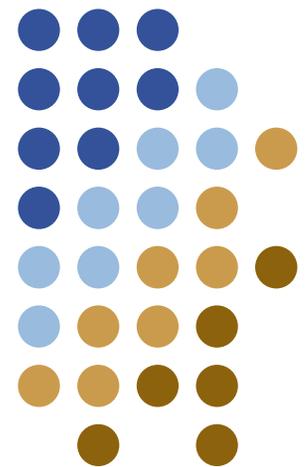
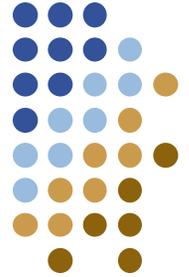


Compilers

Optimization

Yannis Smaragdakis, U. Athens
(original slides by Sam Guyer@Tufts)



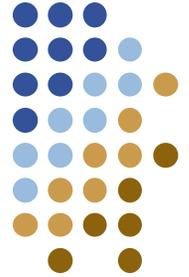


What we already saw

- Lowering
 - From language-level constructs to machine-level constructs*
- At this point we could generate machine code
 - Output of lowering is a correct translation
 - What's left to do?
 - Map from lower-level IR to actual ISA
 - Maybe some register management (*could be required*)
 - Pass off to assembler
- Why have a separate assembler?
 - Handles “packing the bits”

<i>Assembly</i>	<code>addi <target>, <source>, <value></code>
<i>Machine</i>	<code>0010 00ss ssst tttt iiii iiii iiii iiii</code>

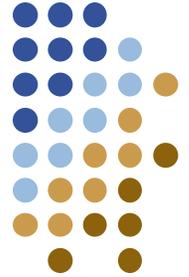




But first...

- The compiler “understands” the program
 - IR captures program semantics
 - Lowering: semantics-preserving transformation
 - Why not do others?
- Compiler optimizations
 - Oh great, now my program will be optimal!
 - Sorry, it’s a misnomer
 - What is an “optimization”?



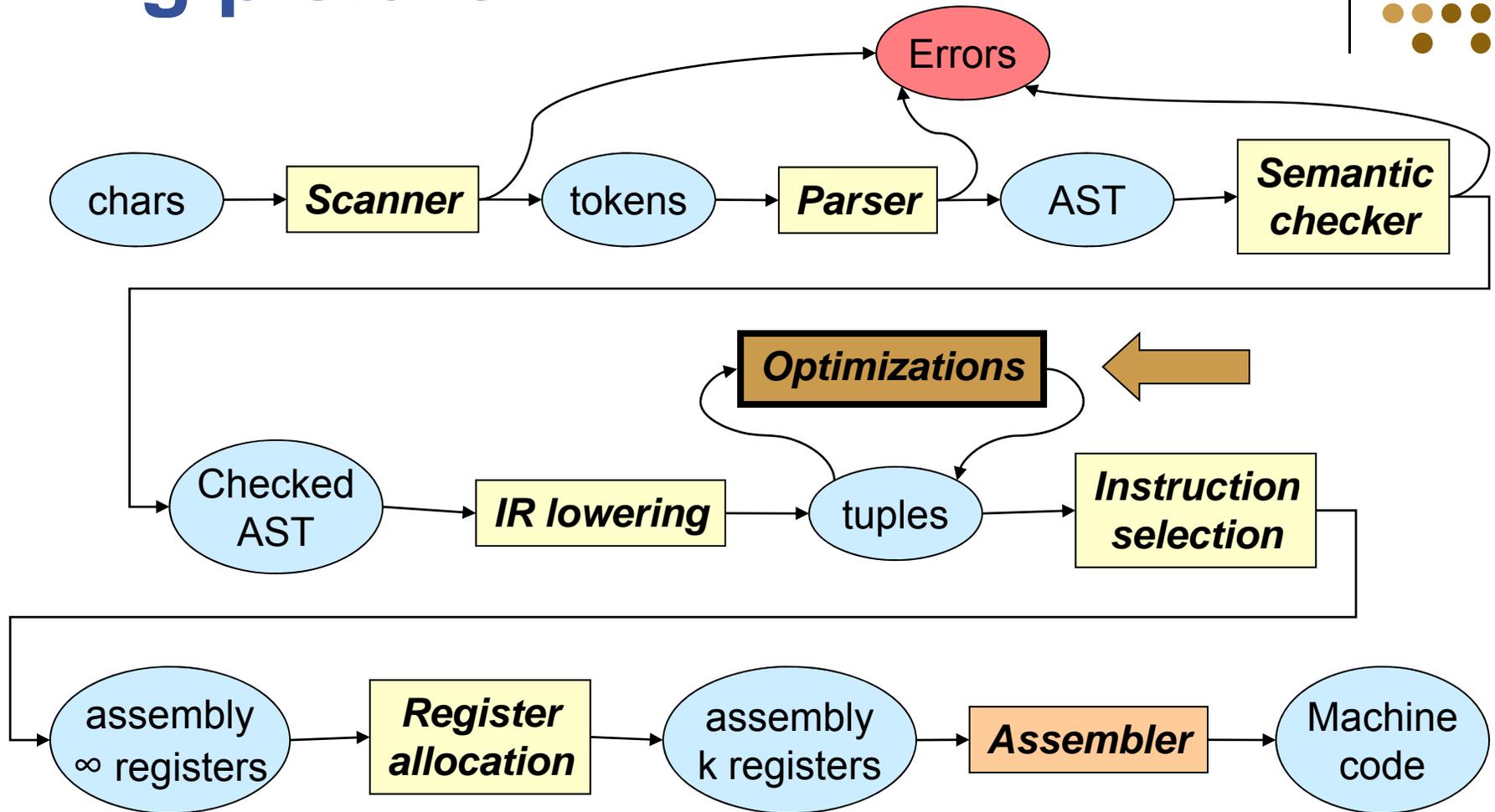
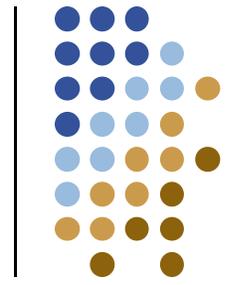


Optimizations

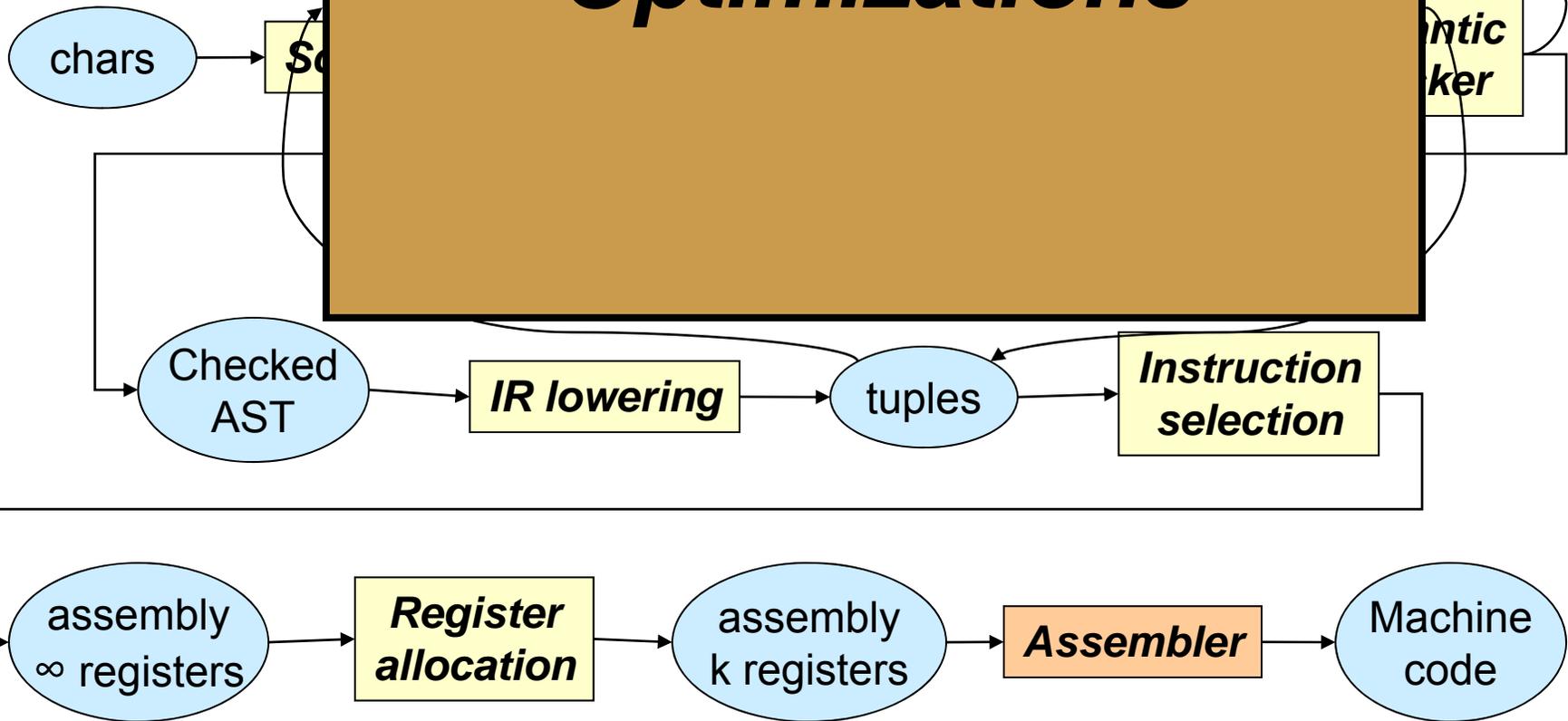
- What are they?
 - Code transformations
 - Improve some metric
- Metrics
 - Performance: time, instructions, cycles
 - Are these metrics equivalent?*
 - Memory
 - Memory hierarchy (reduce cache misses)
 - Reduce memory usage
 - Code Size
 - Energy

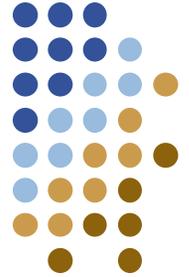


Big picture



Big picture





Why optimize?

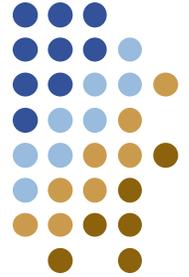
- High-level constructs may make some optimizations difficult or impossible:

```
A[i][j] = A[i][j-1] + 1
```

```
t = A + i*row + j  
s = A + i*row + j - 1  
(*t) = (*s) + 1
```

- High-level code may be more desirable
 - Program at high level
 - Focus on design; clean, modular implementation
 - Let compiler worry about gory details
- Premature optimization is the root of all evil!

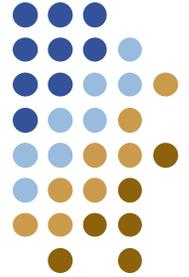




Limitations

- What are optimizers good at?
 - Consistent and thorough
 - Find all opportunities for an optimization
 - Uniformly apply the transformation
- What are they not good at?
 - Asymptotic complexity
 - Compilers can't fix bad algorithms
 - Compilers can't fix bad data structures
- There's no magic

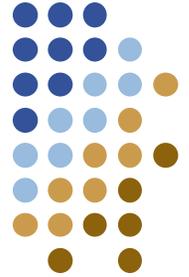




Requirements

- Safety
 - Preserve the semantics of the program
 - What does that mean?
- Profitability
 - Will it help our metric?
 - Do we need a guarantee of improvement?
- Risk
 - How will interact with other optimizations?
 - How will it affect other stages of compilation?





Example: loop unrolling

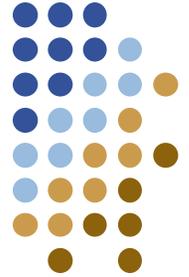
```
for (i=0; i<100; i++)  
    *t++ = *s++;
```

```
for (i=0; i<25; i++) {  
    *t++ = *s++;  
    *t++ = *s++;  
    *t++ = *s++;  
    *t++ = *s++; }  
}
```

- Safety:
 - Always safe; getting loop conditions right can be tricky.
- Profitability
 - Depends on hardware – usually a win – why?
- Risk?
 - Increases size of code in loop
 - May not fit in the instruction cache

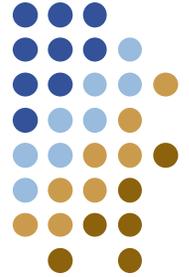


Optimizations



- Many, many optimizations invented
 - *Constant folding, constant propagation, tail-call elimination, redundancy elimination, dead code elimination, loop-invariant code motion, loop splitting, loop fusion, strength reduction, array scalarization, inlining, cloning, data prefetching, parallelization. . .etc . .*
- How do they interact?
 - Optimist: we get the sum of all improvements!
 - Realist: many are in direct opposition

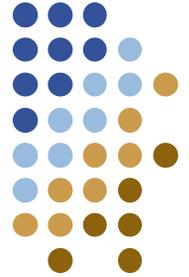




Rough categories

- Traditional optimizations
 - Transform the program to reduce work
 - Don't change the level of abstraction
- Resource allocation
 - Map program to specific hardware properties
 - Register allocation
 - Instruction scheduling, parallelism
 - Data streaming, prefetching
- Enabling transformations
 - Don't necessarily improve code on their own
 - Inlining, loop unrolling





Constant propagation

- **Idea**

- If the value of a variable is known to be a constant at compile-time, replace the use of variable with constant

```
n = 10;  
c = 2;  
for (i=0;i<n;i++)  
    s = s + i*c;
```



```
n = 10;  
c = 2;  
for (i=0;i<10;i++)  
    s = s + i*2;
```

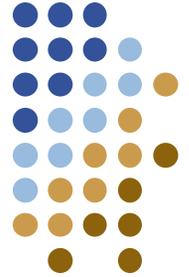
- **Safety**

- Prove the value is constant

- **Notice:**

- May interact favorably with other optimizations, like loop unrolling – now we know the *trip count*





Constant folding

- **Idea**

- If operands are known at compile-time, evaluate expression at compile-time

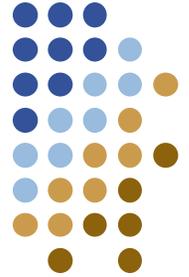
```
r = 3.141 * 10;  →  r = 31.41;
```

- What do we need to be careful of?
 - Is the result the same as if executed at runtime?
 - Overflow/underflow, rounding and numeric stability

- Often repeated throughout compiler

```
x = A[2];  →  t1 = 2*4;  
             t2 = A + t1;  
             x = *t2;
```

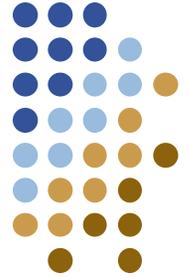




Partial evaluation

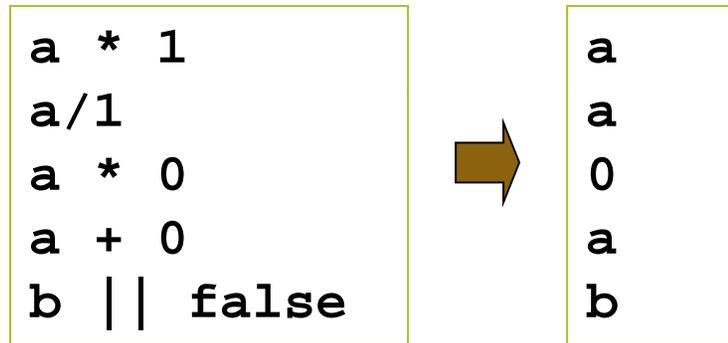
- Constant propagation and folding together
- **Idea:**
 - Evaluate as much of the program at compile-time as possible
 - More sophisticated schemes:
 - Simulate data structures, arrays
 - Symbolic execution of the code
- **Caveat: floating point**
 - Preserving the error characteristics of floating point values





Algebraic simplification

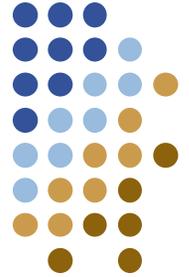
- **Idea:**
 - Apply the usual algebraic rules to simplify expressions



- Repeatedly apply to complex expressions
- Many, many possible rules
 - Associativity and commutativity come into play



Common sub-expression elimination



- **Idea:**

- If program computes the same expression multiple times, reuse the value.

```
a = b + c;  
c = b + c;  
d = b + c;
```



```
t = b + c  
a = t;  
c = t;  
d = b + c;
```

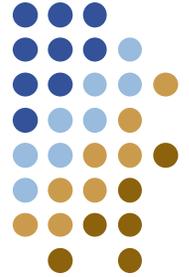
- **Safety:**

- Subexpression can only be reused until operands are redefined

- Often occurs in address computations

- Array indexing and struct/field accesses

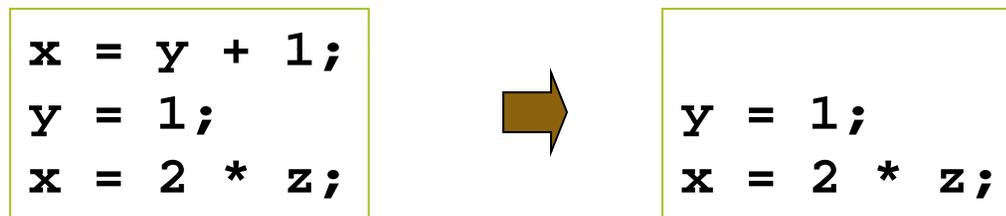




Dead code elimination

- **Idea:**

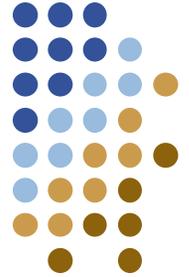
- If the result of a computation is never used, then we can remove the computation



- **Safety**

- Variable is dead if it is never used after defined
- Remove code that assigns to dead variables
- This may, in turn, create more dead code
 - Dead-code elimination usually works transitively

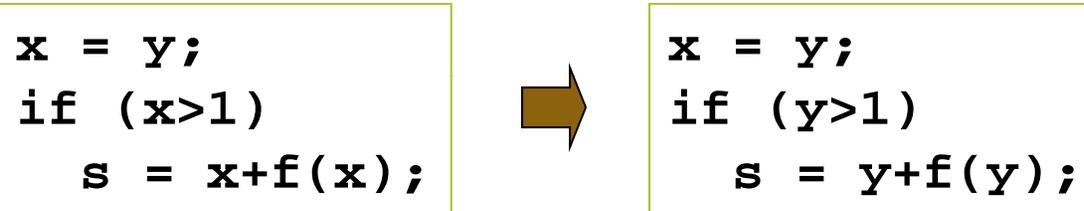




Copy propagation

- **Idea:**

- After an assignment $x = y$, replace any uses of x with y

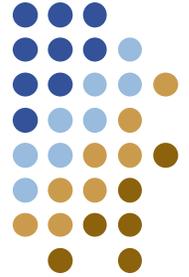


- **Safety:**

- Only apply up to another assignment to x , or
- ...another assignment to y !
- What if there was an assignment $y = z$ earlier?
 - Apply transitively to all assignments



Unreachable code elimination



- **Idea:**

- Eliminate code that can never be executed

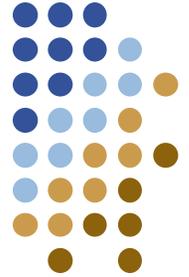
```
#define DEBUG 0
. . .
if (DEBUG)
    print("Current value = ", v);
```

- **Different implementations**

- High-level: look for if (false) or while (false)
- Low-level: more difficult
 - Code is just labels and gotos
 - Traverse the graph, marking reachable blocks

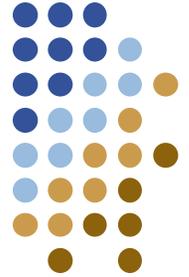


How do these things happen?



- Who would write code with:
 - Dead code
 - Common subexpressions
 - Constant expressions
 - Copies of variables
- Two ways they occur
 - High-level constructs – we've already seen examples
 - Other optimizations
 - Copy propagation often leaves dead code
 - Enabling transformations: inlining, loop unrolling, etc.

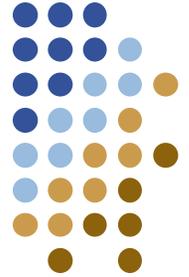




Loop optimizations

- Program hot-spots are usually in loops
 - Most programs: 90% of execution time is in loops
 - What are possible exceptions?
OS kernels, compilers and interpreters
- Loops are a good place to expend extra effort
 - Numerous loop optimizations
 - Often expensive – complex analysis
 - For languages like Fortran, very effective
 - What about C?





Loop-invariant code motion

- **Idea:**

- If a computation won't change from one loop iteration to the next, move it outside the loop

```
for (i=0;i<N;i++)  
    A[i] = A[i] + x*x;
```



```
t1 = x*x;  
for (i=0;i<N;i++)  
    A[i] = A[i] + t1;
```

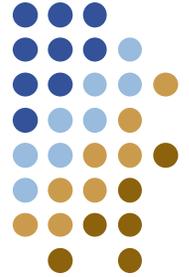
- **Safety:**

- Determine when expressions are invariant
- Just check for variables with no assignments?

- **Useful for array address computations**

- Not visible at source level





Strength reduction

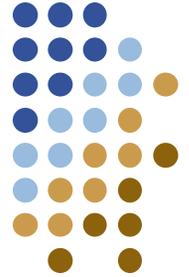
- **Idea:**
 - Replace expensive operations (multiplication, division) with cheaper ones (addition, subtraction, bit shift)
- Traditionally applied to *induction variables*
 - Variables whose value depends linearly on loop count
 - Special analysis to find such variables

```
for (i=0;i<N;i++)  
    v = 4*i;  
    A[v] = . . .
```



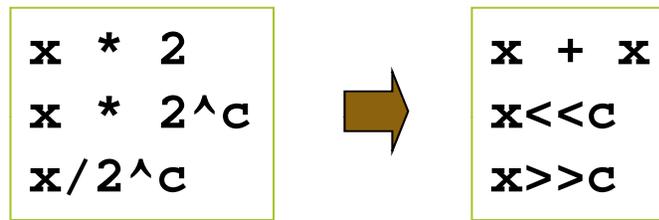
```
v = 0;  
for (i=0;i<N;i++)  
    A[v] = . . .  
    v = v + 4;
```





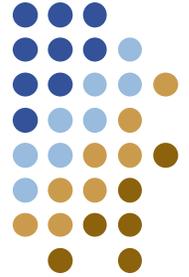
Strength reduction

- Can also be applied to simple arithmetic operations:



- Typical example of premature optimization
 - Programmers use bit-shift instead of multiplication
 - “ $x \ll 2$ ” is harder to understand
 - Most compilers will get it right automatically

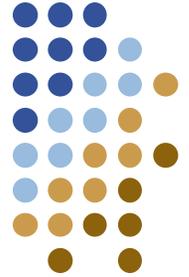




Inlining

- **Idea:**
 - Replace a function call with the body of the callee
- **Safety**
 - What about recursion?
- **Risk**
 - Code size
 - Most compilers use heuristics to decide when
 - Has been cast as a *knapsack problem*

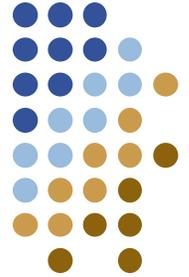




Inlining

- What are the benefits of inlining?
 - Eliminate call/return overhead
 - Customize callee code in the context of the caller
 - Use actual arguments
 - Push into copy of callee code using constant prop
 - Apply other optimizations to reduce code
 - Hardware
 - Eliminate the two jumps
 - Keep the pipeline filled
- Critical for OO languages
 - Methods are often small
 - Encapsulation, modularity force code apart





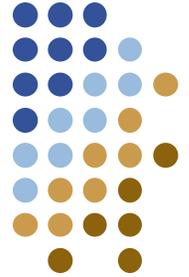
Inlining

- In C:
 - At a call-site, decide whether or not to inline
 - (Often a heuristic about callee/caller size)
 - Look up the callee
 - Replace call with body of callee
- What about Java?
 - What complicates this?
 - Virtual methods
 - Even worse?
 - Dynamic class loading

```
class A { void M() {...} }
class B extends A
        { void M() {...} }

void foo(A x)
{
    x.M(); // which M?
}
```





Inlining in Java

- With guards:

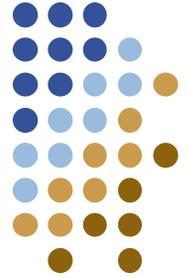
```
void foo(A x)
{
    if (x is type A)
        x.M(); // inline A's M
    if (x is type B)
        x.M(); // inline B's M
}
```

- Specialization

- At a given call, we may be able to determine the type
- Requires fancy analysis

```
y = new A();
foo(y);
z = new B();
foo(z);
```

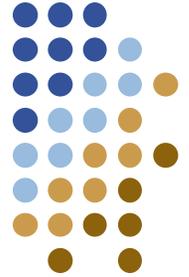




Big picture

- When do we apply these optimizations?
 - High-level:
 - Inlining, cloning
 - Some algebraic simplifications
 - Low-level
 - Everything else
- It's a black art
 - Ordering is often arbitrary
 - Many compilers just repeat the optimization passes over and over





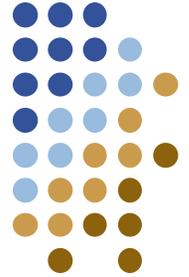
Writing fast programs

In practice:

- Pick the right algorithms and data structures
 - Asymptotic complexity and constants
 - Memory usage, memory layout, data representation
- Turn on optimization and profile
 - Run-time
 - Program counters (e.g., cache misses)
- Evaluate problems
- Tweak source code
 - Help the optimizer do “the right thing”



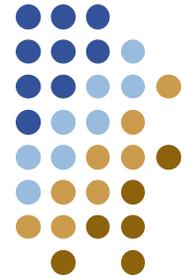
Anatomy of an optimization



Two big parts:

- Program analysis
 - Pass over code to find:*
 - Opportunities
 - Check safety constraints
- Program transformation
 - Change the code to exploit opportunity
- Often: rinse and repeat

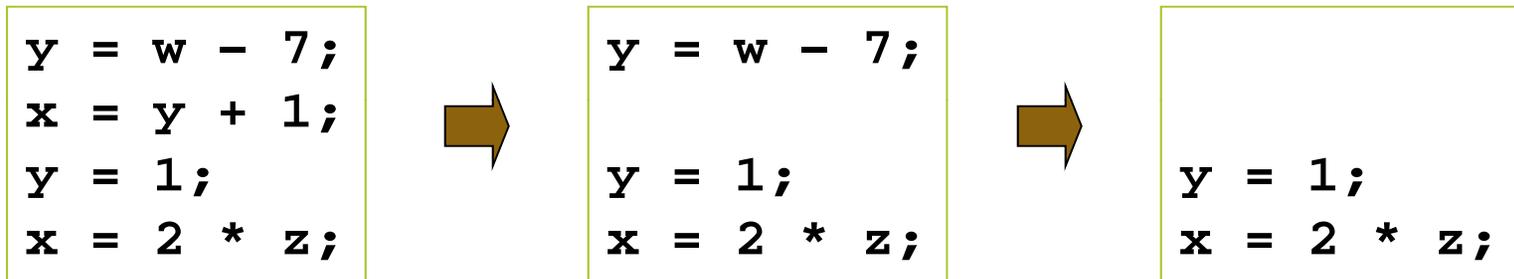




Dead code elimination

- **Idea:**

- Remove a computation if result is never used



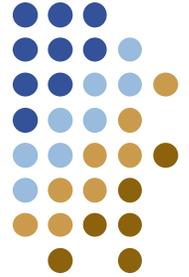
- **Safety**

- Variable is dead if it is never used after defined
- Remove code that assigns to dead variables

- This may, in turn, create more dead code

- Dead-code elimination usually works transitively





Dead code

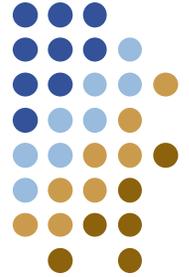
- Another example:

```
x = y + 1;  
y = 2 * z;  
x = y + z;  
z = 1;  
z = x;
```

*Which statements
can be safely
removed?*

- Conditions:
 - Computations whose value is never used
 - Obvious for straight-line code
 - What about control flow?





Dead code

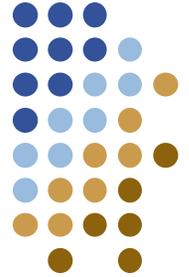
- With if-then-else:

Which statements are can be removed?

```
x = y + 1;  
y = 2 * z;  
if (c) x = y + z;  
z = 1;  
z = x;
```

- Which statements are dead code?
 - What if “c” is false?
 - Dead only on some paths through the code





Dead code

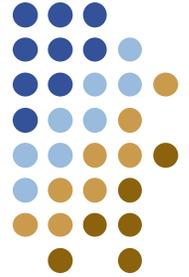
- And a loop:

*Which
statements are
can be removed?*

```
while (p) {  
    x = y + 1;  
    y = 2 * z;  
    if (c) x = y + z;  
    z = 1;  
}  
z = x;
```

- Now which statements are dead code?





Dead code

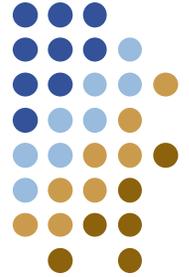
- And a loop:

*Which
statements are
can be removed?*

```
while (p) {  
    x = y + 1;  
    y = 2 * z;  
    if (c) x = y + z;  
    z = 1;  
}  
z = x;
```

- Statement “x = y+1” not dead
- What about “z = 1”?





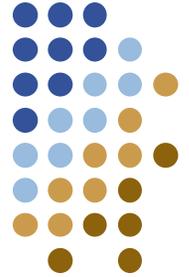
Low-level IR

- Most optimizations performed in low-level IR
 - Labels and jumps
 - No explicit loops
- Even harder to see possible paths

```
label1:  
  jumpifnot p label2  
  x = y + 1  
  y = 2 * z  
  jumpifnot c label3  
  x = y + z  
label3:  
  z = 1  
  jump label1  
label2:  
  z = x
```

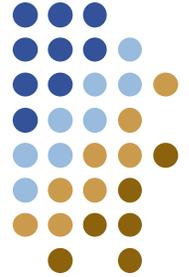


Optimizations and control flow



- Dead code is *flow sensitive*
 - Not obvious from program
 - Dead code example: are there any possible paths that make use of the value?*
 - Must characterize all possible dynamic behavior
 - Must verify conditions at compile-time
- Control flow makes it hard to extract information
 - High-level: different kinds of control structures
 - Low-level: control-flow hard to infer
- Need a unifying data structure



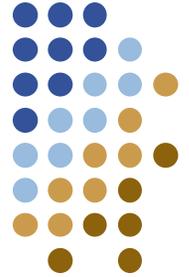


Control flow graph

- ***Control flow graph*** (CFG):
a graph representation of the program
 - Includes both computation and control flow
 - Easy to check control flow properties
 - Provides a framework for global optimizations and other compiler passes
- Nodes are ***basic blocks***
 - Consecutive sequences of non-branching statements
- Edges represent control flow
 - From jump to a label
 - Each block may have multiple incoming/outgoing edges



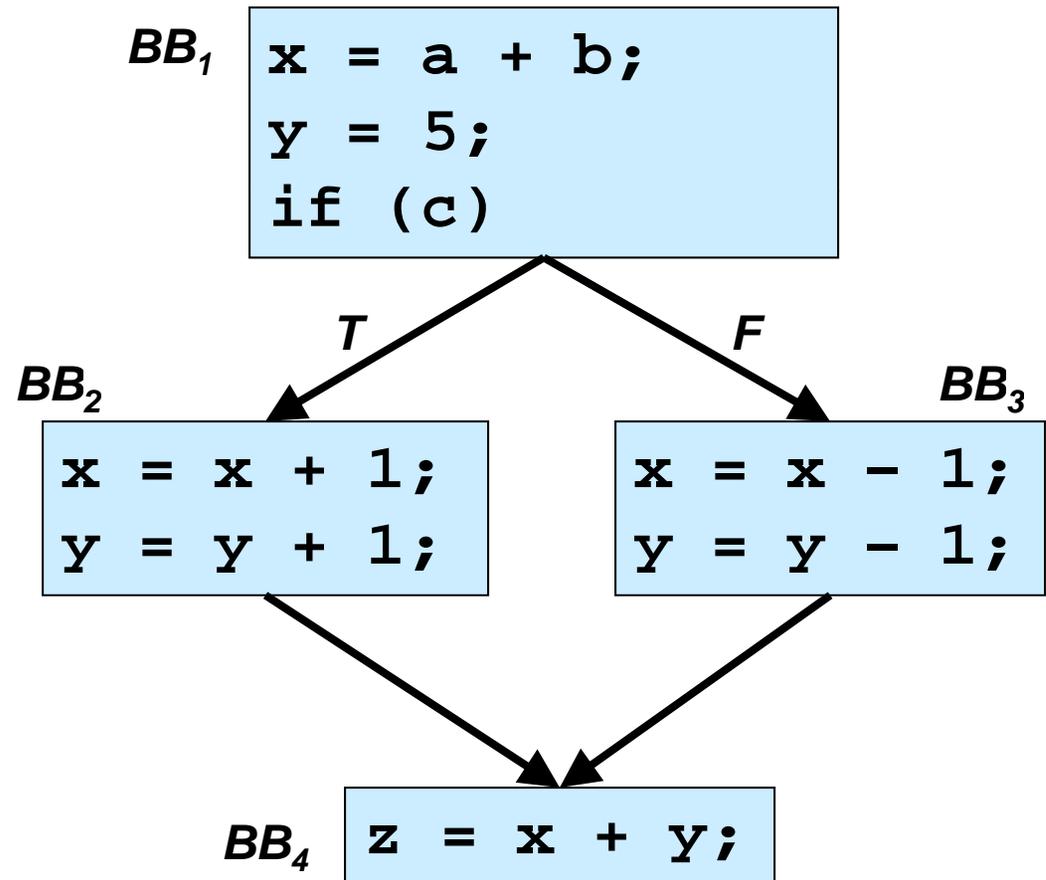
CFG Example

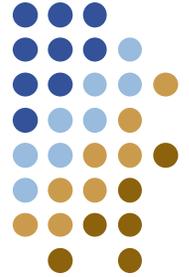


Program

```
x = a + b;  
y = 5;  
if (c) {  
    x = x + 1;  
    y = y + 1;  
} else {  
    x = x - 1;  
    y = y - 1;  
}  
z = x + y;
```

Control flow graph

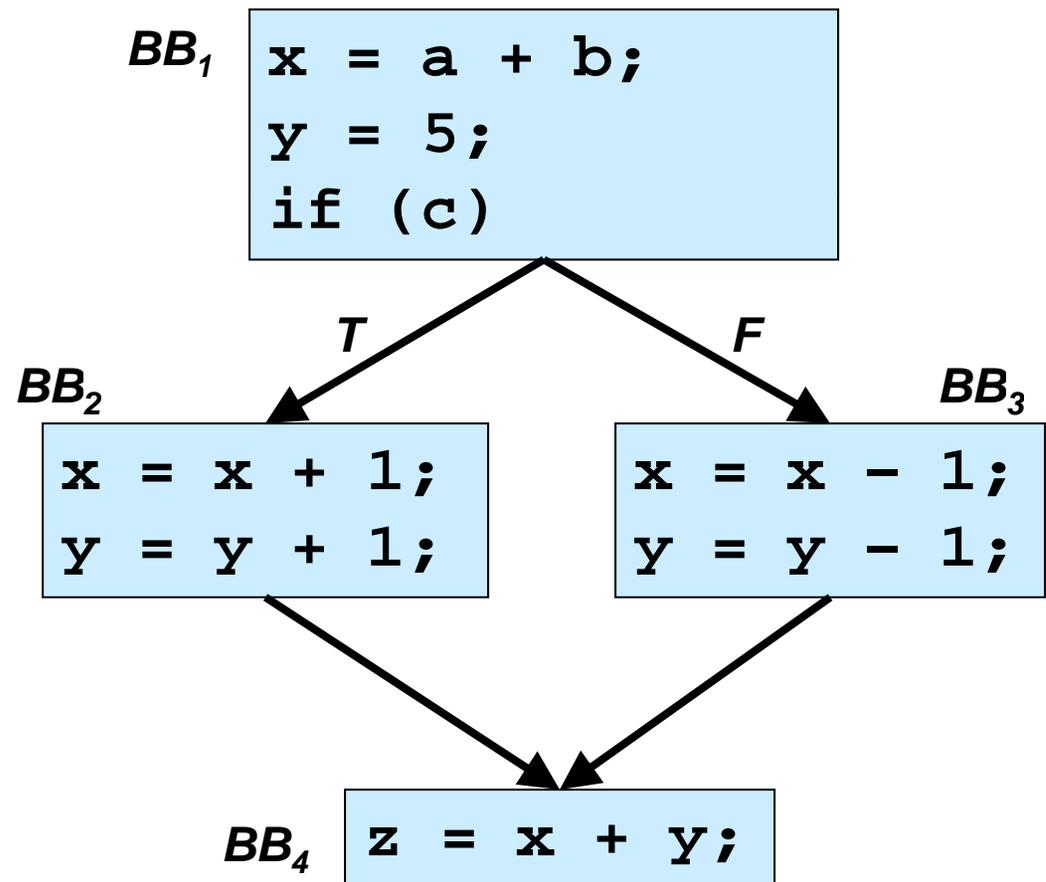




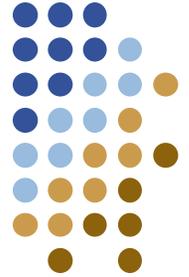
Multiple program executions

Control flow graph

- CFG models all program executions
- An actual execution is a path through the graph
- Multiple paths: multiple possible executions
 - How many?

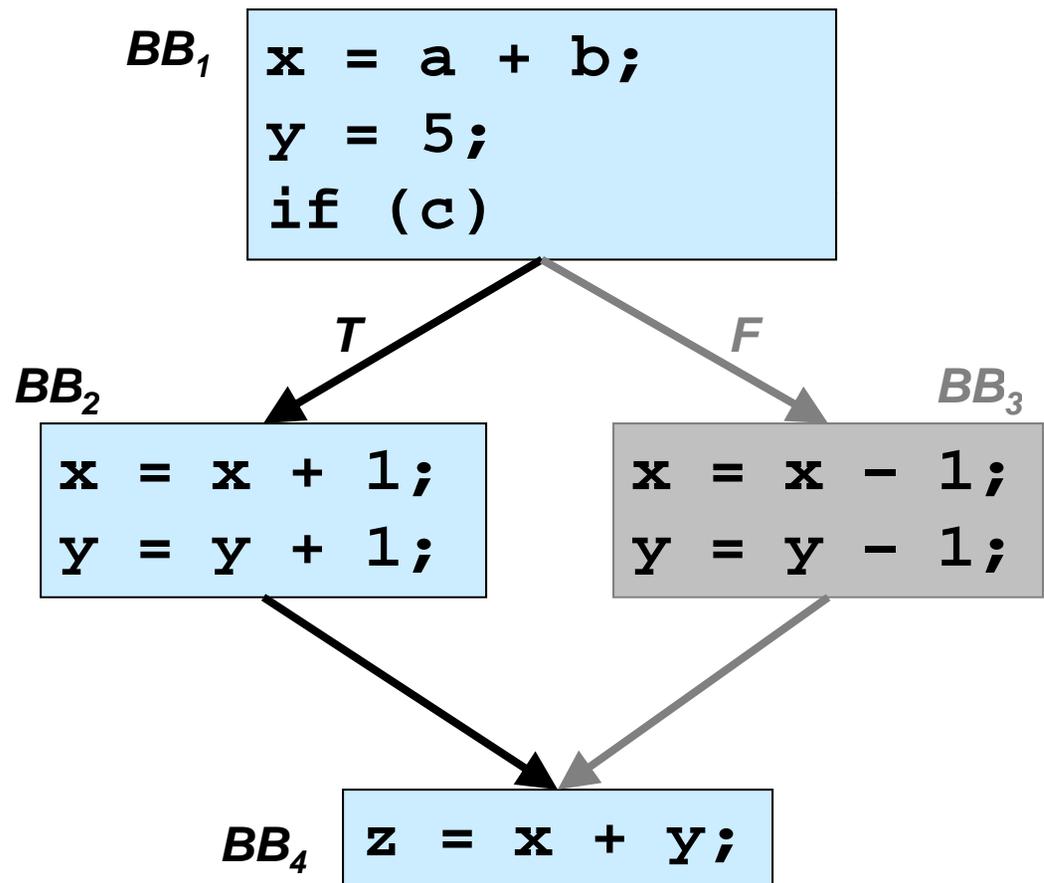


Execution 1

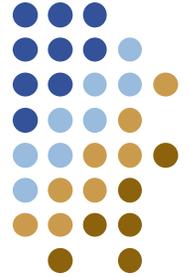


- CFG models all program executions
- Execution 1:
 - c is true
 - Program executes BB₁, BB₂, and BB₄

Control flow graph

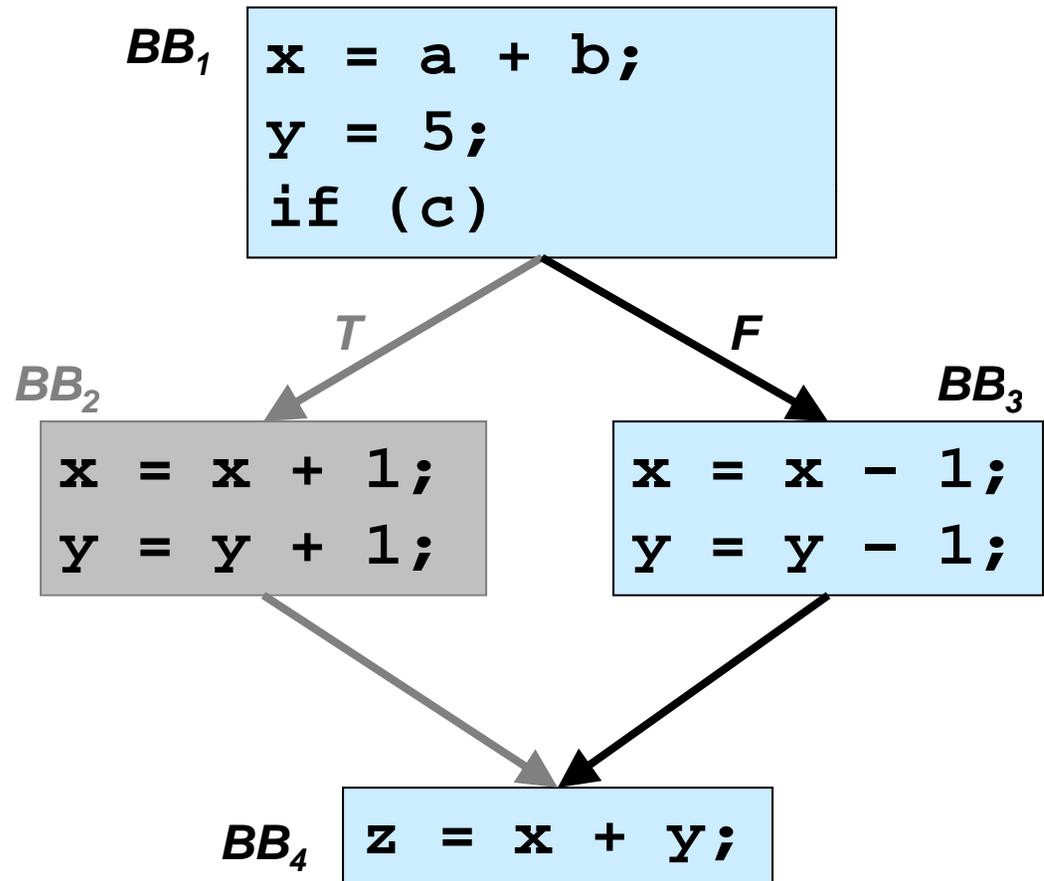


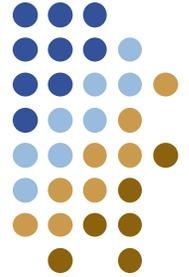
Execution 2



- CFG models all program executions
- Execution 2:
 - c is false
 - Program executes BB_1 , BB_3 , and BB_4

Control flow graph



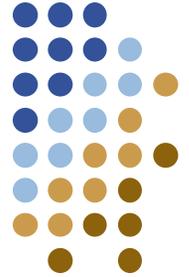


Basic blocks

- **Idea:**
 - Once execution enters the sequence, all statements (or instructions) are executed
 - Single-entry, single-exit region
- **Details**
 - Starts with a label
 - Ends with one or more branches
 - Edges may be labeled with predicates
 - May include special categories of edges*
 - Exception jumps
 - Fall-through edges
 - Computed jumps (jump tables)

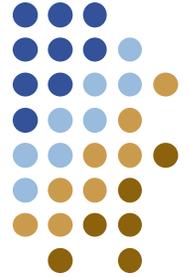


Building the CFG



- Two passes
 - First, group instructions into basic blocks
 - Second, analyze jumps and labels
- How to identify basic blocks?
 - Non-branching instructions
 - Control cannot flow out of a basic block without a jump*
 - Non-label instruction
 - Control cannot enter the middle of a block without a label*



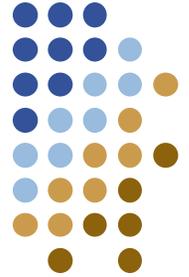


Basic blocks

- Basic block starts:
 - At a label
 - After a jump
- Basic block ends:
 - At a jump
 - Before a label

```
label1:  
jumpifnot p label2  
x = y + 1  
y = 2 * z  
jumpifnot c label3  
x = y + z  
label3:  
z = 1  
jump label1  
label2:  
z = x
```





Basic blocks

- Basic block starts:
 - At a label
 - After a jump
- Basic block ends:
 - At a jump
 - Before a label
- **Note:** order still matters

```
label1:  
jumpifnot p label2
```

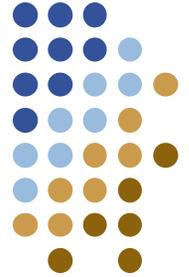
```
x = y + 1  
y = 2 * z  
jumpifnot c label3
```

```
x = y + z
```

```
label3:  
z = 1  
jump label1
```

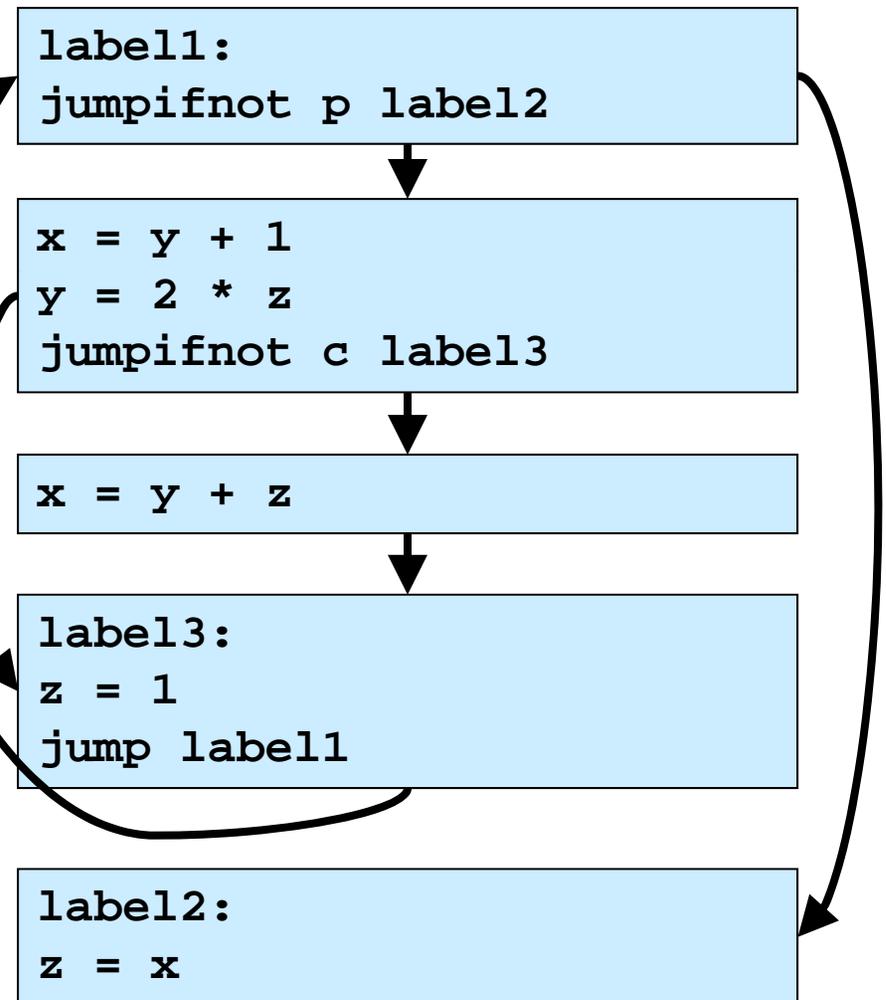
```
label2:  
z = x
```

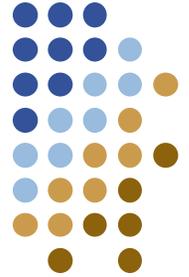




Add edges

- Unconditional jump
 - Add edge from source of jump to the block containing the label
- Conditional jump
 - 2 successors
 - One may be the fall-through block
- Fall-through

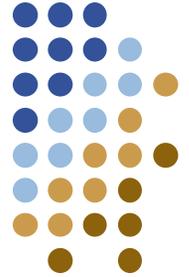




Two CFGs

- From the high-level
 - Break down the complex constructs
 - Stop at sequences of non-control-flow statements
 - Requires special handling of break, continue, goto
- From the low-level
 - Start with lowered IR – tuples, or 3-address ops
 - Build up the graph
 - More general algorithm
 - Most compilers use this approach
- Should lead to roughly the same graph

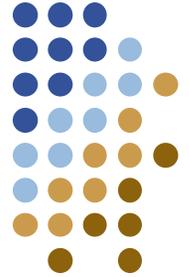




Using the CFG

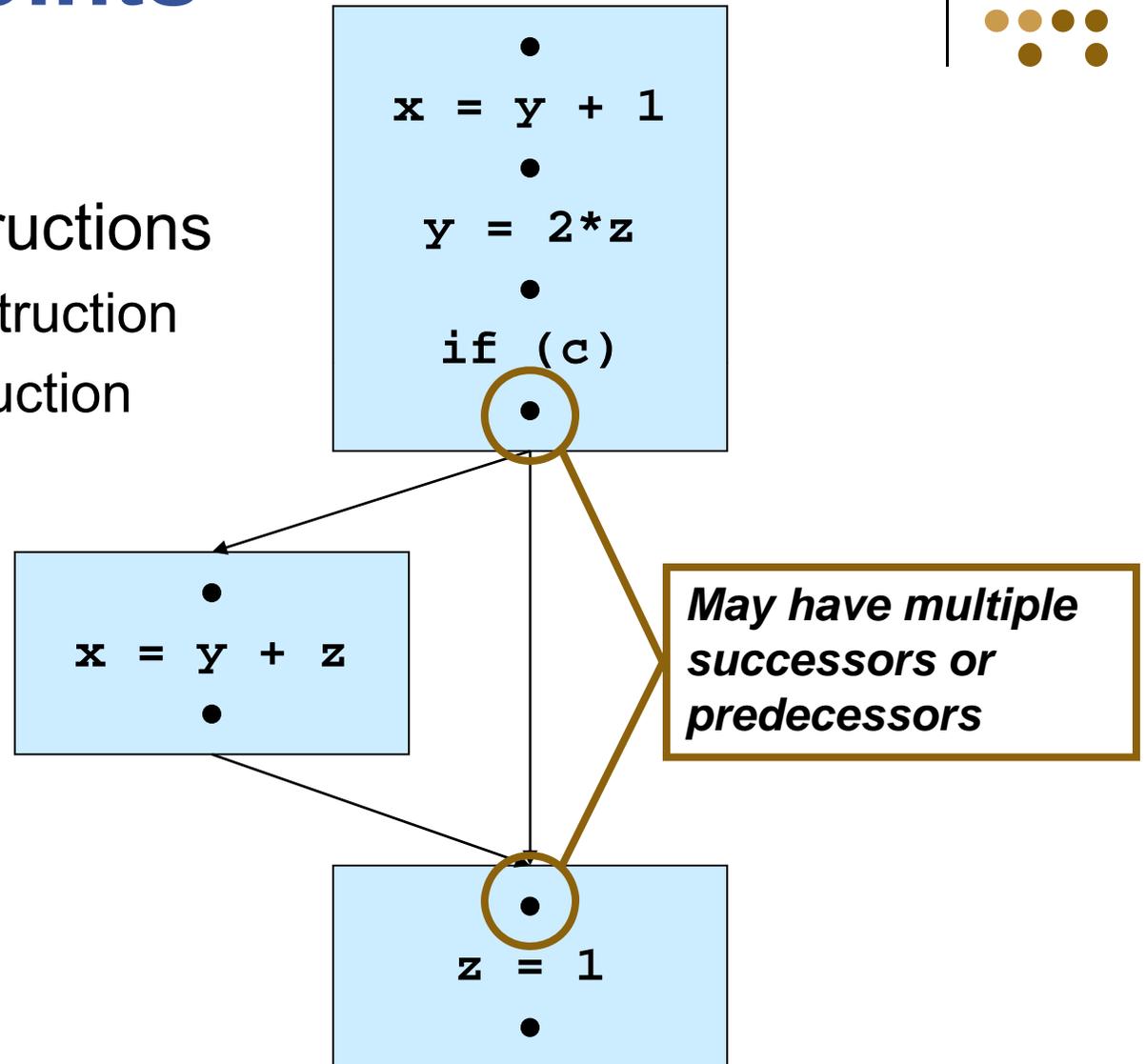
- Uniform representation for program behavior
 - Shows all possible program behavior
 - Each execution represented as a path
 - Can reason about potential behavior
 - Which paths can happen, which can't*
 - Possible paths imply possible values of variables
- Example: ***liveness*** information
- **Idea:**
 - Define program points in CFG
 - Describe how information flows between points

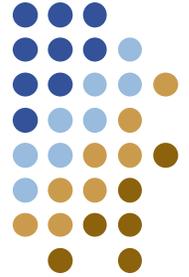




Program points

- In between instructions
 - Before each instruction
 - After each instruction





Live variables analysis

- **Idea**

- Determine *live range* of a variable

Region of the code between when the variable is assigned and when its value is used

- Specifically:

Def: A variable v is live at point p if

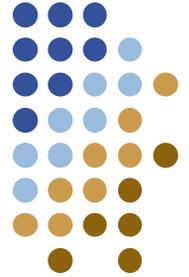
- There is a path through the CFG from p to a use of v
- There are no assignments to v along the path

➡ Compute a set of live variables at each point p

- **Uses of live variables:**

- Dead-code elimination – find unused computations
- Also: register allocation, garbage collection

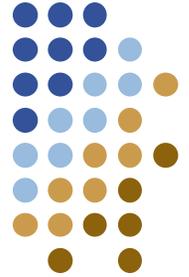




Computing live variables

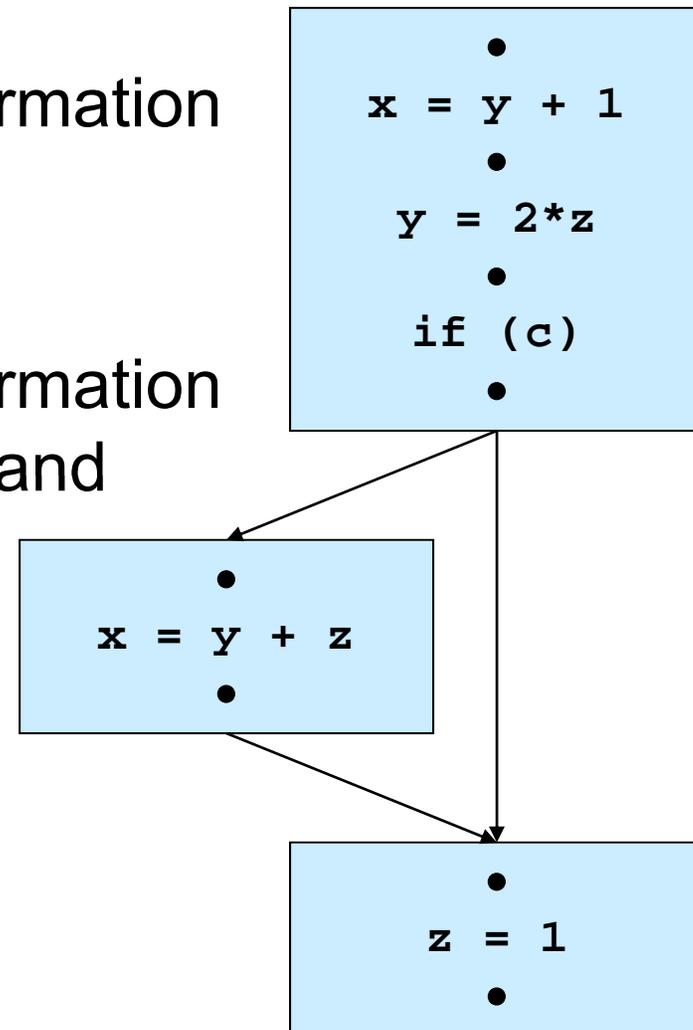
- How do we compute live variables?
(Specifically, a set of live variables at each program point)
- What is a straight-forward algorithm?
 - Start at uses of v , search backward through the CFG
 - Add v to live variable set for each point visited
 - Stop when we hit assignment to v
- Can we do better?
 - Can we compute liveness for all variables at the same time?
 - **Idea:**
 - Maintain a set of live variables
 - Push set through the CFG, updating it at each instruction

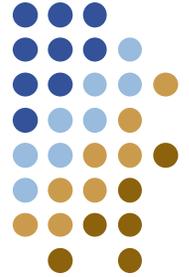




Flow of information

- **Question 1:** how does information flow across instructions?
- **Question 2:** how does information flow between predecessor and successor blocks?

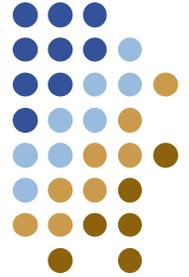




Live variables analysis

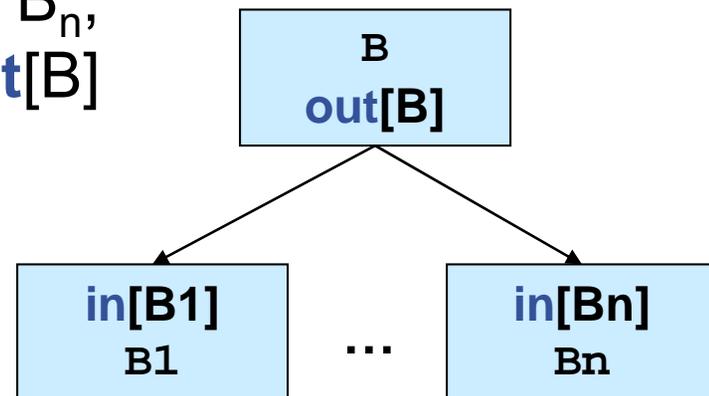
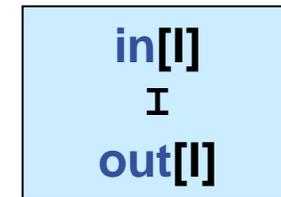
- At each program point:
Which variables contain values computed earlier and needed later
- For instruction I:
 - **in**[I] : live variables at program point before I
 - **out**[I] : live variables at program point after I
- For a basic block B:
 - **in**[B] : live variables at beginning of B
 - **out**[B] : live variables at end of B
- **Note:** **in**[I] = **in**[B] for first instruction of B
out[I] = **out**[B] for last instruction of B



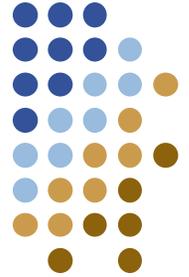


Computing liveness

- **Answer question 1:** for each instruction I , what is relation between $\mathbf{in}[I]$ and $\mathbf{out}[I]$?
- **Answer question 2:** for each basic block B , with successors B_1, \dots, B_n , what is relationship between $\mathbf{out}[B]$ and $\mathbf{in}[B_1] \dots \mathbf{in}[B_n]$



Part 1: Analyze instructions



- Live variables across instructions
- Examples:

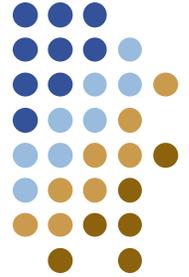
$in[l] = \{y,z\}$
 $x = y + z$
 $out[l] = \{x\}$

$in[l] = \{y,z,t\}$
 $x = y + z$
 $out[l] = \{x,t,y\}$

$in[l] = \{x,t\}$
 $x = x + 1$
 $out[l] = \{x,t\}$

- Is there a general rule?





Liveness across instructions

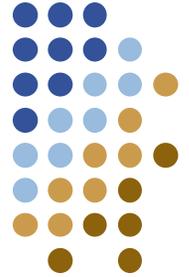
- How is liveness determined?
 - All variables that I uses are live before I
*Called the **uses** of I*
 - All variables live after I are also live before I, unless I writes to them
*Called the **defs** of I*
- Mathematically:

```
in[I] = {b}
a = b + 2
```

```
in[I] = {y,z}
x = 5
out[I] = {x,y,z}
```

$$\mathbf{in[I]} = (\mathbf{out[I]} - \mathbf{def[I]}) \cup \mathbf{use[I]}$$



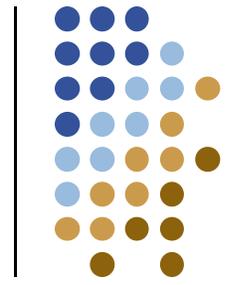


Example

- Single basic block
(obviously: $\text{out}[I] = \text{in}[\text{succ}(I)]$)
 - Live1 = in[B] = in[I1]
 - Live2 = out[I1] = in[I2]
 - Live3 = out[I2] = in[I3]
 - Live4 = out[I3] = out[B]
- Relation between live sets
 - Live1 = (Live2 - {x}) \cup {y}
 - Live2 = (Live3 - {y}) \cup {z}
 - Live3 = (Live4 - {}) \cup {d}

```
Live1  
x = y+1  
Live2  
y = 2*z  
Live3  
if (d)  
Live4
```



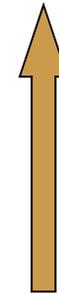


Flow of information

- Equation:

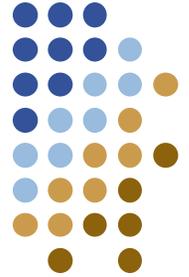
$$\text{in}[l] = (\text{out}[l] - \text{def}[l]) \cup \text{use}[l]$$

- Notice: information flows **backwards**
 - Need out[] sets to compute in[] sets
 - Propagate information up
- Many problems are **forward**
Common sub-expressions, constant propagation, others



```
Live1
x = y+1
Live2
y = 2*z
Live3
if (d)
Live4
```

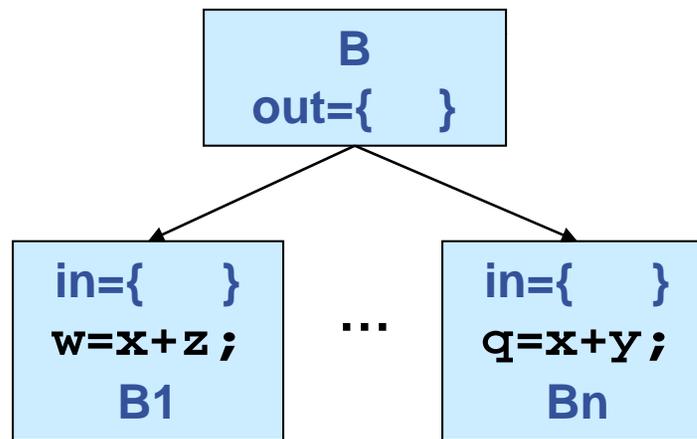




Part 2: Analyze control flow

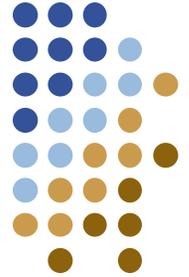
- **Question 2:** for each basic block B , with successors B_1, \dots, B_n , what is relationship between **out**[B] and **in**[B_1] ... **in**[B_n]

- Example:



- What's the general rule?





Control flow

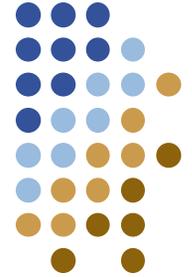
- Rule: A variable is live at end of block B if it is live at the beginning of **any** of the successors
 - Characterizes all possible executions
 - **Conservative**: some paths may not actually happen

- Mathematically:

$$\text{out}[B] = \bigcup_{B' \in \text{succ}(B)} \text{in}[B']$$

- Again: information flows backwards





System of equations

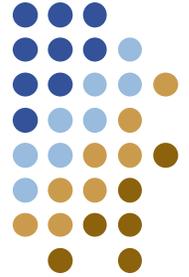
- Put parts together:

$$\begin{aligned} \mathbf{in}[I] &= (\mathbf{out}[I] - \mathbf{def}[I]) \cup \mathbf{use}[I] \\ \mathbf{out}[I] &= \mathbf{in}[\mathbf{succ}(I)] \\ \mathbf{out}[B] &= \bigcup_{B' \in \mathbf{succ}(B)} \mathbf{in}[B'] \end{aligned}$$

Often called a
system of
*Dataflow
Equations*

- Defines a system of equations (or constraints)
 - Consider equation instances for each instruction and each basic block
 - What happens with loops?
 - Circular dependences in the constraints
 - Is that a problem?





Solving the problem

- Iterative solution:
 - Start with empty sets of live variables
 - Iteratively apply constraints
 - Stop when we reach a *fixpoint*

For all instructions $\mathbf{in}[I] = \mathbf{out}[I] = \emptyset$

Repeat

For each instruction I

$$\mathbf{in}[I] = (\mathbf{out}[I] - \mathbf{def}[I]) \cup \mathbf{use}[I]$$

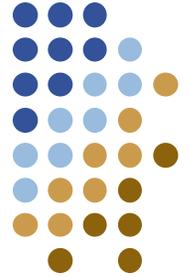
$$\mathbf{out}[I] = \mathbf{in}[\mathbf{succ}(I)]$$

For each basic block B

$$\mathbf{out}[B] = \bigcup_{B' \in \mathbf{succ}(B)} \mathbf{in}[B']$$

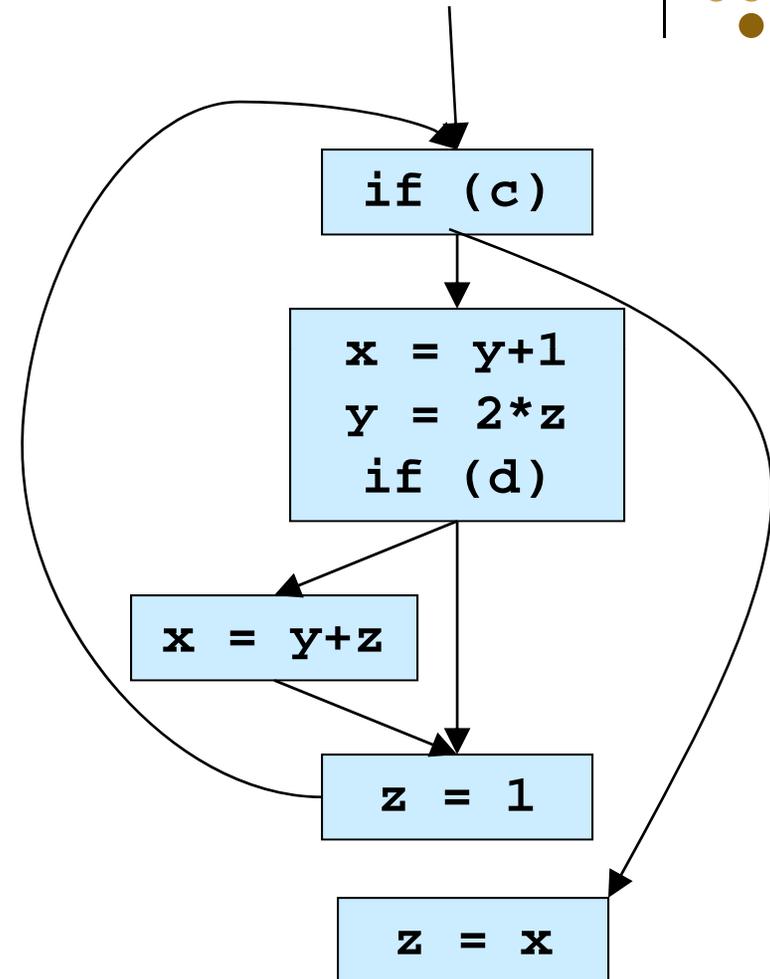
Until no new changes in sets





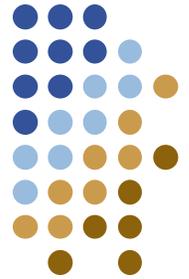
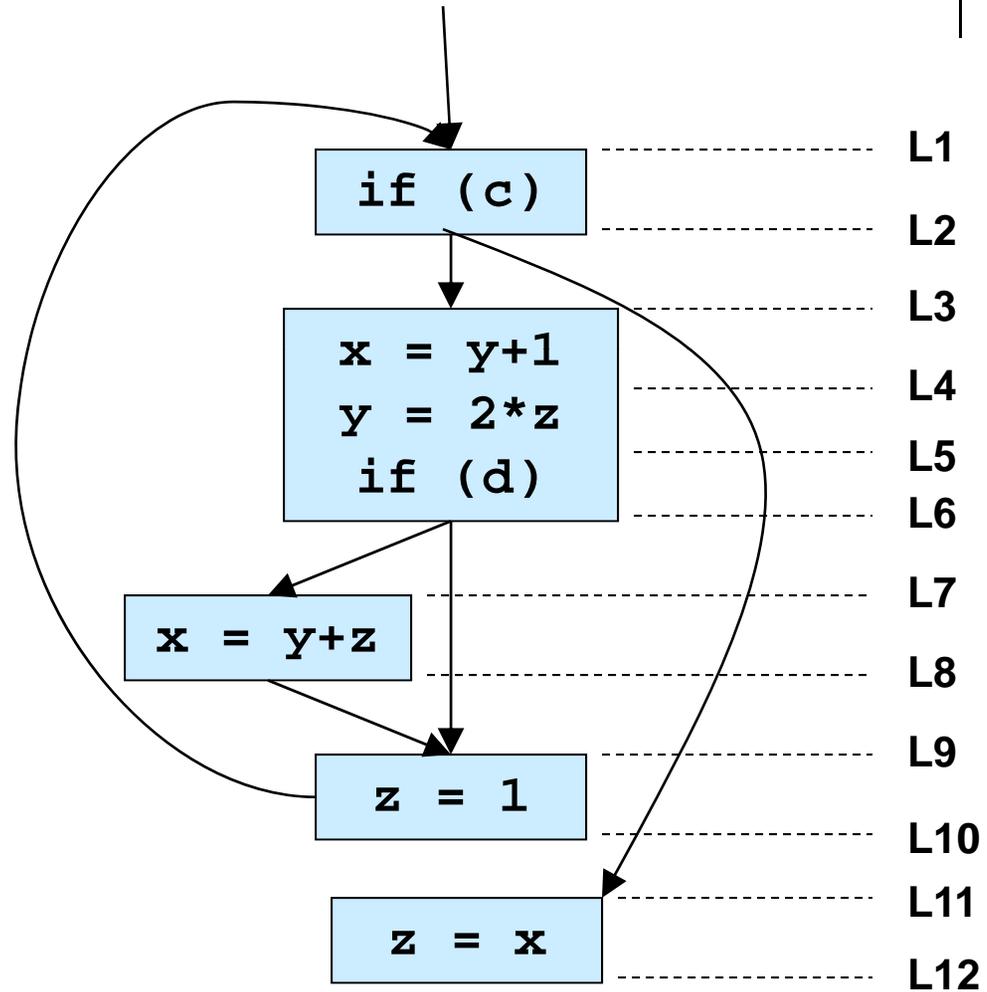
Example

- Steps:
 - Set up live sets for each program point
 - Instantiate equations
 - Solve equations



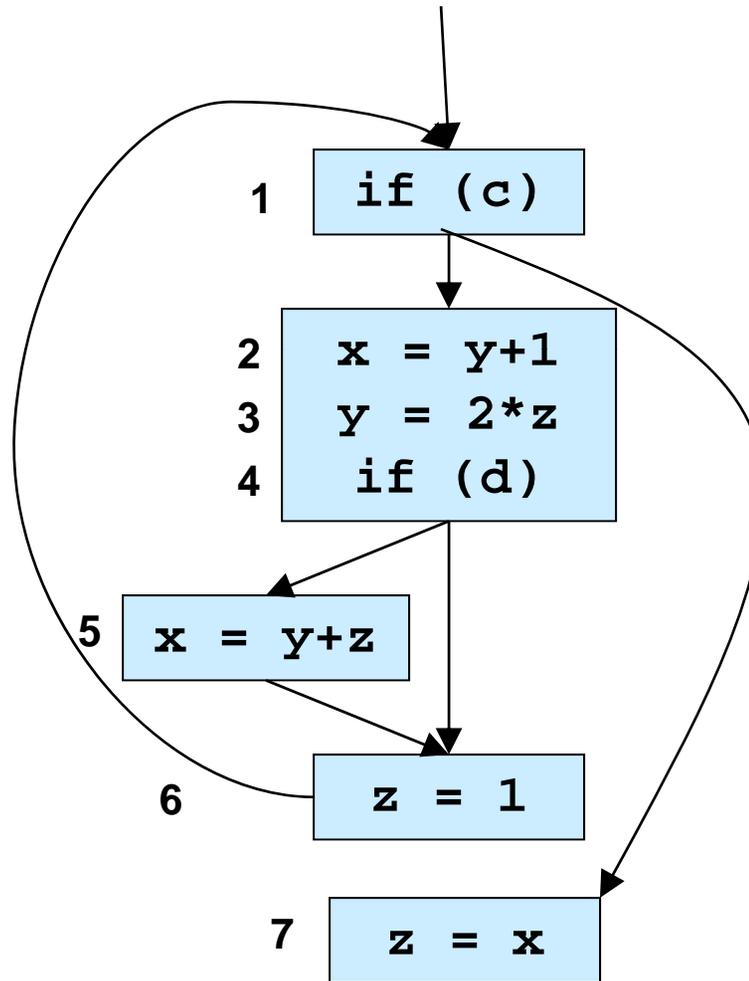
Example

- Program points

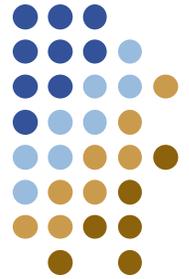


Example

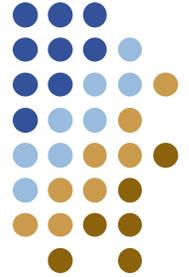
- L1 = L2 \cup {c}
- L2 = L3 \cup L11
- L3 = (L4 - {x}) \cup {y}
- L4 = (L5 - {y}) \cup {z}
- L5 = L6 \cup {d}
- L6 = L7 \cup L9
- L7 = (L8 - {x}) \cup {y,z}
- L8 = L9
- L9 = L10 - {z}
- L10 = L1
- L11 = (L12 - {z}) \cup {x}
- L12 = {}



- L1 = { x, y, z, c, d }
- L2 = { x, y, z, c, d }
- L3 = { y, z, c, d }
- L4 = { x, z, c, d }
- L5 = { x, y, z, c, d }
- L6 = { x, y, z, c, d }
- L7 = { y, z, c, d }
- L8 = { x, y, c, d }
- L9 = { x, y, c, d }
- L10 = { x, y, z, c, d }
- L11 = { x }
- L12 = { }

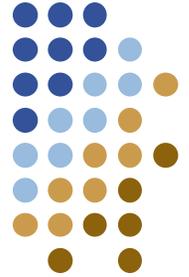


Questions



- Does this terminate?
- Does this compute the right answer?
- How could generalize this scheme for other kinds of analysis?



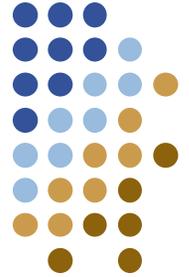


Generalization

- Dataflow analysis
 - A common framework for such analysis
 - Computes information at each program point
 - Conservative: characterizes all possible program behaviors
- Methodology
 - Describe the information (e.g., live variable sets) using a structure called a ***lattice***
 - Build a system of equations based on:
 - How each statement affects information
 - How information flows between basic blocks
 - Solve the system of constraints

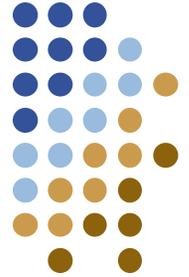


Parts of live variables analysis



- Live variable sets
 - Called *flow values*
 - Associated with program points
 - Start “empty”, eventually contain solution
- Effects of instructions
 - Called *transfer functions*
 - Take a flow value, compute a new flow value that captures the effects
 - One for each instruction – often a schema
- Handling control flow
 - Called *confluence operator*
 - Combines flow values from different paths

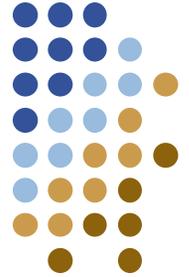




Mathematical model

- Flow values
 - Elements of a lattice $L = (P, \subseteq)$
 - Flow value $v \in P$
- Transfer functions
 - Set of functions (one for each instruction)
 - $F_i : P \rightarrow P$
- Confluence operator
 - Merges lattice values
 - $C : P \times P \rightarrow P$
- How does this help us?

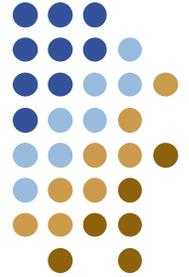




Lattices

- Lattice $L = (P, \subseteq)$
- A partial order relation \subseteq
Reflexive, anti-symmetric, transitive
- Upper and lower bounds
Consider a subset S of P
 - Upper bound of S : $u \in S : \forall x \in S \ x \subseteq u$
 - Lower bound of S : $l \in S : \forall x \in S \ l \subseteq x$
- Lattices are complete
Unique greatest and least elements
 - “Top” $T \in P : \forall x \in P \ x \subseteq T$
 - “Bottom” $\perp \in P : \forall x \in P \ \perp \subseteq x$

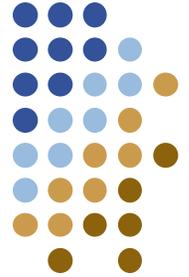




Confluence operator

- Combine flow values
 - “Merge” values on different control-flow paths
 - Result should be a safe over-approximation
 - We use the lattice \subseteq to denote “more safe”
- Example: live variables
 - $v1 = \{x, y, z\}$ and $v2 = \{y, w\}$
 - How do we combine these values?
 - $v = v1 \cup v2 = \{w, x, y, z\}$
 - What is the “ \subseteq ” operator?
 - Superset





Meet and join

- **Goal:**

Combine two values to produce the “best” approximation

- Intuition:

- Given $v1 = \{x, y, z\}$ and $v2 = \{y, w\}$
- A safe over-approximation is “all variables live”
- We want the smallest set

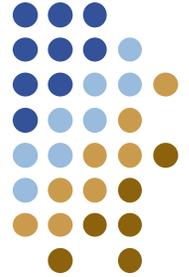
- Greatest lower bound

- Given $x, y \in P$
- $GLB(x, y) = z$ such that
 - $z \subseteq x$ and $z \subseteq y$ and
 - $\forall w w \subseteq x$ and $w \subseteq y \Rightarrow w \subseteq z$

- **Meet** operator: $x \wedge y = GLB(x, y)$

- Natural “opposite”: Least upper bound, **join** operator





Termination

- Monotonicity

Transfer functions F are *monotonic* if

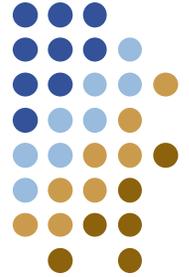
- Given $x, y \in P$
- If $x \subseteq y$ then $F(x) \subseteq F(y)$
- Alternatively: $F(x) \subseteq x$

- Key idea:

Iterative dataflow analysis terminates if

- Transfer functions are monotonic
- Lattice has finite height
- *Intuition*: values only go down, can only go to bottom





Example

- Prove monotonicity of live variables analysis

- Equation: $\text{in}[i] = (\text{out}[i] - \text{def}[i]) \cup \text{use}[i]$

(For each instruction i)

- As a function: $F(x) = (x - \text{def}[i]) \cup \text{use}[i]$

- Obligation: If $x \subseteq y$ then $F(x) \subseteq F(y)$

- Prove:

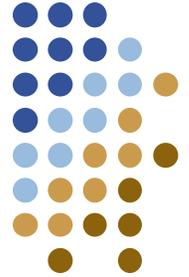
$$x \subseteq y \quad \Rightarrow \quad (x - \text{def}[i]) \cup \text{use}[i] \subseteq (y - \text{def}[i]) \cup \text{use}[i]$$

- Somewhat trivially:

- $x \subseteq y \Rightarrow x - s \subseteq y - s$

- $x \subseteq y \Rightarrow x \cup s \subseteq y \cup s$

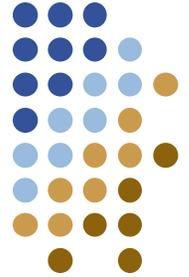




Dataflow solution

- Question:
 - What is the solution we compute?
 - Start at lattice top, move down
 - Called greatest *fixpoint*
 - Where does approximation come from?
 - Confluence of control-flow paths
- Knaster Tarski theorem
 - Every monotonic function F over a complete lattice L has a unique least (and greatest) fixpoint
 - (Actually, the theorem is more general)





Summary

- Dataflow analysis
 - Lattice of flow values
 - Transfer functions (encode program behavior)
 - Iterative fixpoint computation
- **Key insight:**
 - If our dataflow equations have these properties:*
 - Transfer functions are monotonic
 - Lattice has finite height
 - Transfer functions distribute over meet operator
 - Then:*
 - Our fixpoint computation will terminate
 - Will compute meet-over-all-paths solution

