

The code for **Round Robin** Iteration in **Java** looks as follows:

```
for (i = 1; i ≤ n; i++) xi = ⊥;
do {
    finished = true;
    for (i = 1; i ≤ n; i++) {
        new = fi(x1, ..., xn);
        if (!(xi ⊇ new)) {
            finished = false;
            xi = xi ⊔ new;
        }
    }
} while (!finished);
```

Correctness:

Assume $y_i^{(d)}$ is the i -th component of $F^d \underline{\underline{1}}$.

Assume $x_i^{(d)}$ is the value of x_i after the d -th RR-iteration.

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(2) $x_i^{(d)} \sqsubseteq z_i$ for every solution (z_1, \dots, z_n) :-)

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(3) If RR-iteration terminates after d rounds, then
 $(x_1^{(d)}, \dots, x_n^{(d)})$ is a solution :-))

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The efficiency of **RR**-iteration depends on the **ordering** of the unknowns **!!!**

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- u before v , if $u \rightarrow^* v$;
- entry condition before loop body :-)

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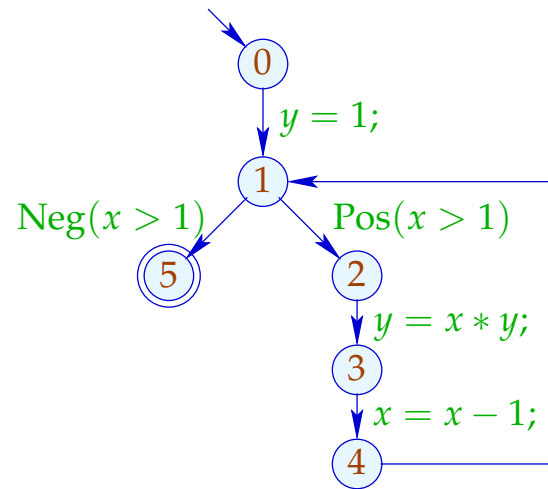
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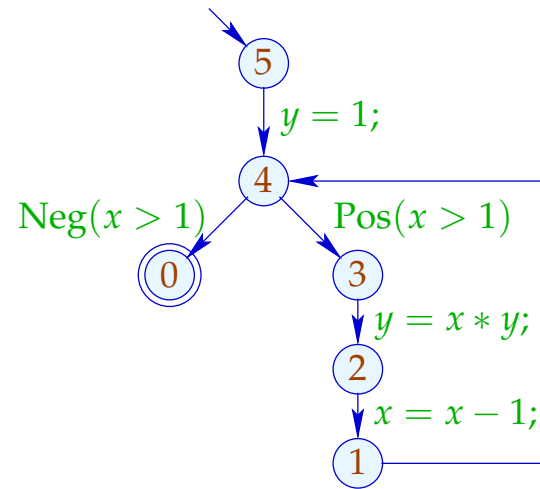
Bad:

e.g., post-order DFS of the CFG, starting at **start** :-)

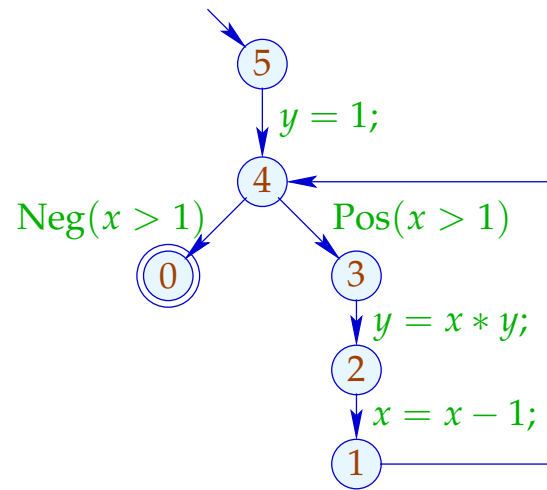
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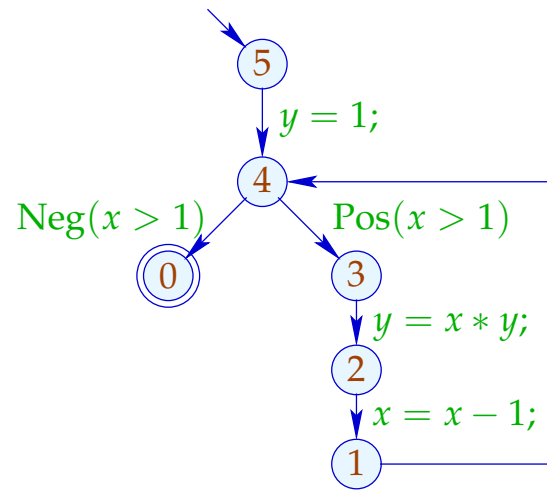


Inefficient Round Robin Iteration:



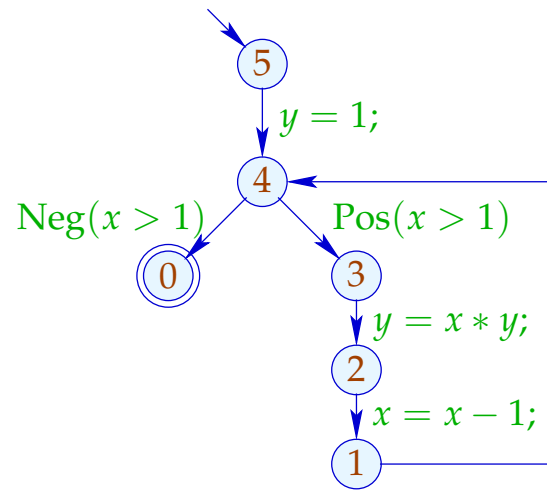
| |
|---|
| |
| 0 |
| 1 |
| 2 |
| 3 |
| 4 |
| 5 |

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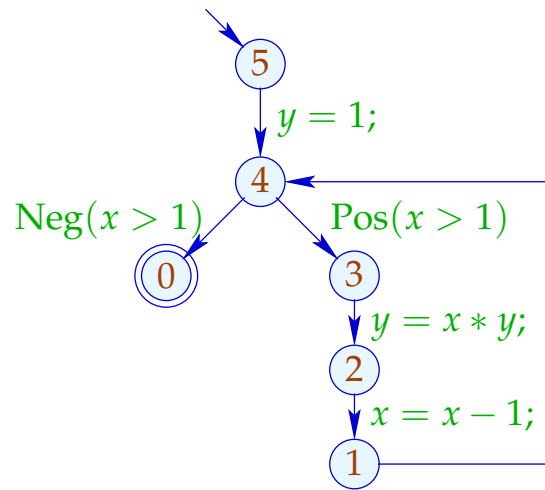
| | |
|---|-------------------------|
| | 1 |
| 0 | <i>Expr</i> |
| 1 | {1} |
| 2 | {1, $x - 1$, $x > 1$ } |
| 3 | <i>Expr</i> |
| 4 | {1} |
| 5 | \emptyset |

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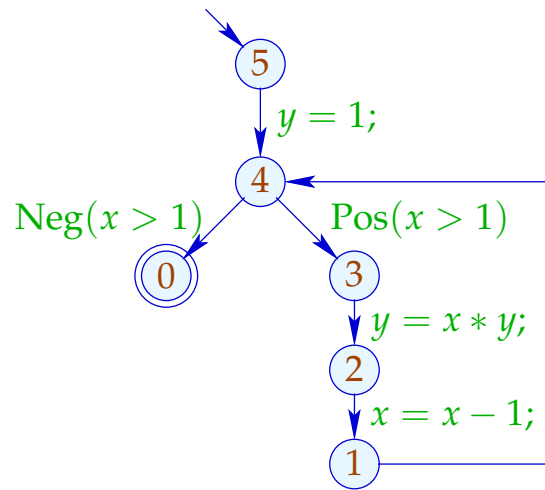
| | 1 | 2 |
|---|----------------------|----------------------|
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Inefficient Round Robin Iteration:



| | 1 | 2 | 3 |
|---|-------------------|-------------------|------------|
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| 5 | ∅ | ∅ | ∅ |

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| | 1 | 2 | 3 | 4 |
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⇒ significantly less efficient :-)

... end of background on: **Complete Lattices**

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For a complete lattice \mathbb{D} , consider systems:

$$\mathcal{I}[\textit{start}] \sqsupseteq d_0$$

$$\mathcal{I}[v] \sqsupseteq \llbracket k \rrbracket^\# (\mathcal{I}[u]) \quad k = (u, _, v) \text{ edge}$$

where $d_0 \in \mathbb{D}$ and all $\llbracket k \rrbracket^\# : \mathbb{D} \rightarrow \mathbb{D}$ are monotonic ...

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Monotonic Analysis Framework

Wanted: **MOP** (Merge Over all Paths)

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Assume \mathcal{I} is a solution of the constraint system. Then:

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Jeffrey D. Ullman, Stanford

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$$\mathcal{I}[v] \supseteq \mathcal{I}^*[v] \quad \text{for every } v$$

In particular: $\mathcal{I}[v] \supseteq \llbracket \pi \rrbracket^\# d_0$ for every $\pi : \textit{start} \rightarrow^* v$

Proof: Induction on the length of π .

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Foundation: $\pi = \epsilon$ (empty path)

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Then:

$$\begin{aligned} [[\pi']]^\# d_0 &\sqsubseteq \mathcal{I}[u] && \text{by I.H. for } \pi \\ \implies [[\pi]]^\# d_0 &= [[k]]^\# ([[\pi']]^\# d_0) \\ &\sqsubseteq [[k]]^\# (\mathcal{I}[u]) && \text{since } [[k]]^\# \text{ monotonic} \\ &\sqsubseteq \mathcal{I}[v] && \text{since } \mathcal{I} \text{ solution } \text{:-)} \end{aligned}$$

Disappointment:

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With the notable exception when all functions $[[k]]^\#$ are **distributive** ... :-)

The function $f : \mathbb{D}_1 \rightarrow \mathbb{D}_2$ is called

- **distributive**, if $f(\sqcup X) = \sqcup\{f x \mid x \in X\}$ for all $\emptyset \neq X \subseteq \mathbb{D}$;
- **strict**, if $f \perp = \perp$.
- **totally distributive**, if f is distributive and strict.

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Distributivity:

$$\begin{aligned} f(x_1 \cup x_2) &= a \cap (x_1 \cup x_2) \cup b \\ &= a \cap x_1 \cup a \cap x_2 \cup b \\ &= f x_1 \cup f x_2 \quad \text{:-)} \end{aligned}$$

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Distributivity:

$$\begin{aligned} f((1,4) \sqcup (4,1)) &= f(4,4) = 8 \\ &\neq 5 = f(1,4) \sqcup f(4,1) \quad :-) \end{aligned}$$

Remark:

If $f : \mathbb{D}_1 \rightarrow \mathbb{D}_2$ is distributive, then also monotonic :-)

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From that follows:

$$\begin{aligned} f b &= f (a \sqcup b) \\ &= f a \sqcup f b \\ \implies f a &\sqsubseteq f b \quad \text{:-)} \end{aligned}$$

Assumption: all v are reachable from *start* .

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Then:

Theorem

Kildall 1972

If **all** effects of edges $[[k]]^\#$ are distributive, then: $\mathcal{I}^*[v] = \mathcal{I}[v]$
for all v .



Gary A. Kildall (1942-1994).

Has developed the operating system CP/M and GUIs for PCs.

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Proof:

It suffices to prove that \mathcal{I}^* is a solution $:-)$

For this, we show that \mathcal{I}^* satisfies all constraints $:-))$

(1) We prove for *start* :

$$\begin{aligned} \mathcal{I}^*[start] &= \bigsqcup \{ \llbracket \pi \rrbracket^\# d_0 \mid \pi : start \rightarrow^* start \} \\ &\supseteq \llbracket \epsilon \rrbracket^\# d_0 \\ &\supseteq d_0 \quad :-) \end{aligned}$$

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(2) For every $k = (u, _, v)$ we prove:

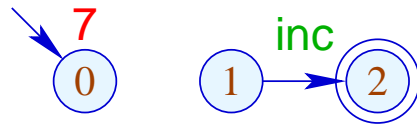
$$\begin{aligned}
\mathcal{I}^*[v] &= \sqcup \{ [[\pi]]^\# d_0 \mid \pi : start \rightarrow^* v \} \\
&\supseteq \sqcup \{ [[\pi'k]]^\# d_0 \mid \pi' : start \rightarrow^* u \} \\
&= \sqcup \{ [[k]]^\# ([[\pi']]^\# d_0) \mid \pi' : start \rightarrow^* u \} \\
&= [[k]]^\# (\sqcup \{ [[\pi']]^\# d_0 \mid \pi' : start \rightarrow^* u \}) \\
&= [[k]]^\# (\mathcal{I}^*[u])
\end{aligned}$$

since $\{ \pi' \mid \pi' : start \rightarrow^* u \}$ is non-empty :-)

Warning:

- **Reachability** of all program points cannot be abandoned!

Consider:

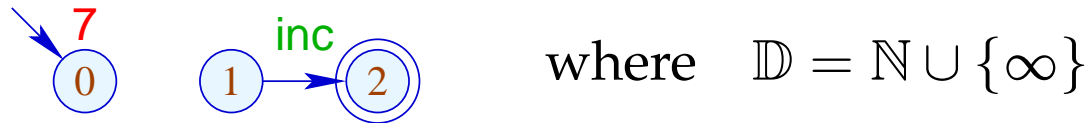


where $\mathbb{D} = \mathbb{N} \cup \{\infty\}$

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Then:

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$$\mathcal{I}[2] = \text{inc } 0 = 1$$

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- **Unreachable** program points can always be thrown away :-)

Summary and Application:

- The effects of edges of the analysis of **availability of expressions** are distributive:

$$\begin{aligned}(a \cup (x_1 \cap x_2)) \setminus b &= ((a \cup x_1) \cap (a \cup x_2)) \setminus b \\ &= ((a \cup x_1) \setminus b) \cap ((a \cup x_2) \setminus b)\end{aligned}$$

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- If all effects of edges are **distributive**, then the **MOP** can be computed by means of the constraint system and **RR-iteration**. :-)
- If **not all** effects of edges are **distributive**, then **RR-iteration** for the constraint system at least returns a **safe** upper bound to the MOP :-)

1.2 Removing Assignments to Dead Variables

Example:

1 : $x = y + 2;$

2 : $y = 5;$

3 : $x = y + 3;$

The value of x at program points 1, 2 is over-written before it can be used.

Therefore, we call the variable x **dead** at these program points :-)

Note:

- Assignments to dead variables can be removed ;-)
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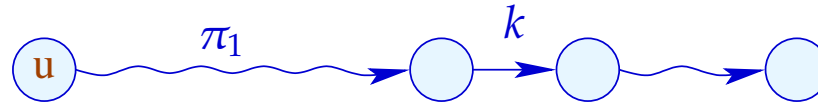
Formal Definition:

The variable x is called **live** at u along the path π starting at u relative to a set X of variables either:

if $x \in X$ and π does not contain a **definition** of x ; or:

if π can be decomposed into: $\pi = \pi_1 k \pi_2$ such that:

- k is a **use** of x ; and
- π_1 does not contain a **definition** of x .

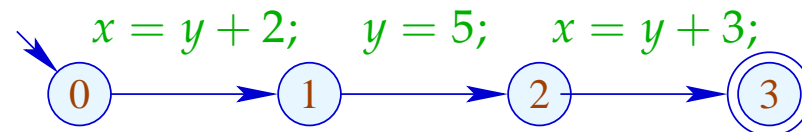


Thereby, the set of all defined or used variables at an edge $k = (_, lab, _)$ is defined by:

| <i>lab</i> | <i>used</i> | <i>defined</i> |
|--|------------------------------|----------------|
| <i>;</i> | \emptyset | \emptyset |
| <i>Pos (e)</i> | $Vars (e)$ | \emptyset |
| <i>Neg (e)</i> | $Vars (e)$ | \emptyset |
| <i>x = e;</i> | $Vars (e)$ | $\{x\}$ |
| <i>x = M[e];</i> | $Vars (e)$ | $\{x\}$ |
| <i>M[e₁] = e₂;</i> | $Vars (e_1) \cup Vars (e_2)$ | \emptyset |

A variable x which is not live at u along π (relative to X) is called **dead** at u along π (relative to X).

Example:



where $X = \emptyset$. Then we observe:

| | live | dead |
|---|-------------|------------|
| 0 | { y } | { x } |
| 1 | \emptyset | { x, y } |
| 2 | { y } | { x } |
| 3 | \emptyset | { x, y } |

The variable x is **live** at u (relative to X) if x is live at u along **some** path to the exit (relative to X). Otherwise, x is called **dead** at u (relative to X).

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Idea:

For every edge $k = (u, _, v)$, define a function $[[k]]^\#$ which transforms the set of variables which are live at v into the set of variables which are live at u ...