Lecture 12

Region-Based Analysis

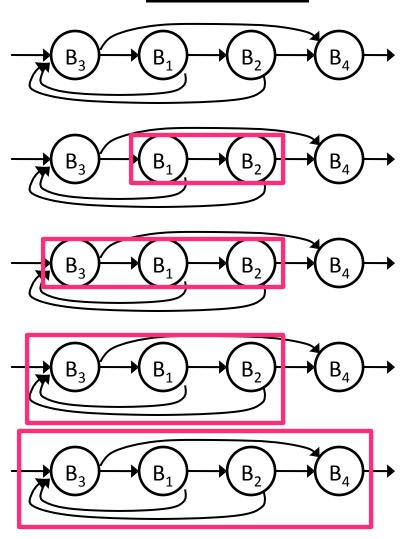
- I. Basic Idea
- II. Algorithm
- III. Optimization and Complexity
- IV. Comparing region-based analysis with iterative algorithms

Reading: ALSU 9.7

Motivation for Studying Region-Based Analysis

- Exploit the structure of block-structured programs in data flow
- Tie in several concepts studied:
 - Use of structure in induction variables, loop invariant
 - motivated by nature of the problem
 - <u>This lecture:</u> can we use structure for speed?
 - Iterative algorithm for data flow
 - This lecture: an alternative algorithm
 - Reducibility
 - all retreating edges of DFST are back edges
 - reducible graphs converge quickly
 - <u>This lecture:</u> algorithm exploits & requires reducibility
- Usefulness in practice
 - Faster for "harder" analyses
 - Useful for analyses related to structure
- Theoretically interesting: better understanding of data flow

I. Big Picture



Basic Idea

- In Iterative Analysis:
 - DEFINITION: Transfer function F_B:
 summarize effect from beginning to end of basic block B
- In Region-Based Analysis:
 - DEFINITION: Transfer function F_{R,B}: summarize effect from beginning of R to end of basic block B
 - Recursively construct a larger region R from smaller regions construct $F_{R,B}$ from transfer functions for smaller regions until the program is one region
 - Let P be the region for the entire program, and v be initial value at entry node
 - $out[B] = F_{P,B}(v)$
 - in [B] = $\Lambda_{B'}$ out[B'], where B' is a predecessor of B

II. Algorithm

- 1. Operations on transfer functions
- 2. How to build nested regions?
- 3. How to construct transfer functions that correspond to the larger regions?

1. Operations on Transfer Functions

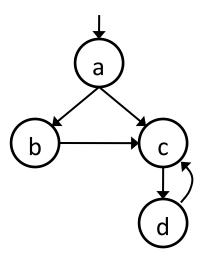
- Example: Reaching Definitions
- $F(x) = Gen \cup (x Kill)$
- $F_2(F_1(x)) = Gen_2 \cup (F_1(x) Kill_2)$ = $Gen_2 \cup (Gen_1 \cup (x - Kill_1)) - Kill_2)$ = $Gen_2 \cup (Gen_1 - Kill_2) \cup (x - (Kill_1 \cup Kill_2))$
- $F_1(x) \wedge F_2(x) = Gen_1 \cup (x Kill_1) \cup Gen_2 \cup (x Kill_2)$ = $(Gen_1 \cup Gen_2) \cup (x - (Kill_1 \cap Kill_2))$
- F*(x) ≤ Fⁿ(x), ∀ n ≥ 0
 = x ∪ F(x) ∪ F(F(x)) ∪ ...
 = x ∪ (Gen ∪ (x Kill)) ∪ (Gen ∪ ((Gen ∪ (x Kill)) Kill)) ∪ ...
 = Gen ∪ (x ∅)

2. Structure of Nested Regions (An Example)

- A region in a flow graph is a set of nodes that
 - includes a header, which dominates all other nodes in a region
- T1-T2 rule (Hecht & Ullman)
 - T1: Remove a loop
 If n is a node with a loop, i.e. an edge n->n, delete that edge

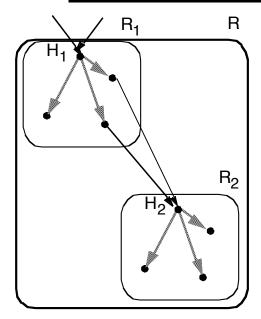
T2: Remove a vertex
 If there is a node n that has a unique predecessor, m, then m may consume n by deleting n and making all successors of n be successors of m.

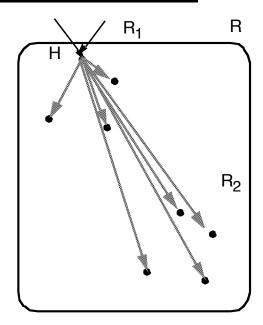
Example



- In reduced graph:
 - each vertex represents a subgraph of original graph (a region).
 - each edge represents an edge in original graph
- Limit flow graph: result of exhaustive application of T1 and T2
 - independent of order of application.
 - if limit flow graph has a single vertex → reducible
- Can define larger regions (e.g. Allen&Cocke's intervals)
 - simple regions → simple composition rules for transfer functions

3. Transfer Functions for T2 Rule



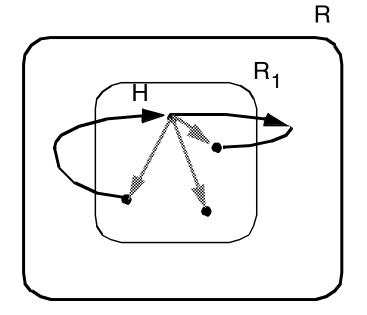


Transfer function

F_{R,B}: summarizes the effect from beginning of R to end of B **F**_{R,in(H2)}: summarizes the effect from beginning of R to beginning of H2

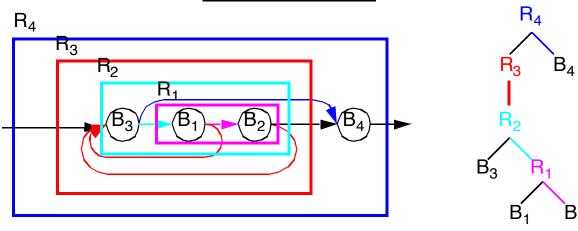
- Unchanged for blocks B in region R_1 ($F_{R,B} = F_{R1,B}$)
- $F_{R,in(H2)} = \Lambda_P F_{R,P}$, where p is a predecessor of H_2
- For blocks B in region R_2 : $F_{R,B} = F_{R2,B} \cdot F_{R,in(H2)}$

Transfer Functions for T1 Rule



- Transfer Function F_{R,B}
 - $F_{R,in(H)} = (\Lambda_P F_{R1,P})^*$, where p is a predecessor of H in R
 - $F_{R,B} = F_{R1,B} \cdot F_{R,in(H)}$

First Example

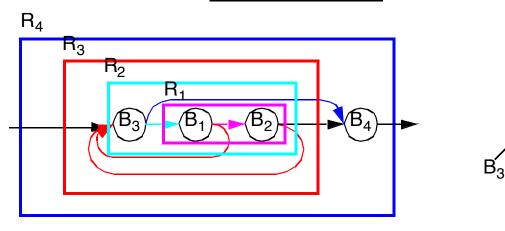


R	T _{1/} T ₂	R'	F _{R,in(R')}	F _{R,B1}	F _{R,B2}	F _{R,B3}	F _{R,B4}
R ₁	T ₂	B ₂					
R ₂	T ₂	R_1					
R ₃	T ₁	R ₂					
R ₄	T ₂	B ₄					

• R: region name

R': region whose header will be subsumed

First Example

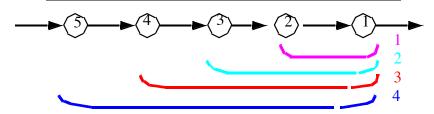


R	T _{1/} T ₂	R'	F _{R,in(R')}	F _{R,B1}	F _{R,B2}	F _{R,B3}	F _{R,B4}
R ₁	T ₂	B ₂	F _{B1}	F _{B1}	F _{B2} •F _{R1,in(B2)}		
R ₂	T ₂	R ₁	F _{B3}	F _{R1,B1} •F _{R2,in(R1)}	F _{R1,B2} •F _{R2,in(R1)}	F _{B3}	
R ₃	T ₁	R ₂	(F _{R2B1} Λ F _{R2B2})*	F _{R2,B1} •F _{R3,in(R2)}	F _{R2,B2} •F _{R3,in(R2)}	F _{R2,B3} •F _{R3,in(R2)}	
R ₄	T ₂	B ₄	F _{R3B3} Λ F _{R3B2}	F _{R3,B1}	F _{R3,B2}	F _{R3,B3}	F _{B4} •F _{R4,in(B4)}

- R: region name
- R': region whose header will be subsumed

 R_4

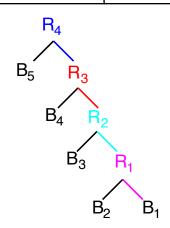
III. Complexity of Algorithm



R	T _{1/} T ₂	R'	F _{R,in(R')}	F _{R,B1}	F _{R,B2}	F _{R,B3}	F _{R,B4}	F _{R,B5}
R ₁	T ₂	B ₂	F _{B2}	F _{B1} •F _{B2}	F _{B2}			
R ₂	T ₂	R ₁	F _{B3}	F _{R1,B1} •F _{B3}	F _{R1,B2} •F _{B3}	F _{B3}		
R ₃	T ₂	R ₂	F _{B4}	F _{R2,B1} •F _{B4}	F _{R2,B2} •F _{B4}	F _{R2,B3} •F _{B4}	F _{B4}	
R ₄	T ₂	R ₃	F _{B5}	F _{R3,B1} •F _{B5}	F _{R3,B2} •F _{B5}	F _{R3,B3} •F _{B5}	F _{B4} •F _{B5}	F _{B5}

R	F _{R4,in(R)}
R_4	1
R_3	F _{B5} •F _{R4,in(R4)}
R ₂	F _{B4} •F _{R4,in(R3)}
R_1	F _{B3} •F _{R4,in(R2)}
B_1	F _{B2} •F _{R4,in(R1)}

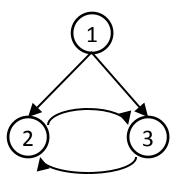
В	F _{R4,B}
B ₅	F _{B5} •I
B ₄	F _{B4} •F _{R4,in(R3)}
B ₃	F _{B3} •F _{R4,in(R2)}
B ₂	F _{B2} •F _{R4,in(R1)}
B_1	F _{B1} •F _{R4,in(B1)}



Optimization

- Let m = number of edges, n = number of nodes
- Ideas for optimization
 - If we compute $F_{R,B}$ for every region B is in, then it is very expensive
 - We are ultimately only interested in the entire region (E); we need to compute only $F_{E,B}$ for every B.
 - There are many common subexpressions between F_{E,B1}, F_{E,B2}, ...
 - Number of F_{E,B} calculated = m
 - Also, we need to compute $F_{R,in(R')}$, where R' represents the region whose header is subsumed.
 - Number of F_{R,B} calculated, where R is not final = n
- Total number of F_{R,B} calculated: (m + n)
 - Data structure keeps "header" relationship
 - Practical algorithm: O(m log n)
 - Complexity: O(m α (m,n)), α is inverse Ackermann function

Reducibility



- If no T1, T2 is applicable before graph is reduced to single node, then **split node** and continue
- Worst case: exponential
- Most graphs (including GOTO programs) are reducible

IV. Comparison with Iterative Data Flow

Applicability

- Definitions of F* can make technique more powerful than iterative algorithms
- Backward flow: reverse graph is not typically reducible.
 - Requires more effort to adapt to backward flow than iterative algorithm
- More important for interprocedural optimization

Speed

- Irreducible graphs
 - Iterative algorithm can process irreducible parts uniformly
 - Serious "irreducibility" can be slow with region-based analysis
- Reducible graph & Cycles do not add information (common)
 - Iterative: (depth + 2) passes depth is 2.75 average, independent of code length
 - Region-based analysis: Theoretically almost linear, typically O(m log n)
- Reducible & Cycles add information
 - Iterative takes longer to converge
 - Region-based analysis remains the same