# CS 477: Dataflow Analysis and Abstract Interpretation

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### **Partial Orders**

### Set P

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

- $\mathbf{x} \leq \mathbf{x}$
- $x \le y$  and  $y \le x$  implies x = y
- $x \le y$  and  $y \le z$  implies  $x \le z$

Can use partial order to define

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

(reflexive)

(antisymmetric) (transitive)

### **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
  - $x \le y$  for all upper bounds y of S
- v join, least upper bound, lub, supremum, sup
  - ${\bf \lor}$  S is the least upper bound of S
  - $x \lor y$  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
  - $y \le x$  for all lower bounds y of S
- A meet, greatest lower bound, glb, infimum, inf
  - $\land$  S is the greatest lower bound of S
  - $x \land y$  is the greatest lower bound of  $\{x,y\}$

# Covering

x < y if  $x \le y$  and  $x \ne y$ 

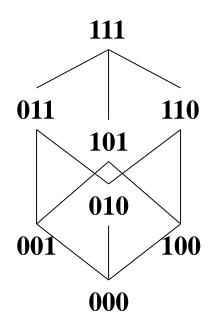
x is covered by y (y covers x) if

- x < y, and
- $x \le z < y$  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)  $x \le y \text{ is equivalent to } (x \text{ bitwise-and } y) = x$ 



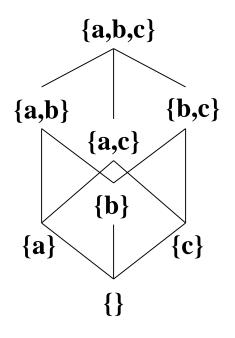
#### Hasse Diagram

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

### **Example: Same as**

(standard powerset lattice, also called hypercube)

 $x \le y$  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 (P,  $\leq$ ) and the operators  $\land$  (meet) and  $\lor$  (join)

If for all  $x,y \in P$  there exist  $x \land y$  and  $x \lor y$ , then P is a **lattice**. If for all  $S \subseteq P$  there exist  $\land S$  and  $\lor S$ then P is a <u>complete</u> lattice. All finite lattices are complete

Example of a lattice that is not complete: Integers Z

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

### **Top and Bottom**

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

- $\forall a \in L . a \lor T = T$
- Note:  $\forall a \in L$ .  $a \leq T$  and  $T \land a = a$

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

- $\forall a \in L. a \land \bot = \bot$
- Note:  $\forall a \in L . \perp \leq a \text{ and } \perp \lor a = a$

### Lattice (Recap)

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

- Set
- Partial order
- Meet
- Join
- Bottom
- Тор

# **Connection Between** $\leq$ , $\land$ , and $\lor$

**Theorem:** The following 3 properties are equivalent:

- $x \le y$
- $\mathbf{x} \lor \mathbf{y} = \mathbf{y}$
- $\mathbf{x} \wedge \mathbf{y} = \mathbf{x}$

Let's prove:

- $x \le y$  implies  $x \lor y = y$  and  $x \land y = x$
- $x \lor y = y$  implies  $x \le y$
- $x \land y = x$  implies  $x \le y$

Then by transitivity, we can obtain

- $x \lor y = y$  implies  $x \land y = x$
- $x \land y = x$  implies  $x \lor y = y$

# **Connecting Lemma Proofs**

#### Lemma: $x \le y$ implies $x \lor y = y$

Proof:

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

#### Lemma: $x \le y$ implies $x \land y = x$

Proof:

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

### **Connecting Lemma Proofs**

#### Lemma: $x \lor y = y$ implies $x \le y$

Proof:

• y is an upper bound of  $\{x,y\}$  implies  $x \le y$ 

#### Lemma: $x \land y = x$ implies $x \le y$

Proof:

• x is a lower bound of  $\{x,y\}$  implies  $x \le y$ 

# Lattices as Algebraic Structures

We have previously defined  $\lor$  and  $\land$  in terms of  $\leq$ We will now define  $\leq$  in terms of  $\lor$  and  $\land$ 

- Start with v and A 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  $\lor$  and  $\land$  as information combination operators (or, and) or set operations (union, intersection)

# **Algebraic Properties of Lattices**

Assume arbitrary operations  $\lor$  and  $\land$  such that

- $(\mathbf{x} \lor \mathbf{y}) \lor \mathbf{z} = \mathbf{x} \lor (\mathbf{y} \lor \mathbf{z})$
- $(x \land y) \land z = x \land (y \land z)$  (associativity of  $\land$ )
- $\mathbf{X} \lor \mathbf{y} = \mathbf{y} \lor \mathbf{X}$
- $\mathbf{X} \wedge \mathbf{y} = \mathbf{y} \wedge \mathbf{X}$
- $\mathbf{X} \lor \mathbf{X} = \mathbf{X}$
- $X \wedge X = X$
- $\mathbf{x} \lor (\mathbf{x} \land \mathbf{y}) = \mathbf{x}$
- $\mathbf{x} \wedge (\mathbf{x} \vee \mathbf{y}) = \mathbf{x}$

- (associativity of  $\vee$ )
- (commutativity of  $\vee$ )
- (commutativity of  $\land$ )
- (idempotence of  $\lor$ )
- (idempotence of  $\land$ )
- (absorption of  $\lor$  over  $\land$ ) (absorption of  $\land$  over  $\lor$ )

### Connection Between $\land$ and $\lor$

Thm:  $x \lor y = y$  if and only if  $x \land y = x$ 

Proof ('if'): 
$$x \lor y = y \implies x = x \land y$$
  
 $x = x \land (x \lor y)$  (by absorption)  
 $= x \land y$  (by assumption)

Proof ('only if'):  $x \land y = x \implies y = x \lor y$   $y = y \lor (y \land x)$  (by absorption)  $= y \lor (x \land y)$  (by commutativity)  $= y \lor x$  (by assumption)  $= x \lor y$  (by commutativity)

### **Properties of** $\leq$

Define:  $x \le y$  if  $x \lor y = y$ Thm :  $x \le y$  is a partial order

Proof of transitive property. Must show that  $x \lor y = y$  and  $y \lor z = z$  implies  $x \lor z = z$   $x \lor z = x \lor (y \lor z)$  (by assumption)  $= (x \lor y) \lor z$  (by associativity)  $= y \lor z$  (by assumption) = z (by assumption)

### Properties of $\leq$

Proof of asymmetry property. Must show that

$$x \lor y = y$$
 and  $y \lor x = x$  implies  $x = y$ 

- $x = y \lor x$  (by assumption)
  - $= x \lor y$  (by commutativity)
  - = y (by assumption)

Proof of reflexivity property. Must show that

- $x \lor x = x$ , which follows directly
- $x \lor x = x$  (by idempotence)

### Properties of $\leq$

Induced operation  $\leq$  agrees with original definitions of  $\lor$  and  $\land$ , i.e.,

- x ∨ y = sup {x, y}
- x ∧ y = inf {x, y}

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

Consider any upper bound u for x and y. Given  $x \lor u = u$  and  $y \lor u = u$ , must show  $x \lor y \le u$ , i.e.,  $(x \lor y) \lor u = u$   $u = x \lor u$  (by assumption)  $= x \lor (y \lor u)$  (by assumption)  $= (x \lor y) \lor u$  (by associativity)

# **Proof of x** ∧ **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.,  $(x \wedge y) \wedge L = L$   $L = x \wedge L$  (by assumption)  $= x \wedge (y \wedge L)$  (by assumption)  $= (x \wedge y) \wedge L$  (by associativity)

# Semi-lattice (P, ∧)

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

- $\mathbf{x} \wedge \mathbf{x} = \mathbf{x}$
- $x \land y = y \land x$  implies x = y
- $(x \land y) \land z = x \land (y \land z)$

(commutative)

(idempotent)

(associative)

The operation  $\land$  imposes a partial order on P

#### If ((L, $\leq$ ), $\land$ , $\lor$ ) is a lattice, then

- (L,  $\land$ ) is a **meet semi-lattice**
- (L,  $\lor$ ) 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  $x_1 \le x_2 \le \dots$  there exists n such that  $x_n = x_{n+1} = \dots$ 

- When a particular ascending chain has the property that x<sub>n</sub> = x<sub>n+1</sub> = ... 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  $(L_1, \leq_1)$  and  $(L_2, \leq_2)$  be posets. Then  $(L^*, \leq^*)$  is also a poset, where  $L^* = \{ (l_1, l_2) \mid l_1 \in L_1, l_2 \in L_2 \}$  and  $(l_{11}, l_{21}) \leq^* (l_{12}, l_{22})$ 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  $\bot^* = (\bot_1, \bot_2)$ , we define

 $glb(Y) = (glb \{ l_1 | (l_{1,-}) \in Y \}, glb \{ l_2 | (_, l_2) \in Y \}$ and same for lub and  $T^*$ 

See Nielsen, Nielsen and Hankin book

### From one variable to more

#### Total function space (S -> L) :

Let  $(L, \leq)$  be a poset, S a set and f <u>total function</u>. Then  $(L^f, \leq^f)$  is also a poset, where

 $L^f = \{f: S \to L\} \text{ and } f' \leq^f f'' \text{ iff } \forall s \in S \, . \, f'(s) \leq f''(s).$ 

To extend to lattices, we define  $\bot^f = \lambda s \bot$  and  $glb(Y) = \lambda s \bot glb_0 \{ f(s) \mid f \in Y \}$  and same for lub and  $\top^f$ 

#### Monotone Function Space $(L_1 \rightarrow L_2)$ :

Let  $(L_1, \leq_1)$  and  $(L_2, \leq_2)$  be posets and f monotone. Then  $(L^f, \leq^f)$  is also a poset, where  $\perp^f = \lambda s \cdot \perp_2$  and

 $L^{f} = \{f: L_{1} \to L_{2}\} \text{ and } f' \leq^{f} f'' \text{ iff } \forall l_{1} \in L_{1} \text{ . } f'(l_{1}) \leq_{2} f''(l_{1})$ 

# **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  $\lor$  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 ∈ F must be monotonic:
   x ≤ y implies f(x) ≤ f(y)
- Sometimes all f ∈ F are distributive\*:
   f(x ∨ y) = f(x) ∨ f(y)
- Note that Distributivity implies monotonicity

\*One can also define distributivity in terms of  $\land$  ("meet"): f(x  $\land$  y) = f(x)  $\land$  f(y)

### **Distributivity Implies Monotonicity**

### Proof.\*

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

Must show:  $x \lor y = y$  implies  $f(x) \lor f(y) = f(y)$   $f(y) = f(x \lor y)$  (by assumption)  $= f(x) \lor f(y)$  (by distributivity)

\*For  $f(x \land y) = f(x) \land f(y)$ , show  $x \land y = x \Rightarrow f(x) \land f(y) = f(x)$ ;  $f(x) = f(x \land y) = f(x) \land f(y)$ 

# **Knaster-Tarsky Fixed-point Theorem**

Let:

- (L,  $\leq$ ,  $\land$ ,  $\lor$ ,  $\top$ ,  $\bot$ ) 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  $\land$ ,  $\lor$  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 ⊥, f(⊥), f(f(⊥)), f(f(f(⊥))), ..., f<sup>n</sup>(⊥)

### **Putting the Pieces Together...**

### **Forward Dataflow Analysis**

Simulates execution of program forward with flow of control

Tuple (G, (L, ≤), F, I) – (graph, (lattice), transfer fs., initial val.)

For each node  $n \in G$ , we have

- in<sub>n</sub> value at program point before n
- out<sub>n</sub> value at program point after n
- $f_n \in \mathbf{F}$  transfer function for n (given in<sub>n</sub>, computes out<sub>n</sub>)
- Signature of  $in_n$ ,  $out_n$ ,  $f_n : L \rightarrow L$

Requires that solution satisfies

- $\forall n.$  out<sub>n</sub> = f<sub>n</sub>(in<sub>n</sub>)
- $\forall n \neq n_0$ .  $in_n = \lor \{out_m . m in pred(n)\}$
- in<sub>n0</sub> = I, summarizes information at the start of program

### **Dataflow Equations**

Compiler processes program to obtain a set of dataflow equations

out<sub>n</sub> := 
$$f_n(in_n)$$
  
in<sub>n</sub> :=  $\lor$  { out<sub>m</sub> . for each m in pred(n) }

Conceptually separates analysis problem from program

### Worklist Algorithm for Solving Forward Dataflow Equations

for each n do out<sub>n</sub> :=  $f_n(\bot)$ 

in<sub>n0</sub> := I; out<sub>n0</sub> := f<sub>n0</sub>(I)
worklist := N - { n<sub>0</sub> }

while worklist ≠ Ø do
 remove a node n from worklist
 in<sub>n</sub> := ∨ { out<sub>m</sub> . m in pred(n) }
 out<sub>n</sub> := f<sub>n</sub>(in<sub>n</sub>)
 if out<sub>n</sub> changed then
 worklist := worklist ∪ succ(n)

### **Correctness Argument**

Why does the result satisfy dataflow equations?

- Whenever it processes a node n, algorithm sets out<sub>n</sub> := f<sub>n</sub>(in<sub>n</sub>)
   Therefore, the algorithm ensures that out<sub>n</sub> = f<sub>n</sub>(in<sub>n</sub>)
- Whenever out<sub>m</sub> changes, it puts succ(m) on worklist. Consider any node n ∈ succ(m). It will eventually come off worklist and algorithm will set

 $in_n := \lor \{ out_m . m in pred(n) \}$ to ensure that  $in_n = \lor \{ out_m . m in pred(n) \}$ 

- 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 IN<sub>n</sub> or OUT<sub>n</sub> 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 (L, ≤)
- Start: each node  $n \in CFG$  has an initial IN set, called  $IN_0[n]$
- When F is **monotone**, for each n, successive values of IN[n] form a non-decreasing sequence.
  - Any chain starting at  $x \in L$  has at most  $c_x$  elements
  - x=IN[n] can increase in value at most c<sub>x</sub> times
  - Then  $C = \max_{n \in CFG} c_{IN[n]}$  is finite
- On every iteration, at least one IN[.] set must increase in value
  - If loop executes N × C times, all IN[.] sets would be  $\top$
  - The algorithm terminates in O(N × 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

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

F is bounded if for all f, the chain  $\{f^{[k]}\}$  is finite, k, bounded if  $k \ge \text{length}$ **K-boundness:**  $f^k \ge f^{[k]}$  (if L has height k, then F will be k-bounded) **Fast: (2-bounded)**  $f \circ f \ge f \land x$ 

**Rapid** (1-semibound):  $\forall f \in F, \forall x, y \in L . f(x) \le y \land x \land f(y)$ which ends up being  $\forall f \in F, \forall x \in L . x \le f(x) \land f(T)$ 

# **Speed of Convergence**

**Loop Connectedness** d(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 d(G) + 3 passes over the graph
- If the lattice L has T, at most d(G) + 2 passes are needed

#### In practice:

• d(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<sub>n</sub> :=  $f_n(\bot)$ 

in<sub>n0</sub> := I; out<sub>n0</sub> := f<sub>n0</sub>(I)
worklist := N - { n<sub>0</sub> }

while worklist ≠ Ø do
 remove a node n from worklist
 in<sub>n</sub> := ∨ { out<sub>m</sub> . m in pred(n) }
 out<sub>n</sub> := f<sub>n</sub>(in<sub>n</sub>)
 if out<sub>n</sub> changed then
 worklist := worklist ∪ 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] \cup OUT[p];
      OUT[n] = GEN[n] \cup (IN[n] - KILL[n]);
      if (OUT[n] changed)
          for all nodes s in successors(n)
          Changed = Changed U { s };
```

### **Reaching Definitions**

```
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] \cup OUT[p];
  OUT[n] = GEN[n] U (IN[n] - KILL[n]);
  if (OUT[n] changed)
    for all nodes s in succ(n)
       Changed = Changed U { s };
```

### **General Worklist**

for each n do out<sub>n</sub> :=  $f_n(\perp)$ 

```
in_{n0} := I; out_{n0} := f_{n0}(I)
worklist := N - { n_0 }
```

```
while worklist \neq \emptyset do
remove a node n from worklist
```

```
in_n := \vee \{ out_m . m in pred(n) \}
```

```
out<sub>n</sub> := f<sub>n</sub>(in<sub>n</sub>)
```

if out<sub>n</sub> changed then
 worklist := worklist ∪ succ(n)

# **Reaching Definitions**

P = powerset of set of all definitions in program (all subsets of set of definitions in program)

- $\vee$  =  $\cup$  (order is  $\subseteq$ )
- $\perp$  = Ø
- $I = in_{n0} = \bot$
- 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

•  $f(x) = GEN \cup (x-KILL)$ 

### **Does Reaching Definitions Framework** Satisfy Properties?

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

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

### **F** satisfies transfer function conditions

- Identity:  $\lambda x. \emptyset \cup (x \emptyset) = \lambda x. x \in F$
- Distributivity: Will show  $f(x \cup y) = f(x) \cup f(y)$   $f(x) \cup f(y) = (a \cup (x - b)) \cup (a \cup (y - b))$   $= a \cup (x - b) \cup (y - b) = a \cup ((x \cup y) - b)$  $= f(x \cup y)$

### **Does Reaching Definitions Framework** Satisfy Properties?

### What about composition of F?

Given  $f_1(x) = a_1 \cup (x-b_1)$  and  $f_2(x) = a_2 \cup (x-b_2)$ we must show  $f_1(f_2(x))$  can be expressed as  $a \cup (x - b)$  $f_1(f_2(x)) = a_1 \cup ((a_2 \cup (x-b_2)) - b_1)$  $= a_1 \cup ((a_2 - b_1) \cup ((x-b_2) - b_1))$  $= (a_1 \cup (a_2 - b_1)) \cup ((x-b_2) - b_1))$  $= (a_1 \cup (a_2 - b_1)) \cup (x-(b_2 \cup b_1))$ 

• Let  $a = (a_1 \cup (a_2 - b_1))$  and  $b = b_2 \cup b_1$ 

• Then  $f_1(f_2(x)) = a \cup (x - b)$ 

### **General Result**

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

- Identity
- Distributivity
- Composition

And all of them converge rapidly