Lecture 21

EECS 483: COMPILER CONSTRUCTION

Announcements

- HW5: OAT v. 2.0
 - Fully released now
 - Due on Friday, April 12
- Guest lectures
 - Lectures on Optimization and Dataflow analysis Eric
 - After that TBA
 - No class on April 8 (eclipse)

CODE ANALYSIS

Liveness information

Consider this program:

```
int f(int x) {
    int a = x + 2;
    int b = a * a;
    int c = b + x;
    return c;
}

    x is live
    a and x are live
    b and x are live
    c is live
```

- The scopes of a,b,c,x all overlap they're all in scope at the end of the block.
- But, a, b, c are never live at the same time.
 - So they can share the same stack slot / register

Live Variable Analysis

- A variable v is *live* at a program point if v is defined before the program point and used after it.
- Liveness is defined in terms of where variables are *defined* and where variables are *used*
- Liveness analysis: Compute the live variables between each statement.
 - May be conservative (i.e. it may claim a variable is live when it isn't) so because that's a safe approximation
 - To be useful, it should be more *precise* than simple scoping rules.
- Liveness analysis is one example of dataflow analysis
 - Other examples: Available Expressions, Reaching Definitions, Constant-Propagation Analysis, ...

Control-flow Graphs Revisited

- For the purposes of dataflow analysis, we use the control-flow graph (CFG) intermediate form.
- Recall that a basic block is a sequence of instructions such that:
 - There is a distinguished, labeled entry point (no jumps into the middle of a basic block)
 - There is a (possibly empty) sequence of non-control-flow instructions
 - The block ends with a single control-flow instruction (jump, conditional branch, return, etc.)
- A control flow graph
 - Nodes are blocks
 - There is an edge from B1 to B2 if the control-flow instruction of B1 might jump to the entry label of B2
 - There are no "dangling" edges there is a block for every jump target.

Note: the following slides are intentionally a bit ambiguous about the exact nature of the code in the control flow graphs:

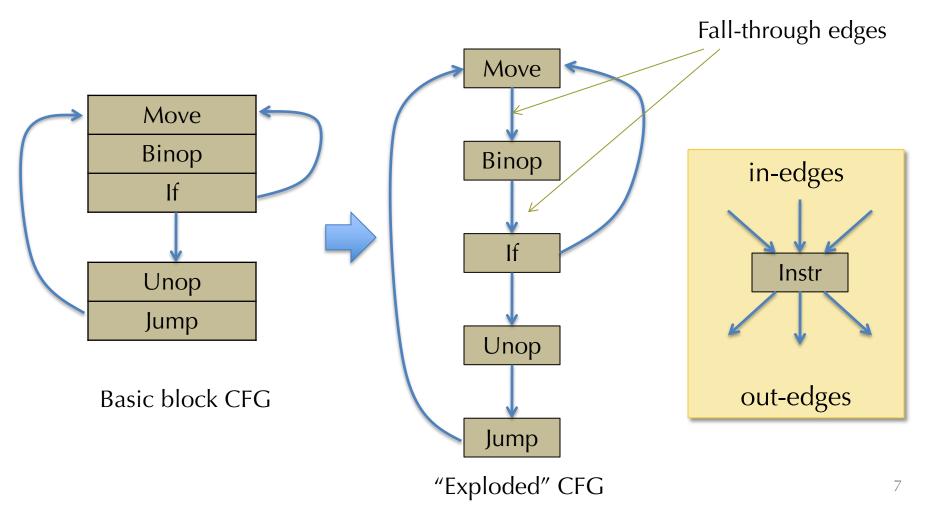
an "imperative" C-like source level at the x86 assembly level the LLVM IR level

Each setting applies the same general idea, but the exact details will differ.

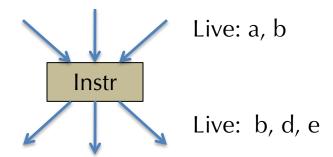
• e.g., LLVM IR doesn't have "imperative" update of %uid temporaries. (The SSA structure of the LLVM IR (by design!) makes some of these analyses simpler.)

Dataflow over CFGs

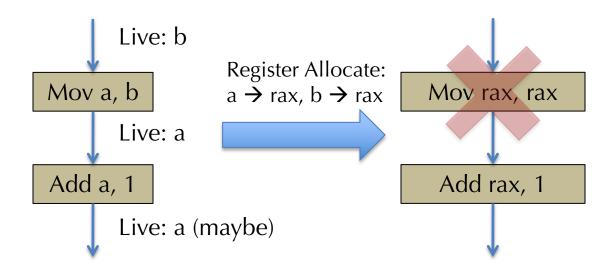
- For precision, it is helpful to think of the "fall through" between sequential instructions as an edge of the control-flow graph too.
 - Different implementation tradeoffs in practice...



Liveness is Associated with Edges



- This is useful so that the same register can be used for different temporaries in the same statement.
- Example: a = b + 1
- Compiles to:



Uses and Definitions

- Every instruction/statement uses some set of variables
 - i.e. reads from them
- Every instruction/statement *defines* some set of variables
 - i.e. writes to them
- For a node/statement s define:
 - use[s] : set of variables used by s
 - def[s] : set of variables defined by s
- Examples:
 - a = b + c $use[s] = \{b,c\}$ $def[s] = \{a\}$
 - a = a + 1 $use[s] = {a}$ $def[s] = {a}$

Liveness, Formally

- A variable v is *live* on edge e if:
 There is
 - a node n in the CFG such that use[n] contains v, and
 - a directed path from e to n such that for every statement s' on the path, def[s'] does not contain v
- The first clause says that v will be used on some path starting from edge e.
- The second clause says that v won't be redefined on that path before the use.
- Questions:
 - How to compute this efficiently?
 - How to use this information (e.g. for register allocation)?
 - How does the choice of IR affect this?
 (e.g. LLVM IR uses SSA, so it doesn't allow redefinition ⇒ simplify liveness analysis)

Simple, inefficient algorithm

- "A variable v is live on an edge e if there is a node n in the CFG using it and a directed path from e to n pasing through no def of v."
- Backtracking Algorithm:
 - For each variable v...
 - Try all paths from each use of v, tracing backwards through the controlflow graph until either v is defined or a previously visited node has been reached.
 - Mark the variable v live across each edge traversed.

• Inefficient because it explores the same paths many times (for different uses and different variables)

Dataflow Analysis

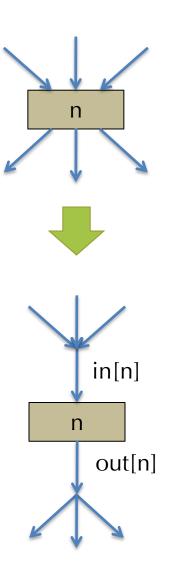
- *Idea*: compute liveness information for all variables simultaneously.
 - Keep track of sets of information about each node
- Approach: define equations that must be satisfied by any liveness determination.
 - Equations based on "obvious" constraints.
- Solve the equations by iteratively converging on a solution.
 - Start with a "rough" approximation to the answer
 - Refine the answer at each iteration
 - Keep going until no more refinement is possible: a fixpoint has been reached
- This is an instance of a general framework for computing program properties: dataflow analysis

Dataflow Value Sets for Liveness

- Nodes are program statements, so:
- use[n] : set of variables used by n
- def[n] : set of variables defined by n
- in[n]: set of variables live on entry to n
- out[n]: set of variables live on exit from n
- Associate in[n] and out[n] with the "collected" information about incoming/outgoing edges
- For Liveness: what constraints are there among these sets?
- Clearly:

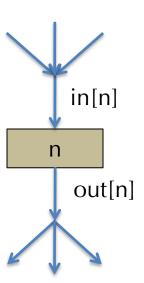
 $in[n] \supseteq use[n]$

What other constraints?



Other Dataflow Constraints

- We have: in[n] ⊇ use[n]
 - "A variable must be live on entry to n if it is used by n"
- Also: in[n] ⊇ out[n] def[n]
 - "If a variable is live on exit from n, and n doesn't define it, it is live on entry to n"
 - Note: here '-' means "set difference"
- And: out[n] ⊇ in[n'] if n' ∈ succ[n]
 - "If a variable is live on entry to a successor node of n, it must be live on exit from n."



Iterative Dataflow Analysis

- Find a solution to those constraints by starting from a rough guess.
 - Start with: $in[n] = \emptyset$ and $out[n] = \emptyset$
- The guesses don't satisfy the constraints:
 - in[n] ⊇ use[n]
 - in[n] \supseteq out[n] def[n]
 - out[n] \supseteq in[n'] if n' ∈ succ[n]
- Idea: iteratively re-compute in[n] and out[n] where forced to by the constraints.
 - Each iteration will add variables to the sets in[n] and out[n]
 (i.e. the live variable sets will increase monotonically)
- We stop when in[n] and out[n] satisfy these equations: (which are derived from the constraints above)
 - in[n] = use[n] \cup (out[n] def[n])
 - out[n] = $U_{n' \in succ[n]}in[n']$

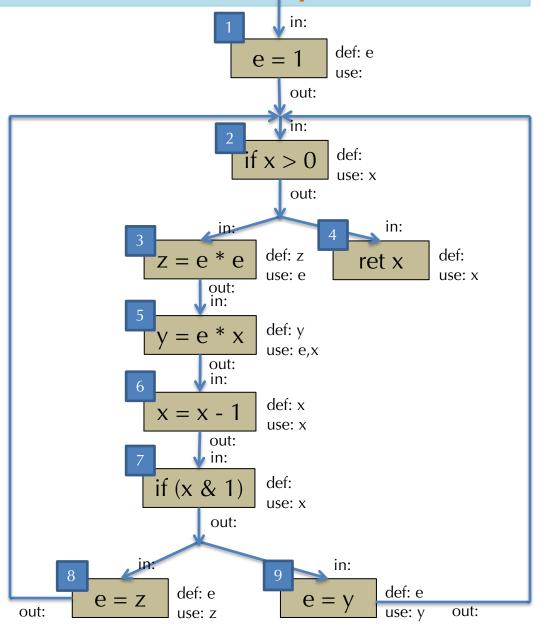
Complete Liveness Analysis Algorithm

```
for all n, in[n] := \emptyset, out[n] := \emptyset
repeat until no change in 'in' and 'out'
for all n
out[n] := \mathbf{U}_{n' \in succ[n]} in[n']
in[n] := use[n] \ \mathbf{U} \ (out[n] - def[n])
end
end
```

- Finds a fixpoint of the in and out equations.
 - The algorithm is guaranteed to terminate... Why?
- Why do we start with Ø?

• Example flow graph:

```
e = 1;
while(x>0) {
  z = e * e;
  y = e * x;
  x = x - 1;
  if (x & 1) {
     e = z;
  } else {
     e = y;
  }
}
return x;
```

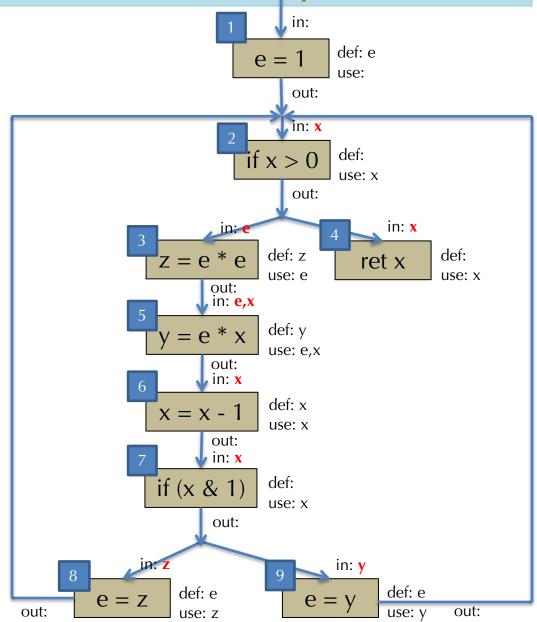


Each iteration update:

```
\label{eq:out_n} \begin{split} \text{out}[n] &:= U_{n' \in \text{succ}[n]} \text{in}[n'] \\ \text{in}[n] &:= \text{use}[n] \ U \ (\text{out}[n] - \text{def}[n]) \end{split}
```

Iteration 1:

(showing only updates that make a change)



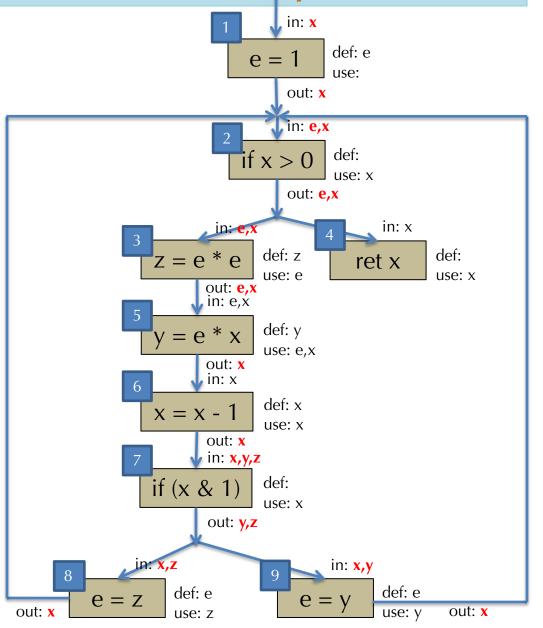
Each iteration update:

out[n] := $U_{n' \in succ[n]}$ in[n'] in[n] := use[n] U (out[n] - def[n])

• Iteration 2:

out[1] =
$$x$$

in[1] = x
out[2] = e, x
out[3] = e, x
out[3] = e, x
out[3] = e, x
out[5] = x
out[6] = x
out[7] = x, z, y
out[8] = x
in[8] = x, z
out[9] = x
in[9] = x, y



Each iteration update:

 $out[n] := U_{n' \in succ[n]} in[n']$ $in[n] := use[n] \cup (out[n] - def[n])$

• Iteration 3:

$$out[1] = e_{x}$$

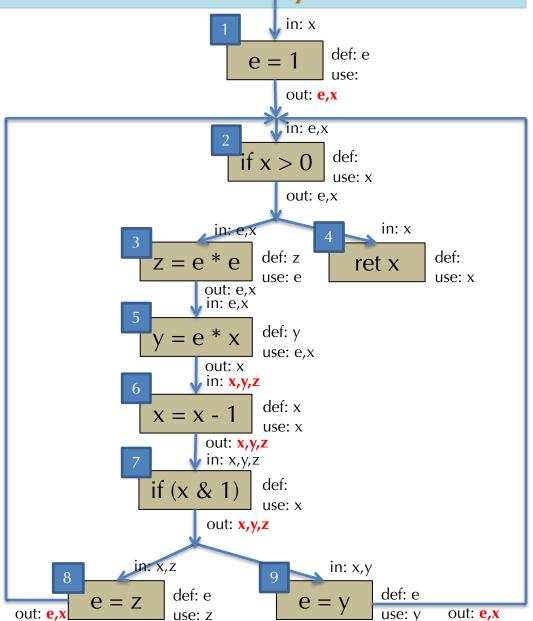
$$out[6] = x,y,z$$

$$in[6] = x,y,z$$

$$out[7] = x,y,z$$

$$out[8] = e,x$$

$$out[9] = e,x$$



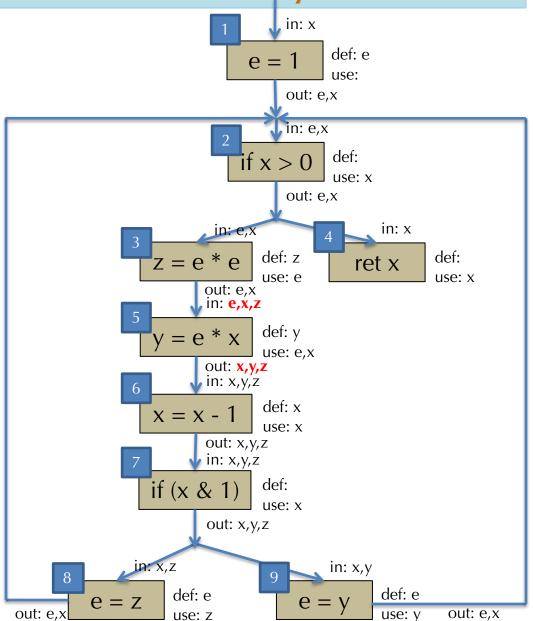
Each iteration update:

 $out[n] := U_{n' \in succ[n]} in[n']$ $in[n] := use[n] \cup (out[n] - def[n])$

• Iteration 4:

out[5]=
$$x,y,z$$

in[5]= e,x,z



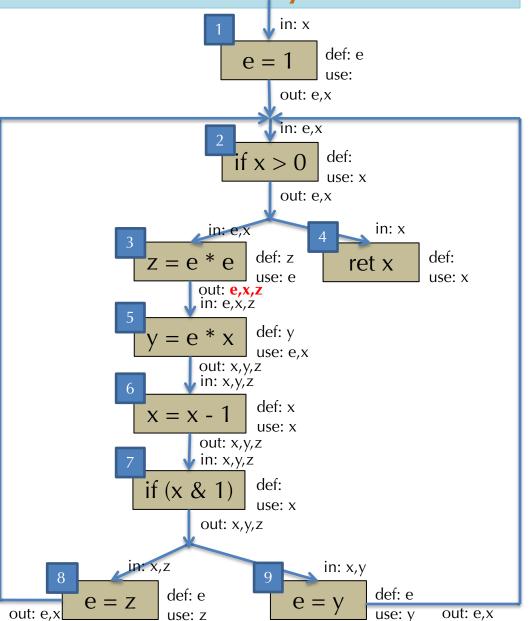
Each iteration update:

 $out[n] := U_{n' \in succ[n]} in[n']$ $in[n] := use[n] \cup (out[n] - def[n])$

• Iteration 5:

out[3] = e,x,z

Done!



Improving the Algorithm

- Can we do better?
- Observe: the only way information propagates from one node to another is using: $out[n] := U_{n' \in succ[n]} in[n']$
 - This is the only rule that involves more than one node
- If a node's successors haven't changed, then the node itself won't change.
- Idea for an improved version of the algorithm:
 - Keep track of which node's successors have changed

A Worklist Algorithm

Use a FIFO queue of nodes that might need to be updated.

```
for all n, in[n] := \emptyset, out[n] := \emptyset
w = new queue with all nodes
repeat until w is empty
   let n = w.pop()
                                             // pull a node off the queue
     old_in = in[n]
                                             // remember old in[n]
     out[n] := \mathbf{U}_{n' \in \text{succ}[n]} \text{in}[n']
     in[n] := use[n] \cup (out[n] - def[n])
     if (old_in != in[n]),
                                             // if in[n] has changed
        for all m in pred[n], w.push(m) // add to worklist
end
```

OTHER DATAFLOW ANALYSES

Generalizing Dataflow Analyses

- The kind of iterative constraint solving used for liveness analysis applies to other kinds of analyses as well.
 - Reaching definitions analysis
 - Available expressions analysis
 - Alias Analysis
 - Constant Propagation
 - These analyses follow the same 3-step approach as for liveness.

- To see these as an instance of the same kind of algorithm, the next few examples to work over a canonical intermediate instruction representation called *quadruples*
 - Allows easy definition of def[n] and use[n]
 - A slightly "looser" variant of LLVM's IR that doesn't require the "static single assignment" – i.e. it has mutable local variables
 - We will use LLVM-IR-like syntax

Quadruples

• A Quadruple sequence is just a control-flow graph (flowgraph) where each node is a quadruple:

•	Quadruple forms n:	def[n]	use[n]	description
	a = b op c	{a}	{b,c}	arithmetic
	a = load b	{a}	{b}	load
	store b, a	Ø	{b}	store
	$a = call f(b_1,, b_n)$	{a}	$\{b_1,\ldots,b_n\}$	call w/return
	call void $f(b_1,,b_n)$	Ø	$\{b_1,\ldots,b_n\}$	call no return
	br L	Ø	Ø	direct jump
	br a, L1, L2	Ø	{a}	branch
	ret a	Ø	{a}	return

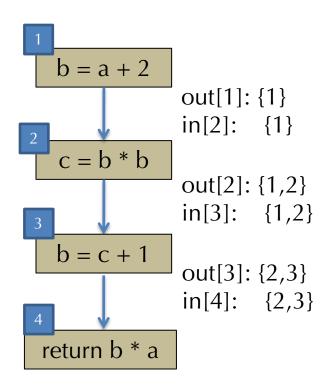
REACHING DEFINITIONS

Reaching Definition Analysis

- Question: what uses in a program does a given variable definition reach?
- This analysis is used for constant propagation & copy prop.
 - If only one definition reaches a particular use, can replace use by the definition (for constant propagation).
 - Copy propagation additionally requires that the copied value still has its same value – computed using an available expressions analysis (next)
- Input: Quadruple CFG
- Output: in[n] (resp. out[n]) is the set of nodes defining some variable such that the definition may reach the beginning (resp. end) of node n

Example of Reaching Definitions

Results of computing reaching definitions on this simple CFG:



Reaching Definitions Step 1

- Define the sets of interest for the analysis
- Let defs[a] be the set of *nodes* that define the variable a
- Define gen[n] and kill[n] as follows:

sn: gen[n]	kill[n]
{n}	defs[a] - {n}
{n}	defs[a] - {n}
Ø	Ø
{n}	defs[a] - {n}
Ø	Ø
Ø	Ø
Ø	Ø
Ø	Ø
	{n} {n} Ø

Reaching Definitions Step 2

- Define the constraints that a reaching definitions solution must satisfy.
- out[n] ⊇ gen[n]
 "The definitions that reach the end of a node at least include the definitions generated by the node"
- in[n] ⊇ out[n'] if n' is in pred[n]
 "The definitions that reach the beginning of a node include those that reach the exit of any predecessor"
- out[n] ∪ kill[n] ⊇ in[n]
 "The definitions that come in to a node either reach the end of the node or are killed by it."
 - Equivalently: out[n] ⊇ in[n] kill[n]

Reaching Definitions Step 3

- Convert constraints to iterated update equations:
- $in[n] := U_{n' \in pred[n]} out[n']$
- out[n] := $gen[n] \cup (in[n] kill[n])$
- Algorithm: initialize in[n] and out[n] to Ø
 - Iterate the update equations until a fixed point is reached
- The algorithm terminates because in[n] and out[n] increase only monotonically
 - At most to a maximum set that includes all variables in the program
- The algorithm is precise because it finds the smallest sets that satisfy the constraints.