

Lecture 12

Region-Based Analysis

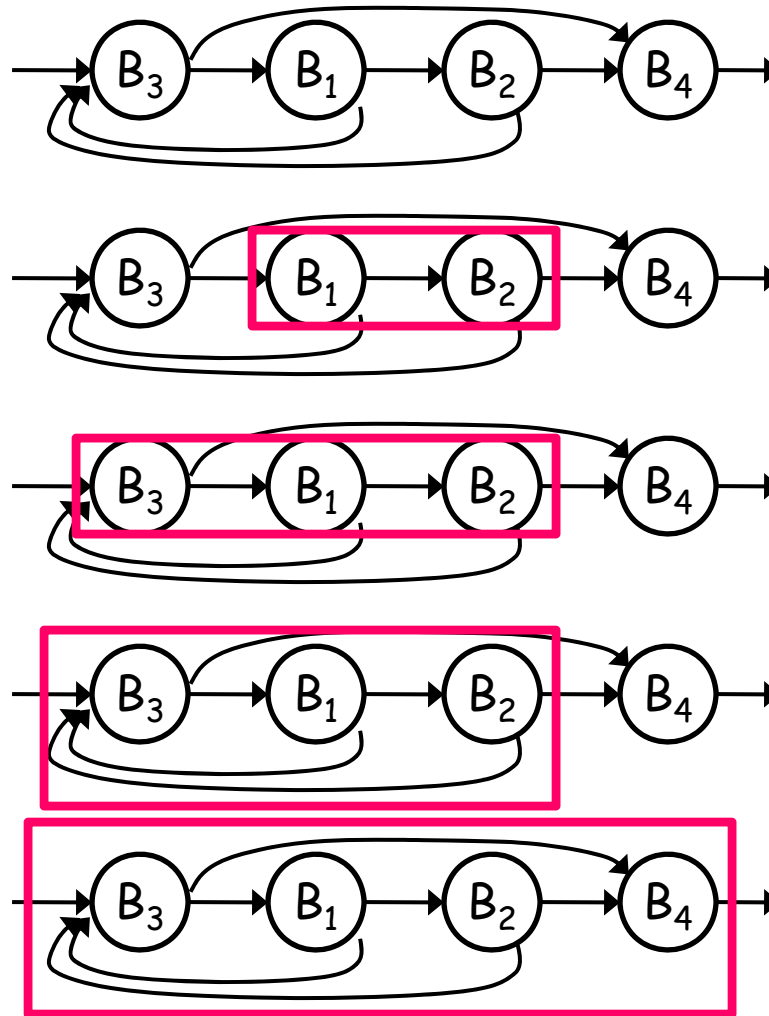
- I. Basic Idea
- II. Algorithm
- III. Optimization and Complexity
- IV. Comparing region-based analysis with iterative algorithms

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
 - *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 regionsuntil 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

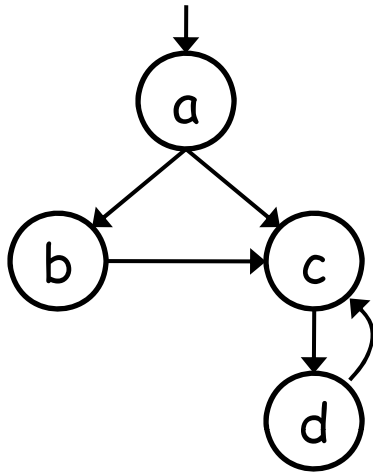
- **Example: Reaching Definitions**
- $F(x) = Gen \cup (x - Kill)$
- $F_2(F_1(x)) = Gen_2 \cup (F_1(x) - Kill_2)$
 $= Gen_2 \cup (Gen_1 \cup (x - Kill_1)) - Kill_2$
 $= Gen_2 \cup (Gen_1 - Kill_2) \cup (x - (Kill_1 \cup Kill_2))$
- $F_1(x) \wedge F_2(x) = Gen_1 \cup (x - Kill_1) \cup Gen_2 \cup (x -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

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**

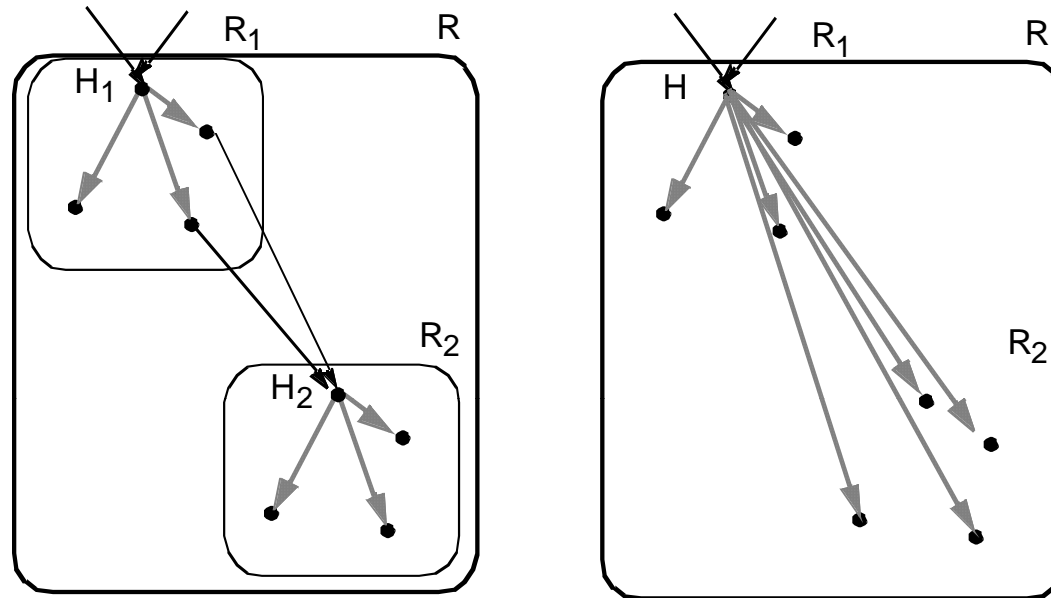
 - **T2: Remove a vertex**
If there is a node n that has a **unique predecessor**, m , then m may consume n by **deleting** n and making **all successors of** n be successors of m .

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** → **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



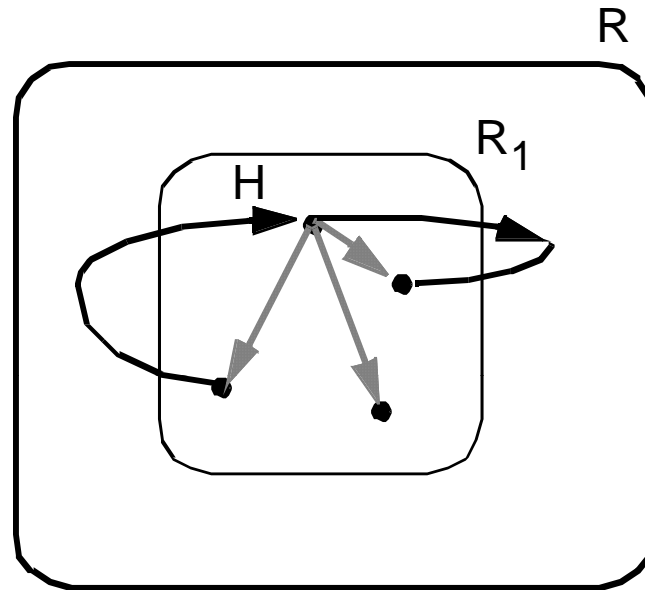
- **Transfer function**

$F_{R,B}$: 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**

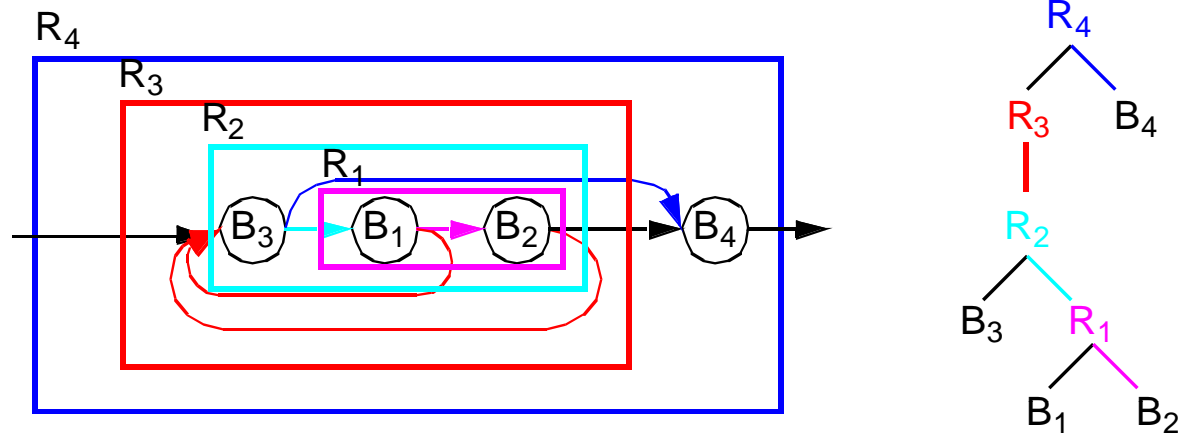
- Unchanged for blocks B in region R_1 ($F_{R,B} = F_{R1,B}$)
- $F_{R,in(H2)} = \bigwedge_p F_{R,p}$, where p is a predecessor of H_2
- For blocks B in region R_2 : $F_{R,B} = F_{R2,B} \cdot F_{R,in(H2)}$

Transfer Functions for T1 Rule



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

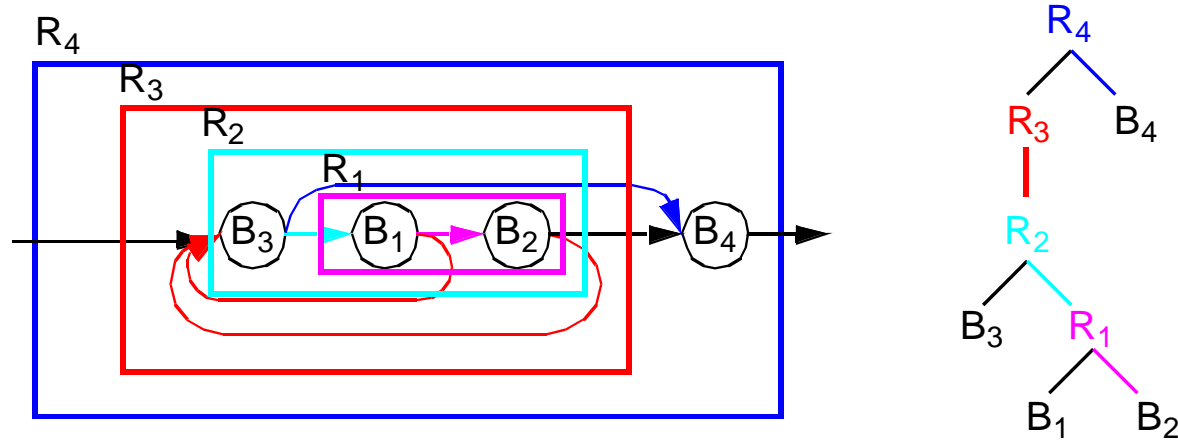
First Example



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

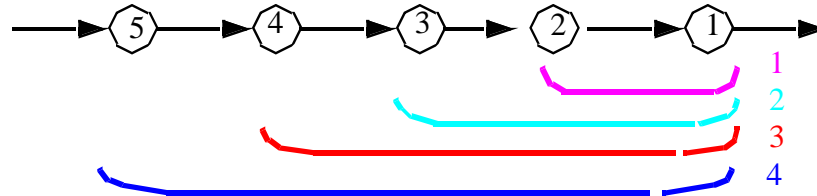
First Example



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} ∧ 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} ∧ 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

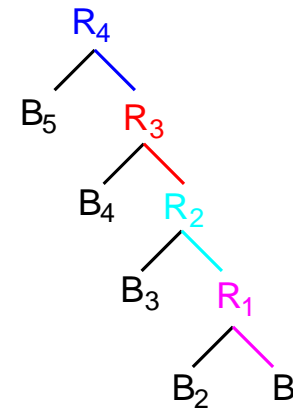
III. Complexity of Algorithm



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	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)}

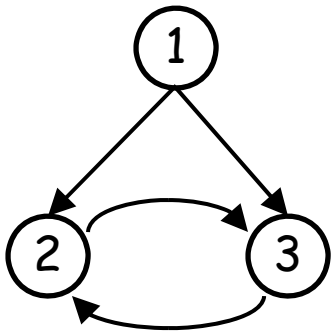
B	$F_{R4,B}$
B ₅	F _{B5} · I
B ₄	F _{B4} · F _{R4,in(R3)}
B ₃	F _{B3} · F _{R4,in(R2)}
B ₂	F _{B2} · F _{R4,in(R1)}
B ₁	F _{B1} · F _{R4,in(B1)}



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