

Lecture 10

Region-Based Analysis

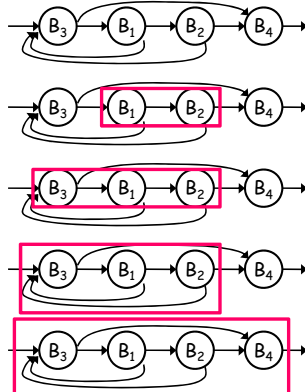
- I. Basic Idea
- II. Algorithm
- III. Optimization and Complexity
- IV. Comparing region-based analysis with iterative algorithms

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
 - *This lecture: 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
 - *This lecture: 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
 - $out[B] = F_{P,B}(v)$
 - $in[B] = \bigwedge_{B'} out[B']$, 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?

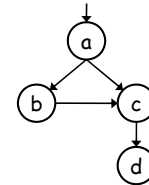
1. Operations on Transfer Functions

- **Example: Reaching Definitions**
- $F(x) = Gen \cup (x - Kill)$
- $F_2(F_1(x)) = Gen_2 \cup (F_1(x) - Kill_2)$
 $= Gen_2 \cup (Gen_1 \cup (x - Kill_1) - Kill_2)$
 $= Gen_2 \cup (Gen_1 - Kill_2) \cup (x - (Kill_1 \cup Kill_2))$
- $F_1(x) \wedge F_2(x) = Gen_1 \cup (x - Kill_1) \cup Gen_2 \cup (x - Kill_2)$
 $= (Gen_1 \cup Gen_2) \cup (x - (Kill_1 \cap Kill_2))$
- $F^*(x) \leq F^n(x), \forall n \geq 0$
 $= x \cup F(x) \cup F(F(x)) \cup \dots$
 $= x \cup (Gen \cup (x - Kill)) \cup (Gen \cup ((Gen \cup (x - Kill)) - Kill)) \cup \dots$
 $= Gen \cup (x - \emptyset)$

2. Structure of Nested Regions (An Example)

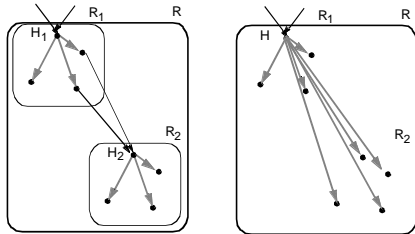
- A **region** in a flow graph is a set of nodes that
 - includes a **header**, which **dominates all other nodes in a region**
- **T1-T2 rule (Hecht & Ullman)**
 - **T1: Remove a loop**
 If n is a node with a **loop**, i.e. an **edge $n \rightarrow n$** , **delete that edge**
 - **T2: Remove a vertex**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



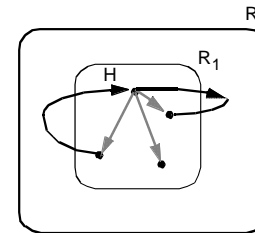
- In reduced graph:
 - each **vertex** represents a **subgraph of original graph (a region)**.
 - each **edge** represents an **edge in original graph**
- **Limit flow graph**: result of **exhaustive application of T1 and T2**
 - independent of order of application.
 - if limit flow graph has a **single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

3. Transfer Functions for T2 Rule



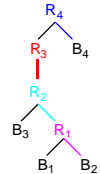
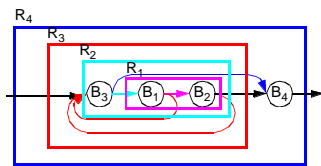
- **Transfer function**
- $F_{R,B}$: summarizes the effect from beginning of R to end of B
- $F_{R,in(H2)}$: summarizes the effect from beginning of R to beginning of H2
 - Unchanged for blocks B in region R_1 ($F_{R,B} = F_{R1,B}$)
 - $F_{R,in(H2)} = \wedge_p F_{R,p}$, where p is a predecessor of H2
 - 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 $F_{R,B}$**
 - $F_{R,in(H)} = (\wedge_p F_{R1,p})^*$, where p is a predecessor of H in R
 - $F_{R,B} = F_{R1,B} \cdot F_{R,in(H)}$

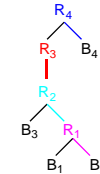
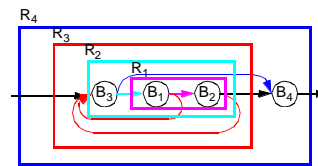
First Example



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

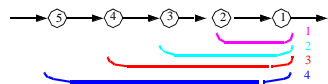
First Example



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}	$F_{B2} \cdot F_{R1,in(B2)}$		
R_2	T_2	R_1	F_{B3}	$F_{R1,B1} \cdot F_{R2,in(R1)}$	$F_{R1,B2} \cdot F_{R2,in(R1)}$	F_{B3}	
R_3	T_1	R_2	$(F_{R2,B1} \wedge F_{R2,B2})^*$	$F_{R2,B1} \cdot F_{R3,in(R2)}$	$F_{R2,B2} \cdot F_{R3,in(R2)}$	$F_{R2,B3} \cdot F_{R3,in(R2)}$	
R_4	T_2	B_4	$F_{R3,B3} \wedge F_{R3,B2}$	$F_{R3,B1}$	$F_{R3,B2}$	$F_{R3,B3}$	$F_{B4} \cdot F_{R4,in(B4)}$

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

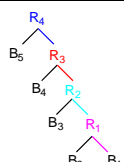
III. Complexity of Algorithm



R	T ₁ /T	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	I
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)}

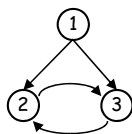
B	F _{R4,B}
B ₅	F _{B5} · I
B ₄	F _{B4} · F _{R4,in(R3)}
B ₃	F _{B3} · F _{R4,in(R2)}
B ₂	F _{B2} · F _{R4,in(R1)}
B ₁	F _{B1} · F _{R4,in(B1)}



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: $O(m\alpha(m,n))$, 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