

Title: Seidl: Programoptimierung (26.11.2015)

Date: Thu Nov 26 08:35:22 CET 2015

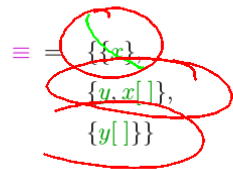
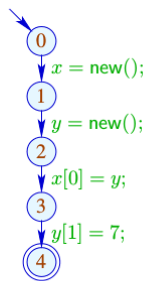
Duration: 90:23 min

Pages: 30

Alias Analysis 3. Idea

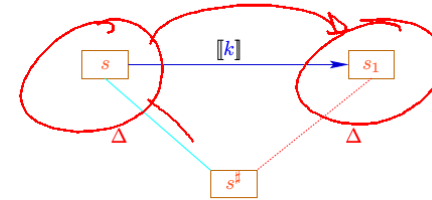
Determine one equivalence relation \equiv on variables x and memory accesses $y[]$ with $s_1 \equiv s_2$ whenever s_1, s_2 may contain the same address at some u_1, u_2

... in the Simple Example



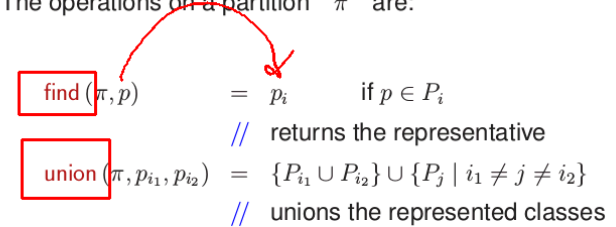
Discussion

- The resulting constraint system has size $\mathcal{O}(k \cdot n)$ for k abstract addresses and n edges.
- The number of necessary iterations is $\mathcal{O}(k(k + \#Vars))$...
- The computed information is perhaps still too **zu precise !!?**
- In order to prove correctness of a solution $s^\# \in States^\#$ we show:

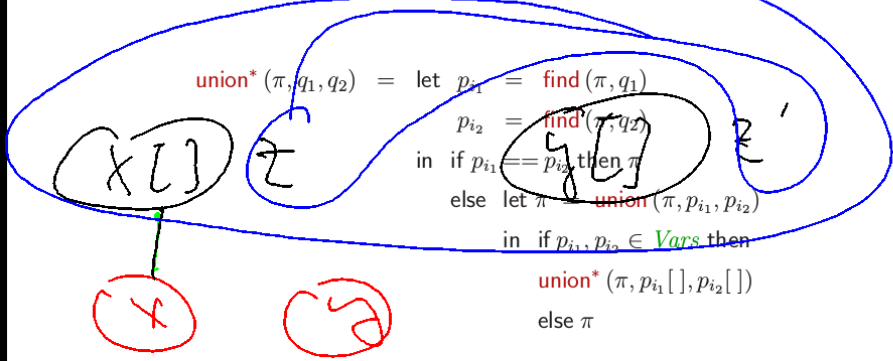


Discussion

- We compute a **single information** for the whole program.
- The computation of this information maintains **partitions** $\pi = \{P_1, \dots, P_m\}$.
- Individual sets P_i are identified by means of **representatives** $p_i \in P_i$.
- The operations on a partition π are:



- If $x_1, x_2 \in Vars$ are equivalent, then also $x_1[]$ and $x_2[]$ must be equivalent.
- If $P_i \cap Vars \neq \emptyset$, then we choose $p_i \in Vars$. Then we can apply **union recursively**:



The analysis iterates over all edges **once**:

```

pi = {{x}, {x[]} | x in Vars};
forall k = (_, lab, _) do pi = [[lab]]#pi;

```

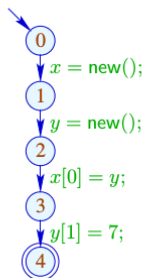
where:

```

[x = y;]#pi = union*(pi, x, y)
[x = y[e];]#pi = union*(pi, x, y[])
[y[e] = x;]#pi = union*(pi, x, y[])
[[lab]]#pi = pi otherwise

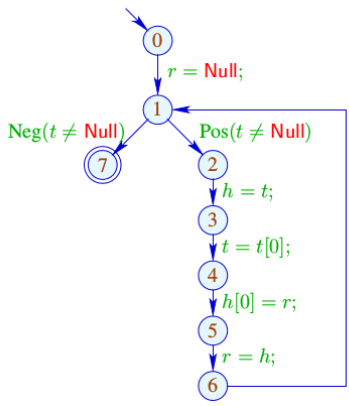
```

... in the Simple Example



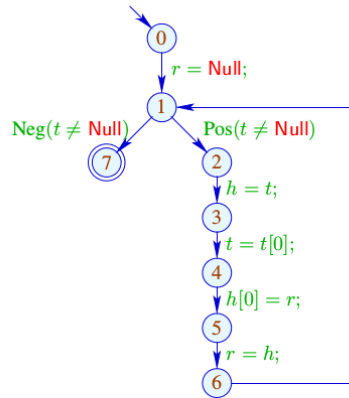
	{{x}, {y}, {x[]}, {y[]}}
(0, 1)	{{x}, {y}, {x[]}, {y[]}}
(1, 2)	{{x}, {y}, {x[]}, {y[]}}
(2, 3)	{{x}, {y, x[]}, {y[]}}
(3, 4)	{{x}, {y, x[]}, {y[]}}

... in the More Complex Example



	{{h}, {t}, {t[]}, {h[]}, {t[]}}
(2, 3)	{{h, t}, {t}, {h[], t[]}}
(3, 4)	{{h, t, h[], t[]}, {r}}
(4, 5)	{{h, t, r, h[], t[]}}
(5, 6)	{{h, t, r, h[], t[]}}

... in the More Complex Example



	$\{\{h\}, \{r\}, \{t\}, \{h[]\}, \{t[]\}\}$
(2, 3)	$\{\{h, t\}, \{r\}, \{h[], t[]\}\}$
(3, 4)	$\{\{h, t, h[], t[]\}, \{r\}\}$
(4, 5)	$\{\{h, t, r, h[], t[]\}\}$
(5, 6)	$\{\{h, t, r, h[], t[]\}\}$

385

Caveat

In order to find something, we must assume that variables / addresses always receive a value before they are accessed.

Complexity

we have:

$O(\# \text{ edges} + \# \text{ Vars})$ calls of **union***

$O(\# \text{ edges} + \# \text{ Vars})$ calls of **find**

$O(\# \text{ Vars})$ calls of **union**

⇒ We require efficient **Union-Find** data-structure ...

386

Idea

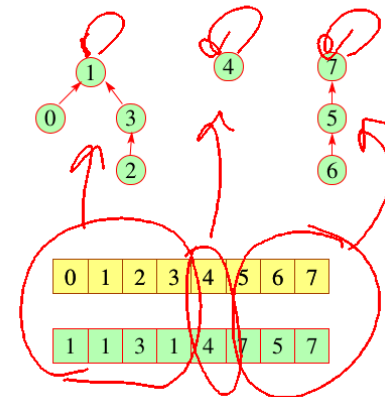
Represent partition of U as directed forest:

- For $u \in U$ a reference $F[u]$ to the father is maintained;
- Roots are elements u with $F[u] = u$.

Single trees represent equivalence classes.

Their roots are their representatives ...

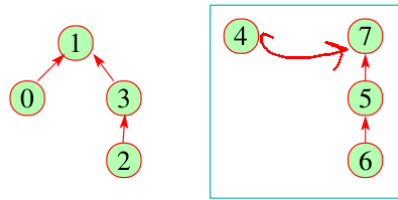
387



→ **find**(π, u) follows the father references.

→ **union**(π, u_1, u_2) re-directs the father reference of one u_i ...

388



0	1	2	3	4	5	6	7
1	1	3	1	4	7	5	7

389

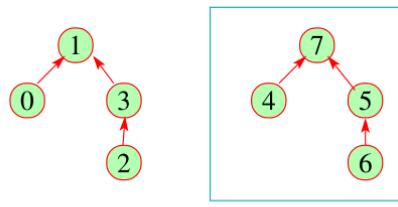
The Costs

union : $\mathcal{O}(1)$
 find : $\mathcal{O}(\text{depth}(\pi))$

Strategy to Avoid Deep Trees

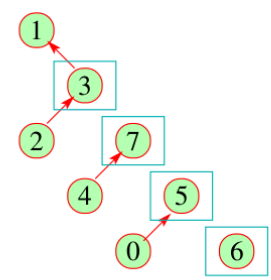
- Put the **smaller** tree below the **bigger** !
- Use **find** to compress paths ...

391



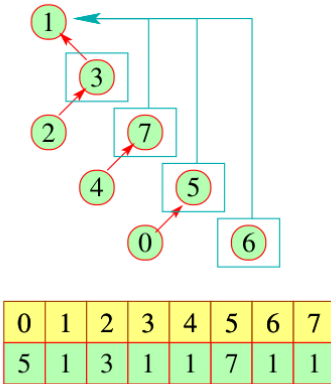
0	1	2	3	4	5	6	7
1	1	3	1	7	7	5	7

393



0	1	2	3	4	5	6	7
5	1	3	1	7	7	5	3

397



398

Remark

- By this data-structure, n union- und m find operations require time $\mathcal{O}(n + m \cdot \alpha(n, n))$
// α the inverse Ackermann-function.
- For our application, we only must modify union such that roots are from Vars whenever possible.
- This modification does not increase the asymptotic run-time.

Summary

The analysis is extremely fast — but may not find very much.

400



Robert Endre Tarjan, Princeton

399

Remark

$n^2 \cdot \ell$

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400

Background 3: Fixpoint Algorithms

Consider: $x_i \sqsupseteq f_i(x_1, \dots, x_n), \quad i = 1, \dots, n$

Observation

RR-Iteration is **inefficient**:

- We require a complete round in order to detect termination.
- If in some round, the value of just one unknown is changed, then we still re-compute all.
- The practical run-time depends on the ordering on the variables.

401

Idea:

Worklist Iteration

If an unknown x_i changes its value, we re-compute all unknowns which depend on x_i . **Technically**, we require:

- the lists $Dep f_i$ of unknowns which are accessed during evaluation of f_i . From that, we compute the lists:

$$I[x_i] = \{x_j \mid x_i \in Dep f_j\}$$

i.e., a list of all x_j which depend on the value of x_i ;

- the values $D[x_i]$ of the x_i where initially $D[x_i] = \perp$;
- a list W of all unknowns whose value must be recomputed ...

402

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401

The Algorithm

```
W = [x1, ..., xn];
while (W ≠ []) {
    xi = extract W;
    t = fi eval;
    if (t ∉ D[xi]) {
        D[xi] = D[xi] ⊔ t;
        W = append I[xi] W;
    }
}
where: eval xj = D[xj]
```

403

Example

$$\begin{aligned}
 x_1 &\supseteq \{a\} \cup x_3 \\
 x_2 &\supseteq x_3 \cap \{a, b\} \\
 x_3 &\supseteq x_1 \cup \{c\}
 \end{aligned}$$

	I
x_1	$\{x_3\}$
x_2	\emptyset
x_3	$\{x_1, x_2\}$

404

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x_2	\emptyset
x_3	$\{x_1, x_2\}$

$D[x_1]$	$D[x_2]$	$D[x_3]$	W
\emptyset	\emptyset	\emptyset	x_1, x_2, x_3
$\{a\}$	\emptyset	\emptyset	x_2, x_3
$\{a\}$	\emptyset	\emptyset	x_3
$\{a\}$	\emptyset	$\{a, c\}$	x_1, x_2
$\{a, c\}$	\emptyset	$\{a, c\}$	x_3, x_2
$\{a, c\}$	\emptyset	$\{a, c\}$	x_2
$\{a, c\}$	$\{a\}$	$\{a, c\}$	$[]$

405

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$D[x_1]$	$D[x_2]$	$D[x_3]$	W
\emptyset	\emptyset	\emptyset	x_1, x_2, x_3
$\{a\}$	\emptyset	\emptyset	x_2, x_3
$\{a\}$	\emptyset	\emptyset	x_3
$\{a\}$	\emptyset	$\{a, c\}$	x_1, x_2
$\{a, c\}$	\emptyset	$\{a, c\}$	x_3, x_2
$\{a, c\}$	\emptyset	$\{a, c\}$	x_2
$\{a, c\}$	$\{a\}$	$\{a, c\}$	$[]$

405

Theorem

Let $x_i \supseteq f_i(x_1, \dots, x_n)$, $i = 1, \dots, n$ denote a constraint system over the complete lattice \mathbb{D} of height $h > 0$.

- (1) The algorithm terminates after at most $h \cdot N$ evaluations of right-hand sides where

$$N = \sum_{i=1}^n (1 + \#(\text{Dep } f_i)) \quad // \text{ size of the system}$$

- (2) The algorithm returns a solution. If all f_i are monotonic, it returns the least one.

406

Proof

Ad (1):

Every unknown x_i may change its value at most h times.

Each time, the list $I[x_i]$ is added to W .

Thus, the total number of evaluations is:

$$\begin{aligned} &\leq n + \sum_{i=1}^n (h \cdot \#(I[x_i])) \\ &= n + h \cdot \sum_{i=1}^n \#(I[x_i]) \\ &= n + h \cdot \sum_{i=1}^n \#(Dep f_i) \\ &\leq h \cdot \sum_{i=1}^n (1 + \#(Dep f_i)) \\ &= h \cdot N \end{aligned}$$

407

Ad (2):

We only consider the assertion for monotonic f_i .

Let D_0 denote the least solution. We show:

- $D_0[x_i] \supseteq D[x_i]$ (all the time)
- $D[x_i] \not\models f_i \text{ eval} \implies x_i \in W$ (at exit of the loop body)
- On termination, the algo returns a solution

408

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407