

Abstract Interpretation

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Fundamental Challenge of Program Analysis

How to infer (loop) invariants ?

Fundamental Challenge of Program Analysis

- ▶ Key issue for any analysis or verification
- ▶ Many algorithms/heuristics
- ▶ See Suzuki & Ishihata, POPL 1977
- ▶ Most formalizable in framework of **Abstract Interpretation**

Abstract Interpretation

“A systematic basis for *approximating* the semantics of programs”

- ▶ Deep and broad area
- ▶ Rich theory
- ▶ Profound practical impact

We look at a tiny slice

- ▶ In context of algorithmic verification of **IMP**

IMP: A Small Imperative Language

Recall the syntax of IMP

```
data Com = Var ' := ' Expr           -- assignment
         | Com ';' Com               -- sequencing
         | Assume Exp                -- assume
         | Com '|' com                -- branch
         | While Pred Exp Com        -- loop
```

Note

We have thrown out If and Skip using the abbreviations:

```
Skip          == Assume True
If e c1 c2    == (Assume e; c1) | (Assume (!e); c2)
```

IMP: Operational Semantics

States

A State is a map from Var to the set of Values

```
type State = Map Var Value
```

IMP: Operational Semantics

Transition Relation

A subset of $\text{State} \times \text{Com} \times \text{State}$ formalized by

- ▶ $\text{eval } s \ c == [s' \mid \text{command } c \text{ transitions state } s \text{ to } s']$

```
eval :: State -> Com -> [State]
eval s (Assume e) = if eval s e then [s] else []
eval s (x := e) = [ add x (eval s e) s ]
eval s (c1 ; c2) = [s2 | s1 <- eval s c1, s2 <- eval s'
eval s (c1 | c2) = eval s c1 ++ eval s c2
eval s w@(While e c) = eval s $ Assume !e | (Assume e; c; w)
```

IMP: Axiomatic Semantics

State Assertions

- ▶ An assertion P is a Predicate over the set of program variables.
- ▶ An assertion corresponds to a **set of states**

states $P = [s \mid \text{eval } s \ P == \text{True}]$

IMP: Axiomatic Semantics

Describe execution via **Predicate Transformers**

Strongest Postcondition

$SP :: \text{Pred} \rightarrow \text{Com} \rightarrow \text{Pred}$

$SP\ P\ c$: States **reachable from** P by executing c

$\text{states}\ (SP\ P\ c) == [s' \mid s \leftarrow \text{states}\ P, s' \leftarrow \text{eval}\ s\ c]$

IMP: Axiomatic Semantics

Describe execution via **Predicate Transformers**

Weakest Precondition

$WP :: Com \rightarrow Pred \rightarrow Pred$

$WP\ c\ Q$: States that **can reach** Q by executing c

$states\ (WP\ c\ Q)' = [s \mid s' \leftarrow eval\ s\ c, eval\ s'\ Q]$

Strongest Postcondition

SP P c : States **reachable from** P by executing c

SP $\text{Pred} \rightarrow \text{Com} \rightarrow \text{Pred}$

SP P (Assume e) = P '&&' e

SP P (x := e) = Exists x'. P[x'/x] '&&' x '==' e[x'/x]

SP P (c1 ; c2) = SP (SP P c1) c2

SP P (c1 | c2) = SP P c1 '||' SP p c2

SP P w@(W e c) = SP s (Assume !e | (Assume e; c; w))

- ▶ **Uh Oh!** last case is non-terminating ...

Weakest Precondition

WP c Q : States that **can reach** Q by executing c

WP $:: \text{Com} \rightarrow \text{Pred} \rightarrow \text{Pred}$

WP (Assume e) $Q = e \text{ '=>' } Q$

WP ($x := e$) $Q = Q[e/x]$

WP ($c1 ; c2$) $Q = \text{WP } c1 (\text{WP } c2 Q)$

WP ($c1 \mid c2$) $Q = \text{WP } c1 Q \text{ '&\&' } \text{WP } c2 Q$

WP $w@(W e c)$ $Q = \text{WP } (\text{Assume } !e \mid (\text{Assume } e; c; w)) Q$

- ▶ **Uh Oh!** last case is non-terminating ...

IMP: Verification (Suspend disbelief regarding loops)

Goal: Verify Hoare-Triples

Given

- ▶ c command
- ▶ P precondition
- ▶ Q postcondition

Prove

- ▶ **Hoare-Triple** $\{P\} c \{Q\}$ which denotes

```
forall s s'. if s 'in' (states P) &&
              s' 'in' (eval s c)
              then
                s' 'in' (states Q)
```

Verification Strategy

(For a moment, suspend disbelief regarding loops)

1. Compute Verification Condition (VC)

- ▶ $(SP \ P \ c) \Rightarrow Q$

- ▶ $P \Rightarrow (WP \ c \ Q)$

2. Use SMT Solver to Check VC is Valid

Verification Strategy

1. Compute Verification Condition (VC)
 - ▶ $(SP \ P \ c) \Rightarrow Q$
 - ▶ $P \Rightarrow (WP \ c \ Q)$
2. Use SMT Solver to Check VC is Valid

Problem: Pesky Loops

- ▶ Cannot compute WP or SP for `While b c ...`
- ▶ ... Require **invariants**

Next: Lets **infer** invariants by **approximation**

Approximate Verification Strategy

0. Compute **Over-approximate** Postcondition SP# s.t.
 - ▶ $(SP \ P \ c) \Rightarrow (SP\# \ P \ c)$
1. Compute Verification Condition (VC)
 - ▶ $(SP\# \ P \ c) \Rightarrow Q$
2. Use SMT Solver to Check VC is Valid
 - ▶ If so, $\{P\} \ c \ \{Q\}$ holds by **Consequence Rule**

Key Requirement

- ▶ Compute SP# **without** computing SP ...
- ▶ But **guaranteeing** over-approximation

What Makes Loops Special?

Why different from other constructs? Let

- ▶ c be a loop-free (i.e. has no `While` inside it)
- ▶ W be the loop `While b c`

Loops as Limits

Inductively define the *infinite* sequence of loop-free Com

$W_0 = \text{Skip}$

$W_1 = W_0 \mid \text{Assume } b; c; W_0$

$W_2 = W_1 \mid \text{Assume } b; c; W_1$

·

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$W_{i+1} = W_i \mid \text{Assume } b; c; W_i$

·

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Loops as Limits

Intuitively

- ▶ W_i is the loop **unrolled upto** i times
- ▶ $W == W_0 \mid W_1 \mid W_2 \mid \dots$

Formally, we can prove (**exercise**)

1. `eval s W == eval s W_0 ++ eval s W_1 ++ ...`

2. `SP P W == SP P W_0 || SP P W_1 || ...`

3. `WP W Q == WP W_0 Q && WP W_1 Q && ...`

So what? Still cannot **compute** SP or WP ...!

Loops as Limits

So what? Still cannot **compute** SP or WP ... but notice

$$\begin{aligned} \text{SP } P \ W_{i+1} &== \text{SP } P \ (W_i \mid \text{assume } b; c; W_i) \\ &== \text{SP } P \ W_i \ \parallel \ \text{SP } (\text{SP } P \ (\text{assume } b; c)) \ W_i \\ &<= \text{SP } P \ W_i \end{aligned}$$

That is, SP P W_i form an **increasing chain**

$$\text{SP } P \ W_0 \Rightarrow \text{SP } P \ W_1 \Rightarrow \dots \Rightarrow \text{SP } P \ W_i \Rightarrow \dots$$

... **Problem:** Chain does not converge! *ONION RINGS*

Approximate Loops as Approximate Limits

To find $SP\#$ such that $SP\ P\ c \Rightarrow SP\#\ P\ c$, we compute chain

$SP\#\ P\ W_0 \Rightarrow SP\#\ P\ W_1 \Rightarrow \dots \Rightarrow SP\#\ P\ W_i \Rightarrow \dots$

where each $SP\#$ over-approximates the corresponding SP

for all i . $SP\ P\ W_i \Rightarrow SP\#\ P\ W_i$

and the chain of $SP\#$ chain converges to a **fixpoint**

exists j . $SP\#\ P\ W_{j+1} == SP\#\ P\ W_j$

This magic $SP\#\ P\ W_{j+1}$ is the loop invariant, and

$SP\#\ P\ W == SP\#\ P\ W_j$

Approximating Loops

Many Questions Remain Around Our Strategy

How to compute SP# so that we can ensure

1. **Convergence** to a fixpoint ?
2. Result is an **over-approximation** of SP ?

Answer: Abstract Interpretation

“Systematic basis for **approximating the semantics** of programs”

Abstract Interpretation

Plan

1. Simple language of arithmetic expressions
2. IMP
3. Predicate Abstraction (AI using SMT)

A Language of Arithmetic

Our language, just has numbers and multiplication

A Language of Arithmetic: Syntax

```
data AExp = N Int | AExp 'Mul' AExp
```

Example Expressions

```
N 7
```

```
N 7 'Mul' N (-3)
```

```
N 0 'Mul' N 7 'Mul' N (-3)
```

Concrete Semantics

To define the **(concrete)** or **exact** semantics, we need

```
type Value = Int
```

and an eval function that maps AExp to Value

```
eval :: AExp -> Value
```

```
eval (N n) = n
```

```
eval (Mul e1 e2) = mul (eval e1) (eval e2)
```

```
mul n m = n * m
```

Signs Abstraction

Suppose that we only care about the **sign** of the number.

Can define an *abstract* semantics

1. Abstract Values
2. Abstract Operators
3. Abstract Evaluators

Signs Abstraction: Abstract Values

Abstract values just preserve the **sign** of the number

```
data Value# = Neg | Zero | Pos
```

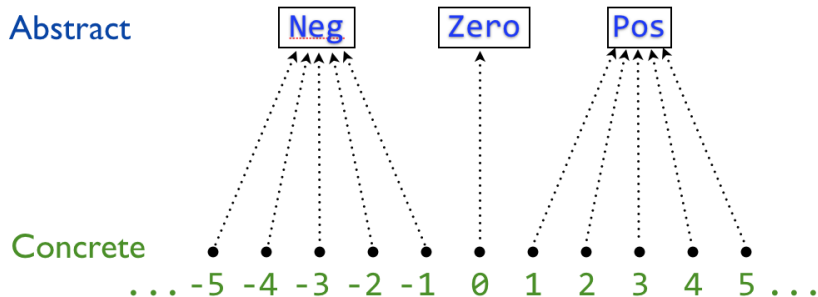


Figure: Abstract and Concrete Values

Signs Abstraction: Abstract Evaluator

Abstract evaluator just uses sign information

```
eval#           :: AExp -> Value#  
eval# | n > 0   = Pos  
      | n < 0   = Neg  
      | otherwise = Zero  
eval# (Mul e1 e2) = mul# (eval# e1) (eval# e2)
```

Signs Abstraction: Abstract Evaluator

mul# is the **abstract multiplication** operators

```
mul#           :: Value# -> Value# -> Value#
mul# Zero _    = Zero
mul# _ Zero    = Zero
mul# Pos Pos   = Pos
mul# Neg Neg   = Pos
mul# Pos Neg   = Neg
mul# Neg Pos   = Neg
```

Connecting the Concrete and Abstract Semantics

Theorem For all $e :: \text{AExp}$ we have

1. $(\text{eval } e) > 0$ iff $(\text{eval}\# e) = \text{Pos}$
2. $(\text{eval } e) < 0$ iff $(\text{eval}\# e) = \text{Neg}$
3. $(\text{eval } e) = 0$ iff $(\text{eval}\# e) = \text{Zero}$

Proof By induction on the structure of e

- ▶ Base Case: $e == N \ n$
- ▶ Ind. Step: Assume above for e_1 and e_2 prove for $\text{Mul } e_1 \ e_2$

Relating the Concrete and Abstract Semantics

Next, let us **generalize** what we did into a **framework**

- ▶ Allows us to use **different** Value#
- ▶ Allows us to get **connection theorem** by construction

Key Idea: Provide Abstraction Function α

We only have to provide connection between Value and Value#

```
alpha :: Value -> Value#
```

Key Idea: Provide Abstraction Function α

We only have to provide connection between Value and Value#

```
alpha :: Value -> Value#
```

For *signs* abstraction

```
alpha n | n > 0      = Pos  
        | n < 0      = Neg  
        | otherwise  = Zero
```

Key Idea: α induces Concretization γ

Given $\alpha :: \text{Value} \rightarrow \text{Value\#}$

we get for free a **concretization function**

```
gamma    :: Value# -> [Value]
gamma v# = [ v | (alpha v) == v# ]
```

For *signs* abstraction

```
gamma Pos  == [1,2..]
gamma Neg  == [-1,-2..]
gamma Zero == [0]
```

Key Idea: α induces Abstract Operator

Given $\alpha :: \text{Value} \rightarrow \text{Value\#}$

we get for free a **abstract operator**

$\text{op\# } x\# \ y\# = \alpha (\text{op } (\text{gamma } x\#) (\text{gamma } y\#))$

(actually, there is some *cheating* above... can you spot it?)

Key Idea: α induces Abstract Operator

Given $\alpha :: \text{Value} \rightarrow \text{Value\#}$

we get for free a **abstract operator**

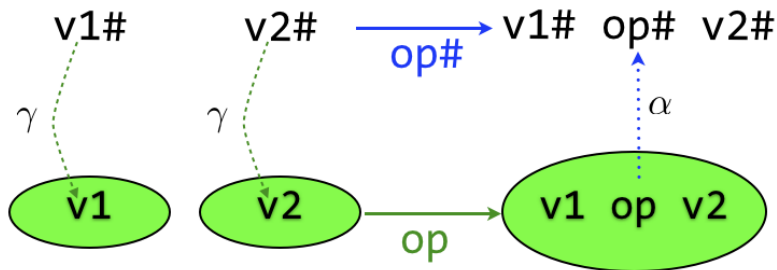


Figure: Abstract Operator

Key Idea: α induces Abstract Evaluator

Given $\alpha :: \text{Value} \rightarrow \text{Value\#}$

we get for free a **abstract evaluator**

$\text{eval\#} :: \text{AExp} \rightarrow \text{Value\#}$

$\text{eval\#} (\text{N } n) = (\alpha n)$

$\text{eval\#} (\text{Op } e1 \ e2) = \text{op\#} (\text{eval\# } e1) (\text{eval\# } e2)$

Key Idea: α induces Connection Theorem

Given $\alpha :: \text{Value} \rightarrow \text{Value\#}$

we get for free a **connection theorem**

Theorem For all $e :: \text{AExp}$ we have

1. $(\text{eval } e) \text{ in } \gamma \text{ (eval\# } e)$
2. $\alpha(\text{eval } e) = (\text{eval\# } e)$

Proof Exercise (same as before, but generalized)

Key Idea: α induces Connection Theorem

Given $\alpha :: \text{Value} \rightarrow \text{Value\#}$

we get for free a **connection theorem**

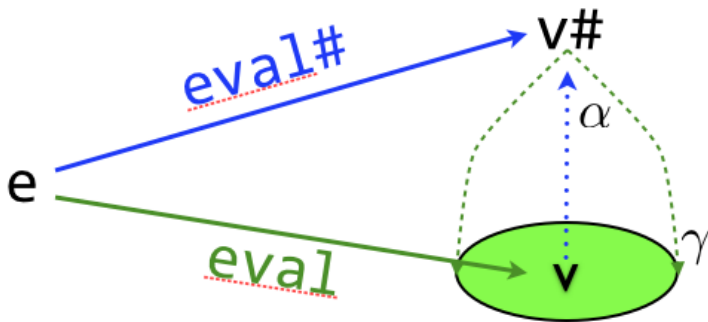


Figure: Connection Theorem

Our First Abstract Interpretation

Given: Language AExp and Concrete Semantics

```
data AExp
data Value

op    :: Value -> Value -> Value
eval :: AExp  -> Value
```

Given: Abstraction

```
data Value#
alpha :: Value -> Value#
```

Our First Abstract Interpretation

Obtain for free: Abstract Semantics

```
op#    :: Value# -> Value# -> Value#  
eval#  :: AExp  -> Value#
```

Obtain for free: Connection

Theorem: Abstract Semantics **approximates** Concrete Semantics

Our Second Abstract Interpretation

Let us extend AExp with new operators

- ▶ **Negation**
- ▶ Addition
- ▶ Division

AExp with Unary Negation

Extended Syntax

```
data AExp = ... | Neg AExp
```

Extended Concrete Semantics

```
eval (Neg e) = neg (eval e)
```

AExp with Unary Negation

Derive Abstract Operator

```
neg# :: Value# -> Value#  
neg# = alpha . neg . gamma
```

Which is equivalent to (if you do the math)

```
neg# Pos  = Neg  
neg# Zero = Zero  
neg# Neg  = Pos
```

Theorem holds as before!

Our Third Abstract Interpretation

Let us extend AExp with new operators

- ▶ Negation
- ▶ **Addition**
- ▶ Division

AExp with Addition

Extended Syntax

```
data AExp = ... | Add AExp AExp
```

Extended Concrete Semantics

```
eval (Add e1 e2) = plus (eval e1) (eval e2)
```

AExp with Addition

Derive Abstract Operator

```
plus#           :: Value# -> Value# -> Value#  
plus# v1# v2# = alpha (plus (gamma v1#) (gamma v2#))
```

That is,

```
plus# Zero v# = v#  
plus# Pos Pos = Pos  
plus# Neg Neg = Neg
```

but ...

```
plus# Pos Neg = ???  
plus# Neg Pos = ???
```


Problem: Require Better Abstract Values

Need new value to represent **union of** *positive* and *negative*

- ▶ T (read: **Top**), denotes **any** integer

Now, we can define

```
plus# Zero v# = v#  
plus# Top  v# = Top  
plus# Pos  Pos = Pos  
plus# Neg  Neg = Neg  
plus# Pos  Neg = Top  
plus# Neg  Pos = Top
```

Semantics is now Over-Approximate

Notice that now,

```
eval (N 1 'Add' N 2 'Add' (Neg 3)) == 0
```

```
eval# (N 1 'Add' N 2 'Add' (Neg 3)) == T
```

That is, we have **lost all information** about the sign!

- ▶ This is **good**
- ▶ Exact semantics **not computable** for real PL!

Our Fourth Abstract Interpretation

Let us extend AExp with new operators

- ▶ Negation
- ▶ Addition
- ▶ **Division**

AExp with Division

Extended Syntax

```
data AExp = ... | Div AExp AExp
```

Extended Concrete Semantics

```
eval (Add e1 e2) = div (eval e1) (eval e2)
```

AExp with Division: Abstract Semantics

How to define

`div# v# Zero = ?`

Need new value to represent **empty set** of integers

- ▶ `_|_` (read: **Bottom**), denotes **no** integer
- ▶ Abstract operator on `_|_` returns `_|_`
- ▶ Wait, this is getting rather *ad-hoc* ...
- ▶ Need more **structure** on `Value#`

Abstract Values Form Complete Partial Order

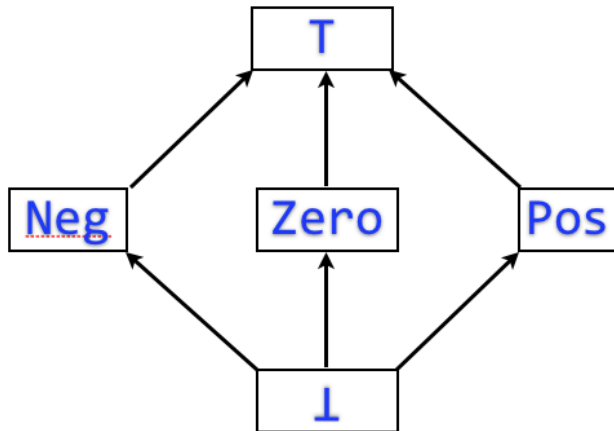


Figure: Value# Forms Complete Partial Order

Abstract Values Form Complete Partial Order

-- Partial Order

`(<=) :: Value# -> Value# -> Bool`

-- Greatest Lower Bound

`glb :: Value# -> Value# -> Value#`

-- Least Upper Bound

`lub :: Value# -> Value# -> Value#`

`leq v1# v2#` means `v1#` corresponds to **fewer** concrete values than `v2#`

Examples

- ▶ `leq _|_ Zero`
- ▶ `leq Pos Top`

Abstract Values: Least Upper Bound

```
forall v1# v2#.  v1# <= lub v1# v2#  
forall v1# v2#.  v2# <= lub v1# v2#  
forall v          .  if v1# <= v && v2# <= v then lub v1# v2# <
```

Examples

- ▶ (lub $_$ | $_$ Zero) == Zero
- ▶ (lub Neg Pos) == Top

~~~~~



## Abstract Values: Greatest Lower Bound

```
forall v1# v2#. glb v1# v2# <= v1#  
forall v1# v2#. glb v1# v2# <= v2#  
forall v      . if v <= v1# && v <= v2# then v <= glb v1#
```

### Examples

- ▶  $(\text{glb Pos Zero}) == \_|\_$
- ▶  $(\text{lub Top Pos}) == \text{Pos}$

## Key Idea: $\alpha$ and CPO induces Concretization $\gamma$

Given

- ▶  $\alpha :: \text{Value} \rightarrow \text{Value\#}$
- ▶  $\sqsubseteq :: \text{Value\#} \rightarrow \text{Value\#} \rightarrow \text{Bool}$

We get for free a **concretization function**

- ▶  $\gamma :: \text{Value\#} \rightarrow [\text{Value}]$

```
gamma      :: Value# -> [Value]
gamma v# = [ v | (alpha v) <= v# ]
```

**Theorem**  $v1\# \sqsubseteq v2\#$  iff  $(\text{gamma } v1\#) \subseteq (\text{gamma } v2\#)$

That is,

- ▶  $v1\# \sqsubseteq v2\#$  means  $v1\#$  represents **fewer** Value than  $v2\#$

## Key Idea: $\alpha$ and CPO induces $\alpha$ over [Value]

We can now lift  $\alpha$  to work on **sets** of values

```
alpha    :: [Value] -> Value#  
alpha vs = lub [alpha v | v <- vs]
```

For example

```
alpha [3, 4] == Pos  
alpha [-3, 4] == Top  
alpha [0]    == Zero
```

## Key Idea: $\alpha + \text{CPO}$ induces Abstract Operator

Given

- ▶  $\alpha :: \text{Value} \rightarrow \text{Value\#}$
- ▶  $\sqsubseteq :: \text{Value\#} \rightarrow \text{Value\#} \rightarrow \text{Bool}$

We get for free a **abstract operator**

$\text{op\# } x\# \ y\# = \text{alpha } [\text{op } x \ y \mid x \leftarrow \text{gamma } x\#, \ y \leftarrow \text{gamma } y\#]$

i.e., lub of results of point-wise concrete operator (*no cheating!*)

For example

$\text{plus\# } \text{Pos } \text{Neg}$

$\text{== alpha } [x + y \mid x \leftarrow \text{gamma } \text{Pos}, \ y \leftarrow \text{gamma } \text{Neg}]$

$\text{== alpha } [x + y \mid x \leftarrow [1, 2..], \ y \leftarrow [-1, -2..]]$

$\text{== alpha } [0, 1, -1, 2, -2..]$

$\text{== Top}$

## Key Idea: $\alpha + \text{CPO}$ induces Abstract Operator

Given  $\alpha :: \text{Value} \rightarrow \text{Value\#}$

we get for free a **abstract operator**

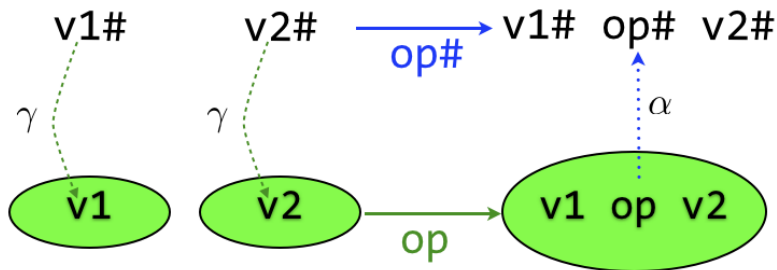


Figure: Abstract Operator

## Key Idea: $\alpha$ + CPO induces Evaluator

As before, we get for free a **abstract evaluator**

```
eval#           :: AExp -> Value#  
eval# (N n)     = (alpha n)  
eval# (Op e1 e2) = op# (eval# e1) (eval# e2)
```

## Key Idea: $\alpha$ + CPO induces Evaluator

And, more importantly, the semantics connection

**Theorem** For all  $e :: \text{AExp}$  we have

1.  $(\text{eval } e) \in \text{gamma } (\text{eval\# } e)$
2.  $\alpha (\text{eval } e) \sqsubseteq (\text{eval\# } e)$

### Over-Approximation

In bare AExp we had **exact** abstract semantics

- ▶  $\alpha (\text{eval } e) = (\text{eval\# } e)$

Now, we have **over-approximate** abstract semantics

- ▶  $\alpha (\text{eval } e) \sqsubseteq (\text{eval\# } e)$

That is, information is lost.

## Next Time: Abstract Interpretation For IMP

So far, abstracted values for AExp

- ▶ Concrete Value = Int
- ▶ Abstract Value# = Signs

Next time: apply these ideas to IMP

- ▶ Concrete Value = State at program points
- ▶ Abstract Value# = ???

Abstract Semantics yields **loop invariants**