

## Cairo University Faculty of Computers and Artificial Intelligence Computer Science Department



Advanced Data Structures

## **Interval and Segment Trees**

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[For more details, refer to "Introduction to Algorithms" by *Thomas Cormen*, et al.] [For more details, refer to "Advanced Data Structures" by *Peter Brass*]

## 1 Interval tree

An *interval* [lo, hi] consists of the *range* between lo and hi inclusive, where  $lo \le hi$ . A *point* is a special case of an *interval* where lo = hi. Two *intervals* [ $a \triangleright lo$ ,  $a \triangleright hi$ ] and [ $b \triangleright lo$ ,  $b \triangleright hi$ ] overlap if and only if the *overlapping condition* is  $a \triangleright lo \le b \triangleright hi$  and  $b \triangleright lo \le a \triangleright hi$ . A *dynamic interval tree* stores a set of *intervals* and allows the following operations:

- Insert an *interval* in  $O(\log n)$  where n is the total number of stored *intervals*.
- Search for an *interval overlapping* with a query *interval* in  $O(\log n)$ .
- Search for all *intervals overlapping* with a query *interval* in  $O(k \log n)$  where k is the number of stored *intervals* satisfying the query.

The *dynamic interval tree* is a *red-black* tree such that each *node* stores the interval *lo* and *hi* and *keyed* by *lo*. That is, an *inorder* (*depth first*) walk of the tree lists the *intervals* sorted by their low endpoints. The tree is *augmented* with an additional member  $node \triangleright max$  which equals to the maximum *hi* of all *intervals* stored in the subtree *rooted* at *node*. The  $node \triangleright max$  attribute can be easily updated while *insertion* and *deletion* without changing their  $O(\log n)$  time complexity by the rule:  $node \triangleright max = \max(node \triangleright hi, node \triangleright left \triangleright max, node \triangleright right \triangleright max)$ .

The following procedure searches for an *interval overlapping* a *query interval* in  $O(\log n)$ :

```
node \leftarrow root
while (query does not overlap node)
if (query \triangleright lo > node \triangleright left \triangleright max)
then node \leftarrow node \triangleright right
else node \leftarrow node \triangleright left
return node \triangleright interval
```

if  $(query \triangleright lo > node \triangleright left \triangleright max)$  there is no point to search in the  $node \triangleright left$  subtree since it can never *overlap* with any *interval* there, thus we only investigate  $node \triangleright right$  subtree.

if  $(query \triangleright lo \le node \triangleright left \triangleright max)$ , it is possible to find *overlapping intervals* with *query* in both  $node \triangleright left$  and  $node \triangleright right$  subtrees. However, if there exists an *interval overlapping* with *query* in  $node \triangleright right$  subtree, there must be an *interval overlapping* with *query* in  $node \triangleright left$  subtree \*. Therefore it is safe to ignore  $node \triangleright right$  subtree and only investigate  $node \triangleright left$  subtree because we are mainly interested in finding one *interval overlapping* with *query*.

\* If  $(query \triangleright lo \le node \triangleright left \triangleright max)$  and there exists an *interval overlapping* with query in  $node \triangleright right$  subtree, there must be an *interval overlapping* with query in  $node \triangleright left$  subtree.

Proof: We will prove that *query* overlaps with  $[m \triangleright lo, m \triangleright hi]$  which is the *interval* in  $node \triangleright left$  subtree having  $m \triangleright hi = node \triangleright left \triangleright max$ . Since  $query \triangleright lo \le node \triangleright left \triangleright max$ , then  $query \triangleright lo \le m \triangleright hi$ . To satisfy the *overlapping condition*, it remains to show that  $m \triangleright lo \le query \triangleright hi$  to prove that the *intervals query* and m *overlap*.

Suppose the *query interval overlaps* with an *interval*  $[r \triangleright lo, r \triangleright hi]$  in  $node \triangleright right$  subtree. From the *overlapping condition*:  $query \triangleright hi \ge r \triangleright lo$ . Since the tree is keyed by  $lo, r \triangleright lo >$  the lo of all *intervals* in  $node \triangleright left$  subtree. Therefore,  $r \triangleright lo > m \triangleright lo$ . Thus,  $query \triangleright hi > m \triangleright lo$ .

The following recursive procedure modifies the above procedure to search for the *interval* having the smallest *lo* endpoint *overlapping* with a *query interval* in  $O(\log n)$ . It differs from the above procedure only in checking the left subtree for *overlapping intervals* before checking the current node *interval*. The initial call should be Search(root, query):

```
Search(node, query)

if (query \triangleright lo \leq node \triangleright left \triangleright max)
then

result = Search(node \triangleright left, query)
if (result \neq null) then return result
if (query \text{ overlaps } node) then return node \triangleright interval
return null
else

if (query \text{ overlaps } node) then return node \triangleright interval
return Search(node \triangleright right, query)
```

Let the number of levels of the tree is  $L = O(\log n)$ . The time complexity of the above procedure is S(L) = S(L-1) + O(1) where S(1) = O(1). Thus, its time complexity is  $L = O(\log n)$ .

To search for all *intervals overlapping* with a *query interval*, we can execute one of the above procedures several times until it returns null, such that after each execution the found *interval* is removed from the tree. Thus, the complexity of such iterative procedure is  $O(k \log n)$  where k is the number of stored *intervals overlapping* with the *query interval*.

It is possible to achieve that target without changing the tree structure, by storing - outside the tree - the max attribute values for nodes that need updating. The last procedure is easier to be used in such implementation, since in that case only the max attribute values of one path of the tree whose length is  $O(\log n)$  need to be stored, and we know that the next *interval* to be retrieved is located somewhere on the stored path or to the <u>right</u> of it.

Since a *point* is a special case of an *interval*, all the above results hold if the *query* is a *point* instead of an *interval*, and it is required to retrieve *intervals* including the *query point*.

There is a variant of *static interval tree* which, given n *static* intervals, can be built in  $O(n \log n)$  time and O(n) space and allows the retrieval of all stored k *intervals overlapping* with a *query interval* in  $O(\log n + k)$  time. But, it does not allow efficient insertion of new *intervals*.

## 2 Segment tree

Segment tree is a static data structure which stores right-open n input intervals in  $O(n \log n)$  space and allows the retrieval of all stored k input intervals overlapping with a query interval in  $O(\log n + k)$  time. It does not allow efficient insertion of new intervals. A right-open interval [lo, hi] consists of the range between its two endpoints: lo and hi, excluding hi, where lo < hi.

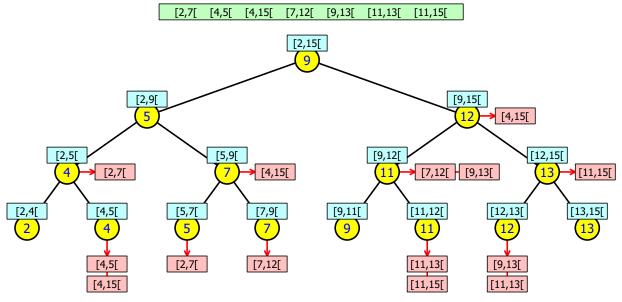
A segment tree is constructed in  $O(n \log n)$  time by sorting the endpoints except the maximum one) of the *n* input intervals as keys of the lowest-level nodes of a static binary tree. Then, higher levels are constructed such that an internal node is keyed by the the smallest key of its right subtree.

Each *leaf node* corresponds to the right-open *segment tree interval* starting from its key to the key of its right adjacent *leaf* (or the *maximum* endpoint it is the right-most leaf). Each *internal node* corresponds to the *segment tree interval* of the union of the *segment tree intervals* of its children.

Note that the union of *segment tree interval* of all nodes in the same level is the right-open *segment tree interval* starting from the minimum *endpoint* to the maximum *endpoint*. Note also that all *segment tree intervals* of a subtree are subsets of the *segment tree interval* of its root.

Then, each *input interval* is represented as the union of the smallest number of *segment tree intervals* corresponding to *segment tree nodes*. Such representation is called the *canonical representation* of the *input interval* relative to that *segment tree*. Then each *input interval* is attached to all such *segment tree nodes* participated in its *canonical representation*.

The following figure contains *input interval* at the top light-green rectangle. Node keys are shown in yellow circles. *Segment tree intervals* corresponding to *segment tree nodes* are shown in light-blue rectangles. *Input Intervals* attached to *segment tree nodes* are shown in light-red rectangles.



The root corresponds to the *interval* [2, 15[ spanning the minimum and maximum endpoints of all *input intervals*. The root has the key 9 to indicate the separator between its left and right children *segment tree intervals* which are [2, 9[ and [9, 15[. The keys of the leaves are the endpoints of all *input intervals* except for the maximum. For example, the leaf having the key 5 corresponds to the *segment tree intervals* [5, 7[ where 7 is the key of the adjacent right leaf.

The canonical representation of the input interval [4, 15] relative to the above segment tree are the segment tree intervals [4, 5], [5, 9], and [9, 15] (light-blue rectangles). Thus, the input interval [4, 15] (light-red rectangles) is attached to the segment tree nodes corresponding to such segment tree intervals. We described how to build the segment tree, except how to find canonical representations of input intervals to attach them to the corresponding segment tree nodes, which is described below:

To construct the *canonical representation* of an *input interval*, start from the *segment tree root* and consider the *segment tree interval* corresponding to the current *segment tree node*:

- If the *segment tree interval* is entirely contained in the *input interval*, attach the *input interval* to the *segment tree node* and stop following the path down.
- If the segment tree interval partially overlaps the input interval, follow both paths down.
- If the segment tree interval is disjoint from the input interval, stop following the path down.

Since the *segment tree* is balanced, the *segment tree* height is  $O(\log n)$ . The above procedure consumes  $O(\log n)$  time because the maximum number of nodes investigated in each level is 4.

Proof: We will prove that there cannot be more than 2 nodes per level that both its paths need to be followed down. Thus, the maximum number of investigated nodes per level is 4. Suppose that there are  $\geq 3$  nodes to be investigated in the same level. All these nodes must be adjacent in the subtree (because *segment tree intervals* in the same level are sorted), and all these nodes (except for the farthest left and farthest right nodes) must be entirely contained in the *input interval* and their paths need not to be followed down.

The above proof also proves that the size of the *canonical representation* of each *input interval* is  $O(\log n)$ . Therefore, each *input interval* is attached to only  $O(\log n)$  segment tree nodes. Thus, the segment tree space is  $O(n \log n)$ . Also, its construction time is  $O(n \log n)$ .

The following  $O(\log n + k)$  procedure reports all k input intervals overlapping with query interval:

- If the *segment tree interval* is entirely contained in the *query interval*, report all the attached *input intervals* to all nodes in the subtree rooted by this *segment tree node*.
- If the *segment tree interval* partially overlaps the *query interval*, report all the attached *input intervals* to this *segment tree node* and follow both paths down.
- If the segment tree interval is disjoint from the query interval, stop following the path down.

If the *query* is a single *point*, *segment tree* allows the retrieval of all k *input intervals* including the *query point* in  $O(\log n + k)$  time, starting from the root we follow one search tree path (as any binary search tree) and report all *input intervals* attached to the traversed *segment tree nodes*.

Segment tree can be recursively generalized to d dimensions by constructing a one-dimensional segment tree based on the first coordinate of all intervals, then attaching to every node a segment tree of the remaining d-1 dimensions built on all intervals in the subtree rooted at that node.

Because all levels are filled with nodes (except for the right places of the last level), *segment trees* are *complete trees* and they can be implemented using arrays (similarly to implementing *heaps*). Simple mathematical formulas are used to traverse from parent to children and vice versa.

Segment tree is a powerful framework that can be modified to be used for other purposes, such as answering dynamic range queries as range sum queries and range minimum queries including point and range updates.