#### The Tree Inclusion Problem: In Optimal Space and Faster

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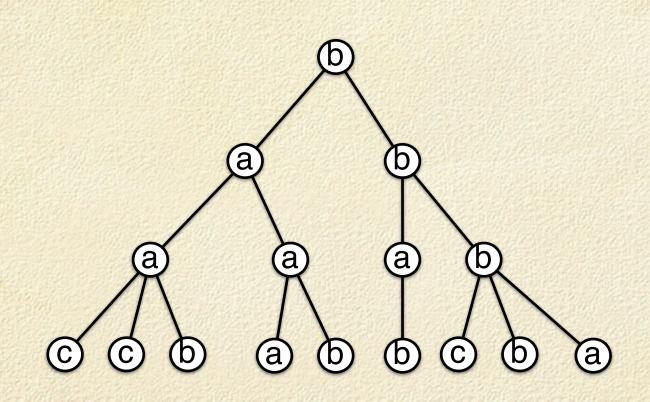
#### Basic setup

Trees are labeled, rooted, and ordered.

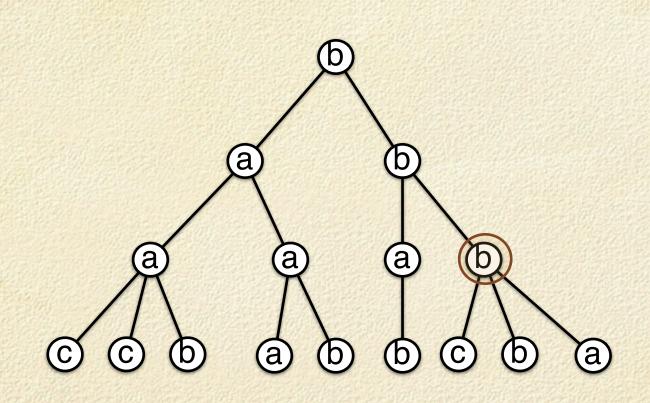
- Rooted: A specific node is designated as the root of the tree.
- □ **Labeled**: Each node is assigned a *label* from some alphabet  $\Sigma$ .
- Ordered: There is a left-to-right order among siblings.

We compare trees by deleting nodes.

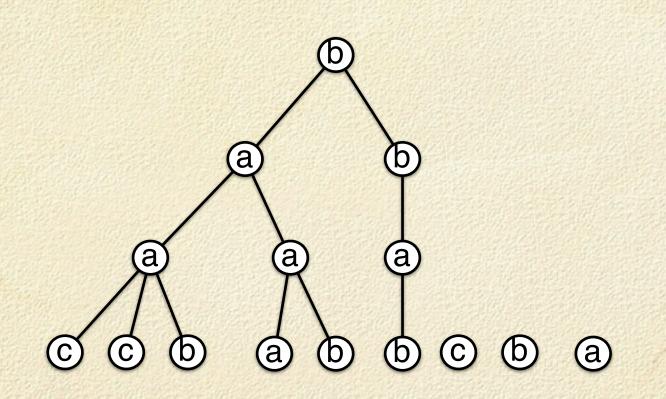




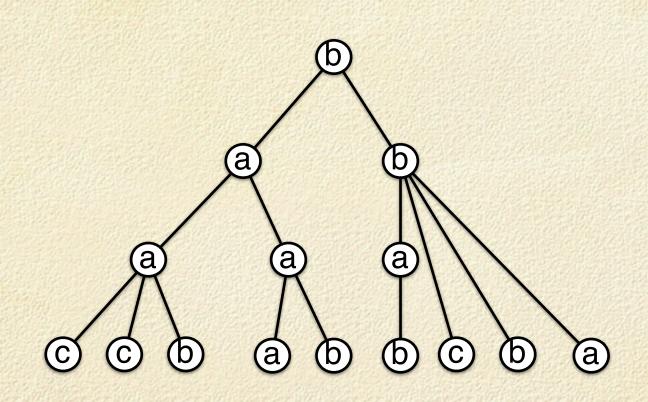






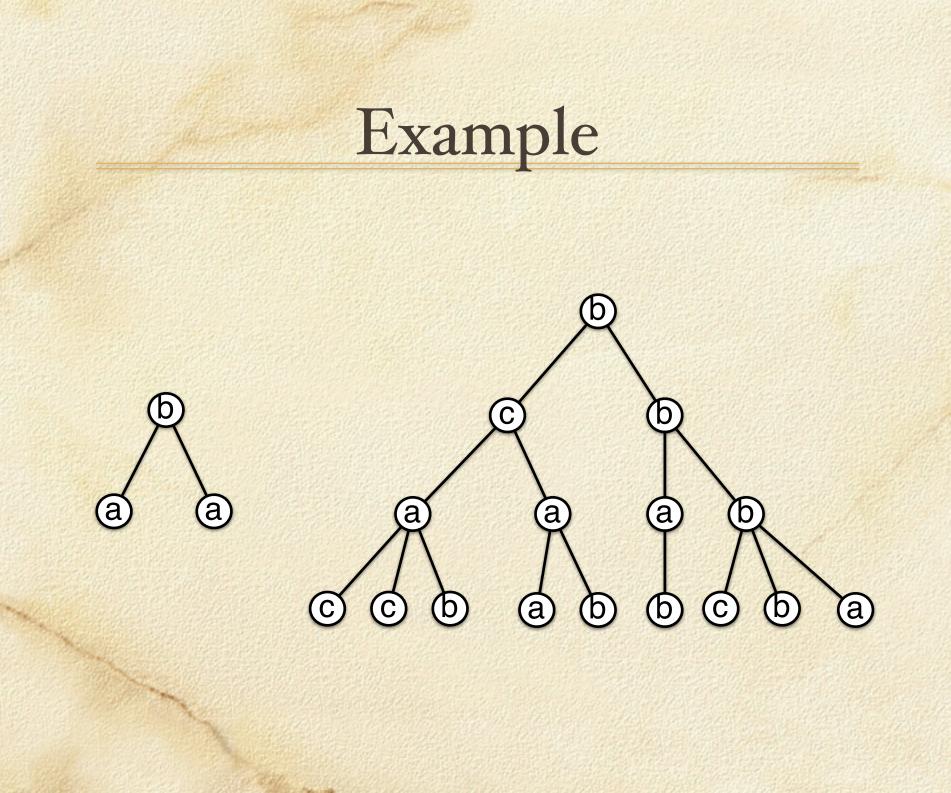


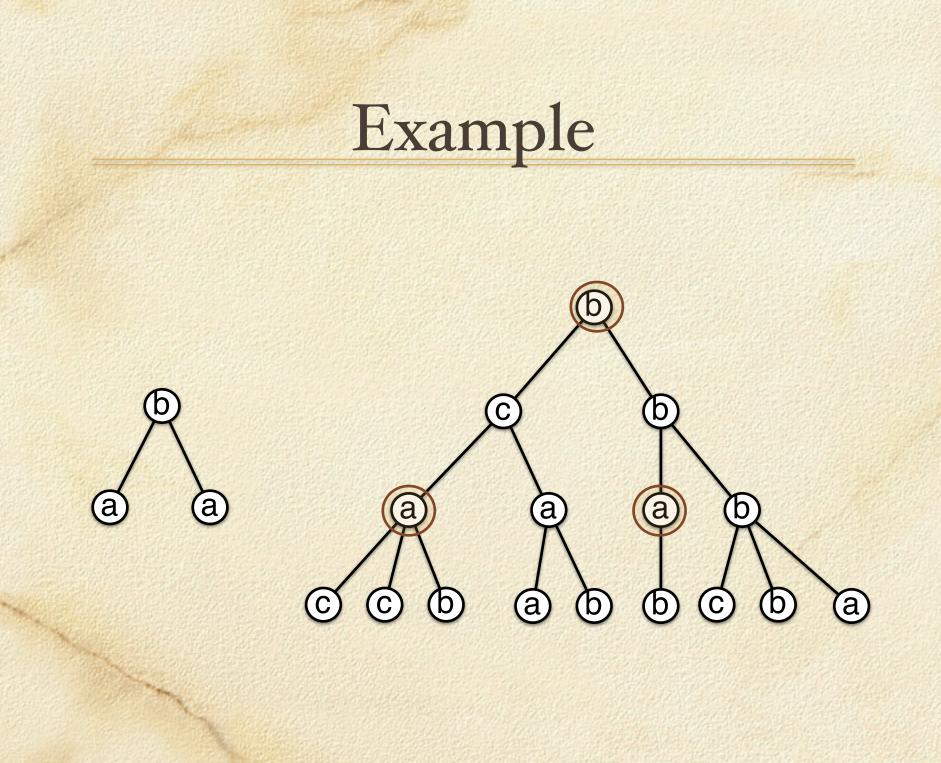


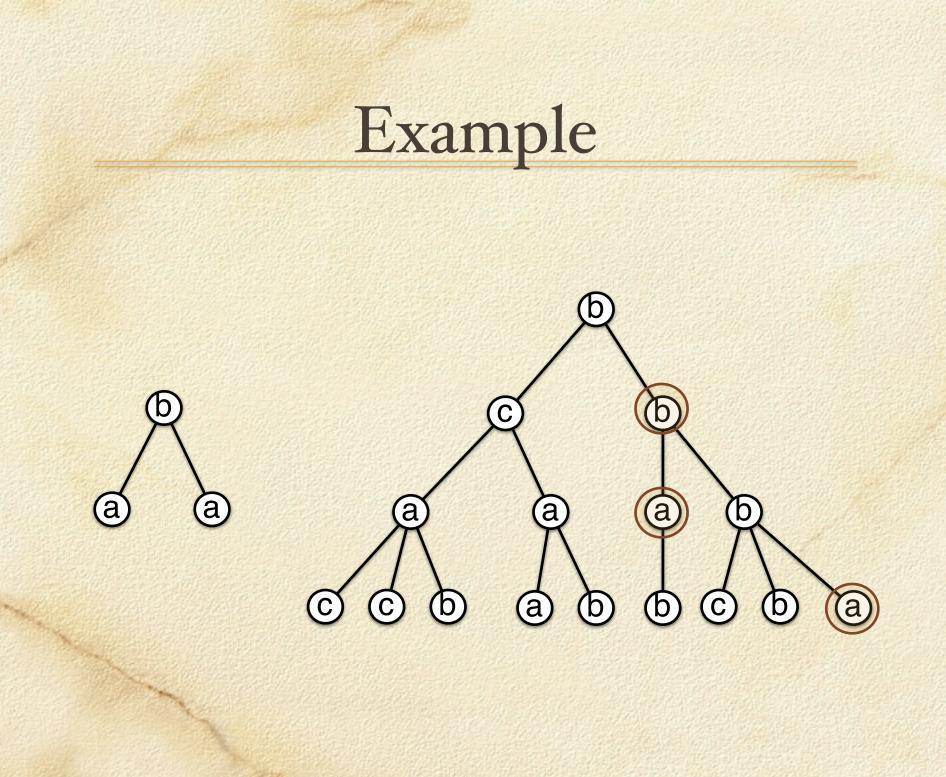


# **Tree Inclusion**

- P is *included* in T if P can be obtained from T by deleting nodes in T.
- P is *minimally included* in T if P is not included in any subtree of T.
- The tree inclusion problem. is to decide if P is included in T, and if so, compute all subtrees of T which minimally includes P.

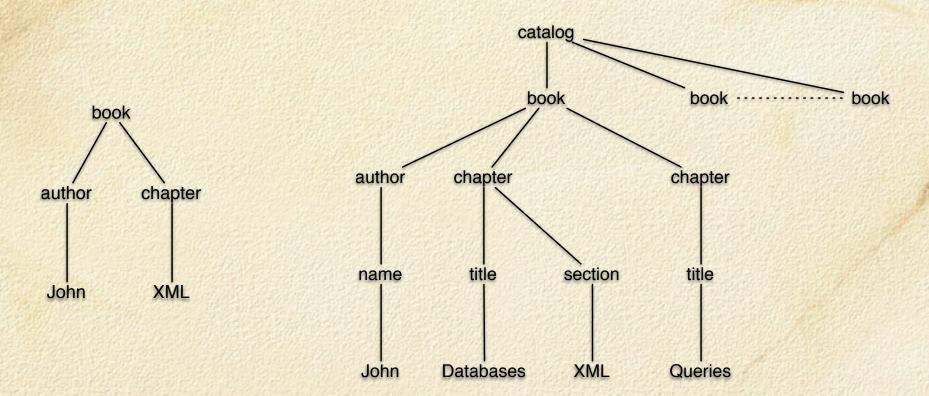






Results		
Time	Space	Reference
$O(n_P n_T)$	$O(n_P n_T)$	[KM92 ]
$O(l_P n_T)$	$O(l_P \min(d_T, l_T))$	[Che98]
$O(l_P n_T)$ $O(n_P l_T \log \log n_T)$ $O(\frac{n_P n_T}{O(\frac{n_P n_T}{\log n_T})}$	$O(n_P + n_T)$	This paper

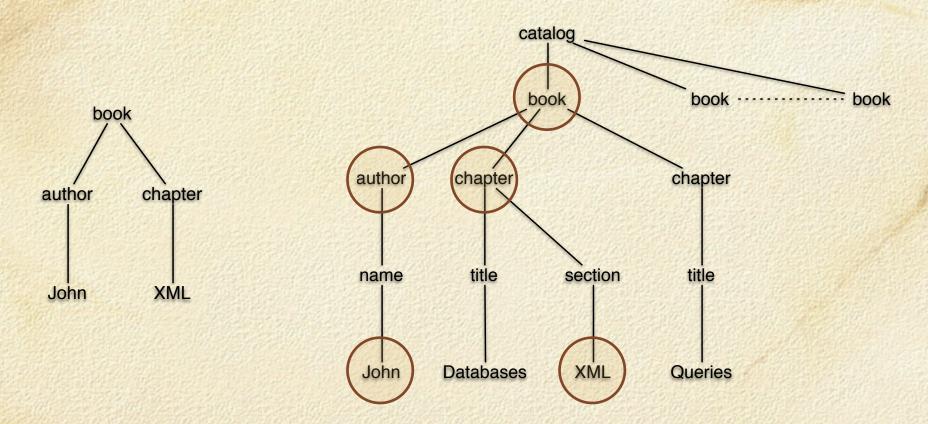
# XML example



Query: "Find all books written by John with a chapter that has something to do with XML".

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# XML example



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#### Practical implications

Space reduction from quadratic to linear:
Possible to query significantly larger XML databases.

Faster query time since more computation can be kept in main memory.

# Embeddings

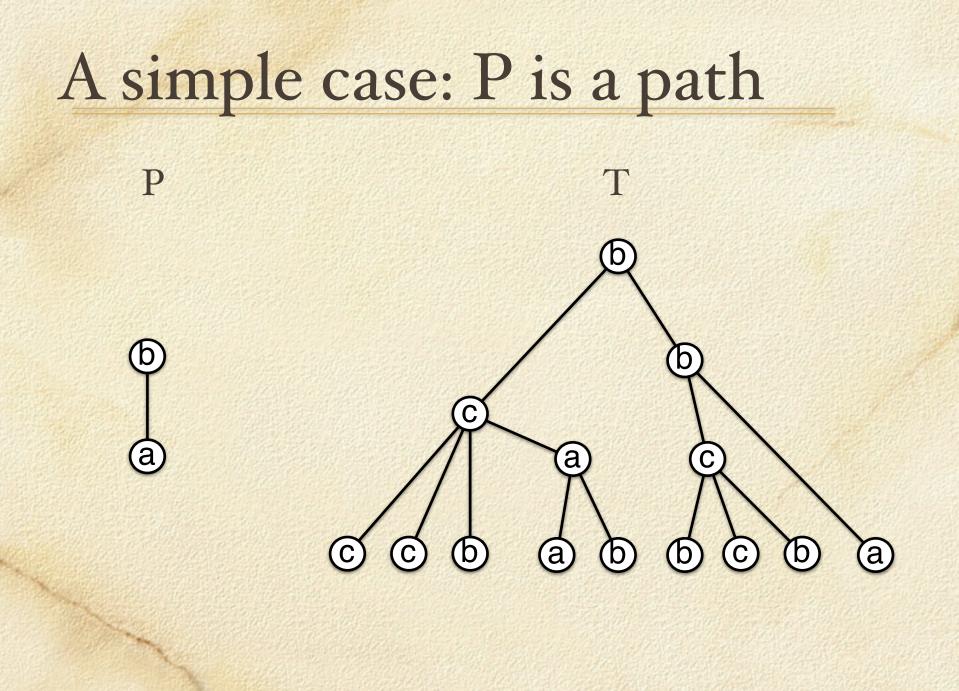
An injective function from the nodes of P to T is an *embedding* if:

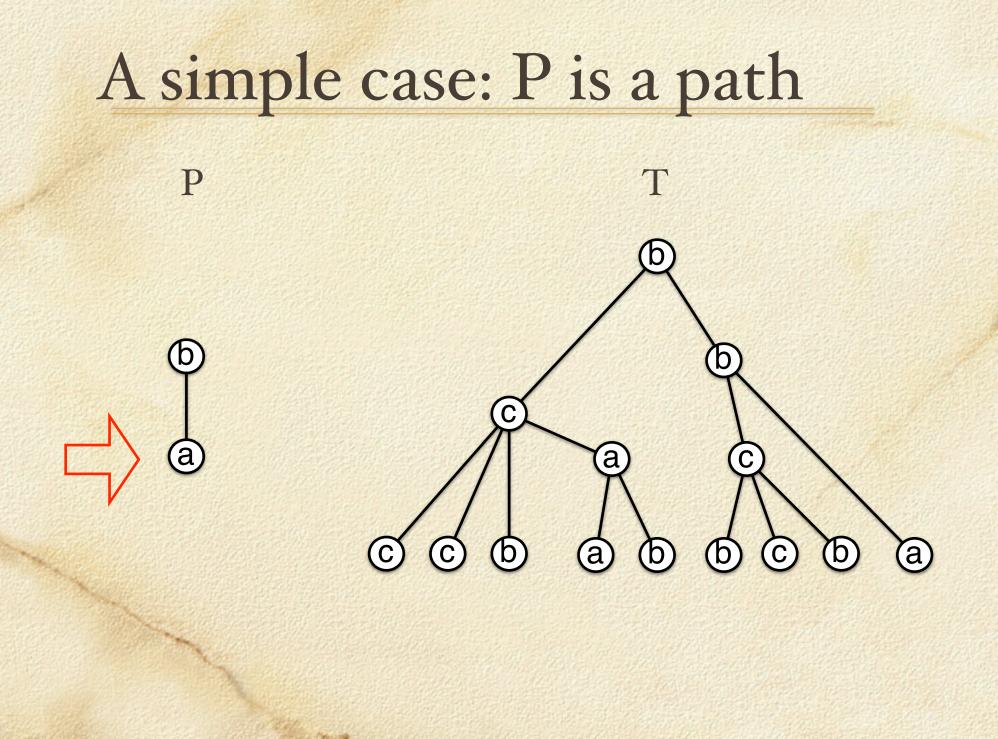
 $\square$  label(v) = label(f(v)),

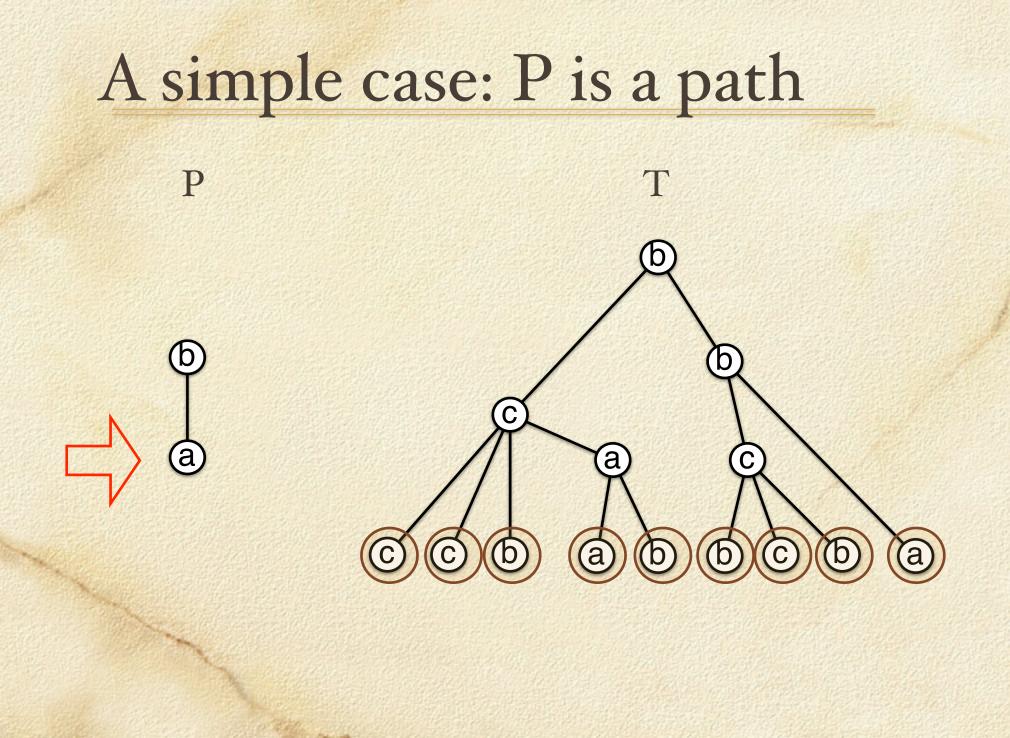
 $\Box$  v is ancestor of w iff f(v) is an ancestor of f(w),

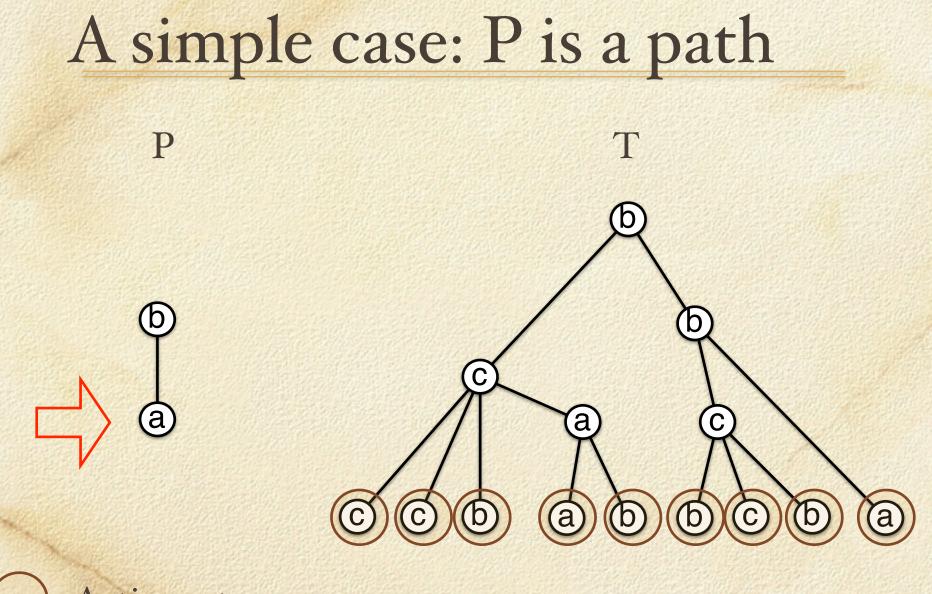
 $\Box$  v is to the left of w iff f(v) is to the left of f(w).

P is included in T iff there is an embedding from P to T.

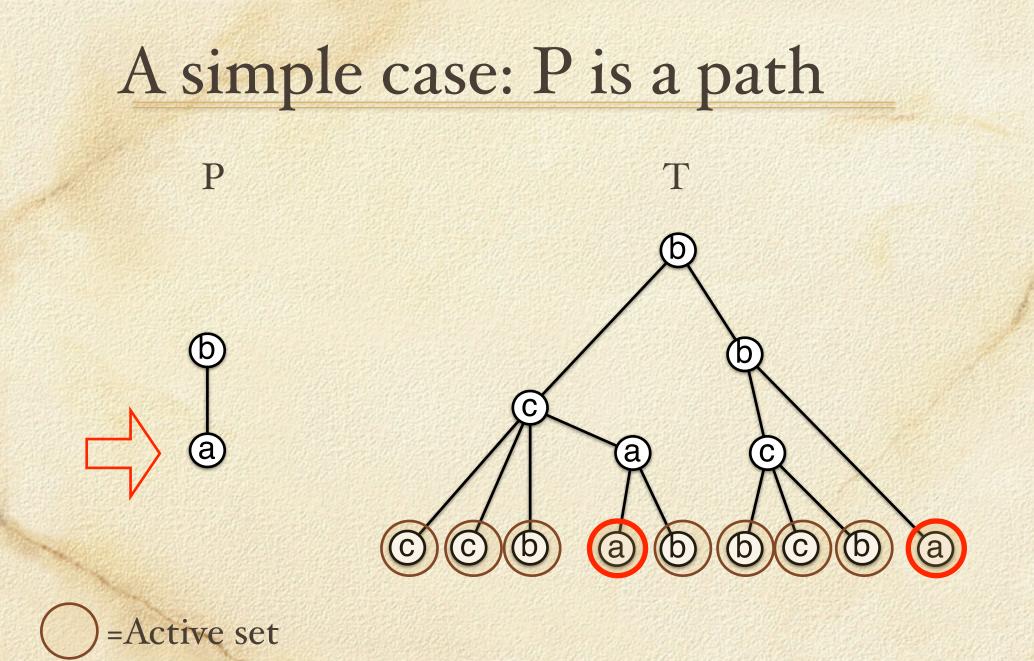


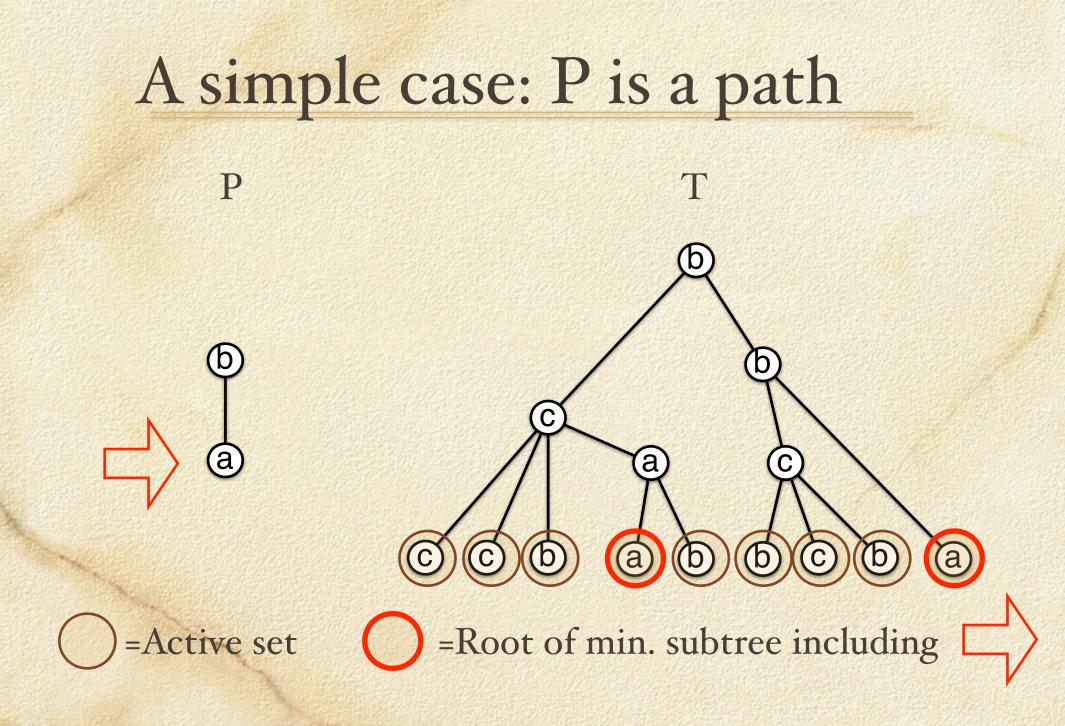


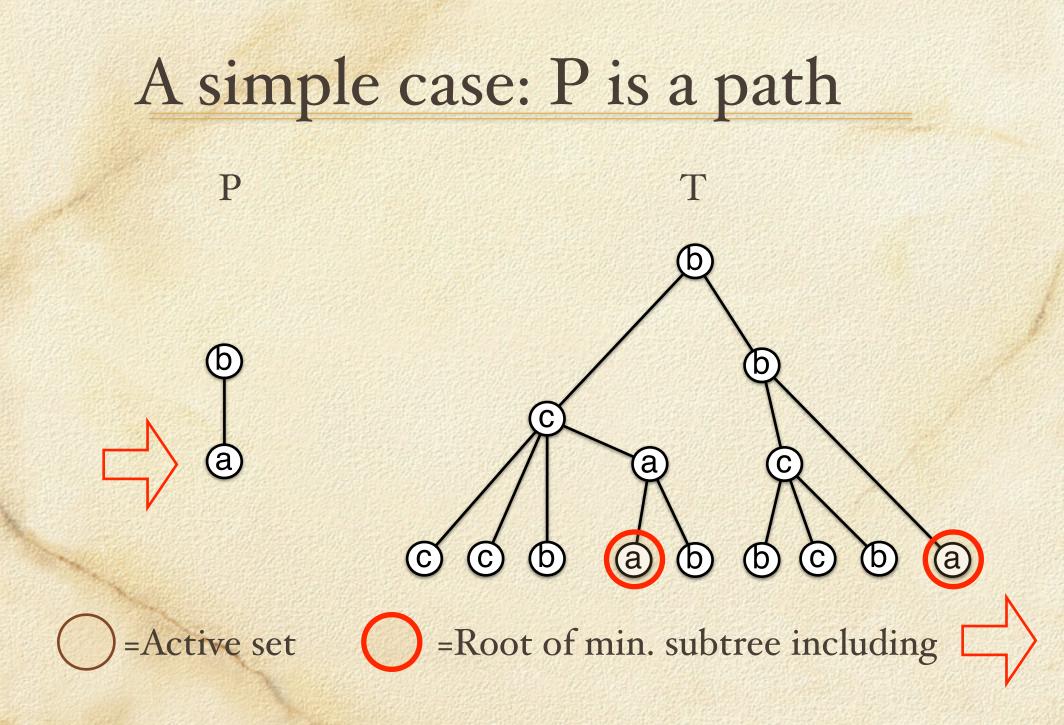


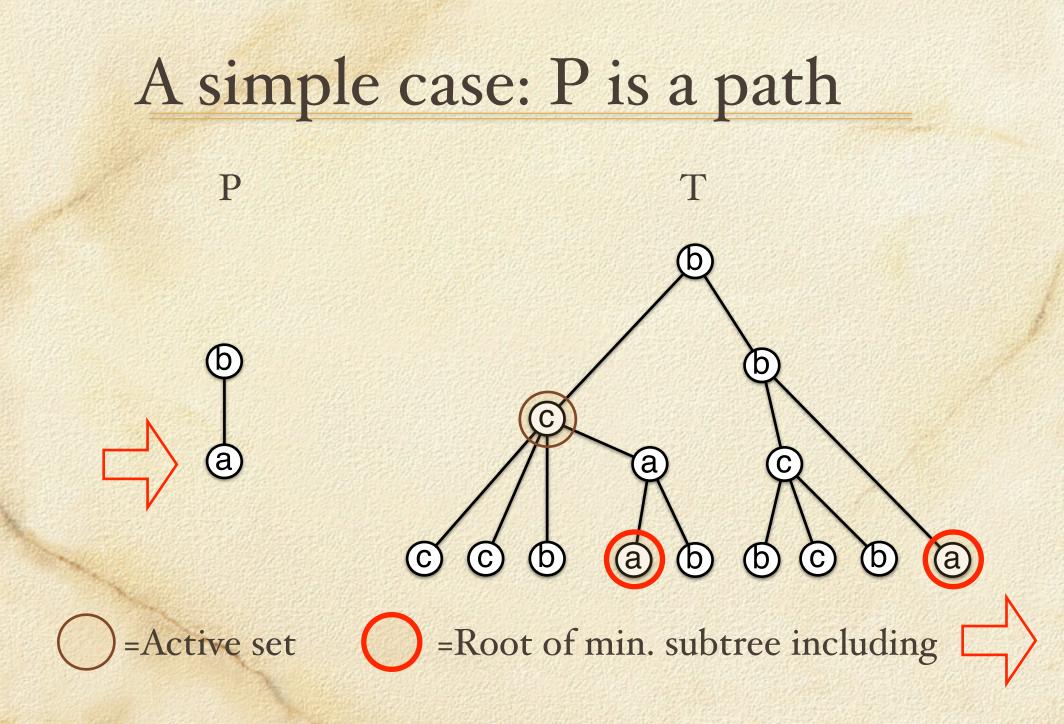


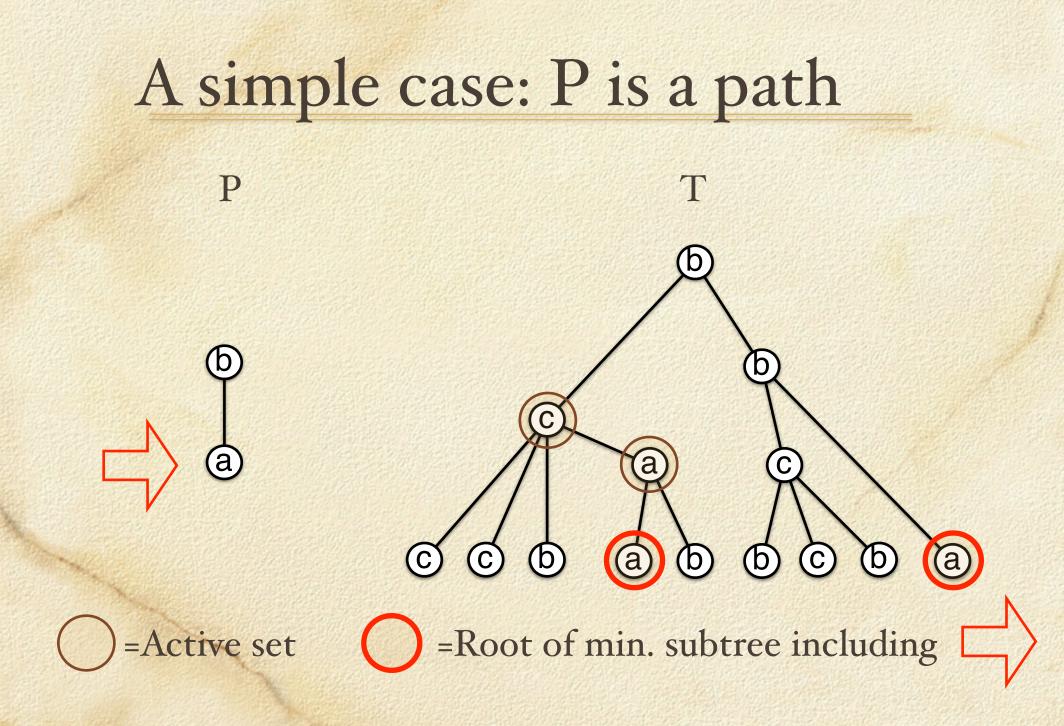


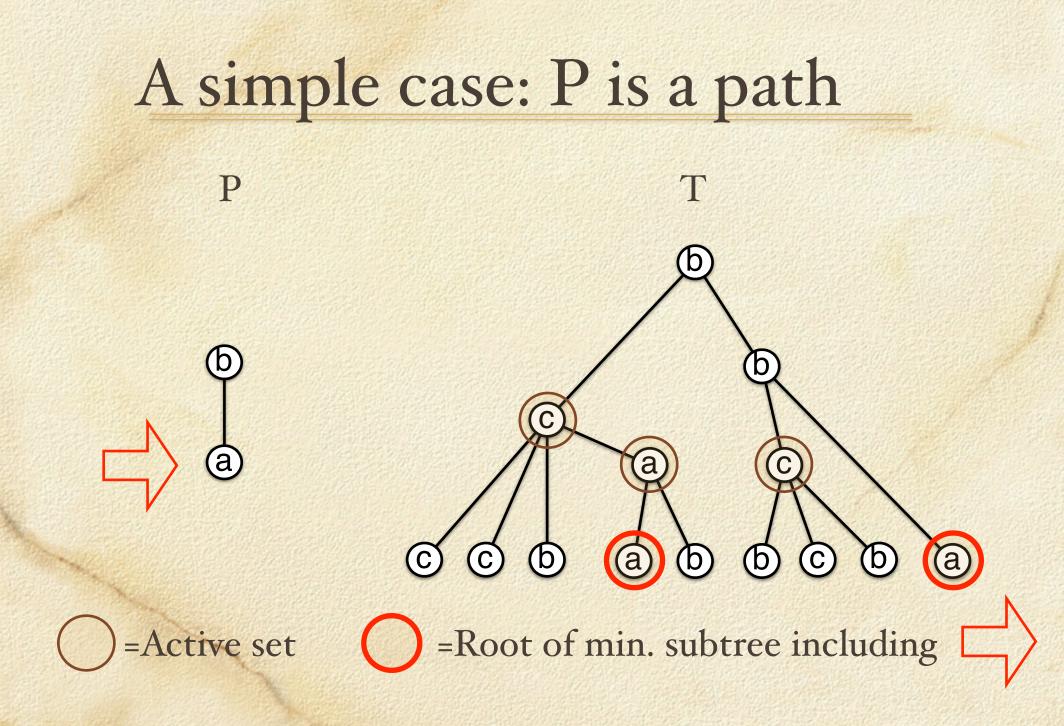


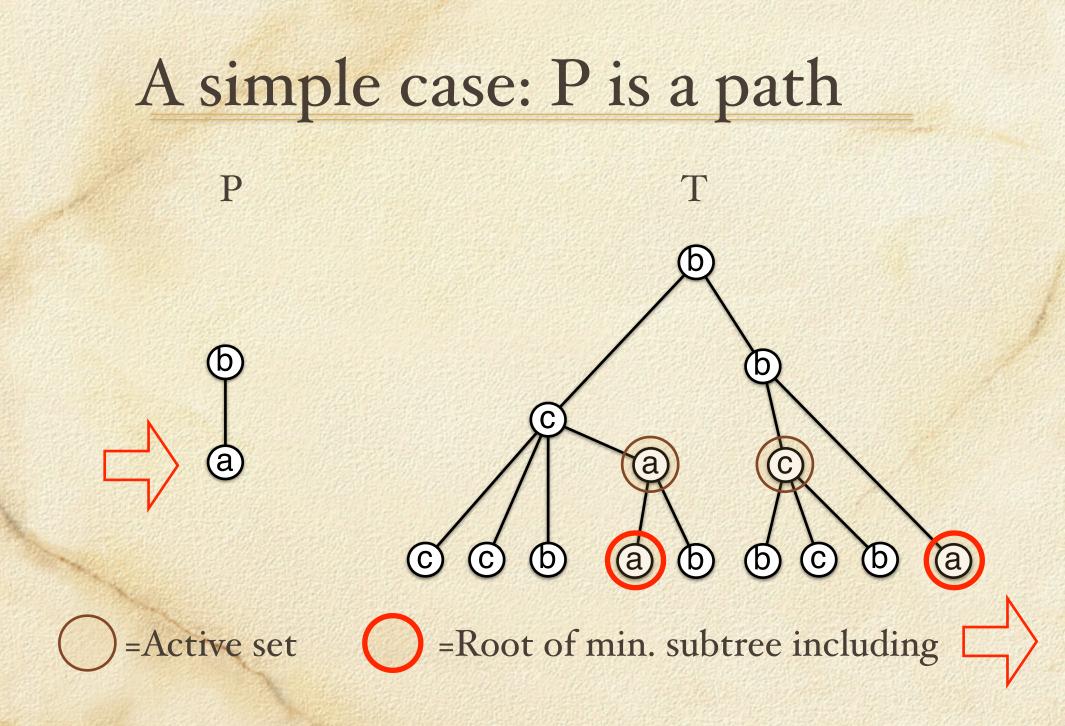


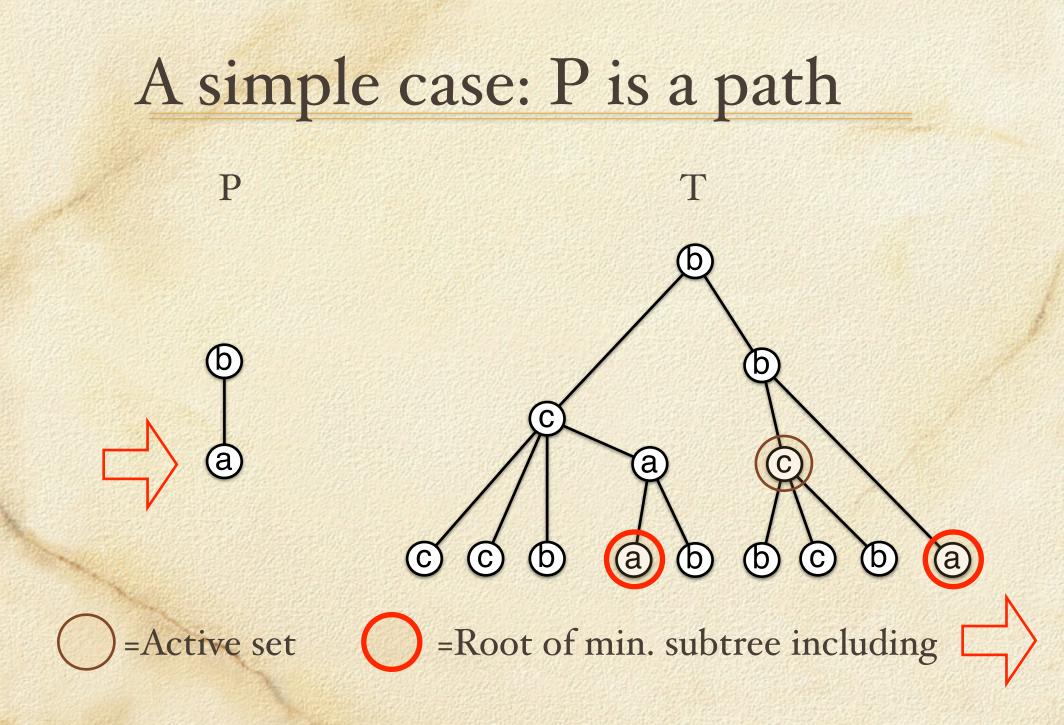


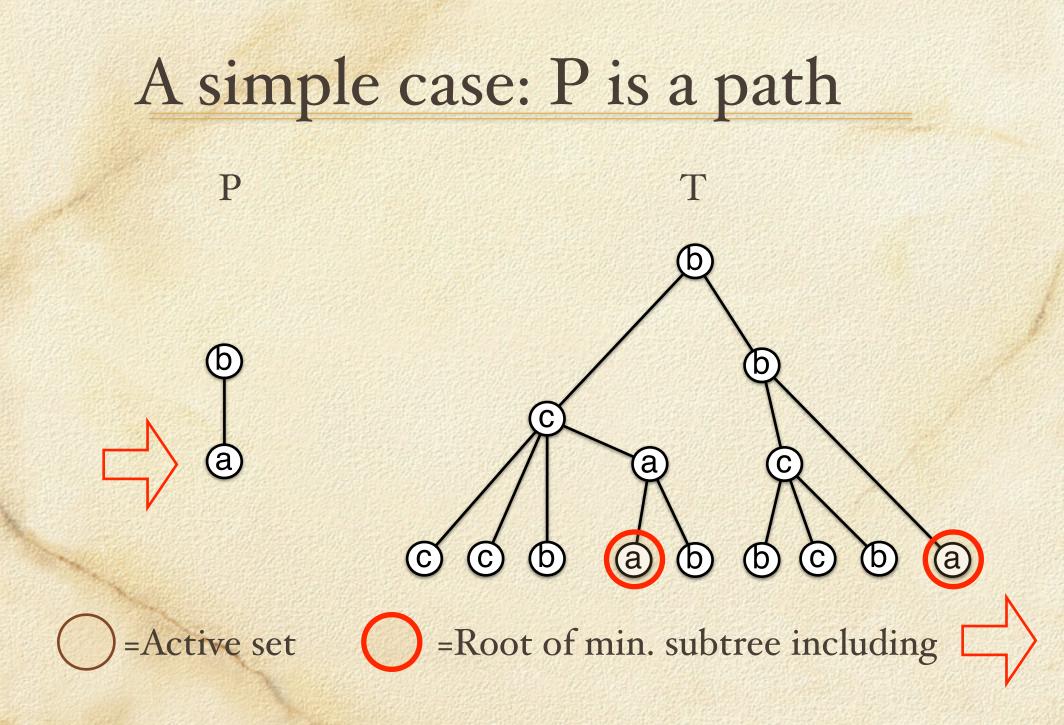


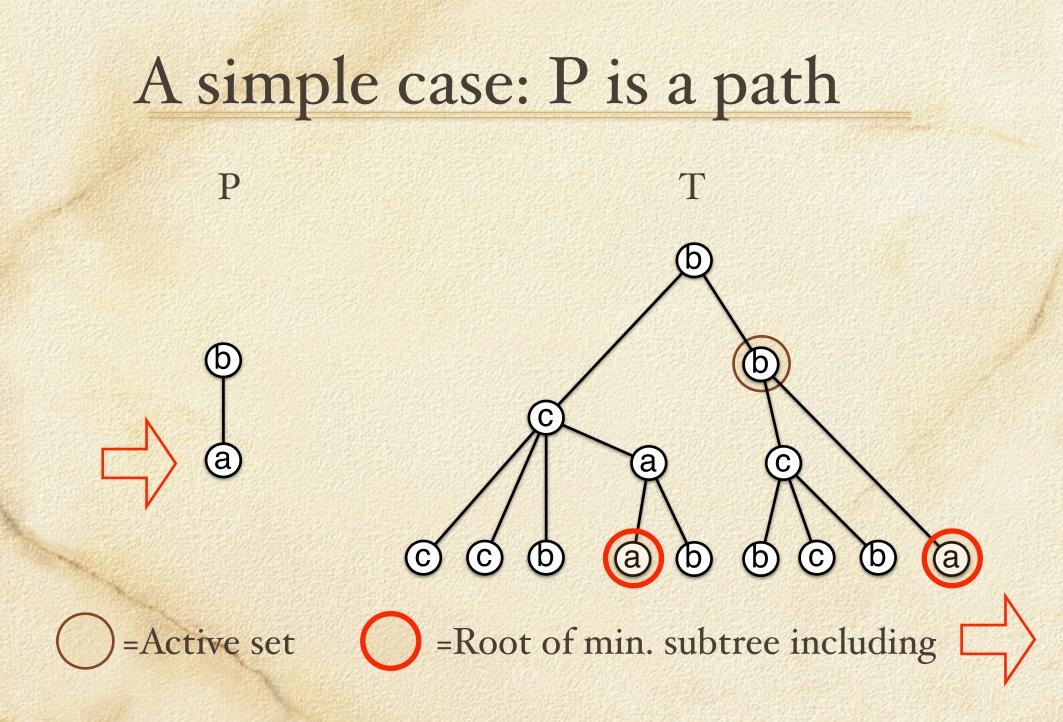


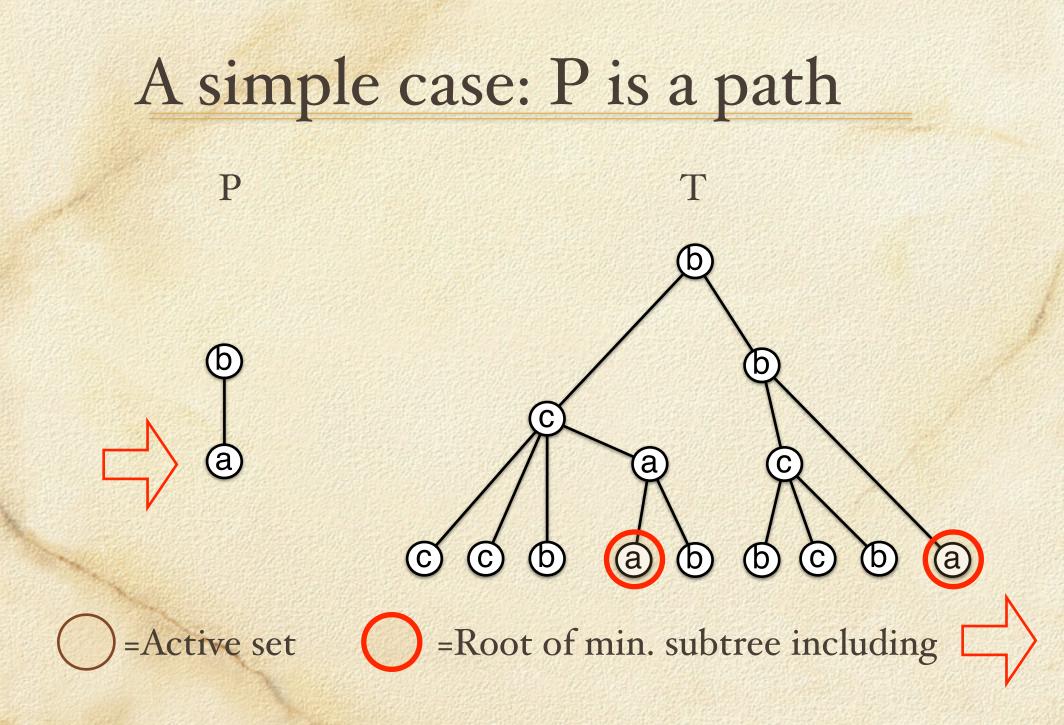


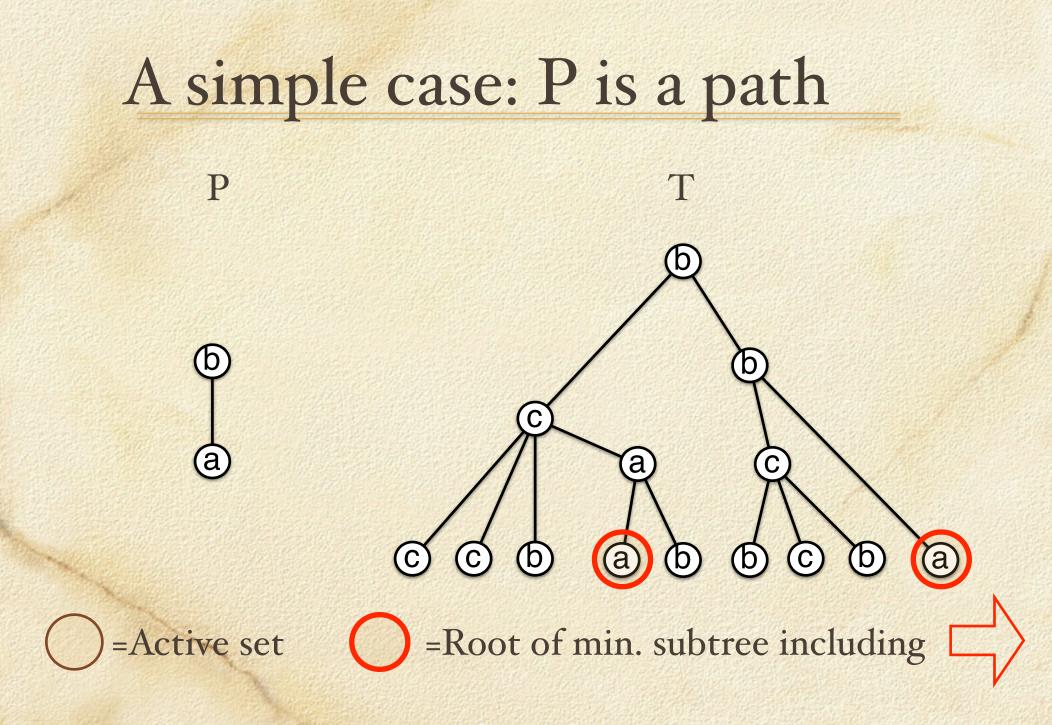


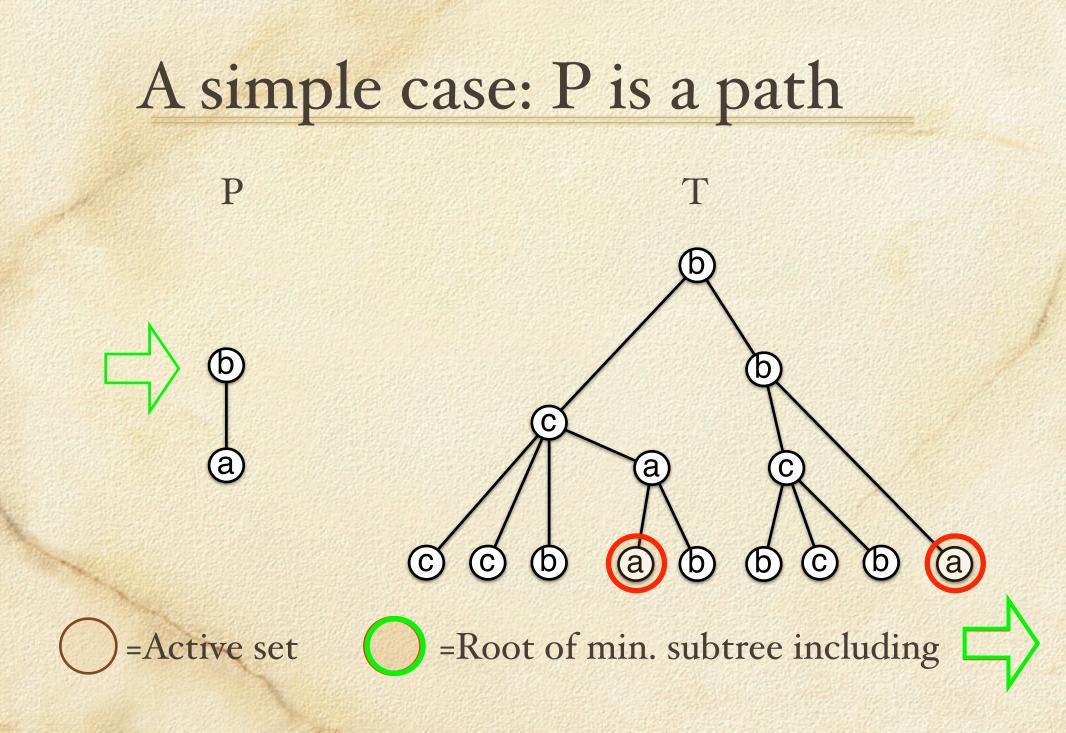


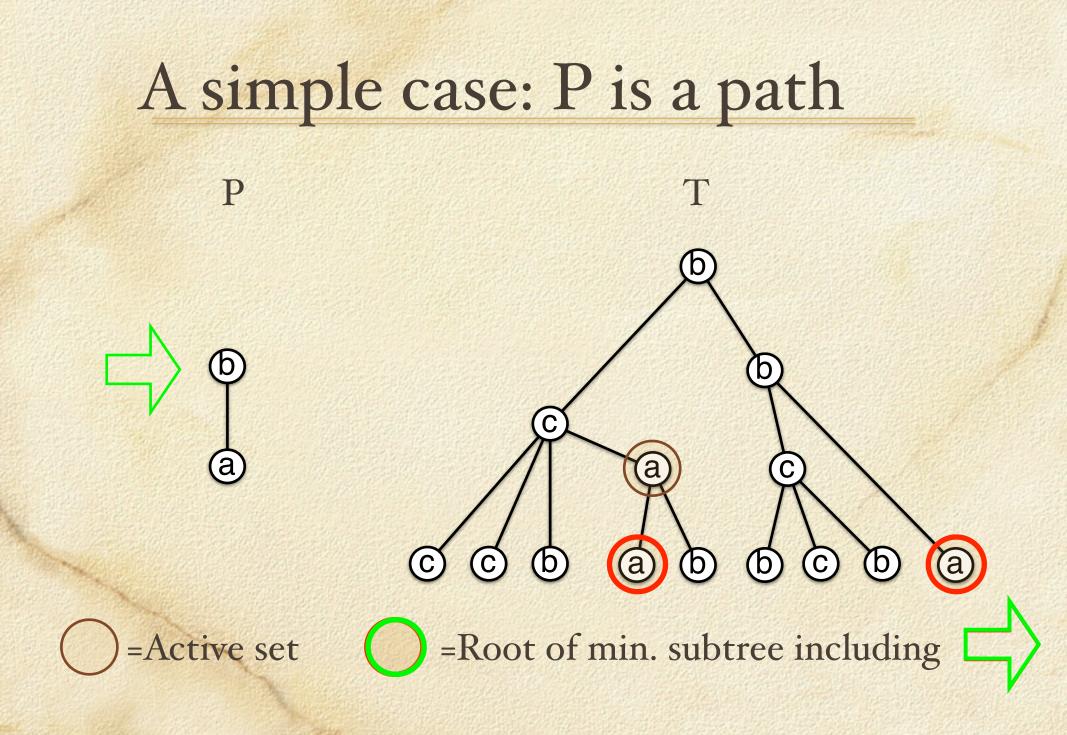


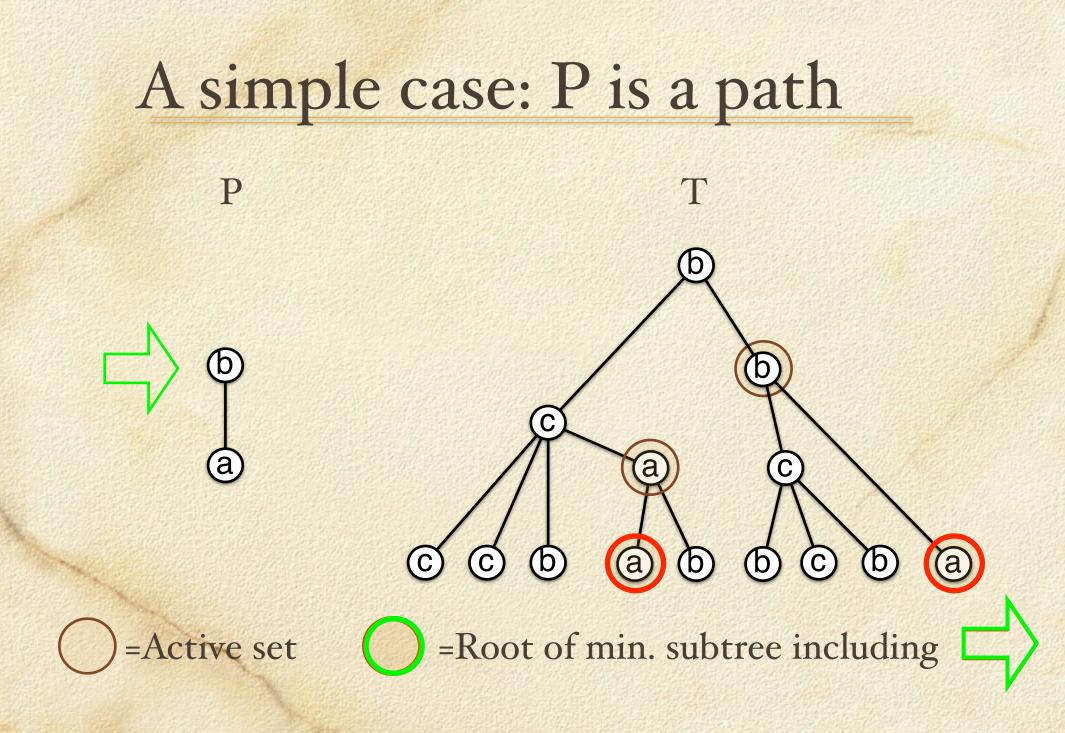


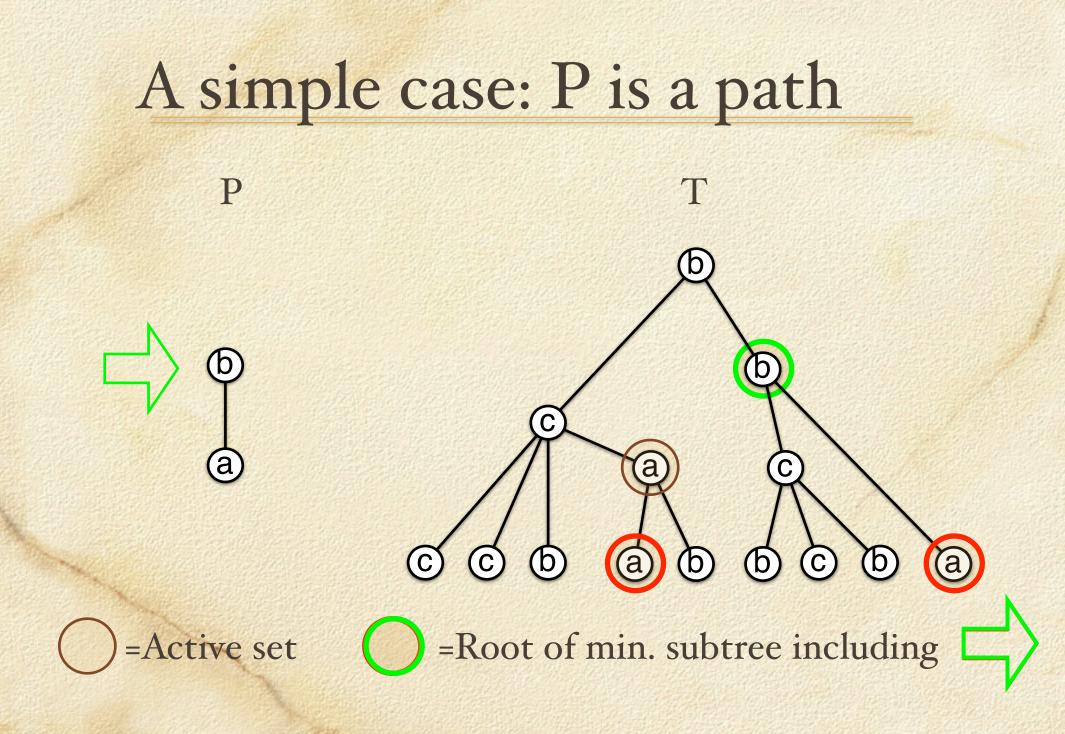


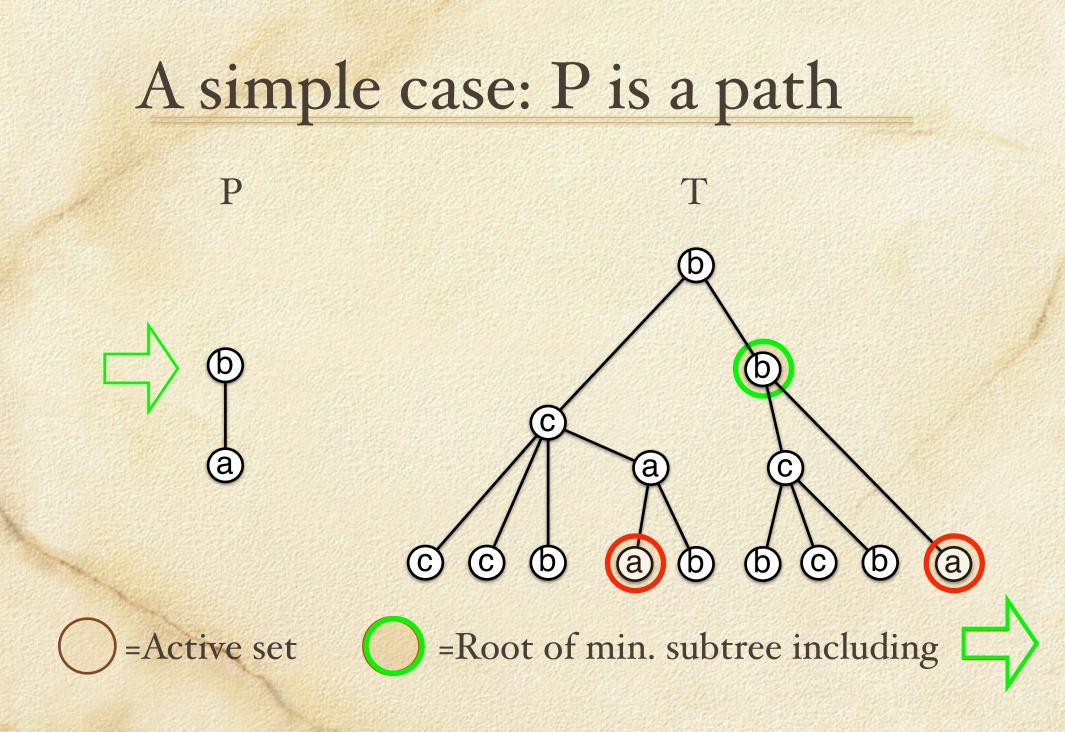


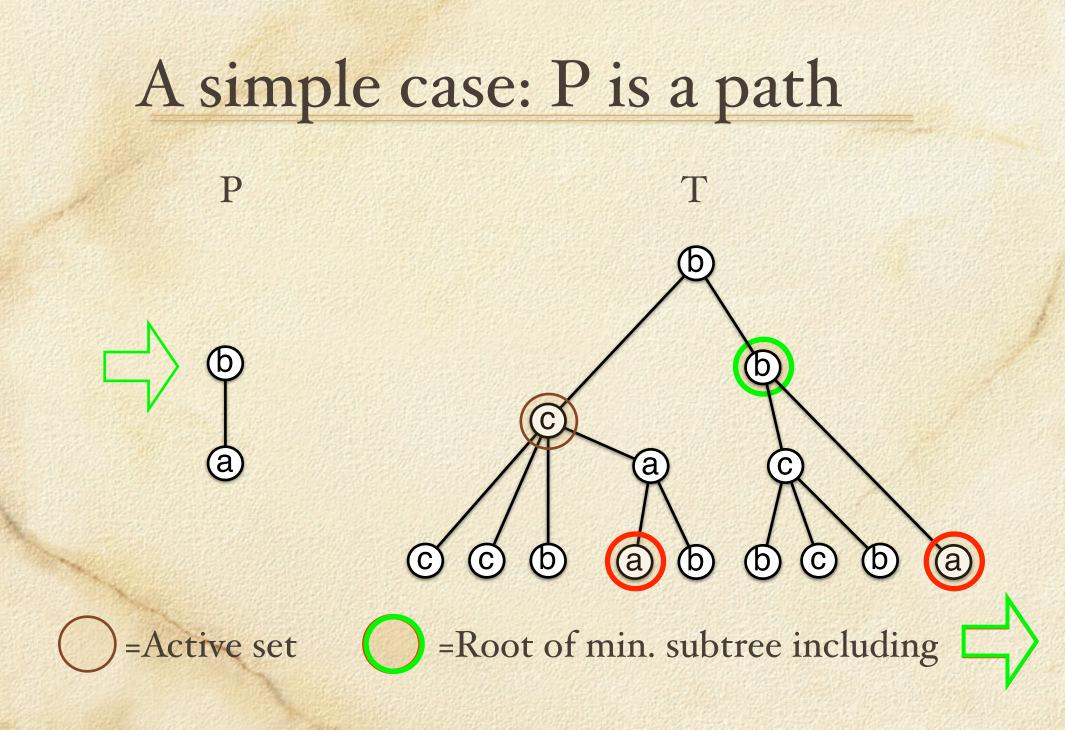


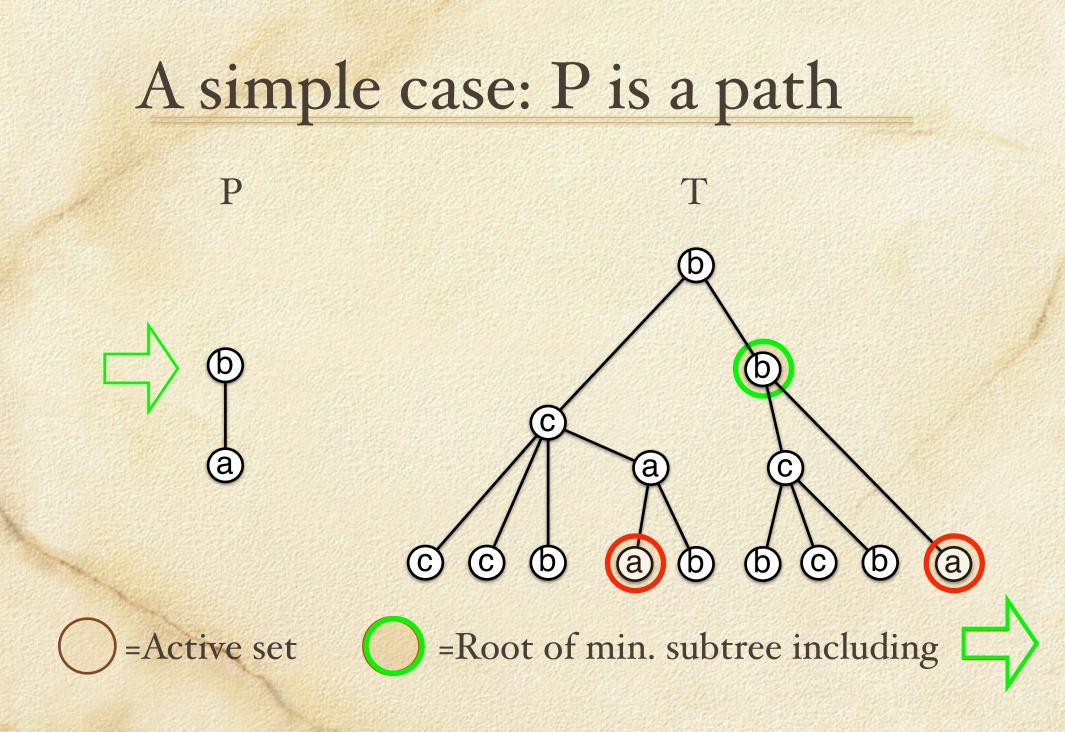


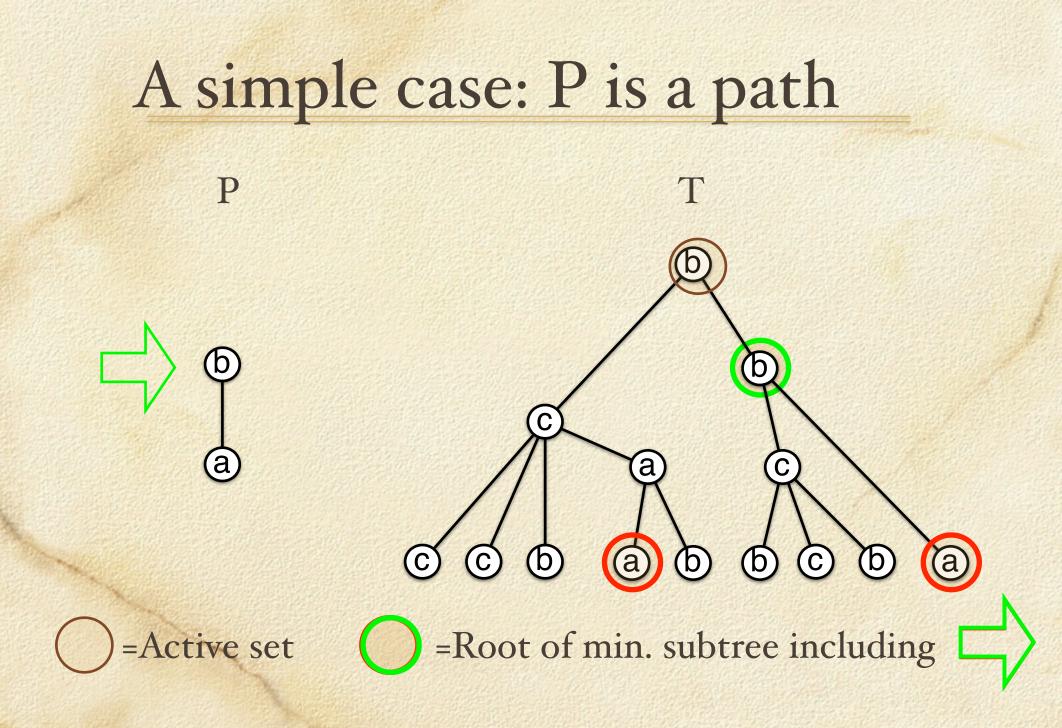


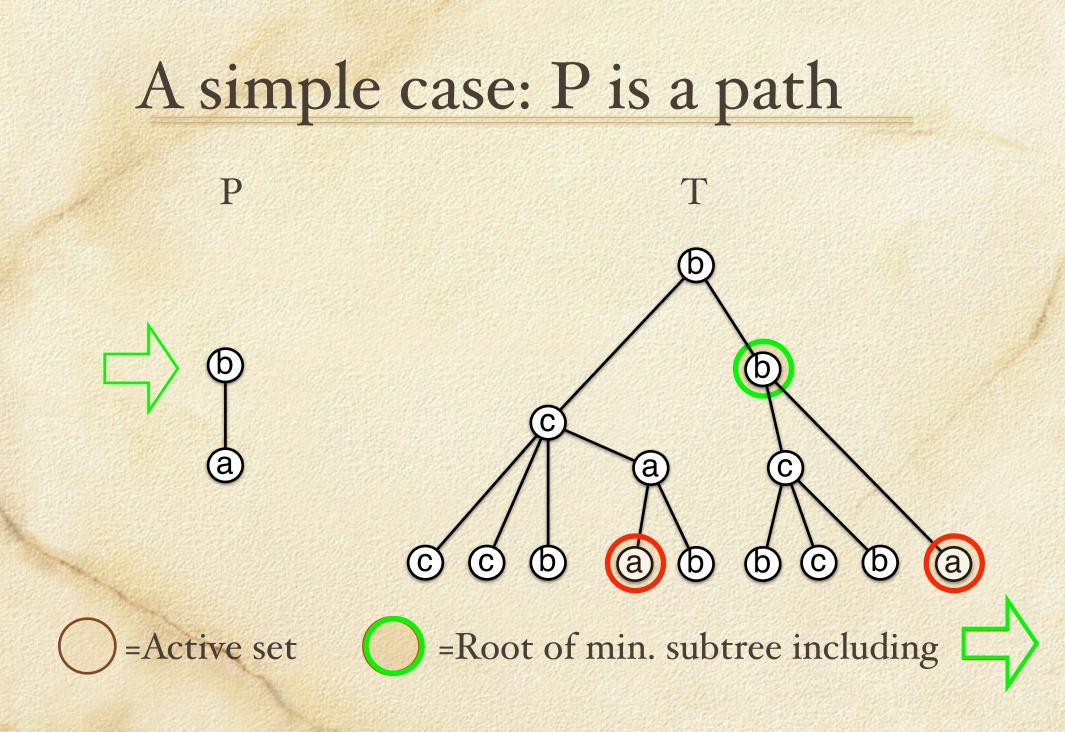






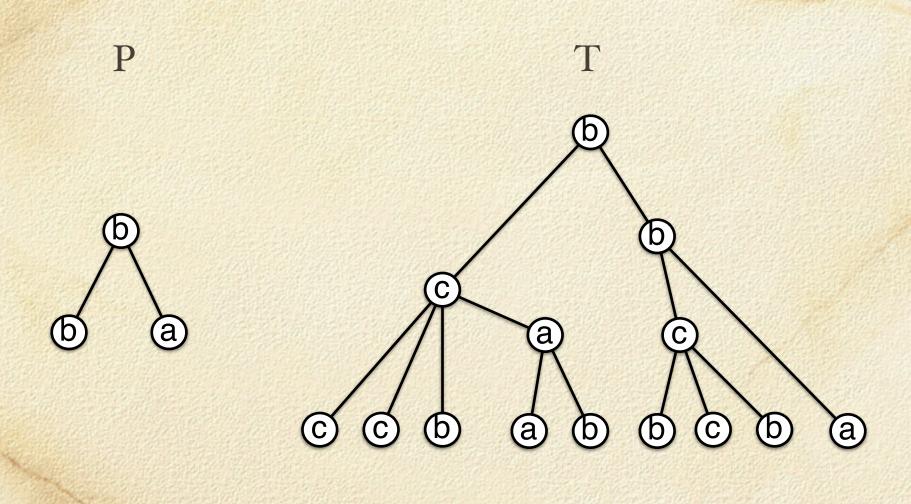


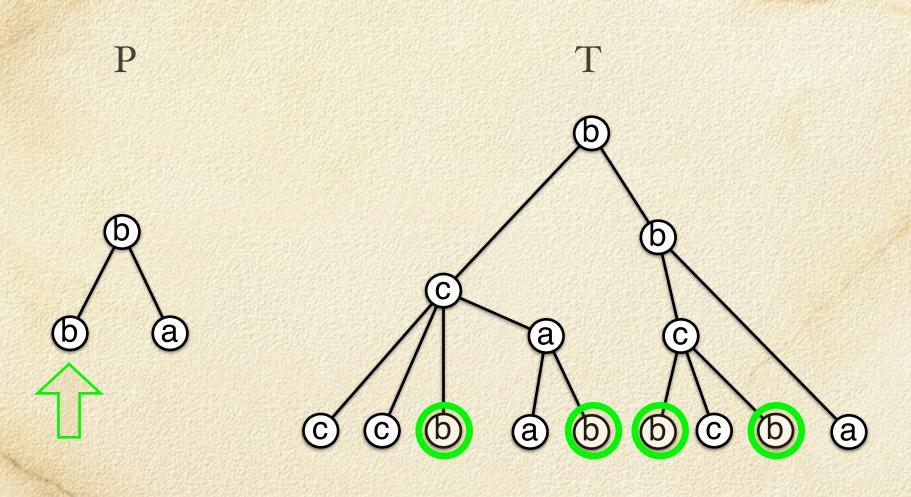


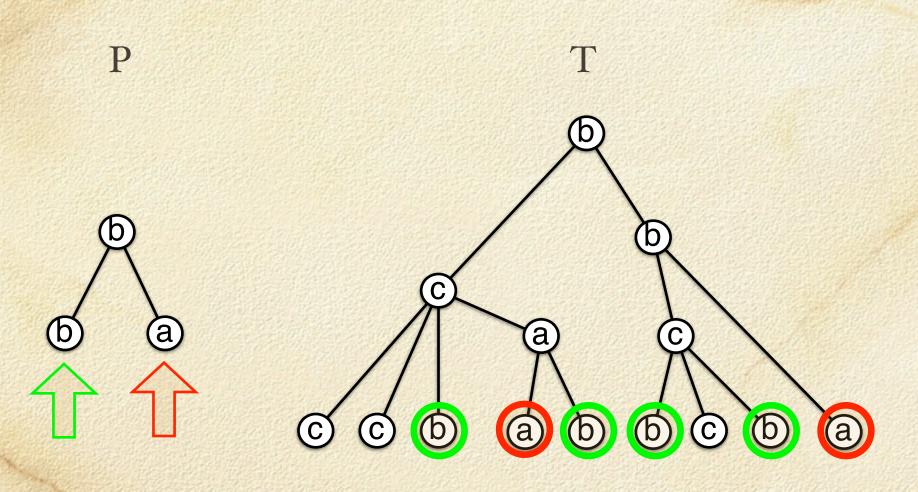


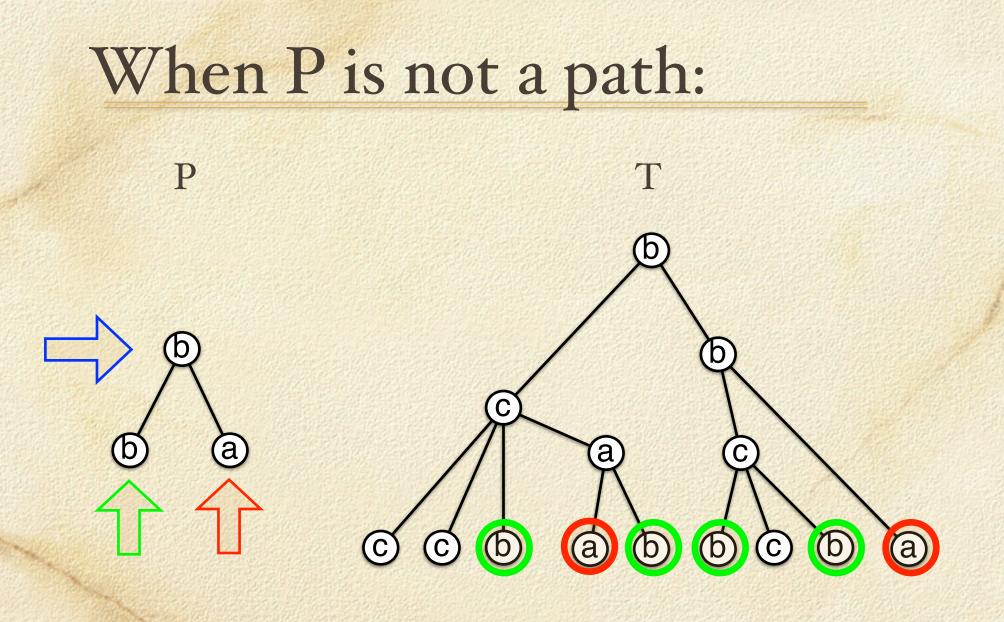
## Complexity

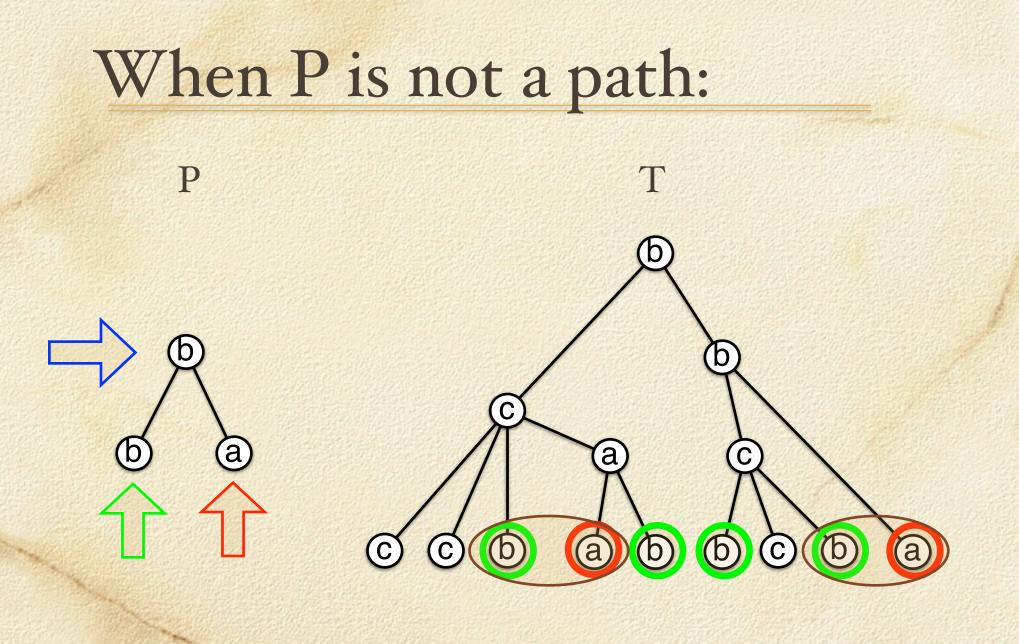
- At each step of the algorithm the active set "moves up".
- Each parent pointer in T is traversed a constant number of times.
- Using a simple data structure and exploiting the ordering of the nodes we get a total running time of O(n<sub>T</sub>).

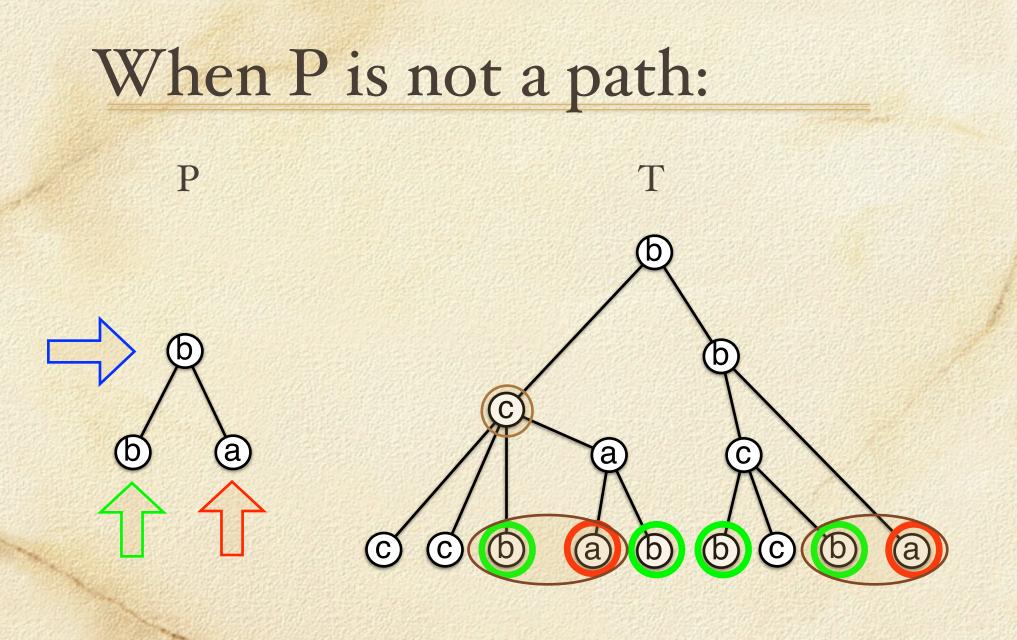


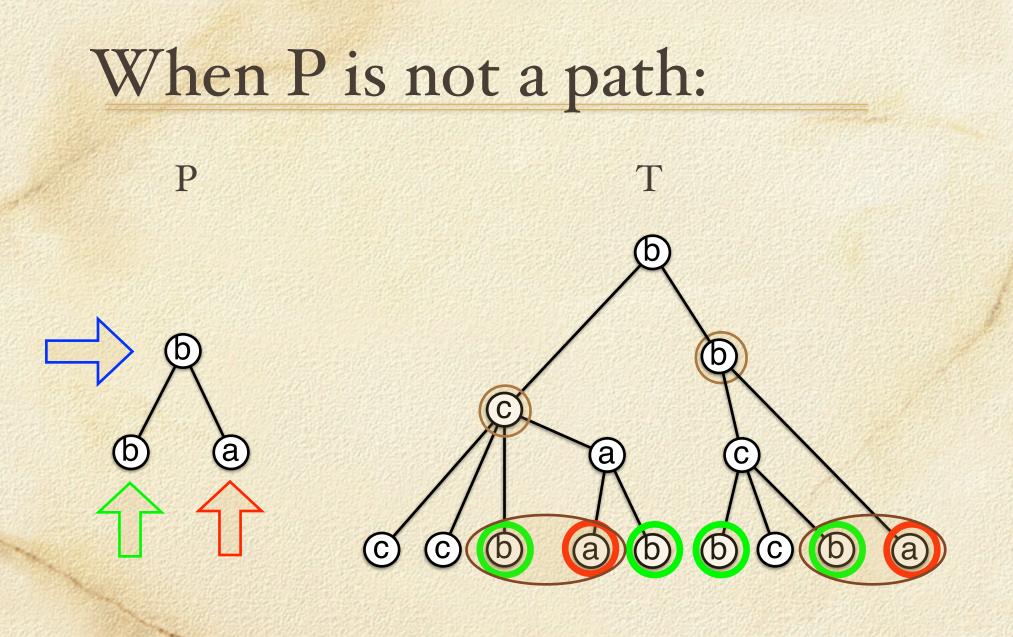


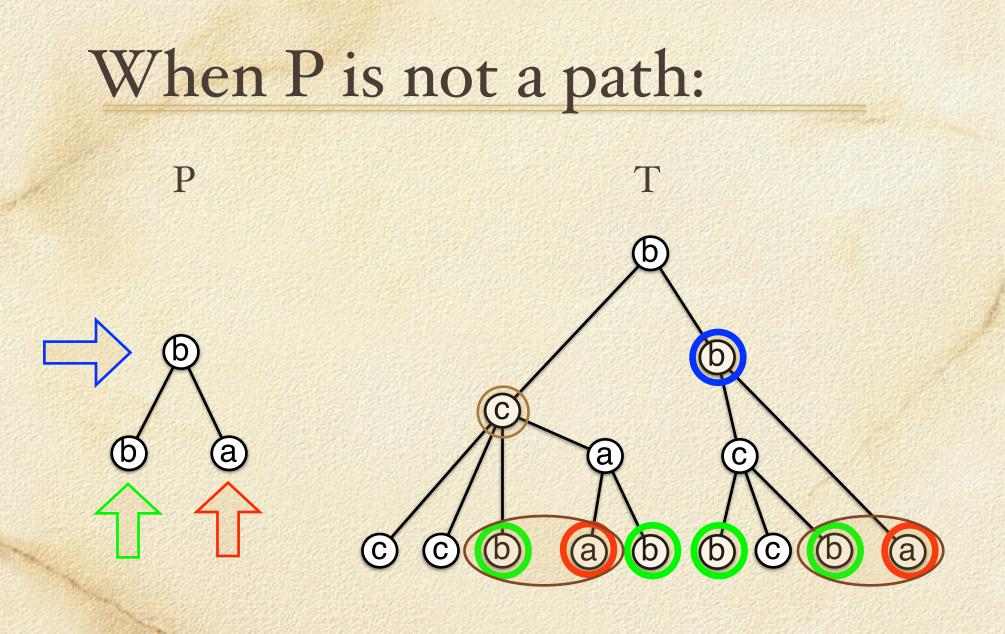


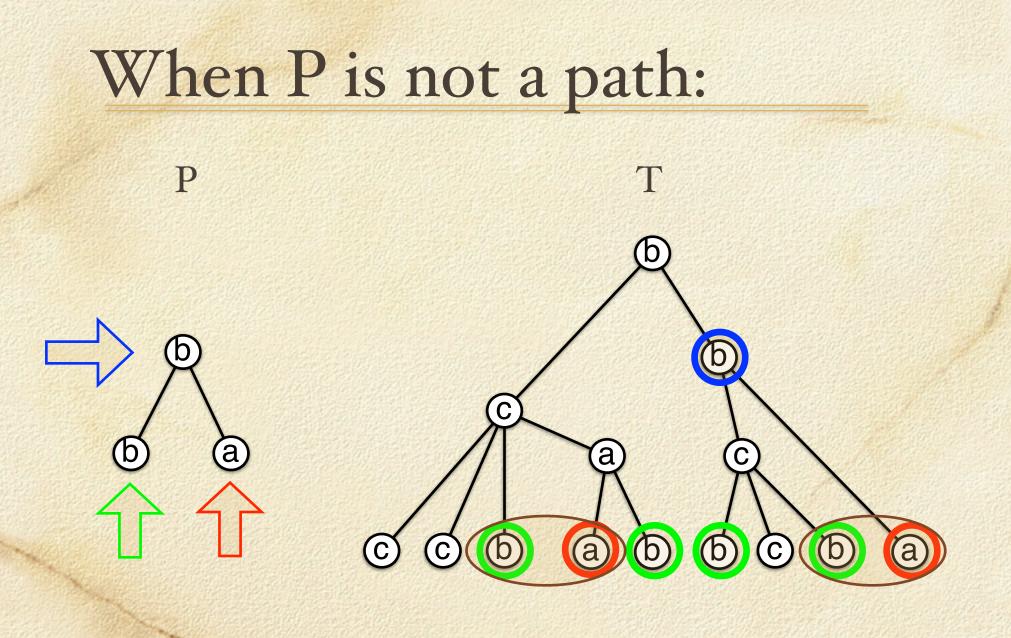


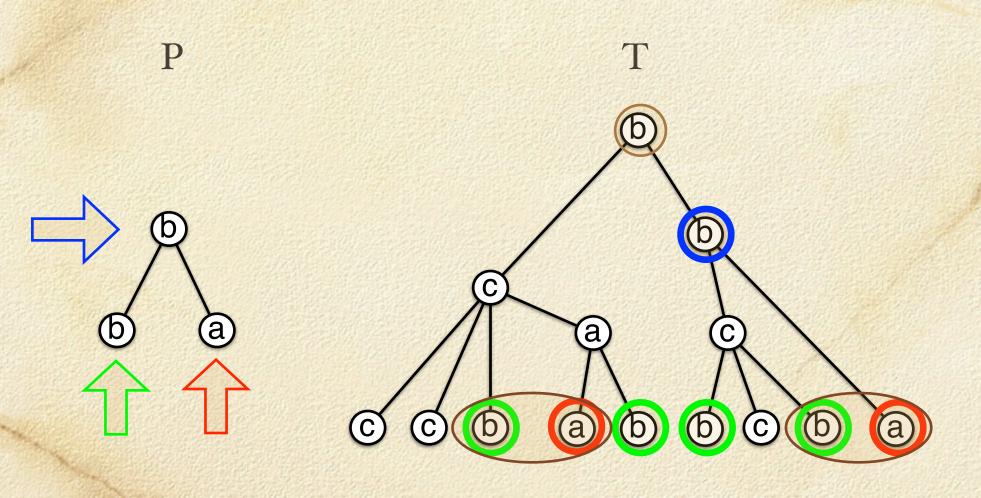


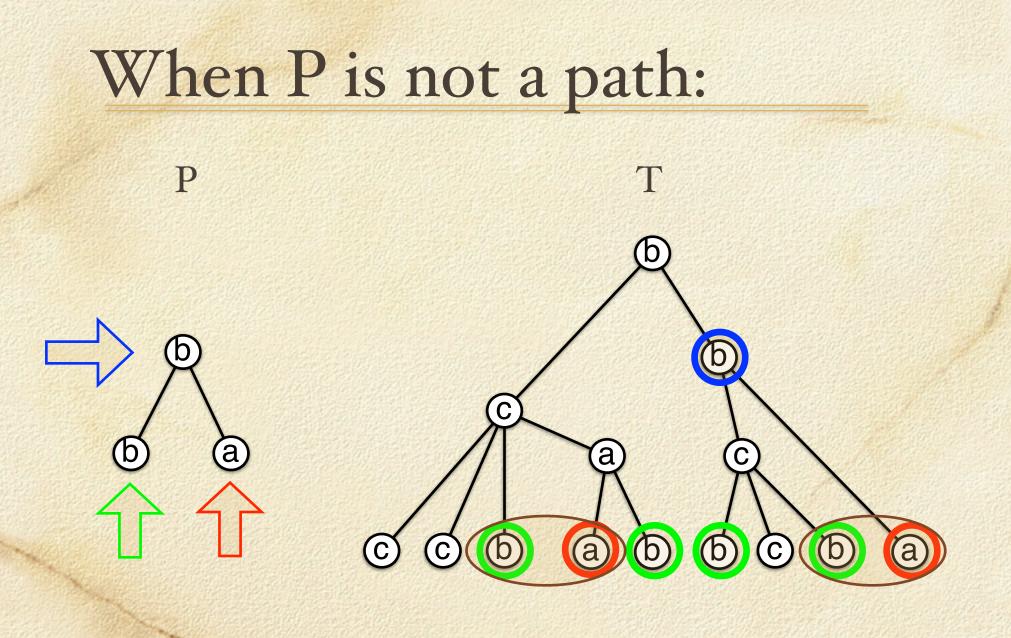












## Complexity

- Let Δ denote the set of all *leaf-to-root*, paths in P.
- Running time is by bounded by the time used to solve the tree inclusion problem on each path in Δ. In total:

$$\sum_{\delta \in \Delta} O(n_T) = O(l_P n_T)$$

Space is  $O(n_P + n_T)$ .

### Alternative algorithm.

Reconsider the case when P is path:

Let firstlabel(v,l) denote the nearest ancestor of the node v in T with label l.

At each step we "essentially" compute firstlabel(v, l) for each v in the active set.

## Alternative algorithm

Idea: Use a fast data structure supporting firstlabel queries. Known as the tree color problem.

**Lemma [Dietz89]** For any tree T there is a data structure using  $O(n_T)$  space,  $O(n_T)$  expected preprocessing time which supports *firstlabel(v,l)* in  $O(\log \log n_T)$  time.

## Complexity

For each node in P there is an active set and for each node in this active set we have to compute a *firstlabel* query.

 $\square$  Size of active set is at most  $l_T$ . Total time:

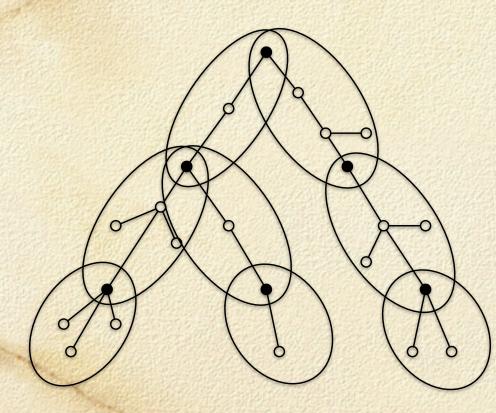
 $O(n_P l_T \log \log n_T)$ 

 $\Box$  Space is still  $O(n_P + n_T)$ .

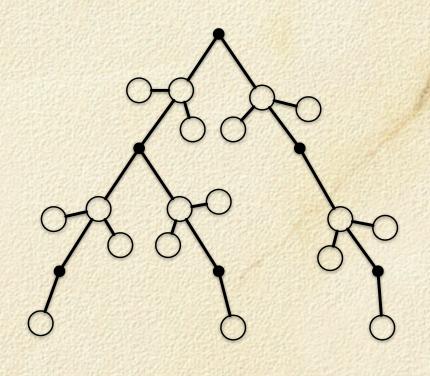
#### Improving the worst-case

- Divide T into O(n<sub>T</sub>/log n<sub>T</sub>) micro trees of size O(log n<sub>T</sub>) which overlap in at most 2 nodes using a clustering technique from [AHT97].
- Each micro tree is represented by a constant number of nodes in a *macro tree* and connected according to the overlap in the micro trees.
- Use a "Four Russian trick" to handle subsets





A Secondary



#### Preprocessing micro trees.

Consider *firstlabel(M,V,l)*, where V is a subset of nodes in a micro tree M.

For all possible. M and V precompute the following:

 $\square$  ancestor(M, V): All ancestors of V in M.

deep(M, V): Subset of V obtained by removing nodes that are ancestors of another node in V.

### Preprocessing micro trees.

- □ Number of different micro trees M of size x is less than  $2^{2x}$ . (Ignoring labels)
- $\square$  Number of different subsets V is less than  $2^x$ .
- Choosing appropriate  $x = \Theta(\log n_T)$  we can compute and tabulate *ancestor* and *deep* for all inputs for in linear time and space.

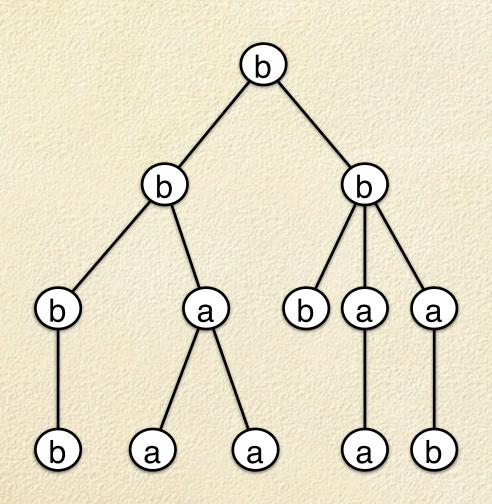
#### Preprocessing micro trees.

□ For each micro tree M (not all possible) store a dictionary (indexed by labels) containing:

 $\square$  mask(l): The set of nodes in M with label l.

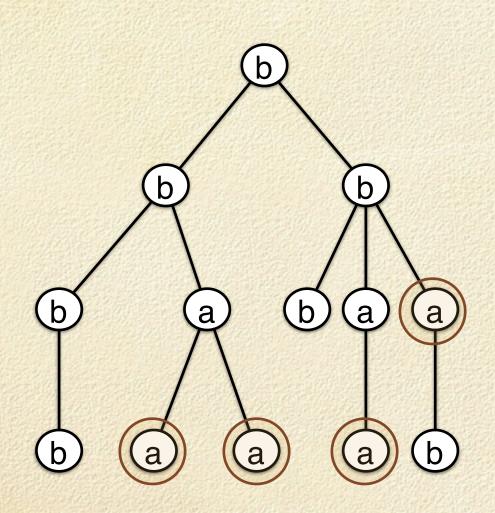
With perfect hashing this gives total linear space, linear expected preprocessing time, and constant lookup time.





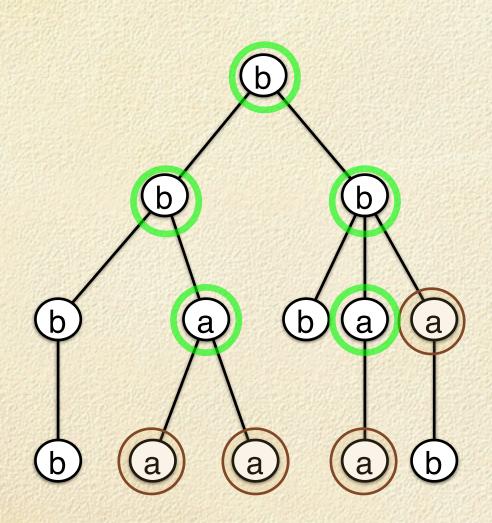
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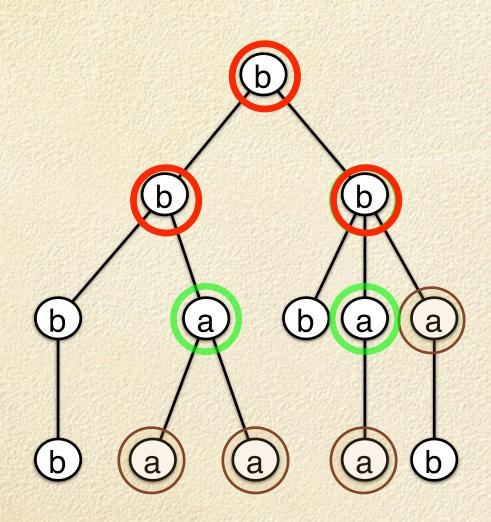
Sugar and





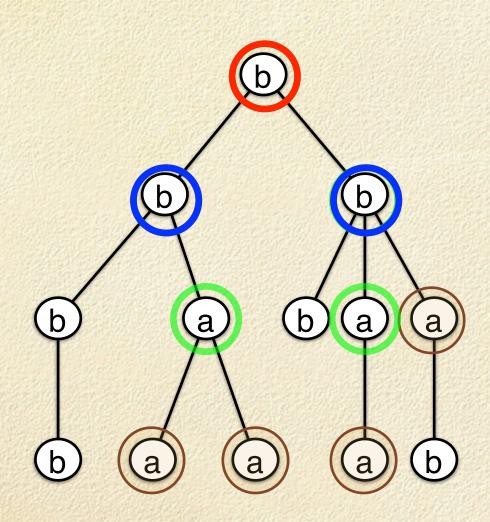
a start where





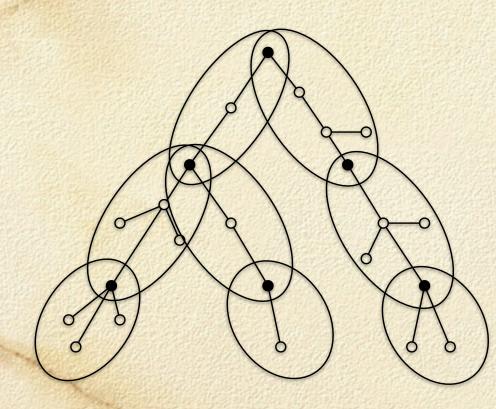
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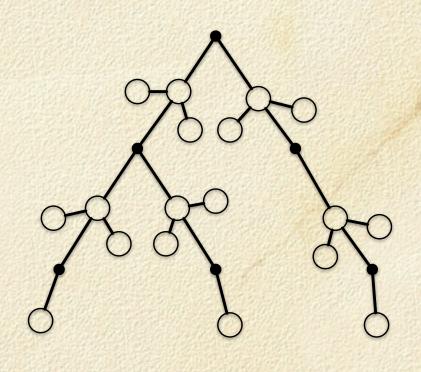


a start where

## Firstlabel not in M?



Color States



### General idea:

Compute *firstlabel* on each micro trees.

This gives a *firstlabel* query on the macro tree which is solved in linear time (in the number of nodes of the macro tree).

## Complexity

 $\Box$  Time for *firstlabel* becomes  $O(n_T / \log n_T)$ .

Same bound for all other needed manipulation of node sets.

Total time becomes O( <sup>n<sub>P</sub>n<sub>T</sub></sup>/<sub>log n<sub>T</sub></sub>).
Space is still O(n<sub>P</sub> + n<sub>T</sub>).

### Conclusion

**Theorem 1** For tree P and T the tree inclusion problem can be solved in time  $O(\min(l_P n_T, n_P l_T \log \log n_T, \frac{n_P n_T}{\log n_T}))$ 

and space  $O(n_P + n_T)$ .