



Dynamic Trees and Log-lists

Irena Rusu

Université de Nantes, LINA

Irena.Rusu@univ-nantes.fr

Outline

- “ Motivations for an Array-List (but not the Java Collection \circ)
- “ Dynamic Trees
- “ A Sequence is a Filiform Tree
- “ A Log-List is a Filiform tree too \circ but Needs Adjustments
- “ Conclusion

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- “ **Motivations for an Array-List** (but not the Java Collection \circ)
- “ Fundamentals: Dynamic Trees
- “ Easy observation: A Sequence is a Filiform Tree
- “ A step further: A Log-List too \circ but Needs Adjustments
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Motivations for an Array-List

Problem. Improve the running time of the existing algorithm for sorting a permutation P by prefix reversals and prefix transpositions (explained below).

Prefix reversal

$7+1=8$

 P: 7 4 5 2 8 9 1 3 6 10 P_8^{-1} needed

P: 7 4 5 2 8 9 1 3 6 10
 Need to:
 " Reverse
 " Update some P_i^{-1}

P: 2 5 4 7 8 9 1 3 6 10

P_8^{-1} better using arrays, reverse better using double-linked lists, update ???

Motivations for an Array-List

Prefix transposition



P_1^{-1}, P_8, P_4^{-1} needed



Need to:

- “ Transpose
- “ Update many P_i
- “ Update many P_i^{-1}



P_1^{-1}, P_8, P_4^{-1} better using arrays, transpose better using double-linked lists, update ??

Motivations for an Array-List : An Idea

Goal:

- choose and perform a reversal or transposition in $O(\log n)$
- improve the running time of the sorting algorithm from $O(n^2)$ to $O(n \log n)$

P: 2 5 4 7 8 9 1 3 6 10

Use trees:

Initial step: the permutation is a big tree

Coloring step: cut the big tree into subtrees

Block operation: change trees order and/or left-right orientation, and globally modify P and P^{-1} values at the root

Final step: link subtrees

Motivations for an Array-List : An Idea

P: 2 5 4 7 8 9 1 3 6 10

Trees need to support:

- “ Cut several edges
- “ Link several subtrees
- “ Left-right flipping
in $O(\log n)$



Binary balanced trees :

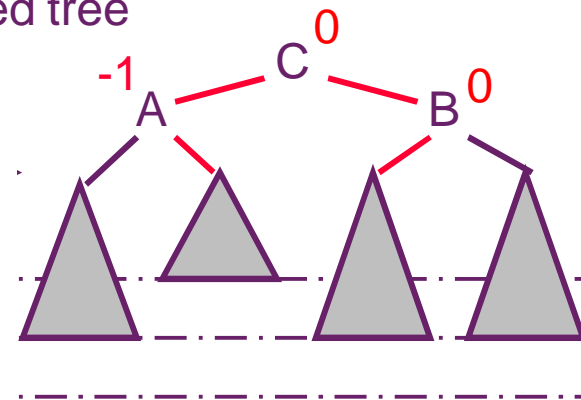
- “ Height balanced trees
- “ Red-Black trees
- “ \tilde{O}

(Any other with height in $O(\log n)$, and which supports re-balancing)

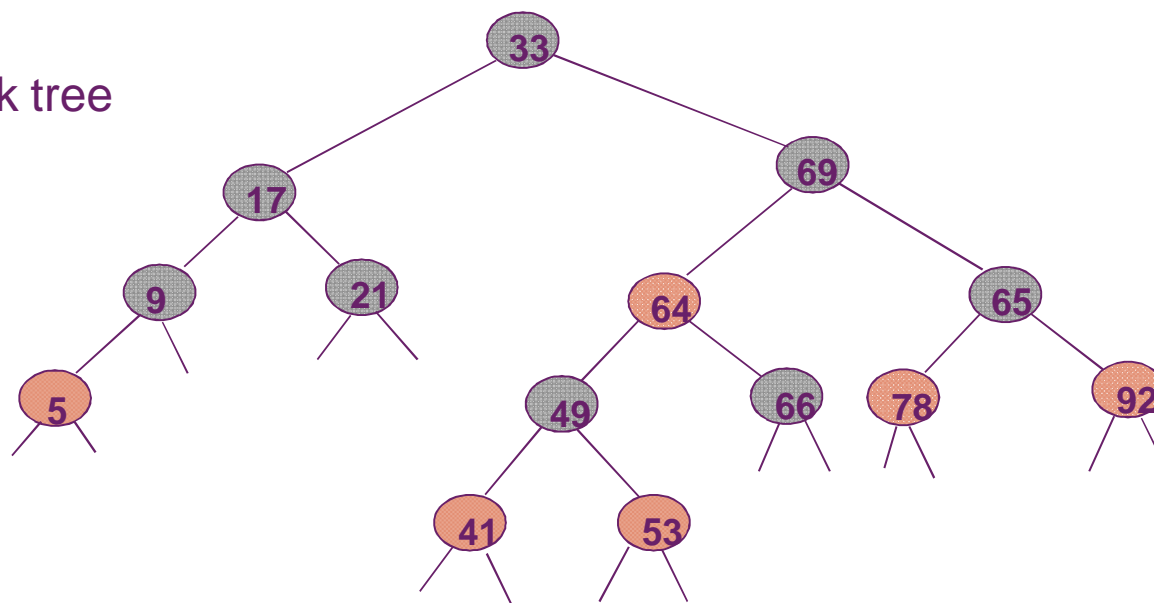
Rappels

(Re-)balance

Height-balanced tree



Red-black tree



Motivations for an Array-List : An Idea

P: 2 5 4 7 8 9 1 3 6 10

Trees need to support:

- “ Cut several edges
 - “ Link several subtrees
 - “ Left-right flipping
- in $O(\log n)$



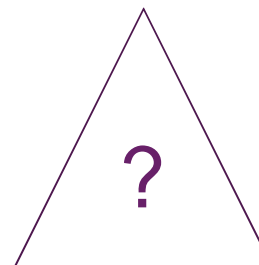
Binary balanced trees :

- “ Height balanced trees
- “ Red-Black trees
- “ \tilde{O}

(Any other with height in $O(\log n)$, and which supports re-balancing)

\tilde{O} but also

- “ global modifications of P
- and P^{-1} values



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- “ Conclusion

Dynamic trees: Introduction (1)

Daniel D. Sleator and Robert E. Tarjan (1983)

Data structure proposed to maintain a forest of vertex disjoint rooted trees under **link** (add an edge) and **cut** (delete an edge) operations between trees, in $O(\log n)$ each.

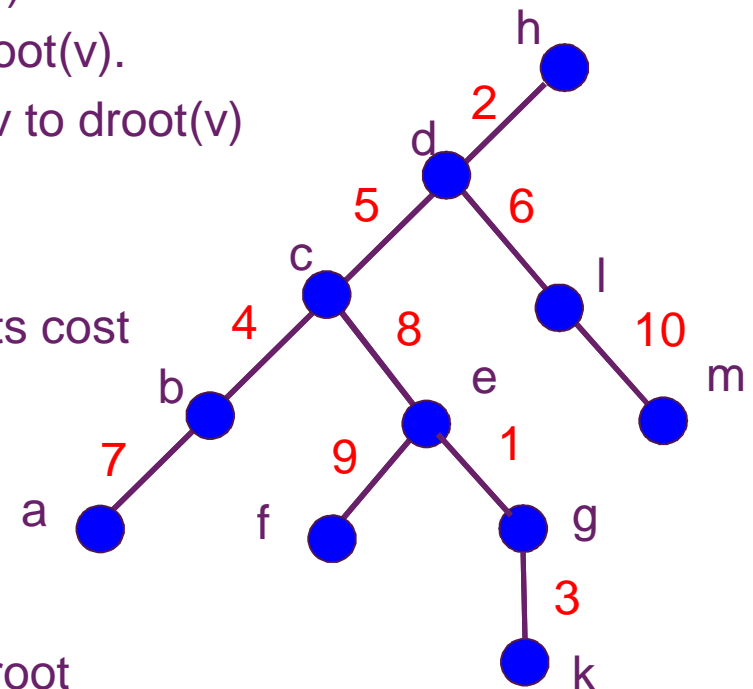
Applications:

- . maximum flow problem and variants (Sleator, Tarjan, 1983)
- . the Least Common Ancestor problem (Sleator, Tarjan, 1983)
- . the transshipment problem (Sleator, Tarjan, 1983)
- . the string matching problem in compressed strings (Farach, Thorup, 1995)
- . ã

Dynamic trees: Introduction (2)

More precisely, do in $O(\log n)$, given the forest of dynamic trees:

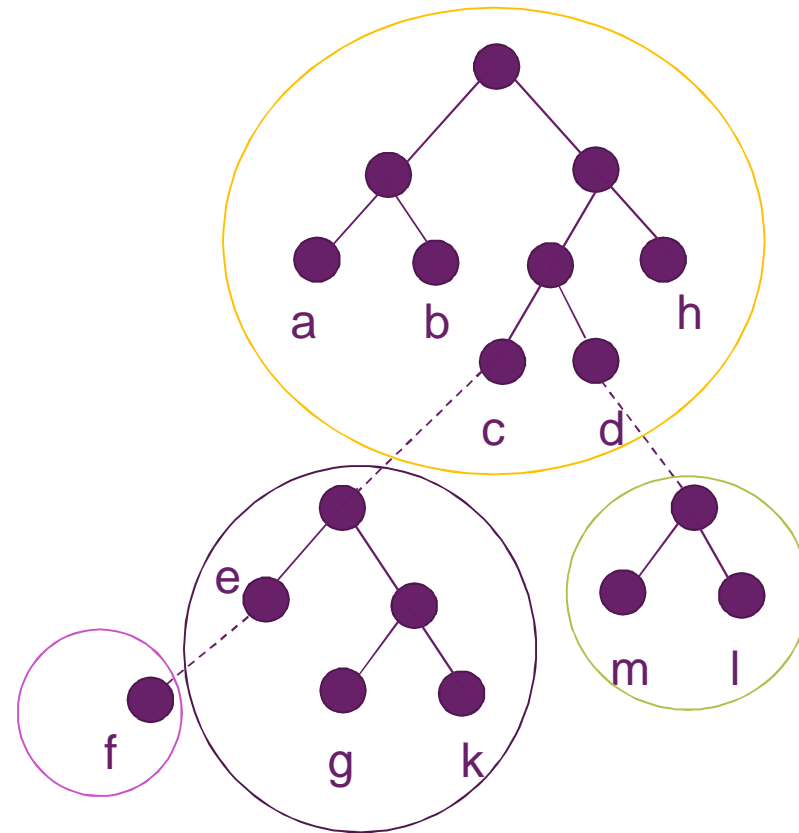
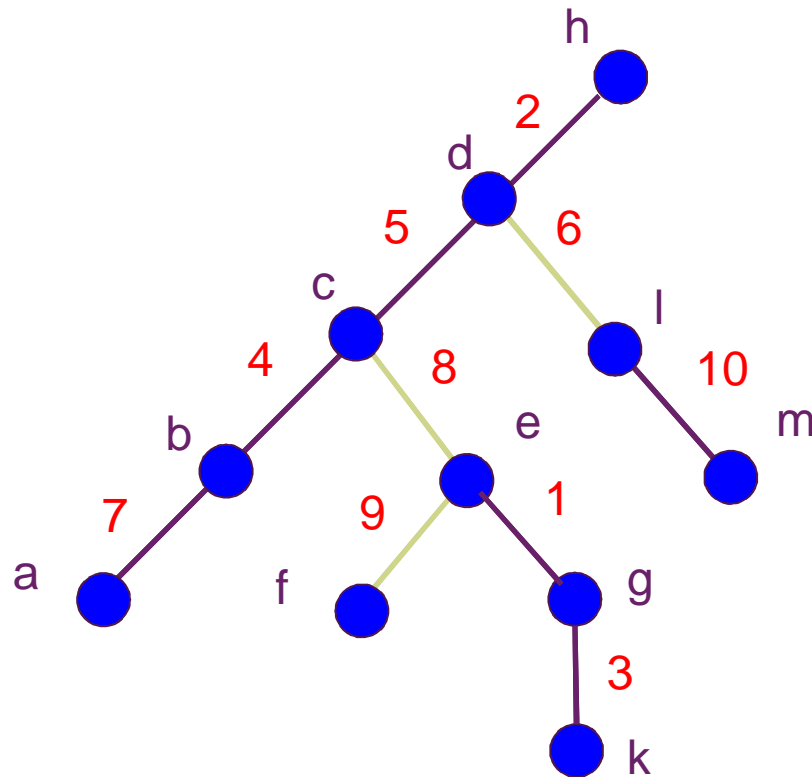
- ” **dparent(v)**: return the parent of v in its tree, or null
- ” **droot(v)**: return the root of the tree containing v
- ” **dcost(v)**: return the cost of the edge $(v, \text{dparent}(v))$
- ” **dmincost(v)**: return the vertex w closest to $\text{droot}(v)$ whose $\text{dcost}(w)$ is minimum on the path from v to $\text{droot}(v)$.
- ” **dupdate(v,u)**: add u to all costs on the path from v to $\text{droot}(v)$
- ” **dlink(v,w,u)**: add edge (v,w) of cost u assuming $v = \text{droot}(v)$, thus making w the parent of v
- ” **dcut(v)**: delete the edge $(v, \text{parent}(v))$ and return its cost
- ” **devert(v)**: make v become the root of its tree



Note. All these are operations on paths towards the root

Dynamic trees: Ideas

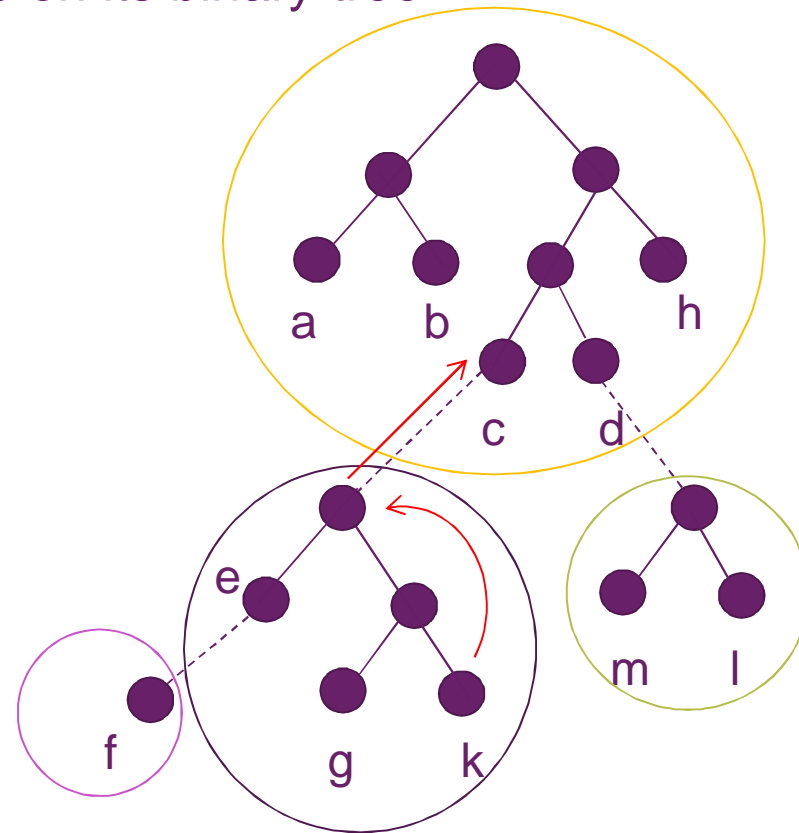
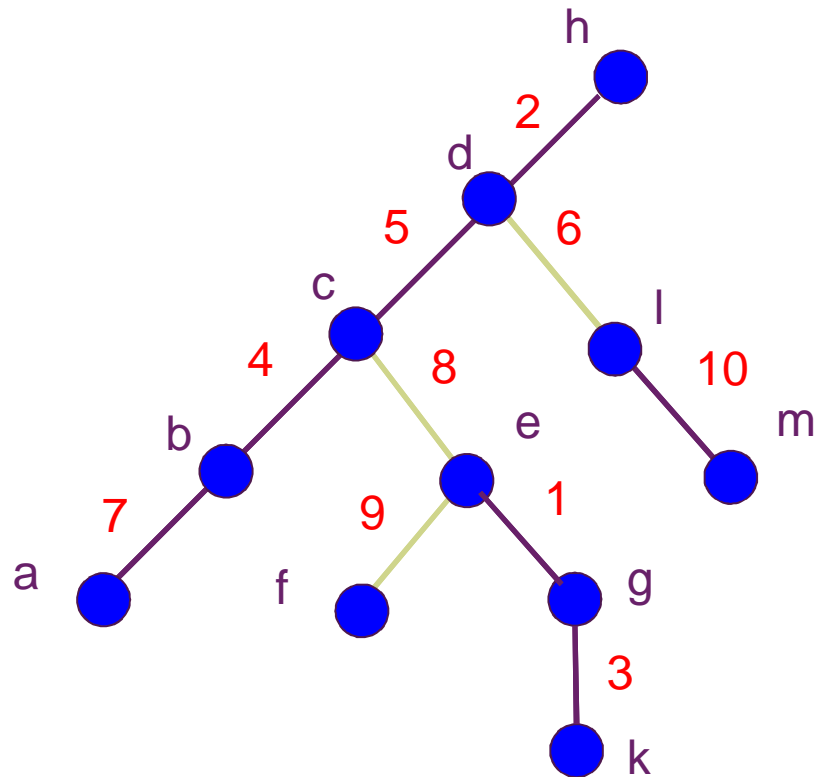
1. A tree is partitioned into « solid » paths going towards the root (edges are « solid » or « dashed »)



2. Each « solid » path is represented as a (particular) binary tree.

Dynamic trees: Ideas

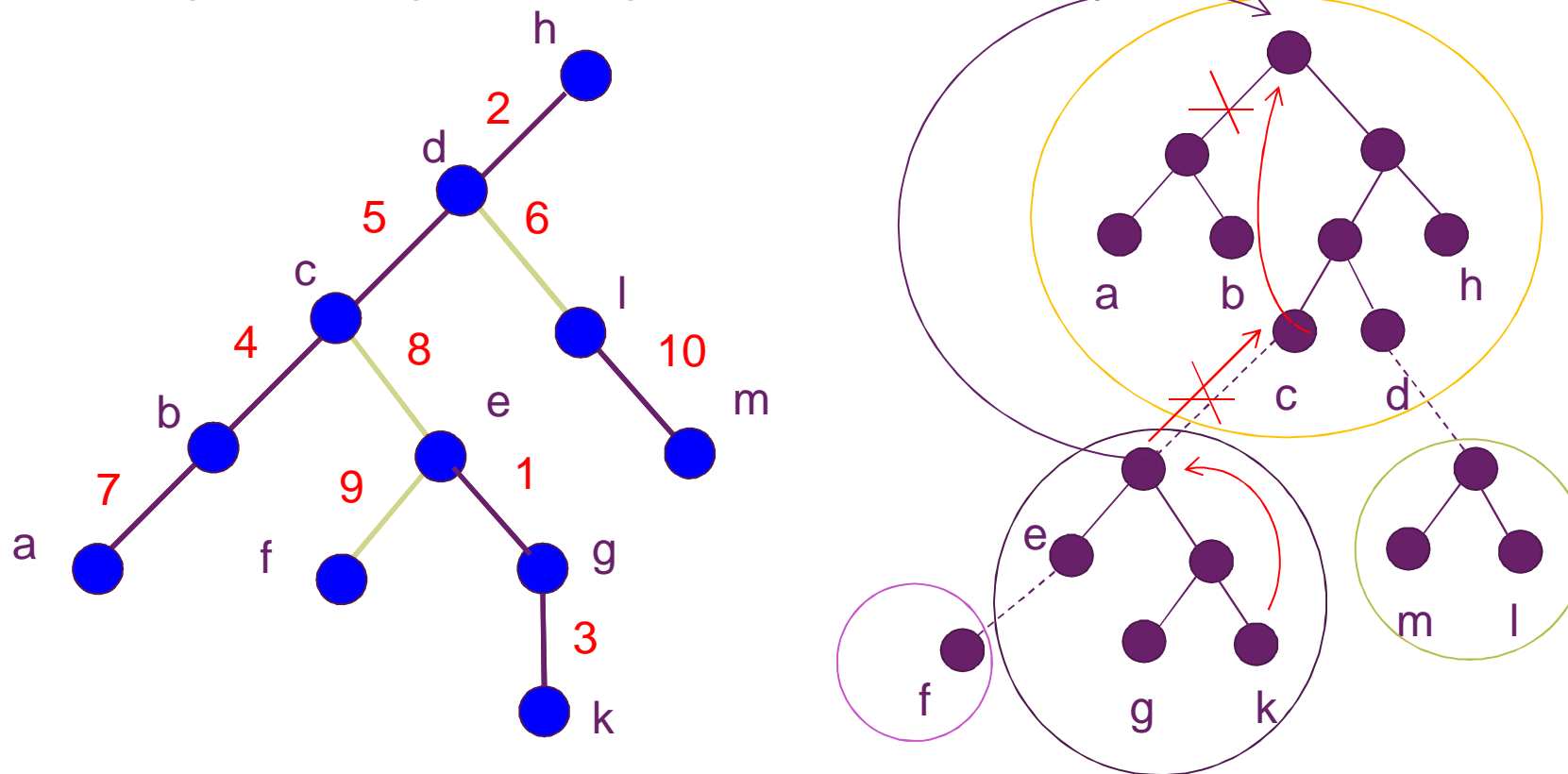
3. Operations on paths towards the root: turn the concerned path into a solid path, and perform operations on its binary tree



Here, the path from k to the root.

Dynamic trees: Ideas

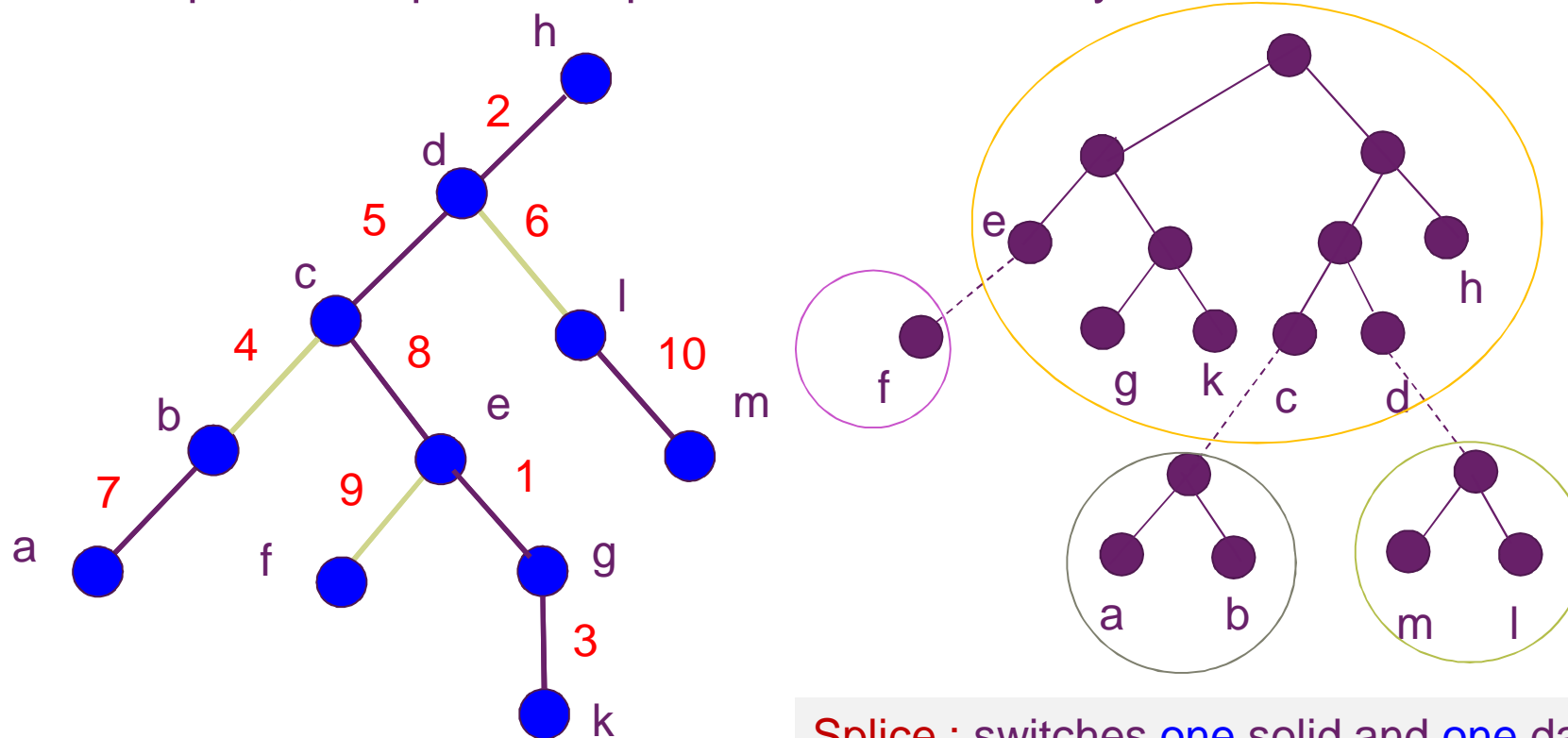
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Here, the path from k to the root.

Dynamic trees: Ideas

3. Operations on paths towards the root: turn the concerned path into a solid path, and perform operations on its binary tree



Here, the path from k to the root.

Splice : switches **one** solid and **one** dashed arc
 Needs $O(\log n)$ for binary balanced trees.
Expose : builds the entire path, uses t splices
 Needs $O(t \log n)$ for binary balanced trees.

Dynamic trees: Choose the appropriate solid paths and binary trees

Trees \ Solid paths	Standard balanced binary trees	Locally biased binary trees/ Splay trees	Globally biased binary trees
Initial: Arbitrary Later: As resulting from the operations performed	$O(\log^2 n)$ amortized	$O(\log n)$ amortized	
Initial and Later: Defined by the structure of the tree (Heavy/light edges*)			$O(\log n)$

Running time of **expose**

*(v,father(v)) is a heavy edge in the dynamic tree if $2 \cdot \text{size}(v) > \text{size}(\text{father}(v))$, where $\text{size}(v) = \# \text{nodes in the tree rooted at } v$

Dynamic trees: Choose the appropriate solid paths and binary trees

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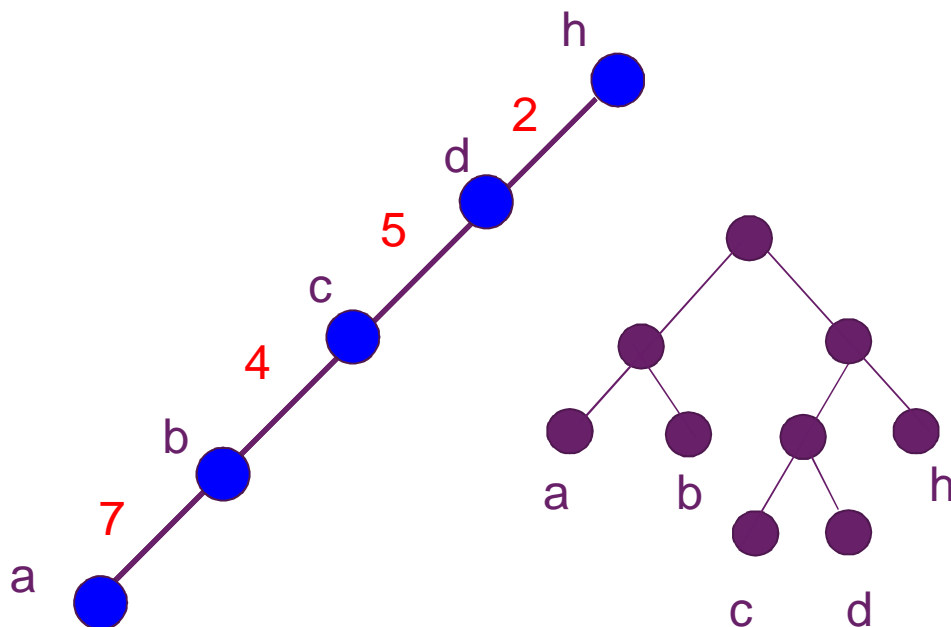
Running time of `expose`

Lemma. With solid paths defined by the heavy edges, a path from v to $\text{droot}(v)$ contains at most $\log n$ dashed edges.

Lemma. With globally biased binary trees, the running time of one splice operation is not constant, but summing over all splices, the total running time of `expose` is in $O(\log n)$.

Dynamic trees: Focus on some operations

Say $v=a$, and assume $\text{expose}(a)$ has been done.



$\text{dparent}(a)$
 $\text{droot}(a)$



$\text{dcost}(a)$
 $\text{dmincost}(a)$
 $\text{dupdate}(a,u)$



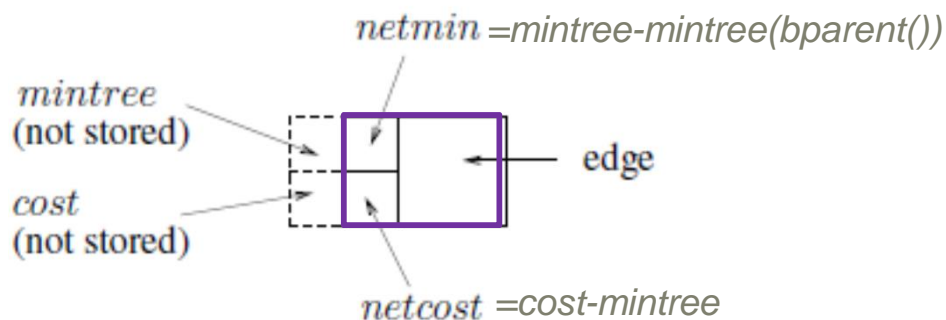
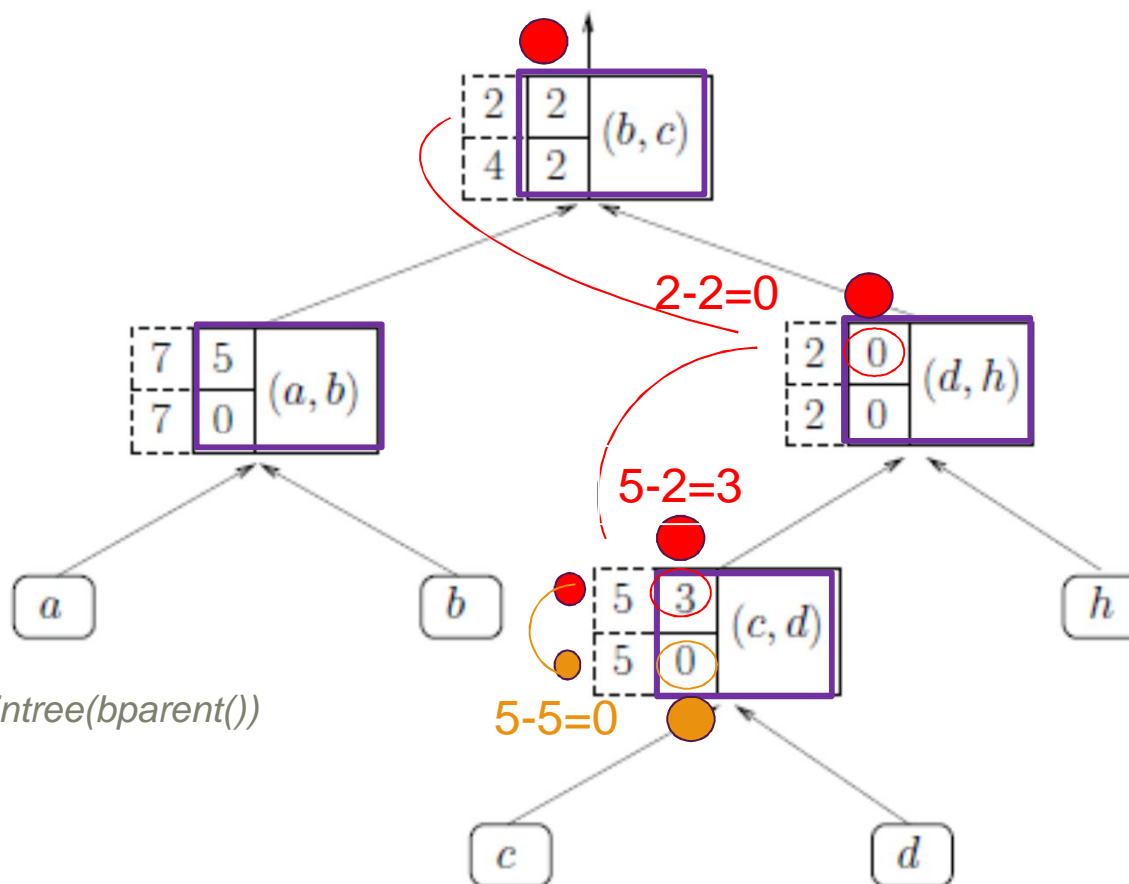
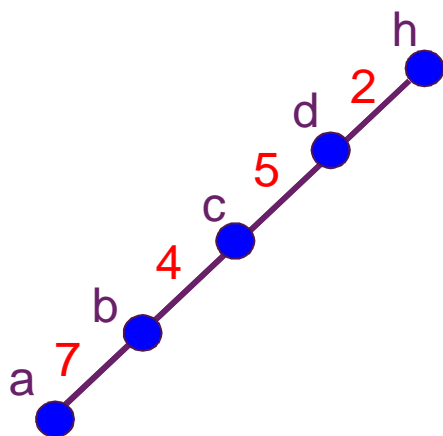
$\text{dlink}(v,c,u)$
 $\text{dcut}(c)$
 $\text{devert}(c)$



whatever the type of the tree, it will have a lot of links and information at nodes

Dynamic trees: Focus on `dcost()`, `dmincost()`, `update()`

Cost-related information in the binary tree.

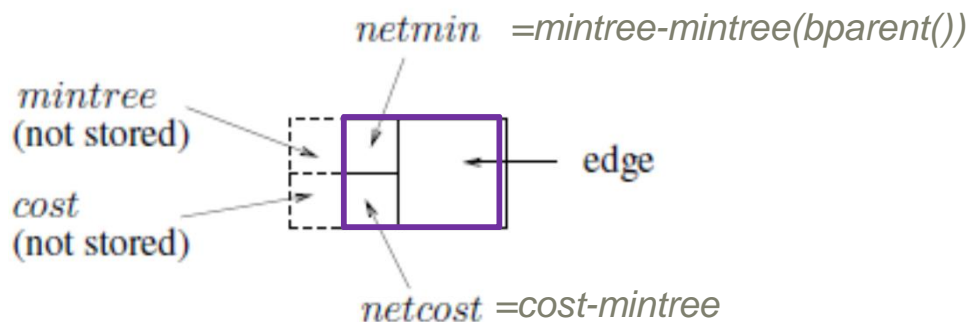
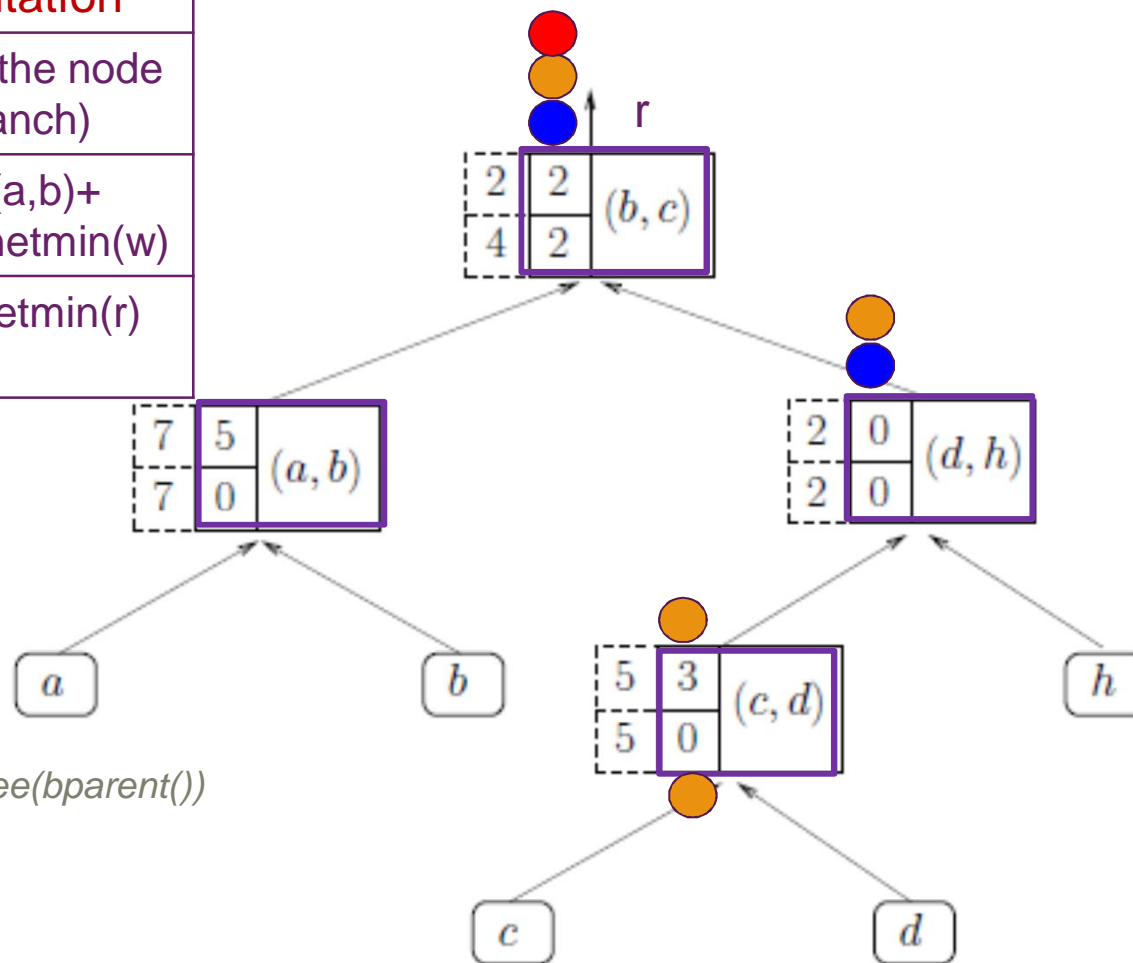


Dynamic trees: Focus on `dcost()`, `dmincost()`, `dupdate()`

Operation	Value	Computation
<code>dmincost(a)</code> ●	<code>mintree(r)</code>	Search the node (one branch)
<code>dcost(a)</code> ●	<code>netcost(a,b) + mintree(a,b)</code>	<code>netcost(a,b) + n_{w=e,...,r} netmin(w)</code>
<code>dupdate(a,x)</code> ●	+u on all costs	+u on <code>netmin(r)</code>

Notes.

- 1) *u on `netmin(r)` does **not** multiply all the values
- 2) several costs are possible



Dynamic trees: Conclusions (1)

With heavy/light edges and globally biased binary trees:

- . **Good news:** All the operations are in $O(\log n)$
- . **Principle:** Create the solid path from v to r , then work on the associated binary tree
- . **Local operations** $d_{\text{parent}}(v)$, $d_{\text{root}}(v)$ search the binary tree
- . **Aggregate operations** $d_{\text{cost}}(v)$, $d_{\text{mincost}}(v)$, $\text{update}(v,u)$ store relative information at nodes and sometimes search the binary tree
- . **Operations** $d_{\text{link}}(v,w,u)$, $d_{\text{cut}}(v)$, $d_{\text{vert}}(v)$ **modifying the structure of the forest** link, cut, reverse, re-balance the binary trees involved.
- . Several costs are possible

Dynamic trees: Conclusions (2)

Cannot be used in all situations:

- . Not adapted to top-down visiting the dynamic trees
- . Not adapted to searching a value in the forest/a dynamic tree
- . Aggregate operations do not allow multiplications

Outline

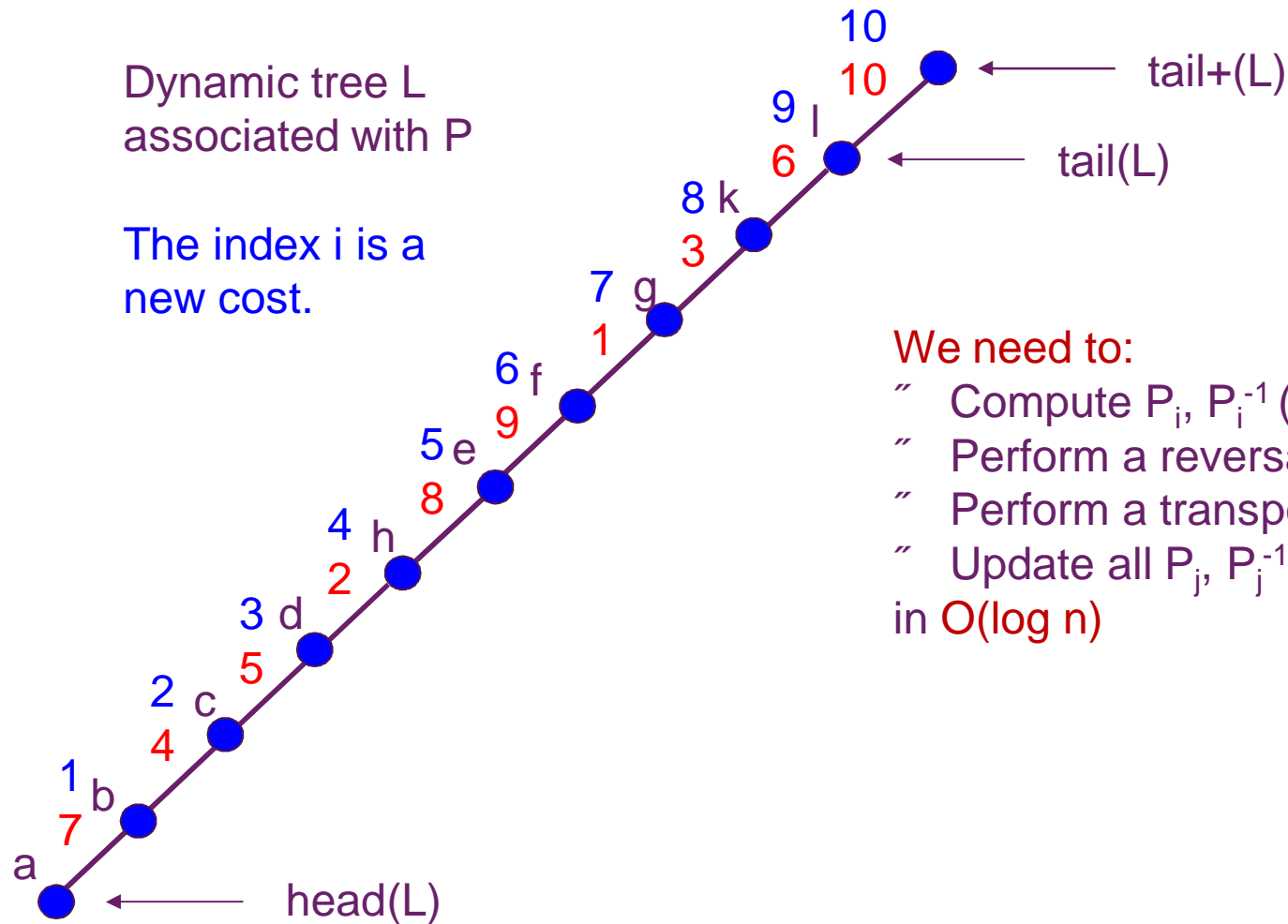
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A sequence is a filiform tree: Implement a sequence as a dynamic tree

P: 7 4 5 2 8 9 1 3 6 10

Dynamic tree L associated with P

The index i is a new cost.



We need to:

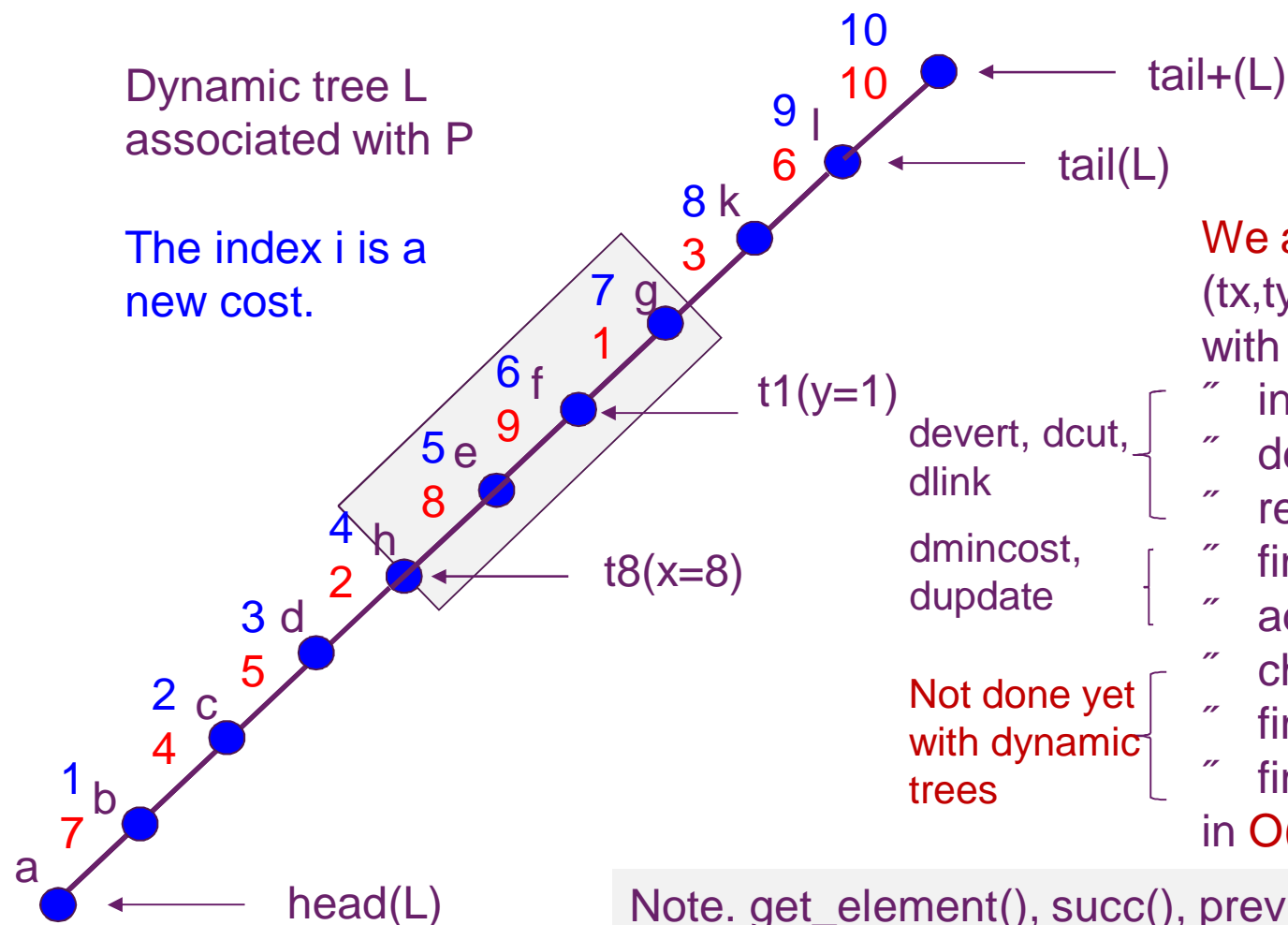
- " Compute P_i, P_i^{-1} (fixed i)
 - " Perform a reversal
 - " Perform a transposition
 - " Update all P_j, P_j^{-1}
- in $O(\log n)$

A sequence is a filiform tree: what we get from dynamic trees

P: 7 4 5 2 8 9 1 3 6 10

Dynamic tree L
associated with P

The index i is a
new cost.



We are able to do more
(tx,ty designate nodes
with edge costs x, y) :

- devert, dcut, { " insert(L,L₁,tx)
- dlink { " delete(L,tx,ty)
- { " reverse(L,tx,ty)
- dmincost, { " find_min(L,tx,ty) (or max)
- dupdate { " add(L,tx,ty,u)
- { " change_sign(L,tx,ty)
- Not done yet { " find_rank(L,tx) (=P⁻¹[x])
- with dynamic { " find_element(L,i) (=P[i])
- trees in **O(log n)**

Note. get_element(), succ(), prev() in O(log n)

A sequence is a filiform tree: what remains to be done

1) Perform

Not done yet
with dynamic
trees

- " change_sign(L,tx,ty)
- " find_rank(L,tx) ($=P^{-1}[x]$)
- " find_element(L,i) ($=P[i]$)

2) Update the index for delete, insert, reverse

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A log-list is a filiform tree too: Adjustments (1)

1) Perform

Not done yet
with dynamic
trees

- ” change_sign(L,tx,ty)
- ” find_rank(L,tx) (=P⁻¹[x])
- ” find_element(L,i) (=P[i])

Quite easy : **find_element(L,i)**

devert(tail+(L))

expose(head(L))

put the list in its standard form

all the list is a solid path with binary
tree B

New: dsearchindex(B,i)

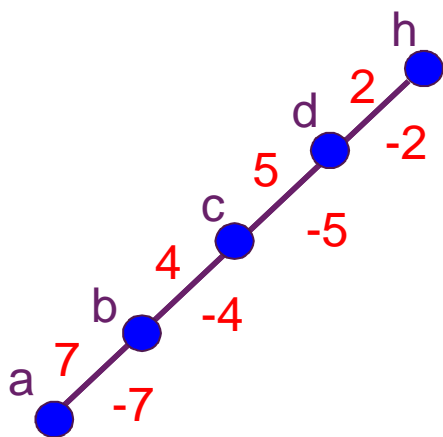
search i in the binary search tree B
with values **index()**

Time: $O(\log n)$

A log-list is a filiform tree too

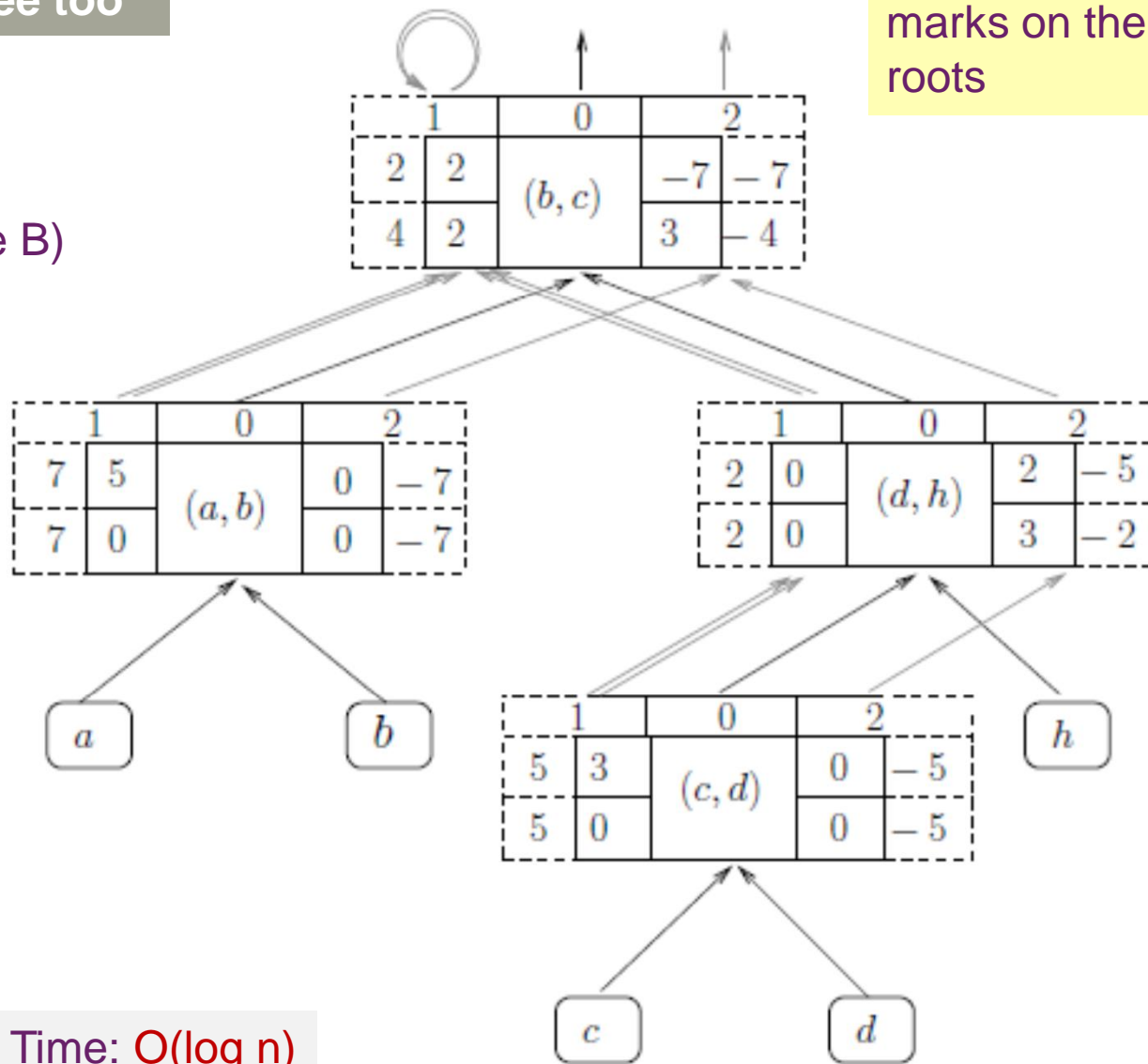
change_sign(L,tx,ty):
 devert(dparent(ty)),
 expose(tx) (binary tree B)
 dminuscost(B) **New**

Idea:
 (±)binary tree



cost() -cost()
 1-tree 2-tree
 positive tree negative tree

Multiply by -1:
 Change the marks on the roots



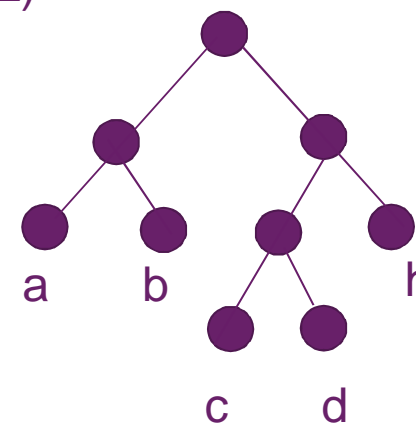
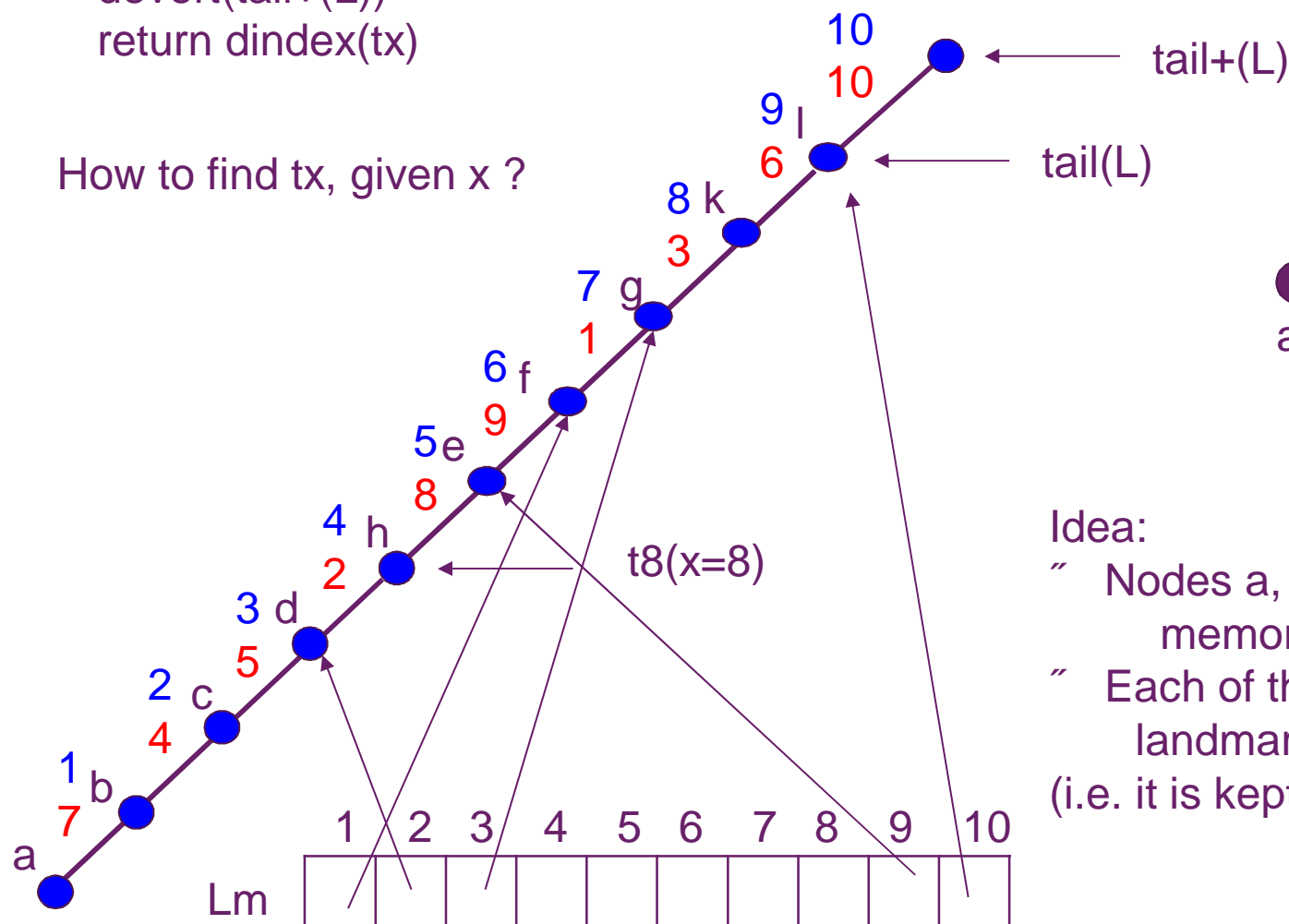
Time: $O(\log n)$

A log-list is a filiform tree too: Adjustments(1ter)

`find_rank(L,tx)`
`devert(tail+(L))`
`return dindex(tx)`

P: 7 4 5 2 8 9 1 3 6 10

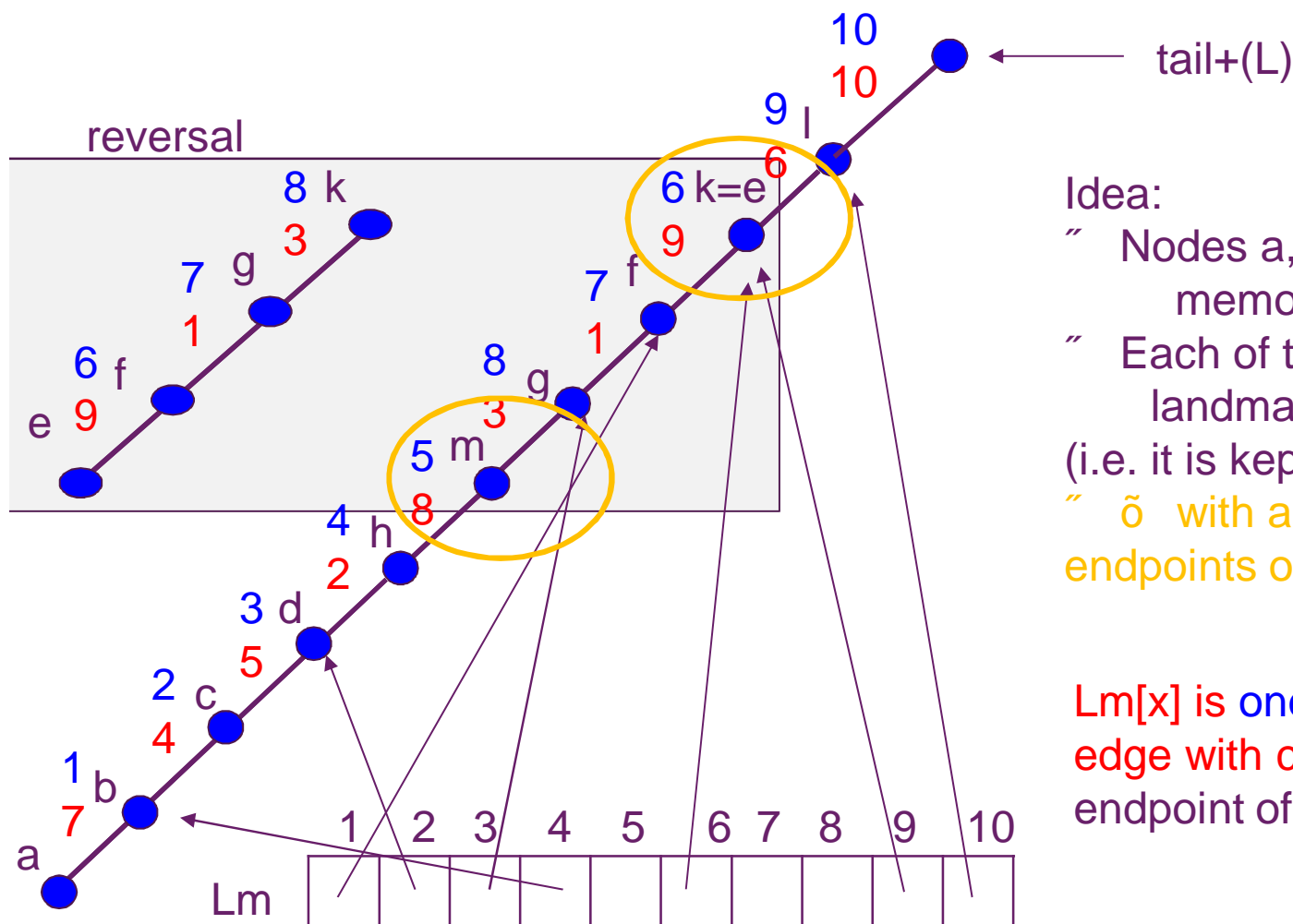
How to find tx, given x ?



- Idea:
- “ Nodes a, b, c, d are fixed memory cases
 - “ Each of them may be a landmark for a given element (i.e. it is kept close to the element)

A log-list is a filiform tree too: Adjustments(1ter)

P: 7 4 5 2 8 9 1 3 6 10



Idea:

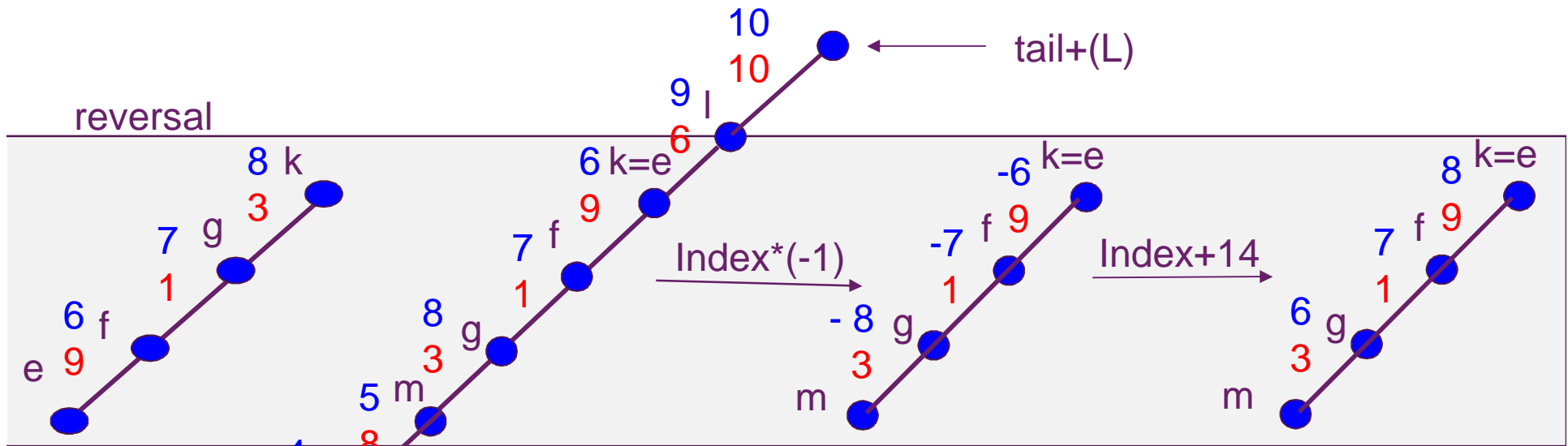
- “ Nodes a, b, c, \tilde{o} are fixed memory cases
- “ Each of them may be a landmark for a given element (i.e. it is kept close to the element)
- “ \tilde{o} with a few exceptions at the endpoints of the moved blocks

$Lm[x]$ is one of the endpoints of the edge with cost x ; tx is the lowest endpoint of it

Time: $O(\log n)$

A log-list is a filiform tree too: Adjustments(2)

2) Update the index for delete, insert, reverse



An example for reverse

Time: $O(\log n)$

A log-list is a filiform tree too: Summary

” What is a log-list ?

- . a filiform dynamic tree L (but temporarily a forest of such trees)
- . three pointers $\text{head}(L)$, $\text{tail}(L)$, $\text{tail}^+(L)$
- . a toolbox of $O(\log n)$ time operations

```
” get_element(tx)
” succ(tx)
” prec(tx)
” insert(L,L1,tx)
” delete(L,tx,ty)
” reverse(L,tx,ty)
” find_min(L,tx,ty) (or max)
” add(L,tx,ty,u)
” change_sign(L,tx,ty)
” find_rank(L,tx) (=P-1[x])
” find_element(L,i) (=P[i])
```

- . (if needed) a landmark table L_m (allows to compute tx)

A log list is a filiform tree too: Summary

” Shall we deduce that visiting all the elements in a log-list takes $O(n \log n)$?

No, visiting directly the binary tree allows to do this in $O(n)$ but needs to go deeper into the implementation

” What about searching a given element in a log-list?

If the list is not ordered, $O(n)$ as above

If the list is ordered and has distinct elements, the dynamic tree operation `dsearchcost` (we used it only for `cost=index`) does it in $O(\log n)$.

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- “ **Conclusion**

Conclusion

- “ Improving the algorithm for sorting by prefix reversals and prefix transpositions
 - . n steps, each identifying and performing a block operation
 - . $O(n \log n)$ time instead of $O(n^2)$ before
- “ In addition: four other variants of the sorting problem are improved
- “ Some of them resist however
- “ Log-list could have many other applications

Bibliography

- “ Daniel D. Sleator and Robert E. Tarjan . *A Data Structure for Dynamic Trees*, Journal of Computer and System Sciences 26, 362-291 (1983)
- “ Irena Rusu . *Log-Lists and Their Applications to Sorting by Transpositions, Reversals and Block-Interchanges*, arXiv 1507.01512 (2015).

Thank you !