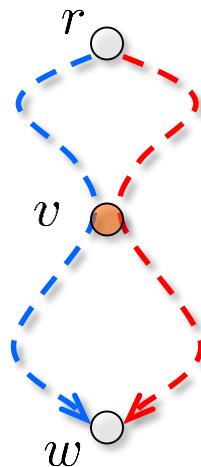


# Dominators in a Flowgraph

Flowgraph:  $G = (V, E, r)$ ; each  $v$  in  $V$  is reachable from  $r$

$v$  dominates  $w$  if every path from  $r$  to  $w$  includes  $v$



Application areas : Program optimization, VLSI testing, theoretical biology, distributed systems, constraint programming, memory profiling, analysis of diffusion networks...

# Dominators in a Flowgraph

---

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Set of dominators:  $\text{Dom}(w) = \{v \mid v \text{ dominates } w\}$

Trivial dominators:  $\forall w \neq r, w, r \in \text{Dom}(w)$

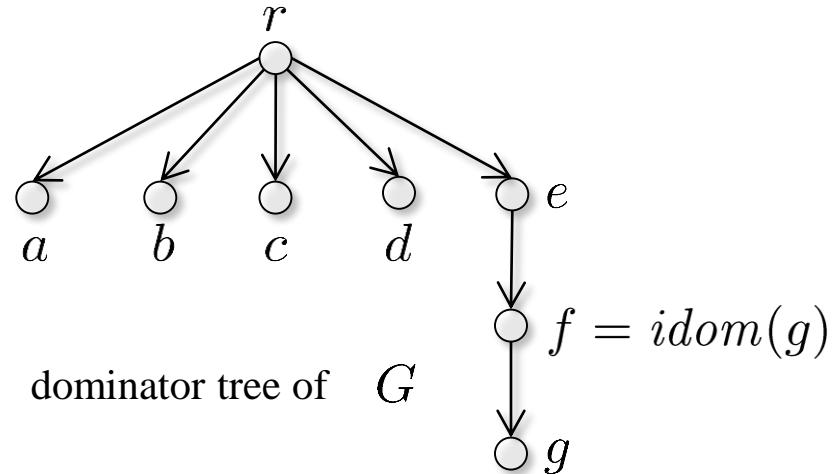
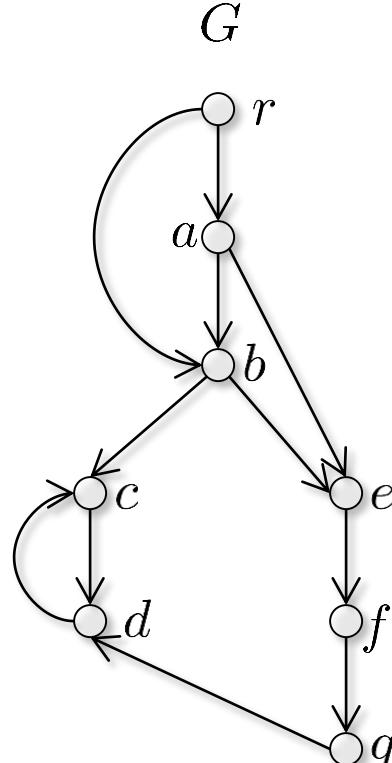
Immediate dominator:  $\text{idom}(w) \in \text{Dom}(w) - w$  and dominated by every  $v$  in  $\text{Dom}(w) - w$

Goal: Find  $\text{idom}(v)$  for each  $v$  in  $V$

# Dominators in a Flowgraph

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$O(m\alpha(m, n))$  algorithm: [Lengauer and Tarjan '79]

$O(m + n)$  algorithms:

[Alstrup, Harel, Lauridsen, and Thorup '97]

[Buchsbaum, Kaplan, Rogers, and Westbrook '04]

[G., and Tarjan '04]

[Buchsbaum et al. '08]

# Application: Loop Optimizations

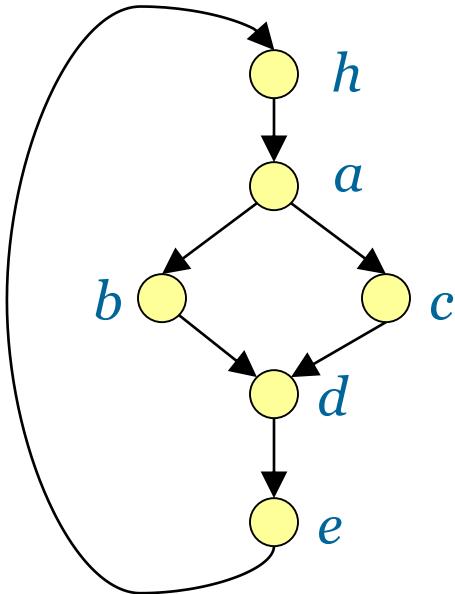
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Loop optimizations typically make a program much more efficient since a large fraction of the total running time is spent on loops.

Dominators can be used to [detect](#) loops.

# Application: Loop Optimizations

Loop  $L$



There is a node  $h \in L$  (loop header), such that

- There is a  $(v,h)$  for some  $v \notin L$
- For any  $w \in L-h$  there is no  $(v,w)$  for  $v \notin L$
- $h$  is reachable from every  $w \in L$
- $h$  reaches every  $w \in L$

Thus  $h$  dominates all nodes in  $L$ .

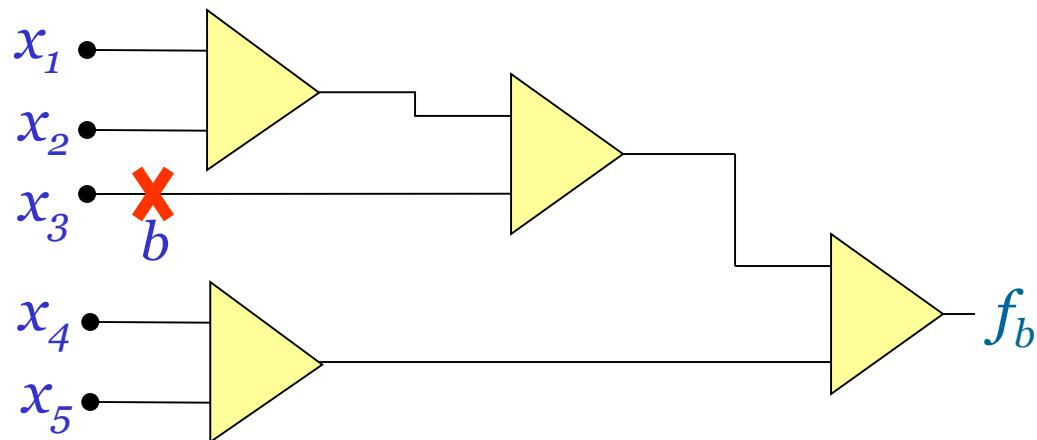
Loop back-edge:  $(v,h) \in A$  and  $h$  dominates  $v$ .

# Application: Identifying Functionally Equivalent Faults

Consider a circuit  $C$ :

Inputs:  $x_1, \dots, x_n$

Output:  $f(x_1, \dots, x_n)$



Suppose there is a fault in wire  $a$ . Then  $C = C_a$  evaluates  $f_a$  instead.

Fault  $a$  and fault  $b$  are functionally equivalent iff

$$f_a(x_1, \dots, x_n) = f_b(x_1, \dots, x_n)$$

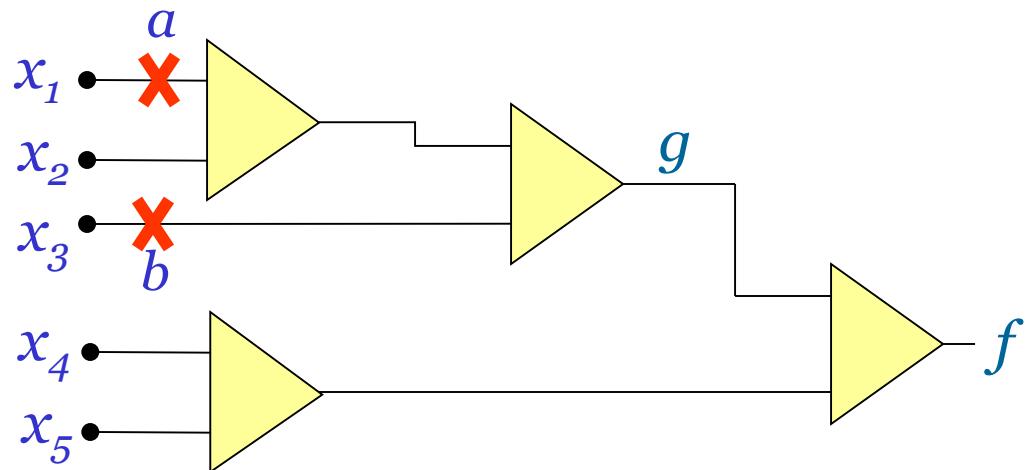
Such pairs of faults are **indistinguishable** and we want to avoid spending time to distinguish them (since it is impossible).

# Application: Identifying Functionally Equivalent Faults

Consider a circuit  $C$ :

Inputs:  $x_1, \dots, x_n$

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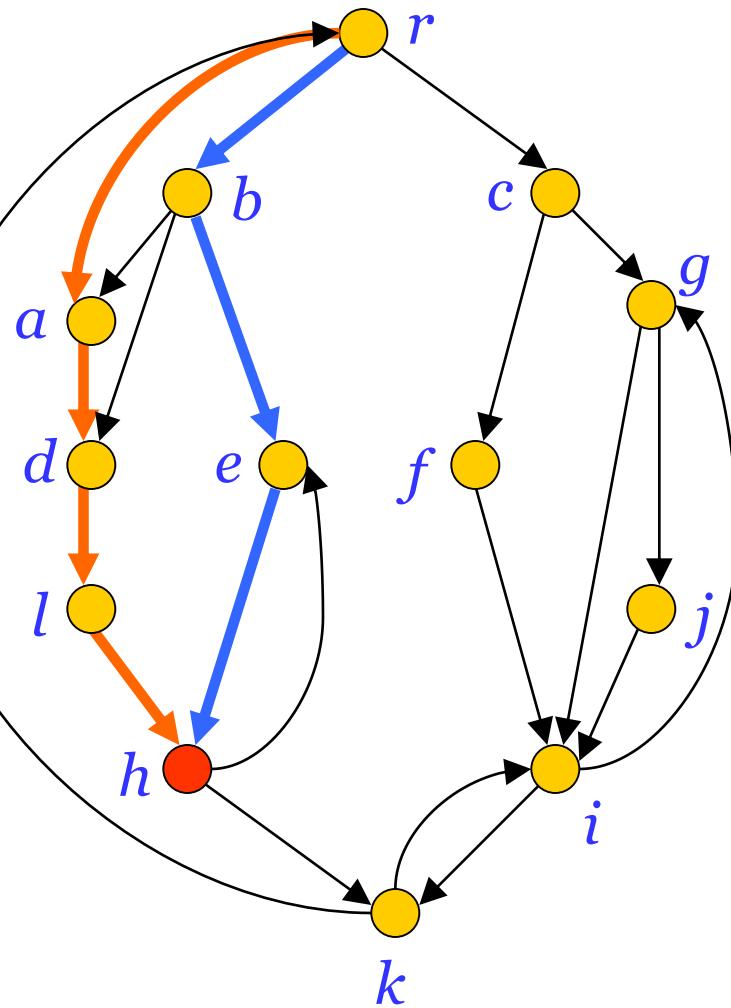
Fault  $a$  and fault  $b$  are functionally equivalent iff

$$f_a(x_1, \dots, x_n) = f_b(x_1, \dots, x_n)$$

It suffices to evaluate the output of a gate  $g$  that dominates  $a$  and  $b$ . This is can be faster than evaluating  $f$  since  $g$  may have fewer inputs.

# Example

---



$$Dom(r) = \{ r \}$$

$$Dom(b) = \{ r, b \}$$

$$Dom(c) = \{ r, c \}$$

$$Dom(a) = \{ r, a \}$$

$$Dom(d) = \{ r, d \}$$

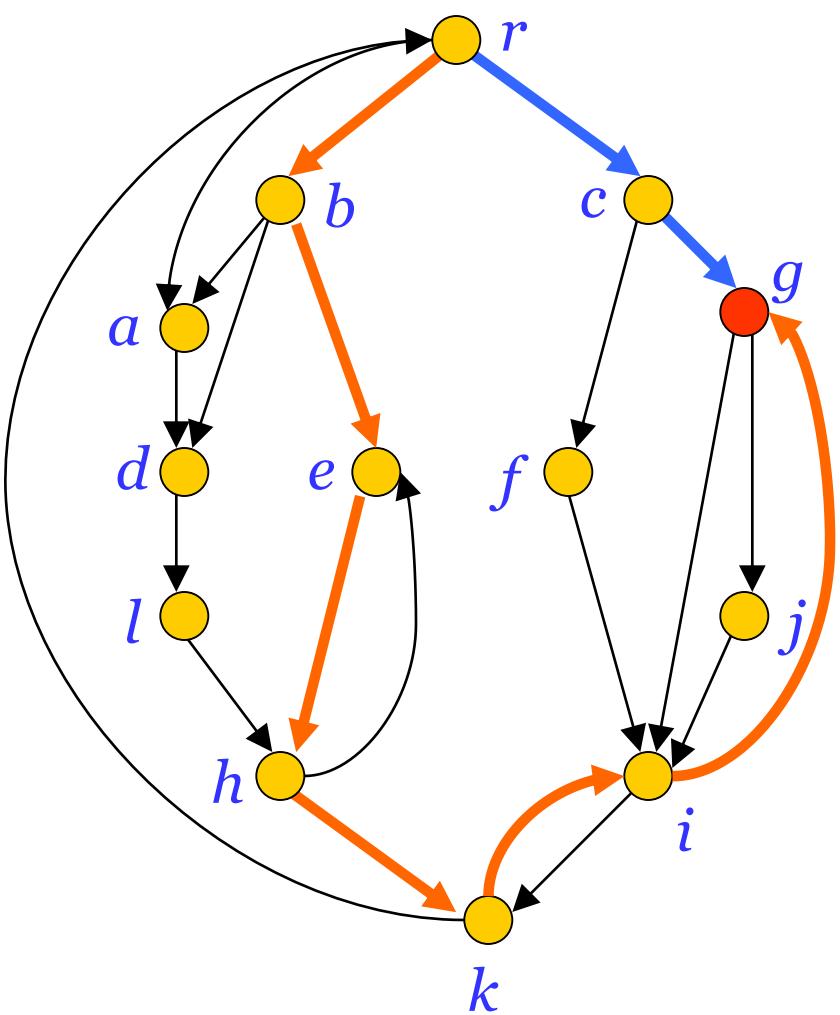
$$Dom(e) = \{ r, e \}$$

$$Dom(l) = \{ r, d, l \}$$

$$Dom(h) = \{ r, h \}$$

# Example

---



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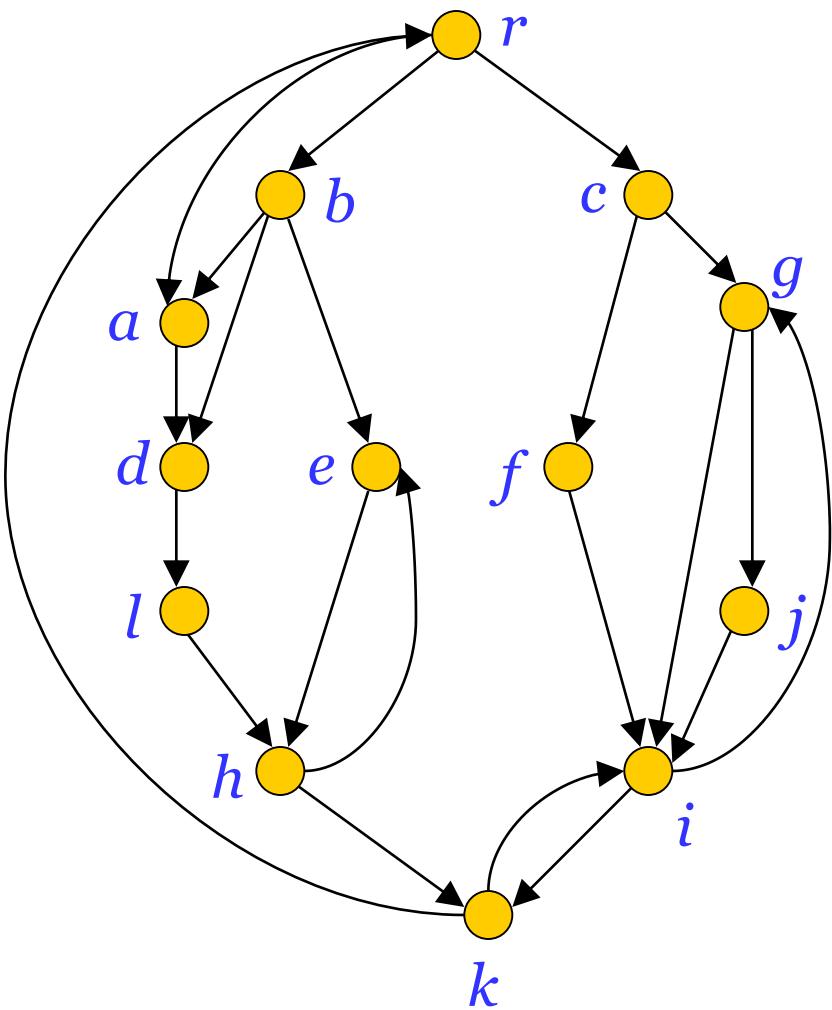
$$Dom(k) = \{ r, k \}$$

$$Dom(f) = \{ r, c, f \}$$

$$Dom(g) = \{ r, g \}$$

# Example

---



$$Dom(b) = \{ r, b \}$$

$$Dom(c) = \{ r, c \}$$

$$Dom(a) = \{ r, a \}$$

$$Dom(d) = \{ r, d \}$$

$$Dom(e) = \{ r, e \}$$

$$Dom(l) = \{ r, d, l \}$$

$$Dom(h) = \{ r, h \}$$

$$Dom(k) = \{ r, k \}$$

$$Dom(f) = \{ r, c, f \}$$

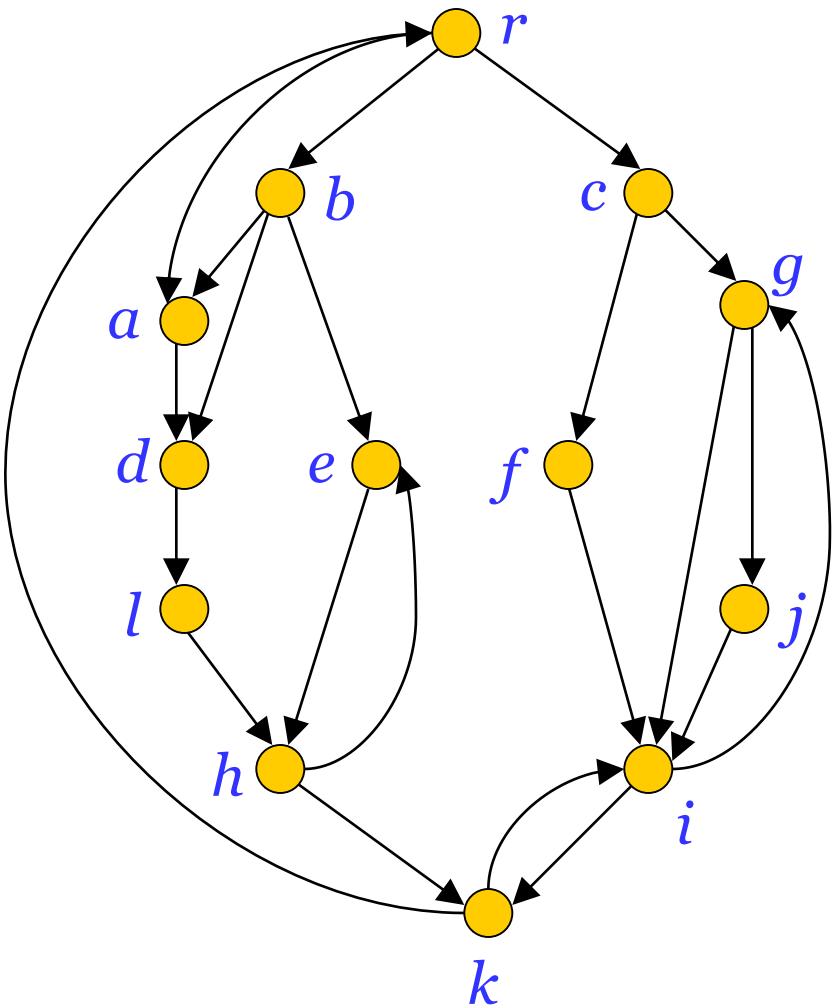
$$Dom(g) = \{ r, g \}$$

$$Dom(j) = \{ r, g, j \}$$

$$Dom(i) = \{ r, i \}$$

# Example

---



$$idom(b) = r$$

$$idom(c) = r$$

$$idom(a) = r$$

$$idom(d) = r$$

$$idom(e) = r$$

$$idom(l) = d$$

$$idom(h) = r$$

$$idom(k) = r$$

$$idom(f) = c$$

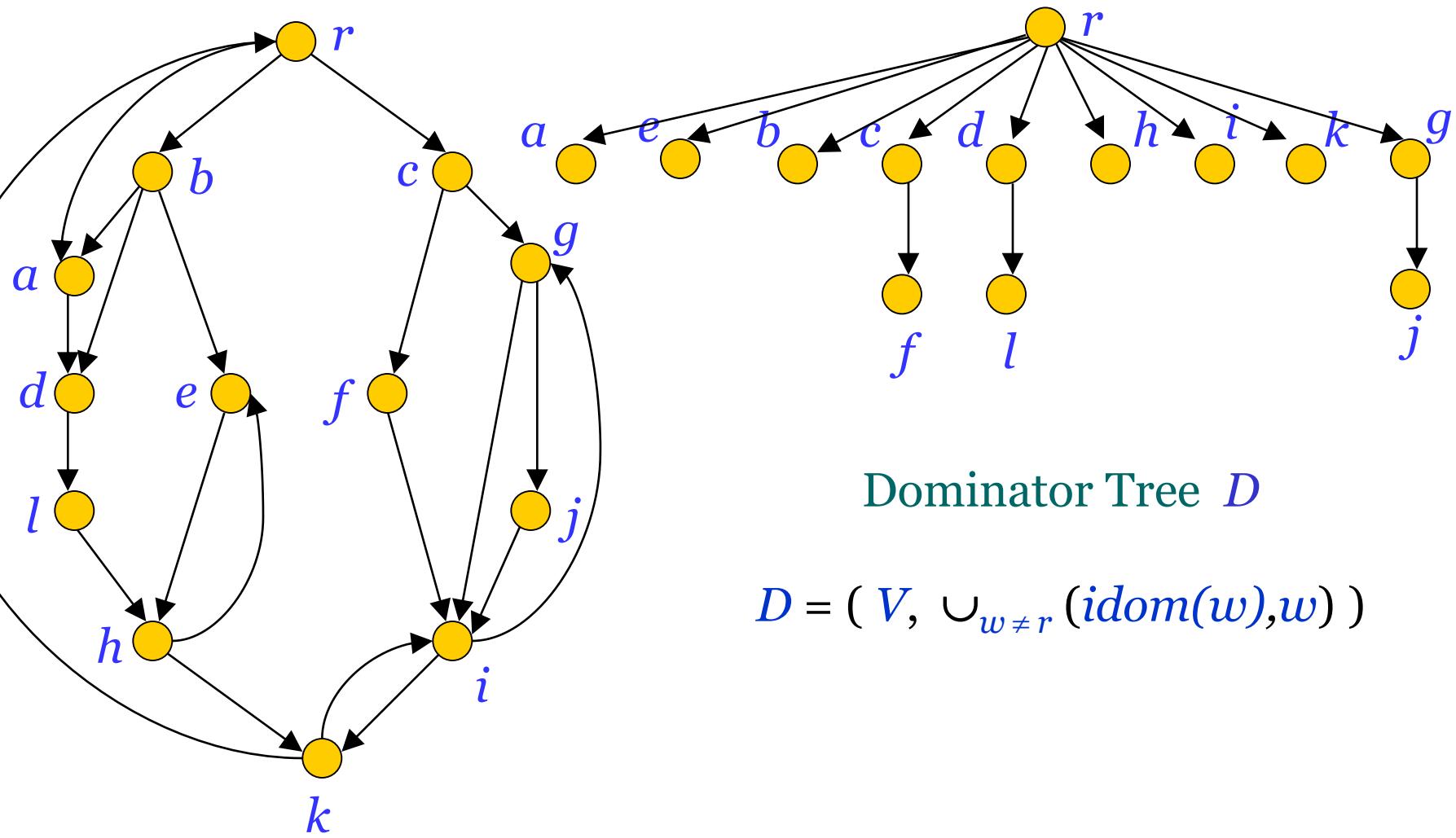
$$idom(g) = r$$

$$idom(j) = g$$

$$idom(i) = r$$

# Example

---



# A Straightforward Algorithm

---

Purdom-Moore [1972]:

```
for all  $v$  in  $V - r$  do
    remove  $v$  from  $G$ 
     $R(v) \leftarrow$  unreachable vertices
    for all  $u$  in  $R(v)$  do
         $Dom(u) \leftarrow Dom(u) \cup \{ v \}$ 
    done
done
```

The running time is  $O(n \cdot m)$ . Also very slow in practice.

# Iterative Algorithm

---

Dominators can be computed by solving **iteratively** the set of equations [Allen and Cocke, 1972]

$$Dom(v) = \left( \cap_{(u,v) \in A} Dom(u) \right) \cup \{v\}, v \neq r$$

Initialization

$$Dom(r) = \{r\}$$

$$Dom(v) = \emptyset, v \neq r$$

In the intersection we consider only the nonempty  $Dom(u)$ .

Each  $Dom(v)$  set can be represented by an  $n$ -bit vector.  
Intersection  $\equiv$  bit-wise AND.

Requires  $n^2$  space. Very slow in practice (but better than PM). 14

# Iterative Algorithm

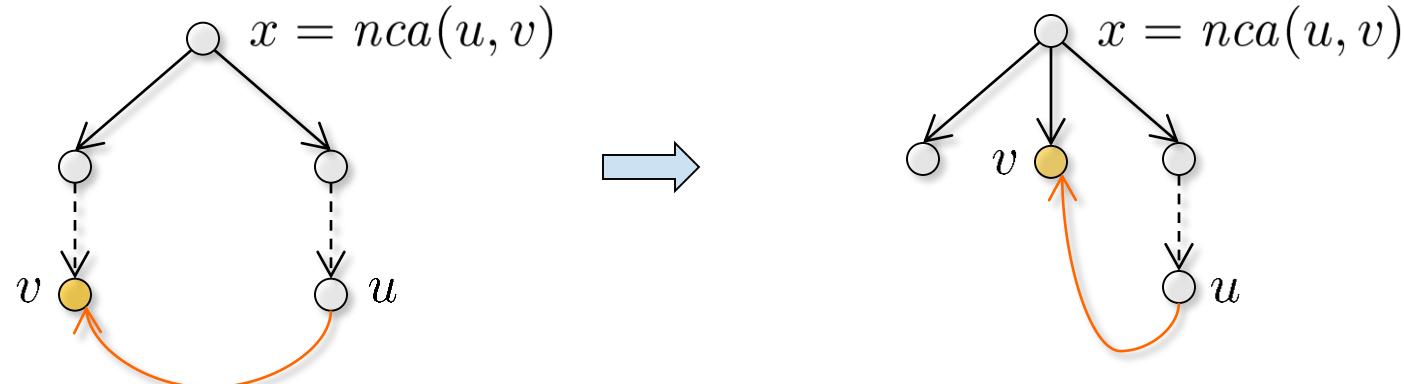
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$$Dom(v) = \left( \cap_{(u,v) \in A} Dom(u) \right) \cup \{v\}, v \neq r$$

Efficient implementation [Cooper, Harvey and Kennedy 2000]:

Maintain tree  $T$ ; process the edges until a fixed-point is reached.

Process  $(u, v)$ : compute  $x = \text{nearest common ancestor of } u \text{ and } v \text{ in } T$ .  
If  $x$  is ancestor of parent of  $v$ , make  $x$  new parent of  $v$ .



# Iterative Algorithm

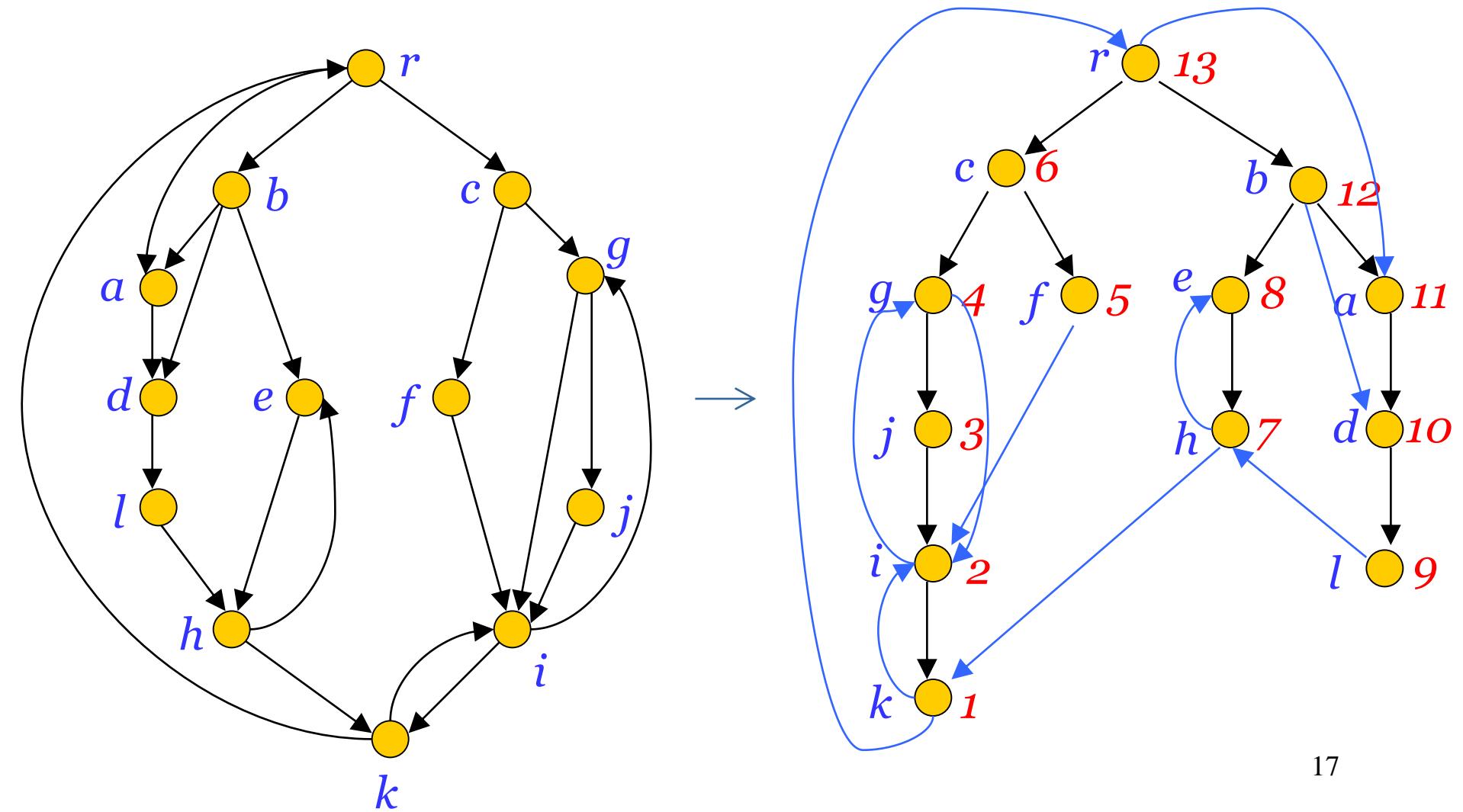
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Efficient implementation [Cooper, Harvey and Kennedy 2000]

```
dfs( $r$ )
 $T \leftarrow \{r\}$ 
 $changed \leftarrow \text{true}$ 
while ( $changed$ ) do
     $changed \leftarrow \text{false}$ 
    for all  $v$  in  $V - r$  in reverse postorder do
         $x \leftarrow \text{nca}(\text{pred}(v))$ 
        if  $x \neq \text{parent}(v)$  then
             $\text{parent}(v) \leftarrow x$ 
             $changed \leftarrow \text{true}$ 
    end
done
done
```

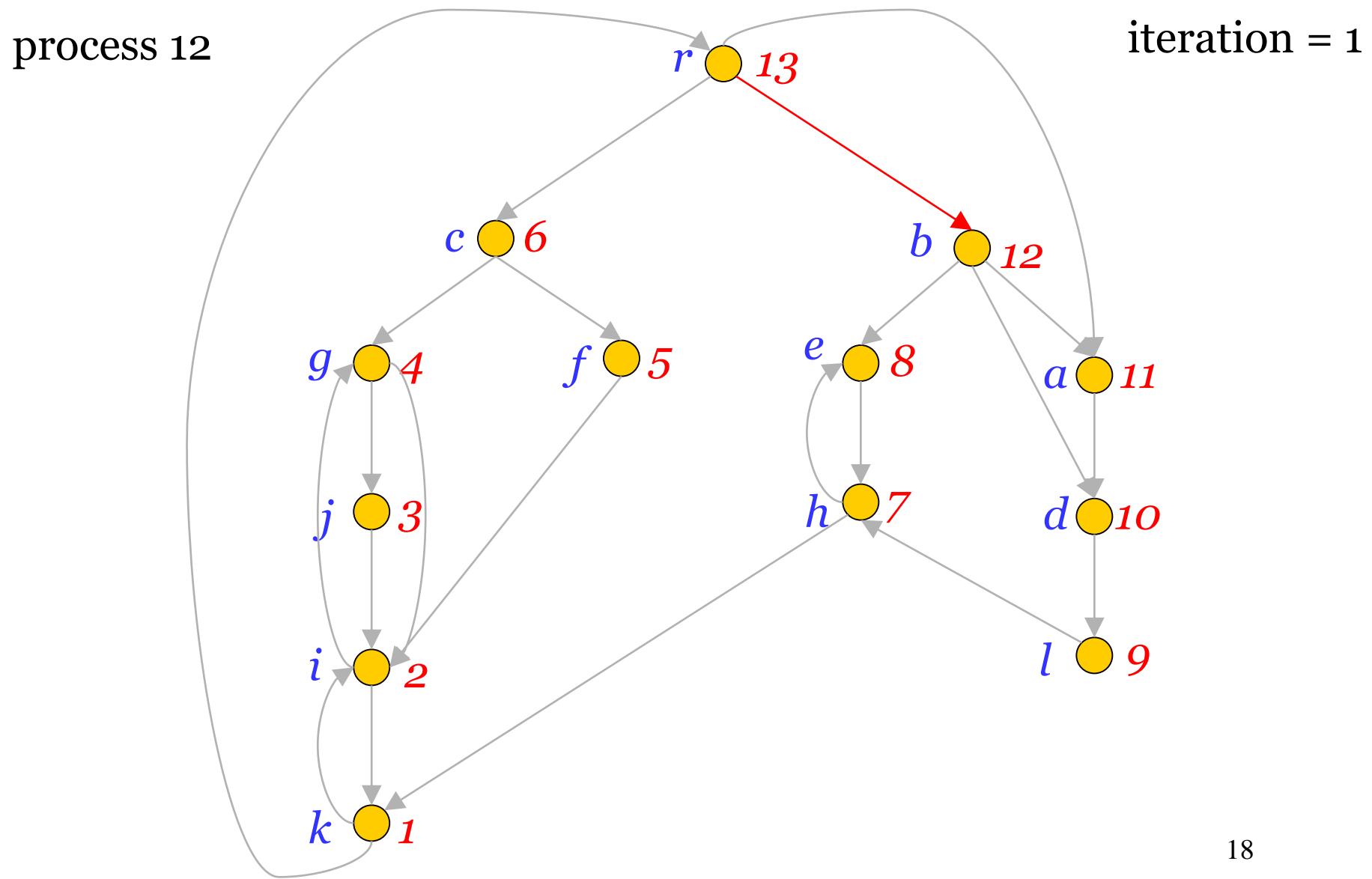
# Iterative Algorithm: Example

Perform a depth-first search on  $G \Rightarrow$  postorder numbers

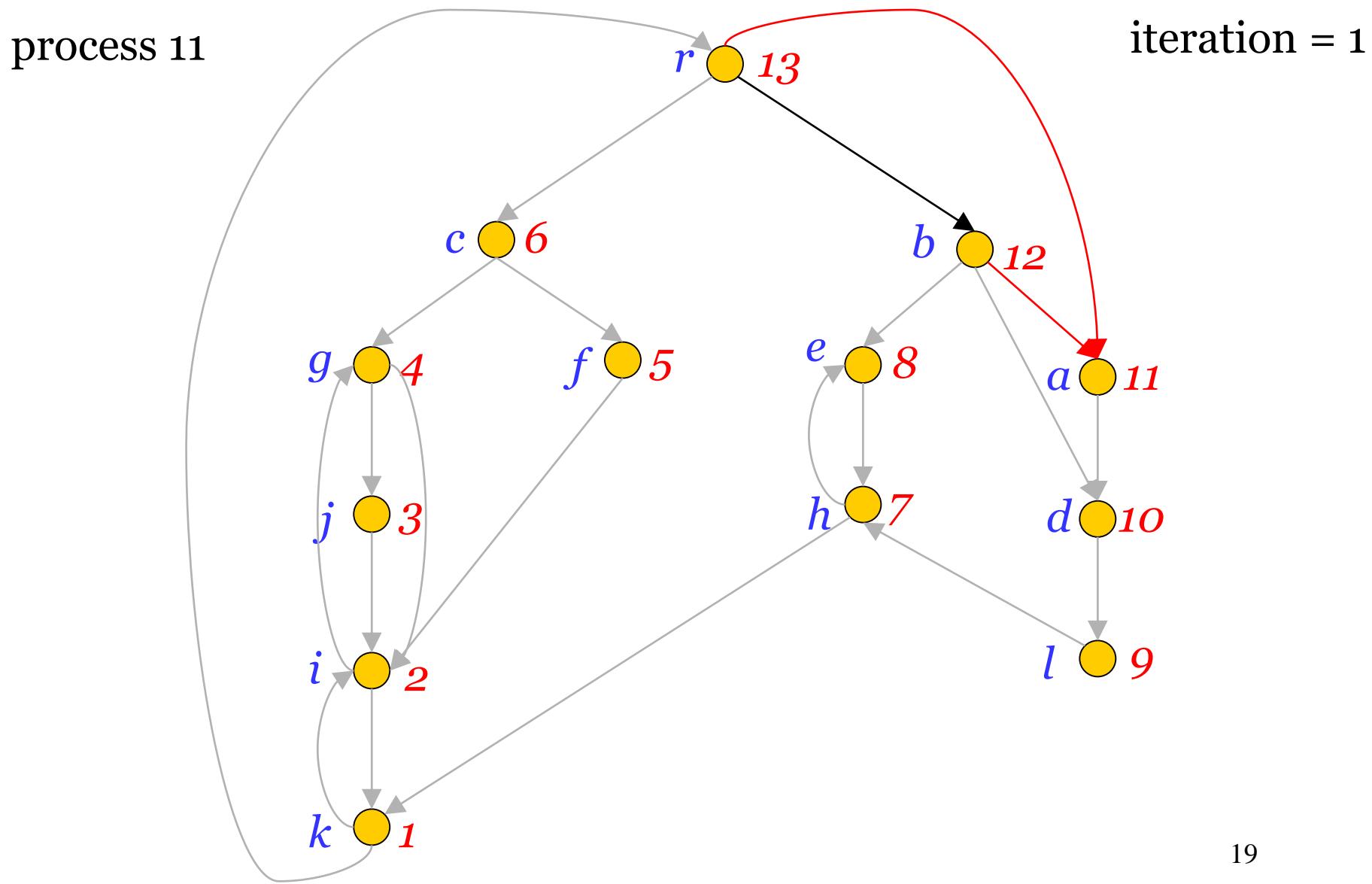


# Iterative Algorithm: Example

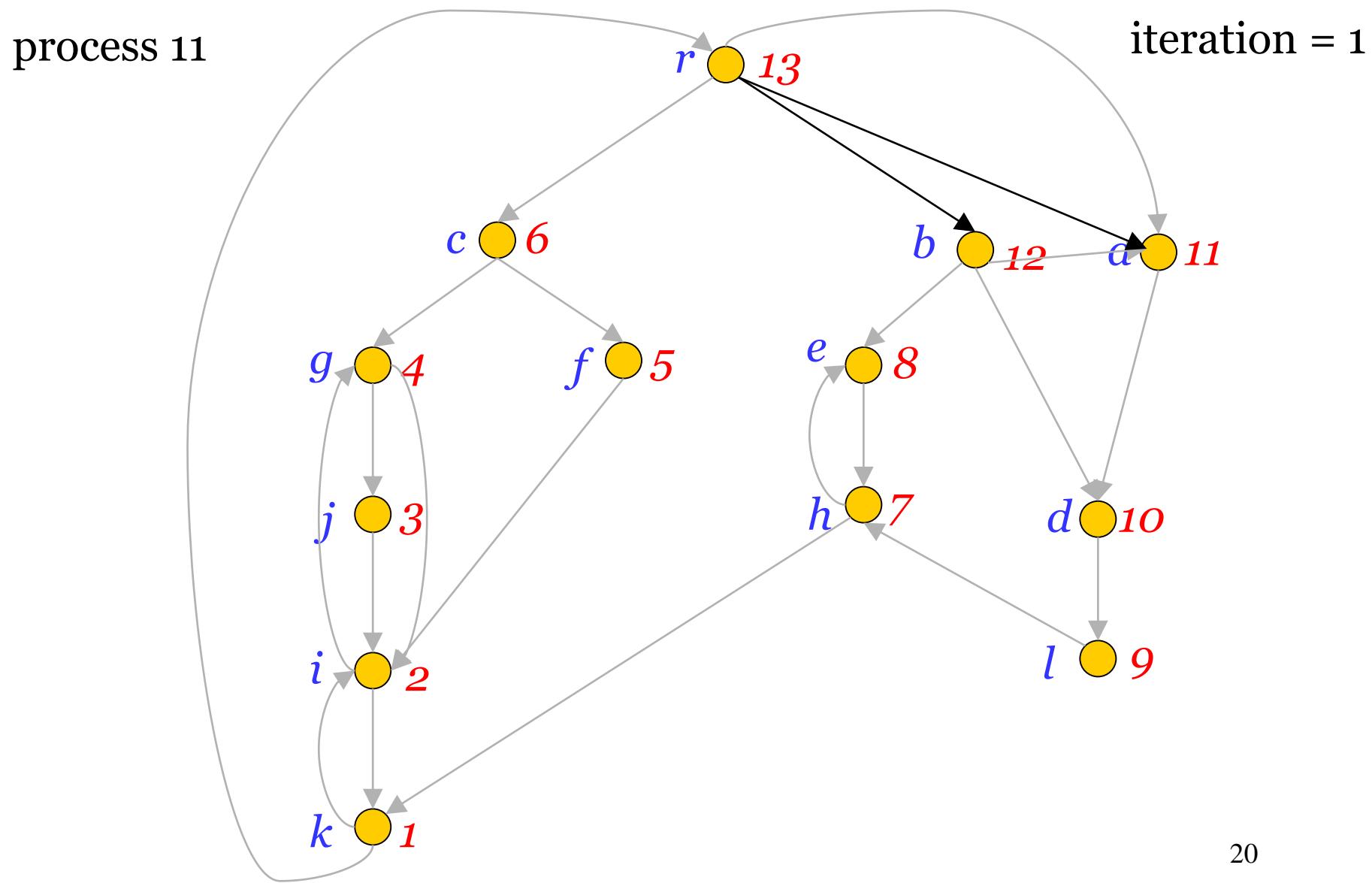
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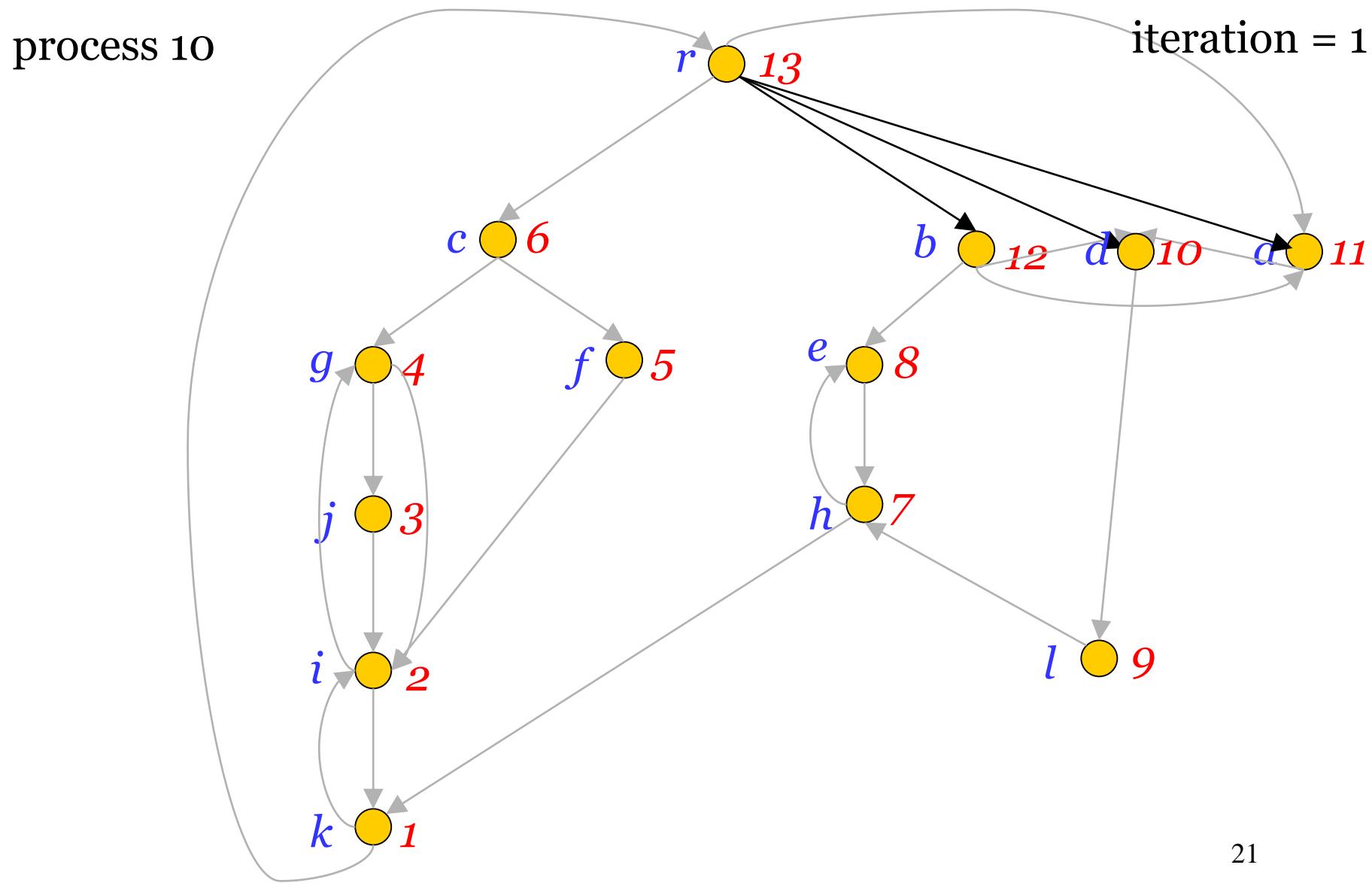
# Iterative Algorithm: Example



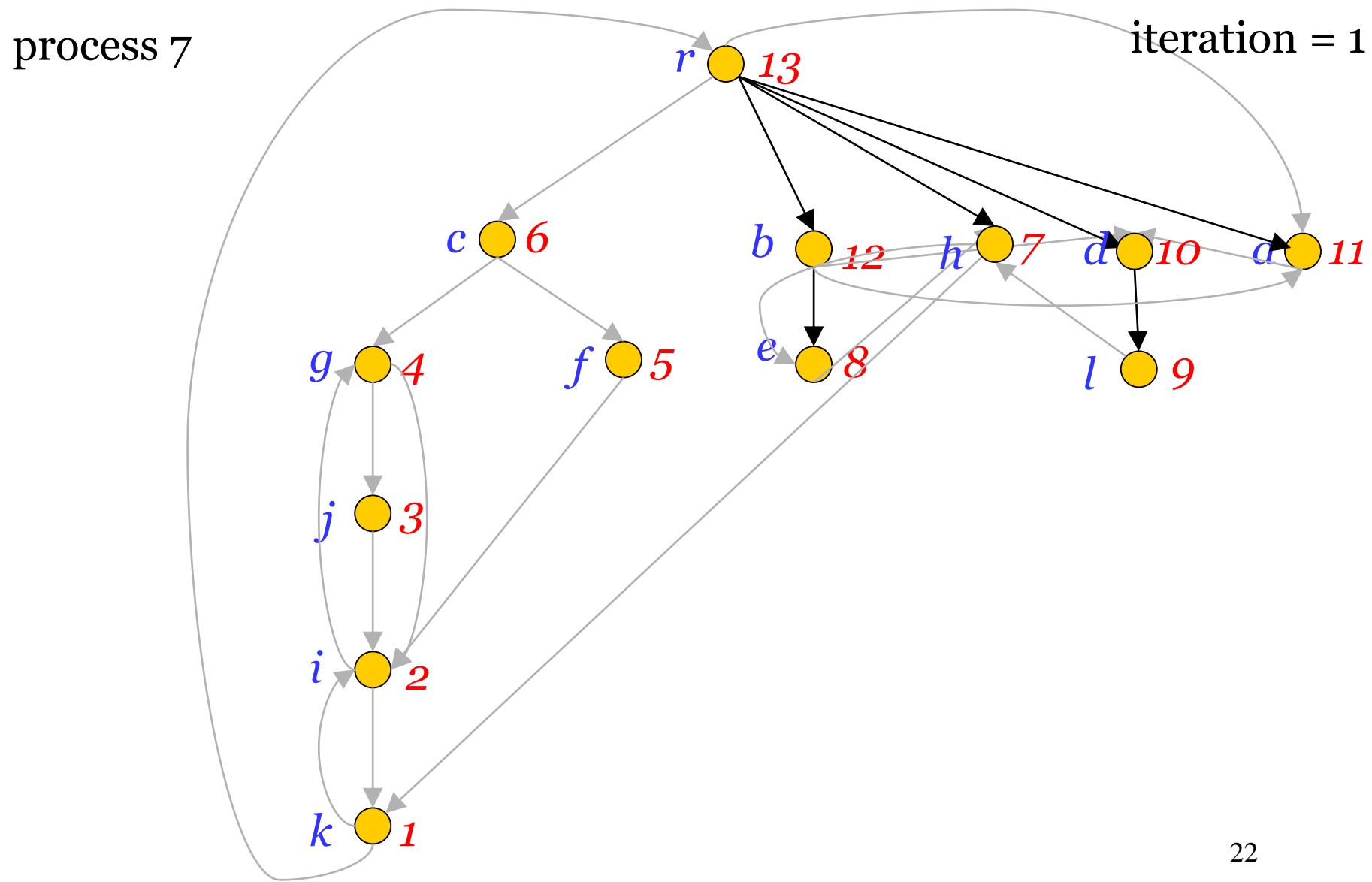
# Iterative Algorithm: Example



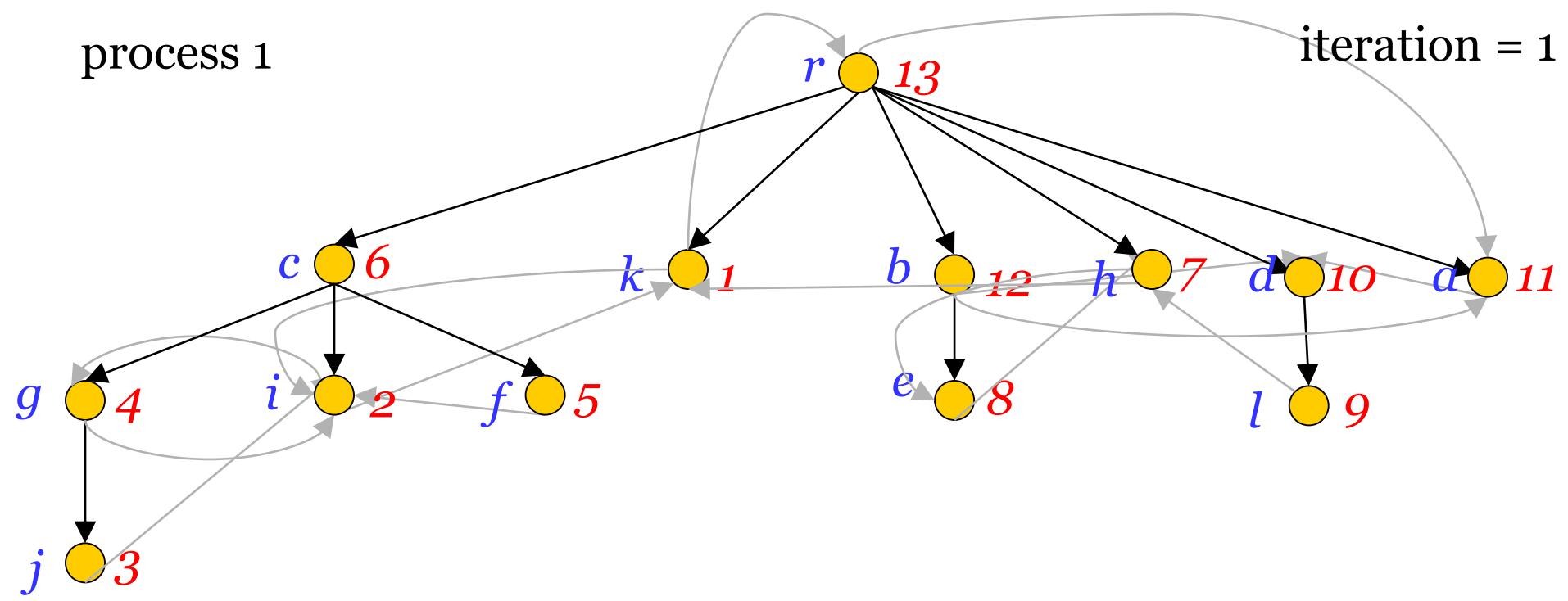
# Iterative Algorithm: Example



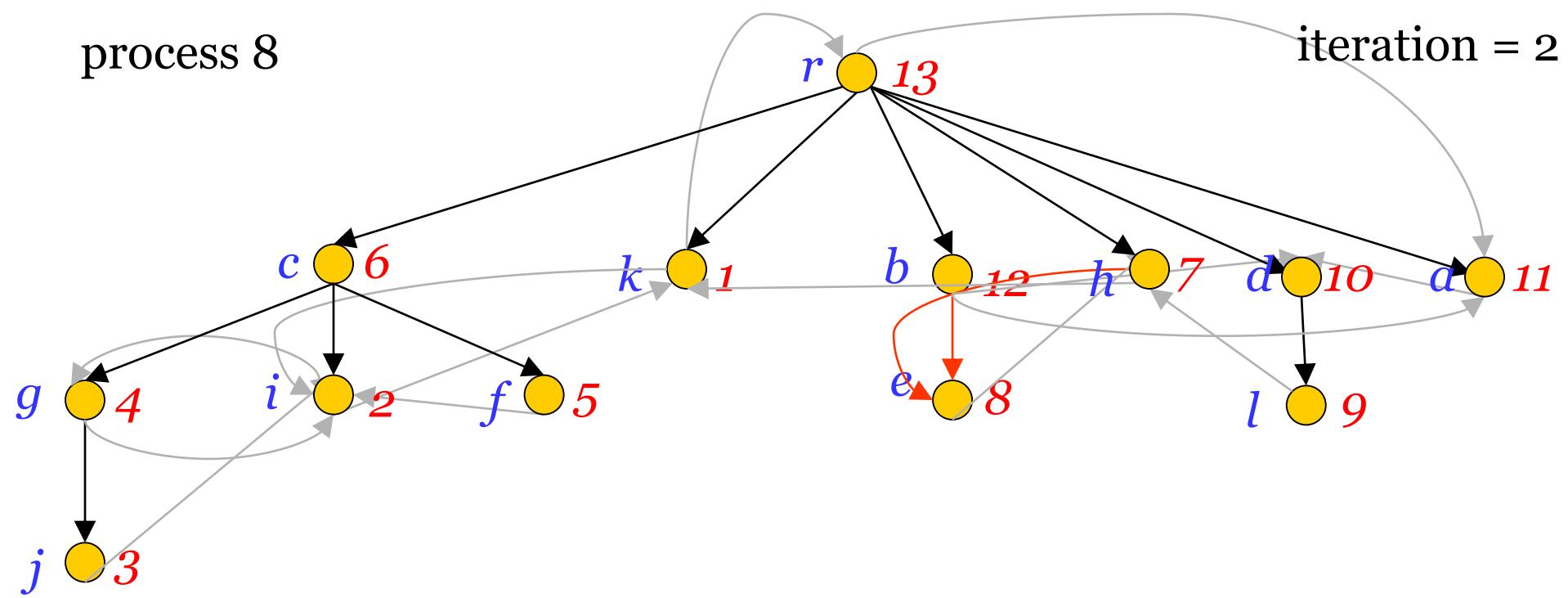
# Iterative Algorithm: Example



# Iterative Algorithm: Example

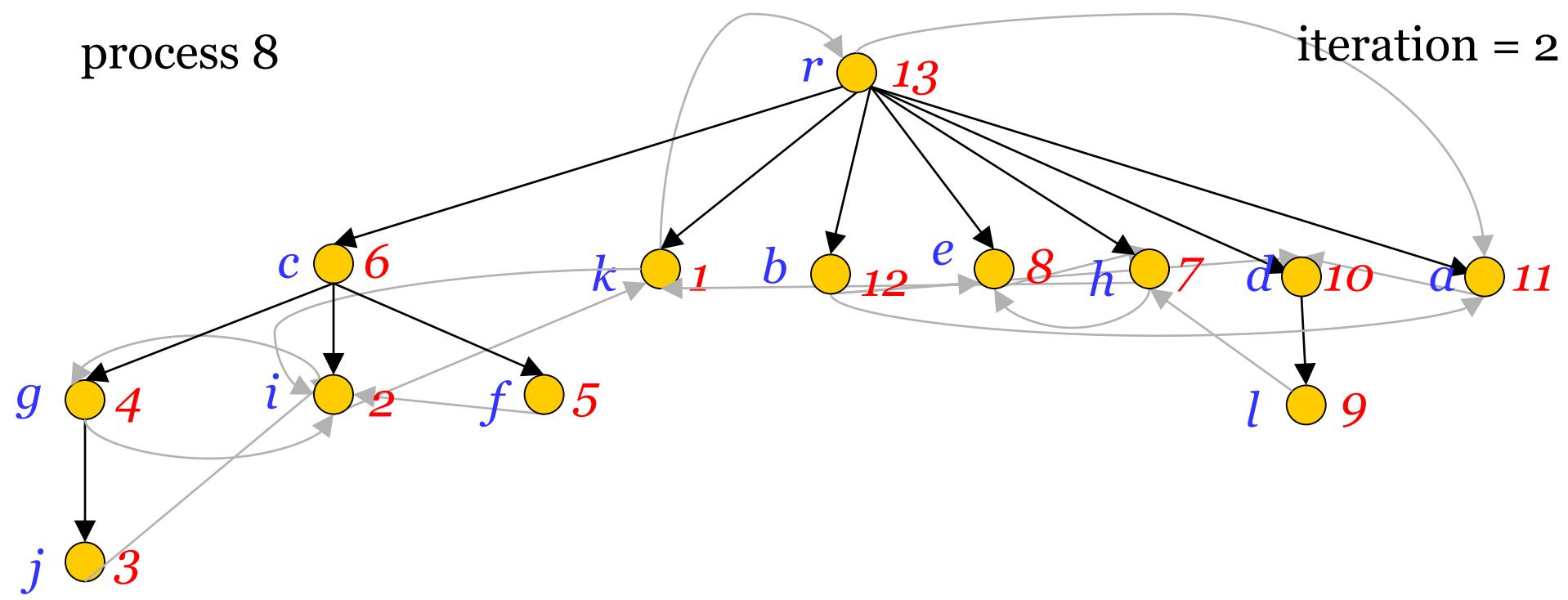


# Iterative Algorithm: Example

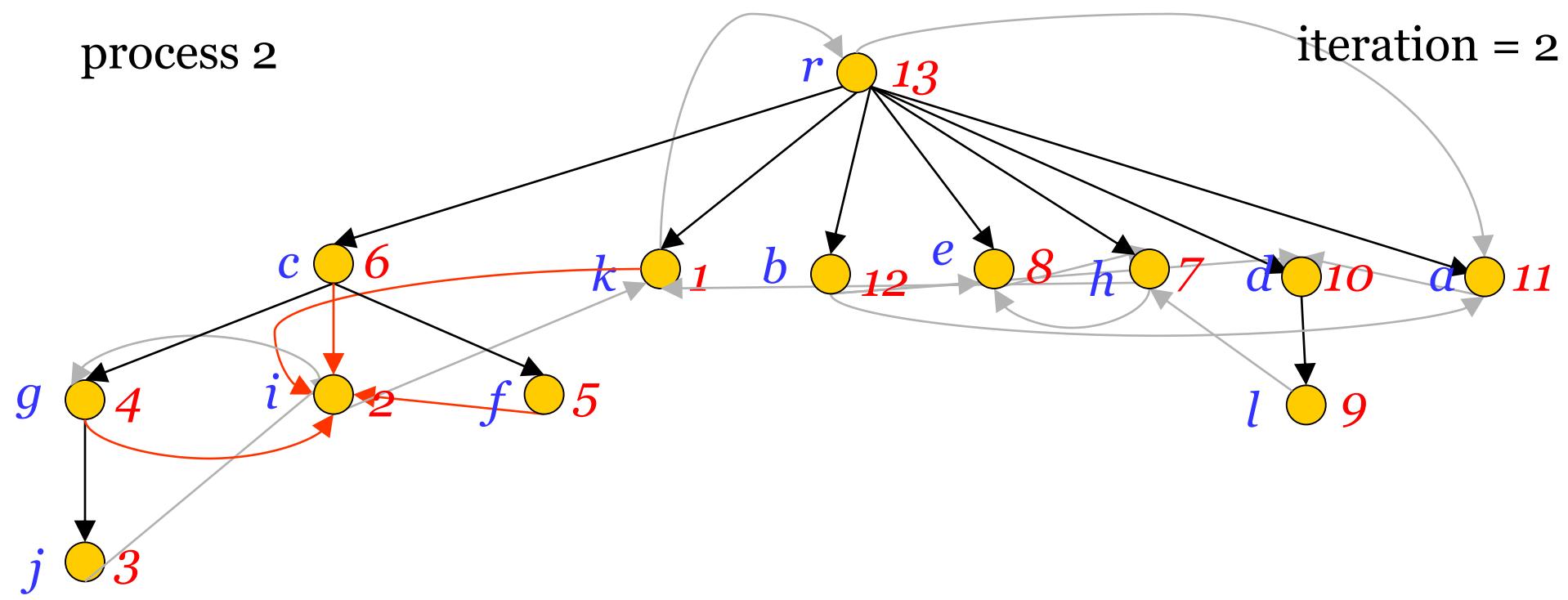


# Iterative Algorithm: Example

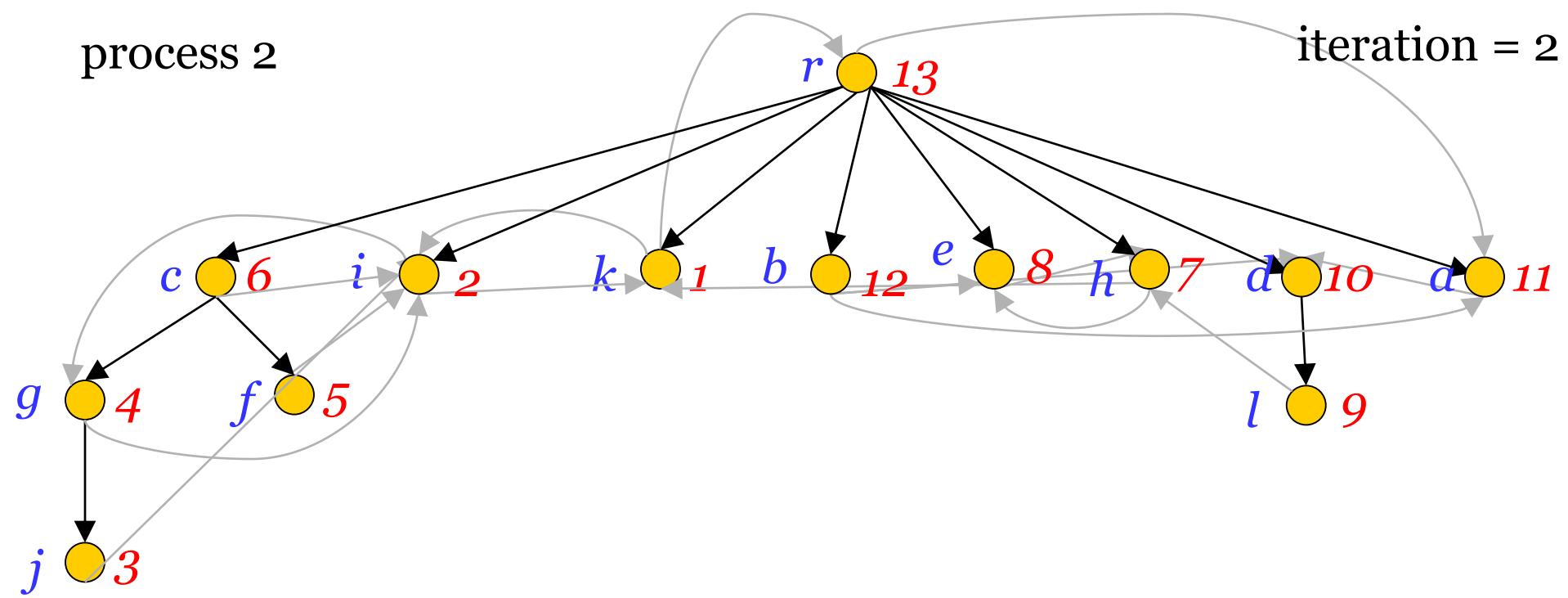
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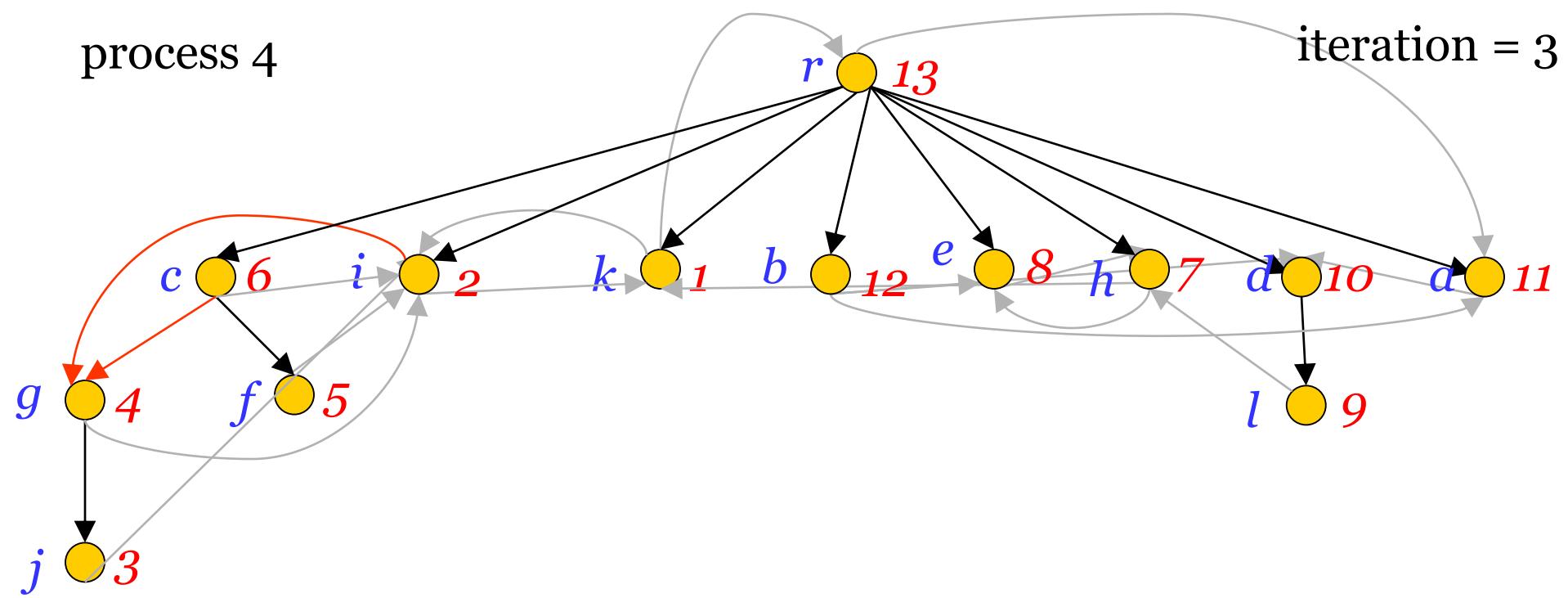
# Iterative Algorithm: Example



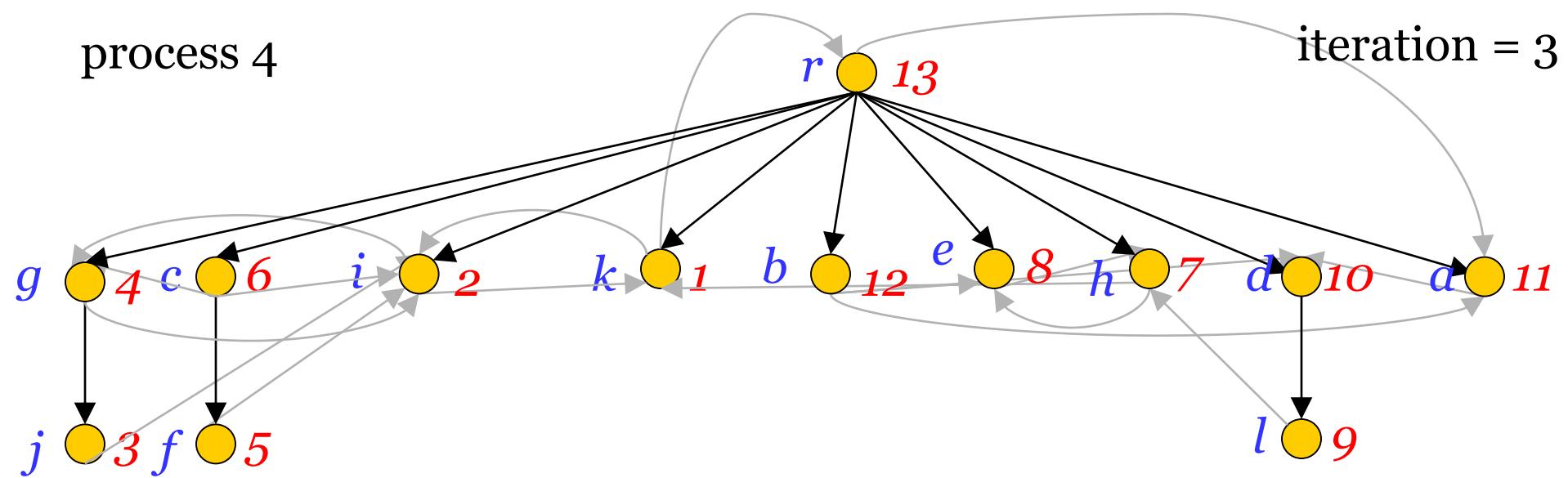
# Iterative Algorithm: Example



# Iterative Algorithm: Example



# Iterative Algorithm: Example



But we need one more iteration to verify  
that nothing changes

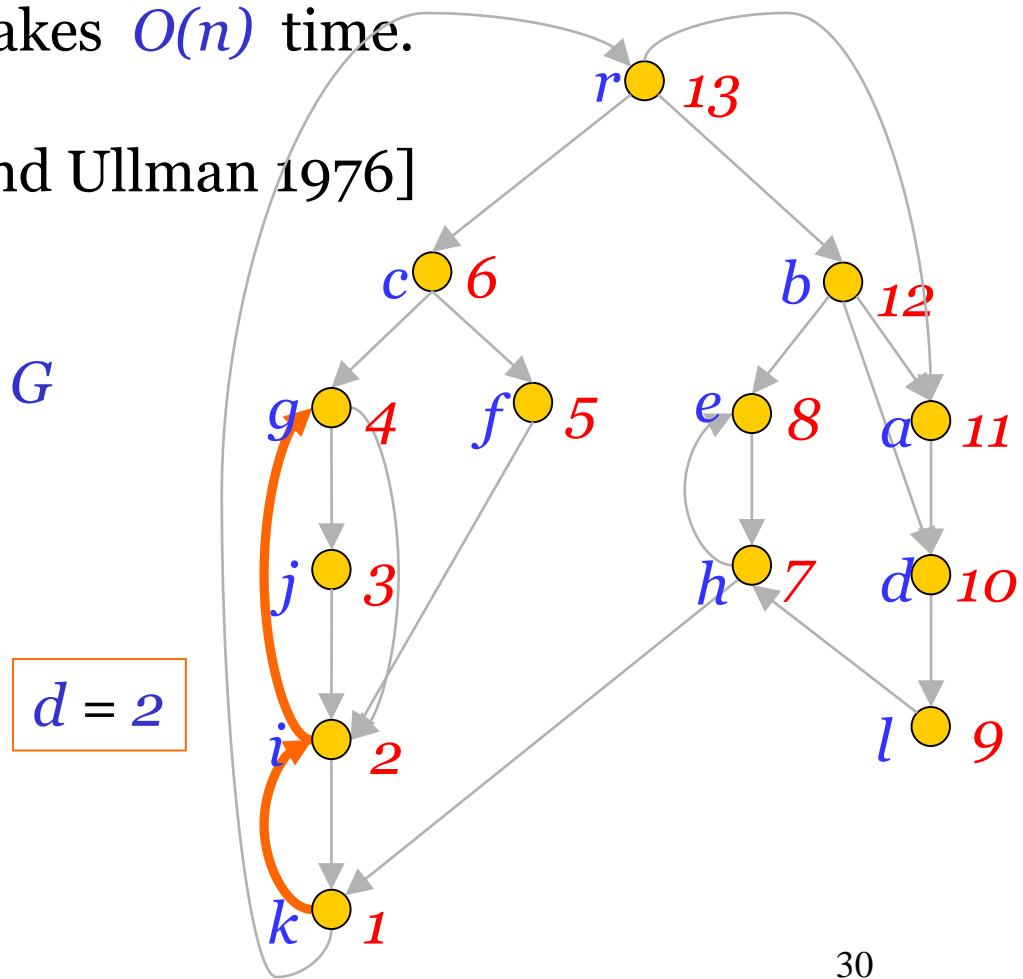
# Iterative Algorithm

## Running Time

Each pairwise intersection takes  $O(n)$  time.

#iterations  $\leq d + 3$ . [Kam and Ullman 1976]

$d = \max$  #back-edges in any  
cycle-free path of  $G$



# Iterative Algorithm

---

## Running Time

Each pairwise intersection takes  $O(n)$  time.

The number of iterations is  $\leq d + 3$ .

$d = \max$  #back-edges in any cycle-free path of  $G$

$$= O(n)$$

Running time =  $O(mn^2)$

This bound is tight, but very pessimistic in practice.

# A Fast Dominator Algorithm

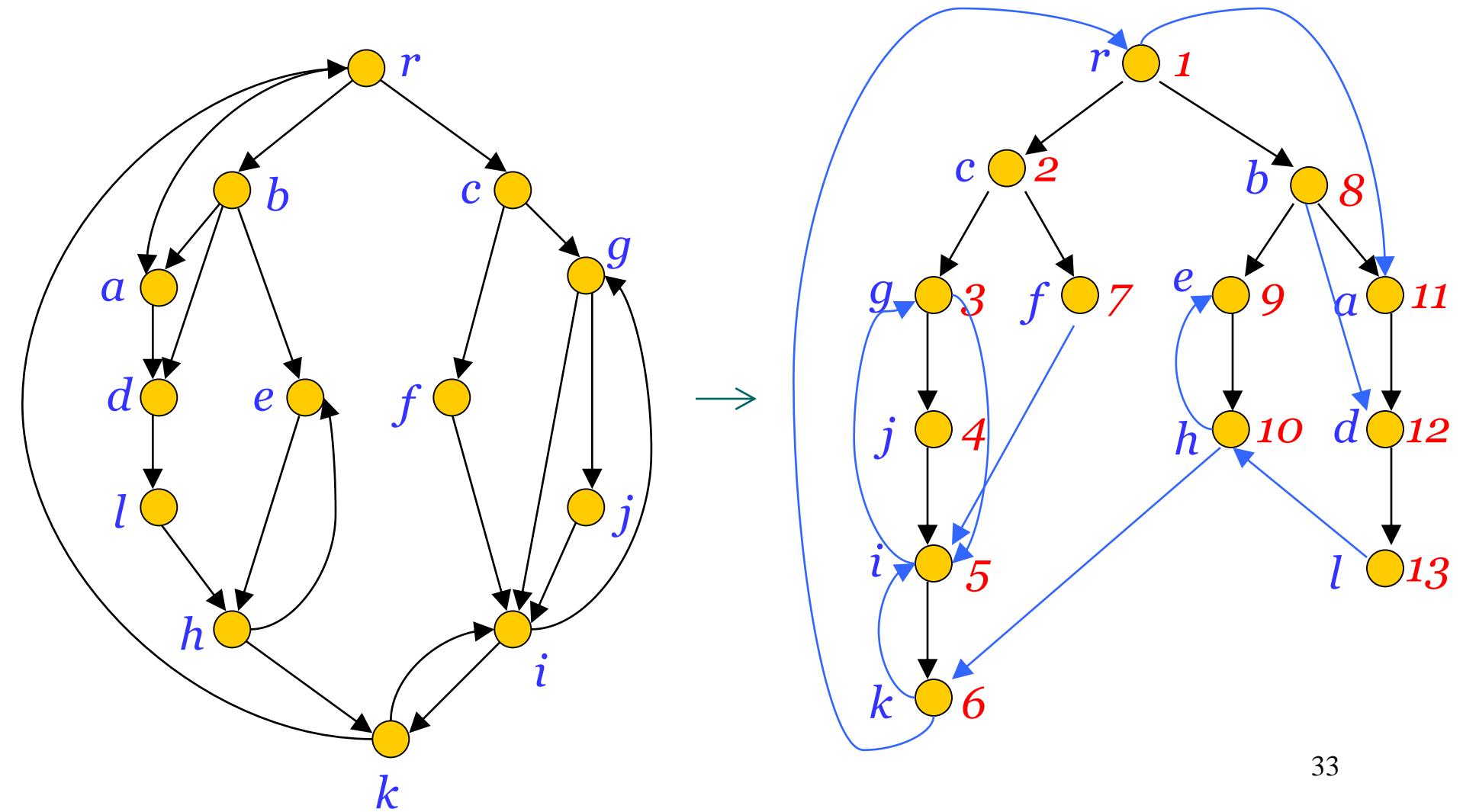
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Lengauer-Tarjan [1979]:  $O(m \cdot \alpha(m,n))$  time

A simpler version runs in  $O(m \cdot \log_{2+\lfloor m/n \rfloor} n)$  time

# The Lengauer-Tarjan Algorithm: Depth-First Search

Perform a depth-first search on  $G \Rightarrow$  DFS-tree  $T$



# The Lengauer-Tarjan Algorithm: Depth-First Search

---

Depth-First Search Tree  $T$ :

We refer to the vertices by their DFS numbers:

$v < w$  :  $v$  was visited by DFS before  $w$

Notation

$v^* \rightarrow w$  :  $v$  is an ancestor of  $w$  in  $T$

$v^+ \rightarrow w$  :  $v$  is a proper ancestor of  $w$  in  $T$

$\text{parent}(v)$  : parent of  $v$  in  $T$

Property 1

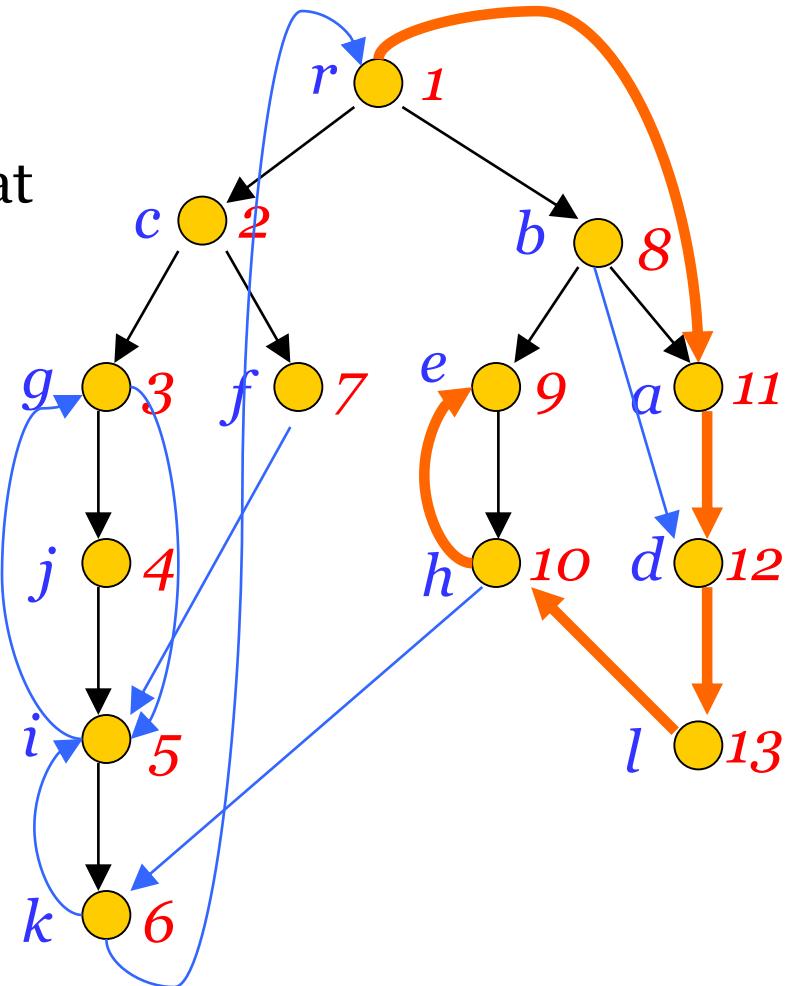
$\forall v, w$  such that  $v \leq w$ ,  $(v, w) \in E \Rightarrow v^* \rightarrow w$

# The Lengauer-Tarjan Algorithm: Semidominators

Semidominator path (SDOM-path):

$P = (v_o = v, v_1, v_2, \dots, v_k = w)$  such that  
 $v_i > w$ , for  $1 \leq i \leq k-1$

$(r, a, d, l, h, e)$  is an SDOM-path for  $e$



# The Lengauer-Tarjan Algorithm: Semidominators

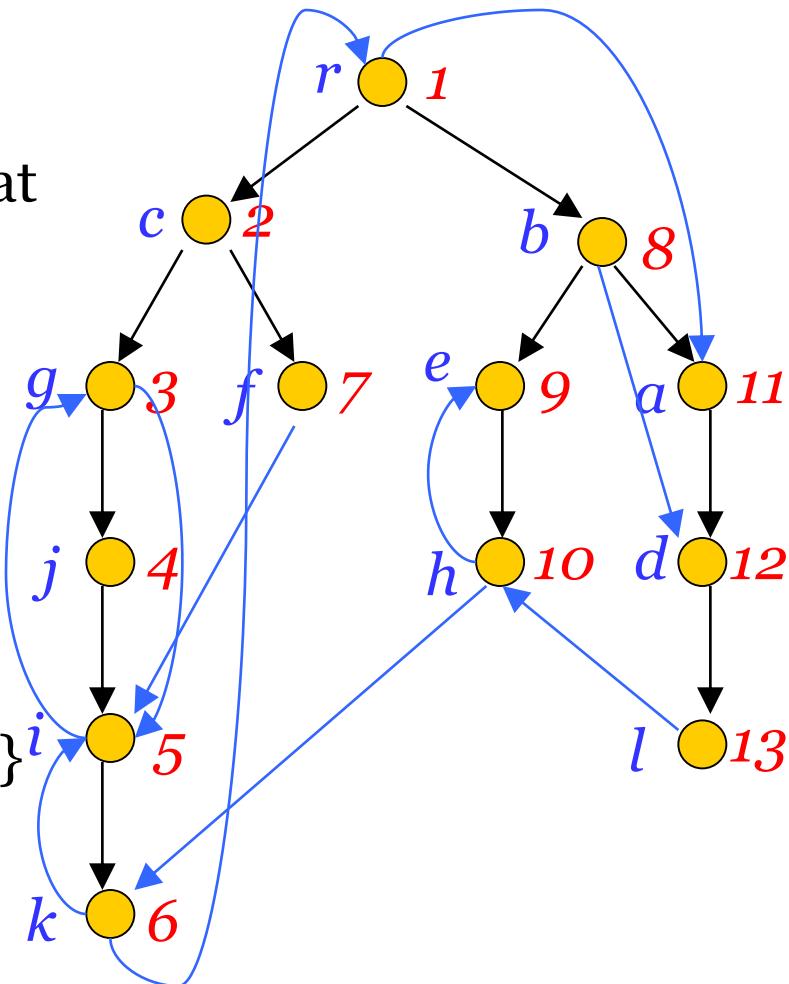
Semidominator path (SDOM-path):

$P = (v_0 = v, v_1, v_2, \dots, v_k = w)$  such that  
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Semidominator:

$sdom(w) =$

$\min \{ v \mid \exists \text{ SDOM-path from } v \text{ to } w \}$



# The Lengauer-Tarjan Algorithm: Semidominators

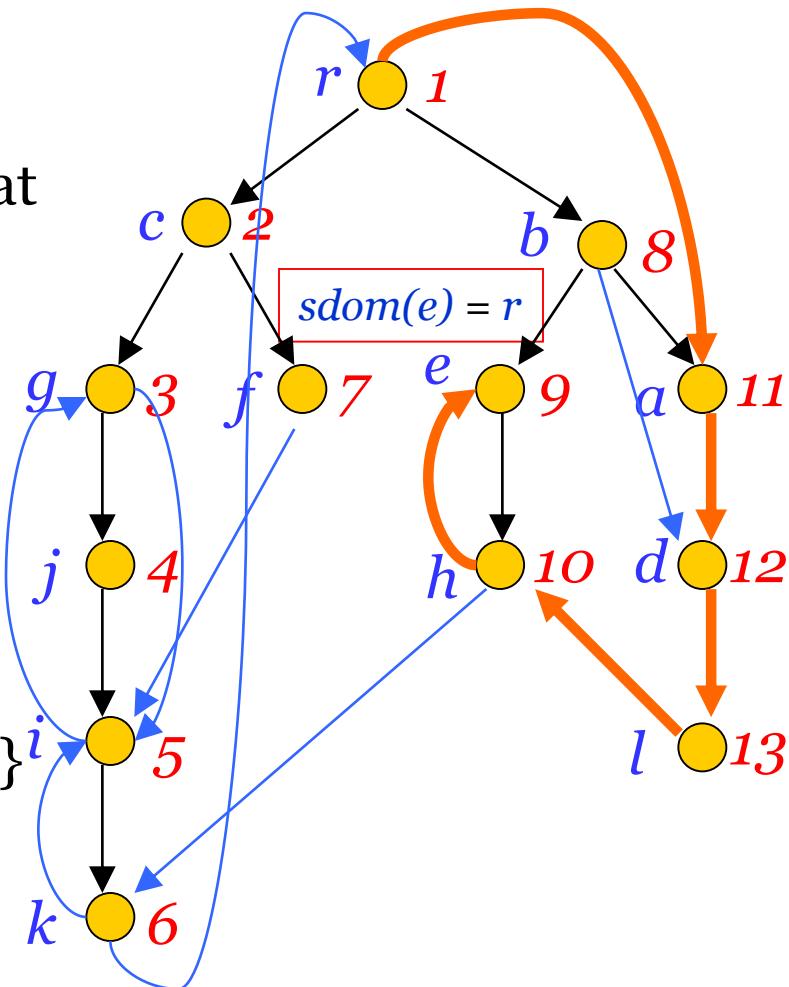
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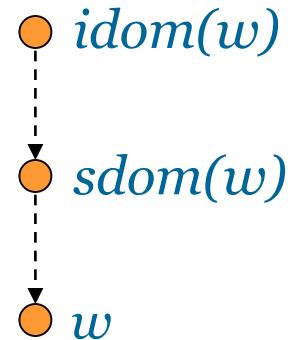
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# The Lengauer-Tarjan Algorithm: Semidominators

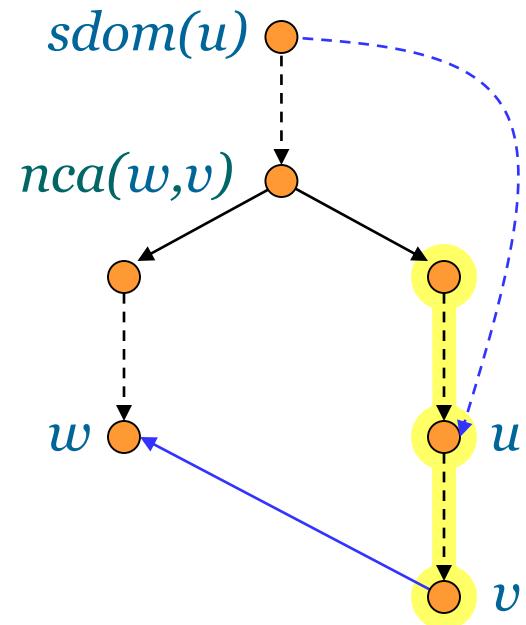
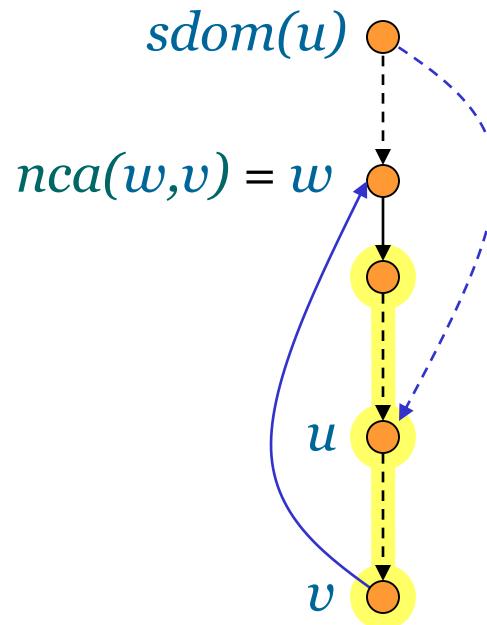
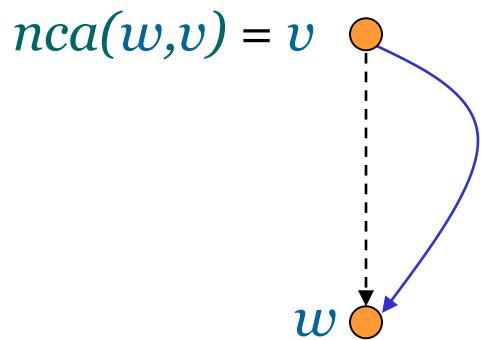
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- For any  $w \neq r$ ,  $idom(w) \xrightarrow{*} sdom(w) \xrightarrow{+} w$ .



# The Lengauer-Tarjan Algorithm: Semidominators

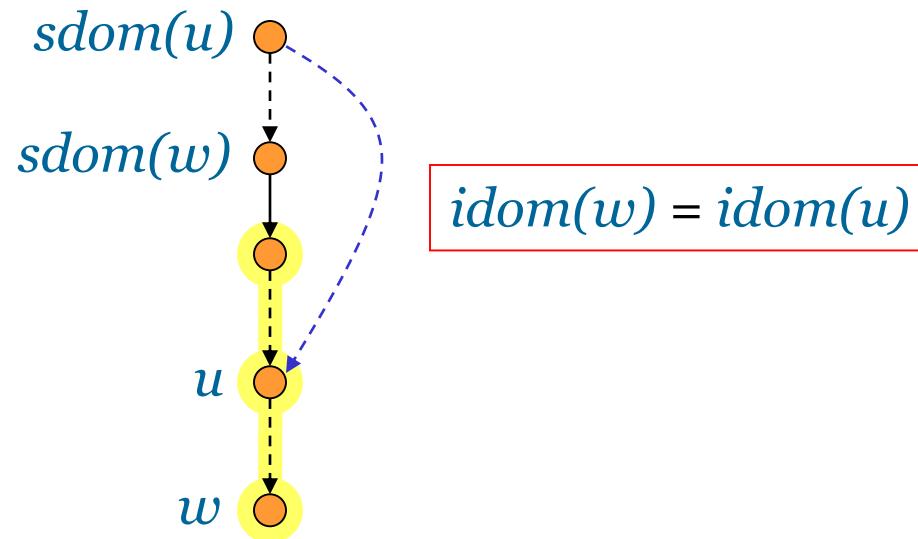
- For any  $w \neq r$ ,  $\text{idom}(w) \xrightarrow{*} \text{sdom}(w) \xrightarrow{+} w$ .
- $\text{sdom}(w) = \min (\{ v \mid (v, w) \in E \text{ and } v < w \} \cup \{ \text{sdom}(u) \mid u > w \text{ and } \exists (v, w) \in E \text{ such that } u \xrightarrow{*} v \})$ .



# The Lengauer-Tarjan Algorithm: Semidominators

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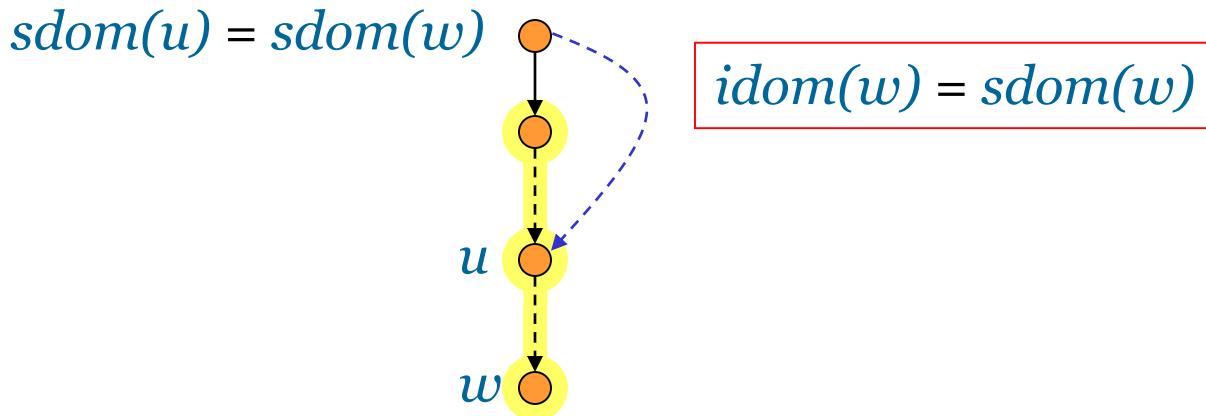
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- Let  $w \neq r$  and let  $u$  be any vertex with  $\min \text{sdom}(u)$  that satisfies  $\text{sdom}(w) \xrightarrow{+} u \xrightarrow{*} w$ . Then  $\text{idom}(w) = \text{idom}(u)$ .



# The Lengauer-Tarjan Algorithm: Semidominators

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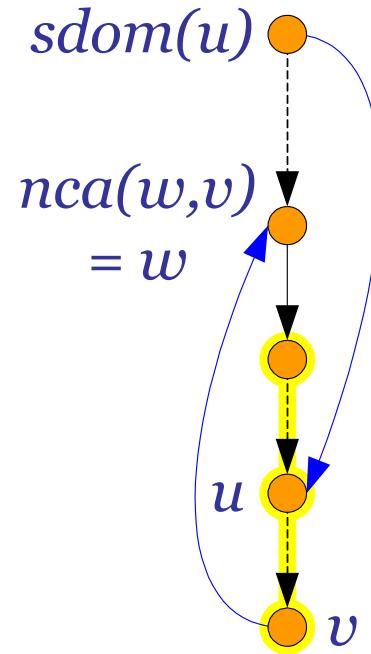
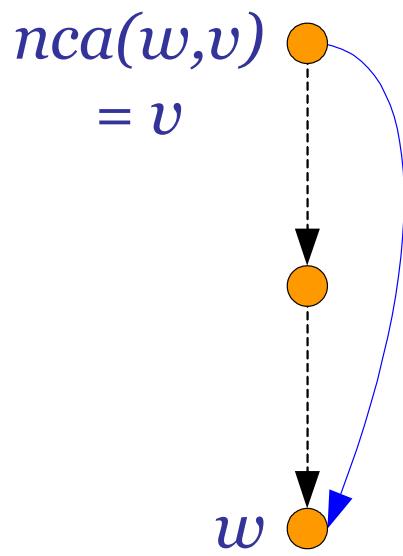
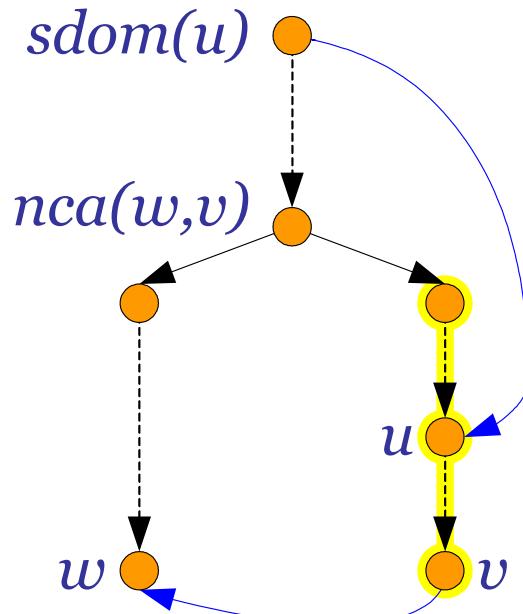
- For any  $w \neq r$ ,  $idom(w) * \rightarrow sdom(w) + \rightarrow w$ .
- $sdom(w) = \min (\{ v \mid (v, w) \in E \text{ and } v < w \} \cup \{ sdom(u) \mid u > w \text{ and } \exists (v, w) \in E \text{ such that } u * \rightarrow v \})$ .
- Let  $w \neq r$  and let  $u$  be any vertex with  $\min sdom(u)$  that satisfies  $sdom(w) + \rightarrow u * \rightarrow w$ . Then  $idom(w) = idom(u)$ . Moreover, if  $sdom(u) = sdom(w)$  then  $idom(w) = sdom(w)$ .



## Overview of the Algorithm

1. Carry out a DFS.
2. Process the vertices in reverse preorder. For vertex  $w$ , compute  $sdom(w)$ .
3. Implicitly define  $idom(w)$ .
4. Explicitly define  $idom(w)$  by a preorder pass.

# Evaluating minima on tree paths



If we process vertices in **reverse preorder** then the *sdom* values we need are known.

# Evaluating minima on tree paths

---

**Data Structure:** Maintain forest  $F$  and supports the operations:

**link( $v, w$ ):** Add the edge  $(v, w)$  to  $F$ .

**eval( $v$ ):** Let  $r$  be the root of the tree that contains  $v$  in  $F$ .  
If  $v = r$  then return  $v$ . Otherwise return any vertex  
with minimum  $sdom$  among the vertices  $u$  that  
satisfy  $r \xrightarrow{+} u \xrightarrow{*} v$ .

Initially every vertex in  $V$  is a root in  $F$ .

# The Lengauer-Tarjan Algorithm

---

$\text{dfs}(r)$

**for all**  $w \in V$  in reverse preorder **do**

**for all**  $v \in \text{pred}(w)$  **do**

$u \leftarrow \text{eval}(v)$

**if**  $\text{semi}(u) < \text{semi}(w)$  **then**  $\text{semi}(w) \leftarrow \text{semi}(u)$

**done**

add  $w$  to the bucket of  $\text{semi}(w)$

$\text{link}(\text{parent}(w), w)$

**for all**  $v$  in the bucket of  $\text{parent}(w)$  **do**

delete  $v$  from the bucket of  $\text{parent}(w)$

$u \leftarrow \text{eval}(v)$

**if**  $\text{semi}(u) < \text{semi}(v)$  **then**  $\text{dom}(v) \leftarrow u$

**else**  $\text{dom}(v) \leftarrow \text{parent}(w)$

**done**

**done**

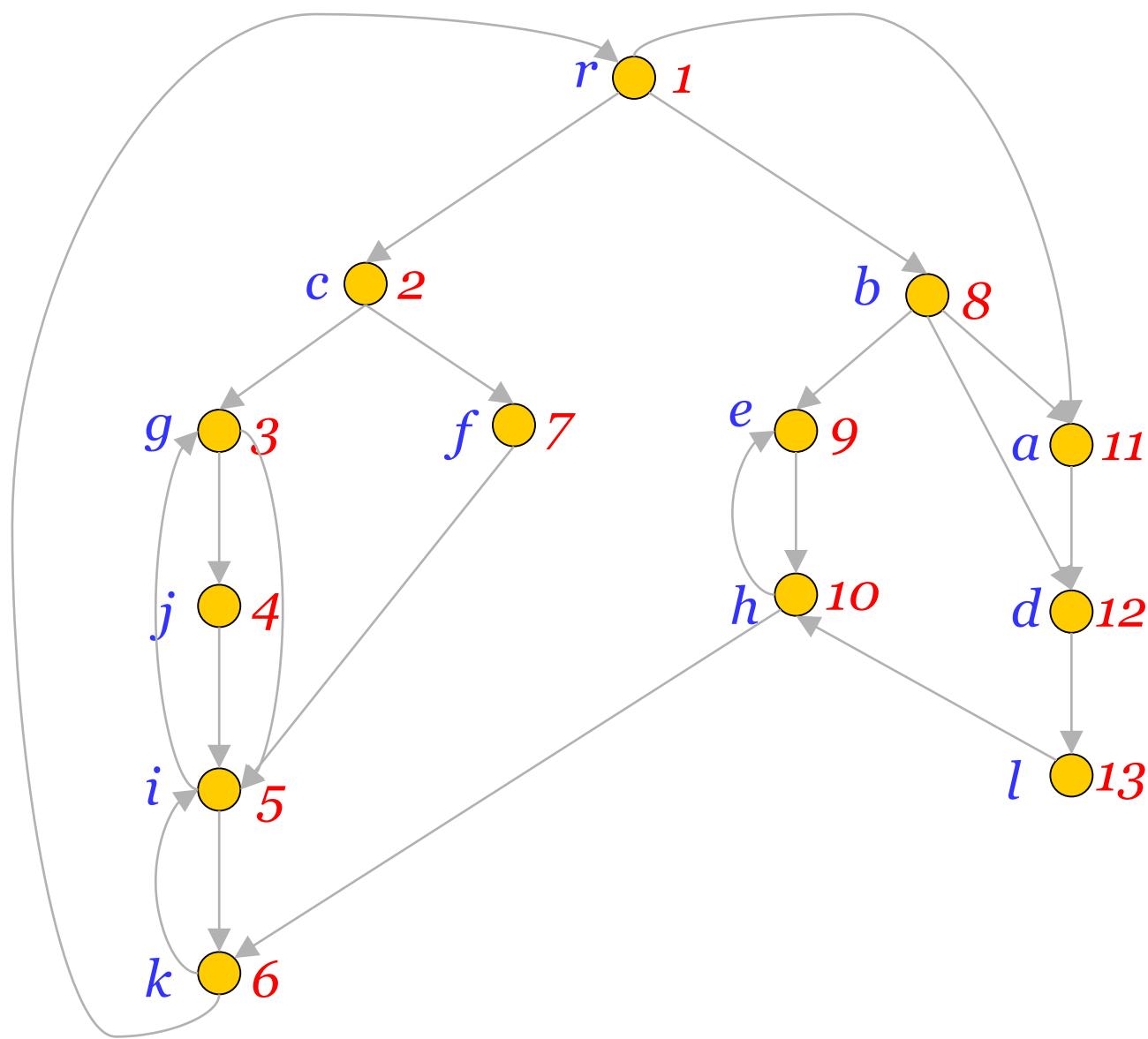
**for all**  $w \in V$  in reverse preorder **do**

**if**  $\text{dom}(w) \neq \text{semi}(w)$  **then**  $\text{dom}(w) \leftarrow \text{dom}(\text{dom}(w))$

**done**

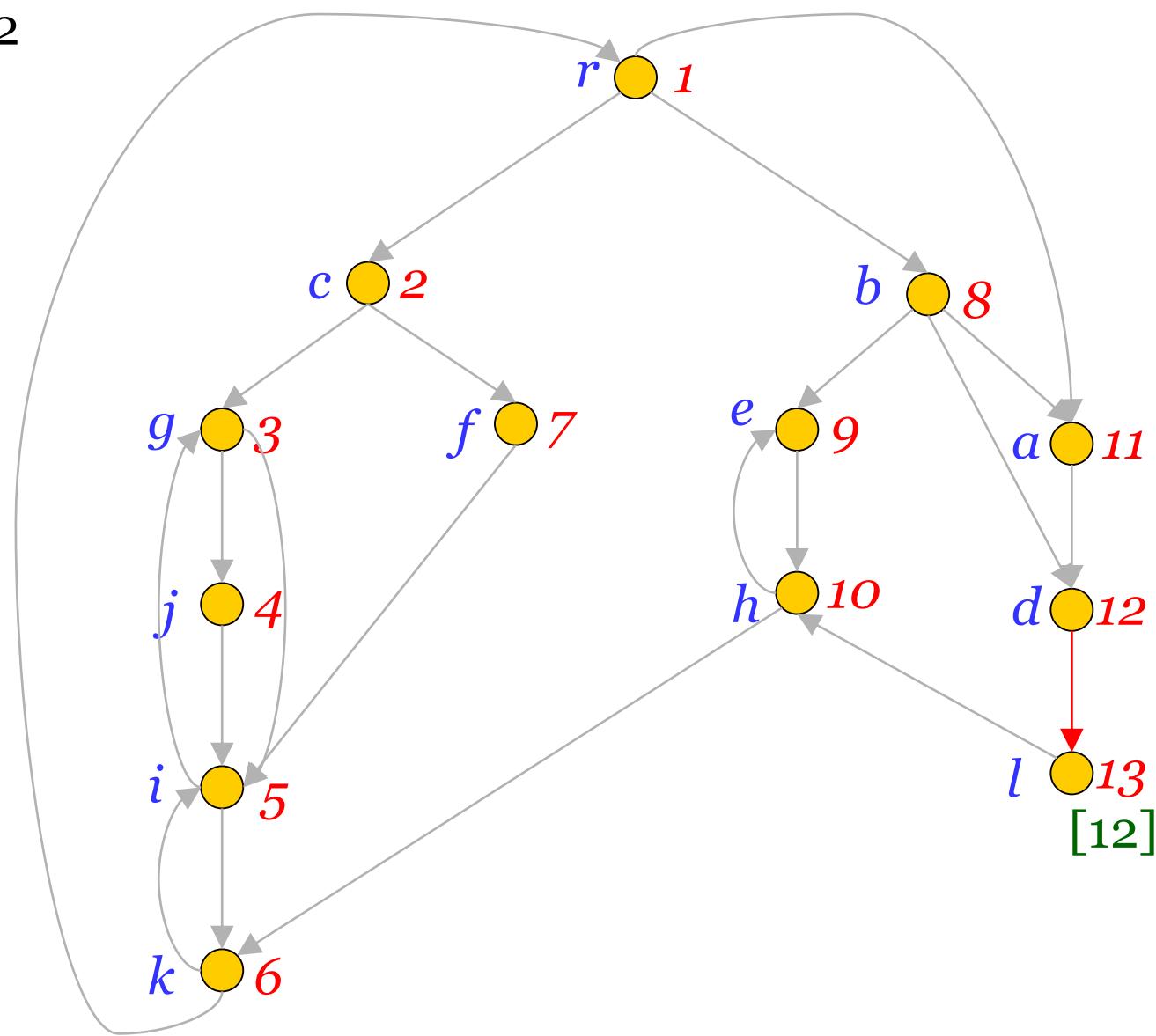
# The Lengauer-Tarjan Algorithm: Example

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# The Lengauer-Tarjan Algorithm: Example

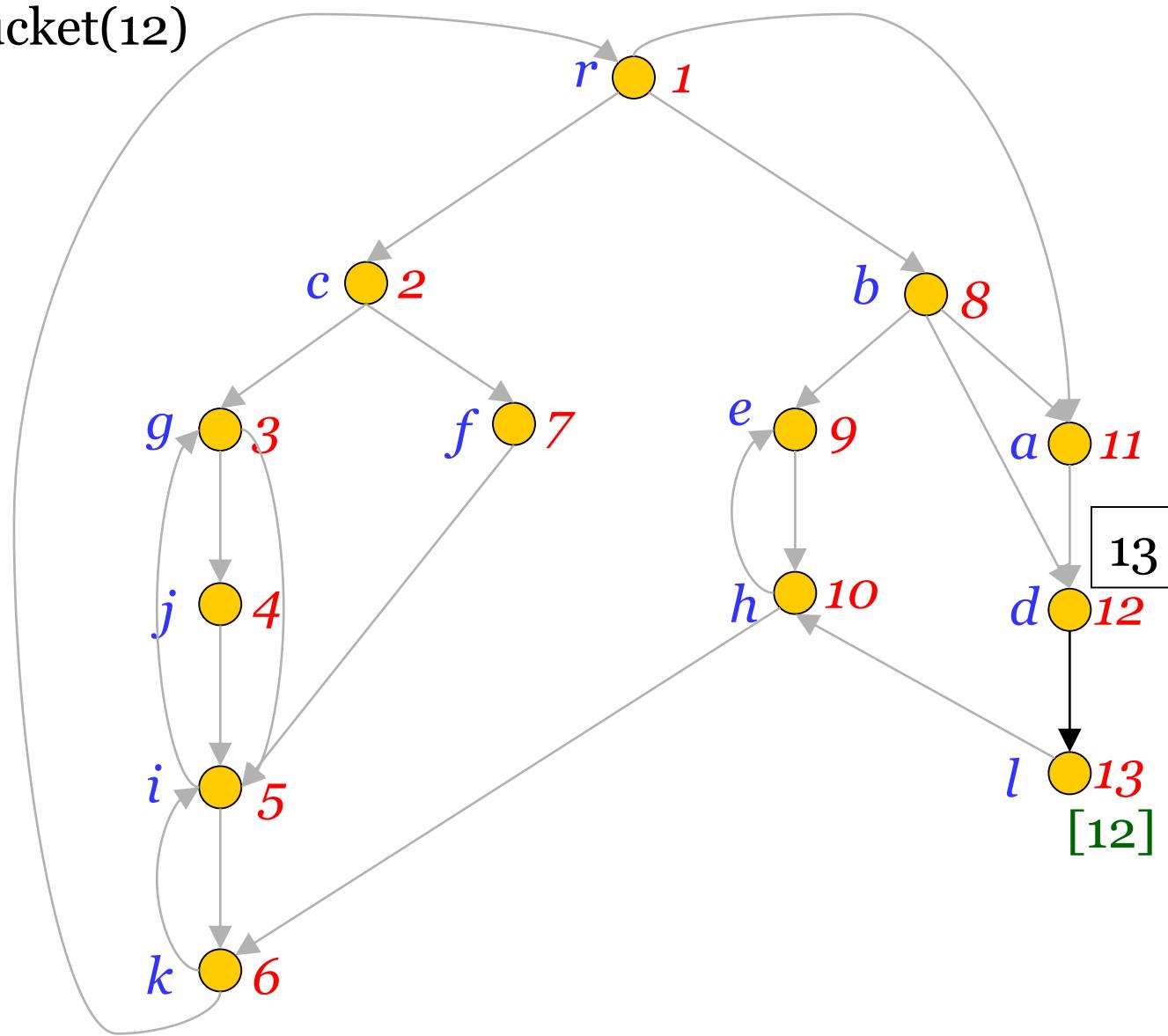
$\text{eval}(12) = 12$



# The Lengauer-Tarjan Algorithm: Example

add 13 to bucket(12)

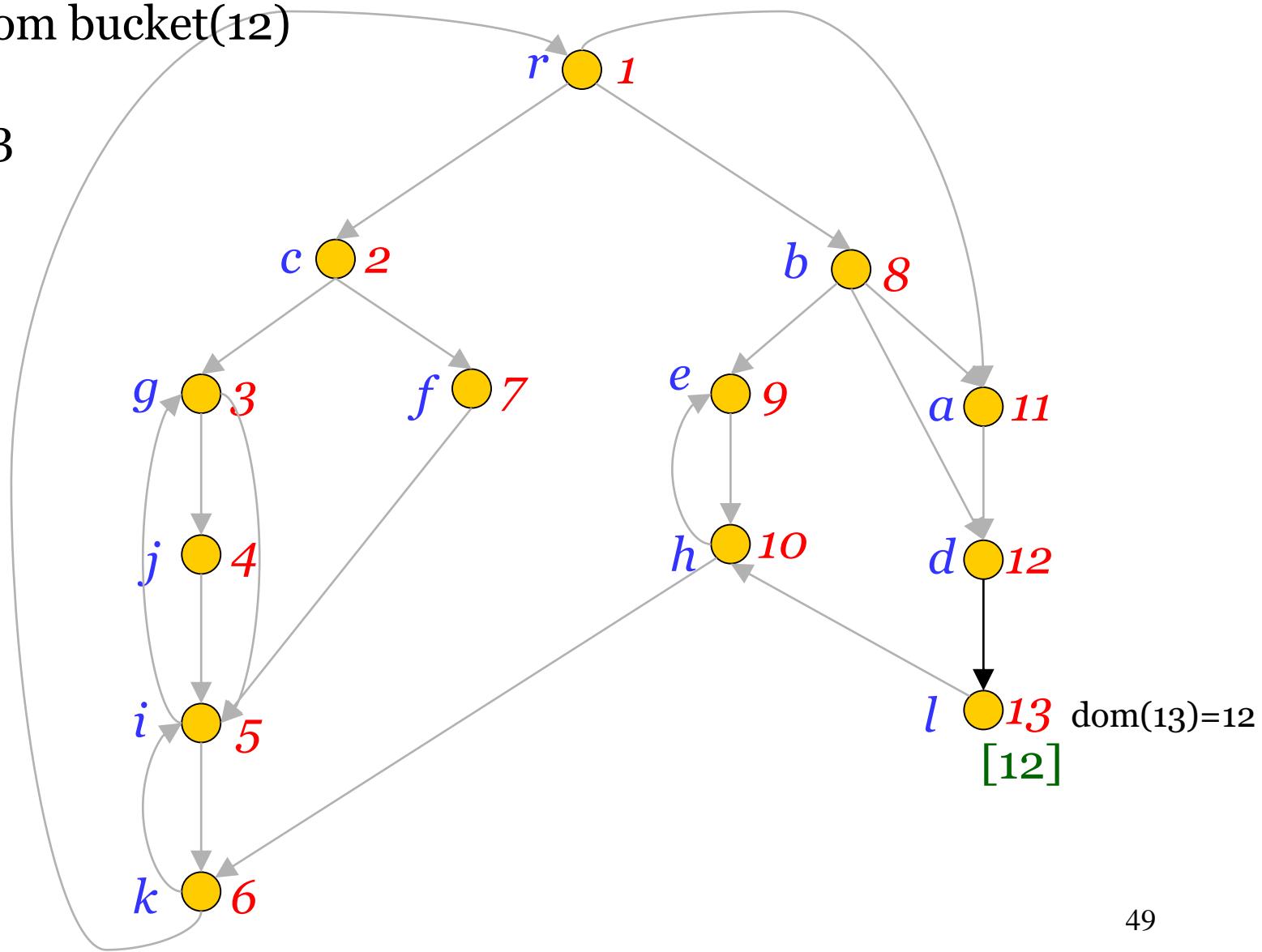
link(13)



# The Lengauer-Tarjan Algorithm: Example

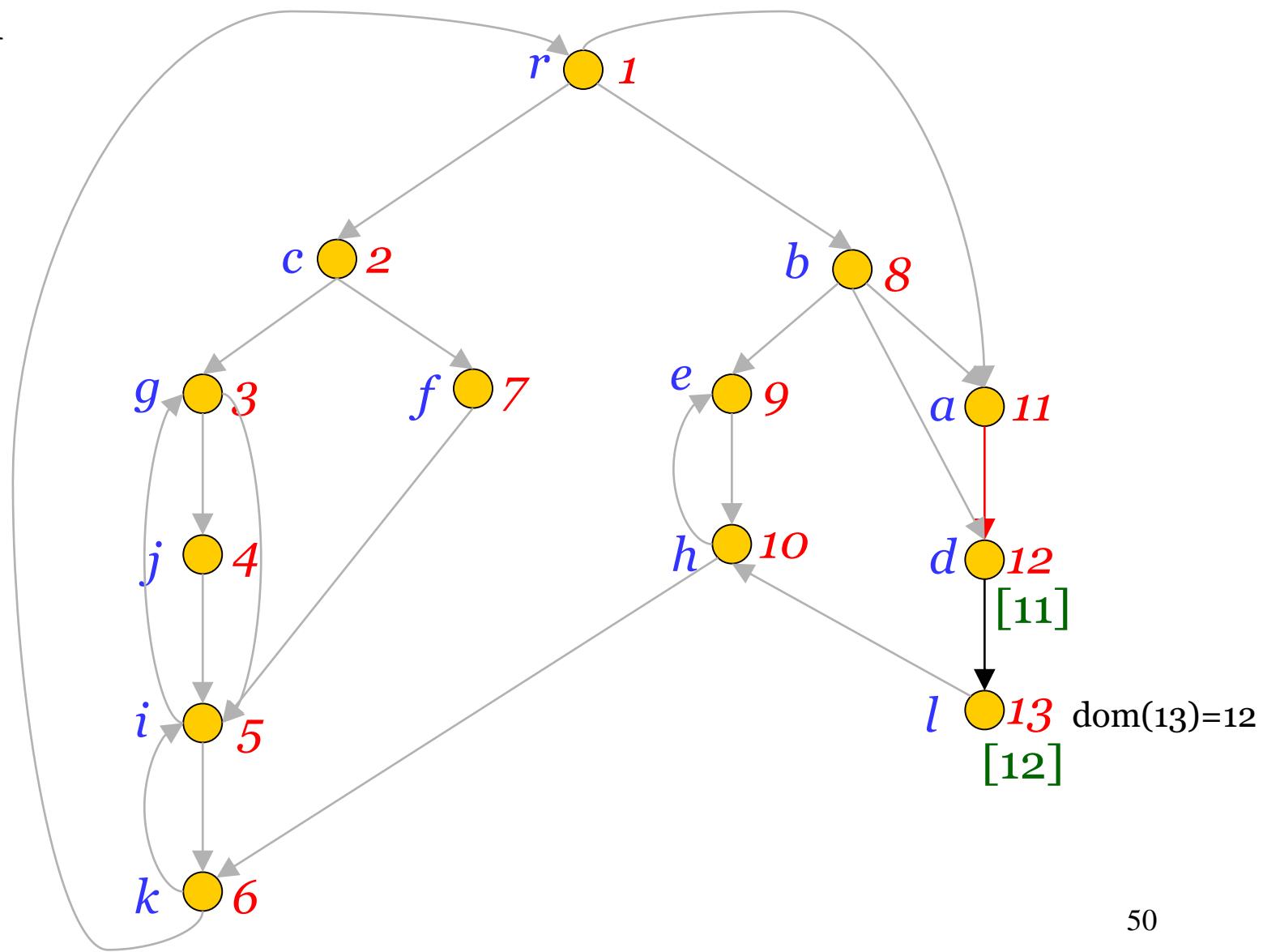
delete 13 from bucket(12)

$\text{eval}(13) = 13$



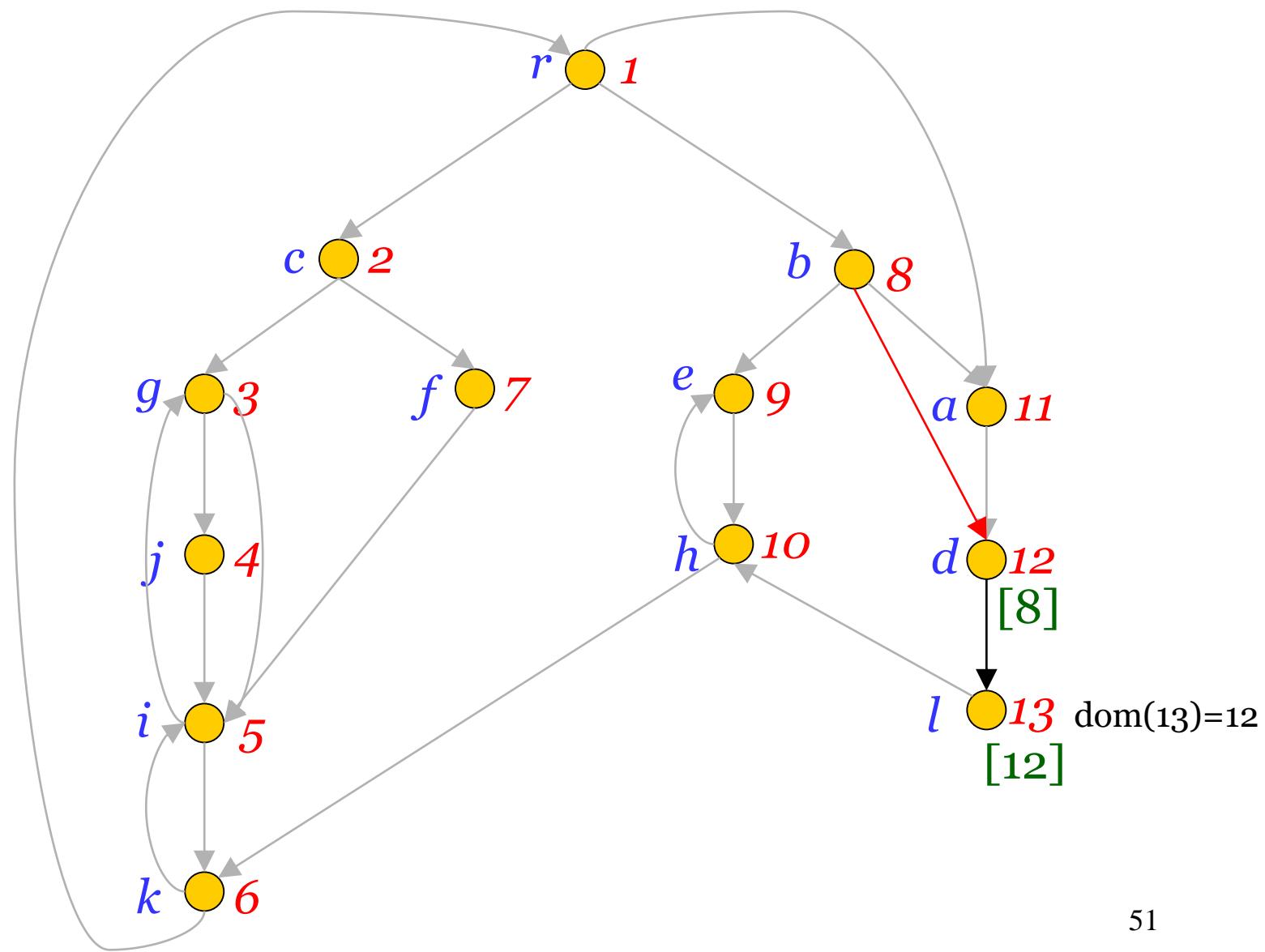
# The Lengauer-Tarjan Algorithm: Example

$\text{eval}(11) = 11$



# The Lengauer-Tarjan Algorithm: Example

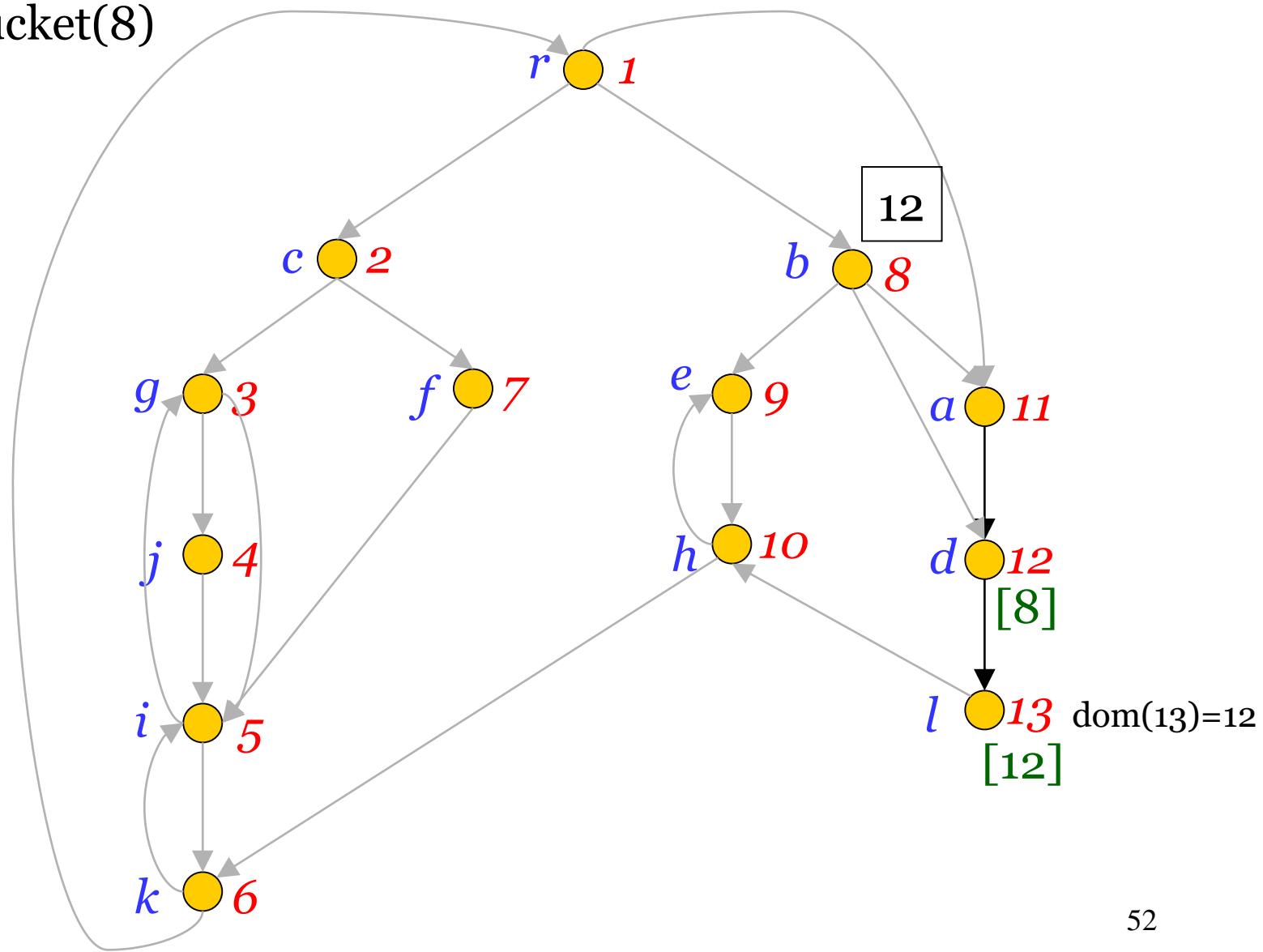
$\text{eval}(8) = 8$



# The Lengauer-Tarjan Algorithm: Example

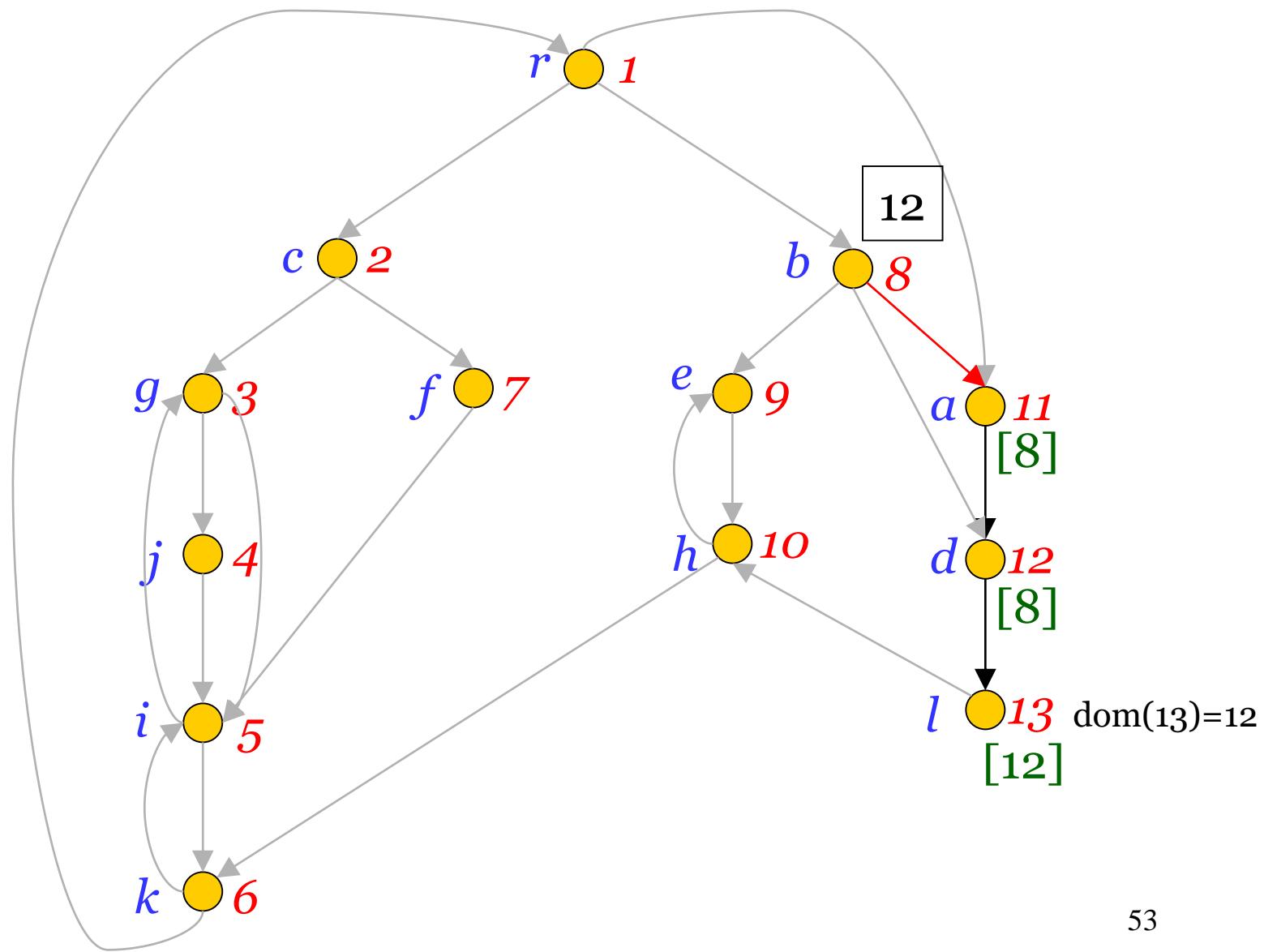
add 12 to bucket(8)

link(12)



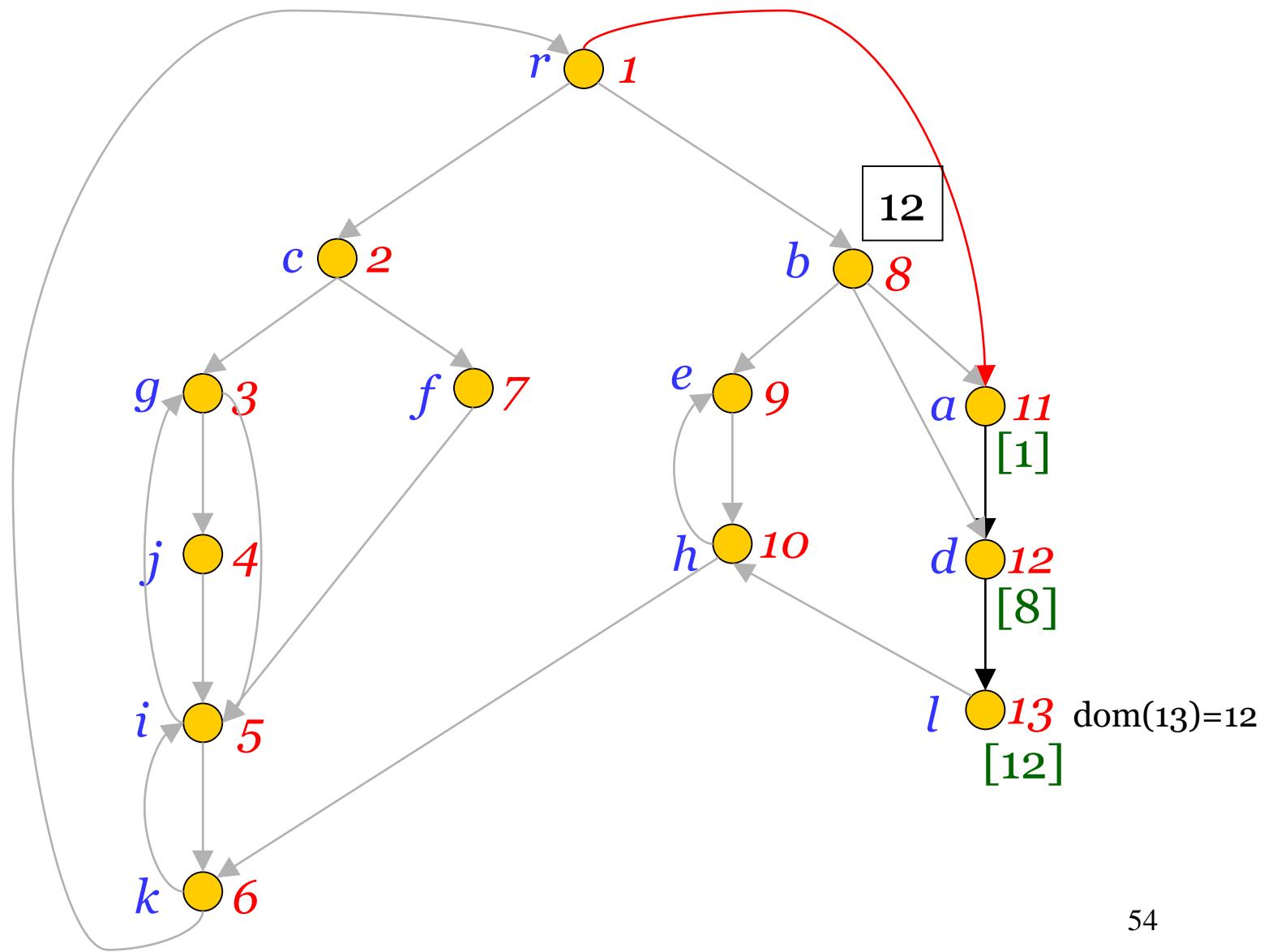
# The Lengauer-Tarjan Algorithm: Example

$\text{eval}(8)=8$



# The Lengauer-Tarjan Algorithm: Example

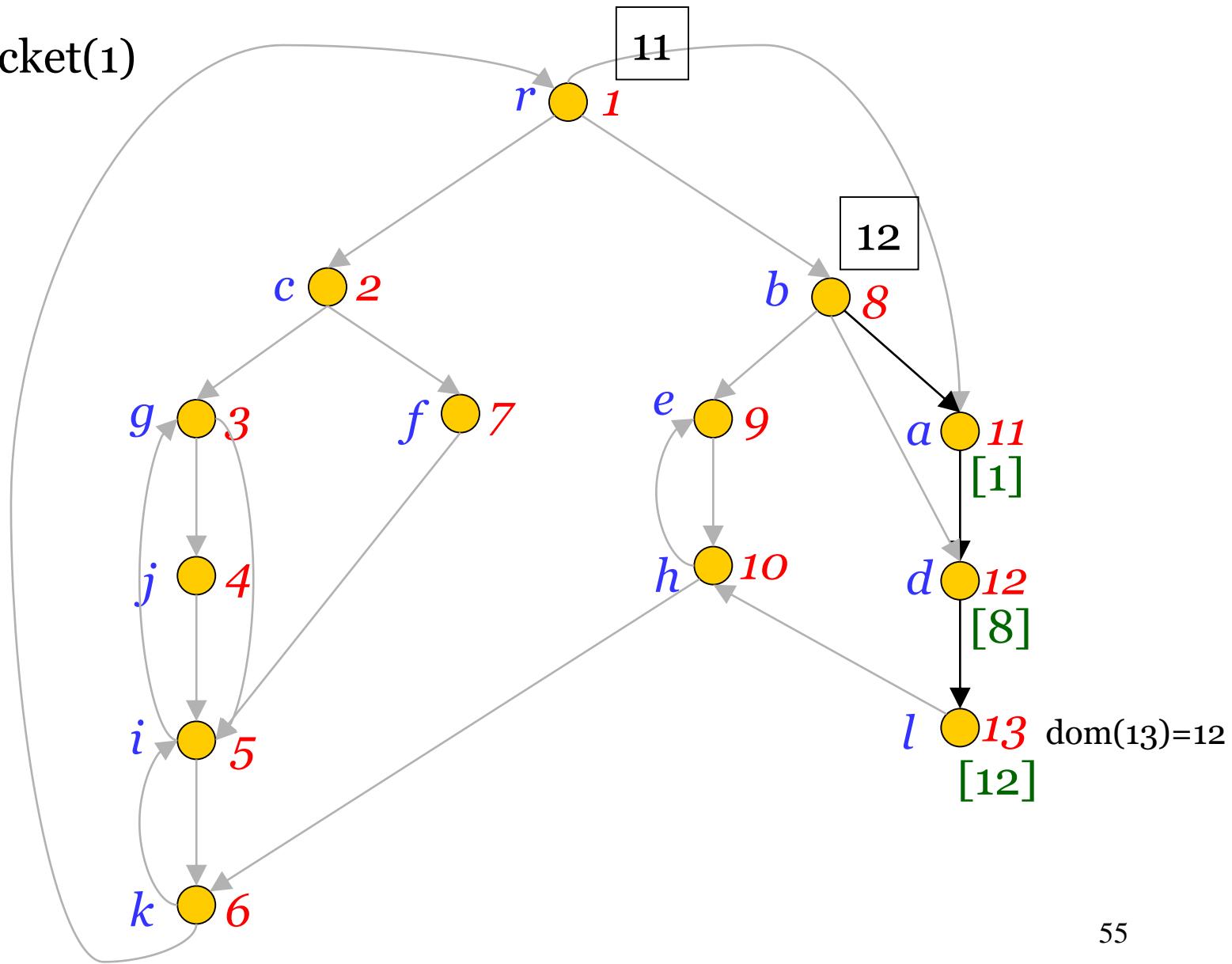
$\text{eval}(1)=1$



# The Lengauer-Tarjan Algorithm: Example

add 11 to bucket(1)

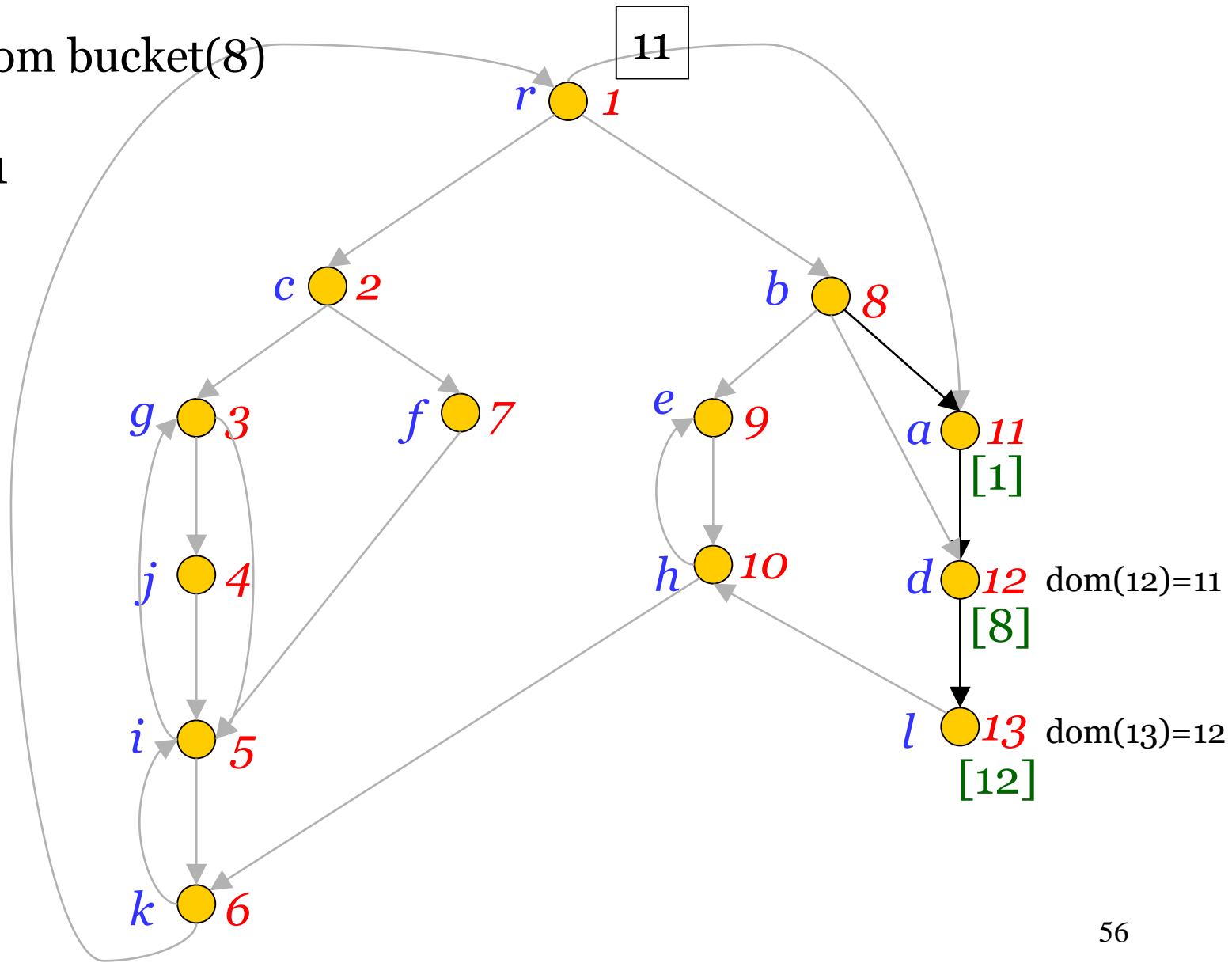
link(11)



# The Lengauer-Tarjan Algorithm: Example

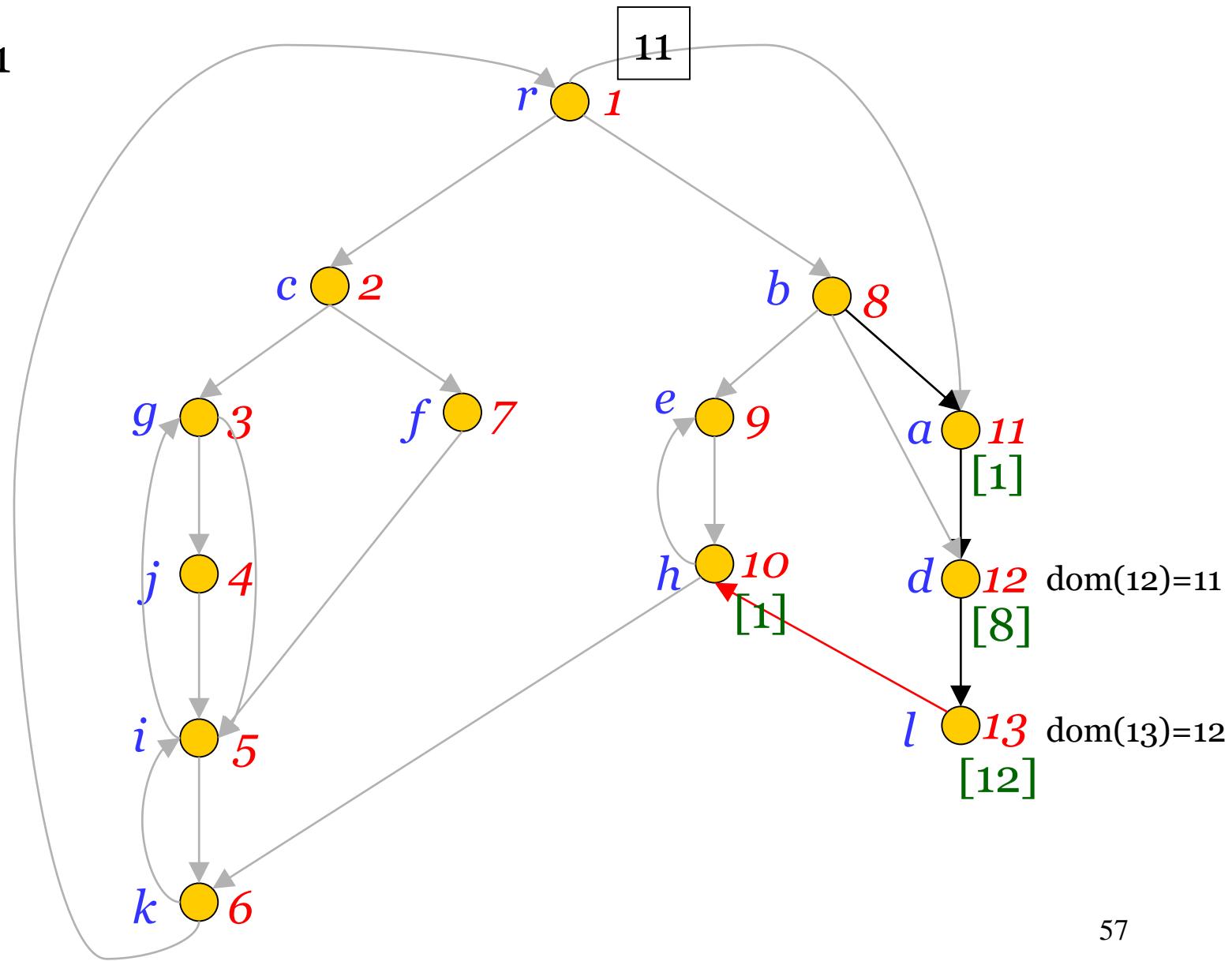
delete 12 from bucket(8)

eval(12) = 11



# The Lengauer-Tarjan Algorithm: Example

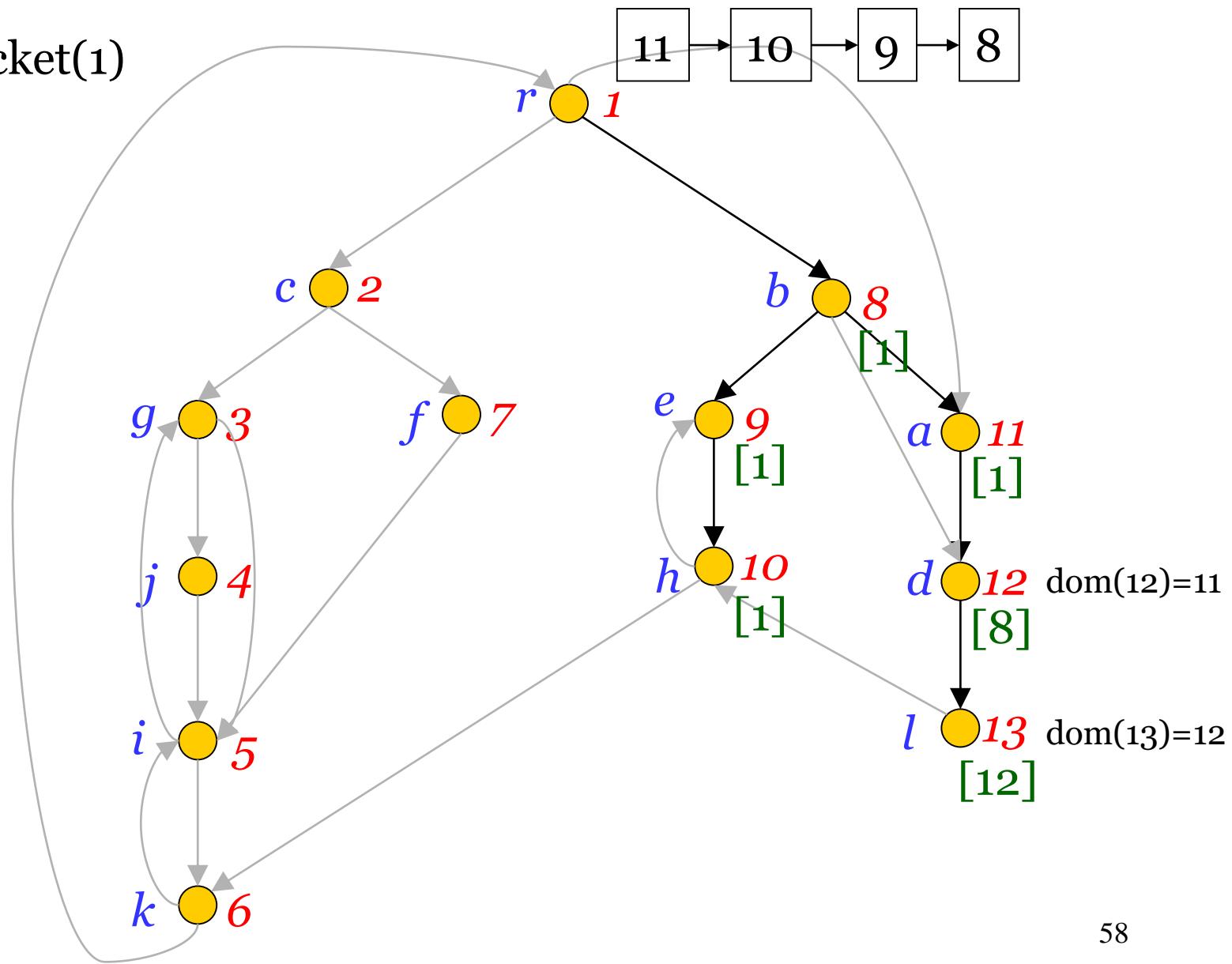
$\text{eval}(13) = 11$



# The Lengauer-Tarjan Algorithm: Example

add 8 to bucket(1)

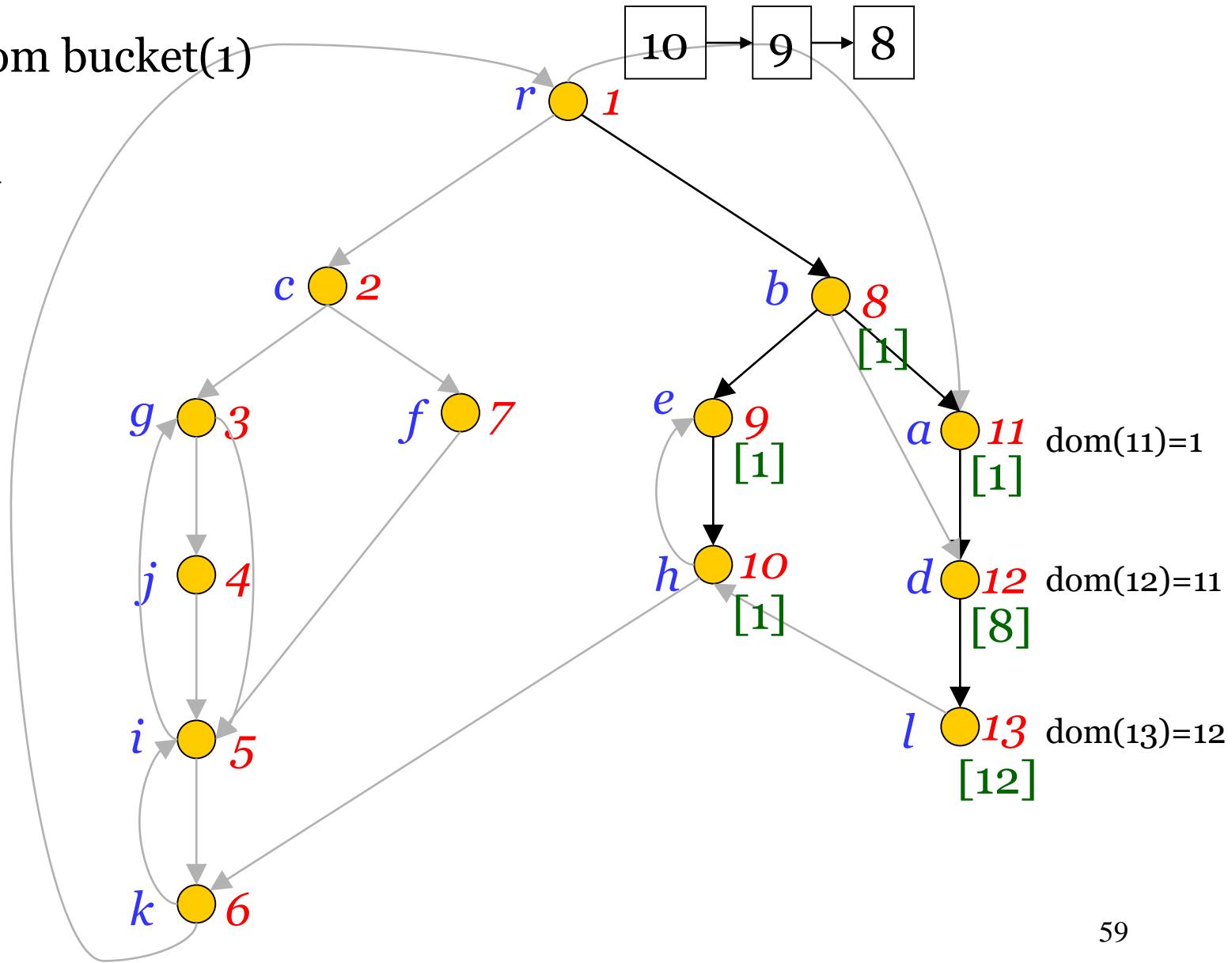
link(8)



# The Lengauer-Tarjan Algorithm: Example

delete 11 from bucket(1)

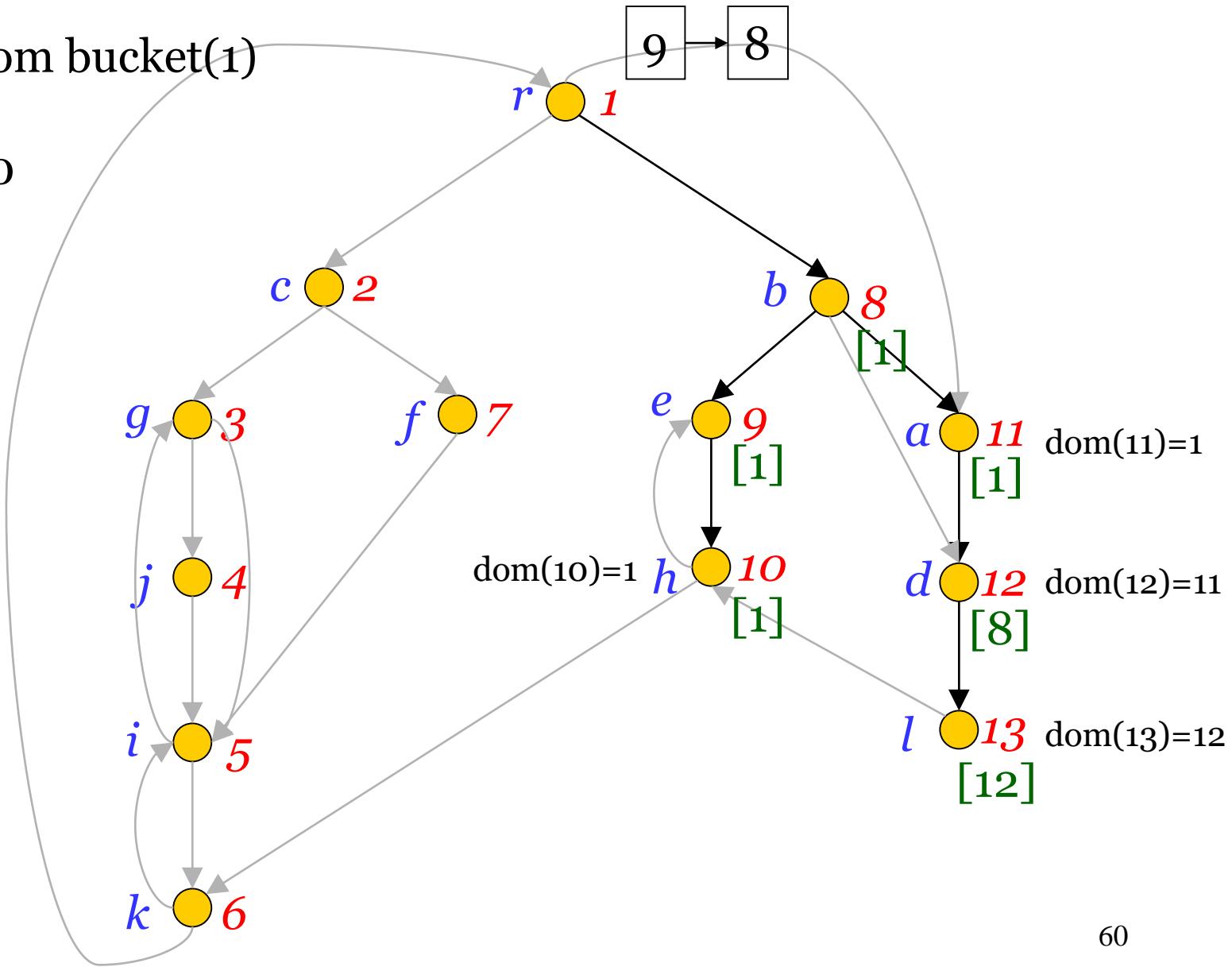
$\text{eval}(11) = 11$



# The Lengauer-Tarjan Algorithm: Example

delete 10 from bucket(1)

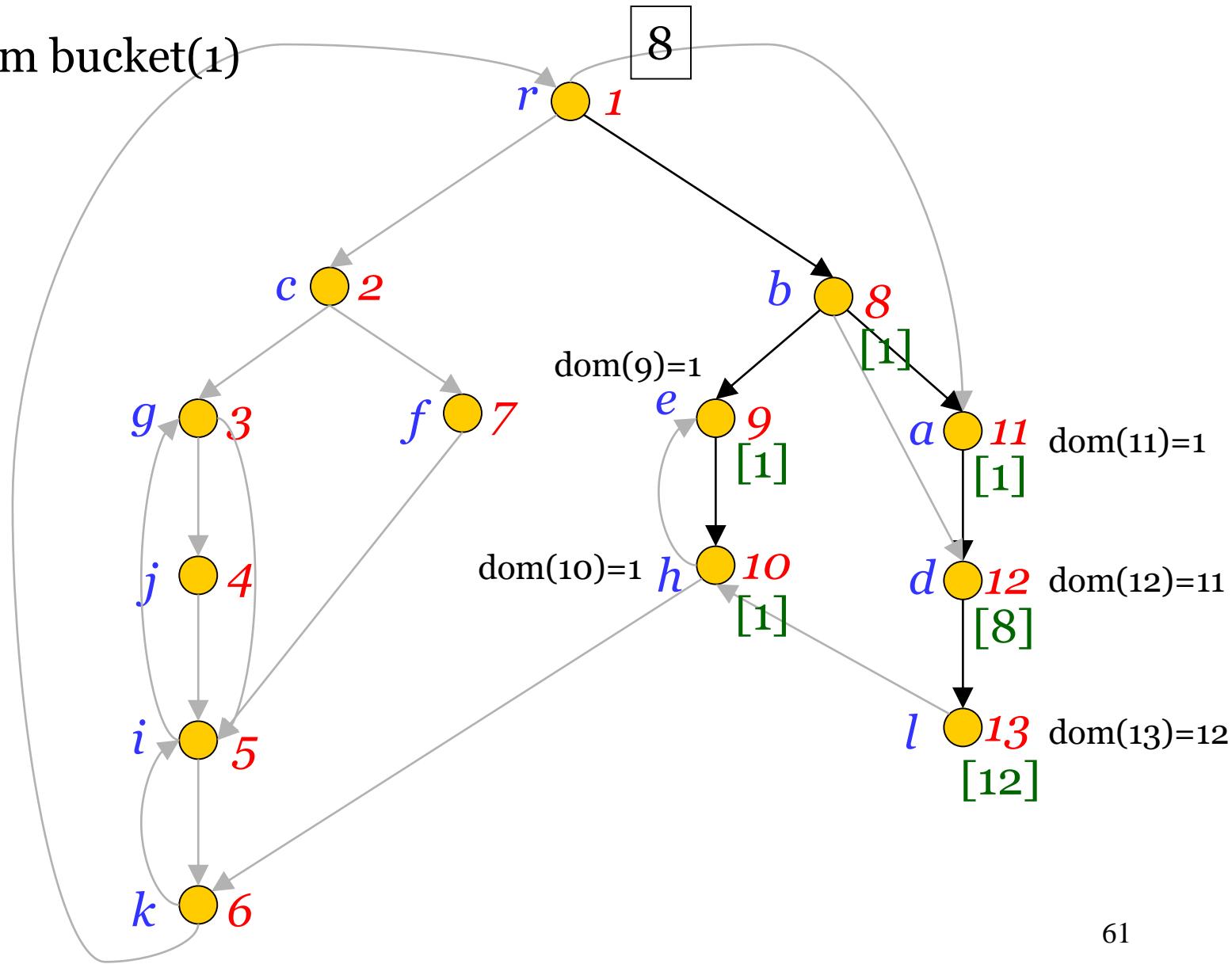
$\text{eval}(10) = 10$



# The Lengauer-Tarjan Algorithm: Example

delete 9 from bucket(1)

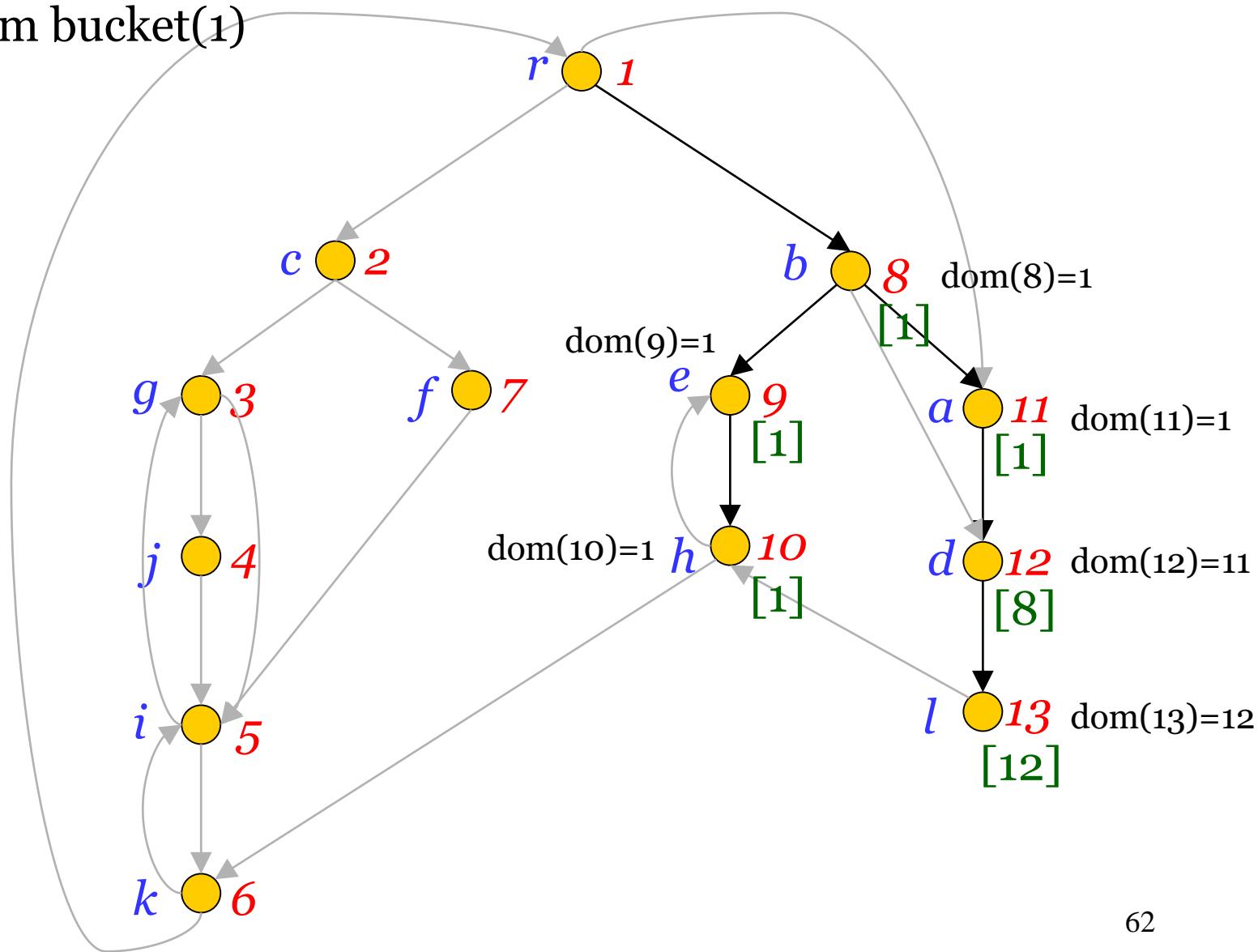
$\text{eval}(9) = 9$



# The Lengauer-Tarjan Algorithm: Example

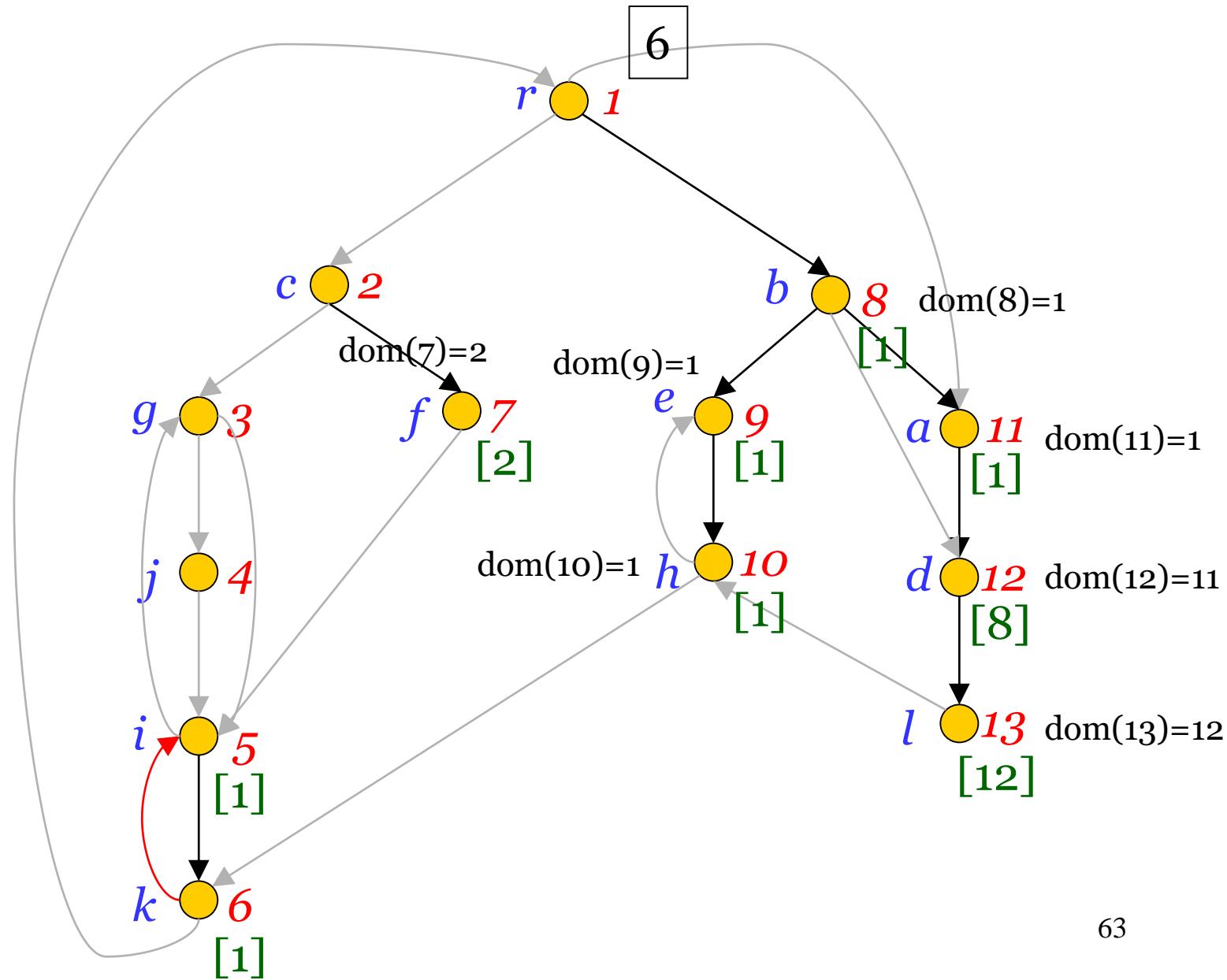
delete 8 from bucket(1)

$\text{eval}(8) = 8$



# The Lengauer-Tarjan Algorithm: Example

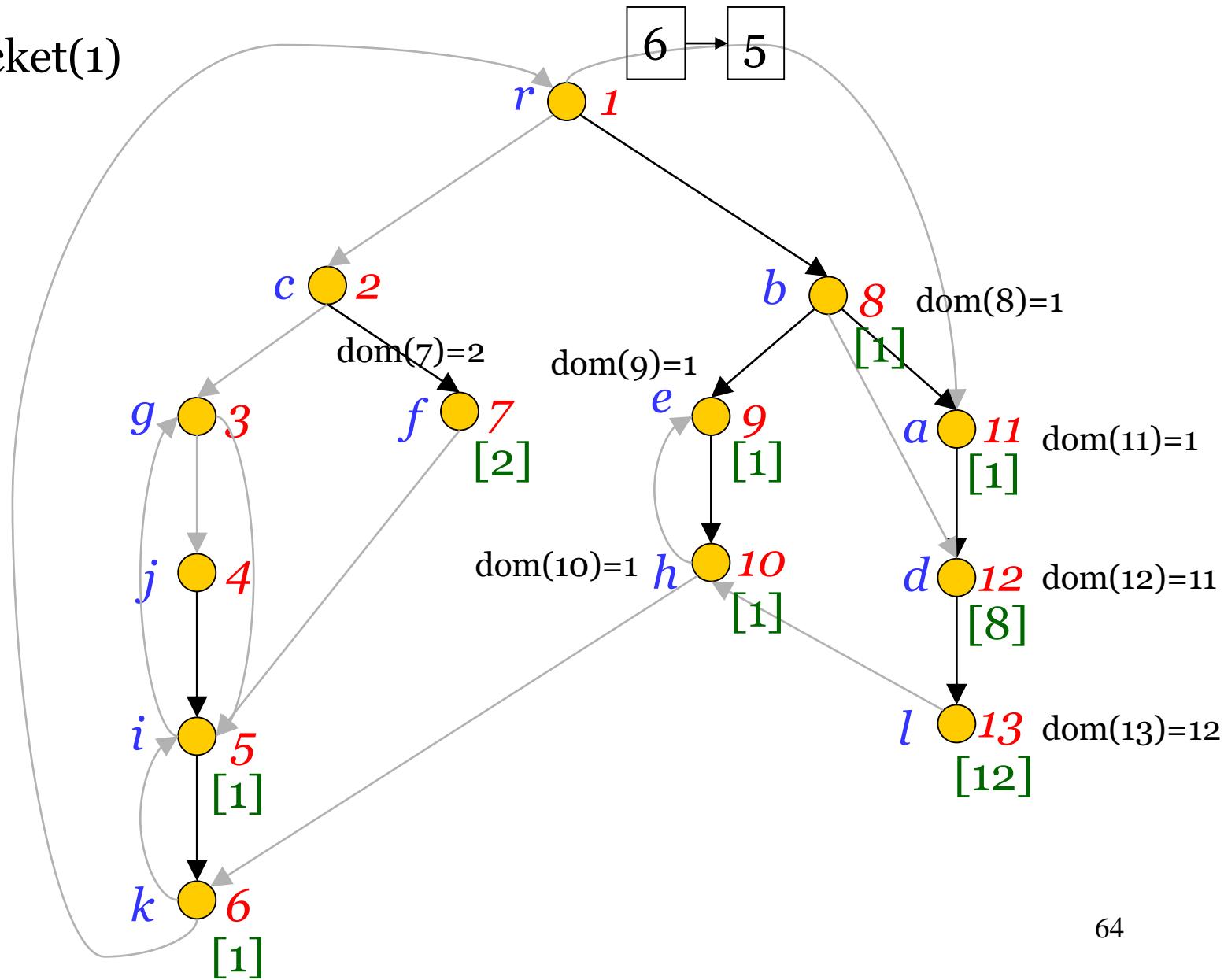
$\text{eval}(6) = 6$



# The Lengauer-Tarjan Algorithm: Example

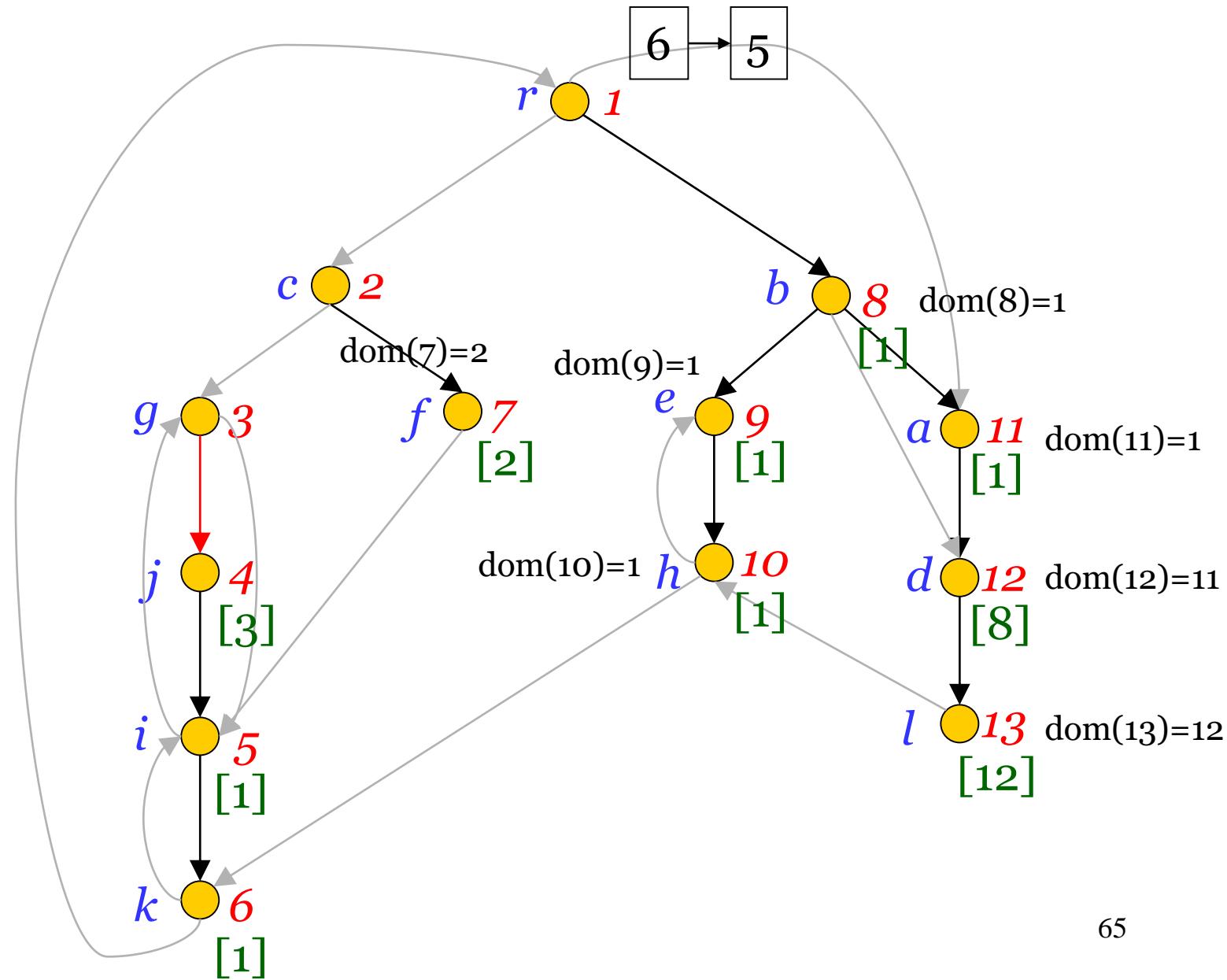
add 5 to bucket(1)

link(5)



# The Lengauer-Tarjan Algorithm: Example

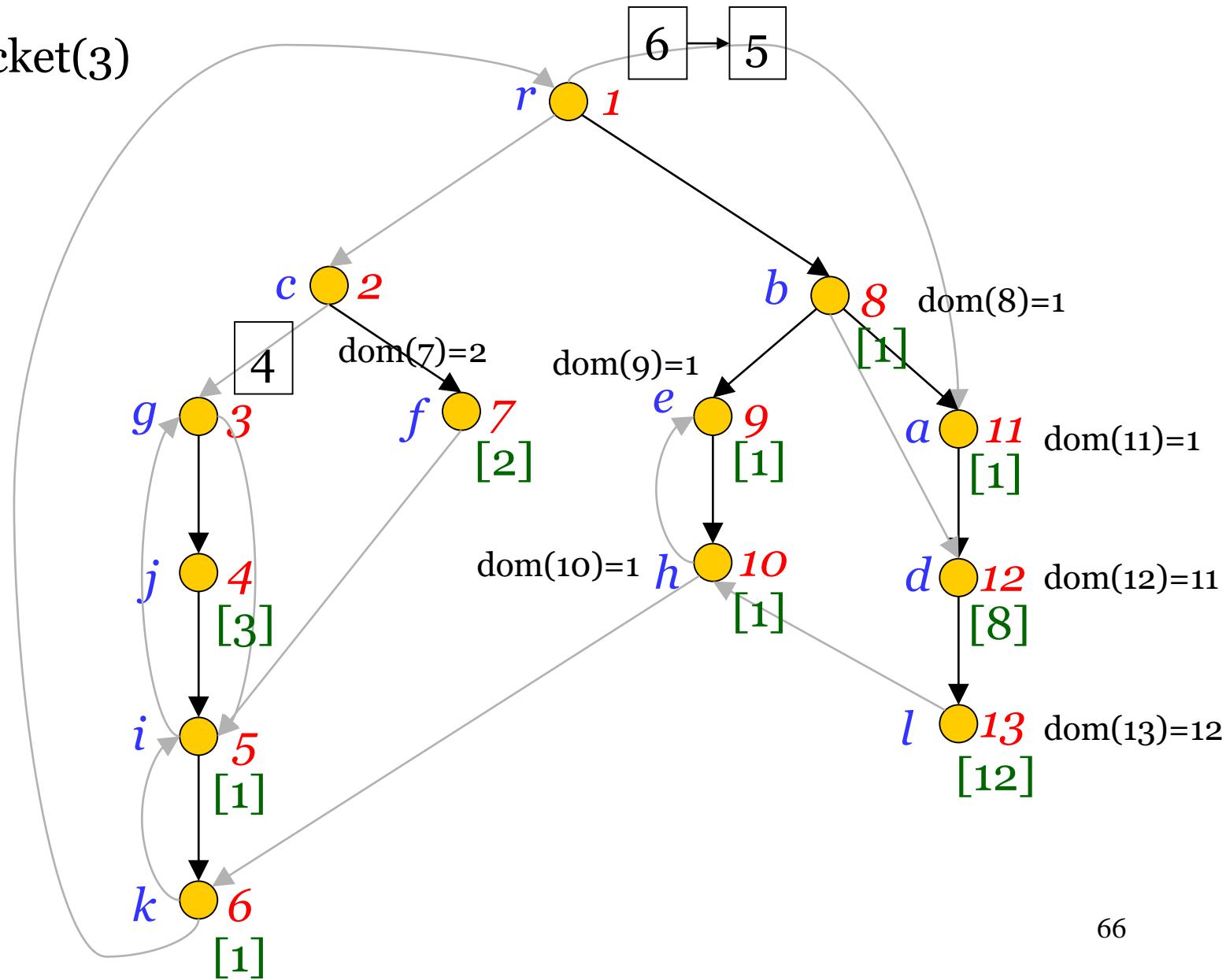
$\text{eval}(3) = 3$



# The Lengauer-Tarjan Algorithm: Example

add 4 to bucket(3)

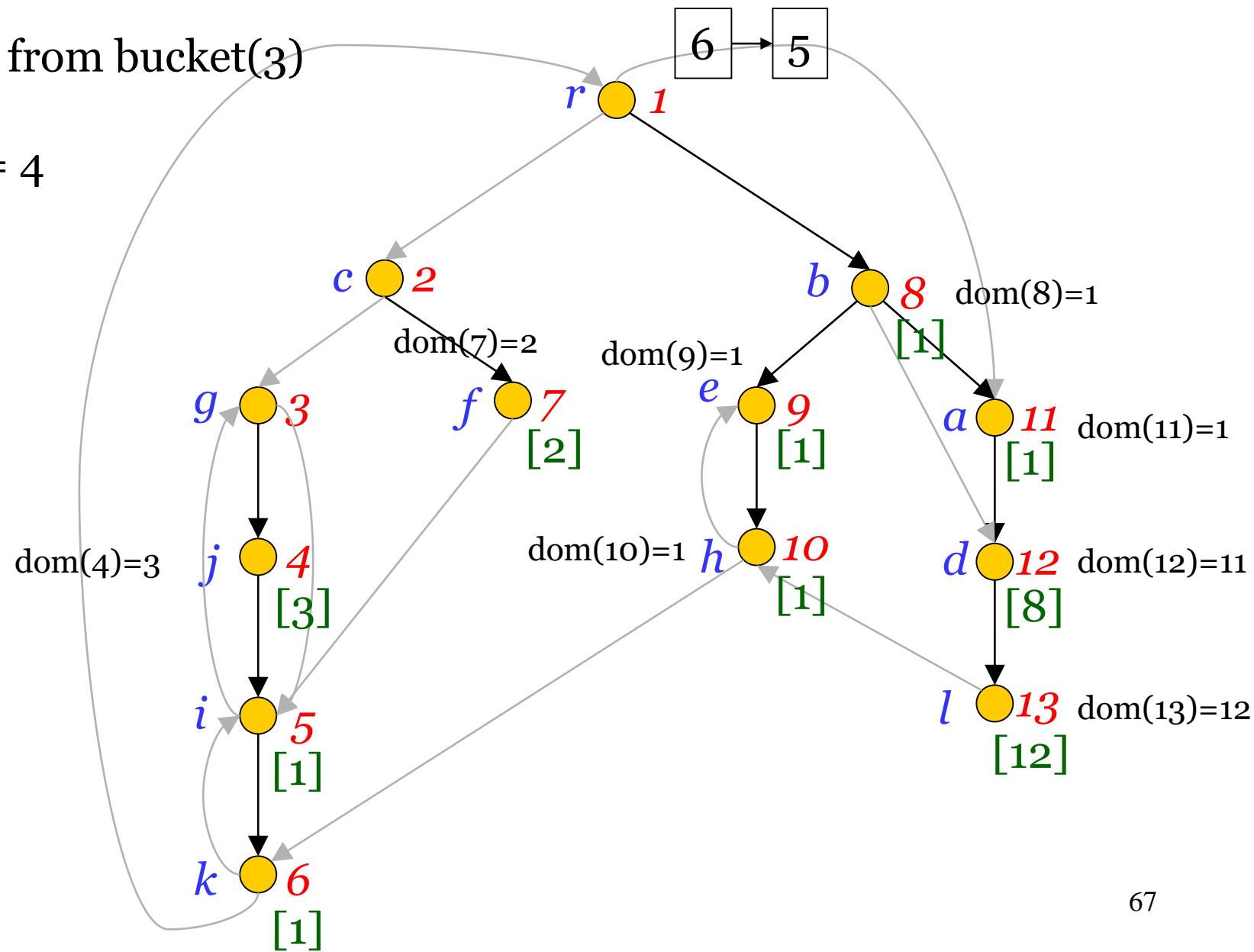
link(4)



# The Lengauer-Tarjan Algorithm: Example

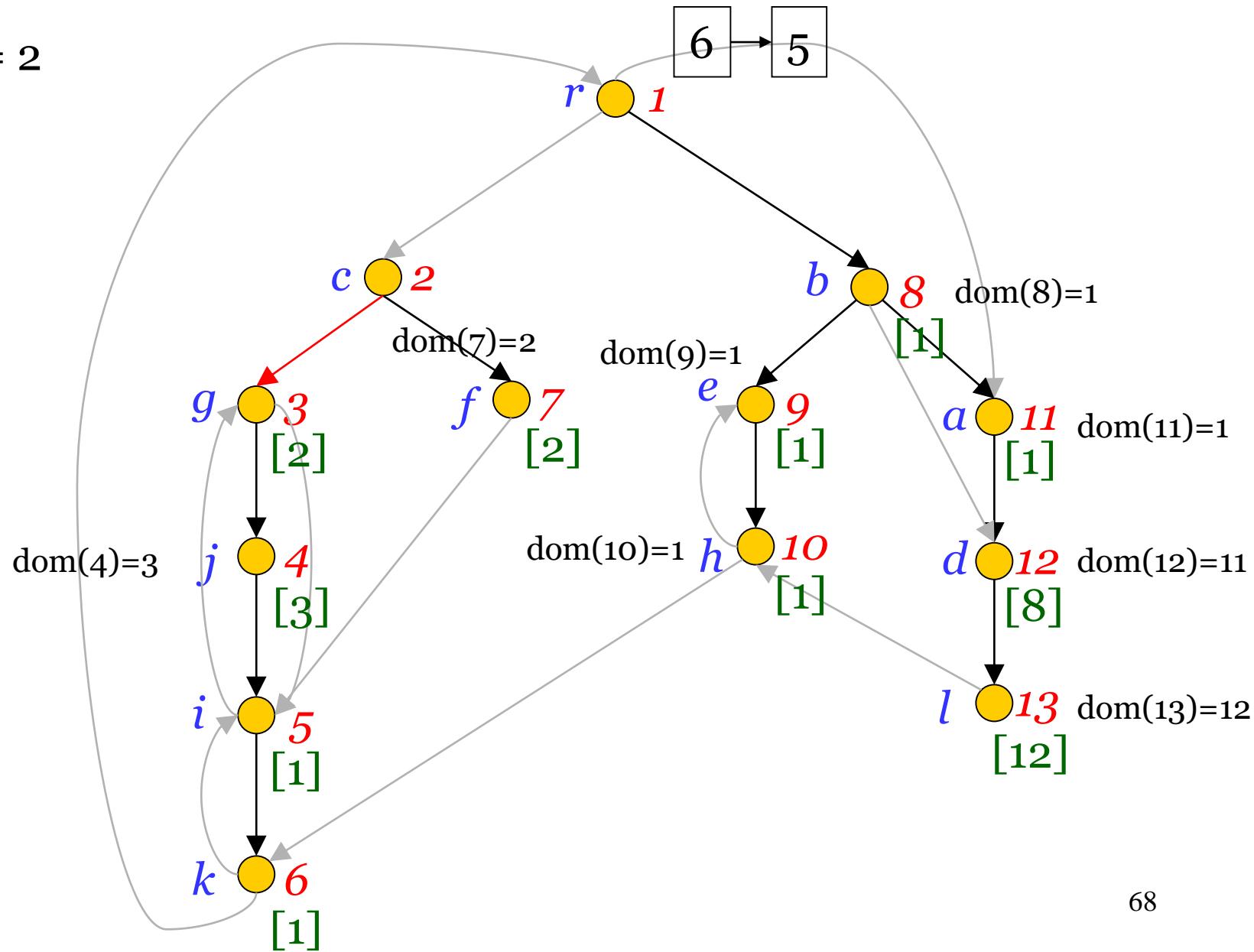
delete 4 from bucket(3)

$\text{eval}(4) = 4$



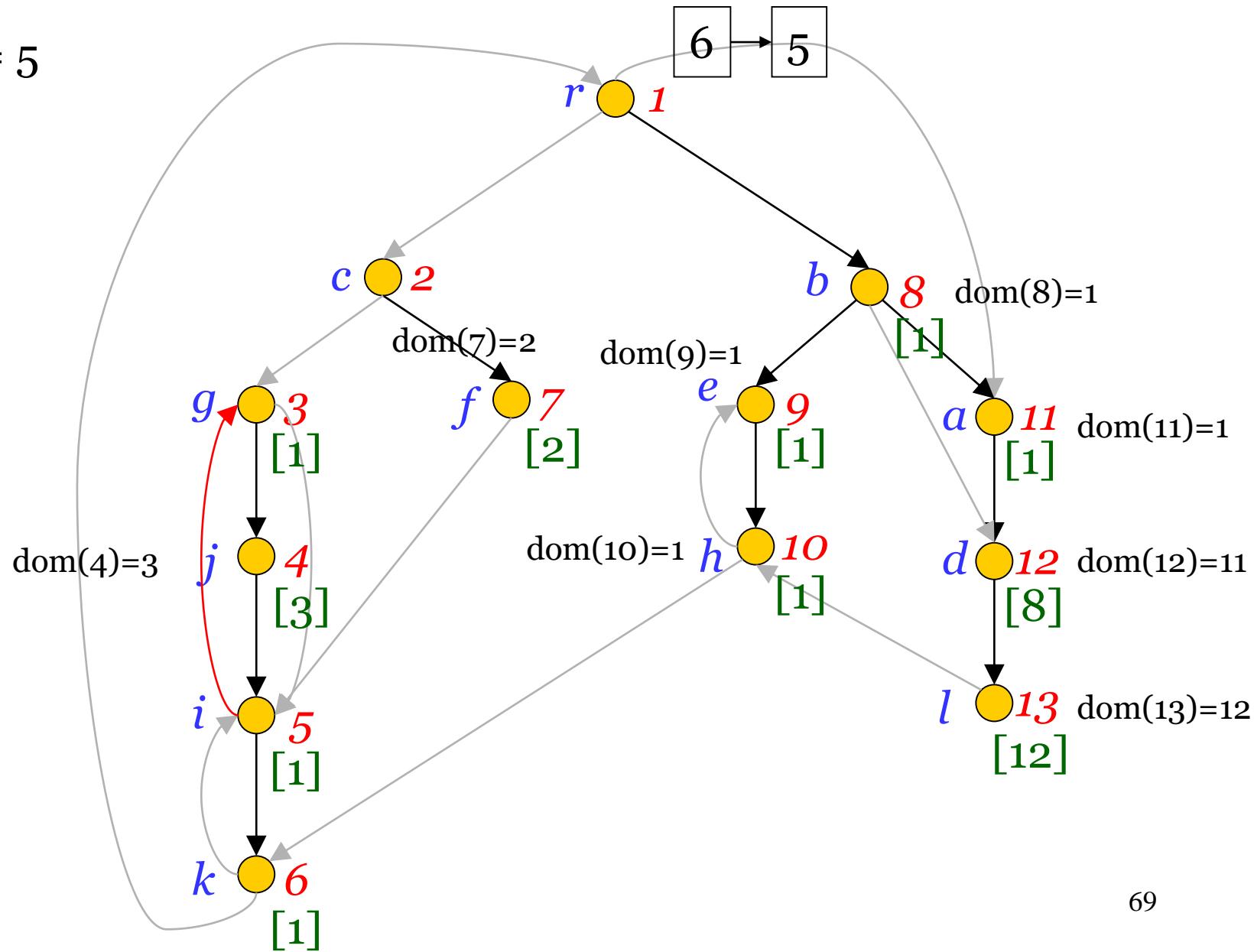
# The Lengauer-Tarjan Algorithm: Example

$\text{eval}(2) = 2$



# The Lengauer-Tarjan Algorithm: Example

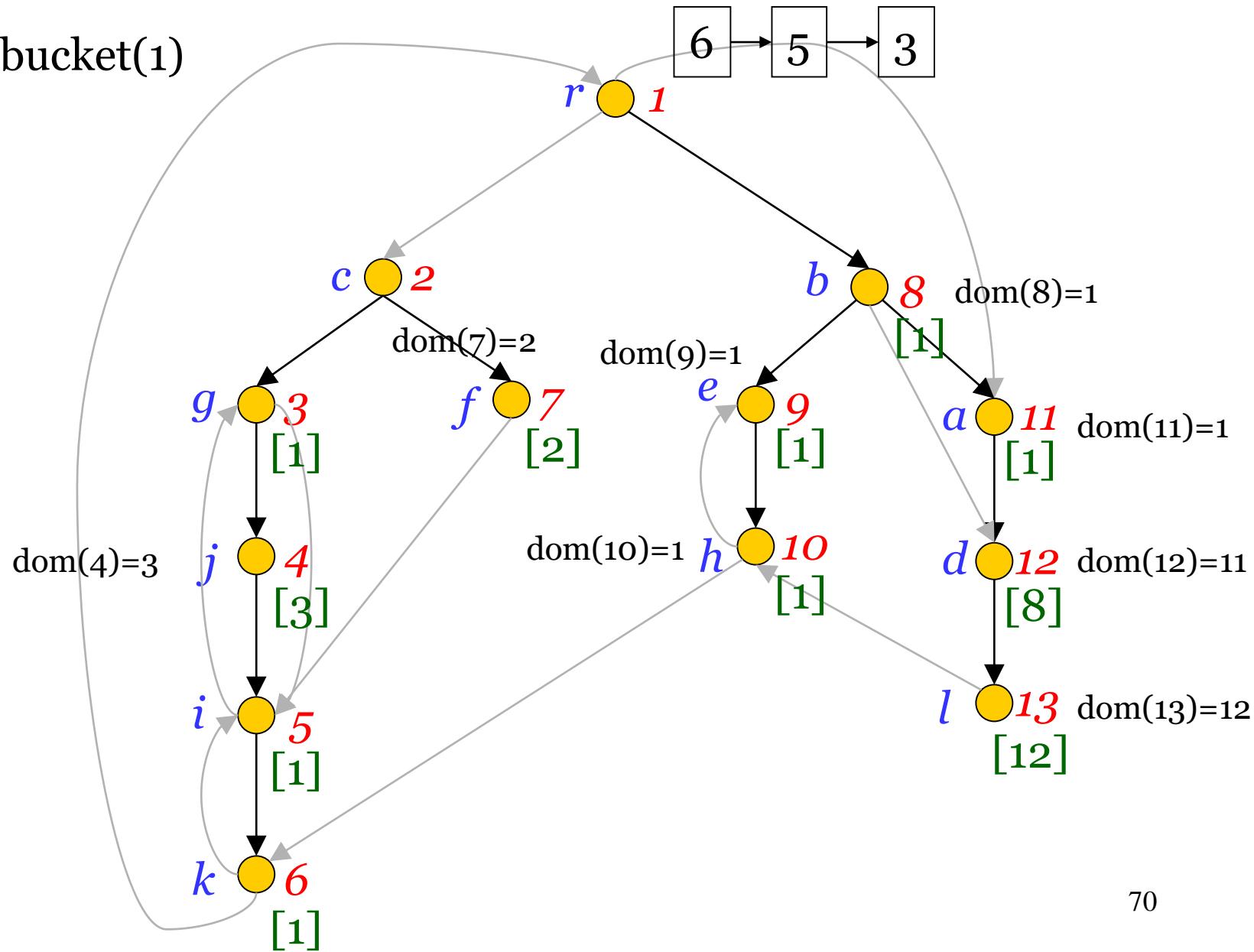
$\text{eval}(5) = 5$



# The Lengauer-Tarjan Algorithm: Example

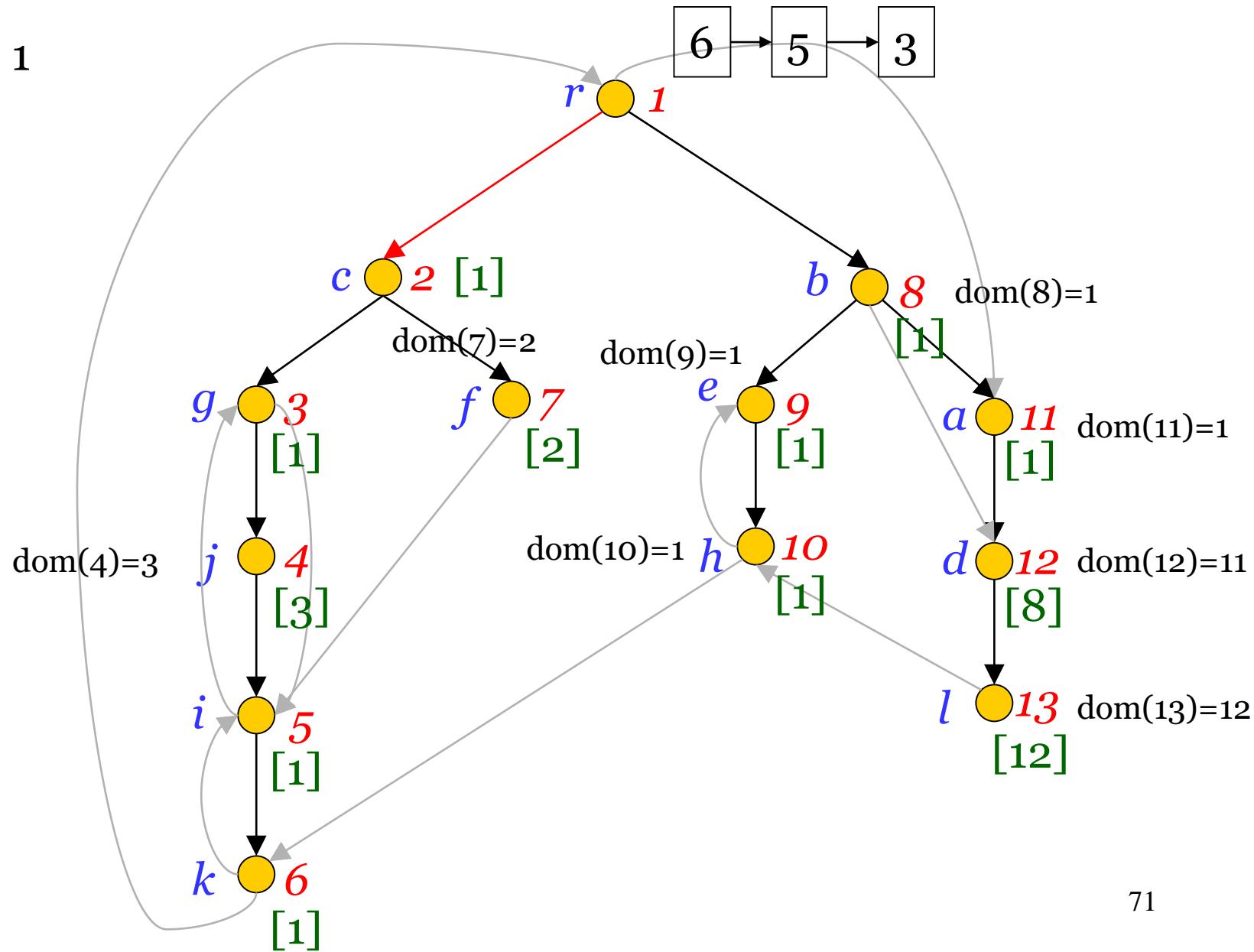
add 3 to bucket(1)

link(3)



# The Lengauer-Tarjan Algorithm: Example

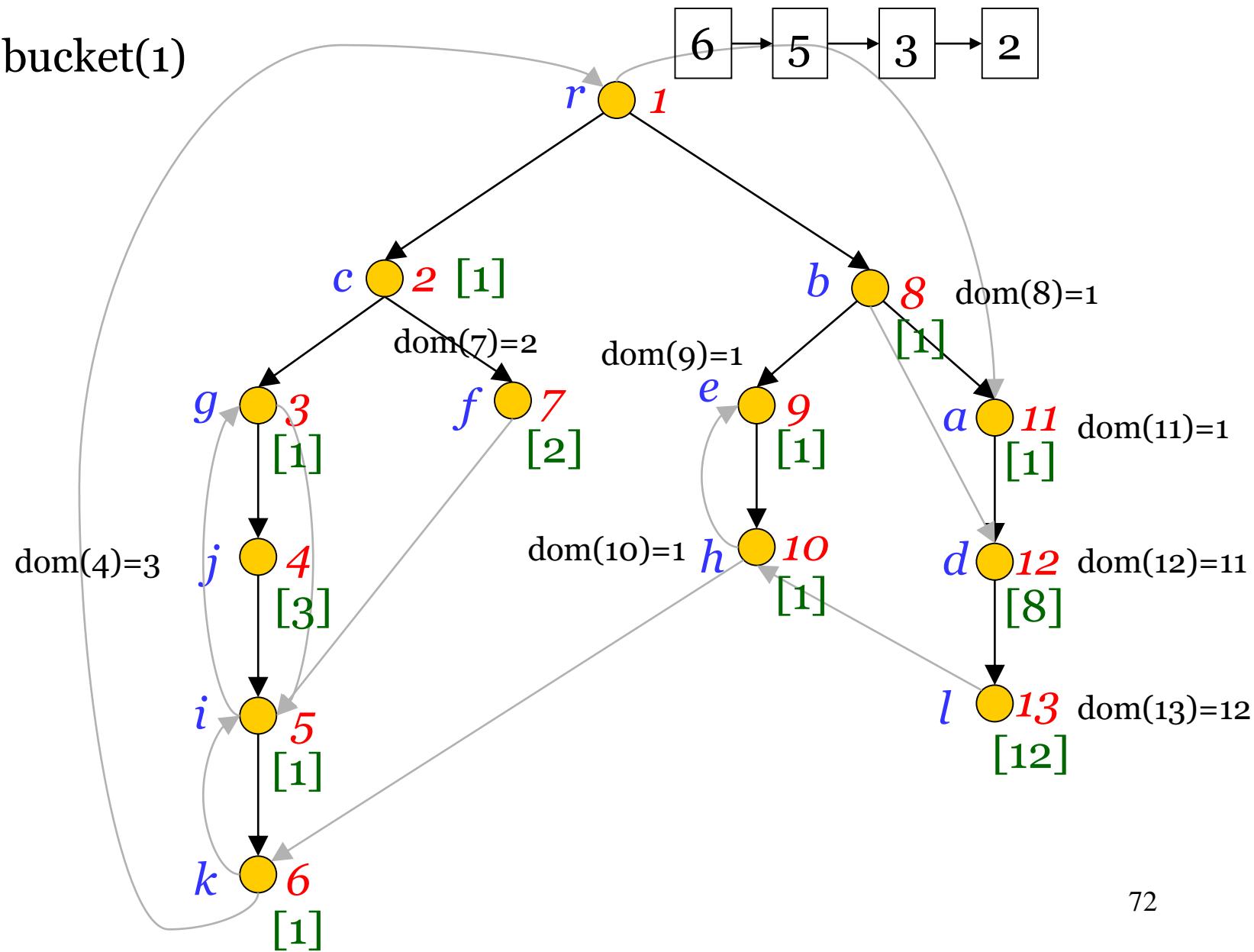
$\text{eval}(1) = 1$



# The Lengauer-Tarjan Algorithm: Example

add 2 to bucket(1)

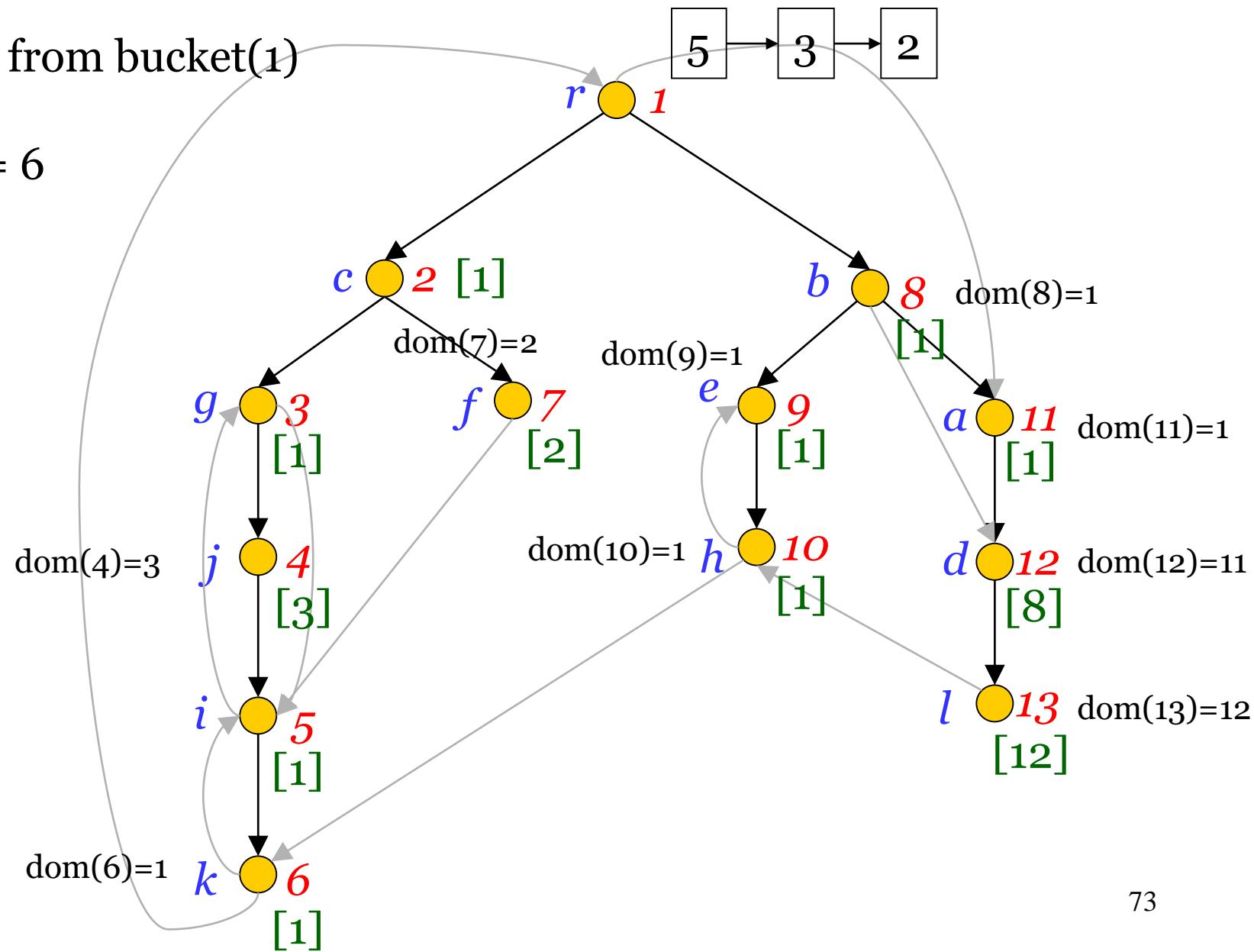
link(2)



# The Lengauer-Tarjan Algorithm: Example

delete 6 from bucket(1)

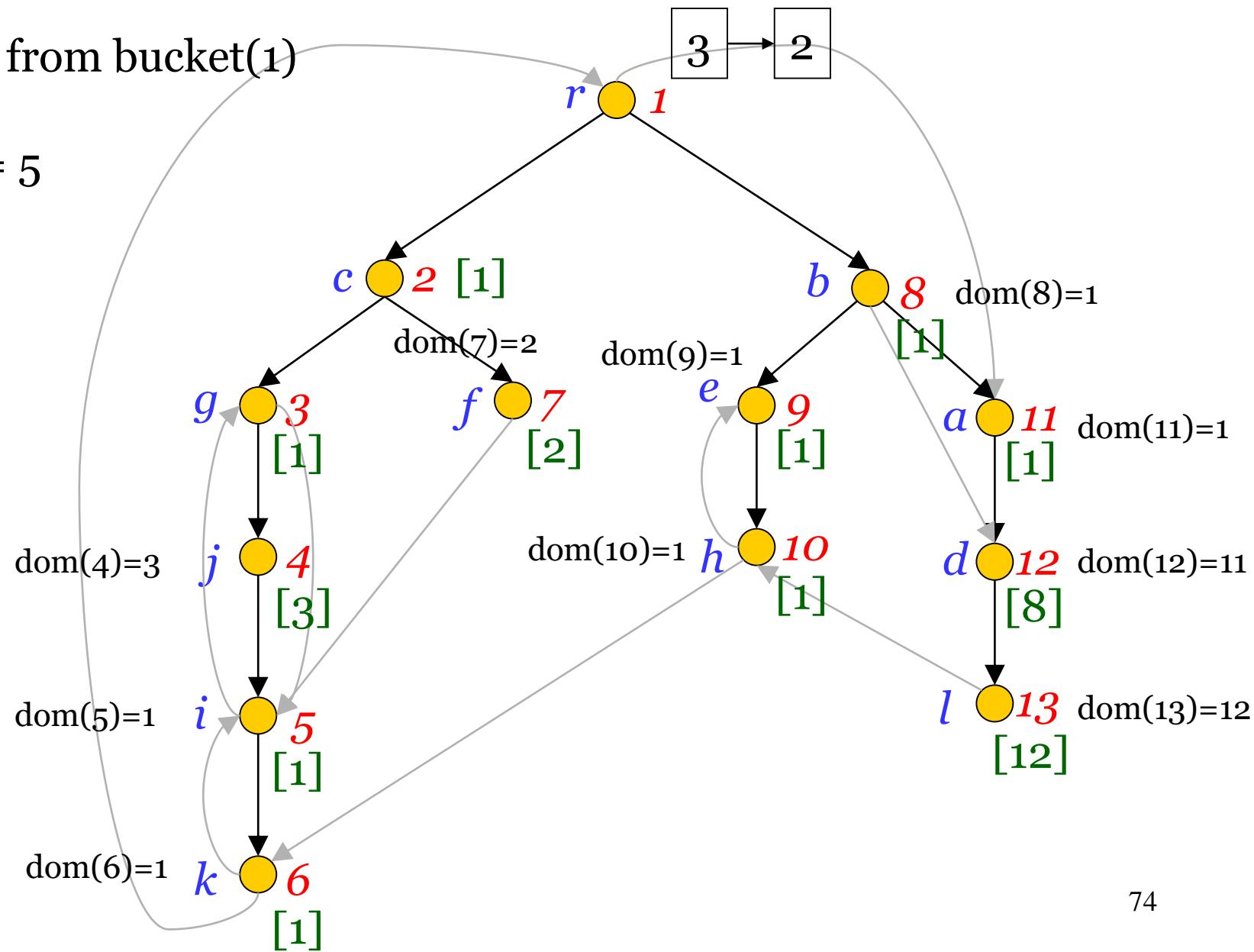
$\text{eval}(6) = 6$



# The Lengauer-Tarjan Algorithm: Example

delete 5 from bucket(1)

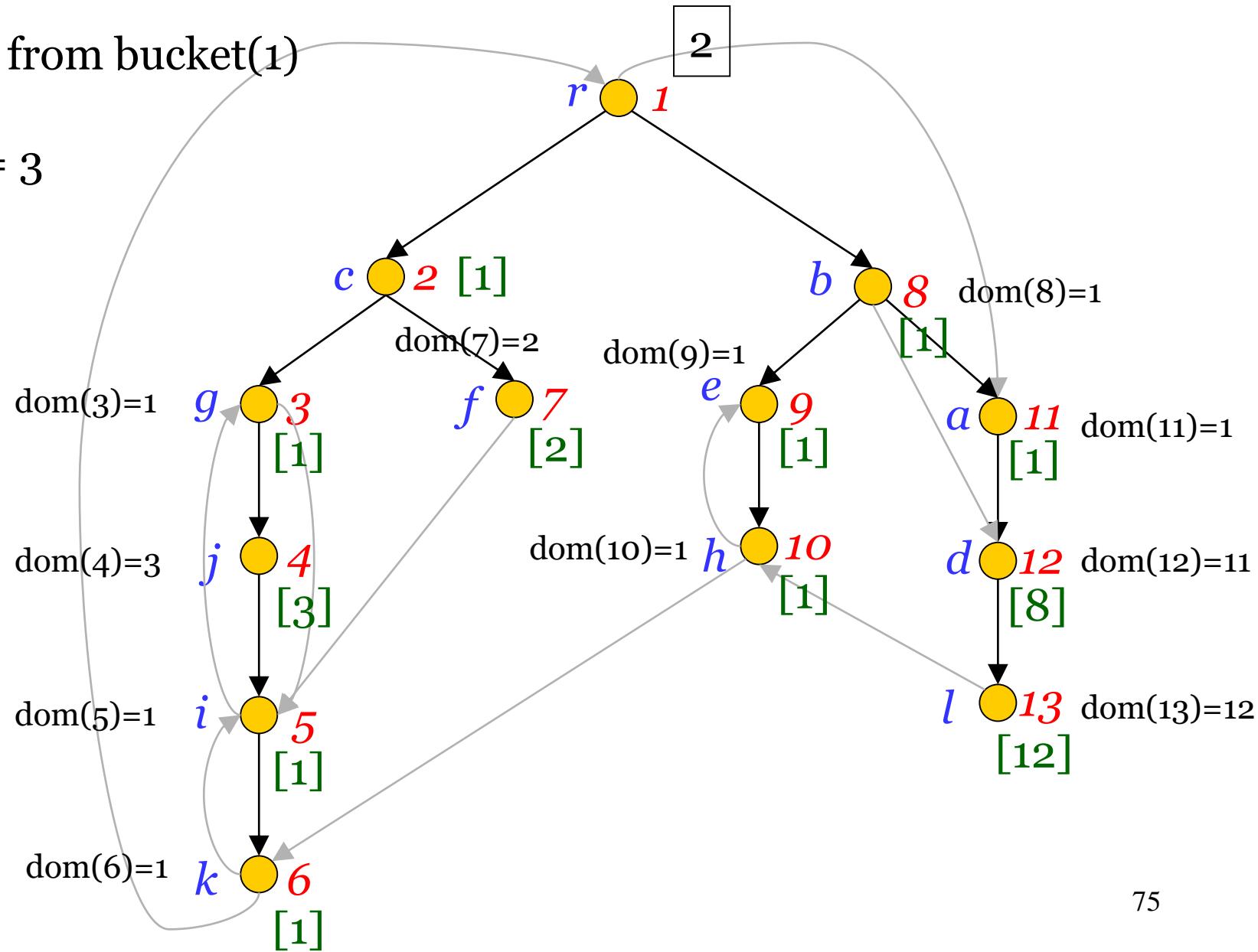
$\text{eval}(5) = 5$



# The Lengauer-Tarjan Algorithm: Example

delete 3 from bucket(1)

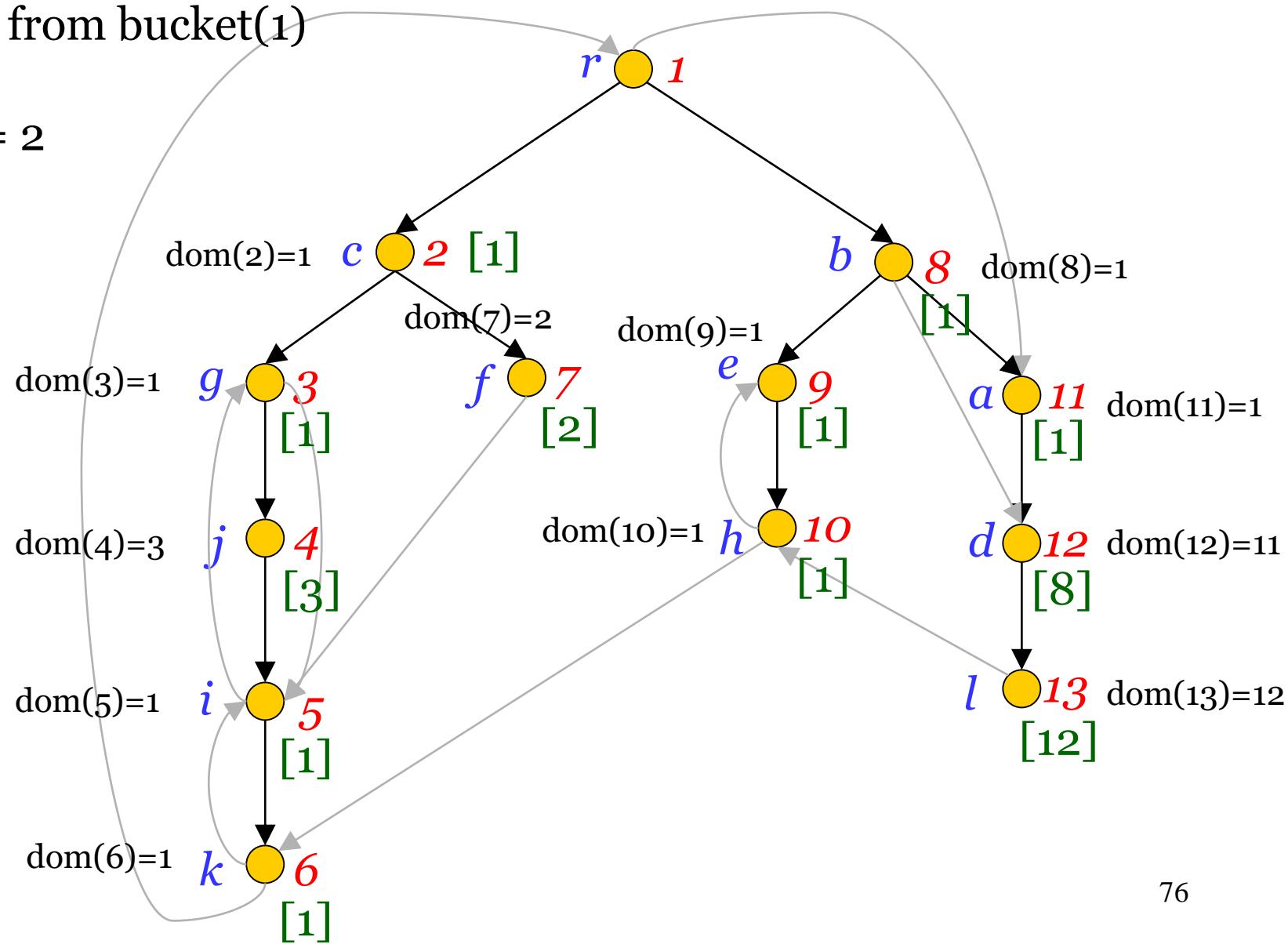
$\text{eval}(3) = 3$



# The Lengauer-Tarjan Algorithm: Example

delete 2 from bucket(1)

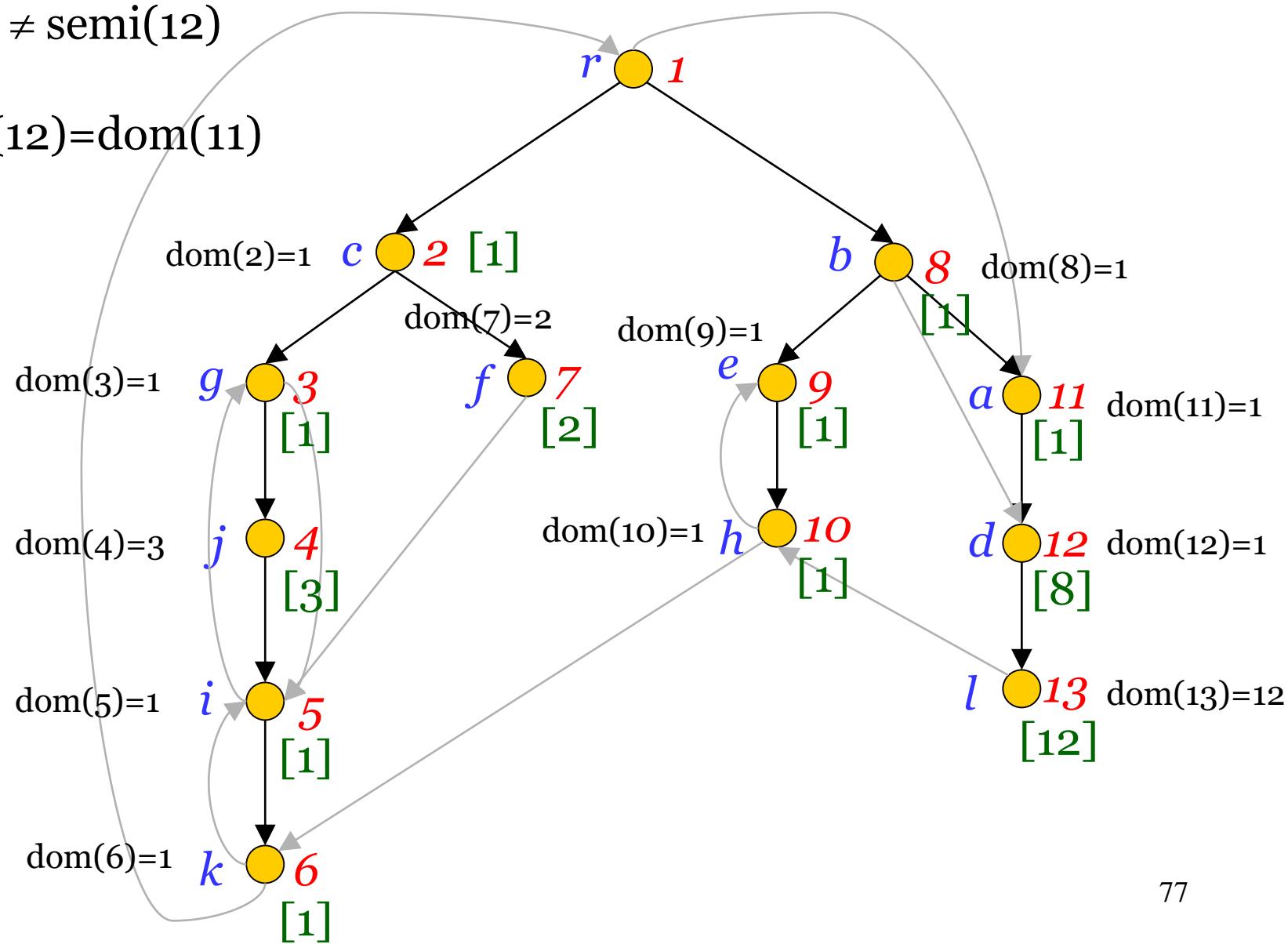
$\text{eval}(2) = 2$



# The Lengauer-Tarjan Algorithm: Example

$\text{dom}(12) \neq \text{semi}(12)$

set  $\text{dom}(12) = \text{dom}(11)$



# The Lengauer-Tarjan Algorithm

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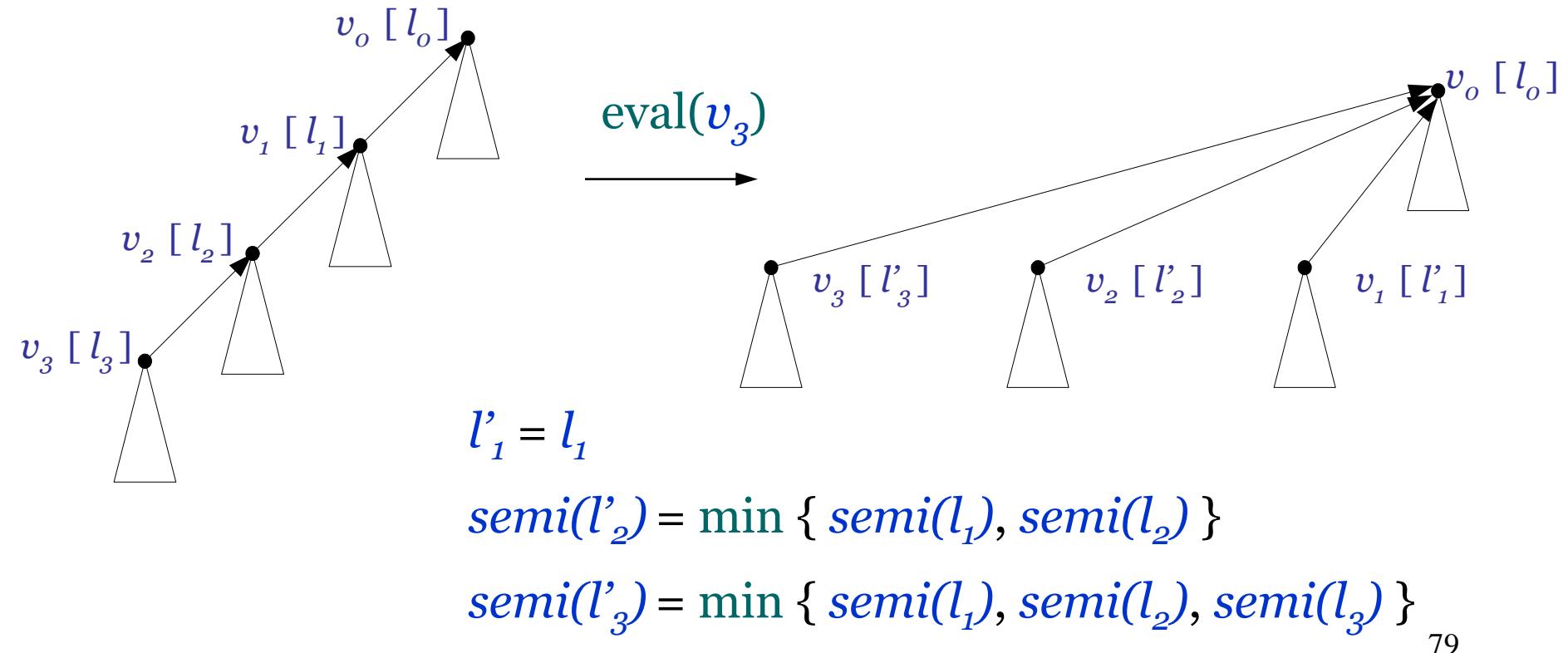
Running Time =  $O(n + m)$  + Time for  $n-1$  calls to `link()`

+ Time for  $m+n-1$  calls to `eval()`

# Data Structure for `link()` and `eval()`

---

We want to apply **Path Compression**:



# Data Structure for `link()` and `eval()`

---

We maintain a virtual forest  $\text{VF}$  such that:

1. For each  $T$  in  $F$  there is a corresponding  $VT$  in  $\text{VF}$  with the same vertices as  $T$ .
2. Corresponding trees  $T$  and  $VT$  have the same root with the same label.
3. If  $v$  is any vertex,  $\text{eval}(v, F) = \text{eval}(v, \text{VF})$ .

Representation:  $\text{ancestor}(v) = \text{parent of } v \text{ in } VT$ .

# Data Structure for `link()` and `eval()`

---

`eval( $v$ )`: Compress the path  $r^* \rightarrow v$  and return the label of  $v$ .

`link( $v, w$ )`: Make  $v$  the parent of  $w$ .

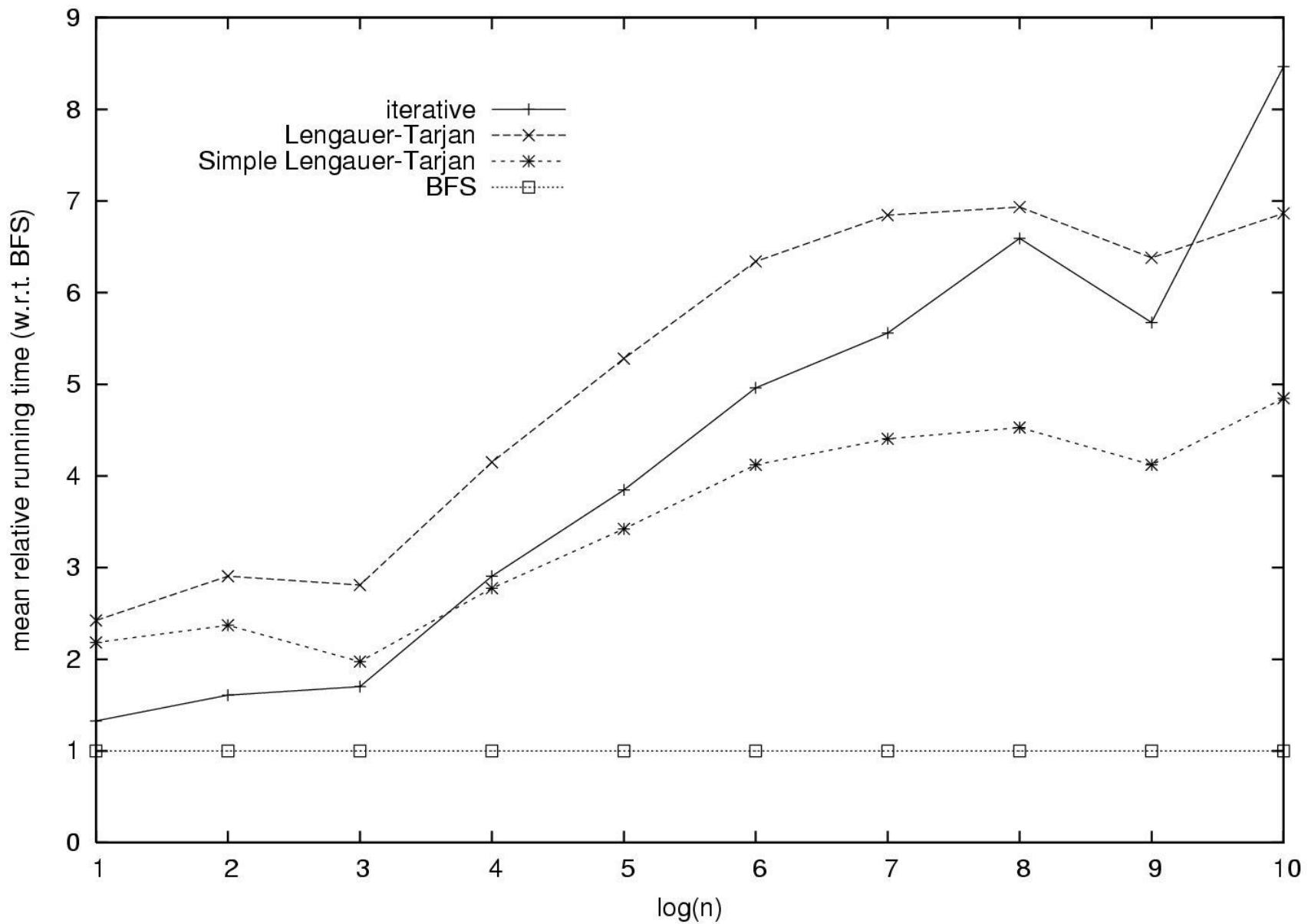
$VF$  satisfies Properties 1-3.

Time for  $n-1$  calls to `link()` +

Time for  $m+n-1$  calls to `eval()` =  $O(m \cdot \log_{2+\lfloor m/n \rfloor} n)$

# Experimental Results (Small Graphs)

---



# Experimental Results (Large Graphs)

---

GRAPH	<i>n</i>	<i>m</i>	density	height of dom. tree
NY	264346	733846	2.78	36
USA	23947347	58333344	2.44	241
ORACLE-16K	15678	48293	3.08	47
ORACLE-4M	4129899	14678595	3.55	1747
SAP-4M	4112973	12029328	2.92	6546
SAP-11M	11156904	36428546	3.27	4569
SAP-32M	32356252	81991806	2.53	7978
SAP-187M	186921678	556264024	2.98	180129504

GRAPH	SLT	LT	SNCA	IDFS	IBFS
NY	67.12	75.06	57.66	906.86	1301.80
USA	6544.00	7686.83	5868.11	393330.20	
ORACLE-16K	1.91	2.58	1.80	3.50	6.42
ORACLE-4M	1298.80	1511.77	1093.83	6827.96	327878.15
SAP-4M	1126.83	1445.78	965.85	1557961.15	1436782.57
SAP-11M	3035.53	3708.43	2610.60		
SAP-32M	7995.78	9843.50	6911.95		
SAP-187M	22422.59	28238.70	21281.76		

Running time in *ms*. Missing values correspond to execution time >1 h. 83

# The Lengauer-Tarjan Algorithm: Correctness

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**Lemma 1:**  $\forall v, w$  such that  $v \leq w$ , any path from  $v$  to  $w$  contains a common ancestor of  $v$  and  $w$  in  $T$ .

---

Follows from **Property 1**

**Lemma 2:** For any  $w \neq r$ ,  $idom(w)$  is an ancestor of  $w$  in  $T$ .

---

$idom(w)$  is contained in every path from  $r$  to  $w$

# The Lengauer-Tarjan Algorithm: Correctness

---

**Lemma 3:** For any  $w \neq r$ ,  $sdom(w)$  is an ancestor of  $w$  in  $T$ .

---

- $(parent(w), w)$  is an SDOM-path  $\Rightarrow sdom(w) \leq parent(w)$ .
- SDOM-path  $P = (v_0 = sdom(w), v_1, \dots, v_k = w)$ ;

Lemma 1  $\Rightarrow$  some  $v_i$  is a common ancestor  
of  $sdom(w)$  and  $w$ .

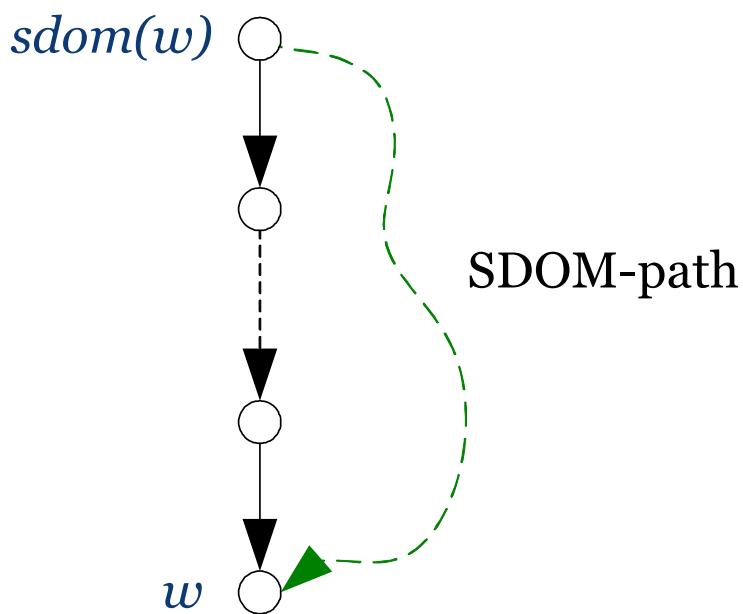
We must have  $v_i \leq sdom(w) \Rightarrow v_i = sdom(w)$ .

# The Lengauer-Tarjan Algorithm: Correctness

---

**Lemma 4:** For any  $w \neq r$ ,  $\text{idom}(w)$  is an ancestor of  $sdom(w)$  in  $T$ .

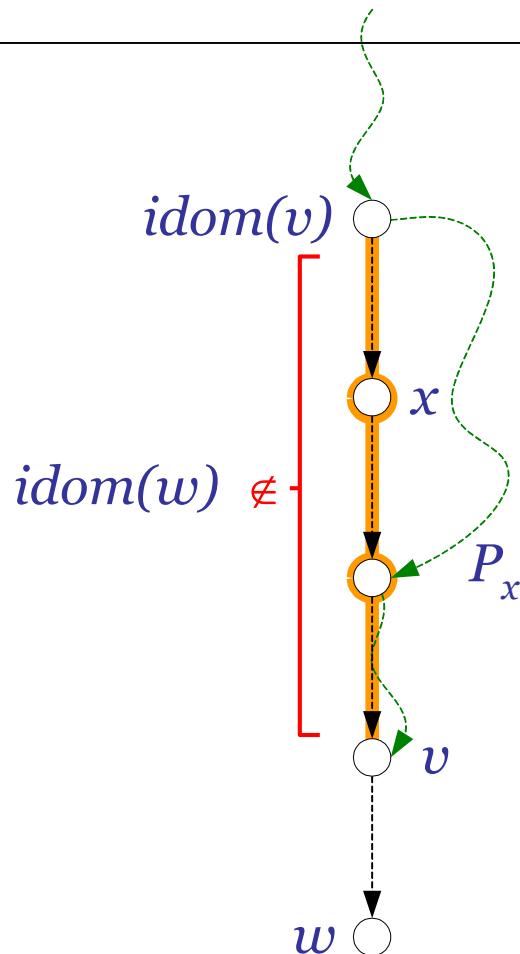
---



The SDOM-path from  $sdom(w)$  to  $w$  avoids the proper ancestors of  $w$  that are proper descendants of  $sdom(w)$ .

# The Lengauer-Tarjan Algorithm: Correctness

Lemma 5: Let  $v, w$  satisfy  $v^* \rightarrow w$ . Then  $v^* \rightarrow idom(w)$  or  $idom(w)^* \rightarrow idom(v)$ .



For each  $x$  that satisfies

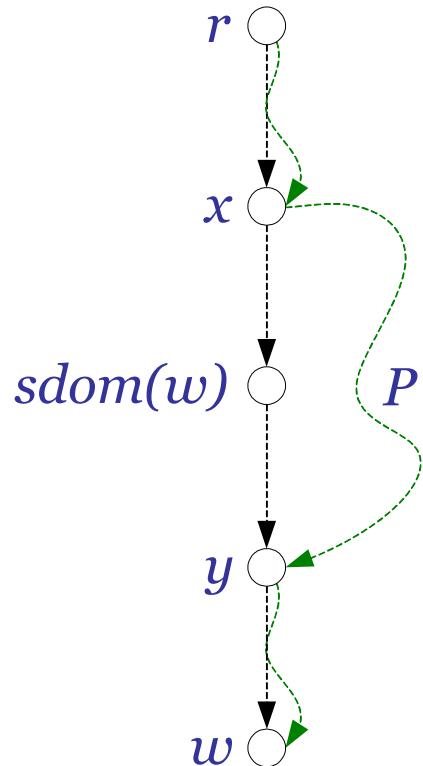
$$idom(v) \xrightarrow{+} x \xrightarrow{+} v$$

there is a path  $P_x$  from  $r$  to  $v$  that avoids  $x$ .

$P_x \cdot v^* \rightarrow w$  is a path from  $r$  to  $w$  that avoids  $x \Rightarrow idom(w) \neq x$ .

# The Lengauer-Tarjan Algorithm: Correctness

**Theorem 2:** Let  $w \neq r$ . If  $sdom(u) \geq sdom(w)$  for every  $u$  that satisfies  $sdom(w) \xrightarrow{+} u \xrightarrow{*} w$  then  $idom(w) = sdom(w)$ .



Suppose for contradiction  $sdom(w) \notin Dom(w)$

$\Rightarrow \exists$  path  $P$  from  $r$  to  $w$  that avoids  $sdom(w)$ .

$x$  = last vertex  $\in P$  such that  $x < sdom(w)$

$y$  = first vertex  $\in P \cap sdom(w) \xrightarrow{*} w$

$Q$  = part of  $P$  from  $x$  to  $y$

Lemma 1  $\Rightarrow y < u, \forall u \in Q - \{x,y\}$

$\Rightarrow sdom(y) < sdom(w)$ .

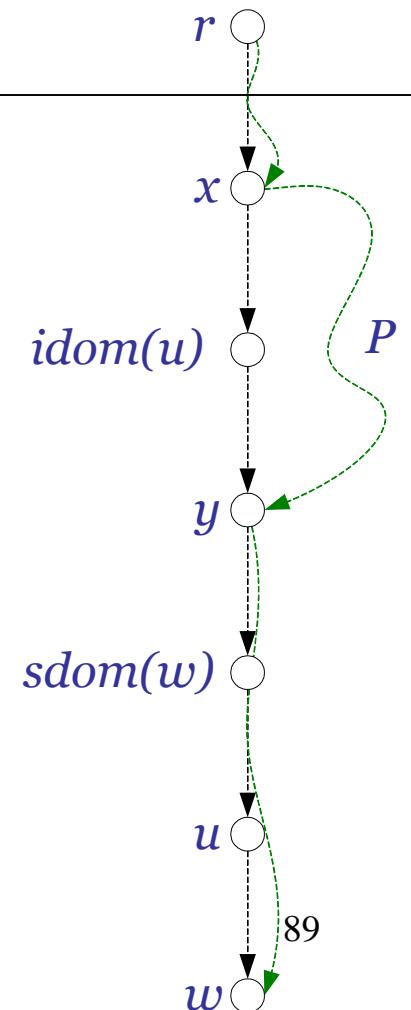
# The Lengauer-Tarjan Algorithm: Correctness

**Theorem 3:** Let  $w \neq r$  and let  $u$  be any vertex for which  $sdom(u)$  is minimum among the vertices  $u$  that satisfy  $sdom(w) \xrightarrow{*} u \xrightarrow{*} w$ . Then  $idom(u) = idom(w)$ .

Lemma 4 and Lemma 5  $\Rightarrow idom(w) \xrightarrow{*} idom(u)$ .

Suppose for contradiction  $idom(u) \neq idom(w)$ .

$\Rightarrow \exists$  path  $P$  from  $r$  to  $w$  that avoids  $idom(u)$ .



# The Lengauer-Tarjan Algorithm: Correctness

**Theorem 3:** Let  $w \neq r$  and let  $u$  be any vertex for which  $sdom(u)$  is minimum among the vertices  $u$  that satisfy  $sdom(w) \xrightarrow{*} u \xrightarrow{*} w$ . Then  $idom(u) = idom(w)$ .

$x$  = last vertex  $\in P$  such that  $x < idom(u)$ .

$y$  = first vertex  $\in P \cap idom(u) \xrightarrow{*} w$ .

$Q$  = part of  $P$  from  $x$  to  $y$ .

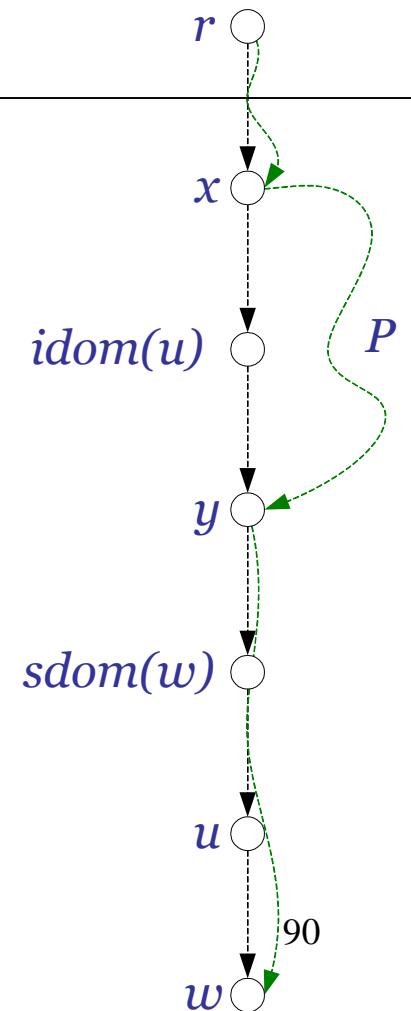
Lemma 1  $\Rightarrow y < u, \forall u \in Q - \{x,y\}$

$\Rightarrow sdom(y) < idom(u) \leq sdom(u)$ .

Therefore  $y \neq v$  for any  $v$  that satisfies

$idom(u) \xrightarrow{*} v \xrightarrow{*} u$ .

But  $y$  cannot be an ancestor of  $idom(u)$ .



# The Lengauer-Tarjan Algorithm: Correctness

---

From Theorem 2 and Theorem 3 we have  $sdom \Rightarrow idom$ :

**Corollary 1:** Let  $w \neq r$  and let  $u$  be any vertex for which  $sdom(u)$  is minimum among the vertices  $u$  that satisfy  $sdom(w) \xrightarrow{+} u \xrightarrow{*} w$ . Then  $idom(w) = sdom(w)$ , if  $sdom(w) = sdom(u)$  and  $idom(w) = idom(u)$  otherwise.

We still need a method to compute  $sdom$ .

# The Lengauer-Tarjan Algorithm: Correctness

---

**Theorem 4:** For any  $w \neq r$ ,

$$sdom(w) = \min (\{v \mid (v, w) \in E \text{ and } v < w\} \cup \\ \{sdom(u) \mid u > w \text{ and } \exists (v, w) \in E \text{ such that } u * \rightarrow v\}).$$

---

Let  $x = \min (\{v \mid (v, w) \in E \text{ and } v < w\} \cup \\ \{sdom(u) \mid u > w \text{ and } \exists (v, w) \in E \text{ such that } u * \rightarrow v\}).$

We first show  $sdom(w) \leq x$  and then  $sdom(w) \geq x$ .

# The Lengauer-Tarjan Algorithm: Correctness

---

**Theorem 4:** For any  $w \neq r$ ,

$$sdom(w) = \min (\{v \mid (v, w) \in E \text{ and } v < w\} \cup \\ \{sdom(u) \mid u > w \text{ and } \exists (v, w) \in E \text{ such that } u * \rightarrow v\}).$$

---

- $sdom(w) \leq x$

Assume  $x = v$  such that  $(v, w) \in E$  and  $v < w \Rightarrow sdom(w) \leq x$ .

# The Lengauer-Tarjan Algorithm: Correctness

**Theorem 4:** For any  $w \neq r$ ,

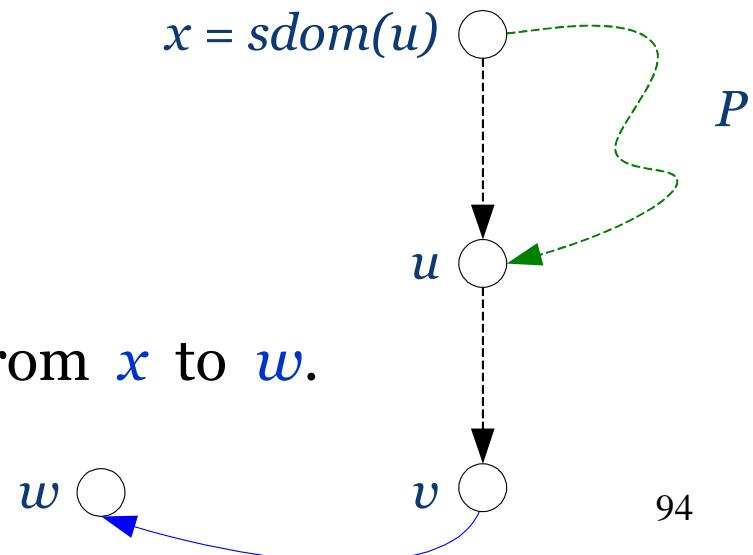
$$sdom(w) = \min (\{v \mid (v, w) \in E \text{ and } v < w\} \cup \{sdom(u) \mid u > w \text{ and } \exists (v, w) \in E \text{ such that } u * \rightarrow v\}).$$

- $sdom(w) \leq x$

Assume  $x = sdom(u)$  such that  $u > w$  and  $(v, w) \in E$  for some descendant  $v$  of  $u$  in  $T$ .

$P = \text{SDOM-path from } x \text{ to } u \Rightarrow$

$P \cdot u * \rightarrow v \cdot (v, w)$  is an SDOM-path from  $x$  to  $w$ .



# The Lengauer-Tarjan Algorithm: Correctness

---

**Theorem 4:** For any  $w \neq r$ ,

$$sdom(w) = \min (\{ v \mid (v, w) \in E \text{ and } v < w \} \cup \\ \{ sdom(u) \mid u > w \text{ and } \exists (v, w) \in E \text{ such that } u * \rightarrow v \}).$$

---

- $sdom(w) \geq x$

Assume that  $(sdom(w), w) \in E \Rightarrow sdom(w) \geq x$ .

# The Lengauer-Tarjan Algorithm: Correctness

**Theorem 4:** For any  $w \neq r$ ,

$$sdom(w) = \min (\{ v \mid (v, w) \in E \text{ and } v < w \} \cup \{ sdom(u) \mid u > w \text{ and } \exists (v, w) \in E \text{ such that } u * \rightarrow v \}).$$

- $sdom(w) \geq x$

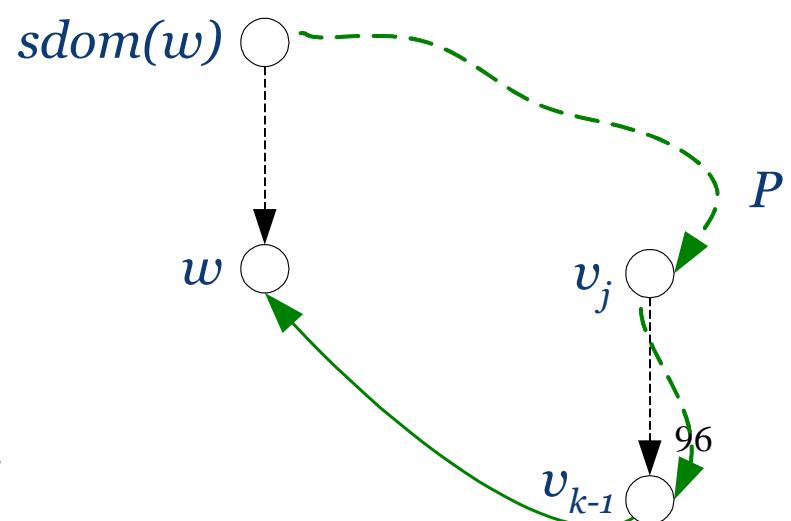
Assume that  $P = (sdom(w) = v_0, v_1, \dots, v_k = w)$  is a simple path

$$v_i > w, 1 \leq i \leq k-1.$$

$$j = \min \{ i \geq 1 \mid v_i * \rightarrow v_{k-1} \}.$$

$$\text{Lemma 1} \Rightarrow v_i > v_j, 1 \leq i \leq j-1$$

$$\Rightarrow x \leq sdom(v_j) \leq sdom(w).$$



# The Lengauer-Tarjan Algorithm: Almost-Linear-Time Version

---

We get better running time if the trees in  $F$  are **balanced** (as in Set-Union).

$F$  is balanced for constants  $a > 1$ ,  $c > 0$  if for all  $i$  we have:

$$\# \text{ vertices in } F \text{ of height } i \leq cn/a^i$$

**Theorem 5** [Tarjan 1975]: The total length of an arbitrary sequence of  $m$  path compressions in an  $n$ -vertex forest balanced for  $a, c$  is  $O((m+n) \cdot \alpha(m+n, n))$ , where the constant depends on  $a$  and  $c$ .

# Linear-Time Algorithms

---

There are linear-time dominators algorithms both for the **RAM Model** and the **Pointer-Machine Model**.

- Based on LT, but **much more complicated**.
- First published algorithms that claimed linear-time, in fact didn't achieve that bound.

**RAM:** Harel [1985]

→ Alstrup, Harel, Lauridsen and Thorup [1999]

**Pointer-Machine:** Buchsbaum, Kaplan, Rogers and Westbrook [1998]

→ G. and Tarjan [2004],

→ Buchsbaum, G., Kaplan, Rogers, Tarjan and Westbrook [2008]<sup>98</sup>

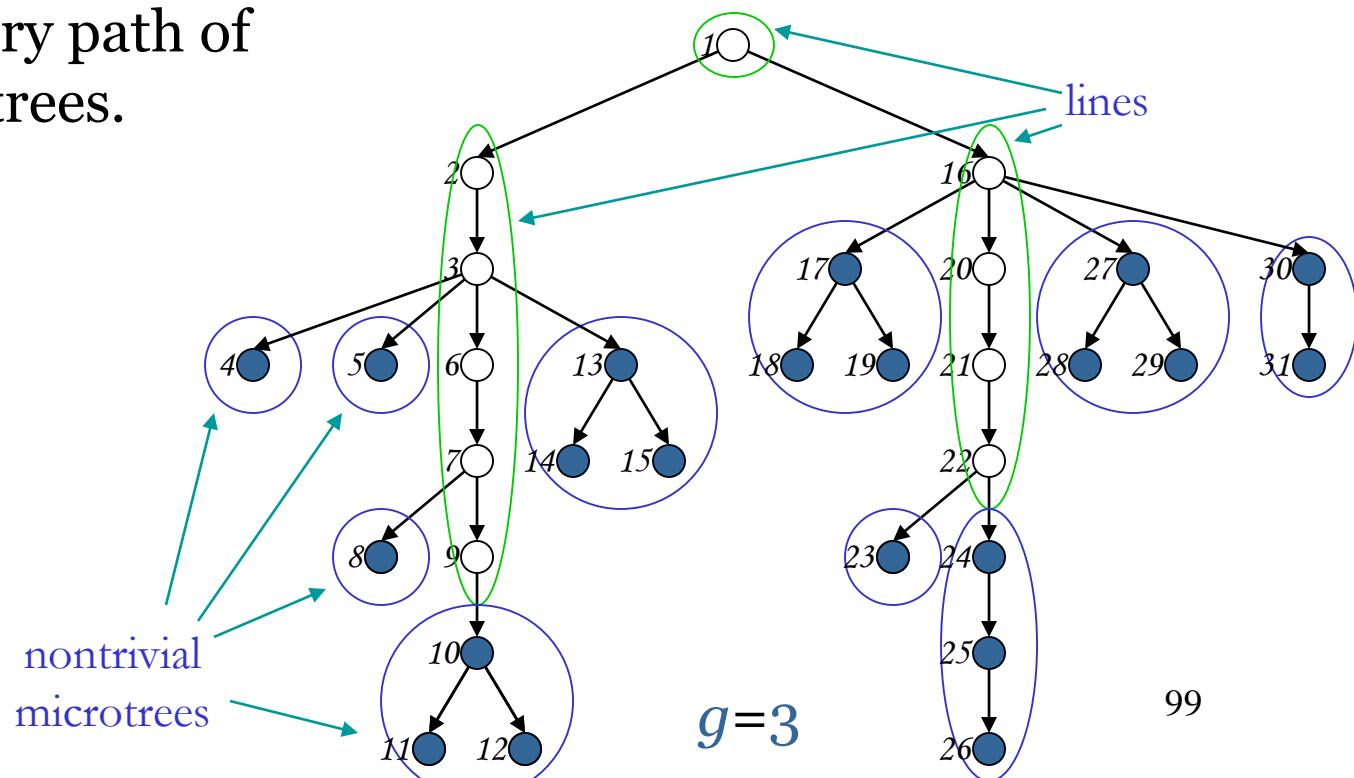
# GT Linear-Time Algorithm: High-Level View

Partition DFS-tree  $D$  into nontrivial microtrees and lines.

**Nontrivial microtree:** Maximal subtree of  $D$  of size  $\leq g$  that contains at least one leaf of  $D$ .

**Trivial microtree:** Single internal vertex of  $D$ .

**Line:** Maximal unary path of trivial microtrees.



# GT Linear-Time Algorithm: High-Level View

Partition DFS-tree  $D$  into nontrivial microtrees and lines.

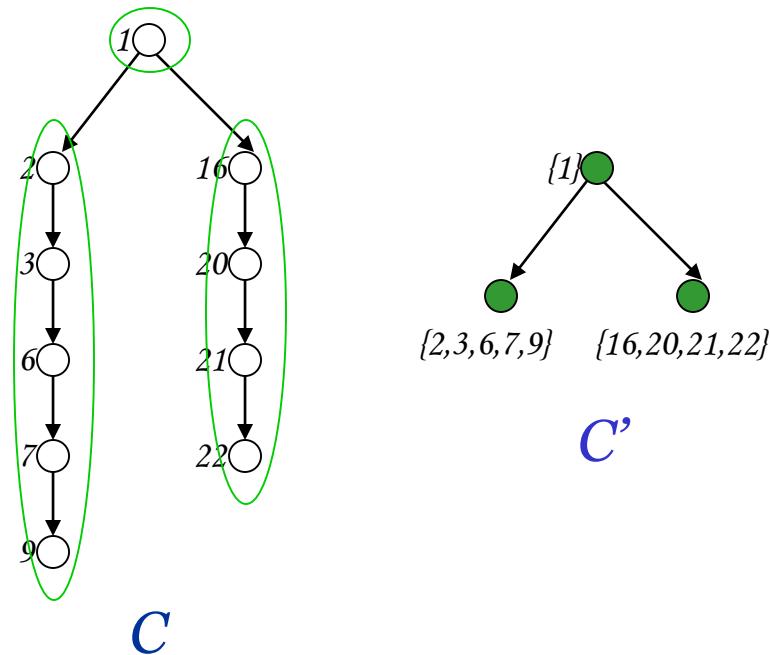
**Nontrivial microtree:** Maximal subtree of  $D$  of size  $\leq g$  that contains at least one leaf of  $D$ .

**Trivial microtree:** Single internal vertex of  $D$ .

**Line:** Maximal unary path of trivial microtrees.

**Core  $C$ :** Tree  $D$  – nontrivial microtrees

$C'$ : contract each line of  $C$  to a single vertex

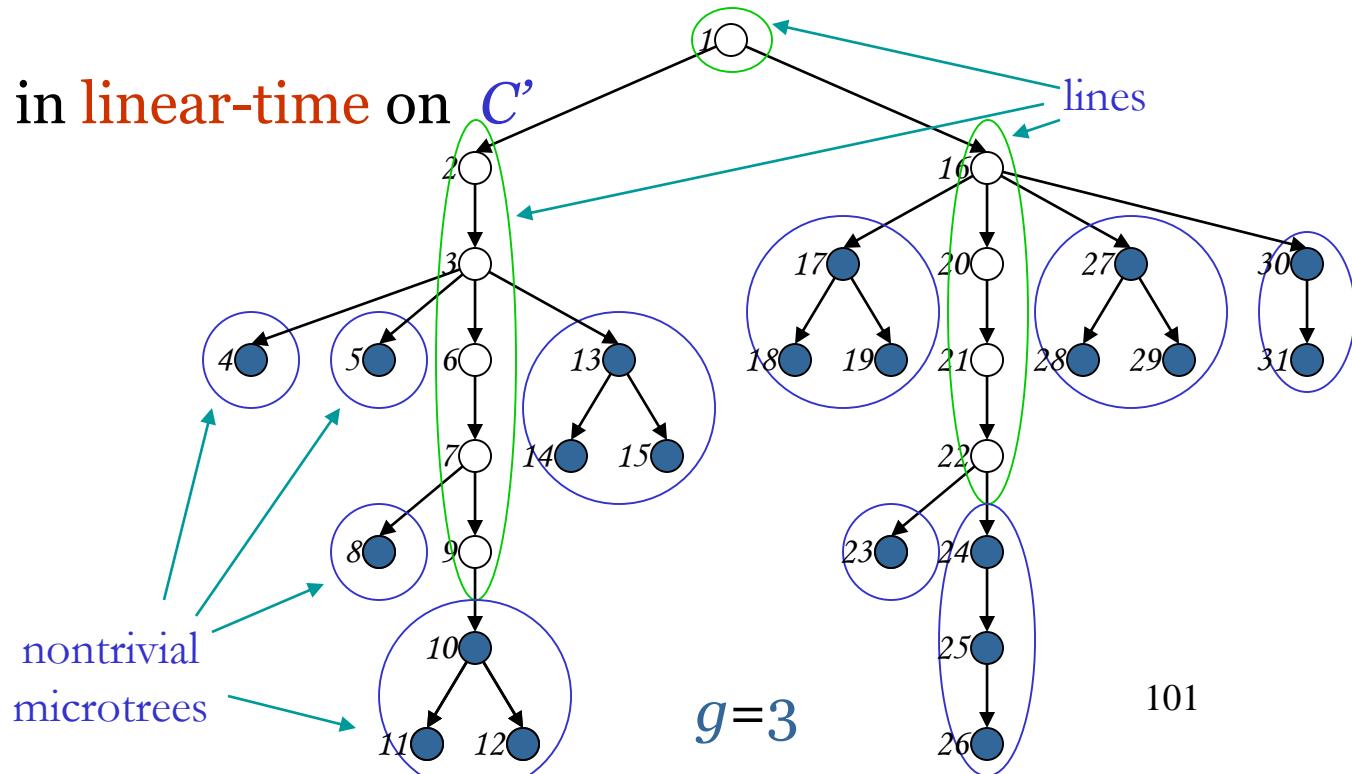


# GT Linear-Time Algorithm: High-Level View

**Basic Idea:** Compute **external dominators** in each nontrivial microtree and semidominators in each line, by running LT on  $C'$

Precompute **internal dominators** in non-identical nontrivial microtrees.

**Remark:** LT runs in **linear-time** on  $C'$



# The Lengauer-Tarjan Algorithm: Almost-Linear-Time Version

---

Back to the  $O(n \cdot \alpha(m, n))$ -time version of LT...

We give the details of a data structure that achieves asymptotically faster `link()` and `eval()`

# A Better Data Structure for `link()` and `eval()`

---

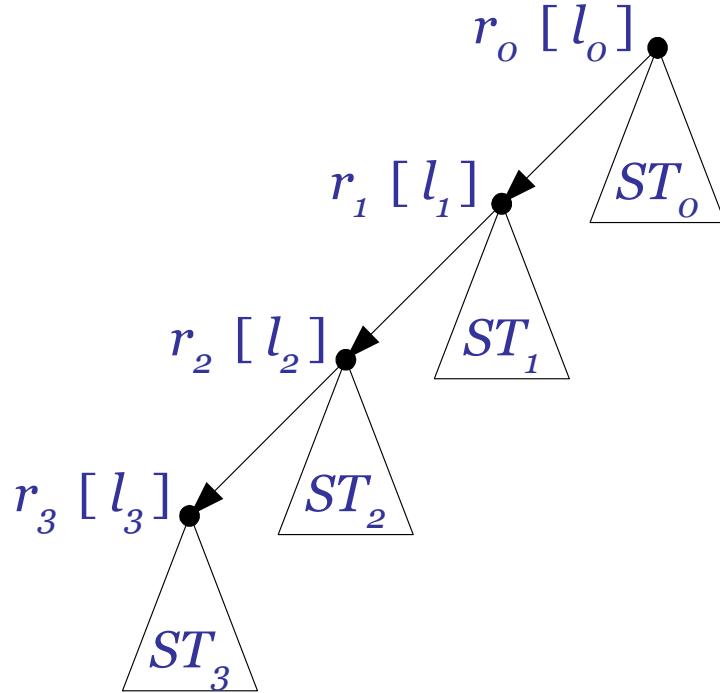
$VF$  must satisfy one additional property:

1. For each  $T$  in  $F$  there is a corresponding  $VT$  in  $VF$  with the same vertices as  $T$ .
2. Corresponding trees  $T$  and  $VT$  have the same root with the same label.
3. If  $v$  is any vertex,  $\text{eval}(v, F) = \text{eval}(v, VF)$ .
4. Each  $VT$  consists of subtrees  $ST_i$  with roots  $r_i$ ,  $0 \leq i \leq k$ , such that  $\text{semi}(\text{label}(r_j)) \geq \text{semi}(\text{label}(r_{j+1}))$ ,  $1 \leq j < k$ .

# A Better Data Structure for `link()` and `eval()`

---

4. Each  $VT$  consists of subtrees  $ST_i$  with roots  $r_i$ ,  $0 \leq i \leq k$ , such that  $\text{semi}(\text{label}(r_j)) \geq \text{semi}(\text{label}(r_{j+1}))$ ,  $1 \leq j < k$ .



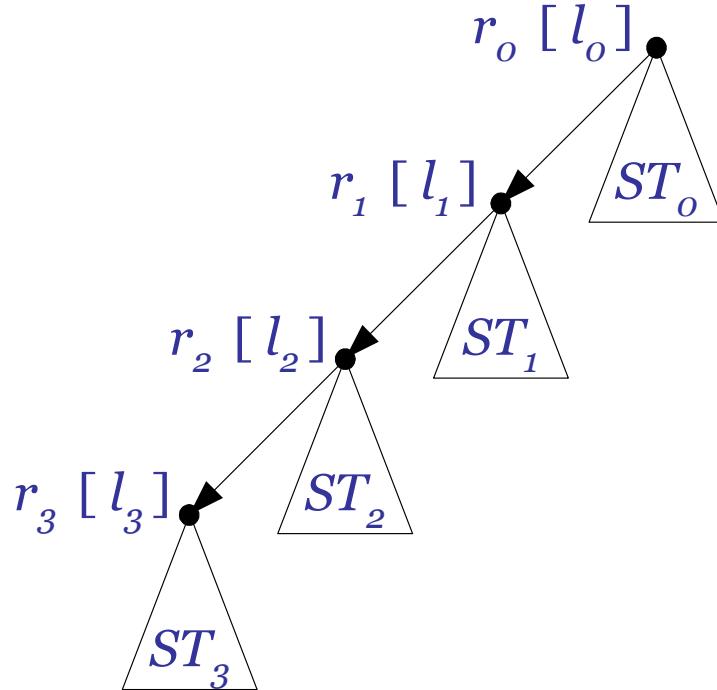
$r_o$  has **not** been processed yet  
i.e.,  $\text{semi}(r_o) \neq \text{sdom}(r_o)$

$$\text{semi}(l_1) \geq \text{semi}(l_2) \geq \text{semi}(l_3)$$

# A Better Data Structure for `link()` and `eval()`

---

4. Each  $VT$  consists of subtrees  $ST_i$  with roots  $r_i$ ,  $0 \leq i \leq k$ , such that  $\text{semi}(\text{label}(r_j)) \geq \text{semi}(\text{label}(r_{j+1}))$ ,  $1 \leq j < k$ .



We need an extra pointer per node:

$$\text{child}(r_j) = r_{j+1}, \quad 0 \leq j < k$$

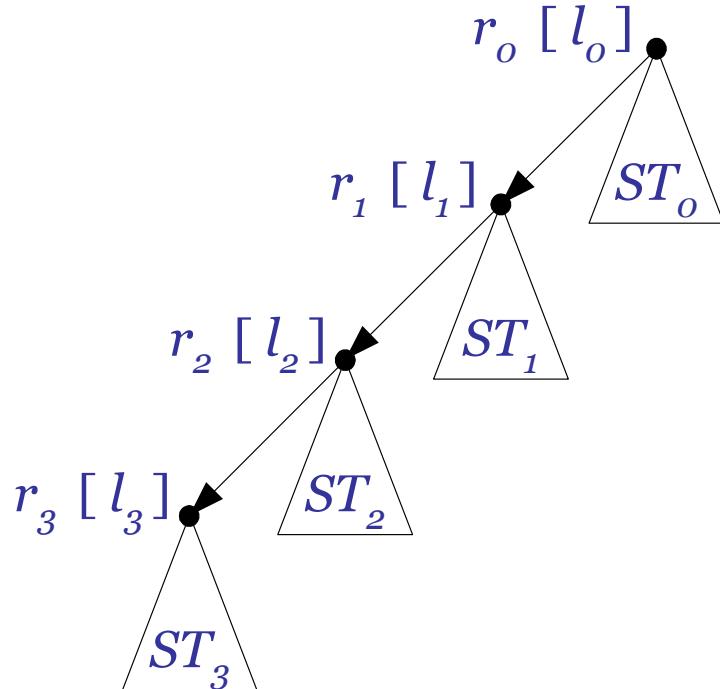
$$\text{ancestor}(r_j) = o, \quad 0 \leq j \leq k$$

$$\text{semi}(l_1) \geq \text{semi}(l_2) \geq \text{semi}(l_3)$$

# A Better Data Structure for `link()` and `eval()`

---

4. Each  $VT$  consists of subtrees  $ST_i$  with roots  $r_i$ ,  $0 \leq i \leq k$ , such that  $\text{semi}(\text{label}(r_j)) \geq \text{semi}(\text{label}(r_{j+1}))$ ,  $1 \leq j < k$ .



For any  $v$  in  $ST_j$ ,  $\text{eval}(v)$  doesn't depend on  $\text{label}(r_i)$ ,  $i < j$ .

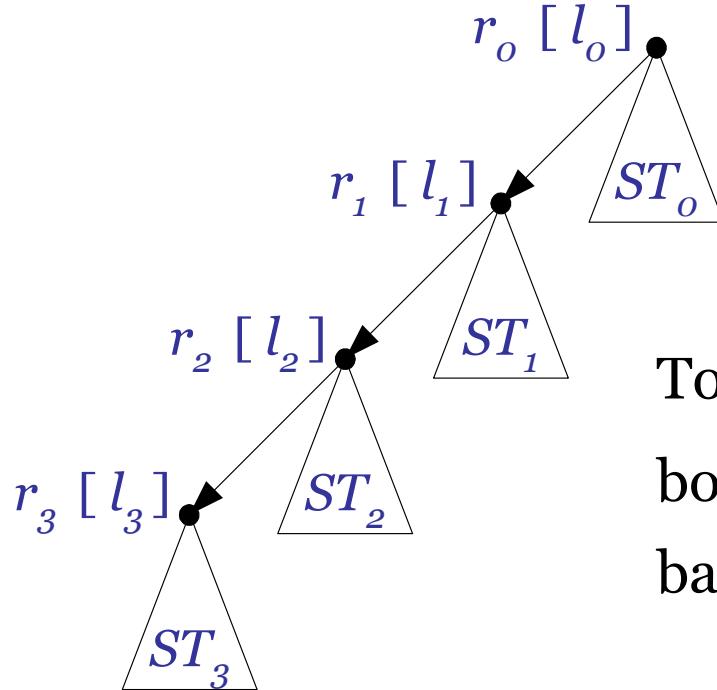
⇒ We can use path compression inside each  $ST_j$ .

$$\text{semi}(l_1) \geq \text{semi}(l_2) \geq \text{semi}(l_3)$$

# A Better Data Structure for `link()` and `eval()`

---

4. Each  $VT$  consists of subtrees  $ST_i$  with roots  $r_i$ ,  $0 \leq i \leq k$ , such that  $\text{semi}(\text{label}(r_j)) \geq \text{semi}(\text{label}(r_{j+1}))$ ,  $1 \leq j < k$ .



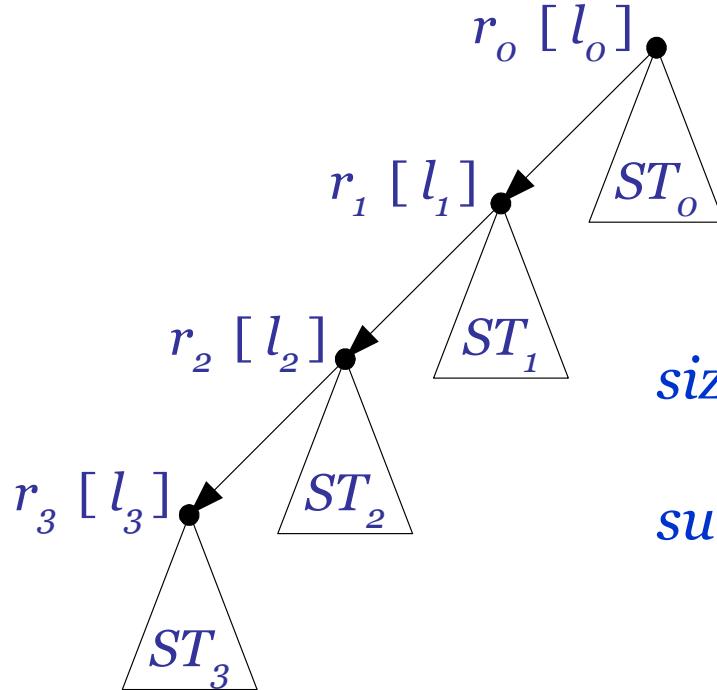
To get the  $O((m+n) \cdot \alpha(m+n, n))$  time bound we want to keep each  $ST_j$  balanced.

$$\text{semi}(l_1) \geq \text{semi}(l_2) \geq \text{semi}(l_3)$$

# A Better Data Structure for `link()` and `eval()`

---

4. Each  $VT$  consists of subtrees  $ST_i$  with roots  $r_i$ ,  $0 \leq i \leq k$ , such that  $\text{semi}(\text{label}(r_j)) \geq \text{semi}(\text{label}(r_{j+1}))$ ,  $1 \leq j < k$ .



$$\text{size}(r_j) = |ST_j| + |ST_{j+1}| + \dots + |ST_k|$$

$$\text{subsize}(r_j) = |ST_j| = \text{size}(r_j) - \text{size}(r_{j+1})$$

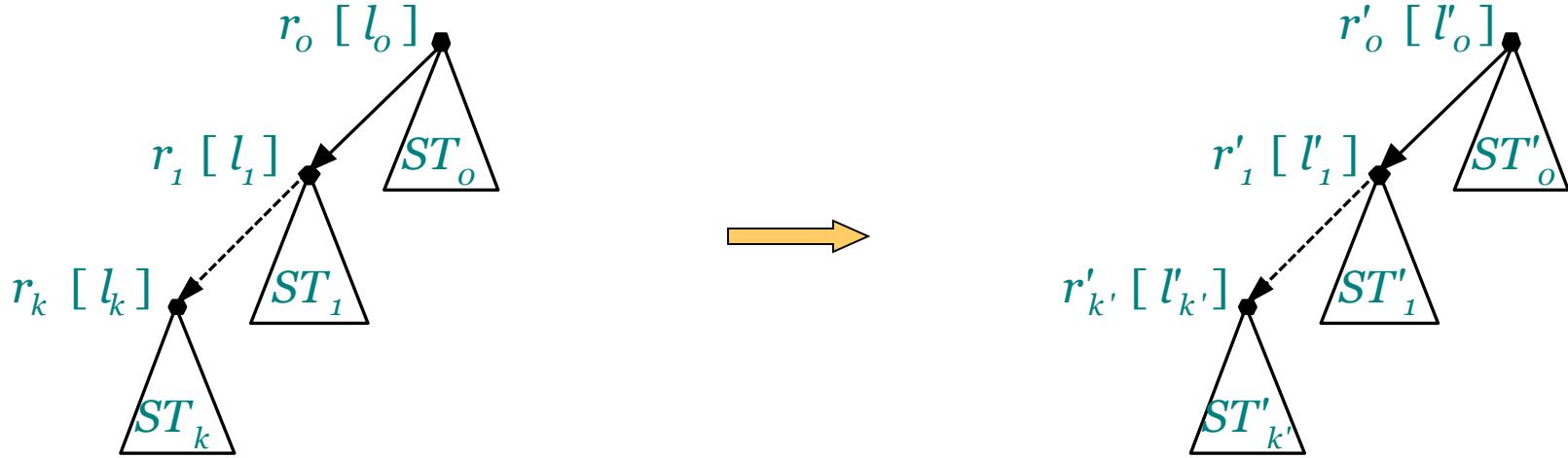
$$\text{semi}(l_1) \geq \text{semi}(l_2) \geq \text{semi}(l_3)$$

# A Better Data Structure for `link()` and `eval()`

---

First we implement the following auxiliary operation:

`update(r)`: If  $r$  is a root in  $F$  and  $l = \text{label}(r)$  then  
restore Property 4 for all subtree roots.



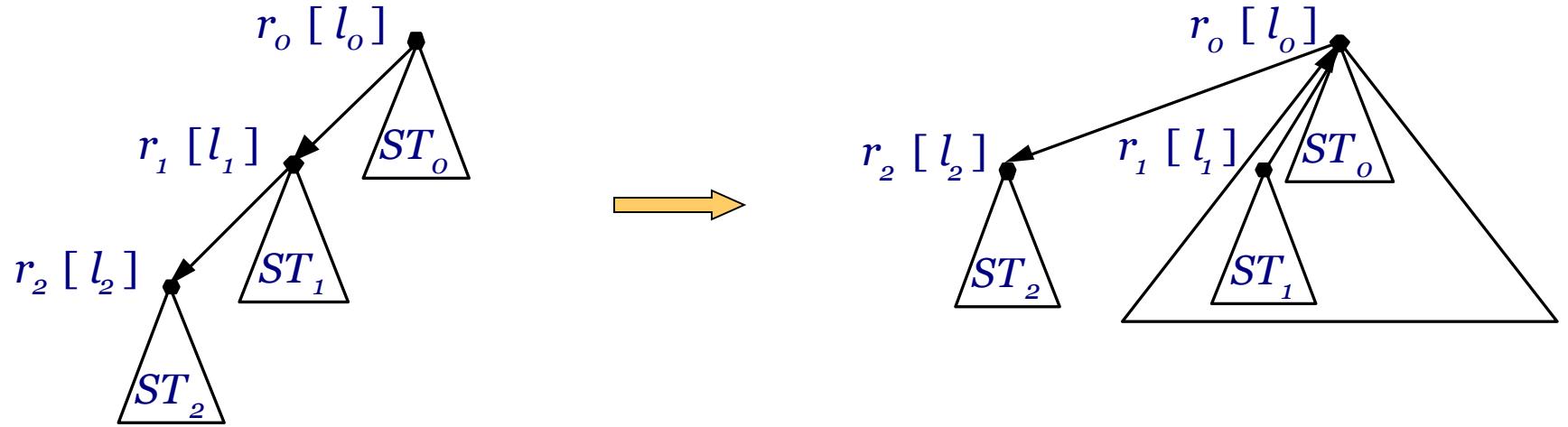
$$\begin{aligned} \text{semi}(\text{label}(r_j)) &\geq \text{semi}(\text{label}(r_{j+1})) \\ 1 \leq j < k. \end{aligned}$$

$$\begin{aligned} \text{semi}(\text{label}(r'_j)) &\geq \text{semi}(\text{label}(r'_{j+1})) \\ 0 \leq j < k'. \end{aligned}$$

# Implementation of `update()`

---

Suppose  $\text{semi}(l_o) < \text{semi}(l_1)$

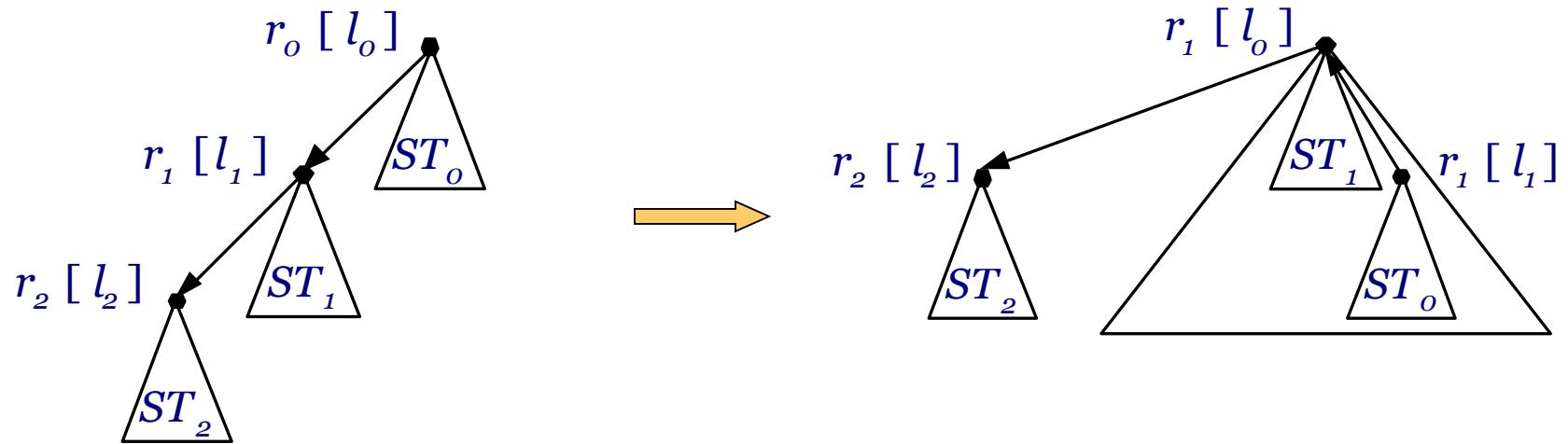


Case (a):  $\text{subsize}(r_1) \geq \text{subsize}(r_2)$

# Implementation of `update()`

---

Suppose  $\text{semi}(l_o) < \text{semi}(l_1)$



Case (b):  $\text{subsize}(r_1) < \text{subsize}(r_2)$

## Implementation of `update()`

---

`update( $r$ )`: Let  $VT$  be the virtual tree rooted at  $r = r_o$ ,  
with subtrees  $ST_i$  and corresponding roots  $r_i$ ,  
 $0 \leq i \leq k$ .

If  $\text{semi}(\text{label}(r)) < \text{semi}(\text{label}(r_1))$  then combine  
 $ST_o$  and  $ST_1$  to a new subtree  $ST'_o$ . Repeat this  
process for  $ST_j$ ,  $i = 2, \dots, j$ , where  $j=k$  or  
 $\text{semi}(\text{label}(r)) \geq \text{semi}(\text{label}(r_j))$ .

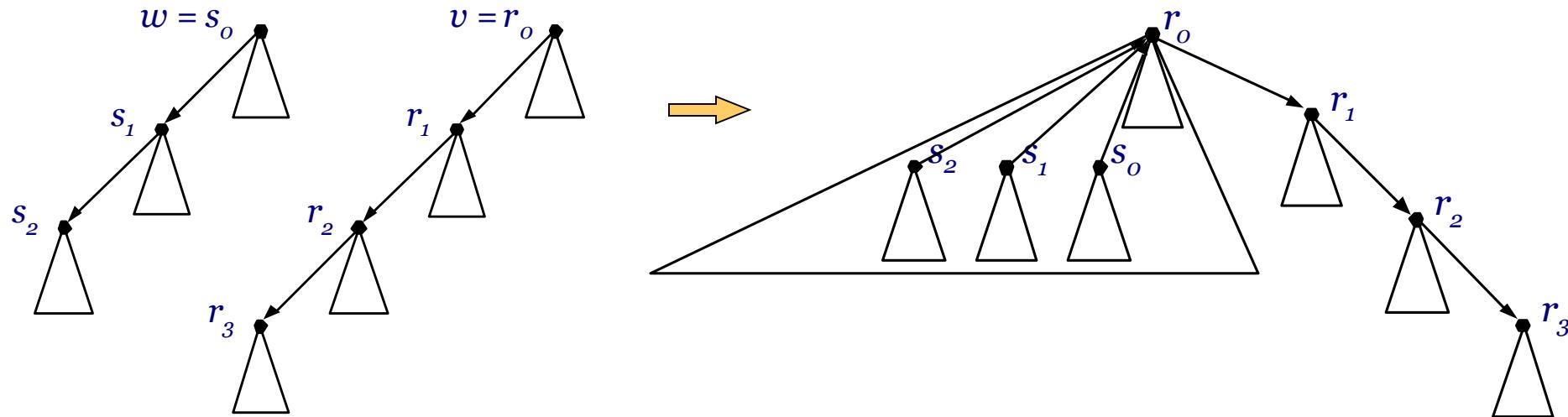
Set the label of the root  $r'_o$  of the final subtree  $ST'_o$   
equal to  $\text{label}(r)$  and set  $\text{child}(r) = r'_1$ .

# Implementation of `link()`

---

$\text{link}(v, w)$ :  $\text{update}(w, \text{label}(v))$

Combine the virtual trees rooted at  $v$  and  $w$ .

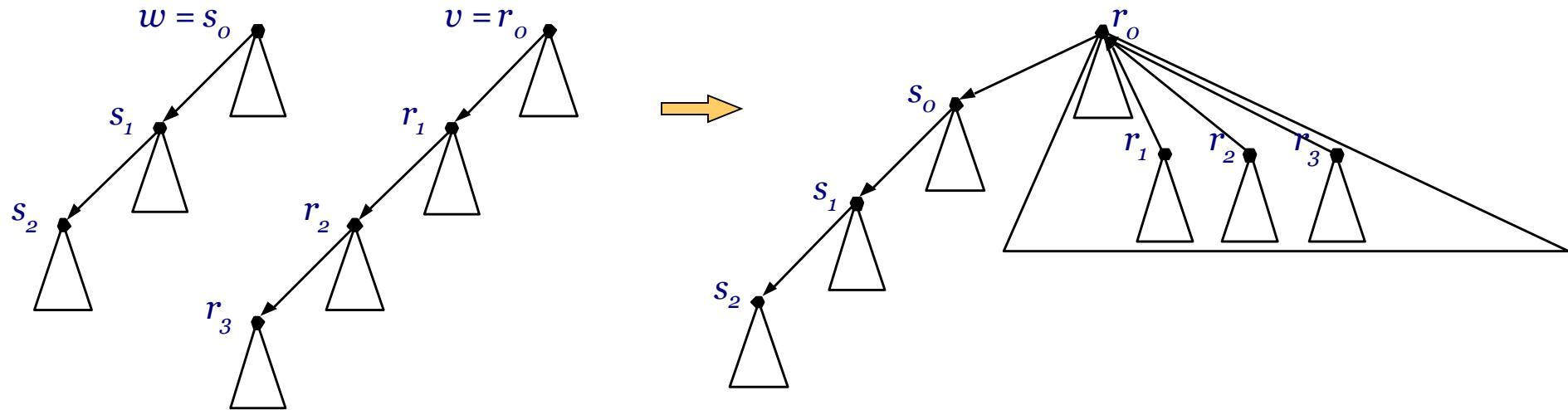


Case (a):  $\text{size}(v) \geq \text{size}(w)$

# Implementation of `link()`

$\text{link}(v, w)$ :  $\text{update}(w, \text{label}(v))$

Combine the virtual trees rooted at  $v$  and  $w$ .



Case (b):  $\text{size}(v) < \text{size}(w)$

## Implementation of `link()`

---

$\text{link}(v, w)$ :  $\text{update}(w, \text{label}(v))$

Let  $VT_1$  be the virtual tree rooted at  $v = r_o$ ,  
with subtree roots  $r_i$ ,  $0 \leq i \leq k$ .

Let  $VT_2$  be the virtual tree rooted at  $w = s_o$ ,  
with subtree roots  $s_i$ ,  $0 \leq i \leq l$ .

If  $\text{size}(v) \geq \text{size}(w)$  make  $v$  parent of  $s_o, s_1, \dots, s_l$ .  
Otherwise make  $v$  parent of  $r_1, r_2, \dots, r_k$  and  
make  $w$  the child of  $v$ .

## Analysis of `link()`

---

We will show that the (uncompressed) subtrees built by `link()` are balanced.

$\mathcal{U}$ : uncompressed forest.

Just before  $\text{ancestor}(y) \leftarrow x \Rightarrow x$  and  $y$  are subtree roots.

Then we add  $(x,y)$  to  $\mathcal{U}$ . Let  $(x,y), (y,z) \in \mathcal{U}$ .

$(x,y)$  is **good** if  $\text{subsize}(x) \geq 2 \cdot \text{subsize}(y)$

$(y,z)$  is **mediocre** if  $\text{subsize}(x) \geq 2 \cdot \text{subsize}(z)$

## Analysis of `link()`

---

$(x,y)$  is **good** if  $\text{subsize}(x) \geq 2 \cdot \text{subsize}(y)$

$(y,z)$  is **mediocre** if  $\text{subsize}(x) \geq 2 \cdot \text{subsize}(z)$

Every edge added by `update()` is good.

Assume  $(x,y)$  and  $(y,z)$  are added outside `update()`.

$(y,z) \Rightarrow \text{size}(y) \geq 2 \cdot \text{size}(z)$

$(x,y) \Rightarrow \text{subsize}(x) \geq 2 \cdot \text{subsize}(z)$

Thus, every edge added by `link()` is mediocre.

## Analysis of link()

---

For any  $x, y$  and  $z$  in  $U$  such that  $x \rightarrow y \rightarrow z$ , we have  
 $\text{subsize}(x) \geq 2 \cdot \text{subsize}(z)$ .

It follows by induction that any vertex of height  $h$  in  $U$  has  
 $\geq 2^{\lfloor h/2 \rfloor}$  descendants.

The number of vertices of height  $h$  in  $U$  is  $\leq n / 2^{\lfloor h/2 \rfloor}$   
 $\leq 2^{1/2} n / (2^{1/2})^h \Rightarrow U$  is balanced for constants  $a = \sqrt{2}$ ,  $b = \sqrt{2}$ .

# Lengauer-Tarjan Running Time

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By Theorem 5,  $m$  calls to `eval()` take  $O((m+n) \cdot \alpha(m+n, n))$  time.

The total time for all the `link()` instructions is proportional to the number of edges in  $U \Rightarrow O(n+m)$  time.

The Lengauer-Tarjan algorithm runs in  $O(m \cdot \alpha(m, n))$  time.