# CS 477: Dataflow Analysis and Abstract Interpretation 

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## Partial Orders

## Set $\mathbf{P}$

Partial order relation $\leq$ such that $\forall x, y, z \in P$

- $x \leq x$
- $x \leq y$ and $y \leq x$ implies $x=y$
- $x \leq y$ and $y \leq z$ implies $x \leq z$
(reflexive)
(antisymmetric)
(transitive)

Can use partial order to define

- Upper and lower bounds
- Least upper bound
- Greatest lower bound


## Upper Bounds

## If $S \subseteq P$ then

- $x \in P$ is an upper bound of $S$ if $\forall y \in S . y \leq x$
- $x \in P$ is the least upper bound of $S$ if
- $x$ is an upper bound of $S$, and
- $x \leq y$ for all upper bounds $y$ of $S$
$\checkmark$ - join, least upper bound, lub, supremum, sup
- $\vee S$ is the least upper bound of $S$
- $x \vee y$ is the least upper bound of $\{x, y\}$


## Lower Bounds

If $S \subseteq P$ then

- $x \in P$ is a lower bound of $S$ if $\forall y \in S . x \leq y$
- $x \in P$ is the greatest lower bound of $S$ if
- $x$ is a lower bound of $S$, and
- $y \leq x$ for all lower bounds $y$ of $S$
- $\wedge$ - meet, greatest lower bound, gllb, infimum, inf
- $\wedge S$ is the greatest lower bound of $S$
- $x \wedge y$ is the greatest lower bound of $\{x, y\}$


## Covering

$x<y$ if $x \leq y$ and $x \neq y$
$x$ is covered by $y$ ( $y$ covers $x$ ) if

- $x<y$, and
- $x \leq z<y$ implies $x=z$

Conceptually, y covers $x$ if there are no elements between $x$ and $y$

## Example

$P=\{000,001,010,011,100,101,110,111\}$
(standard boolean lattice, also called hypercube)
$x \leq y$ is equivalent to ( $x$ bitwise-and $y$ ) $=x$


## Hasse Diagram

- If $y$ covers $x$
- Line from $y$ to $x$
- $y$ above $x$ in diagram


## Example: Same as

$P=\wp(\{a, b, c\})$
(standard powerset lattice, also called hypercube)
$x \leq y$ is equivalent to $x \subseteq y=x$


## Hasse Diagram

- If $y$ covers $x$
- Line from $y$ to $x$
- $y$ above $x$ in diagram


## Lattices

Consider poset $(\mathrm{P}, \leq)$ and the operators $\wedge$ (meet) and $\vee$ (join)
If for all $x, y \in P$ there exist $x \wedge y$ and $x \vee y$, then $P$ is a lattice.
If for all $S \subseteq P$ there exist $\wedge S$ and $\vee S$ then P is a complete lattice.
All finite lattices are complete
Example of a lattice that is not complete: Integers $Z$

- For any $x, y \in Z, x \vee y=\max (x, y), x \wedge y=\min (x, y)$
- But $\vee Z$ and $\wedge Z$ do not exist
- $Z \cup\{+\infty,-\infty\}$ is a complete lattice


## Top and Bottom

Greatest element of $P$ (if it exists) is top ( $T$ )

- $\forall a \in L . a \vee T=T$
- Note: $\forall \mathrm{a} \in \mathrm{L} . \mathrm{a} \leq \mathrm{T}$ and $\mathrm{T} \wedge a=a$

Least element of $P$ (if it exists) is bottom ( $\perp$ )

- $\forall \mathrm{a} \in \mathrm{L} . \mathrm{a} \wedge \perp=\perp$
- Note: $\forall \mathrm{a} \in \mathrm{L} . \perp \leq a$ and $\perp \vee a=a$


## Lattice (Recap)

$$
(\mathrm{P}, \leq, \wedge, \vee, \perp, \mathrm{T})
$$

- Set
- Partial order
- Meet
- Join
- Bottom
- Top


## Connection Between $\leq, \wedge$, and $\vee$

Theorem: The following 3 properties are equivalent:

- $x \leq y$
- $x \vee y=y$
- $x \wedge y=x$

Let's prove:

- $x \leq y$ implies $x \vee y=y$ and $x \wedge y=x$
- $x \vee y=y$ implies $x \leq y$
- $x \wedge y=x$ implies $x \leq y$

Then by transitivity, we can obtain

- $x \vee y=y$ implies $x \wedge y=x$
- $x \wedge y=x$ implies $x \vee y=y$


## Connecting Lemma Proofs

Lemma: $x \leq y$ implies $x \vee y=y$
Proof:

- $x \leq y$ implies $y$ is an upper bound of $\{x, y\}$.
- Any upper bound $z$ of $\{x, y\}$ must satisfy $y \leq z$.
- So $y$ is least upper bound of $\{x, y\}$ and $x \vee y=y$

Lemma: $x \leq y$ implies $x \wedge y=x$
Proof:

- $x \leq y$ implies $x$ is a lower bound of $\{x, y\}$.
- Any lower bound $z$ of $\{x, y\}$ must satisfy $z \leq x$.
- So $x$ is greatest lower bound of $\{x, y\}$ and $x \wedge y=x$


## Connecting Lemma Proofs

Lemma: $x \vee y=y$ implies $x \leq y$
Proof:

- $y$ is an upper bound of $\{x, y\}$ implies $x \leq y$

Lemma: $x \wedge y=x$ implies $x \leq y$
Proof:

- $x$ is a lower bound of $\{x, y\}$ implies $x \leq y$


## Lattices as Algebraic Structures

We have previously defined $\vee$ and $\wedge$ in terms of $\leq$
We will now define $\leq$ in terms of $\vee$ and $\wedge$

- Start with $\vee$ and $\wedge$ as arbitrary algebraic operations that satisfy associative, commutative, idempotence, and absorption laws
- We will define $\leq$ using $\vee$ and $\wedge$
- We will show that $\leq$ is a partial order

Intuitive concept of $\vee$ and $\wedge$ as information combination operators (or, and) or set operations (union, intersection)

## Algebraic Properties of Lattices

Assume arbitrary operations $\vee$ and $\wedge$ such that

- $(x \vee y) \vee z=x \vee(y \vee z) \quad$ (associativity of $\vee$ )
- $(x \wedge y) \wedge z=x \wedge(y \wedge z) \quad$ (associativity of $\wedge)$
- $x \vee y=y \vee x$
- $x \wedge y=y \wedge x$
- $x \vee x=x$
- $x \wedge x=x$
- $x \vee(x \wedge y)=x$
(absorption of $\vee$ over $\wedge$ )
- $x \wedge(x \vee y)=x$
(absorption of $\wedge$ over $\vee$ )


## Connection Between $\wedge$ and $\vee$

Thm: $x \vee y=y$ if and only if $x \wedge y=x$

Proof ('if'): $x \vee y=y \Rightarrow x=x \wedge y$

$$
\begin{aligned}
x & =x \wedge(x \vee y) & & \text { (by absorption) } \\
& =x \wedge y & & \text { (by assumption) }
\end{aligned}
$$

Proof ('only if'): $x \wedge y=x \Rightarrow y=x \vee y$

$$
\begin{array}{rlr}
y & =y \vee(y \wedge x) \text { (by absorption) } \\
& =y \vee(x \wedge y)(\text { by commutativity) } \\
& =y \vee x \quad \text { (by assumption) } \\
& =x \vee y \quad \text { (by commutativity) }
\end{array}
$$

## Properties of $\leq$

Define: $x \leq y$ if $x \vee y=y$

## Thm : $x \leq y$ is a partial order

Proof of transitive property. Must show that

$$
\begin{array}{rlrl}
x \vee y & x \vee y \text { and } y \vee z=z \text { implies } x \vee z=z \\
x \vee z & =x \vee(y \vee z) & & \text { (by assumption) } \\
& =(x \vee y) \vee z & \text { (by associativity) } \\
& =y \vee z & & \text { (by assumption) } \\
& =z & & \text { (by assumption) }
\end{array}
$$

## Properties of $\leq$

Proof of asymmetry property. Must show that

$$
\begin{aligned}
& x \vee y=y \text { and } y \vee x=x \text { implies } x=y \\
& x=y \vee x \\
&=x \vee y \text { (by assumption) } \\
& \text { (by commutativity) } \\
&=y \\
& \text { (by assumption) }
\end{aligned}
$$

Proof of reflexivity property. Must show that $x \vee x=x$, which follows directly $x \vee x=x \quad$ (by idempotence)

## Properties of $\leq$

Induced operation $\leq$ agrees with original definitions of $\vee$ and $\wedge$, i.e.,

- $x \vee y=\sup \{x, y\}$
- $x \wedge y=\inf \{x, y\}$


## Proof of $x \vee y=\sup \{x, y\}$

Consider any upper bound u for x and y .
Given $x \vee u=u$ and $y v u=u$, must show $x \vee y \leq u$, i.e., $(x \vee y) \vee u=u$
$\mathrm{u}=\mathrm{x} \mathrm{v} \mathrm{u}$
(by assumption)
$=x \vee(y \vee u)$
$=(x \vee y) \vee u$
(by assumption)
(by associativity)

## Proof of $x \wedge y=\inf \{x, y\}$

- Consider any lower bound $L$ for $x$ and $y$.
- Given $x \wedge L=L$ and $y \wedge L=L$, must show $L \leq x \wedge$ y, i.e., $(x \wedge y) \wedge L=L$

$$
\begin{aligned}
L & =x \wedge L \\
& =x \wedge(y \wedge L) \\
& =(x \wedge y) \wedge L
\end{aligned}
$$

(by assumption)
(by assumption)
(by associativity)

## Semi-lattice ( $\mathrm{P}, ~ \wedge$ )

Set $P$ and binary operation $\wedge$ such that $\forall x, y, z \in P$

- $x \wedge x=x$
- $x \wedge y=y \wedge x$ implies $x=y$
- $(x \wedge y) \wedge z=x \wedge(y \wedge z)$
(idempotent)
(commutative)
(associative)

The operation $\wedge$ imposes a partial order on $P$
If $((L, \leq), \wedge, \vee)$ is a lattice, then

- $(L, \wedge)$ is a meet semi-lattice
- $(L, v)$ is a join semi-lattice

Give us more flexibility to define the analysis.

- Since our analyses deal with complete lattices, we will represent the framework on them, but it can also be defined on semi-lattices
- Some dataflow analyses can be only represented on semi-lattices


## Chains

## A poset $(S, \leq)$ is a chain if $\forall x, y \in S . y \leq x$ or $x \leq y$

Height of a poset/lattice: the size of the maximum chain.
$(S, \leq)$ is finite if it has the finite height.

P satisfies the ascending chain condition if for all sequences $x_{1} \leq x_{2}$
$\leq \ldots$..there exists $n$ such that $x_{n}=x_{n+1}=\ldots$

- When a particular ascending chain has the property that $x_{n}=$ $x_{n+1}=\ldots$ we say that it stabilizes
- Then ascending chain condition means that all ascending chains stabilize


## From one variable to more

If $L$ is a poset then so is the Cartesian product $L x L$ :
Let $\left(L_{1}, \leq_{1}\right)$ and $\left(L_{2}, \leq_{2}\right)$ be posets.
Then $\left(L^{*}, \leq^{*}\right)$ is also a poset, where

$$
L^{*}=\left\{\left(l_{1}, l_{2}\right) \mid l_{1} \in L_{1}, l_{2} \in L_{2}\right\} \text { and }\left(l_{11}, l_{21}\right) \leq^{*}\left(l_{12}^{\top}, l_{22}\right)
$$

iff $l_{11} \leq_{1} l_{12}$ and $l_{21} \leq_{2} l_{22}$
This construction extends immediately on lattices, so that for $S \subseteq L^{*}$, we define $\perp^{*}=\left(\perp_{1}, \perp_{2}\right)$, we define

$$
g l b(Y)=\left(g l b\left\{l_{1} \mid\left(l_{1,-}\right) \in Y\right\}, \operatorname{glb}\left\{l_{2} \mid\left(\left(_{-}, l_{2}\right) \in Y\right\}\right.\right.
$$

and same for $l u b$ and $\mathrm{T}^{*}$

## From one variable to more

Total function space (S -> L) :
Let ( $L, \leq$ ) be a poset, $S$ a set and $f$ total function. Then $\left(L^{f}, \leq^{f}\right)$ is also a poset, where
$L^{f}=\{f: S \rightarrow L\}$ and $f^{\prime} \leq^{f} f^{\prime \prime}$ iff $\forall s \in S . f^{\prime}(s) \leq f^{\prime \prime}(s)$.
To extend to lattices, we define $\perp^{f}=\lambda s . \perp$ and $g l b(Y)=\lambda s . g l b_{0}\{f(s) \mid f \in Y\}$ and same for $l u b$ and $T^{f}$

## Monotone Function Space ( $L_{1}$-> $L_{2}$ ):

Let $\left(L_{1}, \leq_{1}\right)$ and $\left(L_{2}, \leq_{2}\right)$ be posets and $f$ monotone. Then ( $L^{f}, \leq^{f}$ ) is also a poset, where $\perp^{f}=\lambda s . \perp_{2}$ and
$L^{f}=\left\{f: L_{1} \rightarrow L_{2}\right\}$ and $f^{\prime} \leq^{f} f^{\prime \prime}$ iff $\forall l_{1} \in L_{1} \cdot f^{\prime}\left(l_{1}\right) \leq_{2} f^{\prime \prime}\left(l_{1}\right)$

## Application to Dataflow Analysis

Dataflow information will be lattice values

- Transfer functions operate on lattice values
- Solution algorithm will generate increasing sequence of values at each program point
- Ascending chain condition will ensure termination

We will use $\vee$ to combine values at control-flow join points

## Transfer Functions

Transfer function $f: P \rightarrow P$ is defined for each node in control flow graph

- Maps lattice elements to lattice elements

The function f models effect of the node on the program information

## Transfer Functions

Each dataflow analysis problem has a set F of transfer functions $f: P \rightarrow P$. This set $F$ contains:

- Identity function belongs to the set, $i \in F$
- F must be closed under composition: $\forall f, g \in F$. the function $h=\lambda x . f(g(x)) \in F$
- Each $f \in F$ must be monotonic:

$$
x \leq y \text { implies } f(x) \leq f(y)
$$

- Sometimes all $f \in F$ are distributive*:

$$
f(x \vee y)=f(x) \vee f(y)
$$

- Note that Distributivity implies monotonicity
*One can also define distributivity in terms of $\wedge($ "meet"): $f(x \wedge y)=f(x) \wedge f(y)$


## Distributivity Implies Monotonicity

## Proof.*

Assume distributivity: $f(x \vee y)=f(x) \vee f(y)$

Must show: $x \vee y=y$ implies $f(x) \vee f(y)=f(y)$

$$
\begin{aligned}
f(y) & =f(x \vee y) & & \text { (by assumption) } \\
& =f(x) \vee f(y) & & \text { (by distributivity) }
\end{aligned}
$$

*For $f(x \wedge y)=f(x) \wedge f(y)$, show $x \wedge y=x=>f(x) \wedge f(y)=f(x) ; f(x)=f(x \wedge y)=f(x) \wedge f(y)$

## Knaster-Tarsky Fixed-point Theorem

 Let:- $(L, \leq, \wedge, \vee, T, \perp)$ be a complete lattice
- $f: L \rightarrow L$ be a monotonic function
- fix $(f)$ is the set of fixed points of $f$

The set fix ( $f$ ) with relation $\leq$, and operators $\wedge, \vee$ is forming a complete lattice.

- There will be a least fixed-point and greatest fixed point

Consequences:

- f has at least one fixpoint
- That fixpoint is the largest element in the chain

$$
\perp, f(\perp), f(f(\perp)), f(f(f(\perp))), \ldots, f n(\perp)
$$

## Putting the Pieces Together...

## Forward Dataflow Analysis

Simulates execution of program forward with flow of control
Tuple (G, (L, $\leq$ ), F, I) - (graph, (lattice), transfer fs., initial val.)
For each node $\mathrm{n} \in \mathrm{G}$, we have

- $\mathrm{in}_{\mathrm{n}}$ - value at program point before n
- out ${ }_{n}$ - value at program point after $n$
- $f_{n} \in F-$ transfer function for $n$ (given $\mathrm{in}_{n}$, computes out ${ }_{n}$ )
- Signature of $\mathrm{in}_{\mathrm{n}}$, out $\mathrm{n}_{\mathrm{n}} \mathrm{f}_{\mathrm{n}}: \mathrm{L} \rightarrow \mathrm{L}$

Requires that solution satisfies

- $\forall \mathrm{n}$. out $_{n}=f_{n}\left(\mathrm{in}_{\mathrm{n}}\right)$
- $\forall \mathrm{n} \neq \mathrm{n}_{0}$.

$$
\mathrm{in}_{\mathrm{n}}=\vee\left\{\text { out }_{\mathrm{m}} \cdot \mathrm{~m} \text { in } \operatorname{pred}(\mathrm{n})\right\}
$$

- $\mathrm{in}_{\mathrm{no}}=\mathrm{I}$, summarizes information at the start of program


## Dataflow Equations

Compiler processes program to obtain a set of dataflow equations

$$
\begin{aligned}
& \text { out }_{n}:=\quad f_{n}\left(\mathrm{in}_{n}\right) \\
& \mathrm{in}_{\mathrm{n}}:=\vee\left\{\text { out }_{\mathrm{m}} . \text { for each } \mathrm{m} \text { in } \operatorname{pred}(\mathrm{n})\right\}
\end{aligned}
$$

Conceptually separates analysis problem from program

## Worklist Algorithm for Solving Forward Dataflow Equations

for each $n$ do out $_{n}:=f_{n}(\perp)$

$$
\begin{aligned}
& \mathrm{in}_{\text {ne }}:=\mathrm{I} ; \text { out }_{\mathrm{ne}}:=f_{\mathrm{n} \mathrm{\theta}}(\mathrm{I}) \\
& \text { worklist }:=\mathrm{N}-\left\{\mathrm{n}_{\ominus}\right\}
\end{aligned}
$$

while worklist $\neq \varnothing$ do remove a node n from worklist in $_{n}:=\vee\left\{\right.$ out $_{m} . m$ in pred(n) \} out $_{n}:=f_{n}\left(\right.$ in $\left._{n}\right)$
if out ${ }_{n}$ changed then

$$
\text { worklist }:=\text { worklist } \cup \text { succ(n) }
$$

## Correctness Argument

Why does the result satisfy dataflow equations?

- Whenever it processes a node $n$, algorithm sets out $n_{n}:=f_{n}\left(\mathrm{in}_{n}\right)$ Therefore, the algorithm ensures that out ${ }_{n}=f_{n}\left(\mathrm{in}_{n}\right)$
- Whenever out ${ }_{m}$ changes, it puts succ( $m$ ) on worklist. Consider any node $n \in \operatorname{succ}(m)$. It will eventually come off worklist and algorithm will set

$$
\mathrm{in}_{\mathrm{n}}:=\vee\left\{\operatorname{out}_{\mathrm{m}} \cdot \mathrm{~m} \text { in } \operatorname{pred}(\mathrm{n})\right\}
$$

to ensure that $\mathrm{in}_{\mathrm{n}}=\vee\left\{\right.$ out $_{\mathrm{m}} . \mathrm{m}$ in pred $\left.(\mathrm{n})\right\}$

- So final solution will satisfy dataflow equations
- Need also to ensure that the dataflow equalities correspond to the states in the program execution (this comes later!)


## Termination Argument

Why does algorithm terminate?
Sequence of values taken on by $\mathrm{IN}_{\mathrm{n}}$ or $\mathrm{OUT}_{\mathrm{n}}$ is a chain. If values stop increasing, worklist empties and algorithm terminates.

If lattice has ascending chain property, algorithm terminates

- Algorithm terminates for finite lattices
- For lattices with infinite length, use widening operator
- Detect lattice values that may be part of infinitely ascending chain
- Artificially raise value to least upper bound of chain


## Termination Argument (Details)

- For finite lattice ( $\mathrm{L}, \leq$ )
- Start: each node $n \in C F G$ has an initial $I N$ set, called $\mathrm{IN}_{0}[\mathrm{n}]$
- When $F$ is monotone, for each $n$, successive values of $\operatorname{IN}[n]$ form a non-decreasing sequence.
- Any chain starting at $x \in L$ has at most $c_{x}$ elements
- $x=I N[n]$ can increase in value at most $c_{x}$ times
- Then $\mathrm{C}=\max _{n \in C F G} \mathrm{c}_{I N[n]}$ is finite
- On every iteration, at least one $\operatorname{IN}[$.] set must increase in value
- If loop executes $\mathrm{N} \times \mathrm{C}$ times, all IN[.] sets would be T
- The algorithm terminates in $\mathbf{O}(\mathbf{N} \times \mathbf{C})$ steps (but this is conservative)


## Speed of Convergence

How quickly does the transfer function stabilize over backedge?
If the lattice has ascending chain property, then $\forall f \in F, \forall x \in$ $L f^{[k]}$ stabilizes, where

$$
f^{[k]}=\bigwedge_{i=0 . . k} f^{i}(x) \quad \text { where } f^{0}=x, f^{i}=f \circ f^{i-1}(x)
$$

F is bounded if for all $f$, the chain $\left\{f^{[k]}\right\}$ is finite, k , bounded if $\mathrm{k} \geq$ length
K-boundness: $f^{k} \geq f^{[k]}$ (if $L$ has height $k$, then $F$ will be $k$-bounded)
Fast: (2-bounded) $f \circ f \geq f \wedge x$
Rapid (1-semibound): $\forall f \in F, \forall x, y \in L . f(x) \leq y \wedge x \wedge f(y)$ which ends up being $\forall f \in F, \forall x \in L . x \leq f(x) \wedge f(T)$

## Speed of Convergence

Loop Connectedness $\mathrm{d}(\mathrm{G})$ : for a reducible CFG G , it is the maximum number of back edges in any acyclic path in G .

## Kam \& Ullman, 1976:

- The depth-first version of the iterative algorithm halts in at most $\mathrm{d}(\mathrm{G})+3$ passes over the graph
- If the lattice $L$ has $T$, at most $d(G)+2$ passes are needed

In practice:

- $\mathrm{d}(\mathrm{G})<3$, so the algorithm makes less than 6 passes over the graph

For mode details, see also Properties of data flow frameworks, Marlowe and Ryder (1990)

## General Worklist Algorithm

 (Reminder)for each $n$ do out $_{n}:=f_{n}(\perp)$

$$
\begin{aligned}
& \text { in }_{n \theta}:=I ; \text { out }_{n \theta}:=f_{n \theta}(I) \\
& \text { worklist }:=N-\left\{n_{\theta}\right\}
\end{aligned}
$$

while worklist $\neq \varnothing$ do remove a node n from worklist in $_{n}:=\vee\left\{\right.$ out $_{m} . m$ in pred(n) \} out $_{n}:=f_{n}\left(\right.$ in $\left._{n}\right)$
if out ${ }_{n}$ changed then

$$
\text { worklist }:=\text { worklist } \cup \text { succ(n) }
$$

## Reaching Definitions Algorithm (Reminder)

```
for all nodes n in N
    OUT[n] = emptyset; // OUT[n] = GEN[n];
IN[Entry] = emptyset;
OUT[Entry] = GEN[Entry];
Changed = N - { Entry }; // N = all nodes in graph
while (Changed != emptyset)
    choose a node n in Changed;
    Changed = Changed - { n };
    IN[n] = emptyset;
    for all nodes p in predecessors(n)
    IN[n] = IN[n] U OUT[p];
    OUT[n] = GEN[n] U (IN[n] - KILL[n]);
    if (OUT[n] changed)
    for all nodes s in successors(n)
    Changed = Changed U { s };
```


## Reaching Definitions

## General Worklist

```
for all nodes n in N
    OUT[n] = emptyset;
IN[Entry] = emptyset;
OUT[Entry] = GEN[Entry];
Changed = N - { Entry };
while (Changed != emptyset)
    choose a node n in Changed;
    Changed = Changed - { n };
    IN[n] = emptyset;
    for all nodes p in predecessors(n)
        IN[n] = IN[n] U OUT[p];
    OUT[n] = GEN[n] U (IN[n] - KILL[n]); out m := fon(inn)
    if (OUT[n] changed)
        for all nodes s in succ(n)
        Changed = Changed U { s };
```

```
for each n do out 
inne
worklist := N - { n n }
while worklist }\not=\varnothing\varnothing\mathrm{ do
    remove a node n from worklist
    in
    if outn changed then
        worklist := worklist \cup succ(n)
```


## Reaching Definitions

$P=$ powerset of set of all definitions in program (all subsets of set of definitions in program)
$\checkmark=\cup$ (order is $\subseteq$ )
$\perp=\varnothing$
$\mathrm{I}=\mathrm{in}_{\mathrm{n} 0}=\perp$
$F=$ all functions $f$ of the form $f(x)=a \cup(x-b)$

- $b$ is set of definitions that node kills
- a is set of definitions that node generates

General pattern for many transfer functions

- $f(x)=G E N \cup(x-K I L L)$


## Does Reaching Definitions Framework Satisfy Properties?

## $\subseteq$ satisfies conditions for $\leq$

- Reflexivity: $\mathrm{x} \subseteq \mathrm{x}$
- Antisymmetry: $\mathrm{x} \subseteq \mathrm{y}$ and $\mathrm{y} \subseteq \mathrm{x}$ implies $\mathrm{y}=\mathrm{x}$
- Transitivity: $x \subseteq y$ and $y \subseteq z$ implies $x \subseteq z$


## F satisfies transfer function conditions

- Identity: $\lambda x . ~ \varnothing \cup(x-\varnothing)=\lambda x . x \in F$
- Distributivity: Will show $f(x \cup y)=f(x) \cup f(y)$

$$
\begin{aligned}
f(x) \cup f(y) & =(a \cup(x-b)) \cup(a \cup(y-b)) \\
& =a \cup(x-b) \cup(y-b)=a \cup((x \cup y)-b) \\
& =f(x \cup y)
\end{aligned}
$$

## Does Reaching Definitions Framework Satisfy Properties?

## What about composition of $F$ ?

Given $f_{1}(x)=a_{1} \cup\left(x-b_{1}\right)$ and $f_{2}(x)=a_{2} \cup\left(x-b_{2}\right)$ we must show $f_{1}\left(f_{2}(x)\right)$ can be expressed as a $\cup(x-b)$

$$
\begin{aligned}
f_{1}\left(f_{2}(x)\right) & =a_{1} \cup\left(\left(a_{2} \cup\left(x-b_{2}\right)\right)-b_{1}\right) \\
& =a_{1} \cup\left(\left(a_{2}-b_{1}\right) \cup\left(\left(x-b_{2}\right)-b_{1}\right)\right) \\
& \left.=\left(a_{1} \cup\left(a_{2}-b_{1}\right)\right) \cup\left(\left(x-b_{2}\right)-b_{1}\right)\right) \\
& =\left(a_{1} \cup\left(a_{2}-b_{1}\right)\right) \cup\left(x-\left(b_{2} \cup b_{1}\right)\right)
\end{aligned}
$$

- Let $a=\left(a_{1} \cup\left(a_{2}-b_{1}\right)\right)$ and $b=b_{2} \cup b_{1}$
- Then $f_{1}\left(f_{2}(x)\right)=a \cup(x-b)$


## General Result

All GEN/KILL transfer function frameworks satisfy the three properties:

- Identity
- Distributivity
- Composition

And all of them converge rapidly

