

Principles of Program Analysis:

Algorithms

Transparencies based on Chapter 6 of the book: Flemming Nielson, Hanne Riis Nielson and Chris Hankin: [Principles of Program Analysis](#). Springer Verlag 2005. ©Flemming Nielson & Hanne Riis Nielson & Chris Hankin.

Worklist Algorithms

We abstract away from the details of a particular analysis:

We want to compute the solution to a set of **equations**

$$\{x_1 = t_1, \quad \dots, \quad x_N = t_N\}$$

or **inequations**

$$\{x_1 \supseteq t_1, \quad \dots, \quad x_N \supseteq t_N\}$$

defined in terms of a set of **flow variables** x_1, \dots, x_N ; here t_1, \dots, t_N are **terms** using the flow variables.

Equations or inequations?

It does not really matter:

- A solution of the equation system $\{x_1 = t_1, \dots, x_N = t_N\}$ is also a solution of the inequation system $\{x_1 \sqsupseteq t_1, \dots, x_N \sqsupseteq t_N\}$
- The least solution to the inequation systems $\{x_1 \sqsupseteq t_1, \dots, x_N \sqsupseteq t_N\}$ is also a solution to the equation system $\{x_1 = t_1, \dots, x_N = t_N\}$
 - The inequation system $\{x \sqsupseteq t_1, \dots, x \sqsupseteq t_n\}$ (same left hand sides) and the equation $\{x = x \sqcup t_1 \sqcup \dots \sqcup t_n\}$ have the same solutions.
 - The least solution to the equation $\{x = x \sqcup t_1 \sqcup \dots \sqcup t_n\}$ is also the least solution of $\{x = t_1 \sqcup \dots \sqcup t_n\}$ (where the x component has been removed on the right hand side).

Example While program

Reaching Definitions Analysis of

```
if  $[b_1]^1$  then (while  $[b_2]^2$  do  $[x := a_1]^3$ )  
    else (while  $[b_3]^4$  do  $[x := a_2]^5$ );  
 $[x := a_3]^6$ 
```

gives equations of the form

$$RD_{entry}(1) = X_?$$

$$RD_{entry}(2) = RD_{exit}(1) \cup RD_{exit}(3)$$

$$RD_{entry}(3) = RD_{exit}(2)$$

$$RD_{entry}(4) = RD_{exit}(1) \cup RD_{exit}(5)$$

$$RD_{entry}(5) = RD_{exit}(4)$$

$$RD_{entry}(6) = RD_{exit}(2) \cup RD_{exit}(4)$$

$$RD_{exit}(1) = RD_{entry}(1)$$

$$RD_{exit}(2) = RD_{entry}(2)$$

$$RD_{exit}(3) = (RD_{entry}(3) \setminus X_{356?}) \cup X_3$$

$$RD_{exit}(4) = RD_{entry}(4)$$

$$RD_{exit}(5) = (RD_{entry}(5) \setminus X_{356?}) \cup X_5$$

$$RD_{exit}(6) = (RD_{entry}(6) \setminus X_{356?}) \cup X_6$$

where e.g. $X_{356?}$ denotes the definitions of x at labels 3, 5, 6 and ?

Example (cont.)

Focussing on RD_{entry} and expressed as equations using the flow variables $\{x_1, \dots, x_6\}$:

$$\begin{array}{ll} x_1 = X_? & x_4 = x_1 \cup (x_5 \setminus X_{356?}) \cup X_5 \\ x_2 = x_1 \cup (x_3 \setminus X_{356?}) \cup X_3 & x_5 = x_4 \\ x_3 = x_2 & x_6 = x_2 \cup x_4 \end{array}$$

Alternatively we can use inequations:

$$\begin{array}{llll} x_1 \supseteq X_? & x_2 \supseteq X_3 & x_4 \supseteq x_1 & x_5 \supseteq x_4 \\ x_2 \supseteq x_1 & x_3 \supseteq x_2 & x_4 \supseteq x_5 \setminus X_{356?} & x_6 \supseteq x_2 \\ x_2 \supseteq x_3 \setminus X_{356?} & & x_4 \supseteq X_5 & x_6 \supseteq x_4 \end{array}$$

Assumptions

- There is a finite **constraint system** \mathcal{S} of the form $(x_i \sqsupseteq t_i)_{i=1}^N$ for $N \geq 1$ where the left hand sides x_i are not necessarily distinct; the form of the terms t_i of the right hand sides is left unspecified.
- The set $FV(t_i)$ of flow variables occurring in t_i is a subset of the finite set $X = \{x_i \mid 1 \leq i \leq N\}$.
- A solution is a total function, $\psi : X \rightarrow L$, assigning to each flow variable a value in the complete lattice (L, \sqsubseteq) satisfying the Ascending Chain Condition.
- The terms are interpreted with respect to solutions, $\psi : X \rightarrow L$, and we write $\llbracket t \rrbracket \psi \in L$ to represent the value of t relative to ψ .
- The interpretation $\llbracket t \rrbracket \psi$ of a term t is **monotone** in ψ and its value only depends on the values of the flow variables occurring in t .

Abstract Worklist Algorithm

INPUT: A system \mathcal{S} of constraints: $x_1 \sqsupseteq t_1, \dots, x_N \sqsupseteq t_N$

OUTPUT: The least solution: Analysis

DATA STRUCTURES:

- W : worklist of constraints
- A : array indexed by flow variables containing elements of the lattice L (the current value of the flow variable)
- $Infl$: array indexed by flow variables containing the set of constraints **influenced** by the flow variable

Worklist Algorithm: initialisation

$W := \text{empty};$

for all $x \sqsupseteq t$ in \mathcal{S} do

$W := \text{insert}((x \sqsupseteq t), W);$

initially all constraints in the worklist

$\text{Analysis}[x] := \perp;$

initialised to the least element of L

$\text{infl}[x] := \emptyset;$

for all $x \sqsupseteq t$ in \mathcal{S} do

for all x' in $FV(t)$ do

$\text{infl}[x'] := \text{infl}[x'] \cup \{x \sqsupseteq t\};$

changes to x' might influence x

via the constraint $x \sqsupseteq t$

OBS: After the initialisation we have $\text{infl}[x'] = \{(x \sqsupseteq t) \text{ in } \mathcal{S} \mid x' \in FV(t)\}$

Worklist Algorithm: iteration

while $W \neq \text{empty}$ do

$((x \sqsupseteq t), W) := \text{extract}(W);$

consider the next constraint

$\text{new} := \text{eval}(t, \text{Analysis});$

if $\text{Analysis}[x] \not\sqsupseteq \text{new}$ then

any work to do?

$\text{Analysis}[x] := \text{Analysis}[x] \sqcup \text{new};$

update the analysis information

for all $x' \sqsupseteq t'$ in $\text{infl}[x]$ do

$W := \text{insert}((x' \sqsupseteq t'), W);$

update the worklist

Operations on worklists

- **empty** is the empty worklist;
- **insert** $((x \sqsupseteq t), W)$ returns a new worklist that is as W except that a new constraint $x \sqsupseteq t$ has been added; it is normally used as in

$$W := \text{insert}((x \sqsupseteq t), W)$$

so as to update the worklist W to contain the new constraint $x \sqsupseteq t$;

- **extract** (W) returns a pair whose first component is a constraint $x \sqsupseteq t$ in the worklist and whose second component is the smaller worklist obtained by removing an occurrence of $x \sqsupseteq t$; it is used as in

$$((x \sqsupseteq t), W) := \text{extract}(W)$$

so as to select and remove a constraint from W .

Organising the worklist

In its most abstract form the worklist could be viewed as a **set of constraints** with the following operations:

empty = \emptyset

function **insert**(($x \sqsupseteq t$), **W**)

return **W** \cup { $x \sqsupseteq t$ }

function **extract**(**W**)

return (($x \sqsupseteq t$), **W** \setminus { $x \sqsupseteq t$ }) for some $x \sqsupseteq t$ in **W**

Extraction based on LIFO

The worklist is represented as a **list of constraints** with the following operations:

empty = nil

function **insert**(($x \sqsupseteq t$), **W**)

return cons(($x \sqsupseteq t$), **W**)

function **extract**(**W**)

return (head(**W**), tail(**W**))

Extraction based on FIFO

The worklist is represented as a list of constraints:

```
empty = nil
```

```
function insert( $(x \sqsupseteq t)$ , W)
```

```
return append(W,  $[x \sqsupseteq t]$ )
```

```
function extract(W)
```

```
return (head(W), tail(W))
```

Example: initialisation

Equations:

$$x_1 = X_?$$

$$x_2 = x_1 \cup (x_3 \setminus X_{356?}) \cup X_3$$

$$x_3 = x_2$$

$$x_4 = x_1 \cup (x_5 \setminus X_{356?}) \cup X_5$$

$$x_5 = x_4$$

$$x_6 = x_2 \cup x_4$$

Initialised data structures:

	x_1	x_2	x_3	x_4	x_5	x_6
infl	$\{x_2, x_4\}$	$\{x_3, x_6\}$	$\{x_2\}$	$\{x_5, x_6\}$	$\{x_4\}$	\emptyset
A	\emptyset	\emptyset	\emptyset	\emptyset	\emptyset	\emptyset
W	$[x_1, x_2, x_3, x_4, x_5, x_6]$					

OBS: in this example the left hand sides of the equations uniquely identify the equations

Example: iteration

W	x ₁	x ₂	x ₃	x ₄	x ₅	x ₆
[x ₁ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	∅	∅	∅	∅	∅	∅
[x ₂ , x ₄ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	X?	—	—	—	—	—
[x ₃ , x ₆ , x ₄ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	—	X ₃ ?	—	—	—	—
[x ₂ , x ₆ , x ₄ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	—	—	X ₃ ?	—	—	—
[x ₆ , x ₄ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	—	—	—	—	—	—
[x ₄ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	—	—	—	—	—	X ₃ ?
[x ₅ , x ₆ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	—	—	—	X ₅ ?	—	—
[x ₄ , x ₆ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	—	—	—	—	X ₅ ?	—
[x ₆ , x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	—	—	—	—	—	—
[x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	—	—	—	—	—	X ₃₅ ?
[x ₃ , x ₄ , x ₅ , x ₆]	—	—	—	—	—	—
[x ₄ , x ₅ , x ₆]	—	—	—	—	—	—
[x ₅ , x ₆]	—	—	—	—	—	—
[x ₆]	—	—	—	—	—	—
[]	—	—	—	—	—	—

Correctness of the algorithm

Given a system of constraints, $\mathcal{S} = (x_i \sqsupseteq t_i)_{i=1}^N$, we define

$$F_{\mathcal{S}} : (X \rightarrow L) \rightarrow (X \rightarrow L)$$

by:

$$F_{\mathcal{S}}(\psi)(x) = \bigsqcup \{ \llbracket t \rrbracket \psi \mid x \sqsupseteq t \text{ in } \mathcal{S} \}$$

This is a monotone function over a complete lattice $X \rightarrow L$.

It follows from Tarski's Fixed Point Theorem:

If $f : L \rightarrow L$ is a monotone function on a complete lattice (L, \sqsubseteq) then it has a least fixed point $\text{lfp}(f) = \bigsqcap \text{Red}(f) \in \text{Fix}(f)$

that $F_{\mathcal{S}}$ has a least fixed point, $\mu_{\mathcal{S}}$, which is the least solution to the constraints \mathcal{S} .

Tarski's Fixed Point Theorem (again)

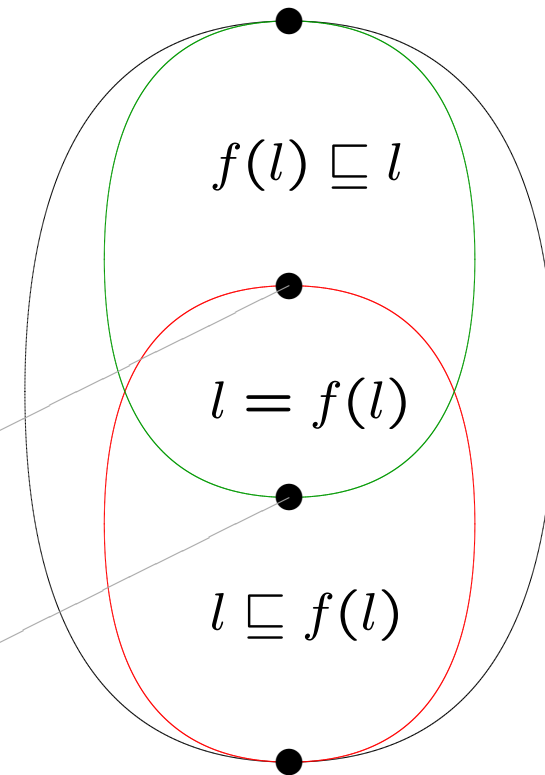
Let $L = (L, \sqsubseteq)$ be a complete lattice and let $f : L \rightarrow L$ be a monotone function.

The **greatest fixed point** $gfp(f)$ satisfy:

$$gfp(f) = \sqcup \{l \mid l \sqsubseteq f(l)\} \in \{l \mid f(l) = l\}$$

The **least fixed point** $lfp(f)$ satisfy:

$$lfp(f) = \sqcap \{l \mid f(l) \sqsubseteq l\} \in \{l \mid f(l) = l\}$$



Correctness of the algorithm (2)

Since L satisfies the Ascending Chain Condition and since X is finite it follows that also $X \rightarrow L$ satisfies the Ascending Chain Condition; therefore μ_S is given by

$$\mu_S = \text{lfp}(F_S) = \bigsqcup_{j \geq 0} F_S^j(\perp)$$

and the chain $(F_S^n(\perp))_n$ eventually stabilises.

Lemma

Given the assumptions, the abstract worklist algorithm computes the least solution of the given constraint system, \mathcal{S} .

Proof

- termination – of initialisation and iteration loop
- correctness is established in three steps:
 - $A \sqsubseteq \mu_{\mathcal{S}}$ – holds initially and is preserved by the loop
 - $F_{\mathcal{S}}(A) \sqsubseteq A$ – proved by contradiction
 - $\mu_{\mathcal{S}} \sqsubseteq A$ – follows from Tarski's fixed point theorem
- complexity: $O(h \cdot M^2 \cdot N)$ for h being the height of L , M being the maximal size of the right hand sides of the constraints and N being the number of constraints

Worklist & Reverse Postorder

- Changes should be propagated throughout the rest of the program before returning to re-evaluate a constraint.
- To ensure that every other constraint is evaluated before re-evaluating the constraint which caused the change is to impose some **total order** on the constraints.
- We shall impose a **graph structure** on the constraints and then use an iteration order based on **reverse postorder**.

Graph structure of constraint system

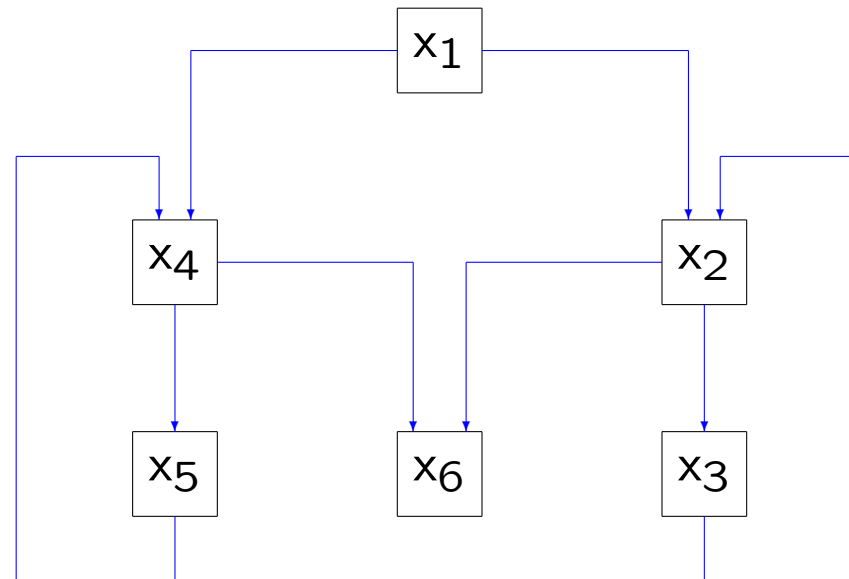
Given a constraint system $\mathcal{S} = (x_i \sqsupseteq t_i)_{i=1}^N$ we can construct a **graphical representation** $G_{\mathcal{S}}$ of the dependencies between the constraints in the following way:

- there is a node for each constraint $x_i \sqsupseteq t_i$, and
- there is a directed edge from the node for $x_i \sqsupseteq t_i$ to the node for $x_j \sqsupseteq t_j$ if x_i appears in t_j (i.e. if $x_j \sqsupseteq t_j$ appears in $\text{infl}[x_i]$).

This constructs a **directed graph**.

Example: graph representation

$$\begin{aligned}x_1 &= X_? \\x_2 &= x_1 \cup (x_3 \setminus X_{356?}) \cup X_3 \\x_3 &= x_2 \\x_4 &= x_1 \cup (x_5 \setminus X_{356?}) \cup X_5 \\x_5 &= x_4 \\x_6 &= x_2 \cup x_4\end{aligned}$$



Handles and roots

Observations:

- A constraint systems corresponding to **forward analyses of While programs** will have a root
- A constraint systems corresponding to **backward analyses for While programs** will not have a single root

A **handle** is a set of nodes such that each node in the graph is reachable through a directed path starting from one of the nodes in the handle.

- A graph G has a **root** r if and only if G has $\{r\}$ as a handle
- Minimal handles always exist (but they need not be unique)

Depth-First Spanning Forest

We can then construct a **depth-first spanning forest** (abbreviated DFSF) from the graph G_S and handle H_S :

INPUT: A directed graph (N, A) with k nodes and handle H

OUTPUT: (1) A DFSF $T = (N, A_T)$, and

(2) a numbering **rPostorder** of the nodes indicating the reverse order in which each node was last visited and represented as an element of array $[N]$ of int

Algorithm for DFSF

METHOD: $i := k$;

mark all nodes of N as unvisited;

let A_T be empty;

while unvisited nodes in H exists do

 choose a node h in H ; DFS(h);

USING: procedure DFS(n) is

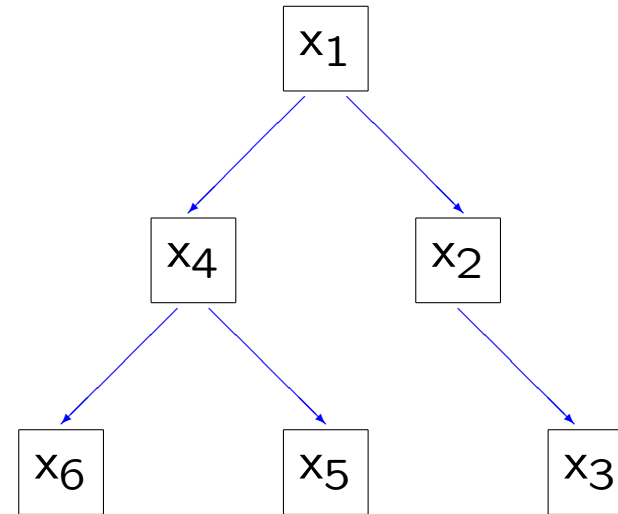
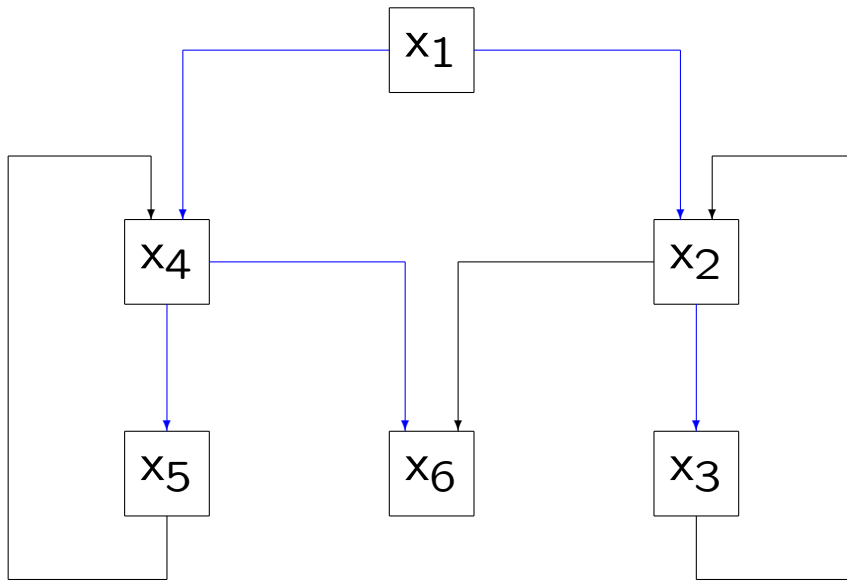
 mark n as visited;

 while $(n, n') \in A$ and n' has not been visited do

 add the edge (n, n') to A_T ; DFS(n');

 rPostorder[n] := i ; $i := i - 1$;

Example: DFST



reverse postorder: $x_1, x_2, x_3, x_4, x_5, x_6$
pre-order: $x_1, x_4, x_6, x_5, x_2, x_3$
breadth-first order: $x_1, x_4, x_2, x_6, x_5, x_3$

Categorisation of edges

Given a spanning forest one can categorise the edges in the original graph as follows:

- **Tree edges**: edges present in the spanning forest.
- **Forward edges**: edges that are not tree edges and that go from a node to a proper descendant in the tree.
- **Back edges**: edges that go from descendants to ancestors (including self-loops).
- **Cross edges**: edges that go between nodes that are unrelated by the ancestor and descendant relations.

Properties of Reverse Postorder

Let $G = (N, A)$ be a directed graph, T a depth-first spanning forest of G and rPostorder the associated ordering computed by the algorithm.

- $(n, n') \in A$ is a **back edge** if and only if $\text{rPostorder}[n] \geq \text{rPostorder}[n']$.
- $(n, n') \in A$ is a **self-loop** if and only if $\text{rPostorder}[n] = \text{rPostorder}[n']$.
- Any cycle of G contains at least one back edge.
- **Reverse postorder** (rPostorder) topologically sorts tree edges as well as the forward and cross edges.
- **Preorder** and **breadth-first order** also sorts tree edges and forward edges but not necessarily cross edges.

Extraction based on Reverse Postorder

Idea: The iteration amounts to an **outer** iteration that contains an **inner** iteration that visits the nodes in reverse postorder:

We organise the worklist **W** as a pair **(W.c, W.p)** of two structures:

- **W.c** is a list of **current** nodes to be visited in the current inner iteration.
- **W.p** is a set of **pending** nodes to be visited in a later inner iteration.

Nodes are always **inserted into W.p** and always **extracted from W.c**.

When **W.c** is exhausted the current inner iteration has finished and in preparation for the next inner iteration we must sort **W.p** in the reverse postorder given by **rPostorder** and assign the result to **W.c**.

Iterating in Reverse Postorder

`empty = (nil, \emptyset)`

function `insert`(($x \sqsupseteq t$), (`W.c`, `W.p`))

return (`W.c`, (`W.p` \cup { $x \sqsupseteq t$ }))

insert into pending set

function `extract`((`W.c`, `W.p`))

if `W.c` = nil then

no more constraints in current list

`W.c` := `sort_rPostorder`(`W.p`);

sort pending set and update

`W.p` := \emptyset

current list and pending set

return (`head`(`W.c`), (`tail`(`W.c`), `W.p`))

extract from current round

Example: Reverse Postorder iteration

W.c	W.p	x ₁	x ₂	x ₃	x ₄	x ₅	x ₆
[]	{x ₁ , ..., x ₆ }	∅	∅	∅	∅	∅	∅
[x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	{x ₂ , x ₄ }	X ₁ ?	—	—	—	—	—
[x ₃ , x ₄ , x ₅ , x ₆]	{x ₂ , x ₃ , x ₄ , x ₆ }	—	X ₃ ?	—	—	—	—
[x ₄ , x ₅ , x ₆]	{x ₂ , x ₃ , x ₄ , x ₆ }	—	—	X ₃ ?	—	—	—
[x ₅ , x ₆]	{x ₂ , ..., x ₆ }	—	—	—	X ₅ ?	—	—
[x ₆]	{x ₂ , ..., x ₆ }	—	—	—	—	X ₅ ?	—
[x ₂ , x ₃ , x ₄ , x ₅ , x ₆]	∅	—	—	—	—	—	X ₃₅ ?
[x ₃ , x ₄ , x ₅ , x ₆]	∅	—	—	—	—	—	—
[x ₄ , x ₅ , x ₆]	∅	—	—	—	—	—	—
[x ₅ , x ₆]	∅	—	—	—	—	—	—
[x ₆]	∅	—	—	—	—	—	—
[]	∅	—	—	—	—	—	—

$$x_1 = X_1?$$

$$x_3 = x_2$$

$$x_5 = x_4$$

$$x_2 = x_1 \cup (x_3 \setminus X_{356}?) \cup X_3$$

$$x_4 = x_1 \cup (x_5 \setminus X_{356}?) \cup X_5$$

$$x_6 = x_2 \cup x_4$$

Complexity

- A list of N elements can be sorted in $O(N \cdot \log_2(N))$ steps.
- If we use a linked list representation of lists then inserting an element to the front of a list and extracting the head of a list can be done in constant time.
- The overall complexity for processing N insertions and N extractions is $O(N \cdot \log_2(N))$.

The Round Robin Algorithm

Assumption: the constraints are sorted in reverse postorder.

- each time **W.c** is exhausted we assign it the list $[1, \dots, N]$
- **W.p** is replaced by a boolean, **change**, that is false whenever **W.p** is empty
- the iterations are split into an outer iteration with an explicit inner iteration; each inner iteration is a simple iteration through all constraints in reverse postorder.

Round Robin Iteration

`empty = (nil, false)`

function `insert`(($x \sqsupseteq t$), (`W.c`, `change`))

return (`W.c`, `true`)

pending constraints

function `extract`((`W.c`, `change`))

if `W.c = nil` then

`W.c := [1, ..., N];`

`change := false`

a new round is needed

all constraints are re-considered

no pending constraints

return (`head(W.c)`, (`tail(W.c)`, `change`))

The Round Robin Algorithm

INPUT: A system \mathcal{S} of constraints: $x_1 \sqsupseteq t_1, \dots, x_N \sqsupseteq t_N$
ordered 1 to N in reverse postorder

OUTPUT: The least solution: **Analysis**

METHOD: Initialisation

for all $x \in X$ do **Analysis** $[x] := \perp$

change := true;

The Round Robin Algorithm (cont.)

METHOD: Iteration (updating *Analysis*)

while *change* do

change := false;

 for $i := 1$ to N do

 new := eval(t_i , *Analysis*);

 if *Analysis*[x_i] $\not\sqsubseteq$ new then

change := true;

Analysis[x_i] := *Analysis*[x_i] \sqcup new;

Lemma:

The Round Robin algorithm computes the least solution of the given constraint system, \mathcal{S} .

Example: Round Robin iteration

change	x ₁	x ₂	x ₃	x ₄	x ₅	x ₆
true	∅	∅	∅	∅	∅	∅
* false						
true	X _?	—	—	—	—	—
true	—	X _{3?}	—	—	—	—
true	—	—	X _{3?}	—	—	—
true	—	—	—	X _{5?}	—	—
true	—	—	—	—	X _{5?}	—
true	—	—	—	—	—	X _{35?}
* false						
false	—	—	—	—	—	—
false	—	—	—	—	—	—
false	—	—	—	—	—	—
false	—	—	—	—	—	—
false	—	—	—	—	—	—
false	—	—	—	—	—	—

$$\begin{aligned}
 x_1 &= X_? \\
 x_2 &= x_1 \cup (x_3 \setminus X_{356?}) \cup X_3 \\
 x_3 &= x_2 \\
 x_4 &= x_1 \cup (x_5 \setminus X_{356?}) \cup X_5 \\
 x_5 &= x_4 \\
 x_6 &= x_2 \cup x_4
 \end{aligned}$$

Loop connectness parameter

Consider a depth-first spanning forest T and a reverse postorder rPost-order constructed for the graph G with handle H .

The **loop connectedness** parameter $d(G, T)$ is defined as the largest number of **back edges** found on any cycle-free path of G .

For While programs the loop connectedness parameter equals the maximal nesting depth of `while` loops.

Empirical studies of Fortran programs show that the loop connectness parameter seldom exceeds 3.

Complexity

The constraint system $(x_i \sqsupseteq t_i)_{i=1}^N$ is an instance of a **Bit Vector Framework** when $L = \mathcal{P}(D)$ for some finite set D and when each right hand side t_i is of the form

$$(x_{j_i} \cap Y_i^1) \cup Y_i^2$$

for sets $Y_i^k \subseteq D$ and variable $x_{j_i} \in X$.

Lemma:

For Bit Vector Frameworks, the Round Robin Algorithm terminates after at most $d(G, T) + 3$ iterations.

It performs at most $O((d(G, T) + 1) \cdot N)$ assignments.

For While programs: the overall complexity is $O((d+1) \cdot b)$ where d is the maximal nesting depth of while-loops and b is the number of elementary blocks.

Worklist & Strong Components

Two nodes n and n' are said to be **strongly connected** whenever there is a (possibly trivial) directed path from n to n' and a (possibly trivial) directed path from n' to n . Defining

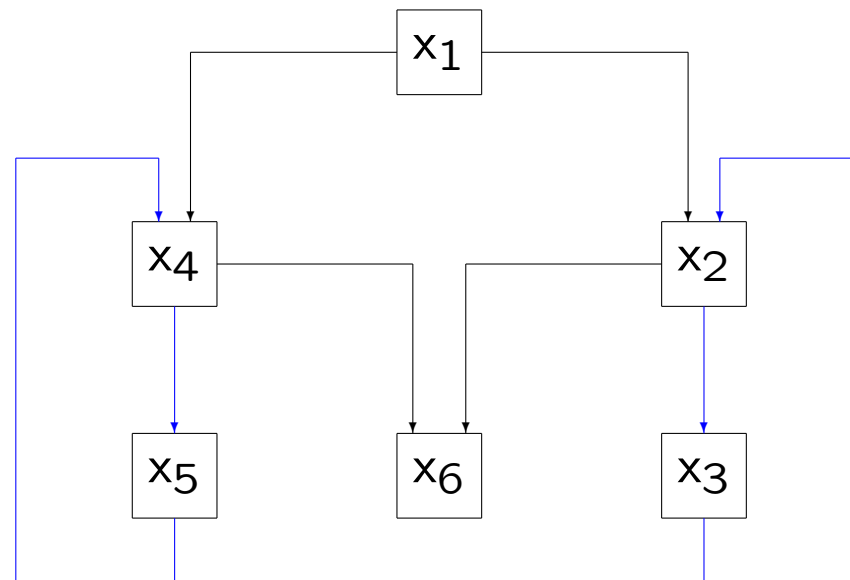
$$\mathcal{SC} = \{(n, n') \mid n \text{ and } n' \text{ are strongly connected}\}$$

we obtain a binary relation $\mathcal{SC} \subseteq N \times N$.

- \mathcal{SC} is an equivalence relation.
- The equivalence classes of \mathcal{SC} are called the **strong components**.

A graph is said to be **strongly connected** whenever it contains exactly one strongly connected component.

Example: Strong Components



Reduced graph

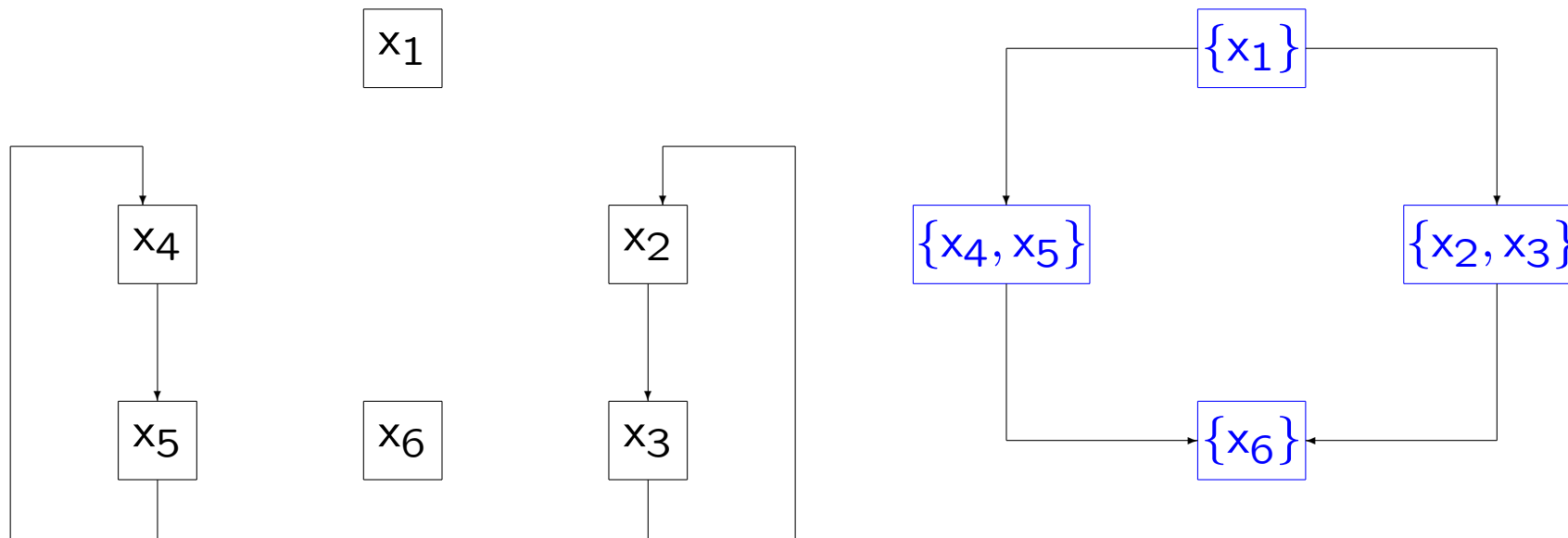
The interconnections between strong components can be represented by the **reduced graph**.

- nodes: the strongly connected components
- edges: there is an **edge** from one node to another distinct node if and only if there is an edge from some node in the first strongly connected component to a node in the second in the original graph.

For any graph G the reduced graph is a **DAG**.

The strong components can be linearly ordered in **topological order**: $SC_1 \leq SC_2$ whenever there is an edge from SC_1 to SC_2 .

Example: Strong Components and reduced graph



The overall idea behind the algorithm

Idea: strong components are visited in topological order with nodes being visited in reverse postorder within each strong component.

The iteration amounts to three levels of iteration:

- the **outermost level** deals with the strong components one by one;
- the **intermediate level** performs a number of passes over the constraints in the current strong component;
- the **inner level** performs one pass in reverse postorder over the appropriate constraints.

To make this work for each constraint we record

- the strong component it occurs in and
- its number in the local reverse postorder for that strong component.

Pseudocode for constraint numbering

INPUT: A graph partitioned into strong components

OUTPUT: `srPostorder`

METHOD: `scc := 1;`
for each `scc` in topological order do
 `rp := 1;`
 for each $x \sqsupseteq t$ in the strong component `scc`
 in local reverse postorder do
 `srPostorder[x \sqsupseteq t] := (scc, rp);`
 `rp := rp + 1`
 `scc := scc + 1;`

Organisation of the worklist

The worklist W as a pair $(W.c, W.p)$ of two structures:

- $W.c$, is a list of **current** nodes to be visited in the current inner iteration.
- $W.p$, is a set of **pending** nodes to be visited in a later intermediate or outer iteration.

Nodes are always inserted into $W.p$ and always extracted from $W.c$.

When $W.c$ is exhausted the current inner iteration has finished and in preparation for the next we must **extract a strong component** from $W.p$, sort it and assign the result to $W.c$.

An inner iteration ends when $W.c$ is exhausted, an intermediate iteration ends when **scc** gets a higher value than last time it was computed, and the outer iteration ends when both $W.c$ and $W.p$ are exhausted.

Iterating through Strong Components

```
empty = (nil,  $\emptyset$ )
```

```
function insert( $(x \sqsupseteq t)$ , ( $W.c, W.p$ ))
```

```
return ( $W.c, (W.p \cup \{x \sqsupseteq t\})$ )
```

```
function extract( $(W.c, W.p)$ )
```

```
local variables: scc, W_scc
```

```
if  $W.c = \text{nil}$  then
```

```
    scc := min{fst(srPostorder[ $x \sqsupseteq t$ ]) | ( $x \sqsupseteq t$ )  $\in W.p$ };
```

```
    W_scc := {( $x \sqsupseteq t$ )  $\in W.p$  | fst(srPostorder[ $x \sqsupseteq t$ ]) = scc};
```

```
    W.c := sort_srPostorder(W_scc);
```

```
     $W.p := W.p \setminus W_scc$ ;
```

```
return ( head( $W.c$ ), (tail( $W.c$ ),  $W.p$ ) )
```

Example: Strong Component iteration

W.c	W.p	x ₁	x ₂	x ₃	x ₄	x ₅	x ₆
[]	{x ₁ , ..., x ₆ }	∅	∅	∅	∅	∅	∅
[]	{x ₂ , ..., x ₆ }	X _?	—	—	—	—	—
[x ₃]	{x ₃ , ..., x ₆ }	—	X _{3?}	—	—	—	—
[]	{x ₂ , ..., x ₆ }	—	—	X _{3?}	—	—	—
[x ₃]	{x ₄ , x ₅ , x ₆ }	—	—	—	—	—	—
[]	{x ₄ , x ₅ , x ₆ }	—	—	—	—	—	—
[x ₅]	{x ₅ , x ₆ }	—	—	—	X _{5?}	—	—
[]	{x ₄ , x ₅ , x ₆ }	—	—	—	—	X _{5?}	—
[x ₅]	{x ₆ }	—	—	—	—	—	—
[]	{x ₆ }	—	—	—	—	—	—
[]	∅	—	—	—	—	—	X _{35?}

$$x_1 = X_?$$

$$x_3 = x_2$$

$$x_5 = x_4$$

$$x_2 = x_1 \cup (x_3 \setminus X_{356?}) \cup X_3$$

$$x_4 = x_1 \cup (x_5 \setminus X_{356?}) \cup X_5$$

$$x_6 = x_2 \cup x_4$$