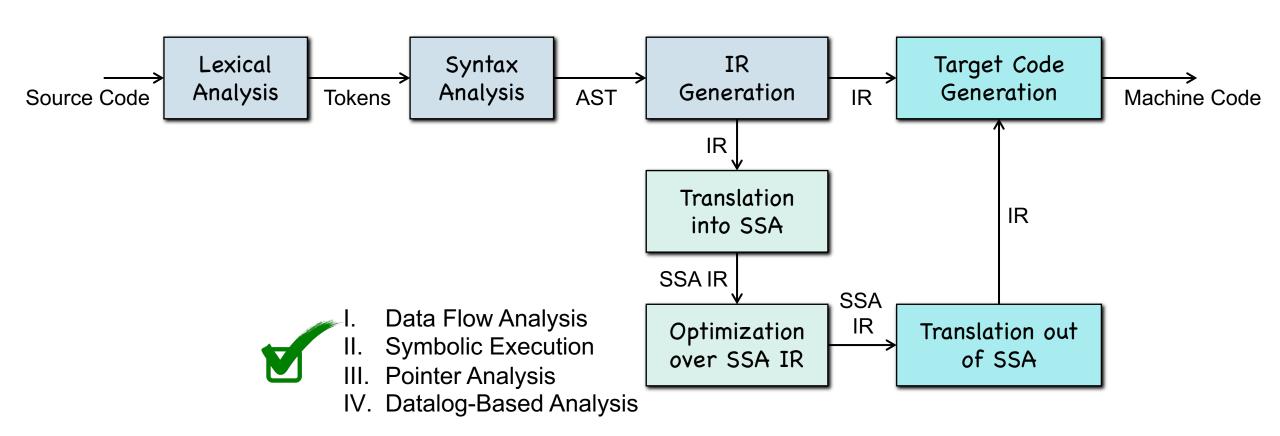
Recap-2 The Compilers' Middle End



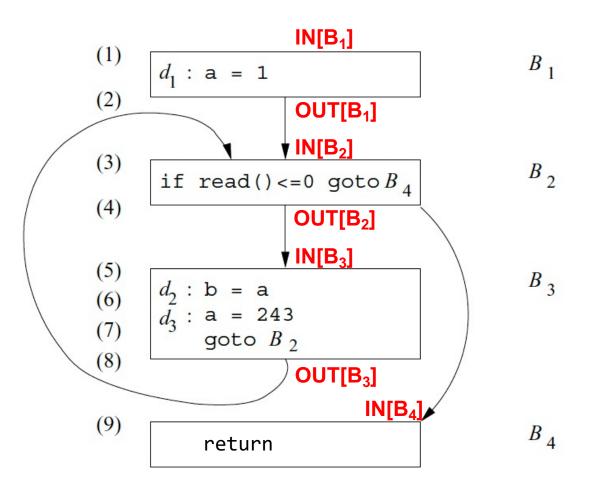
Middle End of Compilers





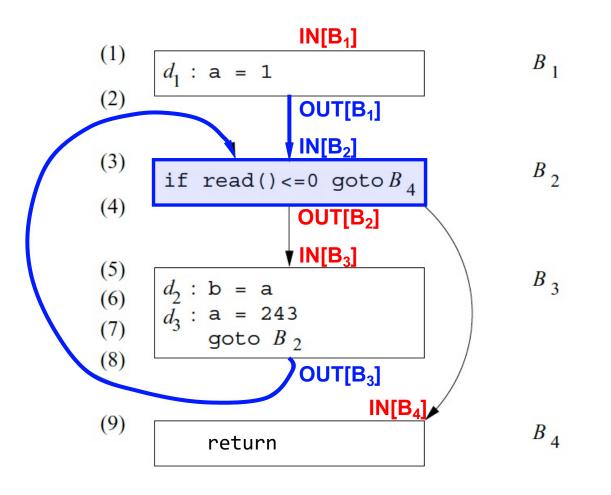
PART I: Data Flow Analysis





- IN[B]/OUT[B]
 - Set of data flow facts before and after a block
- Constraints
 - OUT[B] = $f_B(IN[B])$
 - $IN[B] = \bigwedge_{P \in preds(B)} OUT[P]$





- IN[B]/OUT[B]
 - Set of data flow facts before and after a block
- Constraints
 - OUT[B] = $f_B(IN[B])$
 - $IN[B] = \bigwedge_{P \in preds(B)} OUT[P]$



- To perform data flow analysis, define
 - transfer function f for each statement/block
 - merge function ∧ at the joint point of multiple paths
 - the initial set of data flow facts for each block



- To perform data flow analysis, define
 - transfer function f for each statement/block
 - merge function ∧ at the joint point of multiple paths
 - the initial set of data flow facts for each block

- Data flow analysis produces
 - IN[B]/OUT[B] that include data flow facts at a program point
 - The resulting data flow facts are always true during program execution



The Worklist Algorithm

- ForEach basic block B: initialize IN[B] and OUT[B]; EndFor
- worklist = set of all basic blocks;

- While (!worklist.empty()) Do
- B = worklist.pop();
- IN[B] = $\Lambda_{P \in preds(B)}$ OUT[P]; OUT[B] = $f_B(IN[B])$
- If OUT[B] changed: worklist.push(all B's successors); EndIf
- EndWhile



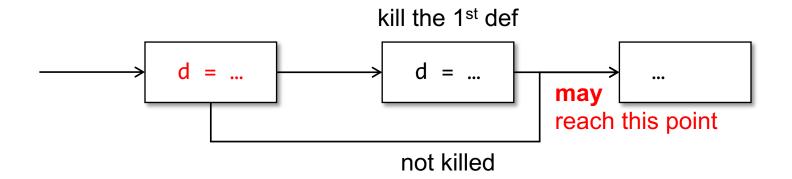
Classic DFA

- Reaching Definition
- Available Expressions
- Live Variables



Reaching Definitions

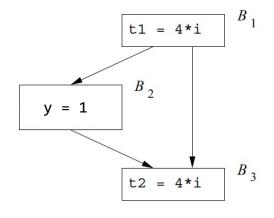
 A definition d may reach a program point, if there is a path from d to the program point such that d is not killed along the path.

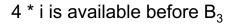


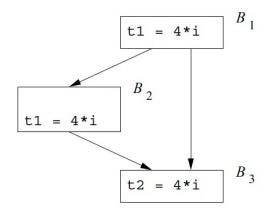


Available Expressions

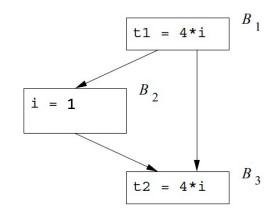
- An expression x + y is available at a point p if
 - every path from the entry node to p evaluates x + y, and
 - after the last such evaluation prior to reaching p, there are no subsequent assignments to x or y.







4 * i is available before B₃



4 * i is **NOT** available before B₃



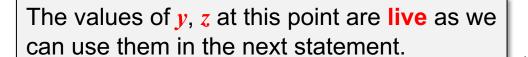
Live Variable Analysis

 We wish to know if the value of a variable x at point p can be used along the path starting from p.

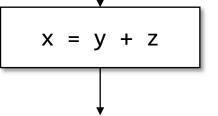


Live Variable Analysis

 We wish to know if the value of a variable x at point p can be used along the path starting from p.



The value of x at this point is **dead** because x will be redefined and the old value cannot be used any longer.





Transfer & Merging Functions

Transfer Functions?

Merge Functions?

Forward or Backward Analysis?



PART II: Symbolic Execution



Symbolic Execution

- Path Sensitivity
- Flow Sensitivity
- Context Sensitivity



Symbolic Execution

- Path Explosion
- Constraint Solving
- External Function Call



Solving Path Constraints

How can we check the satisfiability of a constraint?

$$\alpha_a \neq 0 \land \alpha_b = 0 \land \neg (2(\alpha_a + \alpha_b) - 4 \neq 0)$$

- Constraint solver
 - Satisfiable => A possible solution
 - Unsatisfiable
 - Unknown (due to timeout)







Solving Path Constraints

- Propositional Logic
- Satisfiability (SAT) --- DPLL and CDCL

- First Order Logic
- Satisfiability Modulo Theories (SMT)
- The Bit Vector Theory



The SAT Problem

Deciding satisfiability of a formula F in PL



The SAT Problem

Deciding satisfiability of a formula F in PL

- **Step 1**: Transforming *F* to CNF by Tseitin Transformation
- Step 2: Invoke the DPLL algorithm



The SAT Problem

Deciding satisfiability of a formula F in PL

- **Step 1**: Transforming *F* to CNF by Tseitin Transformation
- Step 2: Invoke the DPLL algorithm What is CNF?
 Why not DNF or NNF?



Davis

Putnam

Logemann

Loveland (DPLL) algorithm

Goal: Find a solution to satisfy the input formula



• $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$

Goal: Find a solution to satisfy the input formula

- Repeat the following steps:
- (1) Choose a variable, p, assign true or false to the variable
- (2) Propagate the value of *p*



• $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$

```
bool DPLL (F, I) {
    if (F == false) return UNSAT;
    if (F == true) return SAT;

    p = choose(F);
    bool ret = DPLL(F[p→true], I[p→true])
    if (ret == SAT) return SAT;

    return DPLL(F[p→false], I[p→ false]);
}
```



• $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$

```
Optimizations may be applied!
bool DPLL (F, I) {
     if (F == false) return UNSAT;
     if (F == true) return SAT;
     p = choose(F);
     bool ret = DPLL(F[p→true]. I[p→true])
     if (ret == SAT) rety Theorem [Resolution]
                                            (a_0 \lor a_1 \lor \cdots \lor a_n \lor c) \land (b_0 \lor b_1 \lor \cdots \lor b_m \lor \neg c)
     return DPLL(F[p→fal
                                                            can be simplified into
                                                   (a_0 \lor a_1 \lor \cdots \lor a_n \lor b_0 \lor b_1 \lor \cdots \lor b_m)
```



• $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$



- $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$
- $(p \lor q) \land (p \lor \neg q) \land \neg p$



- $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$
- $(p \lor q) \land (p \lor \neg q) \land \neg p$



- $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$
- $(p \lor q) \land (p \lor \neg q) \land \neg p$
- $p \land \neg p$



- $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$
- $(p \lor q) \land (p \lor \neg q) \land \neg p$
- $p \land \neg p$ False! Unsatisfiable!



- $(p \lor q \lor \neg r) \land (p \lor q \lor r) \land (p \lor \neg q) \land \neg p$
- $(p \lor q) \land (p \lor \neg q) \land \neg p$
- $p \land \neg p$ False! Unsatisfiable!

Theorem [Resolution]

$$(a_0 \lor a_1 \lor \cdots \lor a_n \lor \mathbf{c}) \land (b_0 \lor b_1 \lor \cdots \lor b_m \lor \neg \mathbf{c})$$

$$can \ be \ simplified \ into$$

$$(a_0 \lor a_1 \lor \cdots \lor a_n \lor b_0 \lor b_1 \lor \cdots \lor b_m)$$



•
$$(r \Rightarrow p) \Rightarrow (\neg(q \land r) \Rightarrow p)$$

On the Complexity of Derivation in Propositional Calculus

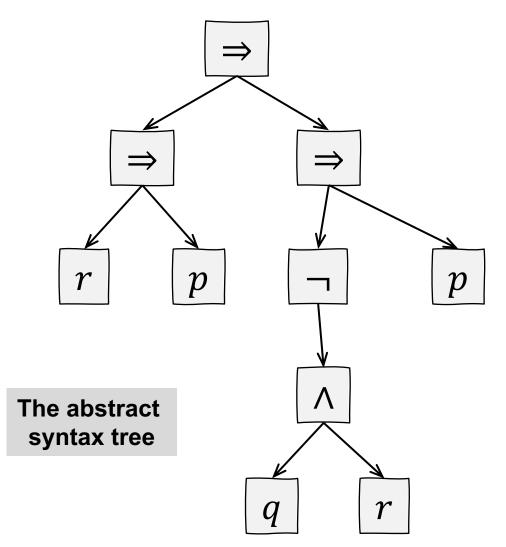
G. S. Tseitin 1966

1. Formulation of the Problem and Principal Results

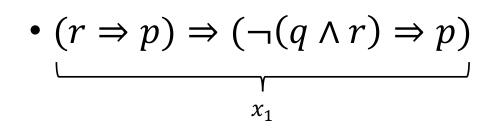
The question of the minimum complexity of derivation of a given formula in classical propositional calculus is considered in this article and it is proved that estimates of complexity may vary considerably among the various forms of propositional calculus. The forms of propositional calculus used in the present article are somewhat unusual, † but the results obtained for them can, in principle, be extended to the usual forms of propositional calculus.

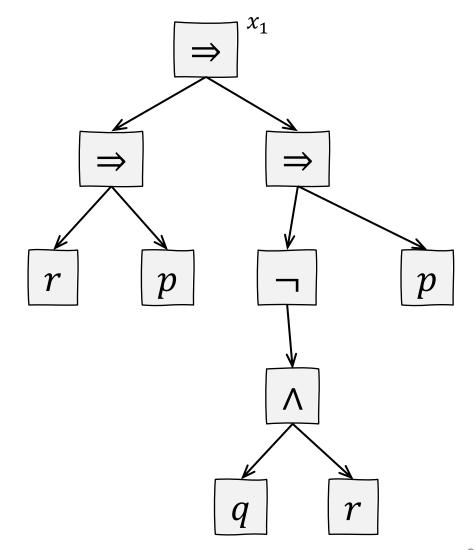


•
$$(r \Rightarrow p) \Rightarrow (\neg (q \land r) \Rightarrow p)$$

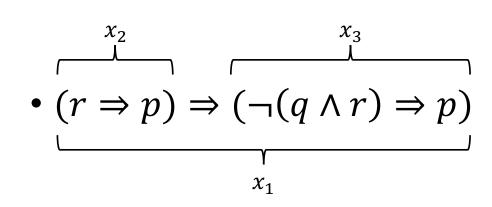


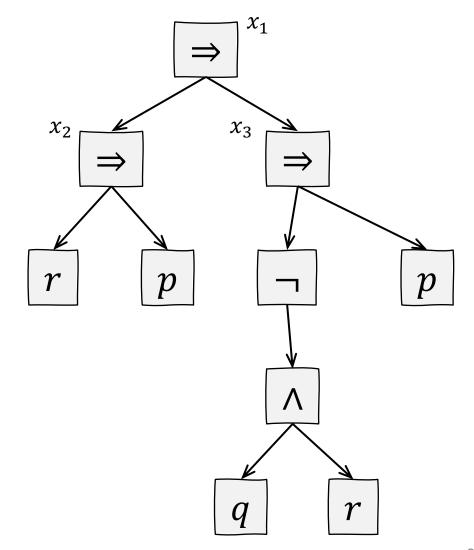




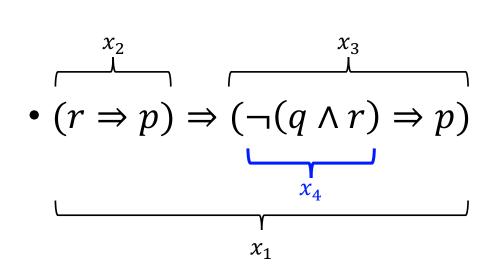


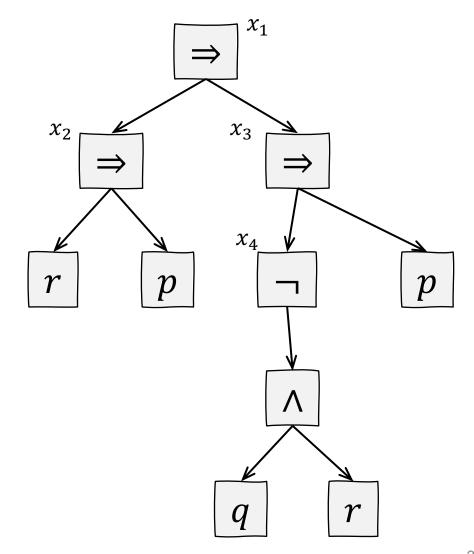




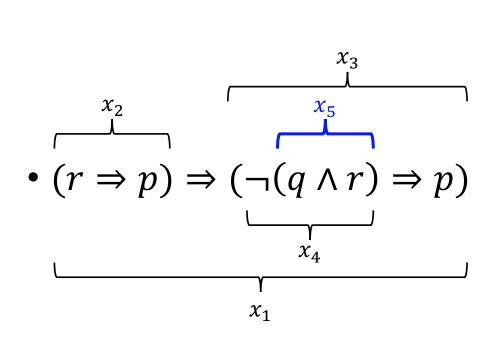


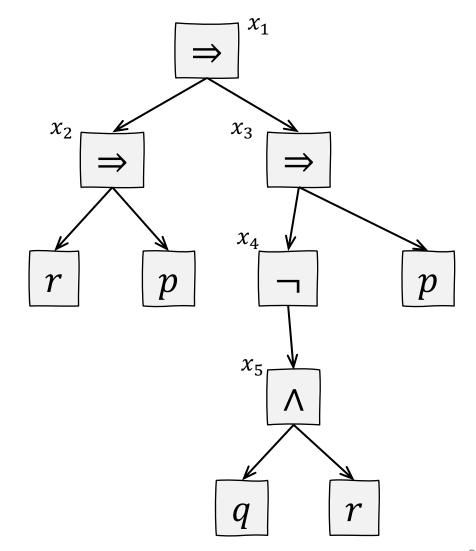




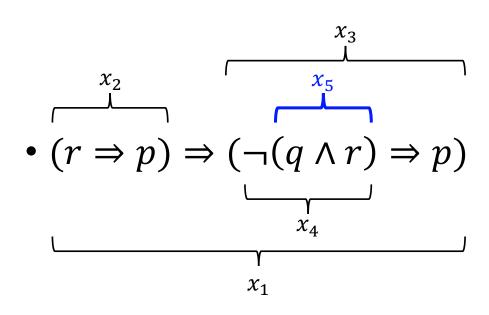






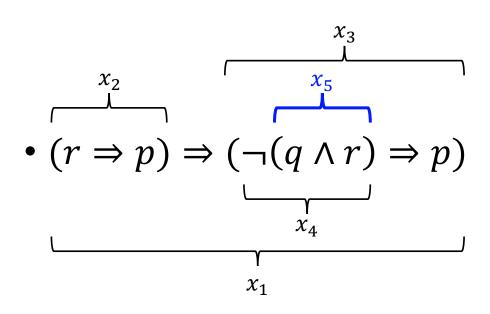






$$x_1 \land (x_1 \Leftrightarrow (x_2 \Rightarrow x_3))$$

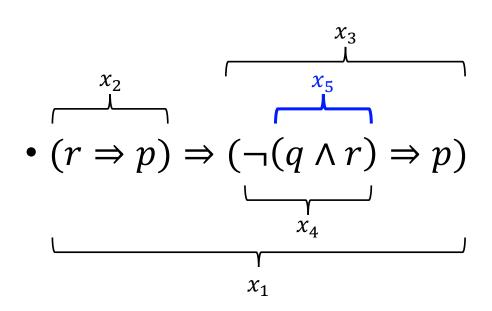




$$x_1 \land (x_1 \Leftrightarrow (x_2 \Rightarrow x_3))$$

 $\land x_2 \Leftrightarrow (r \Rightarrow p)$

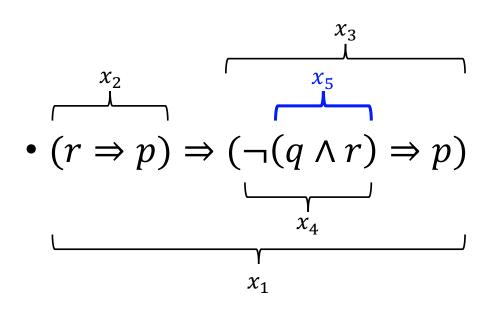




$$x_1 \land (x_1 \Leftrightarrow (x_2 \Rightarrow x_3))$$

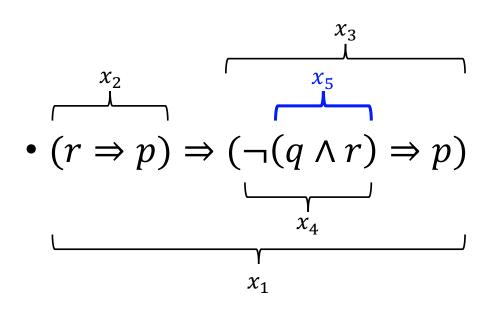
 $\land x_2 \Leftrightarrow (r \Rightarrow p)$
 $\land x_3 \Leftrightarrow (x_4 \Rightarrow p)$





$$x_1 \land (x_1 \Leftrightarrow (x_2 \Rightarrow x_3))$$
 $\land x_2 \Leftrightarrow (r \Rightarrow p)$
 $\land x_3 \Leftrightarrow (x_4 \Rightarrow p)$
 $\land x_4 \Leftrightarrow \neg x_5$





$$x_{1} \wedge (x_{1} \Leftrightarrow (x_{2} \Rightarrow x_{3}))$$

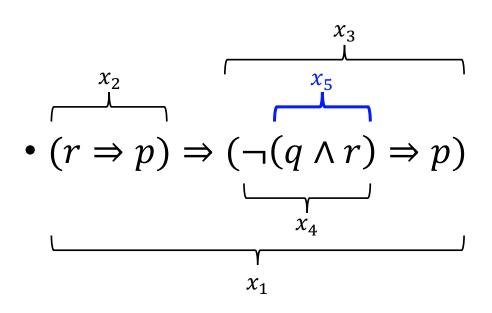
$$\wedge x_{2} \Leftrightarrow (r \Rightarrow p)$$

$$\wedge x_{3} \Leftrightarrow (x_{4} \Rightarrow p)$$

$$\wedge x_{4} \Leftrightarrow \neg x_{5}$$

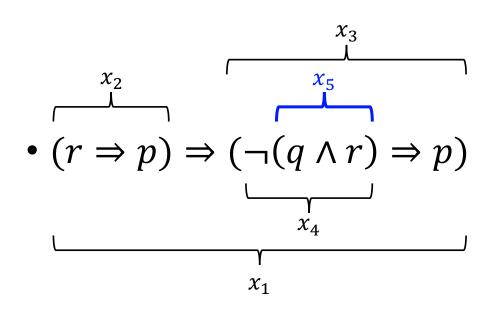
$$\wedge x_{5} \Leftrightarrow q \wedge r$$





$$x_1 \land (x_1 \Leftrightarrow (x_2 \Rightarrow x_3))$$
 $\land x_2 \Leftrightarrow (r \Rightarrow p)$
 $\land x_3 \Leftrightarrow (x_4 \Rightarrow p)$
 $\land x_4 \Leftrightarrow \neg x_5$
 $\land x_5 \Leftrightarrow q \land r$





$$x_{1} \wedge (x_{1} \Leftrightarrow (x_{2} \Rightarrow x_{3}))$$

$$\wedge x_{2} \Leftrightarrow (r \Rightarrow p)$$

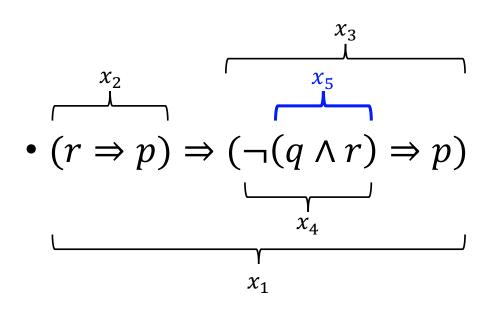
$$\wedge x_{3} \Leftrightarrow (x_{4} \Rightarrow p)$$

$$\wedge x_{4} \Leftrightarrow \neg x_{5}$$

$$\wedge x_{5} \Leftrightarrow q \wedge r$$

$$x_{4} \Rightarrow \neg x_{5} \wedge \neg x_{5} \Rightarrow x_{4}$$





$$x_{1} \wedge (x_{1} \Leftrightarrow (x_{2} \Rightarrow x_{3}))$$

$$\wedge x_{2} \Leftrightarrow (r \Rightarrow p)$$

$$\wedge x_{3} \Leftrightarrow (x_{4} \Rightarrow p)$$

$$\wedge x_{4} \Leftrightarrow \neg x_{5}$$

$$\wedge x_{5} \Leftrightarrow q \wedge r$$

$$x_{4} \Rightarrow \neg x_{5} \wedge \neg x_{5} \Rightarrow x_{4}$$

$$(\neg x_{4} \vee \neg x_{5}) \wedge (x_{5} \vee x_{4})$$



Solving Path Constraints

How can we check the satisfiability of a constraint?

$$\alpha_a \neq 0 \land \alpha_b = 0 \land \neg (2(\alpha_a + \alpha_b) - 4 \neq 0)$$

- Constraint solver
 - Satisfiable => A possible solution
 - Unsatisfiable
 - Unknown (due to timeout)







Satisfiability Modulo Theories

- For the theory of bit vectors
 - Step 1: Word-Level Preprocessing
 - Step 2: Bit-Blasting (From FOL to PL)
 - Step 3: DPLL/CDCL



Satisfiability Modulo Theories

- For the theory of bit vectors
 - Step 1: Word-Level Preprocessing
 - Step 2: Bit-Blasting (From FOL to PL)
 - Step 3: DPLL/CDCL

Example of Step 2:

$$x = [a_7, a_6, ..., a_0], y = [b_7, b_6, ..., b_0]$$
 and $z = [c_7, c_6, ..., c_0]$ are three 8bit bit vectors

$$x \mid y = z \xrightarrow{\text{Step 2}} \bigwedge_{i=0}^{7} ((a_i \lor b_i) \iff c_i)$$



PART III: Pointer Analysis



Andersen & Steensgaard Algorithm

Inclusion-based pointer analysis (Andersen's Algorithm)

Statement	Constraint	Shorthand
y = &x	$pts(y) \supseteq \{x\}$	
y = x	$pts(y) \supseteq pts(x)$	
*y = x	$\forall v \in pts(y): pts(v) \supseteq pts(x)$	pts(*y) ⊇ pts(x)
y = *x	$\forall v \in pts(x): pts(y) \supseteq pts(v)$	pts(y) ⊇ pts(*x)

Unification-based pointer analysis (Steensgaard's Algorithm)

Statement	Constraint	Shorthand
y = &x	$pts(y) \supseteq \{x\}$	
y = x	pts(y) = pts(x)	
*y = x	$\forall v \in pts(y): pts(v) = pts(x)$	pts(*y) = pts(x)
y = *x	$\forall v \in pts(x)$: $pts(y) = pts(v)$	pts(y) = pts(*x)



- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure



- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure

```
p = &a;
q = &b;
*p = q;
r = &c;
s = p;
t = *p;
*s = r;
```



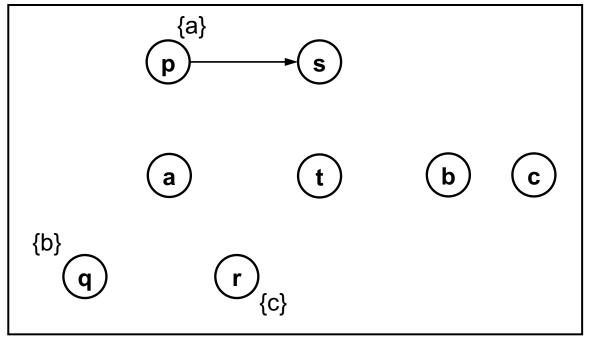
- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure

```
p = &a;
q = &b;
type = q;
r = &c;
s = p;
t = *p;
*s = r;
pts(p) ⊇ {a};
pts(q) ⊇ {b};
...
pts(r) ⊇ {c};
pts(s) ⊇ pts(p);
...
...
```



- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure

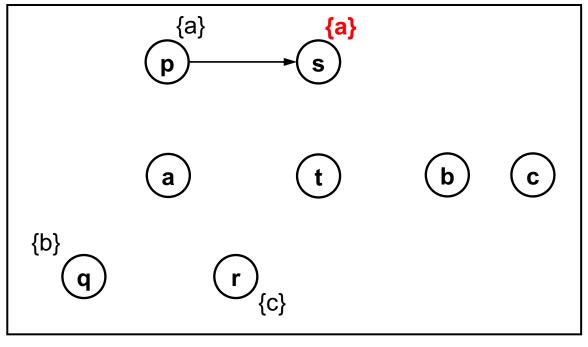
```
p = &a;
q = &b;
type = q;
r = &c;
s = p;
t = *p;
*s = r;
pts(p) ⊇ {a};
pts(q) ⊇ {b};
...
pts(q) ⊇ {b};
...
pts(r) ⊇ {c};
pts(s) ⊇ pts(p);
...
...
```





- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure

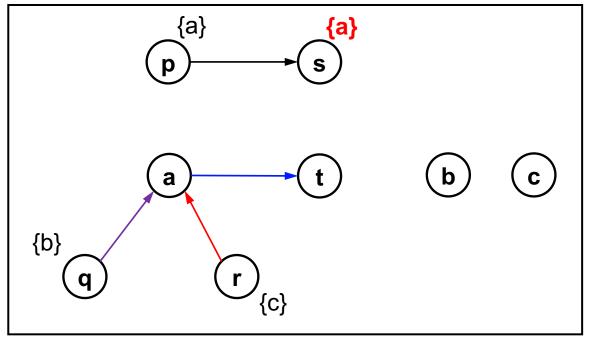
```
p = &a;
q = &b;
type = q;
r = &c;
s = p;
type = q;
```





- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure

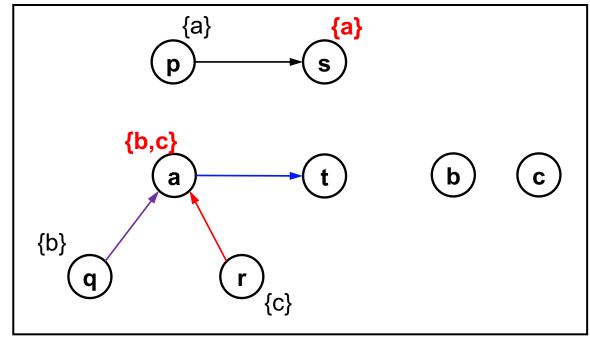
```
p = &a;
q = &b;
*p = q;
r = &c;
s = p;
t = *p;
*s = r;
...
pts(*p) ⊇ pts(q)
...
pts(t) ⊇ pts(*p)
pts(*s) ⊇ pts(*p)
pts(*s) ⊇ pts(r)
```





- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure

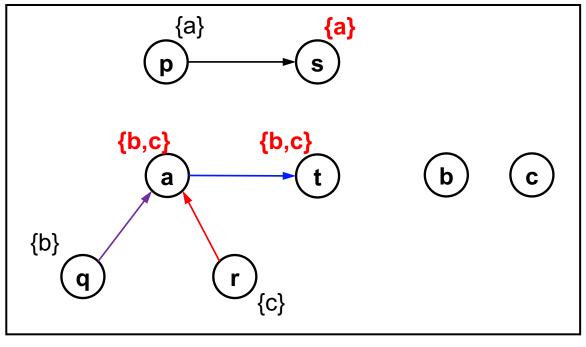
```
p = &a;
q = &b;
*p = q;
r = &c;
s = p;
t = *p;
*s = r;
...
pts(*p) ⊇ pts(q)
...
pts(t) ⊇ pts(*p)
pts(*s) ⊇ pts(*p)
pts(*s) ⊇ pts(r)
```





- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure

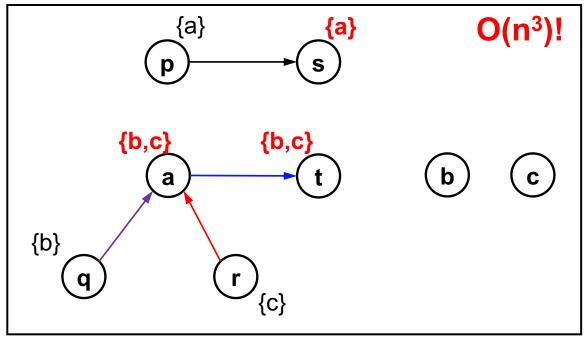
```
p = &a;
q = &b;
*p = q;
r = &c;
s = p;
t = *p;
*s = r;
...
pts(*p) ⊇ pts(q)
...
pts(t) ⊇ pts(*p)
pts(*s) ⊇ pts(*p)
pts(*s) ⊇ pts(r)
```





- Each graph node x denotes the points-to set pts(x)
- Solving the set constraints via a dynamic transitive closure

```
p = &a;
q = &b;
*p = q;
r = &c;
s = p;
t = *p;
*s = r;
...
pts(*p) ⊇ pts(q)
...
pts(t) ⊇ pts(*p)
pts(*s) ⊇ pts(*p)
pts(*s) ⊇ pts(r)
```





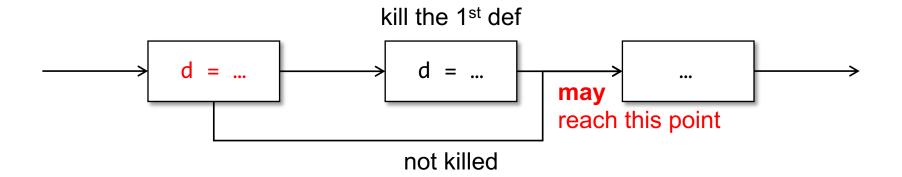
PART IV: Datalog-Based Analysis





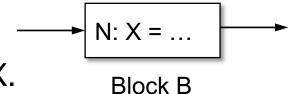
Recap: Reaching Definition

 A definition d may reach a program point, if there is a path from d to the program point such that d is not killed along the path.



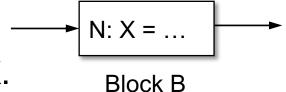


- def(B, N, X)
 - the Nth statement in Block B may define the variable X.





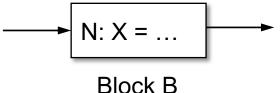
- def(B, N, X)
 - the Nth statement in Block B may define the variable X.



- succ(B, N, C)
 - block C is a successor of block B, and B has N statements.



- def(B, N, X)
 - the Nth statement in Block B may define the variable X.



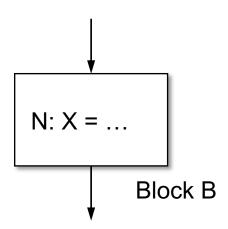
- succ(B, N, C)
 - block C is a successor of block B, and B has N statements.
- rd(B, N, C, M, X)
 - the definition of variable X at the Mth statement of block C reaches the Nth statement in B.



- rd(B, N, B, N, X) :- def(B, N, X)
- rd(B, N, C, M, X) :- rd(B, N-1, C, M, X), def(B, N, Y), X≠Y
- rd(B, 0, C, M, X) :- rd(D, N, C, M, X), succ(D, N, B)

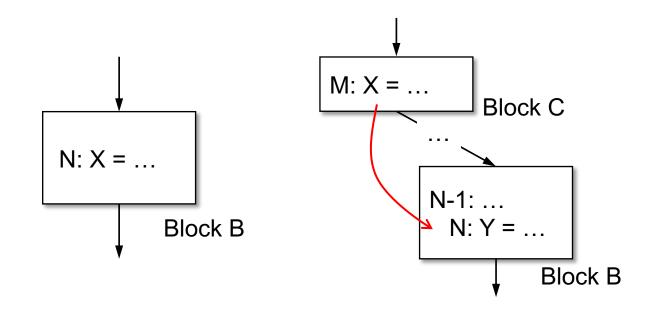


- rd(B, N, B, N, X) :- def(B, N, X)
- rd(B, N, C, M, X):- rd(B, N-1, C, M, X), def(B, N, Y), X≠Y
- rd(B, 0, C, M, X) :- rd(D, N, C, M, X), succ(D, N, B)



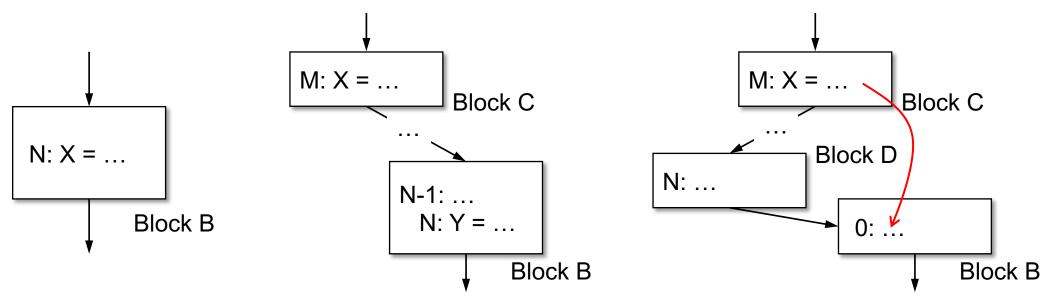


- rd(B, N, B, N, X) :- def(B, N, X)
- rd(B, N, C, M, X) :- rd(B, N-1, C, M, X), def(B, N, Y), X≠Y
- rd(B, 0, C, M, X) :- rd(D, N, C, M, X), succ(D, N, B)

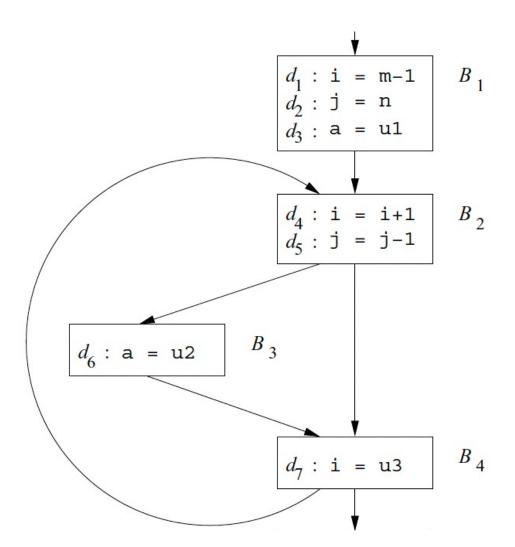




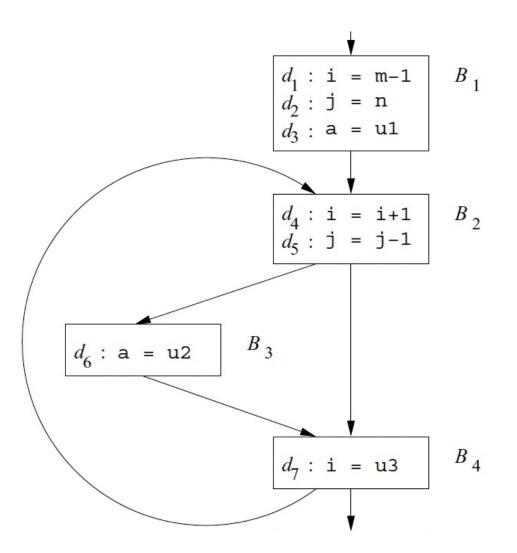
- rd(B, N, B, N, X) :- def(B, N, X)
- rd(B, N, C, M, X):- rd(B, N-1, C, M, X), def(B, N, Y), X≠Y
- rd(B, 0, C, M, X) :- rd(D, N, C, M, X), succ(D, N, B)





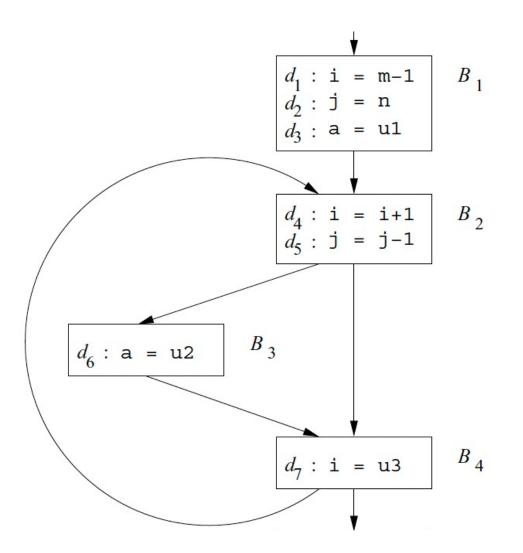






- def(B₁, 1, i)
- $def(B_1, 2, j)$
- def(B₁, 3, a)
- def(B₂, 1, i)
- def(B₂, 2, j)
- def(B₃, 1, a)
- def(B₄, 1, i)





- def(B₁, 1, i)
- $def(B_1, 2, j)$
- def(B₁, 3, a)
- def(B₂, 1, i)
- def(B₂, 2, j)
- def(B₃, 1, a)
- def(B₄, 1, i)

- succ(B₁, 3, B₂)
- succ(B₂, 2, B₃)
- succ(B₂, 2, B₄)
- succ(B₃, 1, B₄)
- succ(B₄, 1, B₂)



- rd(B, N, B, N, X) :- def(B, N, X)
- rd(B, N, C, M, X) :- rd(B, N-1, C, M, X), def(B, N, Y), X≠Y
- rd(B, 0, C, M, X) :- rd(D, N, C, M, X), succ(D, N, B)
- def(B₁, 1, i)
- $def(B_1, 2, j)$
- def(B₁, 3, a)
- def(B₂, 1, i)

- def(B₂, 2, j)
- def(B₃, 1, a)
- def(B₄, 1, i)
- succ(B₁, 3, B₂)

- succ(B₂, 2, B₃)
- succ(B₂, 2, B₄)
- succ(B₃, 1, B₄)
- succ(B₄, 1, B₂)



- rd(B, N, B, N, X) :- def(B, N, X)
- We just define facts and rules.
 - Analysis is automatically done by Datalog engines!

```
O(B_1, 1, i) Query Example: rd(B<sub>4</sub>, 1, B<sub>1</sub>, 1, i)
```

- def(B₁, 2, j)
- $def(B_1, 3, a)$ $def(B_4, 1, i)$ $succ(B_3, 1, B_4)$

• def(B₃, 1, a)

• $def(B_2, 1, i)$ • $succ(B_1, 3, B_2)$ • $succ(B_4, 1, B_2)$



THANKS!