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

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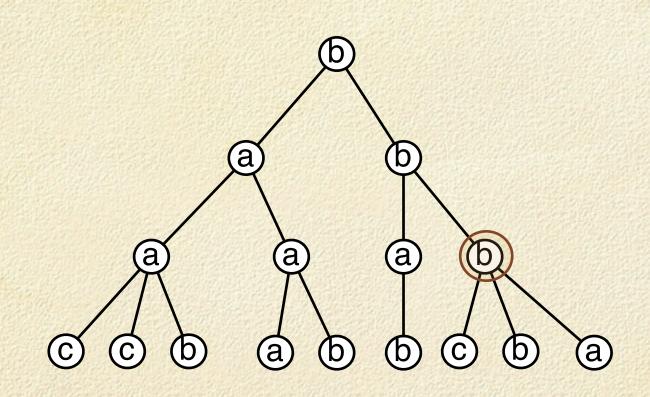
#### Basic setup

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.

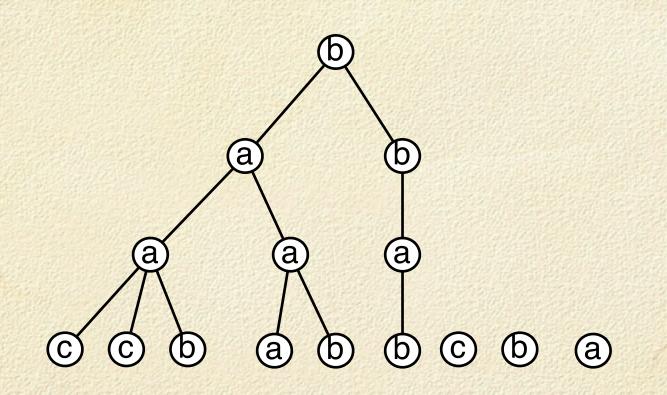
We compare trees by *deleting* nodes.

#### Delete a node



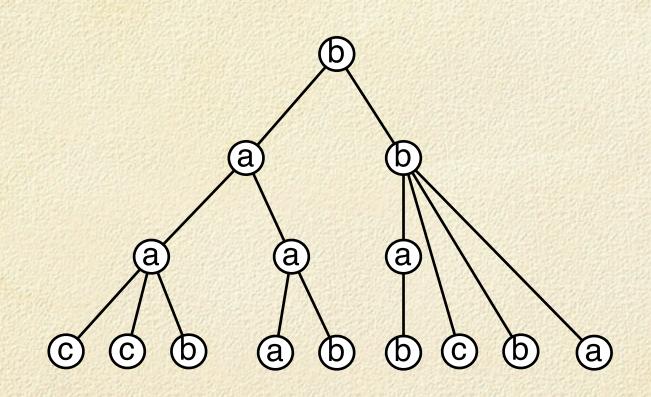
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#### Delete a node



Color States

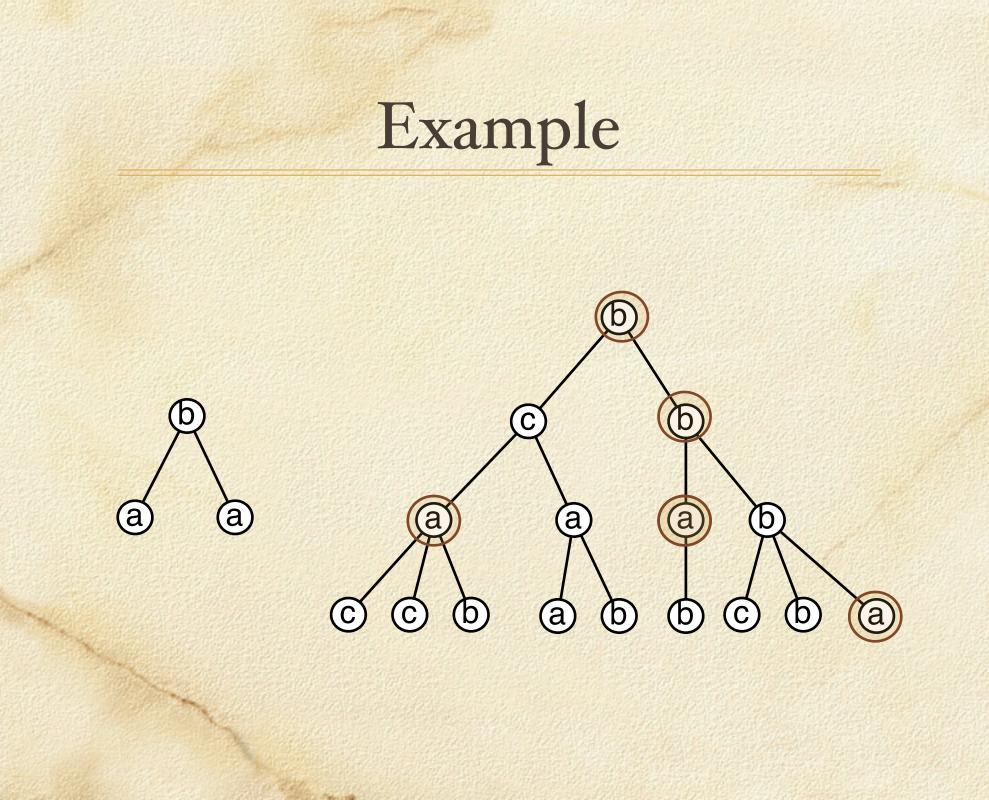
#### Delete a node

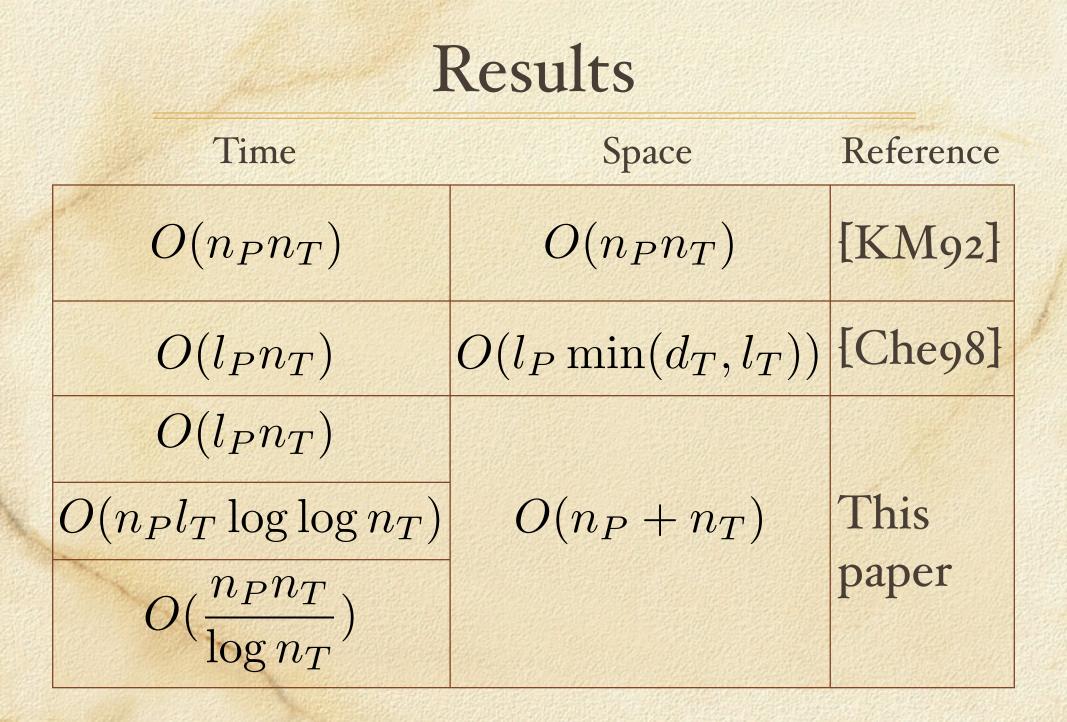


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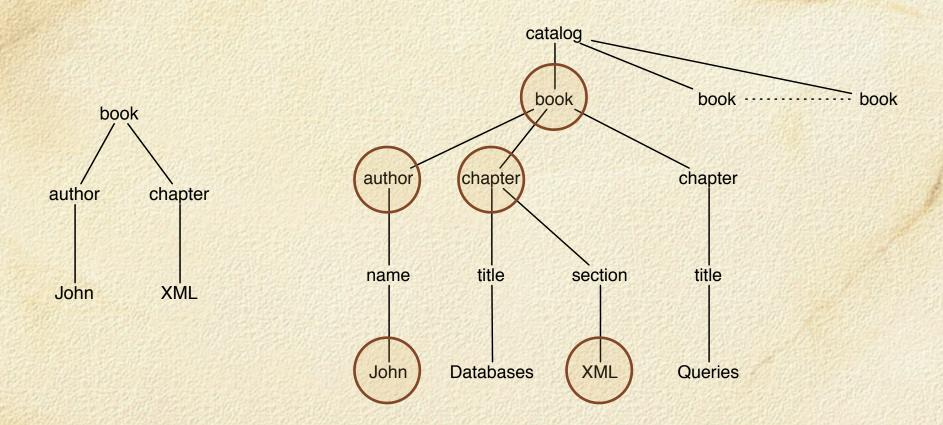
#### **Tree Inclusion**

- 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.





#### XML example



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

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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.

### A simple case: P is a path

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P



=Root of min. subtree including

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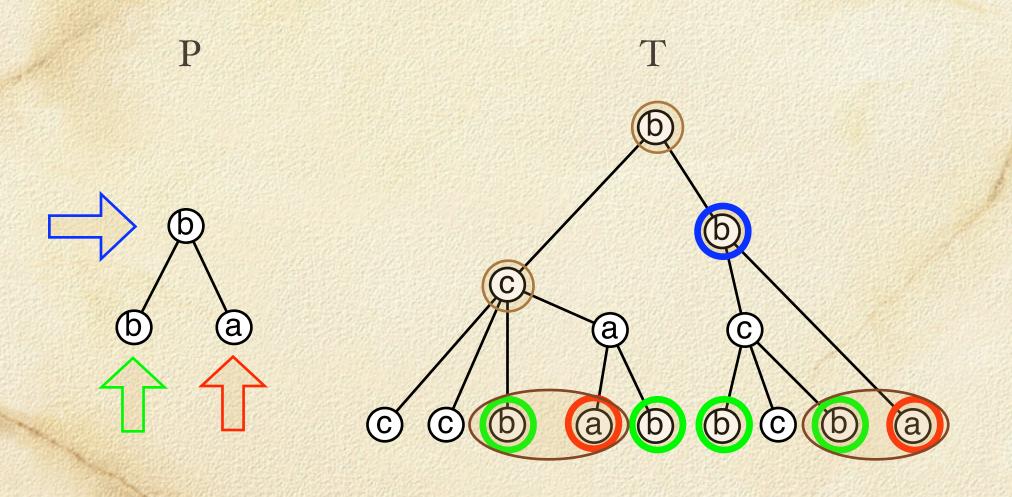
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T

# 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  $O(n_T)$ .

# When P is not a path:



# 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.

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

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

□ Space is still  $O(n_P + n_T)$ .

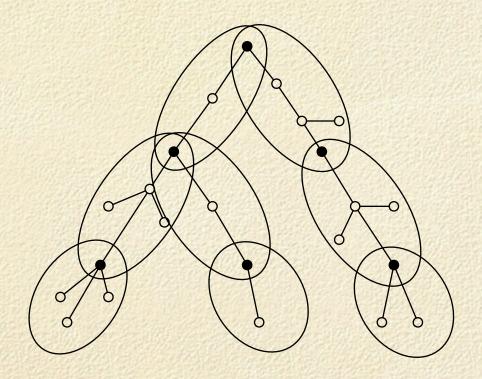
#### Improving the worst-case

Divide T into  $O(n_T/\log n_T)$  micro trees of size  $O(\log n_T)$  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.

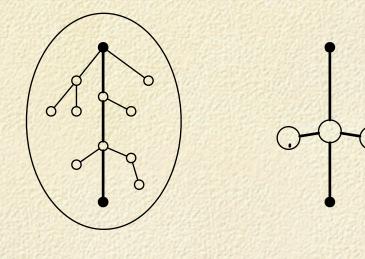
Essentially we do a "four russian" technique to get the speedup.

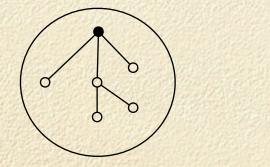
# Clustering example



Contraction of the

#### The Macro Tree





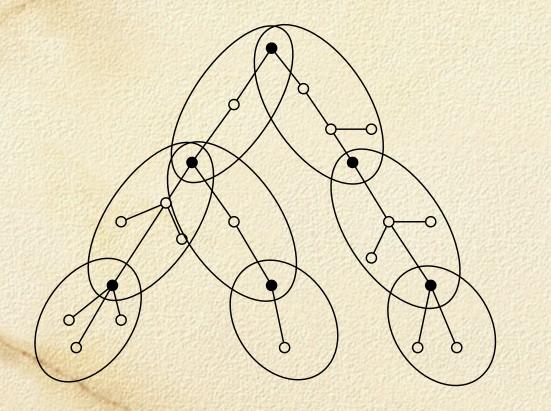
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#### Properties of macro trees

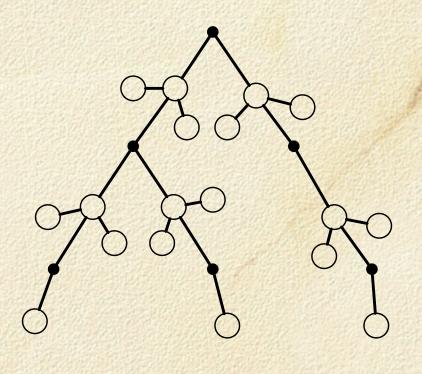
Each node x in the macro tree induces a micro tree (or forest) denoted I(x).

 $\square$  The label of x is the set of labels in I(x).





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#### MM-node sets

- To a represent a node set V in T we represent for each node x in the macro tree the subset of V in I(x).
- Each such subset is represent by a bitstring using a constant number of words.
- $\Box$  V is represented in  $O(n_T / \log n_T)$  space.

#### Firstlabel on mm-node set

- How can we compute *firstlabel*(M, l) for a mmnode set M?
- Use preprocessing to compute firstlabel on the micro trees fast.
- Combine the results using the macro tree to get solution.

### Handling micro trees fast

- Compute *firstlabel*(S,X, I) for a set of nodes X in a micro tree S. S and X are represented compactly in bitstrings.
- For all possible S and X precompute the following:
  - ancestor(S, X): All ancestors of X in S.
  - deep(S,X): Subset of X obtained by removing nodes that are ancestors of another node in X.

Contractor and and

### Handling micro trees fast

- Since S and X are of logarithmic size *ancestor* and *deep* can be computed (using dynamic programming) in linear time and space.
- Tabulating all inputs gives linear time lookup.

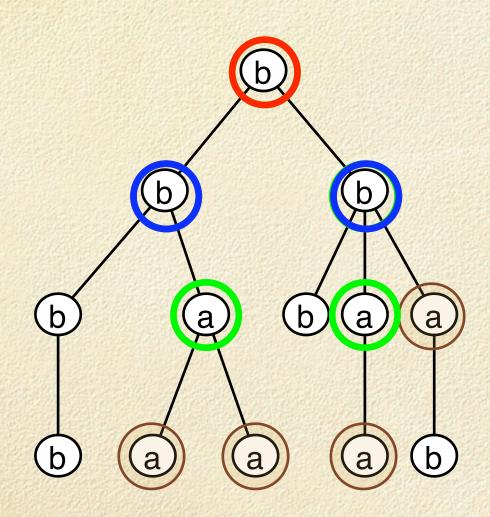
## Handling micro trees fast

For each micro tree S in T compute and store a dictionary (indexed by labels) containing:

mask(1): The set of nodes in S with label 1.

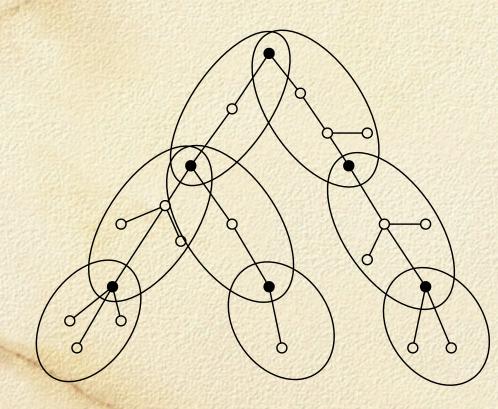
With perfect hashing this gives total linear space, linear expected preprocessing time, and constant lookup time.

# firstlabel(S,X,b)

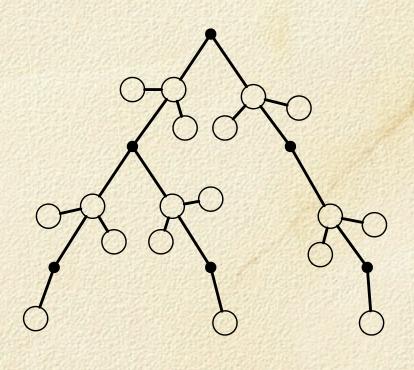


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#### Firstlabel not in S?



Color States



#### General solution

Compute *firstlabel* on each of the micro trees in M.

This gives a *firstlabel* query on the macro tree which is solved in linear time on the macro tree.

# Complexity

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

Similar bound for all other needed manipulation of node sets.

 $\Box \text{ Total time becomes } O(\frac{n_P n_T}{\log n_T}).$ 

• Space is still  $O(n_P + n_T)$ .

#### 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  $O(n_P + n_T)$ .

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