
Introduction to Abstraction and Static Analysis

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Outline

1. What is abstraction?
2. Abstraction and concretization:
Galois-connection-based abstract interpretation
3. Examples of static analyses
4. Logics and static analysis

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)

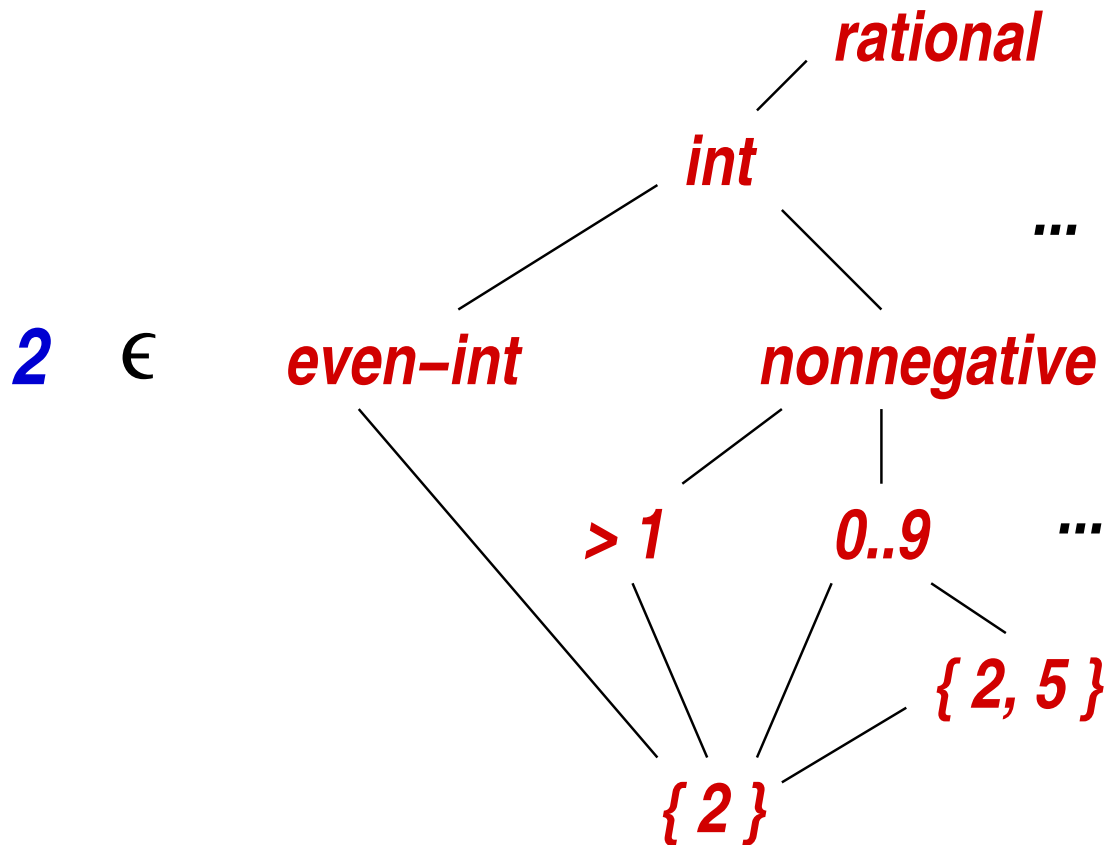
→ **brown** (color)

→ **heavy** (weight)



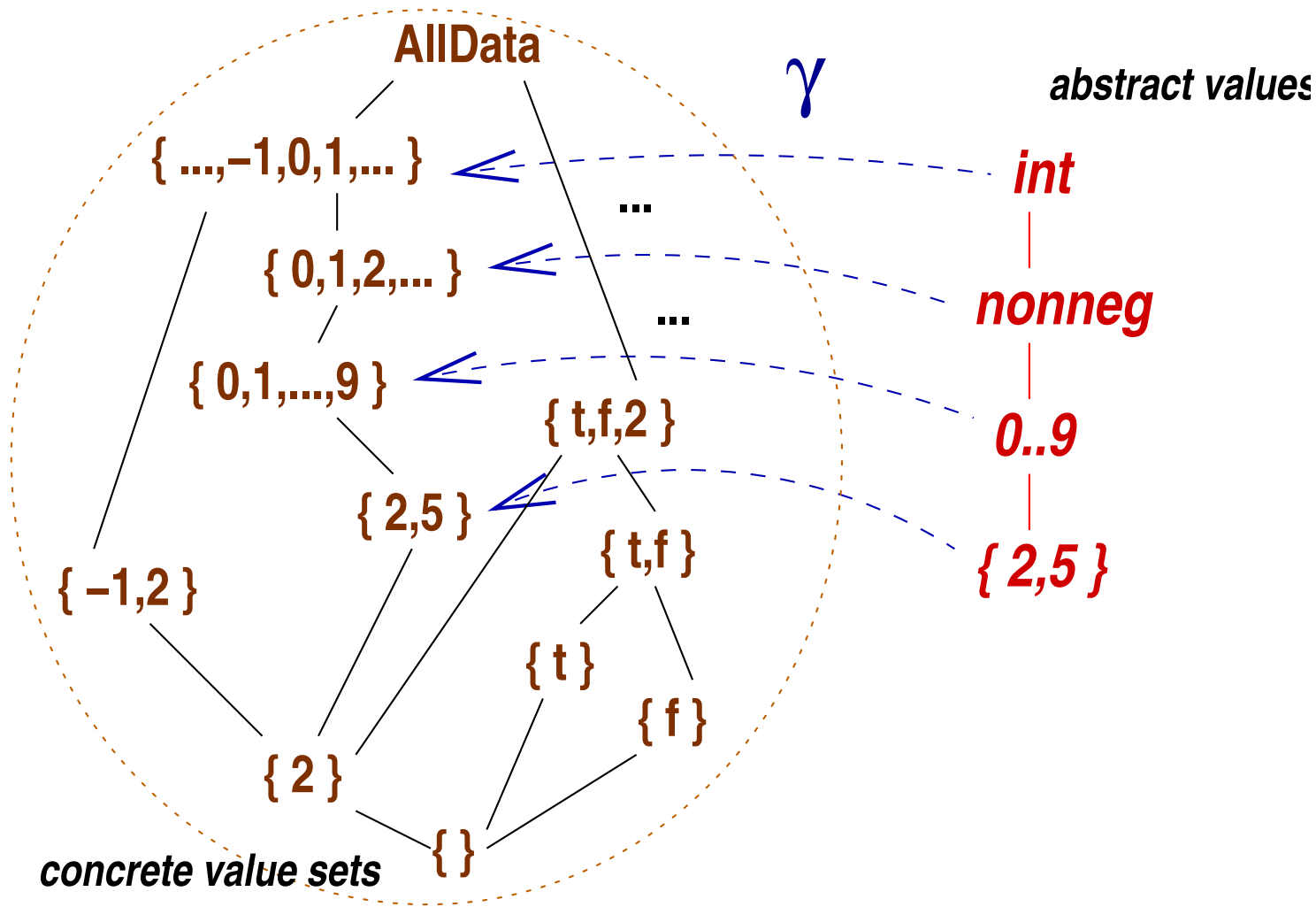
→ **4000..6000 kg.**

Value abstractions are classic to computing



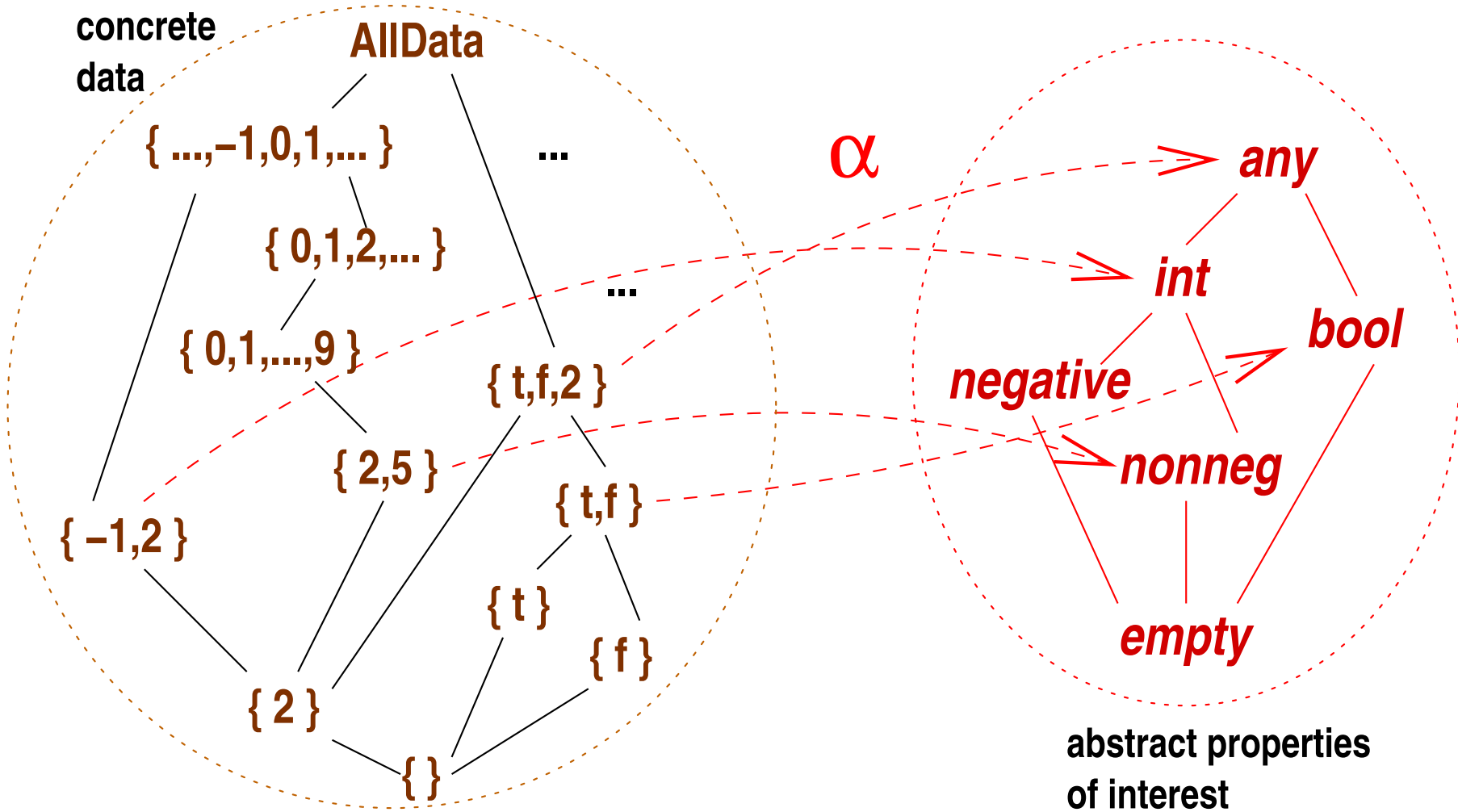
All the properties listed on the right are abstractions of **2**; the upwards lines denote \sqsubseteq , a loss of precision.

Abstract values name sets of concrete values



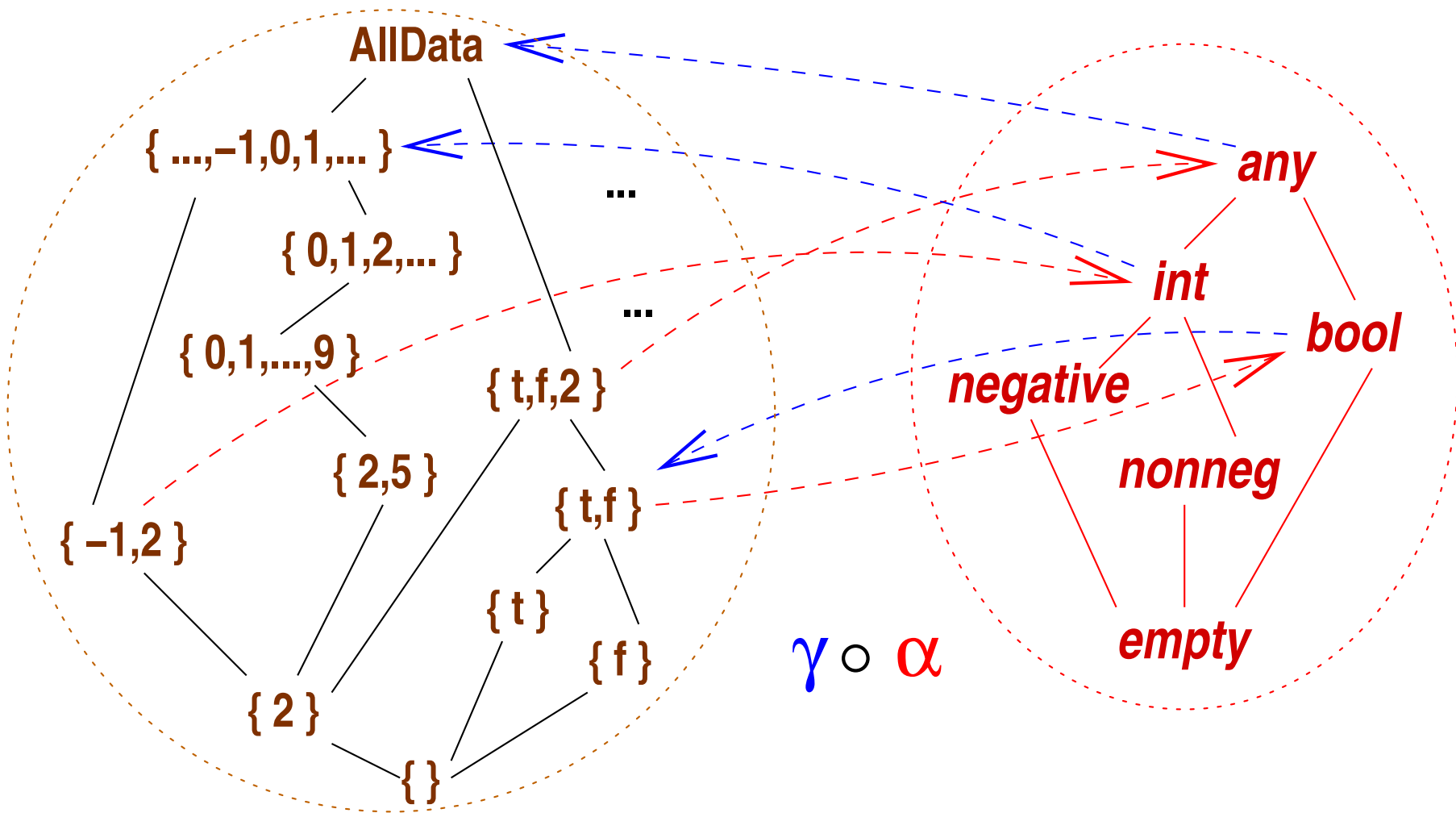
Function γ maps each abstract value to the set of concrete values it represents.

Sets of concrete values are abstracted imprecisely



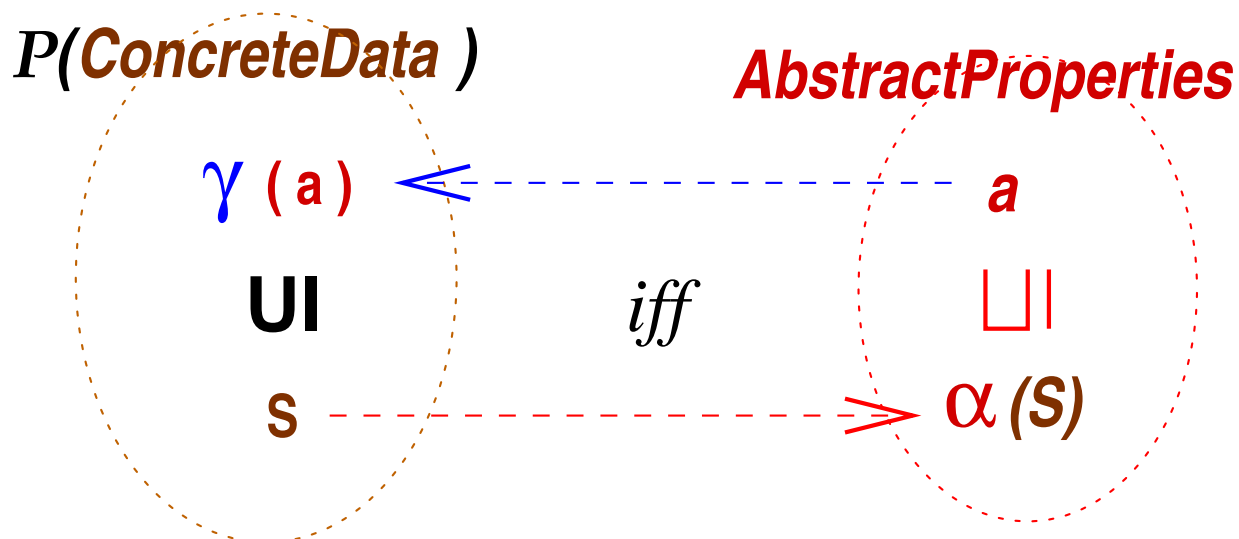
Function α maps each set to the abstract value that best describes it.

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 employ 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 parity

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

*p*₀, even \longrightarrow *p*₁, even

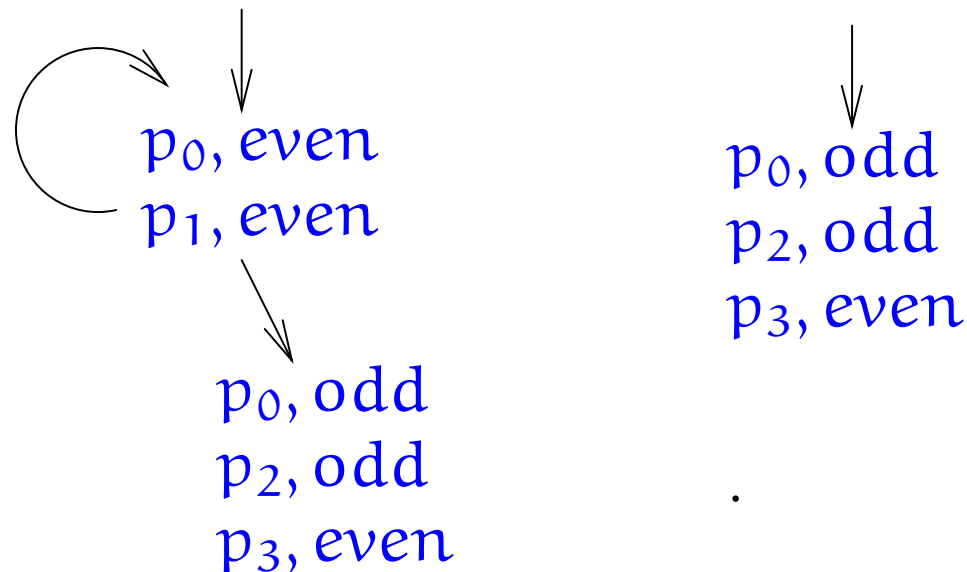
*p*₀, odd \longrightarrow *p*₂, odd

*p*₁, even \longrightarrow *p*₀, even

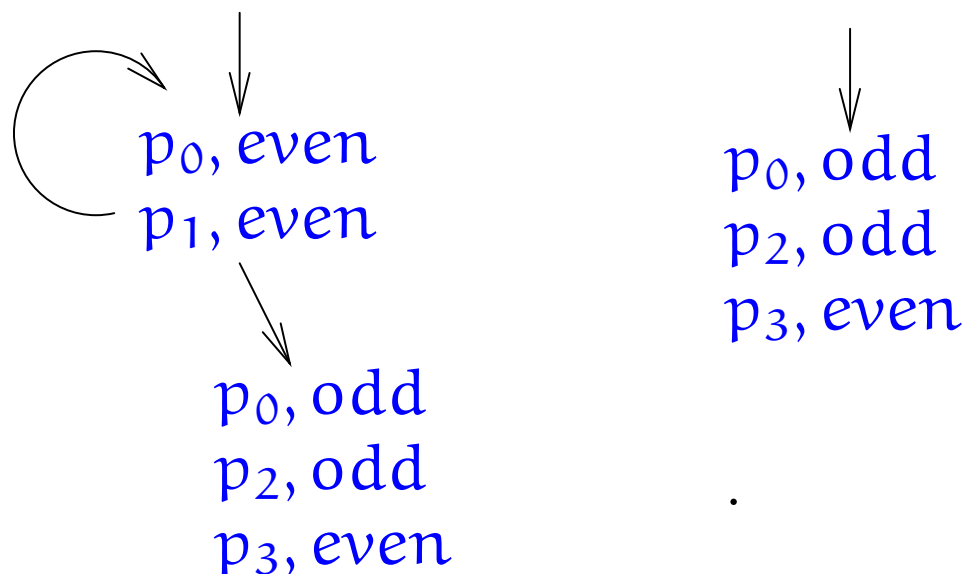
*p*₁, even \longrightarrow *p*₀, odd

*p*₂, a \longrightarrow *p*₃, even

Two trace trees cover the full range of inputs:



The interpretation of the program's semantics with the abstract values is an *abstract interpretation*:

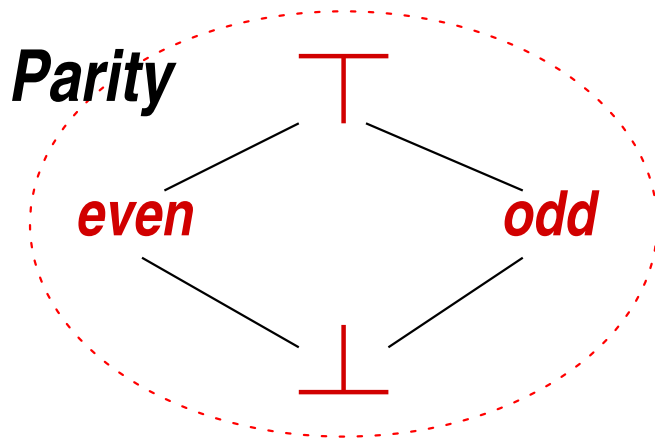


We conclude that

- ◆ if the program terminates, x is even-valued
- ◆ if the input is odd-valued, the loop body, p_1 , will not be entered

Due to the loss of precision, we can not decide termination for almost all the even-valued inputs. (Indeed, only 0 causes nontermination.)

The underlying abstract-interpretation semantics



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

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

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

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

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

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

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 in the abstract semantics.

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

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

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

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

Thus, the abstract semantics simulates all computation traces of the concrete semantics (and due to imprecision, produces more traces 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 underlies most *static analyses*

A *static analysis* of a program is a *sound, finite, and approximate* calculation of the program's executions. The trace trees we just generated for the loop program is an example of a static analysis.

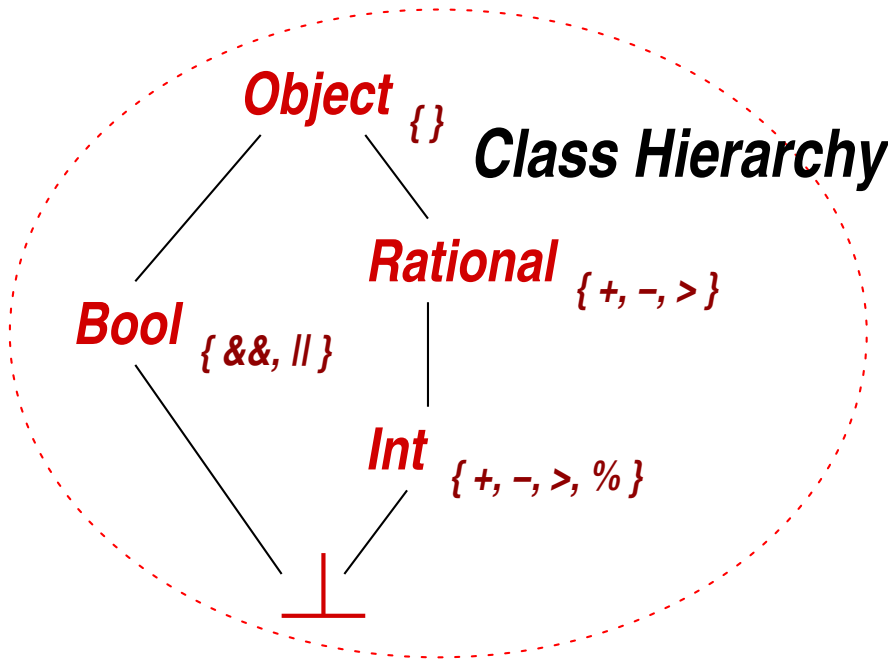
We will survey static analyses for

- ◆ 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
  
```



$p_0, \tau \longrightarrow p_1, \text{Int}$

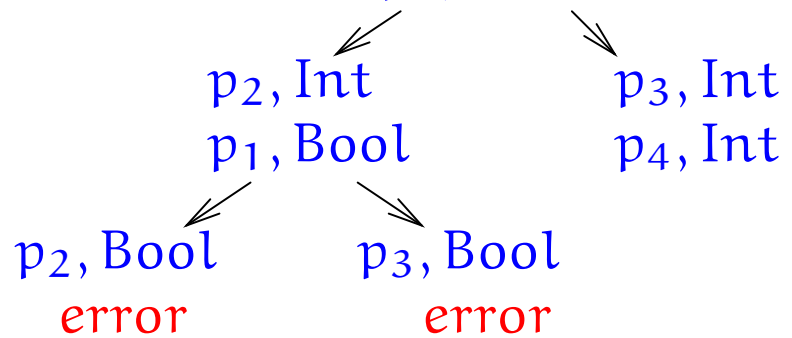
$p_1, \tau \longrightarrow p_2, \tau$

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

$p_2, \tau \longrightarrow p_1, \text{Bool}$, if $\tau \sqsubseteq \text{Rational}$

$p_3, \text{Int} \longrightarrow p_4, \text{Int}$

Abstract trace: p_0, Object
 p_1, Int



Constant propagation analysis

```


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

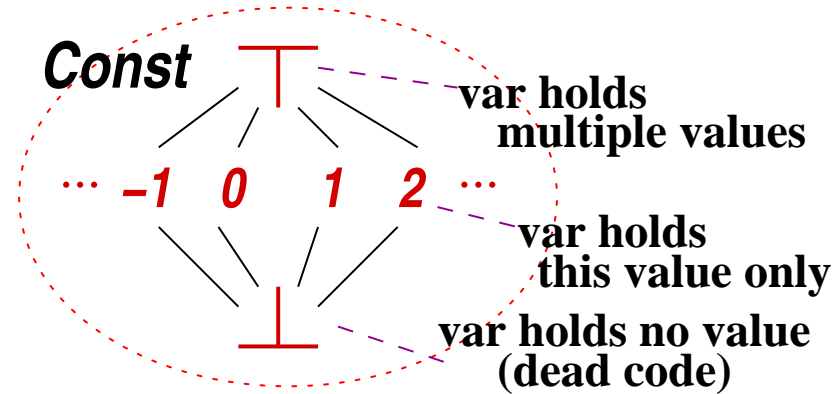
 $p_1$  : while ( $x < y + z$ )  

       $p_2$  :  $x = x + 1;$   

      }  

 $p_3$  : exit


```



where $m + n$ is interpreted

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

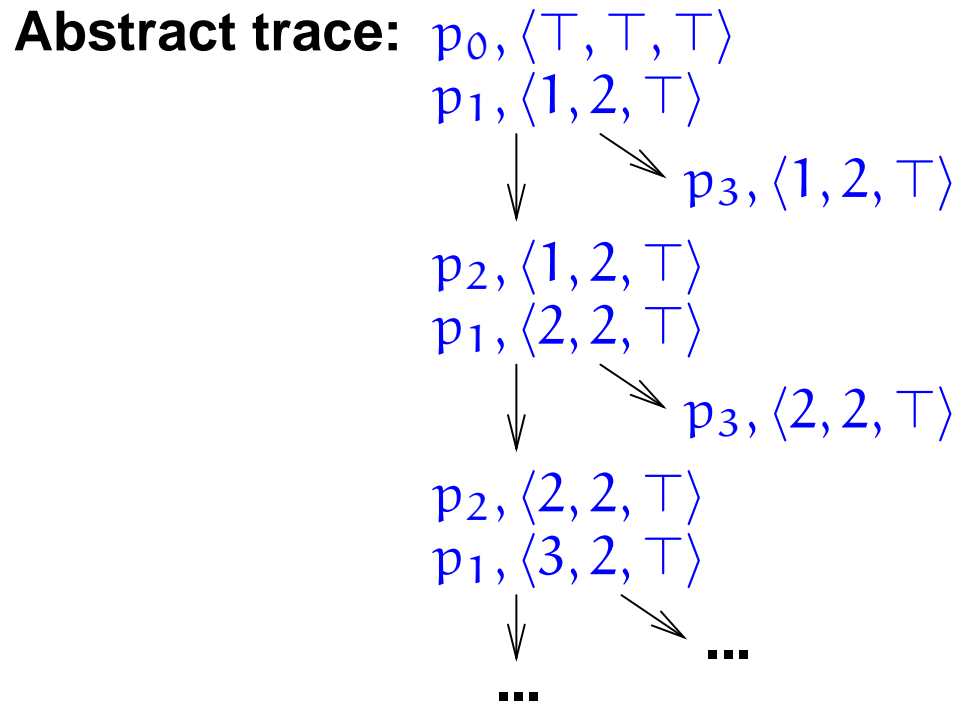
$\top \neq k_i \neq \perp, i \in 1..2$

$\top + k \longrightarrow \top$

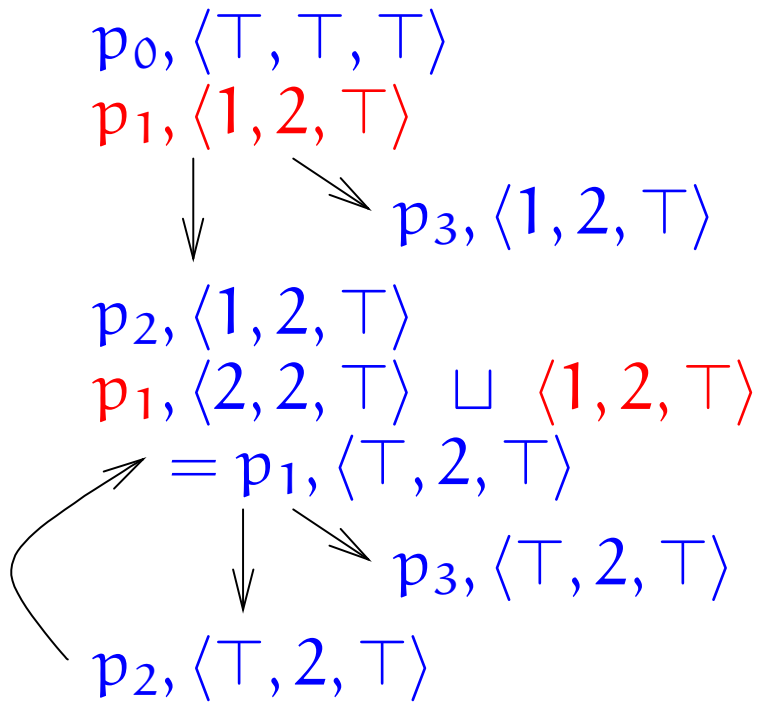
$k + \top \longrightarrow \top$

Let $\langle u, v, w \rangle$ abbreviate

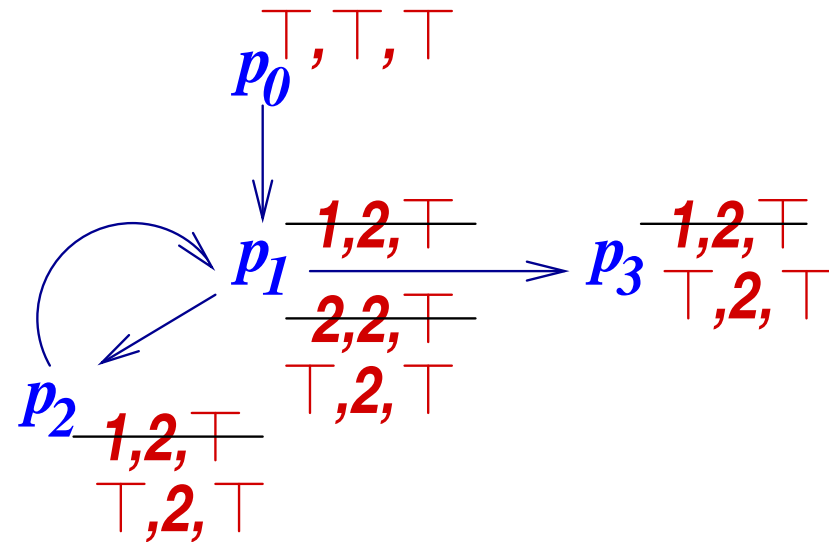
$\langle x : u, y : v, z : w \rangle$



An *acceleration* is needed for finite 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 ( $x < y + z$ ) {  

         $p_2$  :  $x = x + 1;$   

  }  

 $p_3$  : exit


```

2

Array bounds (pre)checking uses intervals

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;
while (i < 10) {
    ... a[i] ...
    i = i + 1;
}
```

$i = [0, 0]$

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

$i = [0, 0] \sqcup [1, 1] \sqcup [-\infty, 9] = [0, 1]$

...

p_2 $i = [1, 1]$

$i = [1, 1] \sqcup [2, 2] = [1, 2]$

...

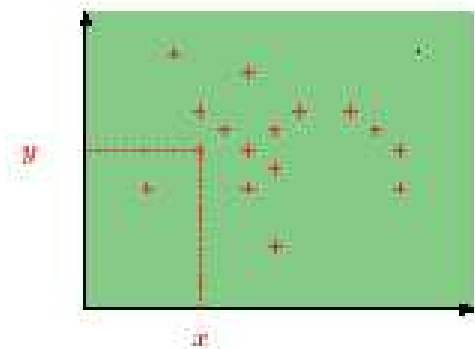
at p_1 : $[0..9]$

At convergence, i 's ranges are at p_2 : $[1..10]$

at loop exit : $[1..10] \sqcap [10, +\infty] = [10, 10]$

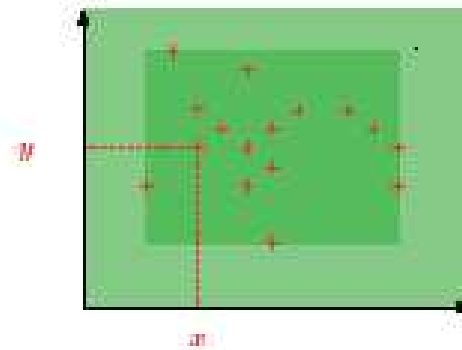
Examples of relations between variables' values

These Figures are from *Abstract Interpretation: Achievements and Perspectives* by Patrick Cousot, Proc. SSGRR 2000.



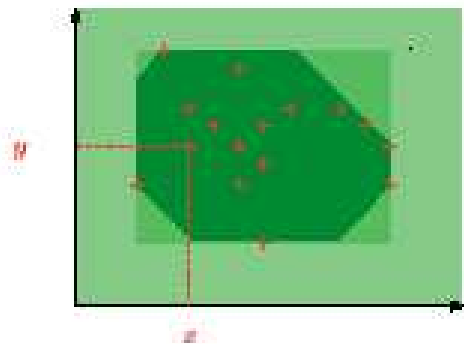
$$\begin{cases} x \geq 0 \\ y \geq 0 \end{cases}$$

Fig. 2
SIGNS



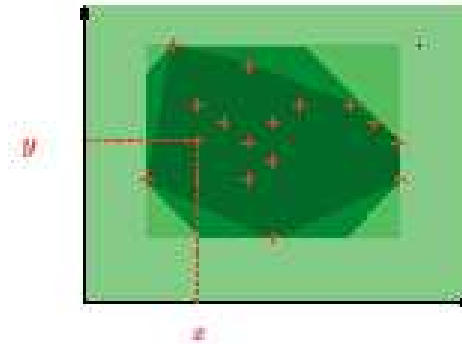
$$\begin{cases} x \in [3, 27] \\ y \in [4, 32] \end{cases}$$

Fig. 3
INTERVALS



$$\begin{cases} 3 \leq x \leq 27 \\ x + y \leq 88 \\ 4 \leq y \leq 32 \\ x - y \leq 61 \end{cases}$$

Fig. 4
OCTAGONS



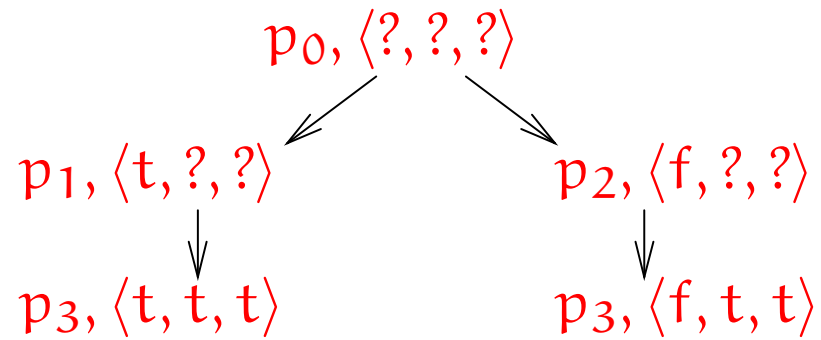
$$\begin{cases} 7x + 31y \leq 325 \\ 21x + 7y \geq 0 \end{cases}$$

Fig. 5
POLYHEDRA

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 choose three predicates, $\phi_2 = z \geq x$

$$\phi_3 = z \geq y$$

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

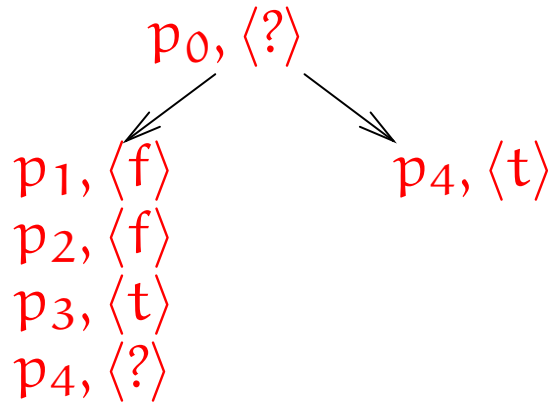
At all occurrences of p_3 in the abstract trace, $\phi_2 \wedge \phi_3$ holds.

When a goal is undecided, *refinement* is necessary

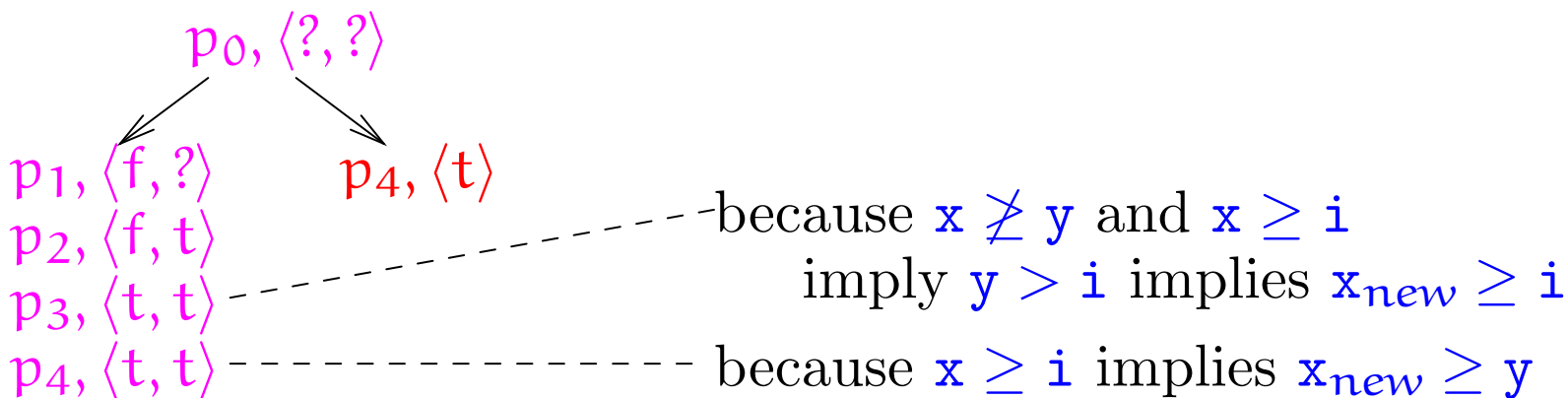
Prove $\phi_0 \equiv x \geq y$ at p_4 :

```

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

The first refinement 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.

Refinement stage j adds predicates $P_j \equiv \{i = j, x = n - j\}$; the refinement process continues forever!

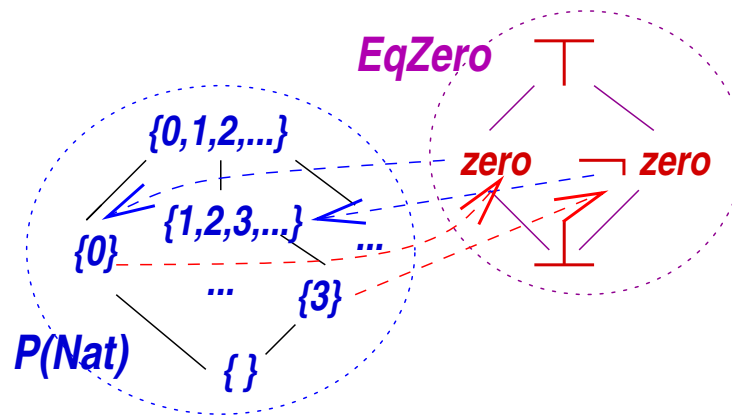
The loop invariant is $x = n - i$:-)

An abstract domain defines a “logic”

For abstract domain A , $a \in A$ is a “property/predicate,” and $\gamma(a) \subseteq C$ defines a subset of concrete states that make a “true.” For $s \in C$,

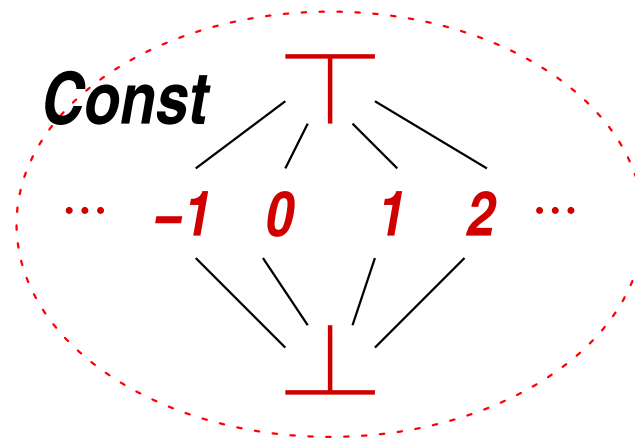
s has a , written $s \models_A a$, iff $s \in \gamma(a)$ iff $\alpha\{s\} \sqsubseteq a$

Example: We might abstract Nat by EqZero :



We have, for example, that $3 \models \neg \text{zero}$; we also have that $3 \models T$; and we have that $3 \models \neg \text{zero} \sqcap T$.

In one sense, *every* analysis based on abstract interpretation is a “predicate abstraction.” But the “logic” is weak — it supports conjunction (\sqcap) but not necessarily disjunction (\sqcup).



For **Const**, we have that

$$2 \models_{\text{Const}} 2 \sqcap T \text{ iff } 2 \models_{\text{Const}} 2 \text{ and } 2 \models_{\text{Const}} T.$$

In general, $n \models_{\text{Const}} a \sqcap a'$ iff $n \models_{\text{Const}} a$ and $n \models_{\text{Const}} a'$

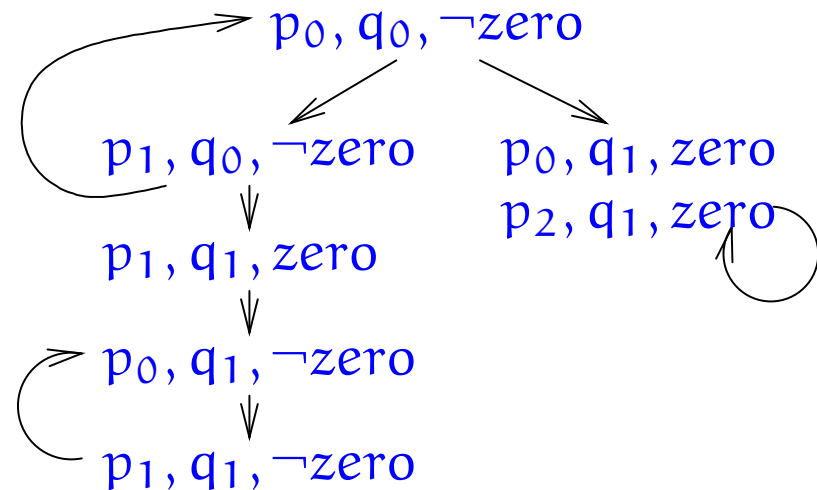
But **Const** does not support disjunction: $2 \models_{\text{Const}} T$, and
 $T = 2 \sqcup 3 = 3 \sqcup 4 = 2 \sqcup 3 \sqcup 4$, etc.

Hence $2 \models_{\text{Const}} 3 \sqcup 4$, but this does *not* imply that $2 \models_{\text{Const}} 3$ or
 $2 \models_{\text{Const}} 4$!

Abstract traces can be *model checked*

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

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



Starting from p_0, q_0, k , for $k > 0$, will every execution “Generally/Globally” avoid resource misuse ?

$$p_0, q_0, k \models G \neg(p_1 \wedge q_1) ?$$

Will every execution reach a Future state where x is permanently (Generally/Globally) zero?

$$p_0, q_0, k \models FG \text{ zero} ?$$

The logical operators, **F** and **G**, 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: $\llbracket \phi \rrbracket \subseteq \mathcal{P}(\text{Trace})$

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

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

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

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, $(\mathcal{P}(\text{Trace}), \subseteq) \leftrightarrow (\mathcal{P}(\text{MiniLTL}), \supseteq)$,

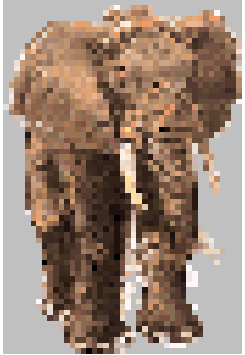
where $\sqcup = \cap$ in $\mathcal{P}(\text{MiniLTL})$:

$\gamma(P) = \bigcap \{ \llbracket \phi \rrbracket \mid \phi \in P \}$ – the traces that have all the properties in P

$\alpha(S) = \{ \phi \mid S \subseteq \llbracket \phi \rrbracket \}$ – properties held by all traces 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:

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

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

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

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

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

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

References

- ◆ The papers of Patrick and Radhia Cousot (www.di.ens.fr/~cousot), including
 1. Abstract interpretation: a unified lattice model for static analysis of programs by construction or approximation of fixpoints. ACM POPL 1977.
 2. Systematic design of program analysis frameworks. ACM POPL, 1979.
 3. Abstract interpretation: achievements and perspectives. Proc. SSGRR 2000.
- ◆ 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.
- ◆ Hanne Nielson, Flemming Nielson, and Chris Hankin. [Principles of Program Analysis](#). Springer 1999.
- ◆ 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.