

17-654

Analysis of Software Artifacts

Midterm Review

Jonathan Aldrich

Analysis as an Approximation

- If S doesn't terminate normally, y cannot be 0
- Problem: undecidable to tell if S terminates!
- In general program analysis must compute an approximation

Safety and Precision

- ***Conservative/Safe Analysis***

- *Computes a larger set of possibilities than will actually occur in program execution*

- ***Precise Analysis***

- *Computes as small a set of possibilities for program execution as it can*

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Finding the Fixpoint

- Why should we think we will find an n such that $F^{n+1}(\overrightarrow{RD}_{\emptyset}) = F^n(\overrightarrow{RD}_{\emptyset})$?

Finding the Fixpoint

- Why should we think we will find an n such that $F^{n+1}(\overrightarrow{RD}_{\emptyset}) = F^n(\overrightarrow{RD}_{\emptyset})$?
 - F is monotone
 - Therefore, every application of F either:
 - Does not change \overrightarrow{RD} (and so we have a fixpoint)
 - Or increases the size of a set in \overrightarrow{RD}
 - The set of definitions is finite so the sets in \overrightarrow{RD} cannot increase in size forever
 - Therefore the algorithm terminates with a fixpoint at some finite n

Reaching Defs. vs. Available Exp.

- Reaching Defs. **May analysis**
 - Initial dataflow values: *empty* sets
 - *Union* at control flow merge
 - Precision: want *least* fixed point
 - Safety: err on the side of *larger* sets
- Available Exp. **Must analysis**
 - Initial dataflow values: *universal* sets
 - *Intersection* at control flow merge
 - Precision: want *greatest* fixed point
 - Safety: err on the side of *smaller* sets

Monotone Framework

$$\begin{aligned} \text{Analysis}_\circ(\ell) &= \iota && \text{if } \ell \in E \\ &= \sqcup \{ \text{Analysis}_\bullet(\ell') \mid (\ell', \ell) \in F \} && \text{otherwise} \end{aligned}$$

$$\text{Analysis}_\bullet(\ell) = f_\ell(\text{Analysis}_\circ(\ell))$$

where:

- \circ means entry (forward) or exit (backward)
- \bullet means exit (forward) or entry (backward)
- \sqcup is \cup (may) or \cap (must)
- F is $\text{flow}(S_*)$ (forward) or $\text{flow}^R(S_*)$ (backward)
- E is $\{ \text{init}(S_*) \}$ (forward) or $\text{final}(S_*)$ (backward)
- ι specifies initial or final analysis information, and
- f_ℓ is a transfer function
 - Typically $f_\ell(x) = x \setminus \text{kill}_{\text{Analysis}}(B^\ell) \cup \text{gen}_{\text{Analysis}}(B^\ell)$

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	RD	AE	LV
\sqcup			
<i>F</i>			
<i>E</i>			
ι			

Monotone Framework

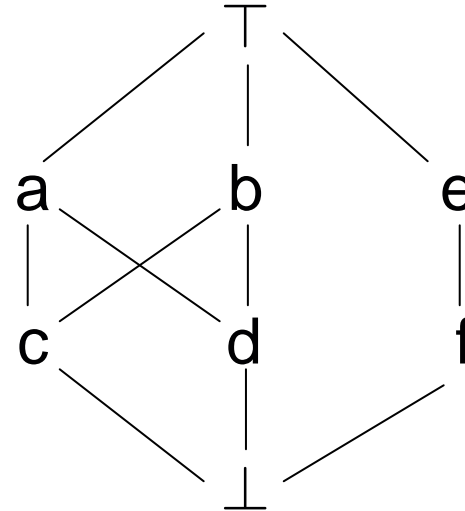
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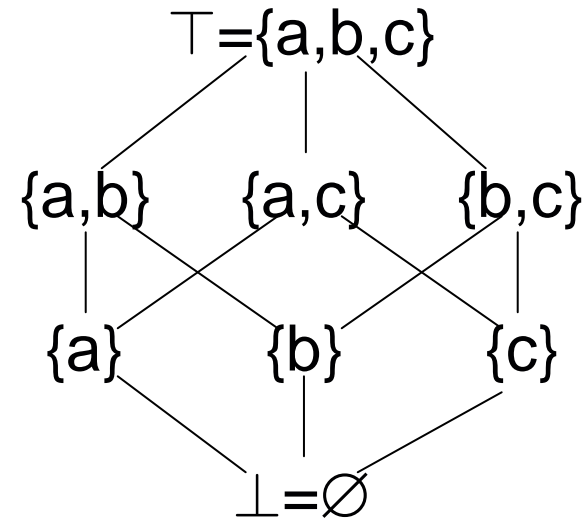
	RD	AE	LV
\sqcup	\cup	\cap	\cup
<i>F</i>	$\text{flow}(S_*)$	$\text{flow}(S_*)$	$\text{flow}^R(S_*)$
<i>E</i>	$\{ \text{init}(S_*) \}$	$\{ \text{init}(S_*) \}$	$\text{final}(S_*)$
ι	$\{ (x, ?) \mid x \in \text{FV}(S_*) \}$	\emptyset	\emptyset

Complete Lattice

- Not all data flow analyses use sets
 - Lattice: a more general concept
- A set L with:
 - A partial order \sqsubseteq
 - A combination operator \sqcup
 - A least element $\perp = \sqcup (\emptyset)$
 - A greatest element $\top = \sqcup (L)$
 - Each subset Y of L has a least upper bound $\sqcup (Y)$
- Typically we want the lattice to have finite height
 - A finite number of elements on each path from \perp to \top
 - See NNH Appendix A.3

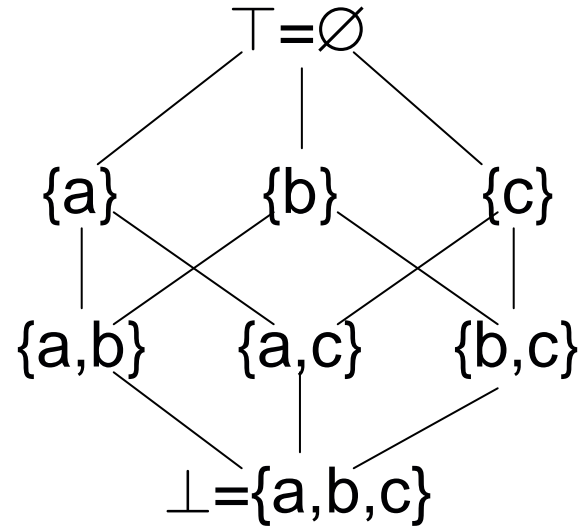


Example: Subset Lattice



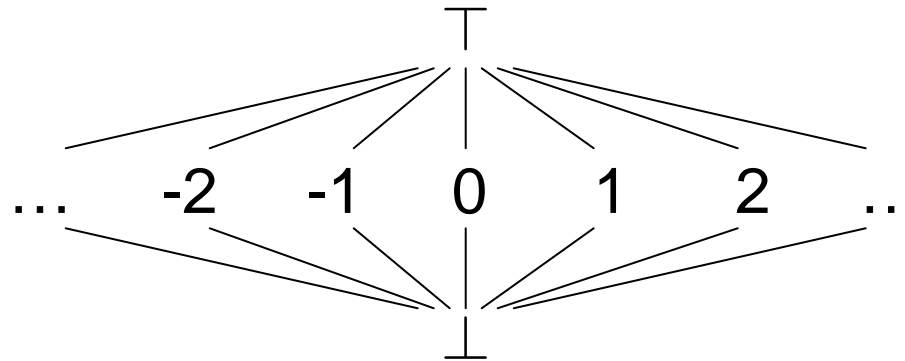
- Reaching Definitions
- The set $L = \mathcal{P}(\{a, b, c\})$ with:
 - $\sqsubseteq = \subseteq$
 - $\sqcup = \cup$ (may analysis)
 - $\perp = \emptyset$ (the most precise and starting element)
 - $\top = \{a, b, c\}$ (the least precise element)

Example: Superset Lattice



- Available Expressions
- The set $L = \mathcal{P}(\{a,b,c\})$ with:
 - $\sqsubseteq = \supseteq$
 - $\sqcup = \cap$ (must analysis)
 - $\perp = \{a,b,c\}$ (the most precise and starting element)
 - $\top = \emptyset$ (the least precise element)

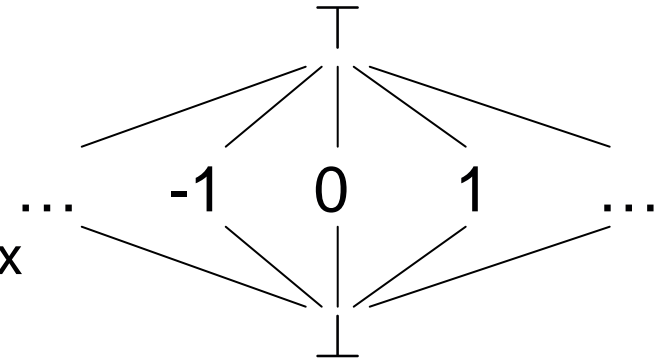
Constant Propagation Lattice



- More efficient than the set of possible values
 - Don't want to store sets
 - If more than one value, give up and assume any (\top)
- The set $L = \{\perp, \top\} \cup \text{NAT}$ with:
 - $x \sqsubseteq \top$, $\perp \sqsubseteq x$, $x \sqsubseteq x$
 - $x \sqcup \perp = x$, $x \sqcup \top = \top$, $n \sqcup m = \top$ (for $n \neq m$)
- $\perp = \top$

Constant Propagation Transfer Fns

- Can't use gen and kill sets
 - Data flow values aren't sets anymore!
- Instead, define function by cases on syntax
 - Input is incoming data flow value σ



- $f^{CP}([x := a], \sigma) = \sigma [x \mapsto CP(a, \sigma)]$
- $f^{CP}([skip], \sigma) = \sigma$
- $f^{CP}([b], \sigma) = \sigma$
- $CP(n, \sigma) = n$
- $CP(x, \sigma) = \sigma(x)$
- $CP(a_1 \ op_a \ a_2, \sigma) = CP(a_1, \sigma) \widehat{op}_a \ CP(a_2, \sigma)$
- $z_1 \widehat{op}_a \ z_2 = z_1 \widehat{op}_a \ z_2$ if $z_1, z_2 \in \text{NAT}$
- $= \top$ if $z_1 = \top$ or $z_2 = \top$
- $= z_1(z_2)$ if $z_2(z_1) = \perp$

Execution Traces

- Sequence of $\langle pp, mem \rangle$ pairs
 - pp is a program point
 - Just before statement pp
 - mem is the state of variables in memory

```
[y := x]1;  
[z := 1]2;  
while [y > 1]3 do  
  [z := z * y]4;  
  [y := y - 1]5;  
[y := 0]6;
```

<u>pp</u>	<u>x</u>	<u>y</u>	<u>z</u>
1	2	0	0
2	2	2	0
3	2	2	1
4	2	2	1
5	2	2	2
3	2	1	2
6	2	1	2
-	2	0	2

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1	1	0	0
2	1	1	0
3	1	1	1
6	1	1	1
-	1	0	1

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3	3	3	1
4	3	3	1
5	3	3	3
3	3	2	3
4	3	2	3
5	3	2	6
3	3	1	6
6	3	1	6
-	3	0	6

Execution Traces

- Sequence of $\langle pp, mem \rangle$ pairs
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pp x y z

Repeat for all possible initial values of $x, y, z!$

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[y := x]1;  
[z := 1]2;  
while [y > 1]3 do  
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  [y := y - 1]5;  
[y := 0]6;
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Abstraction

- Abstraction function α
 - maps traces to data flow values at a certain time t in the trace
- $\alpha_{CP}(\langle p_1, m_1 \rangle \dots \langle p_n, m_n \rangle, t)$
 $= m_t$
- $\alpha_{SA}(\langle p_1, m_1 \rangle \dots \langle p_n, m_n \rangle, t)$
 $= sign(m_t)$
- Also define program point function pp
- $pp(\langle p_1, m_1 \rangle \dots \langle p_n, m_n \rangle, t)$
 $= p_t$

<u>t</u>	<u>pp</u>	<u>x</u>	<u>y</u>	<u>z</u>
0	1	3	0	0
1	2	3	3	0
2	3	3	3	1
3	4	3	3	1
4	5	3	3	3
5	3	3	2	3
6	4	3	2	3
7	5	3	2	6
8	3	3	1	6
9	6	3	1	6
10	-	3	0	6

$$\alpha_{CP}(T, 0) = (x=3, y=0, z=0)$$

$$\alpha_{CP}(T, 10) = (x=3, y=0, z=6)$$

$$\alpha_{SA}(T, 10) = (x=+, y=0, z=+)$$

WHILE Traces, Formally

- A trace for program S_1 and initial state σ_1 is either:
 - a finite sequence $(S_1, \sigma_1), \dots, ([], \sigma_n)$,
where $(S_i, \sigma_i) \rightarrow (S_{i+1}, \sigma_{i+1})$ for $i \in 1, \dots, n-1$
 - an infinite sequence $(S_1, \sigma_1), \dots, (S_i, \sigma_i), \dots$
where $(S_i, \sigma_i) \rightarrow (S_{i+1}, \sigma_{i+1})$ for $i \geq 1$
- Slight notational simplification
 - We will abbreviate $(S_1, \sigma_1), \dots, (S_n, \sigma_n)$
as $(\text{first}(S_1), \sigma_1), \dots, (\text{first}(S_n), \sigma_n)$
 - Uses program counter labels instead of complete programs

Local Soundness

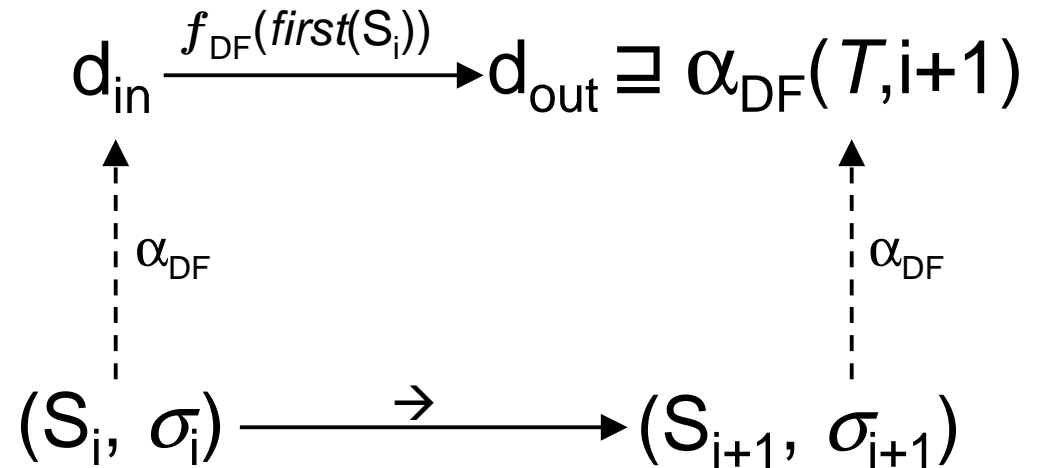
To show:

if $(S_i, \sigma_i) \rightarrow (S_{i+1}, \sigma_{i+1}) \in T$

and $d_{in} = \alpha_{DF}(T, i)$

and $d_{out} = f_{DF}(first(S_i), d_{in})$

then $\alpha_{DF}(T, i+1) \sqsubseteq d_{out}$



Intuitively, translating from concrete to abstract and applying the flow function will safely approximate (\sqsupseteq) taking a step in the trace and translating from concrete to abstract

What does Correctness Mean?

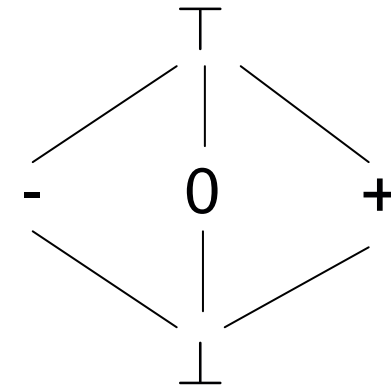
- Intuition
 - At a fixed point, analysis results are a *conservative abstraction of program execution*
- *Soundness* condition
 - When data flow analysis reaches a fixed point F , then for all traces T and all times t in each trace, $\alpha(T, t) \sqsubseteq F(pp(T, t))$

Global Soundness

- Intuition
 - We begin with initial dataflow facts ι that safely approximate (\sqsupseteq) all initial stores σ_1
 - By local soundness, each transfer function when given safe input information yields safe output information
 - By induction, any fixed point of the analysis is sound

Soundness Example: Sign Analysis

Custom Lattice



- Transfer functions

- σ is input data flow value

- $f^{SA}([x := a], \sigma) = \sigma [x \mapsto SA(a, \sigma)]$

- $f^{SA}([\text{skip}], \sigma) = \sigma$

- $f^{SA}([b], \sigma) = \sigma$

- $SA(n, \sigma) = \text{sign}(n)$

// returns sign of n

- $SA(x, \sigma) = \sigma(x)$

- $SA(a_1 + a_2, \sigma) = +$

// is this sound?

- $SA(a_1 \text{ op}_a a_2, \sigma) = \top$

for op_a ≠ +

Local Soundness

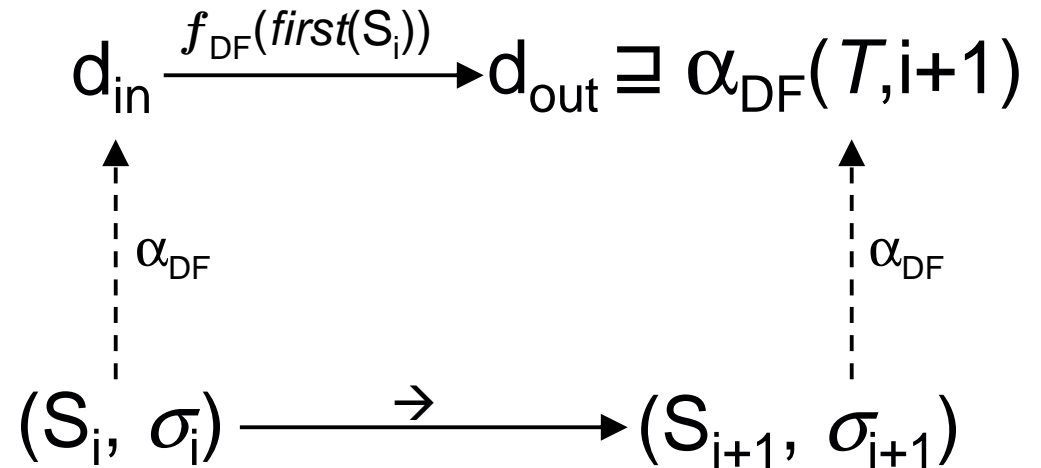
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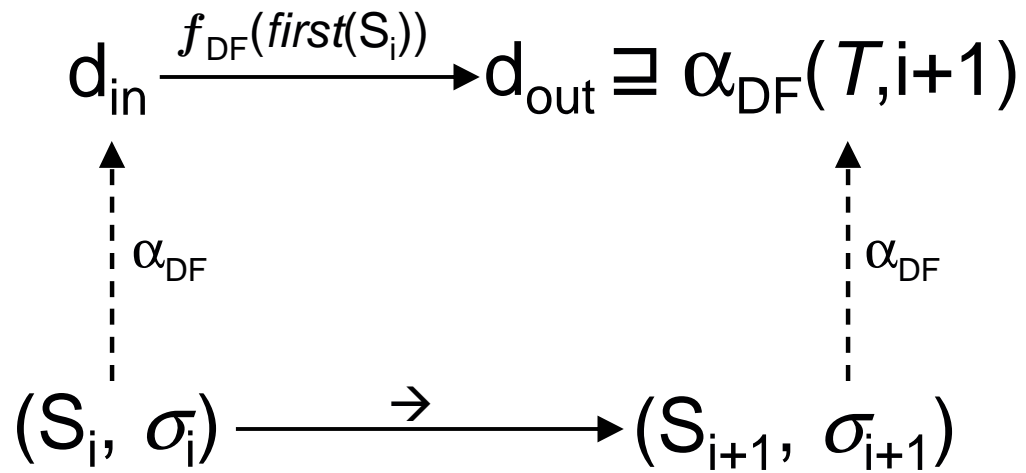
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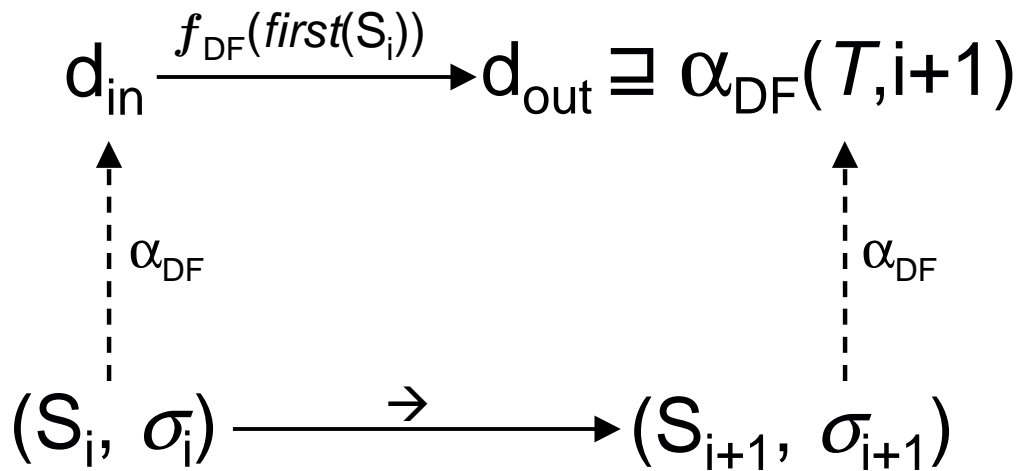


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Local Soundness Fails



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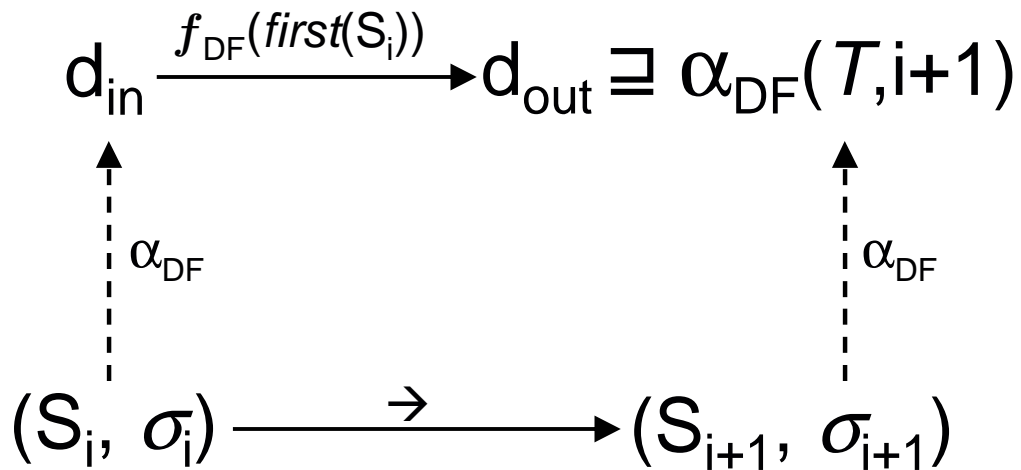


Program:

$[z := y-7]^1$

$[x := y+z]^2$

Local Soundness Fails



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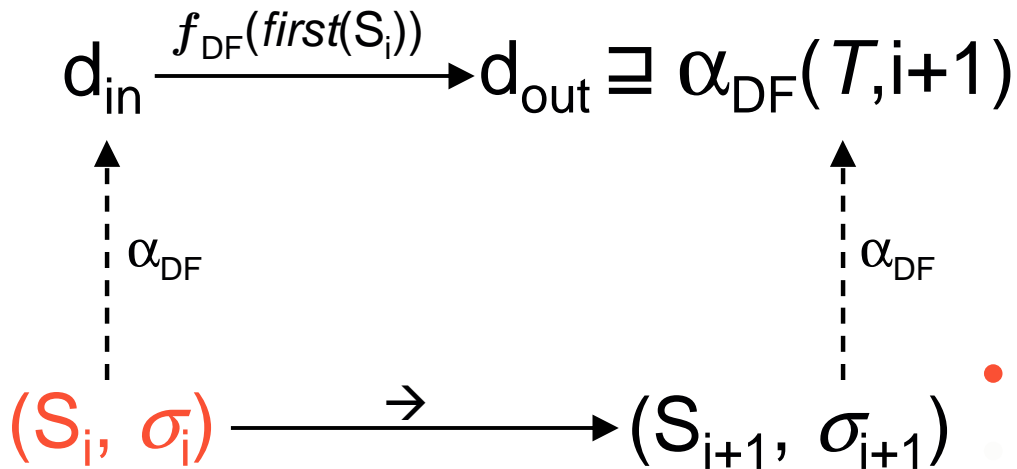
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Trace T :

<u>t</u>	<u>pp</u>	<u>x</u>	<u>y</u>	<u>z</u>
0	1	0	3	0
1	2	0	3	-4
2	-	-1	3	-4

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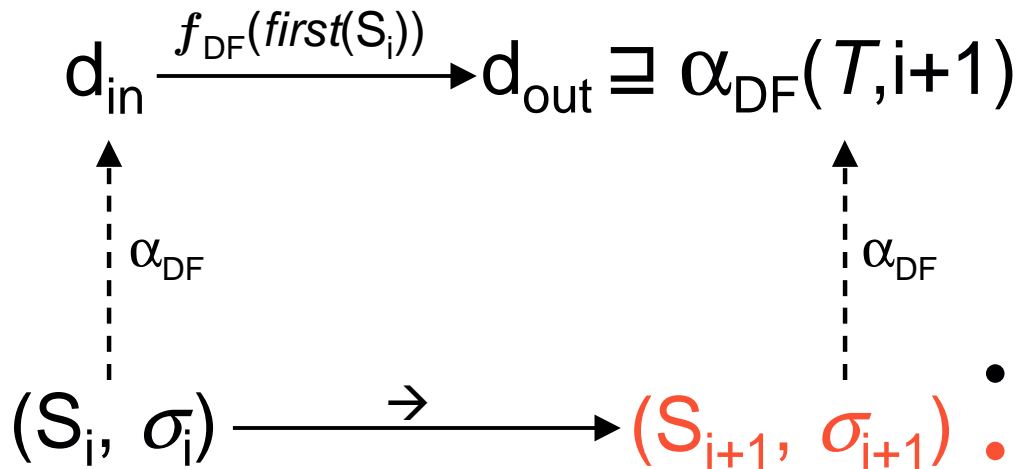
$[x := y+z]^2$

- $\sigma_1 = (x=0, y=3, z=-4)$
- $\sigma_2 = (x=-1, y=3, z=-4)$
- $d_{\text{in}} = \alpha_{\text{DF}}(T, 1) = (x=0, y=+, z=-)$
- $\alpha_{\text{DF}}(T, 2) = (x=-, y=+, z=-)$
- $d_{\text{out}} = f^{\text{SA}}([x := y+z]^2, d_{\text{in}})$
 $= d_{\text{in}} [x \mapsto \text{SA}(y+z, d_{\text{in}})]$
 $= d_{\text{in}} [x \mapsto +]$
 $= (x=+, y=+, z=-)$
- $(x=+, y=+, z=-) \not\cong (x=-, y=+, z=-)$

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[done]

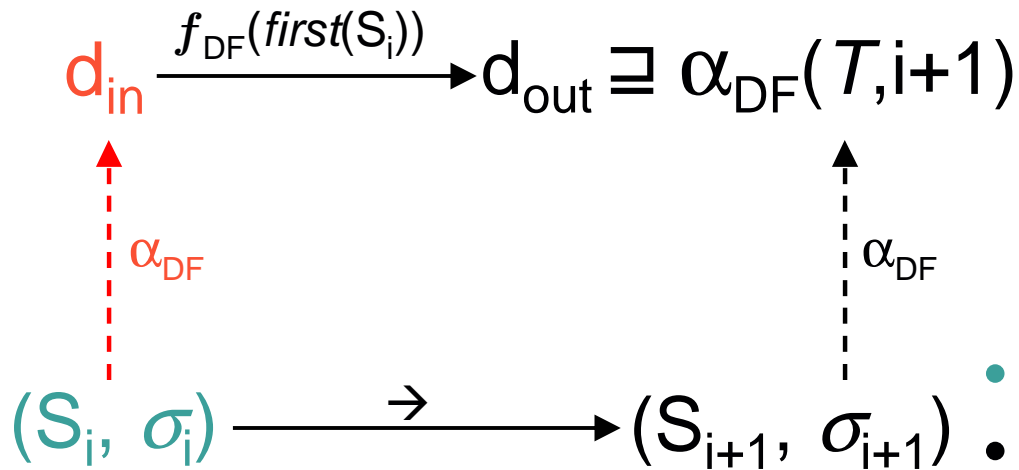
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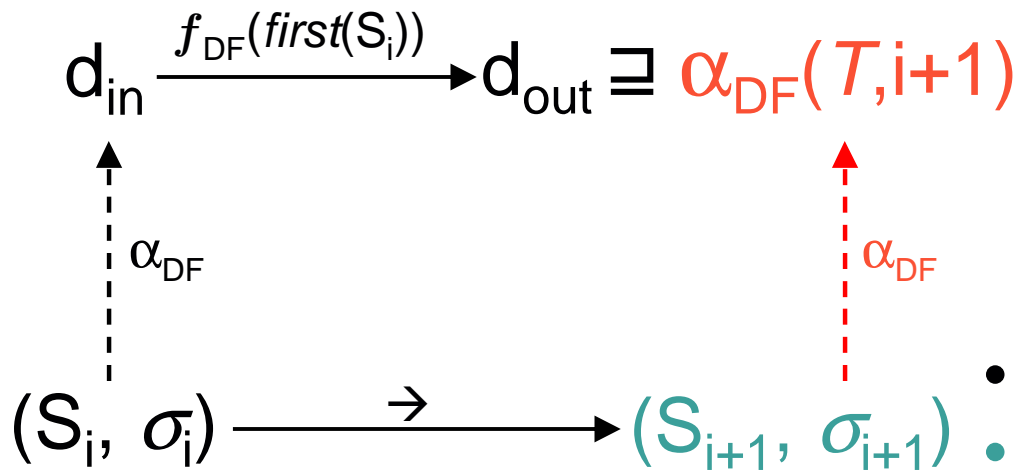
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Trace T :

t	pp	x	y	z
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1	2	0	3	-4
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- $\alpha_{DF}(T, 2) = (x=+, y=+, z=-)$
- $d_{out} = f^{SA}([x := y+z]^2, d_{in})$
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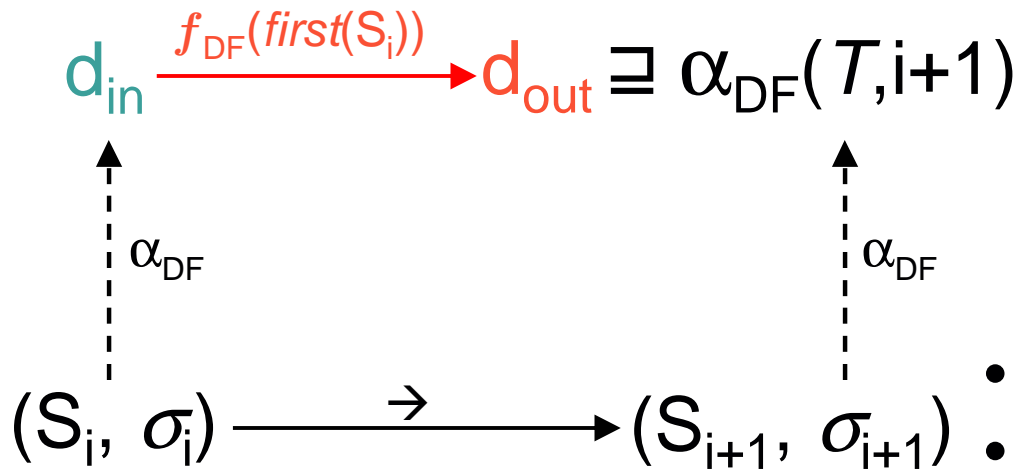
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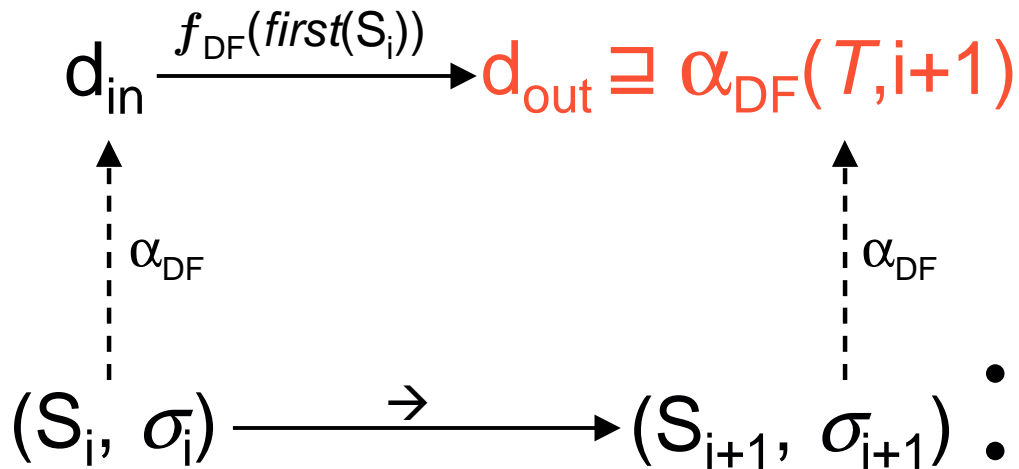
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Local Soundness for Constant Propagation

To show:

if $(S_i, \sigma_i) \rightarrow (S_{i+1}, \sigma_{i+1}) \in T$

and $d_{in} = \alpha_{CP}(T, i)$

and $d_{out} = f_{CP}(first(S_i), d_{in})$

then $\alpha_{CP}(T, i+1) \sqsubseteq d_{out}$

- Case: $S_i = [x := a]^\ell$
 - $\sigma_{i+1} = \sigma_i [x \mapsto \mathcal{A}(a, \sigma_i)]$
 - $d_{in} = \alpha_{CP}(T, i) = \sigma_i$
 - $d_{out} = f_{CP}([x := a]^\ell, \sigma_i)$
 $= \sigma_i [x \mapsto CP(a, \sigma_i)]$
 - $\alpha_{CP}(T, i+1) = \sigma_{i+1}$
 $= \sigma_i [x \mapsto \mathcal{A}(a, \sigma_i)]$
 - Lemma: $\mathcal{A}(a, \sigma_i) = CP(a, \sigma_i)$
 - Thus $\sigma_i [x \mapsto \mathcal{A}(a, \sigma_i)]$
 $\sqsubseteq \sigma_i [x \mapsto CP(a, \sigma_i)]$

Abstraction for Reaching Definitions

- $\alpha_{RD}(\langle p_1, m_1 \rangle \dots \langle p_n, m_n \rangle, t) =$
 $\{ (x, p_k) \mid x \in FV(S_*)$
 and $k < t$
 and $stmt(p_k) = [x := a]$
 and $\forall j, k < j < t \text{ } stmt(p_j) \neq [x := a'] \}$

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To show:

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- Case: $S_i = [x := a]^\ell$
 - $d_{in} = \alpha_{RD}(T, i)$
 - $d_{out} = f_{RD}([x := a]^\ell, d_{in})$
 $= (\alpha_{RD}(T, i) \setminus \{(x, *)\}) \cup \{(x, \ell)\}$
 - Lemma: $\alpha_{RD}(T, i+1)$
 $= (\alpha_{RD}(T, i) \setminus \{(x, *)\}) \cup \{(x, \ell)\}$
 - So $\alpha_{RD}(T, i+1) = d_{out}$
 - Thus $\alpha_{RD}(T, i+1) \sqsubseteq d_{out}$

Abstraction for Live Variables

- $\alpha_{LV}(\langle p_1, m_1 \rangle \dots \langle p_n, m_n \rangle, t) =$
 $\{ x \mid x \in FV(stmt(p_k)) \text{ where } k > t$
 and $\forall j, t < j < k \text{ } stmt(p_j) \neq [x := a'] \}$

Local Soundness for Live Variables

To show:

if $(S_i, \sigma_i) \rightarrow (S_{i+1}, \sigma_{i+1}) \in T$

and $d_{in} = \alpha_{LV}(T, i+1)$

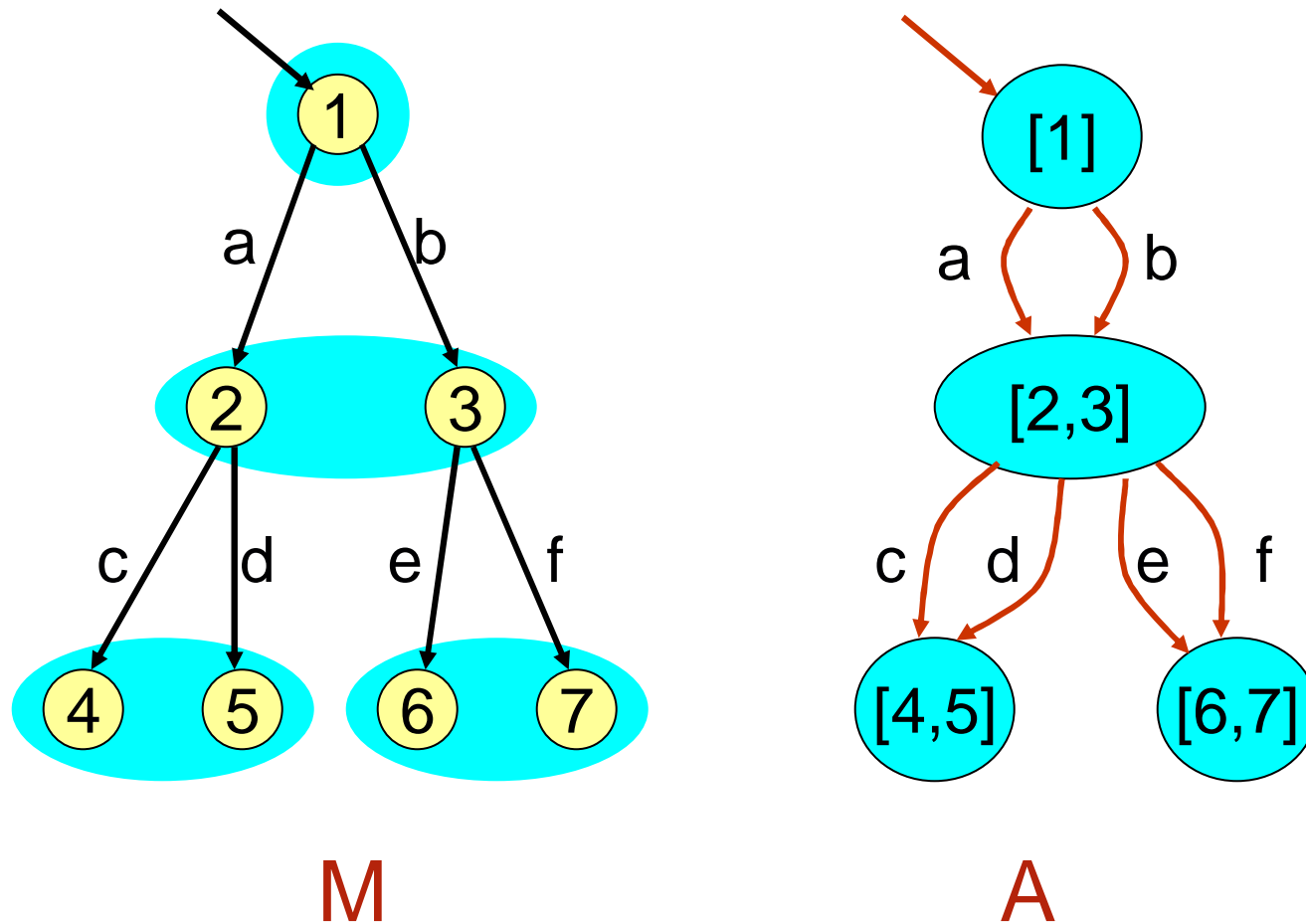
and $d_{out} = f_{LV}(first(S_{i+1}), d_{in})$

then $\alpha_{LV}(T, i) \sqsubseteq d_{out}$

Note: i and $i+1$ are
swapped due to reverse
analysis

- Case: $S_{i+1} = [x := a]^\ell$
 - $d_{in} = \alpha_{RD}(T, i+1)$
 - $d_{out} = f_{RD}([x := a]^\ell, d_{in})$
 $= (\alpha_{RD}(T, i+1) \setminus \{x\}) \cup FV(a)$
 - Lemma: $\alpha_{RD}(T, i)$
 $= (\alpha_{RD}(T, i+1) \setminus \{x\}) \cup FV(a)$
 - So $\alpha_{RD}(T, i) = d_{out}$
 - Thus $\alpha_{RD}(T, i) \sqsubseteq d_{out}$

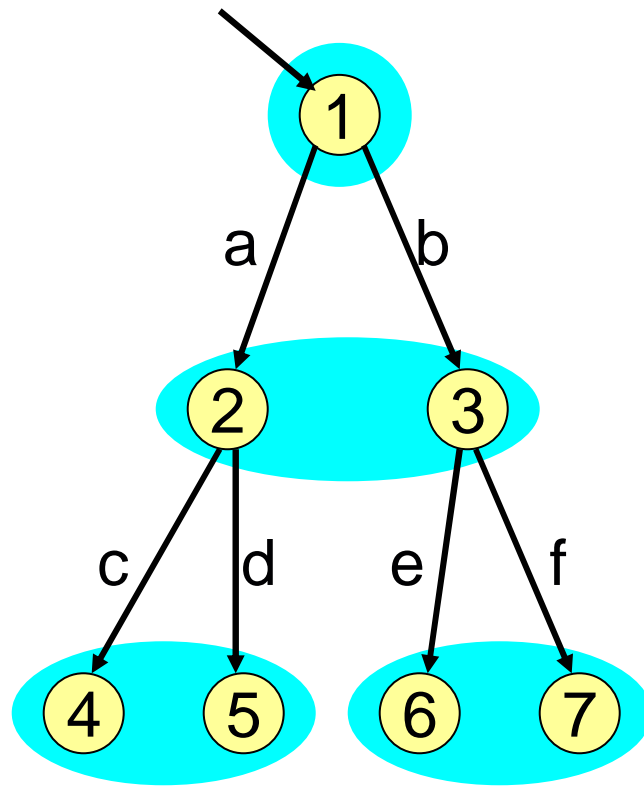
Conservative Abstraction



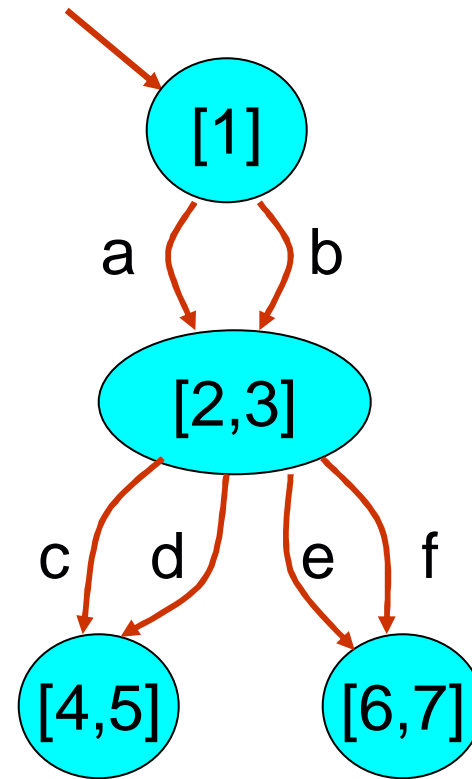
Conservative Abstraction

- Every trace of M is a trace of A
 - A over-approximates what M can do
(Preserves safety properties!): $A \models \phi \Rightarrow M \models \phi$
- Some traces of A may not be traces of M
 - May yield spurious counterexamples - $\langle a, e \rangle$
- Eliminated via abstraction refinement
 - Splitting some clusters in smaller ones
 - Refinement can be automated

Original Abstraction

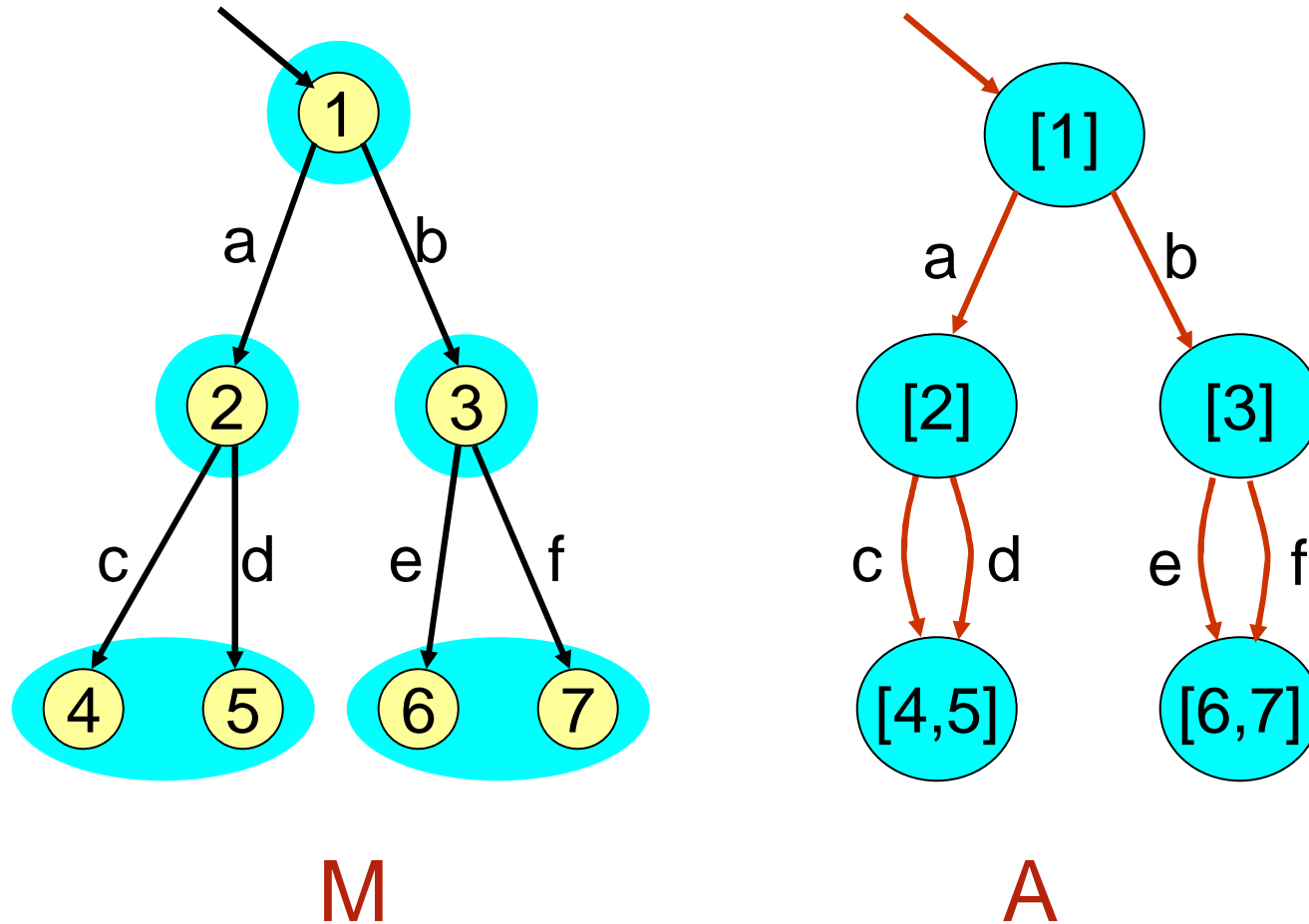


M



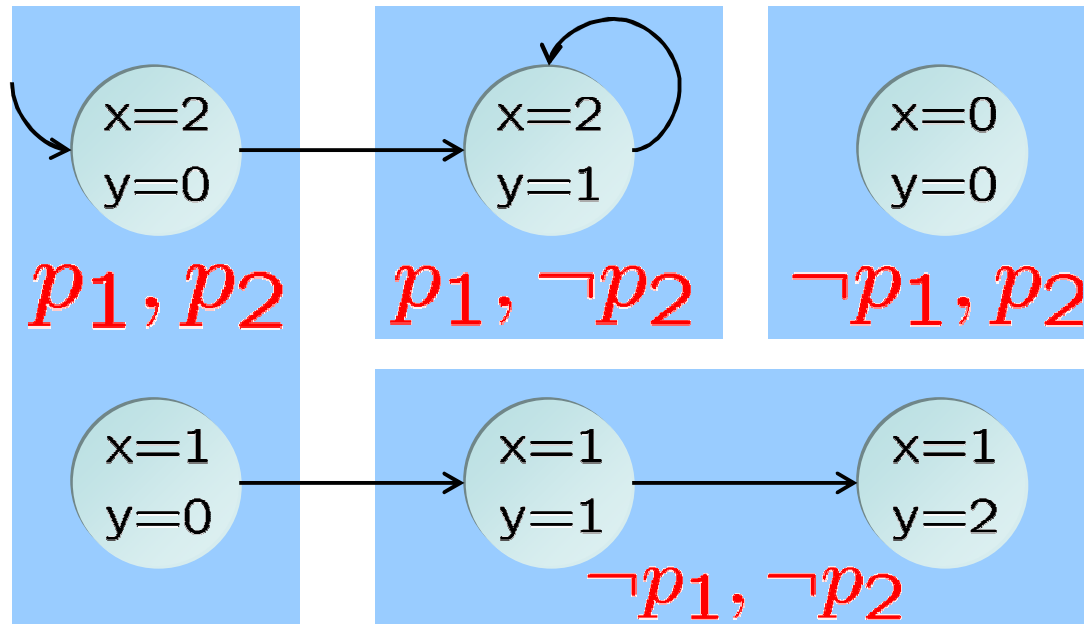
A

Refined Abstraction



Predicate Abstraction

Concrete States:



Predicates:

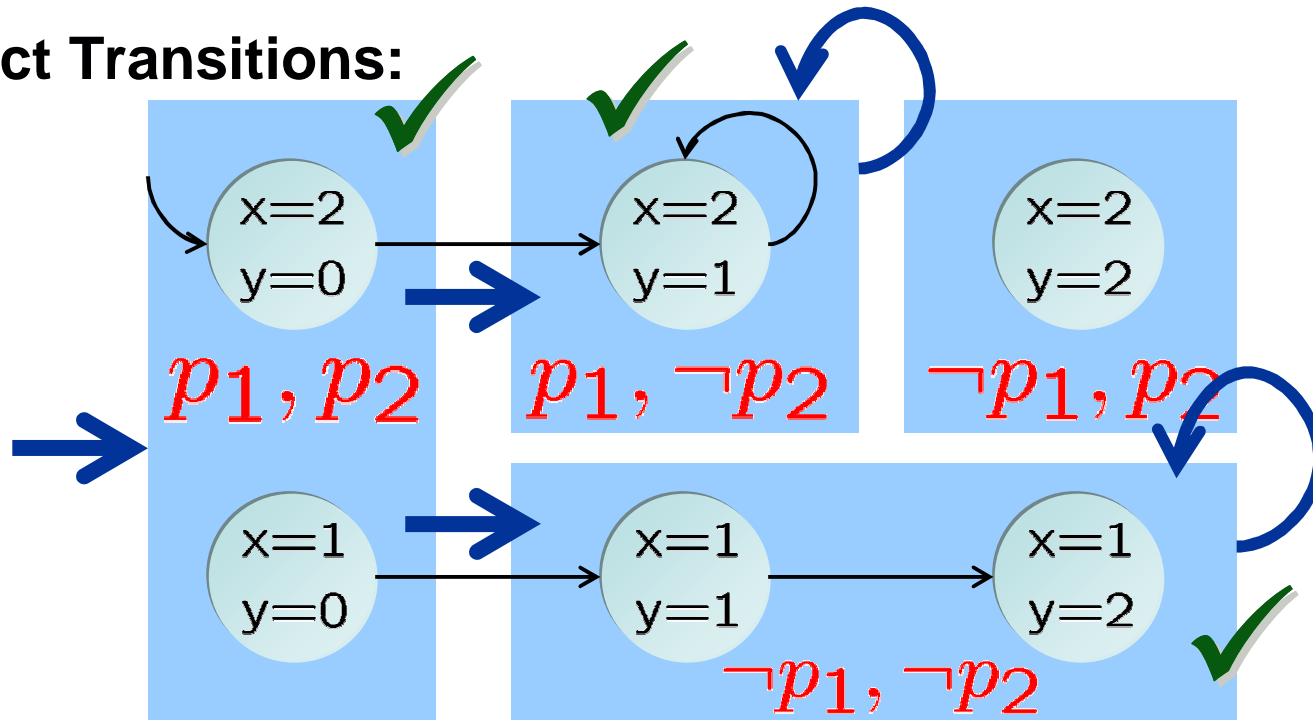
$$p_1(s) = (s.x > s.y)$$

$$p_2(s) = (s.y = 0)$$

Abstract transitions?

Predicate Abstraction

Abstract Transitions:



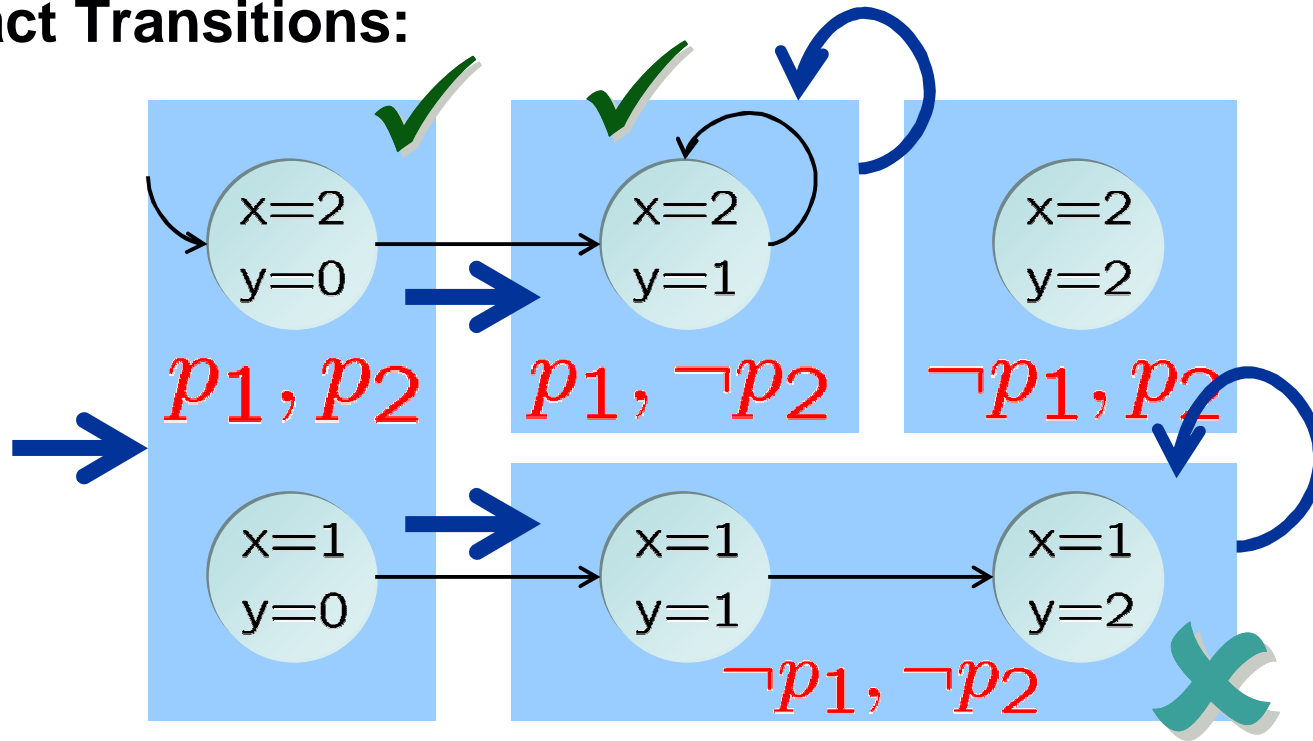
Property:

$$p_1 \vee \neg p_2 \quad \iff \quad (s.x > s.y) \vee (s.y \neq 0)$$

Property holds. Ok.

Predicate Abstraction

Abstract Transitions:



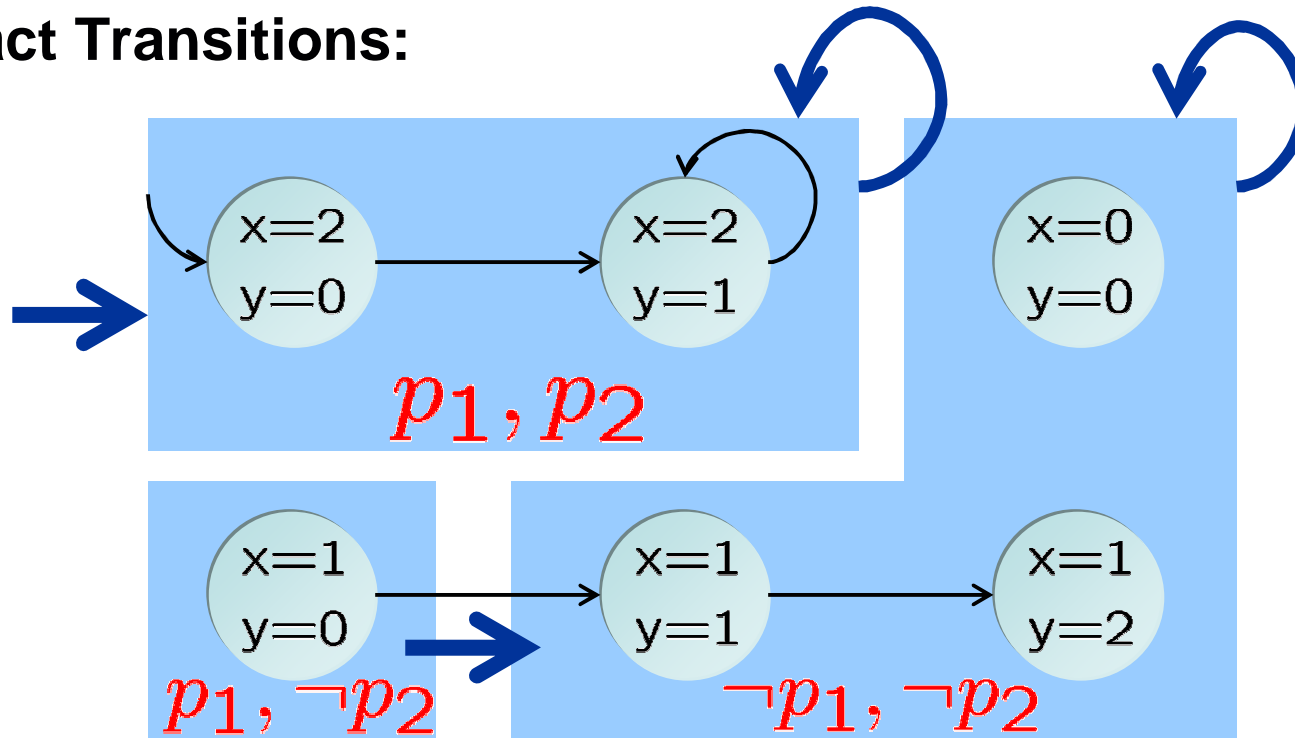
Property:

$$p_1 \iff (s.x > s.y)$$

**This trace is
spurious!**

Predicate Abstraction

Abstract Transitions:



Property:

$$p_1 \iff (s.x > s.y) \quad \checkmark$$

New Predicates:

$$p_1(s) = (s.x > s.y)$$
$$p_2(s) = (s.x = 2)$$