CS 432 Fall 2024

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Data-Flow Analysis

Compilers

Optimization

```
 int a;
  a = 0;while (a < 10) {
       a = a + 1;
  }
  loadI \theta \Rightarrow r1loadI 10 \Rightarrow r211:cmp_LT r1, r2 \Rightarrow r4cbr r4 \implies l2, l3l2:
  addI r1, 1 \Rightarrow r1 jump l1
l3:
  storeAI r1 \Rightarrow [bp-4]
```

```
loadI 0 \Rightarrow r1storeAI r1 \Rightarrow [bp-4]
l1:
  loadAI [bp-4] \Rightarrow r2loadI 10 => r3
  cmp_LT r2, r3 \Rightarrow r4cbr r4 \implies l2, l3l2:
  loadAI [bp-4] => r5loadI 1 \Rightarrow r6add r5, r6 \Rightarrow r7storeAI r7 \Rightarrow [bp-4]
   jump l1
l3:
```

```
loadI 10 => r1
storeAI r1 \Rightarrow [bp-4]
```
Optimization is Hard

- **Problem**: it's hard to reason about all possible executions
	- Preconditions and inputs may differ
	- Optimizations should be correct and efficient in all cases
- Optimization tradeoff: investment vs. payoff
	- "Better than naïve" is fairly easy
	- "Optimal" is impossible
	- Real world: somewhere in between
		- Better speedups with more static analysis
		- Usually worth the added compile time
- Also: linear IRs (e.g., ILOC) don't explicitly expose control flow
	- This makes analysis and optimization difficult
	- Need to re-introduce some representation of possible control flow

Aside: Verifying Returns in P3

- Is is tempting to try to verify that functions end with a return statement in P3, but this is not possible with a naive approach
- Consider cases like this:

```
def int foo(bool x)
\{ // other code here
    if (x) {
          return 5;
     } else {
          return 10;
     }
}
```
This is **guaranteed** to be **safe** (every path has a return statement) but requires nontrivial, non-local static analysis to verify (i.e., can't just check the last statement in the function)

Control-Flow Graphs

• Basic blocks

- "Maximal-length sequence of branch-free code"
- "Atomic" sequences (instructions that always execute together)
- Control-flow graph (CFG) *(note overloaded acronym)*
	- Nodes/vertices for basic blocks
	- Edges for control transfer
		- Branch/jump instructions (explicit) or fallthrough (implicit)
		- p is a predecessor of q if there is a path from p to q
			- p is an immediate predecessor if there is an edge directly from p to q
		- q is a successor of p if there is a path from p to q
			- q is an immediate successor if there is an edge directly from p to q

Control-Flow Graphs

- \bullet Conversion: linear IR to CFG
	- Find leaders (initial instruction of a basic block) and build blocks
		- In ILOC, every call or jump target is a leader
	- Add edges between blocks based on jumps/branches and fallthrough
	- Complicated by indirect jumps (none in our ILOC!)

Static CFG Analysis

- Single block analysis is easy, and trees are too
- General CFGs are harder
	- Which branch of a conditional will execute?
	- How many times will a loop execute?
- How do we handle this?
	- One method: iterative data-flow analysis
	- Simulate all possible paths through a region of code
	- "Meet-over-all-paths" conservative solution
	- Meet operator combines information across paths

Semilattices

- In general, a semilattice is a set of values L, special values ⊤ (top) and \perp (bottom), and a meet operator \wedge such that
	- $-$ a \geq b iff a \wedge b = b
	- a > b iff a ≥ b and a ≠ b
	- a^{\wedge} ⊤ = a for all $a \in L$
	- $-$ a \wedge \perp = \perp for all a \in L
- **Partial ordering**
	- Monotonic

Figure 9.22 from Dragon book: semilattice of definitions using U (set union) as the meet operation

Constant propagation

- For sparse simple constant propagation (SSCP), the lattice is very shallow
	- $-$ c_i \wedge \top = c_i for all c_i
	- $-$ c_i ^ \perp = \perp for all c_i

$$
- c_i^{\prime} \wedge c_j = c_i^{\prime} \text{ if } c_i = c_j^{\prime}
$$

$$
c_i \wedge c_j = \perp
$$
 if $c_i \neq c_j$

- Basically: each SSA value is either unknown (⊤), a known constant (c,), or it is a variable (\perp)
	- Initialize to unknown $($ T) for all SSA values
	- Interpret operations over lattice values (always lowering)
	- Propagate information until convergence

Constant propagation example

Data-Flow Analysis

- Define properties of interest for basic blocks
	- Usually **sets** of blocks, variables, definitions, etc.
- Define a formula for how those properties change within a block
	- F(B) is based on F(A) where A is a predecessor or successor of B
	- This is basically the *meet* operator for a particular problem
- Specify initial information for all blocks
	- Entry/exit blocks usually have special initial values
- Run an *iterative update algorithm to propagate changes*
	- Keep running until the properties converge for all basic blocks
- Key concept: finite descending chain property
	- Properties must be monotonically increasing or decreasing
	- Otherwise, termination is not guaranteed

Data-Flow Analysis

- This kind of algorithm is called fixed-point iteration
	- It runs until it converges to a "fixed point"
- Forward vs. backward data-flow analysis
	- Forward: along graph edges (based on predecessors)
	- Backward: reverse of forward (based on successors)
- Particular data-flow analyses:
	- Constant propagation
	- Dominance
	- Liveness
	- Available expressions
	- Reaching definitions
	- Anticipable expressions

Review: Set Theory

A∪ B
A ∩ B

Dominance

- Block A dominates block B if A is on every path from the entry to B
	- Block A immediately dominates block B if there are no blocks between them
	- Block B postdominates block A if B is on every path from A to an exit
	- Every block both dominates and postdominates itself
- Simple dataflow analysis formulation
	- *preds*(b) is the set of blocks that are immediate predecessors of block b
	- *Dom*(b) is the set of blocks that dominate block b
		- intersection of *Dom* for all immediate predecessors (*p* in *preds(b)*)
	- *PostDom*(b) is the set of blocks that postdominate block b
		- (similar definition using immediate successors)

Initial conditions : $Dom(**entry**) = {**entry**}$ $\forall b \neq$ **entry**, $Dom(b) = \{$ **all** blocks $\}$

$$
\text{Updates: } \text{Dom}(b) = \{b\} \cup \bigcap_{p \in \text{preds}(b)} \text{Dom}(p)
$$

Dominance example

Initial conditions: *Dom*(*entry*) = {*entry* } $\forall b \neq$ **entry**, $Dom(b) = \{$ **all** blocks $\}$

> $Updates: Dom(b) = {b} \cup \bigcap_{p \in grad(b)} Dom(p)$ *p*∈ *preds*(*b*)

Liveness

- Variable *v* is live at point *p* if there is a path from *p* to a use of *v* with no intervening assignment to *v*
	- Useful for finding uninitialized variables (live at function entry)
	- Useful for optimization (remove unused assignments)
	- Useful for register allocation (keep live vars in registers)
- Initial information: *UEVar* and *VarKill*
	- *UEVar*(B): variables read in B before any corresponding write in B
		- ("upwards exposed" variables)
	- *VarKill*(B): variables that are written to ("killed") in B
- Textbook notation note: $X \cap \overline{Y} = X Y$

Initial conditions: $\forall b$, *LiveOut* $(b) = \emptyset$

 $Updates: LiveOut(b) = \bigcup_{s \in suoc(b)} UEVar(s) \cup (LiveOut(s) - VarKill(s))$ *s*∈ *succs*(*b*)

Liveness example

(c) Initial Information

 ∇b , *LiveOut*(*b*) = \emptyset *LiveOut*(*b*) = $\bigcup_{s \in \text{crys}(b)} \text{UEVar}(s) \cup (\text{LiveOut}(s) - \text{VarKill}(s))$ *s*∈ *succs*(*b*)

Alternative definition

- Define LiveIn as well as LiveOut
	- Two formulas for each basic block
	- Makes things a bit simpler to reason about
		- Separates change *within* block from change *between* blocks

 $\forall b, \text{LiveIn}(b) = \emptyset, \text{LiveOut}(b) = \emptyset$

 $LiveIn(b) = UEVar(b) \cup (LiveOut(b) - VarKill(b))$ $LiveOut(b) = \bigcup_{(b)} LiveIn(s)$ $s \in$ *succs* (b)

Liveness example

 $\forall b, \text{LiveIn}(b) = \emptyset, \text{LiveOut}(b) = \emptyset$

loadAI $[bp-4] \Rightarrow r1$ cbr $r1 \implies l1, l2$ loadI $5 \Rightarrow r2$ $\begin{array}{c|c|c|c|c|c|c} \text{loadI} & \text{10 =} & \text{12} \\ \text{jump} & \text{13} & & \end{array}$ storeAI $r2 \Rightarrow$ [bp-4] LiveIn $(foo) = \{\}$ LiveOut(foo) = $\{ \}$ LiveIn $(l1) = \{\}$ LiveOut $(l1)$ = $\{r2\}$ LiveIn $(l2) = \{\}$ $LiveOut(12) = {r2}$ LiveIn $(l3) = \{r2\}$ LiveOut(13) = $\{$ } foo $\frac{11}{2}$ $\frac{1}{2}$ $\frac{12}{2}$ $l3$ $LiveIn(b) = UEVar(b) \cup (LiveOut(b) - VarKill(b))$ $LiveOut(b) = \bigcup_{s \in succ(b)}LiveIn(s)$ *s*∈ *succs*(*b*) $UEVar(foo) = \{\}$ $VarKill(foo) = {r1}$ $UEVar(11) = \{\}$ $VarKill(11) = \{r2\}$ $UEVar(12) = { }$ $Varkill(12) = {r2}$

 $UEVar(13) = {r2}$ $VarKill(13) = \{\}$

Block orderings

- Forwards dataflow analyses converge faster with reverse postorder processing of CFG blocks
	- Visit as many of a block's predecessors as possible before visiting that block
	- Strict reversal of normal postorder traversal
	- Similar to concept of topological sorting on DAGs
	- NOT EQUIVALENT to preorder traversal!
	- Backwards analyses should use reverse postorder on reverse CFG

Depth-first search:

A, B, D, B, A, **C**, A (left first) **D, B, C, A** (left first) **A, C, D,** C**,** A**, B,** A (right first)

Valid *postorderings*:

D, C, B, A (right first)

Valid *preorderings*:

A, B, D, C (left first) **A, C, D, B** (right first)

- Valid *reverse postorderings*:
- **A, C, B, D A, B, C, D**

Summary

 $Dom(b) = \{b\} \cup \bigcap_{p \in \text{pred}(b)} Dom(p)$ *p*∈ *preds*(*b*) *Dom*(*entry*) = {*entry* } ∀ *b*≠*entry , Dom*(*b*) = {*all blocks* } **Dominance**

$$
\forall b, \text{ LiveOut}(b) = \emptyset
$$
\n
$$
\text{LiveOut}(b) = \bigcup_{s \in success(b)} \text{UEVar}(s) \cup (\text{LiveOut}(s) - \text{VarKill}(s))
$$
\n**LiveOut**

\n(EAC version)

 $LiveIn(b) = UEVar(b) \cup (LiveOut(b) - VarKill(b))$ $LiveOut(b) = \bigcup_{s \in cures(b)}LiveIn(s)$ *s*∈ *succs*(*b*) **Liveness (Dragon version)** $\forall b, \text{LiveIn}(b) = \emptyset, \text{LiveOut}(b) = \emptyset$