\{\mu(e)\}) \leq \sigma(e)$$

$[\alpha \circ \gamma = \text{id}]$

Another Notion of Correctness

$$\mu(e) \in \gamma(\sigma(e)) \Leftrightarrow \alpha(\{\mu(e)\}) \leq \sigma(e)$$

Proof for \Leftarrow direction:

$$\alpha(\{\mu(e)\}) \leq \sigma(e)$$

$$\gamma(\alpha(\{\mu(e)\})) \leq \gamma(\sigma(e)) \quad [\text{monotonicity}]$$

$$\mu(e) \in \gamma(\sigma(e)) \quad [\text{id} \leq \gamma \circ \alpha]$$

Extending Our Language

- Add input to the language
 - Modeled as a single free variable x in expressions

$$e = i \mid e * e \mid -e \mid e + e \mid \dots \mid \mathbf{x}$$

Semantics

- The meaning function now has type

$$\mu : \text{Exp} \rightarrow \text{Int} \rightarrow \text{Int}$$

- We write the function with the expression as a subscript.

$$\mu_i(j) = i$$

$$\mu_x(j) = j$$

$$\mu_{e_1 * e_2}(j) = \mu_{e_1}(j) * \mu_{e_2}(j)$$

$$\mu_{e_1 + e_2}(j) = \mu_{e_1}(j) + \mu_{e_2}(j)$$

...

Abstract Semantics

- Abstract semantic function:

$$\sigma : \text{Exp} \rightarrow A \rightarrow A$$

- Also write this semantics in the same form.

$$\sigma_i(\underline{j}) = \underline{j}$$

$$\sigma_x(\underline{j}) = \underline{j}$$

$$\sigma_{e_1 * e_2}(\underline{j}) = \sigma_{e_1}(\underline{j}) * \sigma_{e_2}(\underline{j})$$

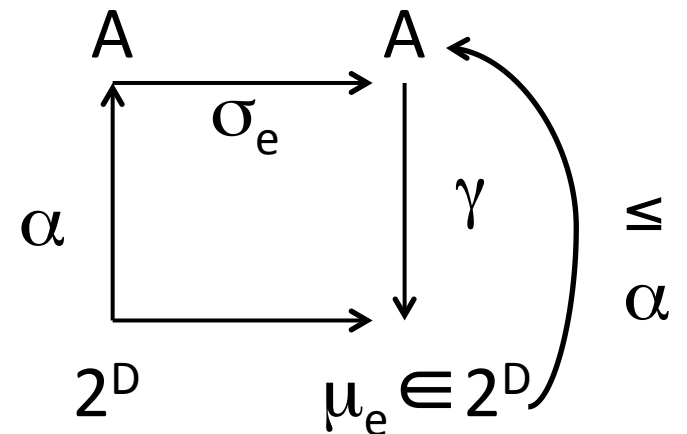
$$\sigma_{e_1 + e_2}(\underline{j}) = \sigma_{e_1}(\underline{j}) \pm \sigma_{e_2}(\underline{j})$$

... ..

$$\underline{j} = \alpha(\{i\})$$

Correctness

- The correctness condition needs to be generalized.
- This is the first real use of the abstraction function.
- The following are all equivalent:
 - $\forall i . \mu_e(i) \in \gamma(\sigma_e(\alpha(\{i\})))$
 - $\mu_e \leq_D \gamma \circ \sigma_e \circ \alpha$
 - $\alpha \circ \mu_e \leq_A \sigma_e \circ \alpha$



Local Correctness

- We also need a modified local correctness condition.

$$\text{op}(\gamma(\sigma_{e_1}(j)), \dots, \gamma(\sigma_{e_n}(j))) \subseteq \gamma(\underline{\text{op}}(\sigma_{e_1}(j), \dots, \sigma_{e_n}(j)))$$

Proof of Correctness

- Theorem: $\mu_e(j) \in \gamma(\sigma_e(j))$
- Proof (by induction on the structure of e):

Base case: $\mu_i(j) = i \in \gamma(j) = \gamma(\sigma_i(j))$

$$\mu_x(j) = j \in \gamma(j) = \gamma(\sigma_x(j))$$

Induction on $\mu_{\text{op}(e_1, \dots, e_n)}(j)$:

$$= \text{op}(\mu_{e_1}(j), \dots, \mu_{e_n}(j)) \quad [\text{definition of } \mu]$$

$$\in \text{op}(\gamma(\sigma_{e_1}(j)), \dots, \gamma(\sigma_{e_n}(j))) \quad [\text{induction}]$$

$$\subseteq \gamma(\underline{\text{op}}(\sigma_{e_1}(j), \dots, \sigma_{e_n}(j))) \quad [\text{local correctness}]$$

$$= \gamma(\sigma_{\text{op}(e_1, \dots, e_n)}(j)) \quad [\text{definition of } \sigma]$$

If-Then-Else

- $e = \dots \mid \text{if } e = e \text{ then } e \text{ else } e \mid \dots$
- $\mu_{\text{if } e_1=e_2 \text{ then } e_3 \text{ else } e_4}(\underline{i}) = \mu_{e_3}(\underline{i}), \text{ if } \mu_{e_1}(\underline{i}) = \mu_{e_2}(\underline{i})$
 $= \mu_{e_4}(\underline{i}), \text{ otherwise}$
- $\sigma_{\text{if } e_1=e_2 \text{ then } e_3 \text{ else } e_4}(\underline{i}) = \sigma_{e_3}(\underline{i}) \sqcup \sigma_{e_4}(\underline{i})$
- Recall that the abstract domain forms a complete lattice

Correctness of If-Then-Else

- Need to show that: $\mu_e(j) \in \gamma(\sigma_e(\underline{j}))$
 - Where e is an if-then-else
- Assume the true branch is taken.
- (The argument for the false branch is symmetric.)

$\mu_{e3}(i)$

$\in \gamma(\sigma_{e3}(\underline{i}))$ [by induction]

$\subseteq \gamma(\sigma_{e3}(\underline{i})) \sqcup \gamma(\sigma_{e4}(\underline{i}))$

$\subseteq \gamma(\sigma_{e3}(\underline{i}) \sqcup \sigma_{e4}(\underline{i}))$ [by monotonicity of γ]

Designing an Abstract Interpretation

- Define abstract domain
 - Needs to be a lattice
- Define the abstraction and concretization functions
 - $\sigma : \mathbf{D} \rightarrow \mathbf{A}$
 - $\alpha : 2^{\mathbf{D}} \rightarrow \mathbf{A}$
 - $\alpha(S) = \text{lub}(\sigma(s)), \text{ for all } s \in S$
 - $\gamma : \mathbf{A} \rightarrow 2^{\mathbf{D}}$
- For every expression, define how to operate in the abstract domain