Special Topics in Logic and Security 1 Dataflow Analysis

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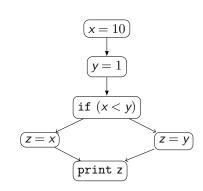
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What is Static Analysis?

Static analysis involves the analysis of program properties without executing them.

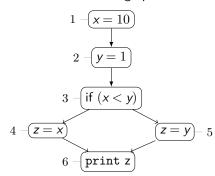
Properties examples: type analysis, null pointers, unused assignments, code vulnerabilities (e.g. buffer overflow), etc.

```
x = 10;
y = 1;
if (x < y) z = x;
    else z = y;
print z
```



```
x = 10;
y = 1;
if (x < y) z = x;
    else z=y;
print z
```

We label each of the graph's nodes.



Example: determining live variables

https://cs420.epfl.ch/c/06_dataflow-analysis.html

A variable is active (live) at a given point in the program's execution if it is possible for the current variable value to be read before it is overwritten.

For a node n from the CFG we will denote with v_n the set of variables that are live before the execution of n.

With that

$$v_n = \left(\bigcup \{v_i \mid i \text{ succesor al lui } n\} \setminus \textit{Written}(n)\right) \cup \textit{Read}(n)$$

where

- Written(n) is the set of variables that are redefined in n,
- Read(n) is the set of variables that are read in n.

Example: determining live variables

Taking into account the above identified condition, in each node we obtain a set of constraints:

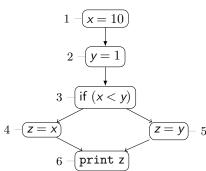
x = 10

$$v_{1} = v_{2} \setminus \{x\}$$
 $v_{2} = v_{3} \setminus \{y\}$
 $v_{3} = v_{4} \cup v_{5} \cup \{x, y\}$
 $v_{4} = (v_{6} \setminus \{z\}) \cup \{x\}$
 $v_{5} = (v_{6} \setminus \{z\}) \cup \{y\}$
 $v_{6} = \{z\}$
 $v_{6} = \{z\}$
 $v_{7} = v_{1} \cup v_{2} \cup v_{3} \cup v_{4} \cup v_{5} \cup$

Example: determining live variables

Solving the system we obtain the solution:

$$\begin{array}{rcl}
 v_1 & = & \emptyset \\
 v_2 & = & \{x\} \\
 v_3 & = & \{x, y\} \\
 v_4 & = & \{x\} \\
 v_5 & = & \{y\} \\
 v_6 & = & \{z\}
 \end{array}$$



CFG - remarks

- In the problem analysis from above, we associated a set of variables to each of the node from the control flow graph that represent the solution of a system of equations from $\mathcal{P}(\mathit{Var})$, where Var is the set of variables from the entire program.
- In general, the analysis of problems from this class leads to a set of equations of latice *L*. Is there a generic method for solving such a system of equations?

The Product Lattice

Definition

We define the pointwise order relationship:

$$(x_1,\ldots,x_n) \leq (y_1,\ldots,y_n) \iff x_i \leq y_i, \ \forall i \in \{1,\ldots,n\}$$

Theorem

If L_1, L_2, \ldots, L_n are lattices then so is the product

$$L_1 \times L_2 \times \cdots \times L_n = \{(x_1, x_2, \dots, x_n) \mid x_i \in L_i\}$$

where the lattice order \leq is defined pointwise.

Remark

Let L be a lattice and $n \ge 1$, note that (L^n, \le) is a lattice. More so, if L is a complete lattice then L^n is also a complete lattice.

Equation Systems in Complete Lattices

Let L be a complete lattice and $n \ge 1$ and $F_1, \dots, F_n : L^n \to L$ a set of monotone functions. We want to solve the equation system:

$$x_1 = F_1(x_1, \dots, x_n)$$

$$x_2 = F_2(x_1, \dots, x_n)$$

$$\dots$$

$$x_n = F_n(x_1, \dots, x_n)$$
with $x_1, \dots, x_n \in L$.

Definition

Denote the function $F: L^n \to L^n$ to be

$$F(x_1,...,x_n) = (F_1(x_1,...,x_n),...,F_1(x_1,...,x_n))$$

and observe that the system can be rewritten as

$$F(x_1,\ldots,x_n)=(x_1,\ldots,x_n)$$

The equation system can be solved using the Fixed Point Theorem!

Determining live variables

It results that, in order to solve the system:

$$v_1 = v_2 \setminus \{x\}
 v_2 = v_3 \setminus \{y\}
 v_3 = v_4 \cup v_5 \cup \{x, y\}
 v_4 = (v_6 \setminus \{z\}) \cup \{x\}
 v_5 = (v_6 \setminus \{z\}) \cup \{y\}
 v_6 = \{z\}$$

we need to find the solution to the equation:

$$F(x_{1}, x_{2}, x_{3}, x_{4}, x_{5}, x_{6}) = (x_{2} \setminus \{x\}, x_{3} \setminus \{y\}, x_{4} \cup x_{5} \cup \{x, y\}, (x_{6} \setminus \{z\}) \cup \{x\}, (x_{6} \setminus \{z\}) \cup \{y\}, \{z\})$$
 in the complet lattice $(\mathcal{P}(Var), \subset, \emptyset, Var)$.

Fixed Point Theorem Solution

$$F(x_1, x_2, x_3, x_4, x_5, x_6) = (x_2 \setminus \{x\}, x_3 \setminus \{y\}, x_3 \setminus \{y\}, (x_6 \setminus \{z\}) \cup \{x\}, (x_6 \setminus \{z\}) \cup \{y\}, \{z\})$$

Determining the least fixed-point:

| | <i>x</i> ₁ | x_2 | <i>x</i> ₃ | | <i>X</i> ₅ | |
|-------------------------|-----------------------|--------------|--|--------------|-----------------------|--------------|
| | Ø | Ø | Ø | Ø | Ø | Ø |
| $F(\perp)$ | Ø | Ø | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $\mathit{F}^{2}(\perp)$ | Ø | { <i>x</i> } | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $\mathit{F}^{3}(\perp)$ | Ø | { <i>x</i> } | $\begin{cases} x, y \\ \{x, y\} \\ \{x, y\} \end{cases}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |

where
$$\bot = (\emptyset, \dots, \emptyset)$$

IMP and live variables

Let the set of the variables successor to the node n be:

$$Join(v_n) = \bigcup \{v_i \mid i \text{ succesor al lui } n\}$$

Then, the general formulation for live variables in node k is:

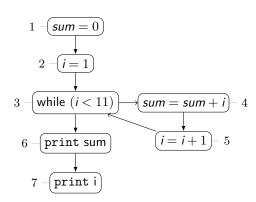
$$v_n = Join(v_n) \setminus Written(n) \cup Read(n)$$

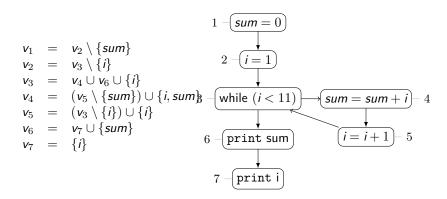
Particular formulations for IMP instructions:

- Expressions (E) E x + 3, (x > 7)
- Instruction blocks
 - Assignment x = E; $v_n = Join(v_n) \setminus \{x\} \cup Var(E)$ Conditional if (E) $v_n = Join(v_n) \cup Var(E)$ Loops while (E) $v_n = Join(v_n) \cup Var(E)$ Output print E; $v_n = Join(v_n) \cup Var(E)$

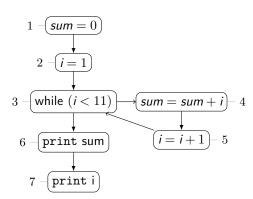
where Var(E) represents the set of variables present in expression E.

```
sum = 0;
i = 1;
while (i < 11) {
    sum = sum + i;
    i = i + 1;
}
print sum
print i</pre>
```





```
\begin{array}{lll} v_1 & = & \emptyset \\ v_2 & = & \{sum\} \\ v_3 & = & \{i, sum\} \\ v_4 & = & \{i, sum\} \\ v_5 & = & \{i, sum\} \\ v_6 & = & \{i, sum\} \\ v_7 & = & \{i\} \end{array}
```



Solving with the Fixed Point Theorem

$$\begin{split} F(x_1, x_2, x_3, x_4, x_5, x_6, x_7) &= \\ (x_2 \setminus \{\text{sum}\}, x_3 \setminus \{i\}, x_4 \cup x_6 \cup \{i\}, (x_5 \setminus \{\text{sum}\}) \cup \{i, \text{sum}\}, x_3 \setminus \{i\}) \cup \{i\}, x_7 \cup \{\text{sum}\}, \{i\}) \end{split}$$

Determining the least fixed-point:

| | x_1 | x_2 | <i>x</i> ₃ | X_4 | <i>x</i> ₅ | <i>x</i> ₆ | X7 |
|--------------|-------|-------|-----------------------|--------------|-----------------------|-----------------------|--------------|
| | Ø | Ø | Ø | Ø | Ø | Ø | Ø |
| $F(\perp)$ | Ø | Ø | Ø | Ø | Ø | {sum} | { <i>i</i> } |
| $F^2(\perp)$ | Ø | Ø | $\{sum, i\}$ | $\{sum, i\}$ | Ø | {sum} | { <i>i</i> } |
| $F^3(\perp)$ | Ø | {sum} | $\{sum, i\}$ | $\{sum, i\}$ | $\{sum, i\}$ | {sum} | $\{i\}$ |

where
$$\bot = (\emptyset, \dots, \emptyset)$$

Exercise

Perform liveness analysis for the following program:

```
x = 42;
while (x > 1) {
    y = x / 2;
    if (y > 3)
        x = x - y;
    z = x - 4;
    if (z > 0)
        x = x / 2;
    z = z - 1;
}
print x
```

Algorithms for determining the Ifp

In order to obtain the least-fixed point (Ifp) we use the following chain until we converge:

$$\mathbf{F}^0(\perp) = \perp \leq \mathbf{F}(\perp) \leq \mathbf{F}^2(\perp) \leq \cdots \leq \mathbf{F}^n(\perp) \leq \cdots$$

Which can be formalized in pseudo-code as:

Algorithm 1: NaiveLFP

- 1 $\mathbf{x} = (\bot, \bot, \ldots, \bot);$
- 2 while $\underline{x \neq F(x)}$ do
- x = F(x);
- 4 return x;

Complexity: depends on the height of lattice L^n and the evaluation cost of F(x).

Convergence: requires *n* iterations.

Algorithms for determining the Ifp

Algorithm 1 does not profit from the lattice structure:

- recomputes the inputs that were not changed in the former iteration
- does not take into consideration the inputs that were already computed in the current iteration

In the first example we had:

$$F(x_1, x_2, x_3, x_4, x_5, x_6) = (x_2 \setminus \{x\}, x_3 \setminus \{y\}, x_4 \cup x_5 \cup \{x, y\}, (x_6 \setminus \{z\}) \cup \{x\}, (x_6 \setminus \{z\}) \cup \{y\}, \{z\})$$

and the fixed-point is computed in 6 iterations

| | x_1 | x_2 | <i>x</i> ₃ | x_4 | <i>x</i> ₅ | <i>x</i> ₆ |
|-------------------------|-------|--------------|-----------------------|--------------|-----------------------|-----------------------|
| $\overline{\bot}$ | Ø | Ø | Ø | Ø | Ø | Ø |
| $F(\perp)$ | Ø | Ø | Ø | Ø | Ø | { <i>z</i> } |
| $F^2(\perp)$ | Ø | Ø | Ø | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $F^3(\perp)$ | Ø | Ø | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $\mathit{F}^{4}(\perp)$ | Ø | { <i>x</i> } | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $F^5(\perp)$ | Ø | { <i>x</i> } | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |

Comparisson of *Ifp* algorithms

Ideally, an efficient algorithm would reduce the number of iterations:

| | x_1 | x_2 | <i>x</i> ₃ | X_4 | <i>X</i> ₅ | <i>x</i> ₆ |
|-------------------------|-------|--------------|-----------------------|--------------|-----------------------|-----------------------|
| | Ø | Ø | Ø | Ø | Ø | Ø |
| $F(\perp)$ | Ø | Ø | Ø | | Ø | { <i>z</i> } |
| $\mathit{F}^{2}(\perp)$ | Ø | Ø | Ø | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $F^3(\perp)$ | Ø | Ø | | { <i>x</i> } | | { <i>z</i> } |
| $\mathit{F}^{4}(ot)$ | | { <i>x</i> } | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $F^5(\perp)$ | Ø | { <i>x</i> } | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |

versus

| | x_1 | x_2 | <i>x</i> ₃ | x_4 | <i>X</i> ₅ | <i>x</i> ₆ |
|--------------|-------|--------------|--|--------------|-----------------------|-----------------------|
| | Ø | Ø | Ø | Ø | Ø | Ø |
| $F(\perp)$ | Ø | Ø | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $F^2(\perp)$ | Ø | { <i>x</i> } | $\{x,y\}$ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |
| $F^3(\perp)$ | Ø | { <i>x</i> } | $ \begin{cases} x, y \\ x, y \\ x, y \end{cases} $ | { <i>x</i> } | { <i>y</i> } | { <i>z</i> } |

Round Robin Ifp

Gauss-Seidel-like component-wise updates:

- computes x_i based on $x_{j < i}$ from the current iteration
- Ifp is attained even though the outer iteration may provide a different output than the outer iteration in Alg. 1

```
Algorithm 2: RoundRobinLFP (x_1, x_2, \dots, x_n) = (\bot, \bot, \dots, \bot); while (x_1, x_2, \dots, x_n) \neq F(x_1, x_2, \dots, x_n) do
```

3 **for** $i \in 1 \rightarrow n$ **do** $x_i = \mathbf{F}_i(x_1, x_2, \dots, x_n);$

5 return $(x_1, x_2, ..., x_n)$;

Complexity: depends on the height of lattice L^n and the cost of evaluating $F_i(x)$.

Convergence: requires at most *n* iterations.

Chaotic Ifp

Stochastic component wise updates:

- the order of operations is not important in Alg. 2
- computes x_i based on the x_j 's from the current iteration; in no particular order
- Alg. 2 still recomputes the unchanged inputs from the former iteration
- the constraints need to be applied until convergence

Algorithm 3: ChaoticLFP

```
1 (x_1, x_2, ..., x_n) = (\bot, \bot, ..., \bot);

2 while (x_1, x_2, ..., x_n) \neq F(x_1, x_2, ..., x_n) do

3 i = random(1 \rightarrow n);

4 x_i = F_i(x_1, x_2, ..., x_n);

5 return (x_1, x_2, ..., x_n);
```

Complexity: first, it depends on how *i* is chosen at each iteartion, then it depends on the height of lattice L^n and then on the cost of evaluating $F_i(\mathbf{x})$.

Convergence: the algorithm is not guaranteed to stop, but if it does the result is correct.

SimpleWorkList Ifp

Define the map $dep: Nodes \rightarrow 2^{Nodes}$, where dep(v) is the set of nodes whose information depends on the information of node v.

Algorithm 4: SimpleWorkListLFP

```
1 (x_1, x_2, ..., x_n) = (\bot, \bot, ..., \bot);

2 W = \{v_1, v_2, ..., v_n\};

3 while \underline{W \neq \emptyset} do

4 v_i = W.removeNext();

5 y = F_i(x_1, x_2, ..., x_n);

6 if \underline{y \neq x_i} then

7 x_i = y;

8 x_i = y;

9 w_i = W
```

The functions removeNext() and add() choose a random node from W and add a node in W, respectively.

SimpleWorkList Ifp

Properties

- computes only the particular constraints for the current node
- does not recompute the inputs unchanged during the former iteration
- computes x_i based on the x_j 's from the current iteration; in no particular order
- constraints are applied until convergence
- each iteration has the same effect as the Alg. 3 iteration

Complexity: depends on the number of nodes n, the lattice height h, and the cost of evaluating $\mathbf{F}_i(\mathbf{x})$.

Convergence: a single iteration either decreases the work volume in W or climbs up the lattice; the algorithm stops in finite time because the lattice height is finite and the iterations stop when W is void.