
All the world is
an abstract interpretation
(of all the world)

David Schmidt
Kansas State University

An **abstraction** is a property from some domain



→ ***brown*** (color)

An abstraction is a property (cont.)



→ *brown*
(color)

→ *heavy*
(weight)

An abstraction is a property (cont.)



→ *brown* (color)

→ *heavy* (weight)



→ *4000..6000 kg.*

An abstraction is a property (concl.)



→ *elephant* (species)

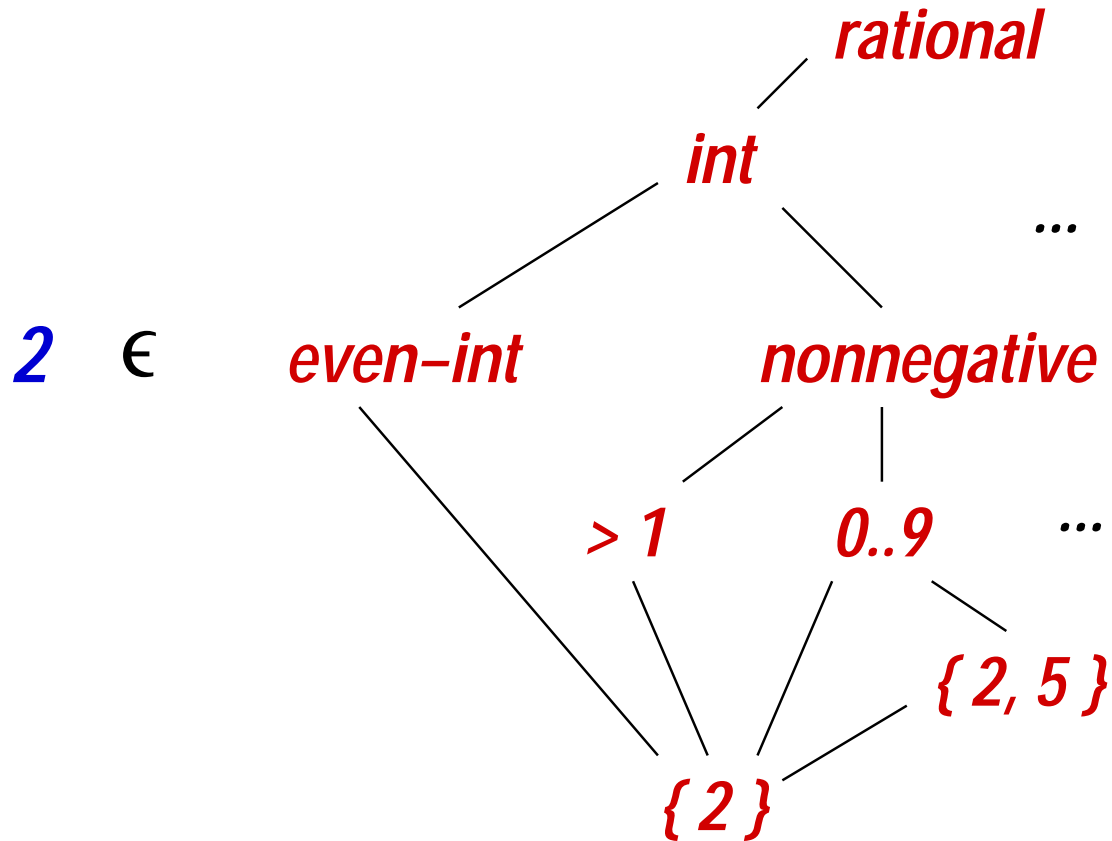
→ *brown* (color)

→ *heavy* (weight)



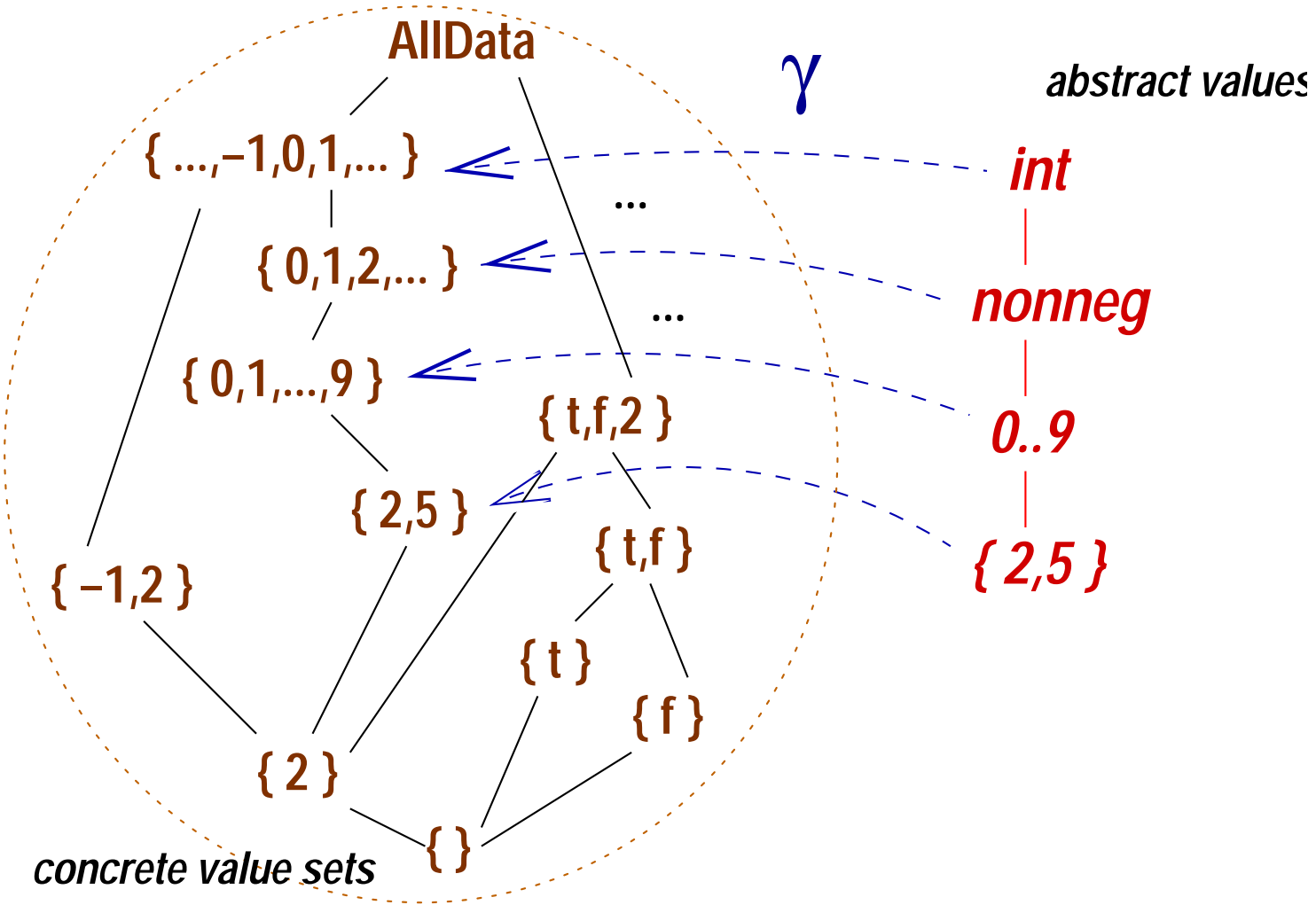
→ *4000..6000 kg.*

In computing, we use value abstractions



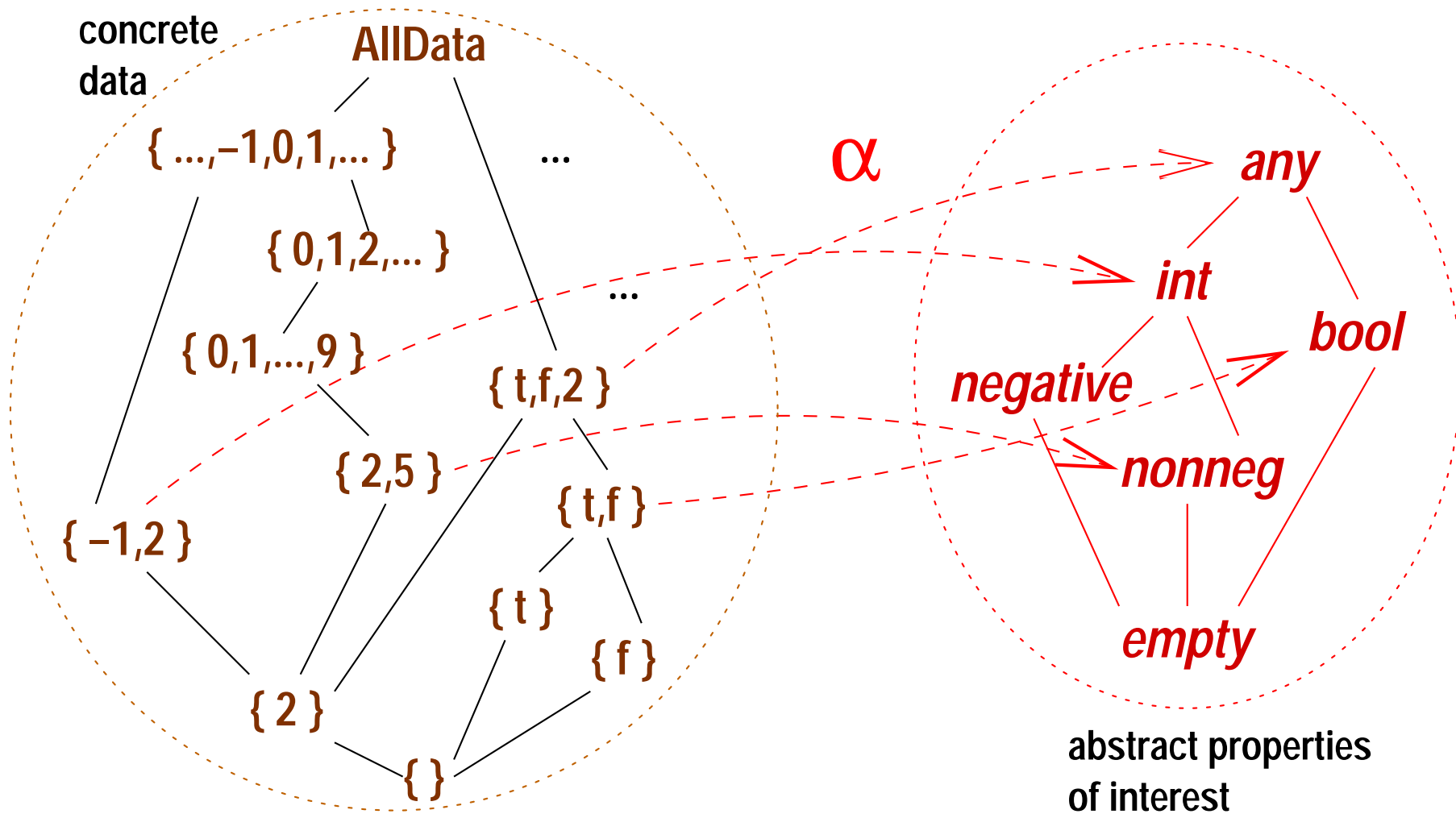
All the properties listed on the right are abstractions of **2**; the upwards lines denote \sqsubseteq , a loss of precision.

Abstract values name sets of concrete values



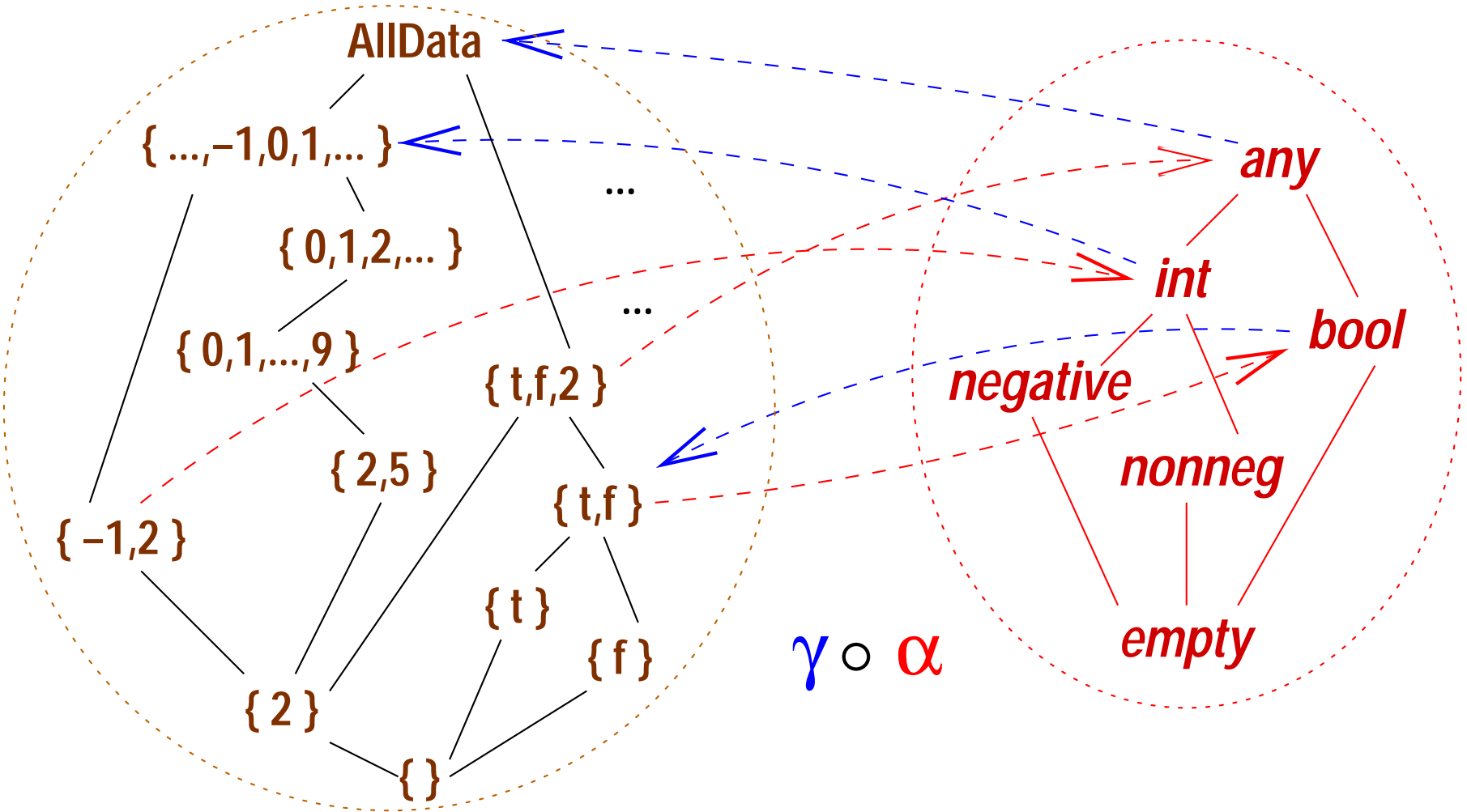
Function γ maps each abstract value to the set of concrete values it represents.

Sets of concrete values are abstracted imprecisely



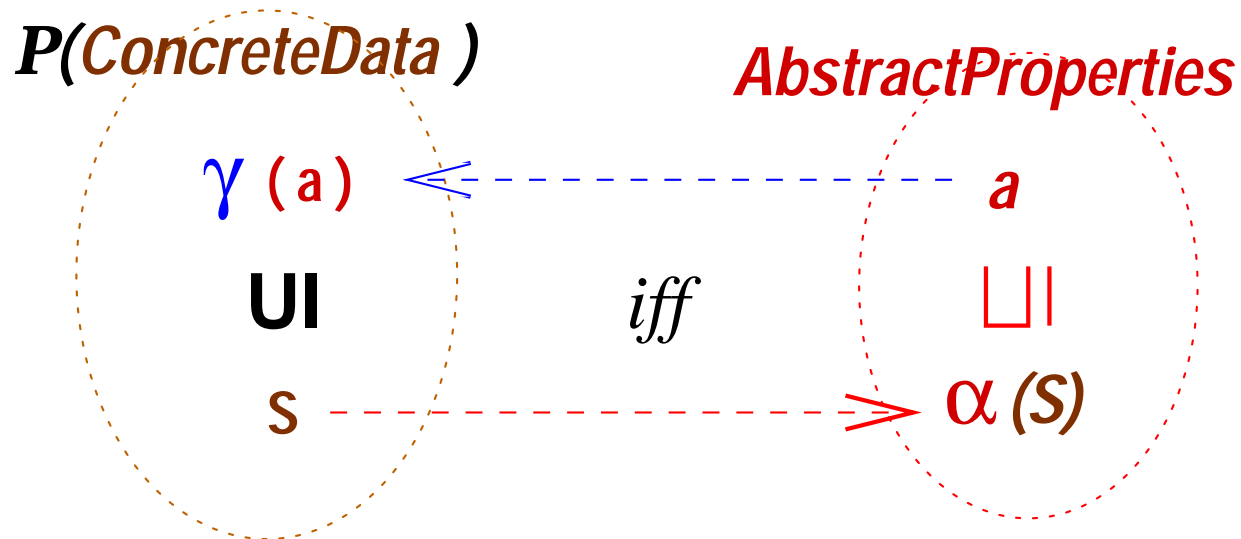
Function α maps each set to the abstract value that best describes it.

Abstraction followed by concretization demonstrates that α is sound but not exact



Nonetheless, the α given here is as precise as it possibly can be, given the abstract value domain and γ .

A Galois connection formalizes the situation



That is, for all $S \in \mathcal{P}(\text{ConcreteData})$, $a \in \text{AbstractProperties}$,

$$S \subseteq \gamma(a) \text{ iff } \alpha(S) \sqsubseteq a$$

When α and γ are monotone, this is equivalent to

$$S \subseteq \gamma \circ \alpha(S) \quad \text{and} \quad \alpha \circ \gamma(a) \sqsubseteq a$$

For practical reasons, the second inequality is usually restricted to $\alpha \circ \gamma(a) = a$, meaning that all abstract properties are “exact.”

Perhaps the oldest application of abstract interpretation is to data-type checking

```
int x;  
int[] a = new int[10];  
...  
a[0] = x + 2; // Whatever x's run-time value might  
...         // be, we know it is an int.  
a[1] = (!x); // Erroneous --- an int cannot be  
             // negated, nor can a bool be  
             // saved in an int cell.
```

But compilers struggle with imprecise abstractions

```
int x;  
int[] a = new int[10];  
...      // Because x's value is described  
a[2 * x] = 3; // imprecisely, we cannot decide  
           // whether 2 * x falls in the  
           // interval, [0,9].
```

We might address array-indexing calculation by

1. making the abstraction more precise, e.g., declaring x with the abstract value (“data type”) $[0, 9]$;
2. computing a “symbolic execution” of the program with the abstract values

These extensions underlie data-flow analyses and many sophisticated program analysis techniques.

A starting point: Trace-based operational semantics

```
 $p_0$  : while isEven(x) {  
     $p_1$  : x = x div 2;  
}  
 $p_2$  : x = 4 * x;  
 $p_3$  : exit
```

The operational semantics updates a program-point, storage-cell pair, pp, x , using these four transition rules:

$$\begin{array}{ll} p_0, 2n \longrightarrow p_1, 2n & p_1, n \longrightarrow p_0, n/2 \\ p_0, 2n + 1 \longrightarrow p_2, 2n + 1 & p_2, n \longrightarrow p_3, 4n \end{array}$$

A program's operational semantics is written as a trace:

$$p_0, 12 \longrightarrow p_1, 12 \longrightarrow p_0, 6 \longrightarrow p_1, 6 \longrightarrow p_0, 3 \longrightarrow p_2, 3 \longrightarrow p_3, 12$$

We can abstractly interpret, say, for polarity

```
p0 : while isEven(x) {  
    p1 : x = x div 2;  
}  
p2 : x = 4 * x;  
p3 : exit
```

*p*_{0, even} \longrightarrow *p*_{1, even}

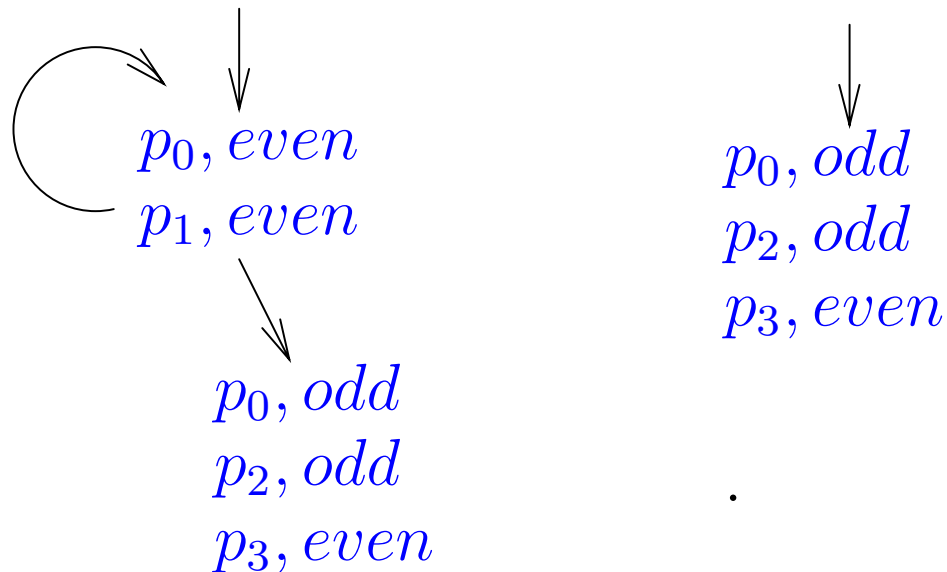
*p*_{0, odd} \longrightarrow *p*_{2, odd}

*p*_{1, even} \longrightarrow *p*_{0, even}

*p*_{1, even} \longrightarrow *p*_{0, odd}

*p*_{2, a} \longrightarrow *p*_{3, even}

Two trace trees cover the full range of inputs:

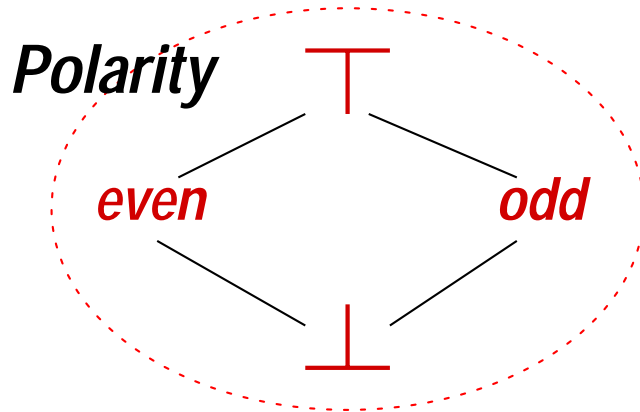


We conclude that

- ◆ if the program terminates, x is even-valued
- ◆ if the input is odd-valued, the loop will not be entered

Due to the loss of precision, we have not decided termination for even-valued inputs.

The underlying abstract interpretation



$$\gamma : \text{Polarity} \rightarrow \mathcal{P}(\text{Int})$$

$$\gamma(\text{even}) = \{\dots, -2, 0, 2, \dots\}$$

$$\gamma(\text{odd}) = \{\dots, -1, 1, 3, \dots\}$$

$$\gamma(\top) = \text{Int}, \quad \gamma(\perp) = \{\}$$

$$\alpha : \mathcal{P}(\text{Int}) \rightarrow \text{Polarity}$$

$$\alpha(S) = \sqcup \{\beta(v) \mid v \in S\}, \text{ where } \beta(2n) = \text{even} \text{ and } \beta(2n + 1) = \text{odd}$$

The abstract transition rules are synthesized from the originals:

$$p_i, a \longrightarrow p_j, \alpha(v'), \text{ if } v \in \gamma(a) \text{ and } p_i, v \longrightarrow p_j, v'$$

This recipe ensures that every transition in the original, “concrete” semantics is simulated by one the abstract semantics.

To elaborate, remember that an abstract state, p_i, a , represents (abstracts) the set of concrete states,

$$\gamma_{State}(p_i, a) = \{p_i, c \mid c \in \gamma(a)\}$$

So, if some p_i, c in the above set can transit to p_j, c' , then its abstraction must make a similar move:

$$p_i, c \longrightarrow p_j, c' \text{ implies } p_i, a \longrightarrow p_j, a', \text{ where } p_j, c' \in \gamma_{State}(p_j, a').$$

Thus, the abstract semantics simulates all computations of the concrete semantics (and due to imprecision, produces more computations than are concretely possible).

Given a Galois connection, α, γ , we synthesize the most precise abstract semantics that simulates the concrete one as defined on the previous slide.

Abstract interpretation is flexible

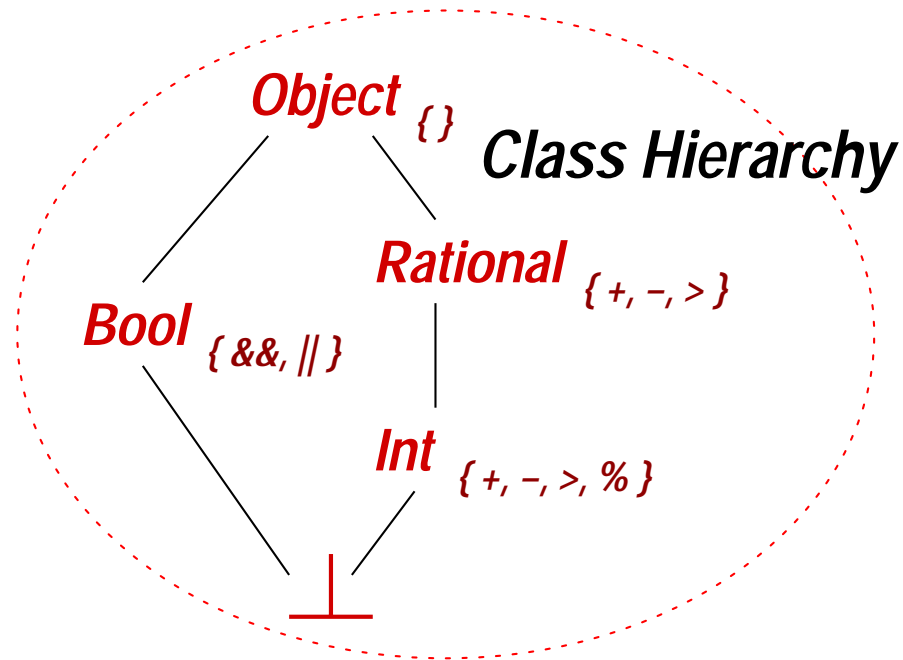
We will apply abstract interpretation to

- ◆ data-type inference
- ◆ code improvement
- ◆ debugging
- ◆ assertion synthesis and program proving
- ◆ model-checking temporal logic formulas

Data-type compatibility inference

```

p0 : x = 4;
p1 : while ... {
      p2 : x = (x > 0)
    }
p3 : x = x % 2;
p4 : exit
  
```



$p_0, \tau \longrightarrow p_1, Int$

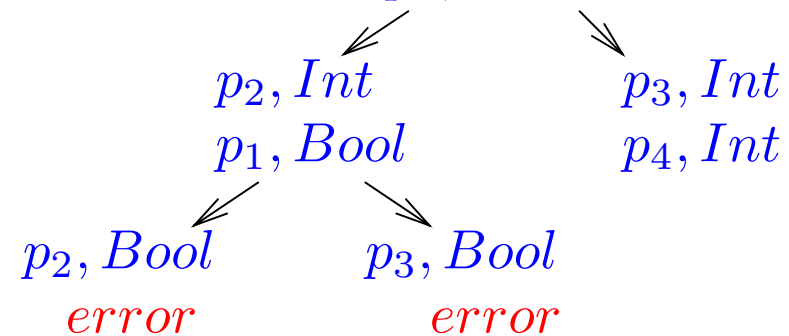
$p_1, \tau \longrightarrow p_2, \tau$

$p_1, \tau \longrightarrow p_3, \tau$

$p_2, \tau \longrightarrow p_1, Bool$, if $\tau \sqsubseteq Rational$

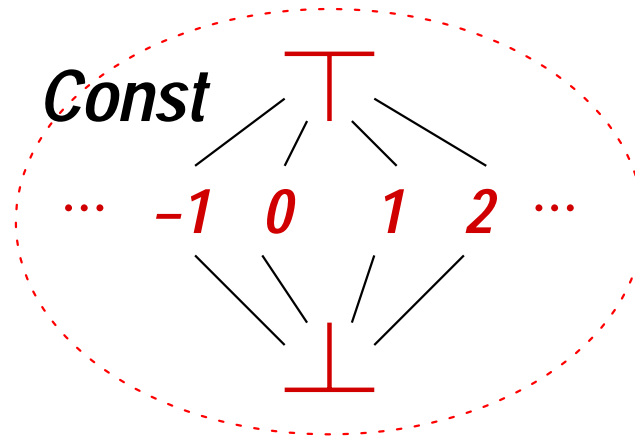
$p_3, Int \longrightarrow p_4, Int$

Abstract trace:



Constant propagation analysis

p_0 : $x = 1; y = 2;$
 p_1 : **while** ($x < y + z$)
 p_2 : $x = x + 1;$
 }
 p_3 : *exit*



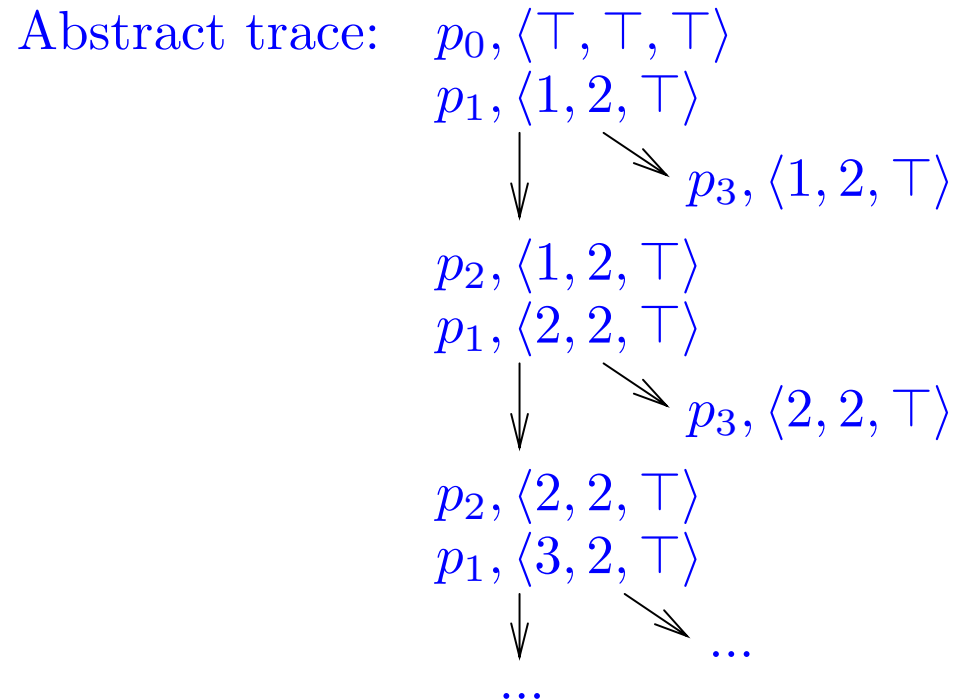
where $m + n$ is interpreted

$$k_1 + k_2 \longrightarrow \text{sum}(k_1, k_2),$$

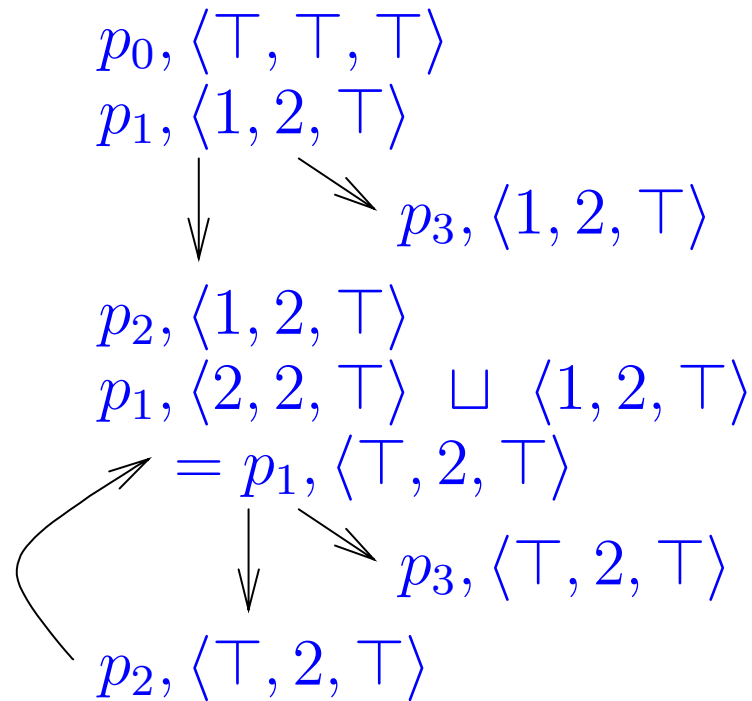
$$\top \neq k_i \neq \perp, i \in 1..2$$

$$\top + k \longrightarrow \top$$

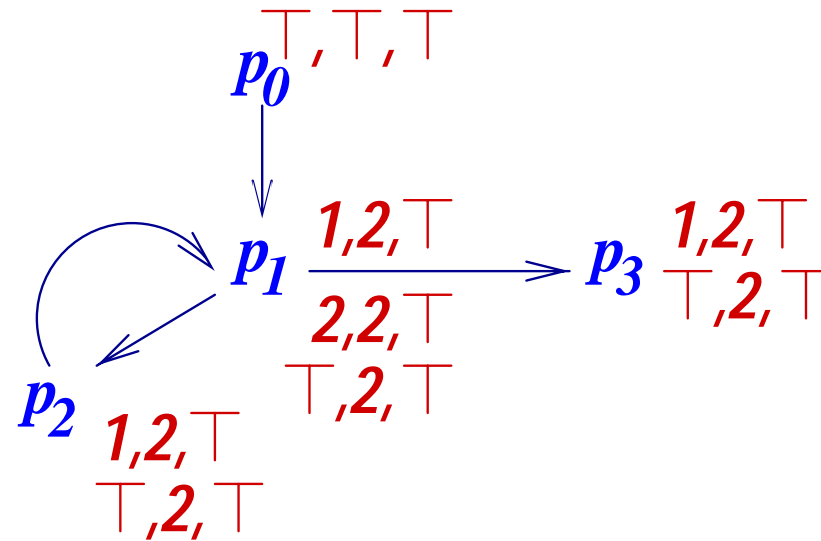
$$k + \top \longrightarrow \top$$



An **acceleration** is needed for £nite convergence



Drawn as a data-flow analysis:



The analysis tells us to replace y at p_1 by 2:

```


$p_0$  :  $x = 1; y = 2;$   

 $p_1$  : while ( $x < 2 + z$ )  

         $p_2$  :  $x = x + 1;$   

        }  

 $p_3$  : exit


```

Array bounds (pre)checking

Integer variables receive values from the **interval domain**,

$$I = \{[i, j] \mid i, j \in \text{Int} \cup \{-\infty, +\infty\}\}.$$

We define $[a, b] \sqcup [a', b'] = [\min(a, a'), \max(b, b')]$.

```
int a = new int[10];  
i = 0;  $\Leftarrow$   $i = [0, 0]$   
while (i < 10) {  
    ... a[i] ...  $\Leftarrow$   $p_1$   $i = [0, 0] \sqcap [-\infty, 9] = [0, 0]$   
    i = i + 1;  $\Leftarrow$   $i = ([0, 0] \sqcup [1, 1]) \sqcap [-\infty, 9] = [0, 1]$   
}  $\Leftarrow$   $p_2$   $i = [1, 1]$   
 $\Leftarrow$   $i = [1, 1] \sqcup [2, 2] = [1, 2]$ 
```

This generates an infinite sequence; we must **accelerate** with ∇ :

$$p_1: i = [0, 0] \sqcap [-\infty, 9] = [0, 0]$$

$$\begin{aligned} i &= ([0, 0] \nabla [1, 1]) \sqcap [-\infty, 9] \\ &= [0, +\infty] \sqcap [-\infty, 9] = [0, 9] \end{aligned}$$

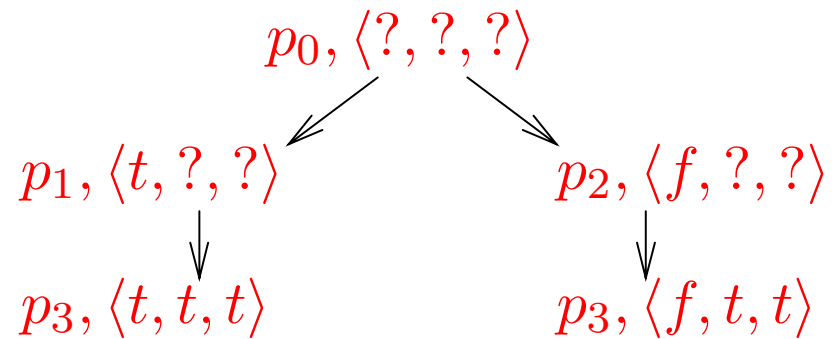
$$p_2: i = [1, 1]$$

$$i = [1, 1] \nabla [1, 10] = [1, +\infty]$$

Program verification via predicate abstraction

We wish to prove that $z \geq x \wedge z \geq y$ at p_3 :

```
 $p_0$  : if  $x < y$   
 $p_1$  :   then  $z = y$   
 $p_2$  :   else  $z = x$   
 $p_3$  : exit
```



$$\phi_1 = x < y$$

We chose three predicates, $\phi_2 = z \geq x$

$$\phi_3 = z \geq y$$

and computed their values at the program's points. The predicates' values come from the domain, $\{t, f, ?\}$. (Read ? as $t \vee f$.)

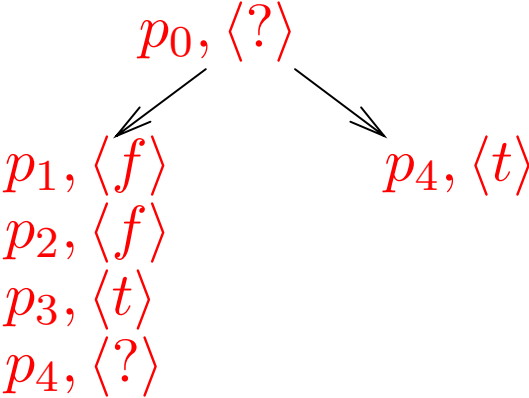
At all occurrences of p_3 in the abstract trace, $\phi_2 \wedge \phi_3$ holds.

When a goal is undecided, refinement is necessary

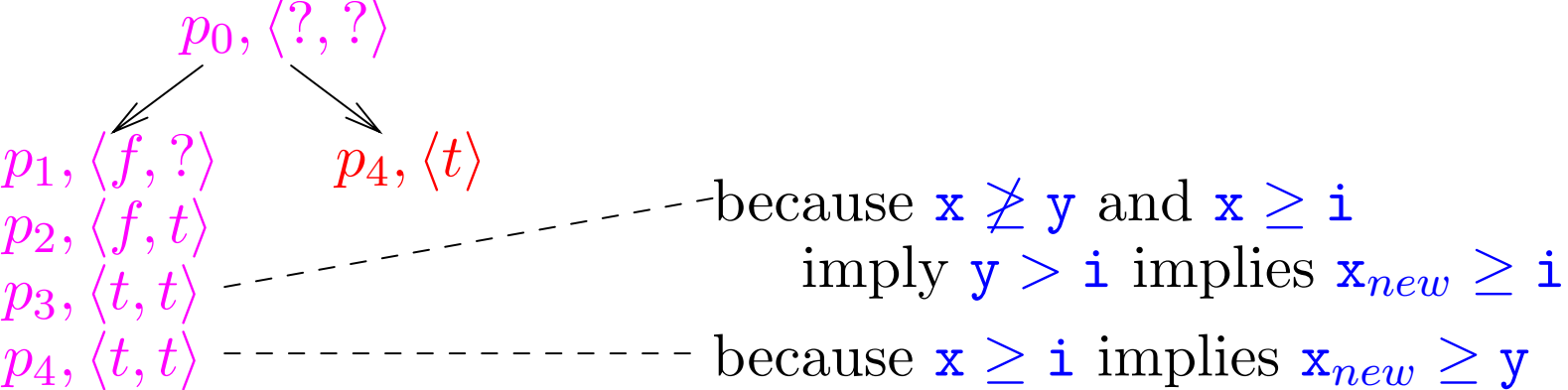
Prove $\phi_0 \equiv x \geq y$ at p_4 :

```

p0 : if !(x >= y)
p1 : then { i = x;
          p2 : x = y;
          p3 : y = i;
p4 : }
    
```



To decide the goal, we must refine the state by adding a needed auxiliary predicate: $wp(y = i, x \geq y) = (x \geq i) \equiv \phi_1$.



But incremental predicate re£nement cannot synthesize many interesting loop invariants. For this example:

```
 $p_0$  :  $i = n$ ;  $x = 0$ ;  
 $p_1$  : while  $i \neq 0$  {  
     $p_2$  :  $x = x + 1$ ;  $i = i - 1$ ;  
}  
 $p_3$  : goal:  $x = n$ 
```

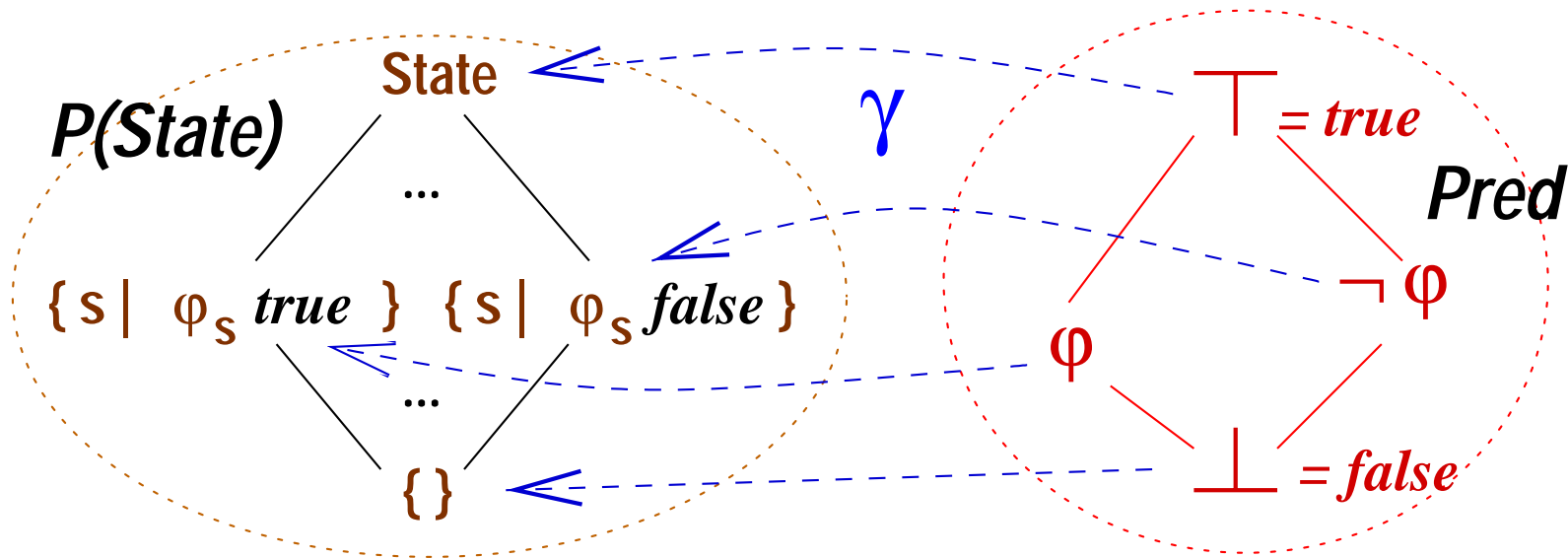
We £nd that the initial predicate set, $P_0 \equiv \{i = 0, x = n\}$, does not validate the loop body.

The £rst re£nement suggests we add $P_1 \equiv \{i = 1, x = n - 1\}$ to the program state, but this fails to validate a loop that iterates more than once.

Re£nement stage j adds predicates $P_j \equiv \{i = j, x = n - j\}$; the re£nement process continues forever!

The loop invariant is $x = n - i$:-)

Predicates live in an abstract domain with a negation operation



The domain is a **boolean lattice**; a stronger condition is **boolean algebra** (where the distributive laws hold).

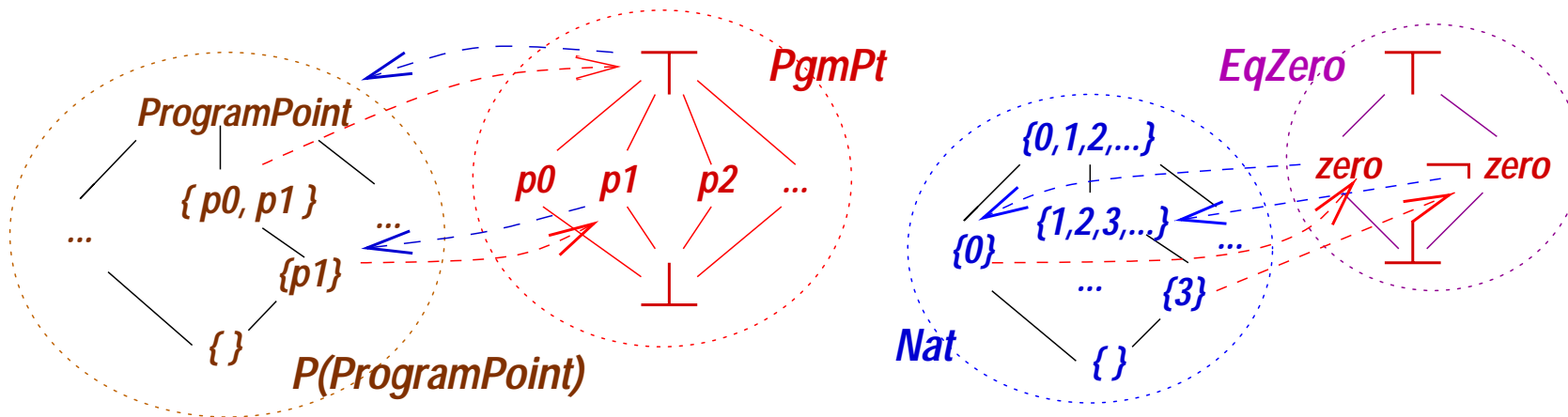
If we check for **correctness criterion**, $S_c \subseteq \mathcal{P}(\text{State})$, then $\alpha(S_c)$ must be **exact** — $\gamma \circ \alpha(S_c) = S_c$ — to use $\phi_c = \alpha(S_c)$ in the abstract analysis. (Otherwise, we might incorrectly infer, from p', ϕ_c , that all such p', s' , where $s' \in \gamma(\phi_c)$, are “correct.”)

Every abstract domain defines a “logic”

For abstract domain A , $a \in A$ is a “property/predicate,” and $\gamma(a)$ defines those concrete states that make a “true”:

$$s \text{ has } a, \text{ written } s \models_A a, \text{ iff } \alpha\{s\} \sqsubseteq a$$

Example: for concrete states, $ProgramPoint \times Nat$, and its abstraction, $PgmPt \times EqZero$:

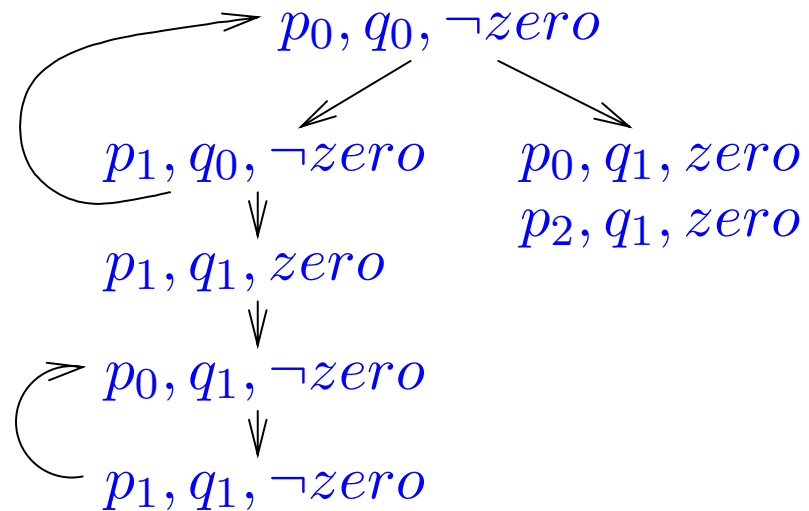


We have these facts: $p_1 \models_{PgmPt} p_1$ and $3 \models_{EqZero} \neg zero$
 therefore, $p_1, 3 \models_{PP \times EqZero} p_1, \neg zero$

Abstract traces are examples of model checks

p_0 : while $x > 0$ {
 p_1 : use resource
 $x = x + 1$; }
 p_2 : sleep

q_0 : $x = 0$;
 q_1 : use resource



Say that p_1, q_1 is an error state; is there a path from the start to it:

$$p_0, q_0, k \models_{LTL(PgmPt \times EqZero)} F(p_1, q_1, \top) ?$$

Say that p_2 is a rest state; for all states reached from the start, can we progress to it:

$$p_0, q_0, k \models_{LTL(PgmPt \times EqZero)} GF(p_2, \top, \top) ?$$

The logical operators, G and F , describe reachability properties in the temporal logic, **LTL**.

A state, s_0 , names the set of traces that begin with it. An LTL property, ϕ , describes a pattern of states in a trace.

$s_0 \models \phi$ means that all traces, $s_0 \rightarrow s_1 \rightarrow \dots$, contain pattern ϕ .

MiniLTL: $\phi ::= a \mid G\phi \mid F\phi$ **Semantics:** $\llbracket \phi \rrbracket \subseteq \mathcal{P}(Trace)$

$$\llbracket a \rrbracket = \{ \pi \mid \pi_0 \models_A a \}$$

$$\llbracket G\phi \rrbracket = \{ \pi \mid \forall i \geq 0, \pi \downarrow i \in \llbracket \phi \rrbracket \}$$

$$\llbracket F\phi \rrbracket = \{ \pi \mid \exists i \geq 0, \pi \downarrow i \in \llbracket \phi \rrbracket \}$$

where, for $\pi = s_0 \rightarrow s_1 \rightarrow \dots$, let $\pi_0 = s_0$ and $\pi \downarrow i = s_i \rightarrow s_{i+1} \rightarrow \dots$.

There is a Galois connection, $(\mathcal{P}(Trace), \subseteq) \leftrightarrow (\mathcal{P}(\text{MiniLTL}), \supseteq)$,

where $\sqcup = \cap = \wedge$ in $\mathcal{P}(\text{MiniLTL})$:

$$\gamma(P) = \bigcap \{ \llbracket \phi \rrbracket \mid \phi \in P \}$$

$$\beta(\pi) = \{ \phi \mid \pi \in \gamma\{\phi\} \}$$

$$\alpha(S) = \bigcap \{ \beta(\pi) \mid \pi \in S \}$$

But this is just the beginning of a long story about the relationship of abstract interpretation to temporal-logic model checking!

Every concrete value is the conjunction of its abstractions (its “abstract-interpretation DNA”)



$$= \text{elephant}_{\text{species}} \wedge \text{brown}_{\text{color}} \wedge \text{heavy}_{\text{weight}} \\ \wedge 4000..6000\text{kg}_{\text{weight}} \wedge \dots$$

There is even a pattern of Galois connection for this:

$$\gamma : \text{AllPossibleProperties} \rightarrow \mathcal{P}(\text{RealWorldObjects})$$

$$\gamma(p) = \{c \in \text{RealWorldObjects} \mid c \text{ has property } p\}$$

$$\beta : \text{RealWorldObjects} \rightarrow \text{AllPossibleProperties}$$

$$\beta(c) = \sqcap \{p \in \text{AllPossibleProperties} \mid c \in \gamma(p)\}$$

$$\alpha : \mathcal{P}(\text{RealWorldObjects}) \rightarrow \text{AllPossibleProperties}$$

$$\alpha(S) = \sqcup \{\beta(s) \mid s \in S\}$$

Some references

- ◆ The papers of Patrick and Radhia Cousot (www.di.ens.fr/~cousot), but especially
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- ◆ Neil Jones and Flemming Nielson. Abstract Interpretation: a Semantics-Based Tool for Program Analysis. In [Handbook of Logic in Computer Science, Vol. 4](#), Oxford University Press, 1994.
- ◆ A few of my papers, found at www.cis.ksu.edu/~schmidt/papers:
 1. Trace-Based Abstract Interpretation of Operational Semantics. J. Lisp and Symbolic Computation 10-3 (1998).
 2. Data-flow analysis is model checking of abstract interpretations. ACM POPL 1998.
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Interpretation. Workshop on Structure Preserving Relations, Amagasaaki, Japan, 2001.

4. Structure-preserving binary relations for program abstraction. In *The Essence of Computation: Complexity, Analysis, Transformation*. Springer LNCS 2566, 2002.