Chapter 14: Augmenting Data Structures

By augmenting already existing data structures one can build new data structures

Augmenting a red-black tree

For each node x, add a new field **size(x)**, the number of non-nil nodes in the subtree rooted at x

Now with the size information, we can fast compute the **dynamic order statistics** and the **rank**, the position in the linear order.



What is the size of the root of the above RB-tree?

Selection

To find the *i*th order statistics, run binary search.

Let a and b be the size of the left child and that of the right child, respectively. Then we do the following

- If i = a + 1, the current node holds the i-th o.s.
- If i < a + 1, search for the *i*-th o.s. in the left subtree.
- If i > a + 1, search for the (i a 1)-st
 o.s. in the right subtree.

What is the running time of this selection procedure?



Computing the Rank of a Given Node \boldsymbol{x}

Rank
$$(T, x)$$

1: $m \leftarrow size[left[x]] + 1$
2: $y \leftarrow x$
3: while $y \neq root[T]$ do
4: { if $right[p[y]] = x$
5: then $m \leftarrow m + size[left[p[y]]] + 1$
6: $y \leftarrow p[y]$
7: }
8: Return (m)



Here the rank of x is 1 + (c+1) + (e+1) = c + e + 3.

What is the running time of rank?

Maintaining the Size Information During RB-Tree Operations

1. Rotation

The size has to be changed for only one node:

- the left-child of the rotated node in the case of right rotation and
- the right-child of the rotated node in the case left rotation.



2. Insertion/Deletion

Climb up the tree from the actual point of insertion (respectively, deletion) all the way to the root. For each of the node that is encountered, and 1 to the size (respectively, subtract 1 from the size).

An Augmentation Strategy

Augmenting a data structure can be broken into the following four steps:

- 1. choosing an underlying data structure,
- determining what kind of additional information should be maintained in the underlying data structure,
- verify that the additional information can be maintained during the execution of each basic modifying operation of the underlying data structure, and
- 4. developing new operations.

The third step is easy for red-black trees.

Theorem A Let f be a field that augments a red-black tree T of n nodes, and suppose that the f-value of a node x can be computed solely from the information stored at x and at its children.

Then, maintaining the f-values of all nodes in T during insertion and deletion can be done in $O(\lg n)$ steps.

Proof Sketch Suppose that an operation has applied to an RB-tree T. Let T' the RB-tree after this operation.

There is a **downward path** π in T' such that every node that has been "touched" (its or its children's information has been modified) is within distance three from the path.

Thus, there are only $O(\log n)$ nodes for which the *f*-field has to be modified.

So, store π and update the *f*-fields of all the nodes within distance 3 from the path in a bottom-up fashion.

An Illustrating Example: Interval trees

For an interval i = [l, t], call l the low end and t the high end of i.

The trichotomy of intervals

For every pair of intervals i and j, exactly one of the following conditions holds:

- 1. i and j overlap
- 2. high[i] < low[j], i.e., j is to the right of i
- 3. high[j] < low[i], i.e., j is to the left of i



How can we maintain a dynamic set of closed intervals?

Step 1: Underlying Data Structure

Use the **RB tree**, where each node holds an interval. Use $int[\cdot]$ to refer to the interval. Use $lowint[\cdot]$ as the key.

Step 2: Additional Information

At each node store as additional information **the largest high end of the intervals** in the subtree rooted at the node. Use $max[\cdot]$ to refer to this information.

Step 3: Maintaining max

For all nodes x, max[x] is equal to

 $\max\{high[int[x]], max[left[x]], max[right[x]]\}.$ By the previous theorem, max can be maintained in $O(\lg n)$ steps.

Step 4: Developing New Operation

The only new operation needed is **searching** for an interval that overlaps an interval *i*.

Let T be the tree and i be the input.

Then set x to the root and execute the following loop:

- If int[x] ∩ i ≠ Ø, output int[x]. The search is over.;->
- Otherwise, if x is a leaf, then output "no intersecting intervals found." :-(
- Otherwise, if x has a unique child, then set x to the unique child.
- Otherwise, if the max[left[x]] ≥ low[i], then set x to left[x].
- Otherwise, set x to right[x].

Theorem B The algorithm works correctly.

Proof Call a subtree U good if it contains an interval overlapping i and bad otherwise.

We have only to show that if

(*) if T is good then T_x is good

holds during the course of the algorithm.

For initialization, the property (*) holds at the beginning of the search.

For the induction step, suppose that we are at non-leaf x and (*) holds. Suppose that Tis good. Then by (*) T_x is good. Suppose that the interval at x does not intersect with i. Let y be the node that is visited at the next round. We will show that T_y is good.

Since the interval at x does not intersect with i, either the left subtree of x is good or the right subtree of x is good.

This means that if there is only one child of x, then the unique child is good. Since y is this unique child (*) when x has only one child.

So, assume that x has two children.

(Case 1) $max[left[x]] \ge low[i]$: Here y = left[x].

(Case 1a) subtree(left[x]) is good: This implies T_y is good and T is good. So, (*) holds for y.

(Case 1b) *subtree*(*left*[x]) is bad:

Since $max[left[x]] \ge low[i]$ there is an interval to the right of *i* in the left subtree of *x*. This means that every interval in the right subtree is to the right of *i*. Thus, the right subtree is bad. So, both subtrees are bad, which is impossible. So, (Case 1b) never occurs.

(Case 2) max[left[x]] < low[i]:

Here y = right[x]. Since max[left[x]] < low[i], there is no interval that intersects with *i* in the left subtree tree. So T_y is good.