#### **Design and Analysis of Algorithms**

CSE 5311

Lecture 11 Augmenting Data Structures

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## Dynamic order statistics

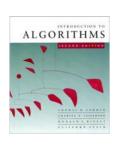
OS-SELECT(i, S): returns the ith smallest element in the dynamic set S.

OS-RANK(x, S): returns the rank of  $x \in S$  in the sorted order of S's elements.

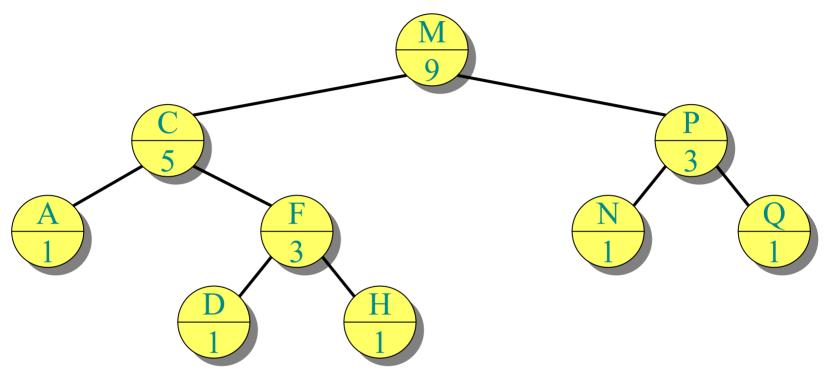
**IDEA:** Use a red-black tree for the set *S*, but keep subtree sizes in the nodes.

Notation for nodes:





#### Example of an OS-tree



$$size[x] = size[left[x]] + size[right[x]] + 1$$



#### Selection

Implementation trick: Use a *sentinel* (dummy record) for NIL such that size[NIL] = 0.

OS-SELECT(x, i) > ith smallest element in the subtree rooted at x

```
k \leftarrow size[left[x]] + 1 \triangleright k = rank(x)
if i = k then return x
if i < k
```

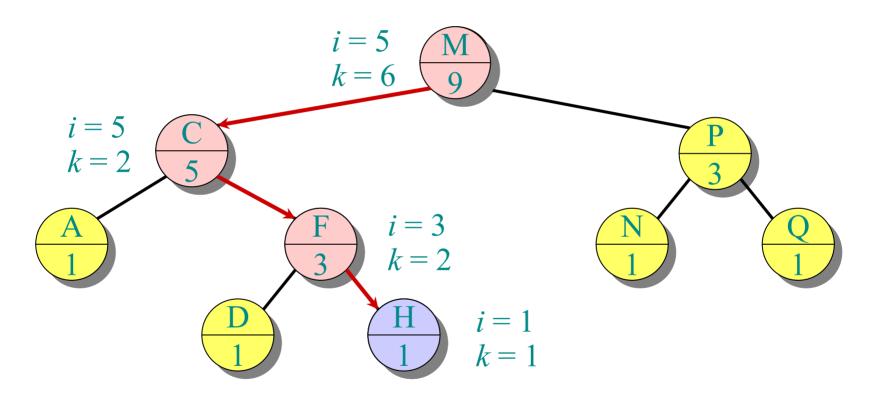
then return OS-SELECT(left[x], i) else return OS-SELECT(right[x], i-k)

(OS-RANK is in the textbook.)

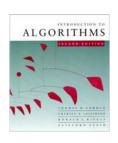


#### Example

#### OS-SELECT(*root*, 5)



Running time =  $O(h) = O(\lg n)$  for red-black trees.



## Data structure maintenance

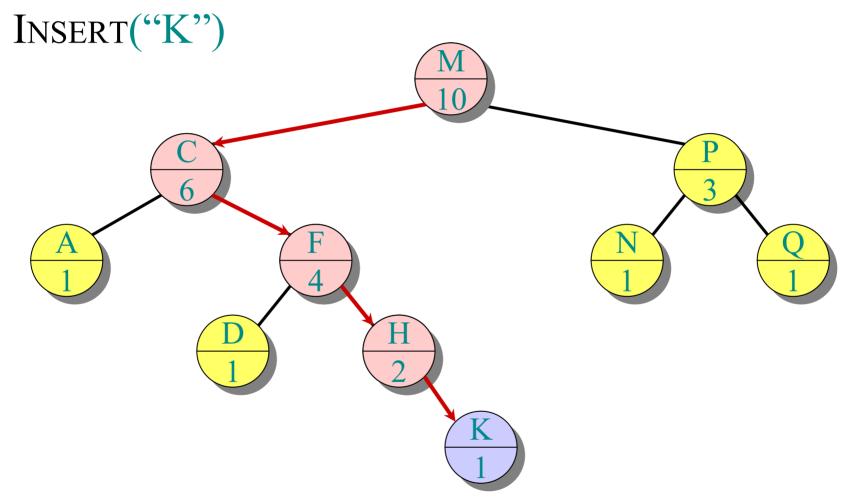
- **Q.** Why not keep the ranks themselves in the nodes instead of subtree sizes?
- A. They are **hard** to maintain when the red-black tree is modified.

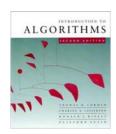
Modifying operations: Insert and Delete.

**Strategy:** Update subtree sizes when inserting or deleting.



#### **Example of insertion**

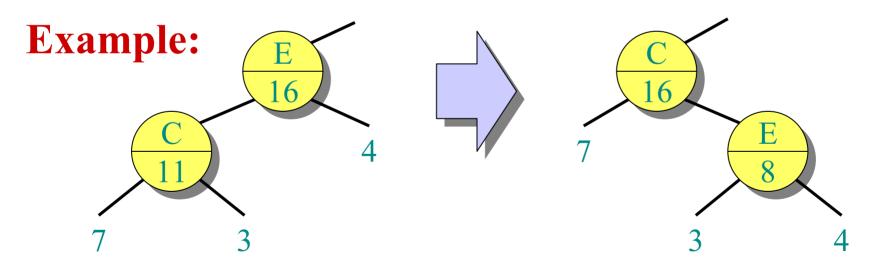




# Handling rebalancing

Don't forget that RB-INSERT and RB-DELETE may also need to modify the red-black tree in order to maintain balance.

- *Recolorings*: no effect on subtree sizes.
- *Rotations*: fix up subtree sizes in O(1) time.



∴ RB-INSERT and RB-DELETE still run in  $O(\lg n)$  time.

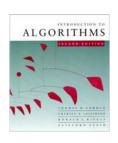


# Data-structure augmentation

#### Methodology: (e.g., order-statistics trees)

- 1. Choose an underlying data structure (*red-black trees*).
- 2. Determine additional information to be stored in the data structure (*subtree sizes*).
- 3. Verify that this information can be maintained for modifying operations (*RB-INSERT, RB-DELETE don't forget rotations*).
- 4. Develop new dynamic-set operations that use the information (*OS-SELECT and OS-RANK*).

These steps are guidelines, not rigid rules.



#### Interval trees

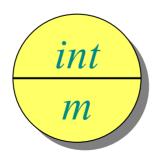
Goal: To maintain a dynamic set of intervals, such as time intervals.

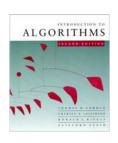
Query: For a given query interval *i*, find an interval in the set that overlaps *i*.



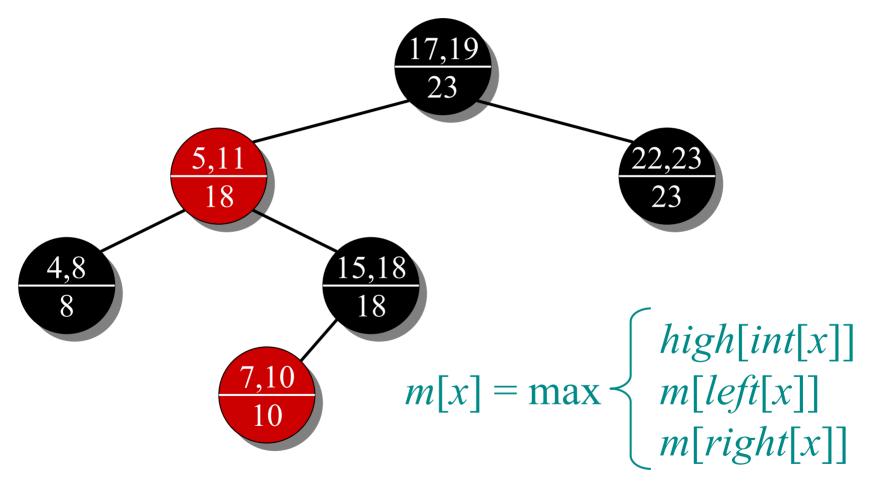
# Following the methodology

- 1. Choose an underlying data structure.
  - Red-black tree keyed on low (left) endpoint.
- 2. Determine additional information to be stored in the data structure.
  - Store in each node x the largest value m[x] in the subtree rooted at x, as well as the interval int[x] corresponding to the key.





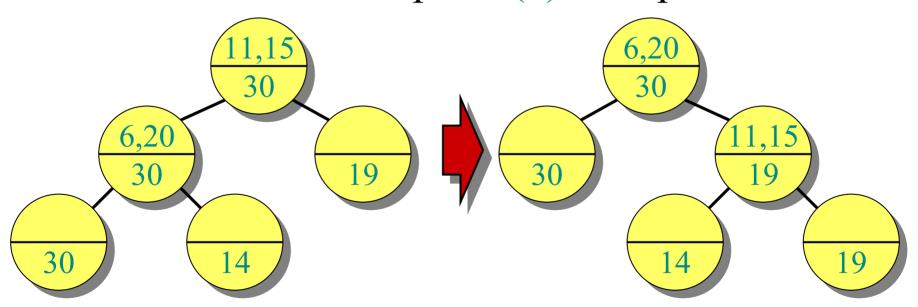
## Example interval tree





# **Modifying operations**

- 3. Verify that this information can be maintained for modifying operations.
  - Insert: Fix *m*'s on the way down.
  - Rotations Fixup = O(1) time per rotation:



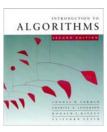
Total Insert time =  $O(\lg n)$ ; Delete similar.

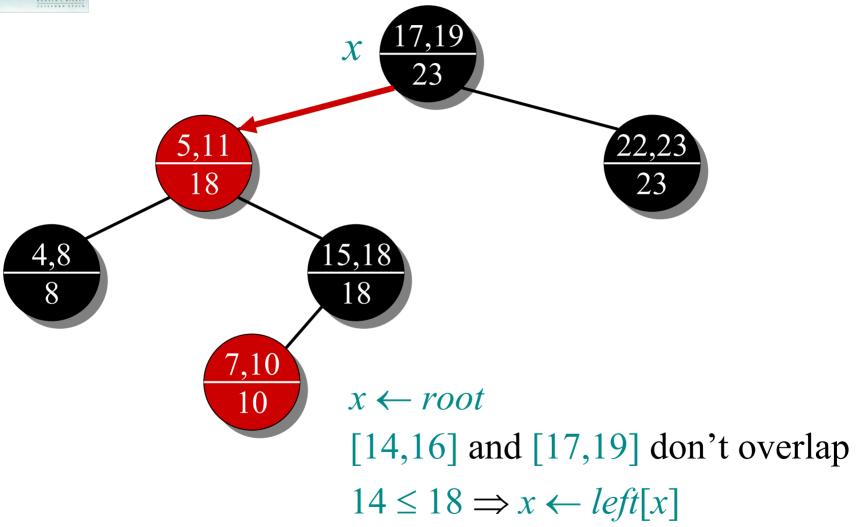


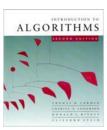
## New operations

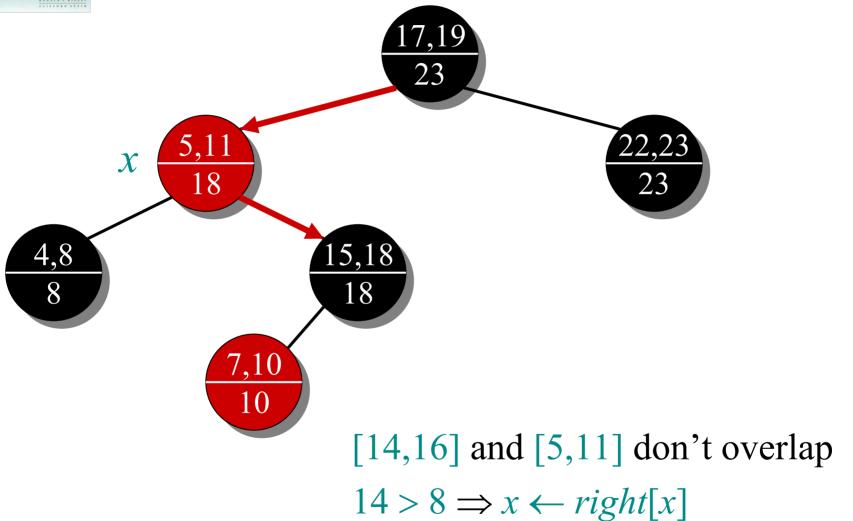
4. Develop new dynamic-set operations that use the information.

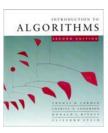
```
INTERVAL-SEARCH(i)
    x \leftarrow root
    while x \neq \text{NIL} and (low[i] > high[int[x]])
                              or low[int[x]] > high[i])
        \mathbf{do} \triangleright i and int[x] don't overlap
            if left[x] \neq NIL and low[i] \leq m[left[x]]
                 then x \leftarrow left[x]
                 else x \leftarrow right[x]
    return x
```

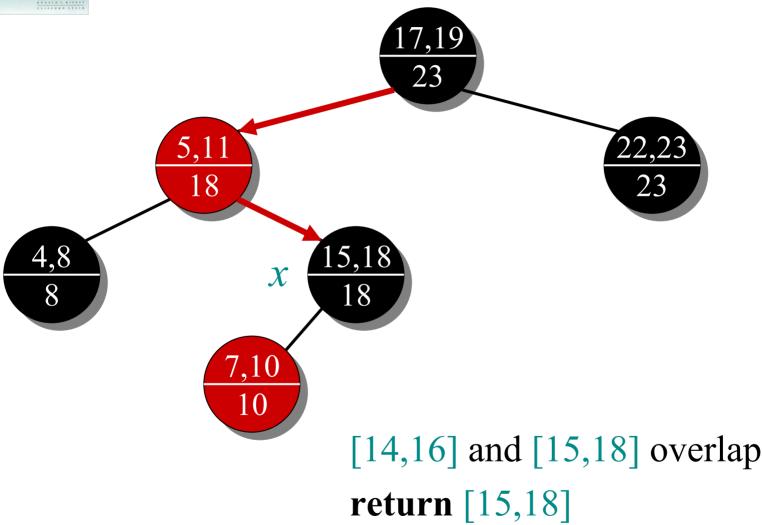


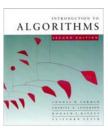


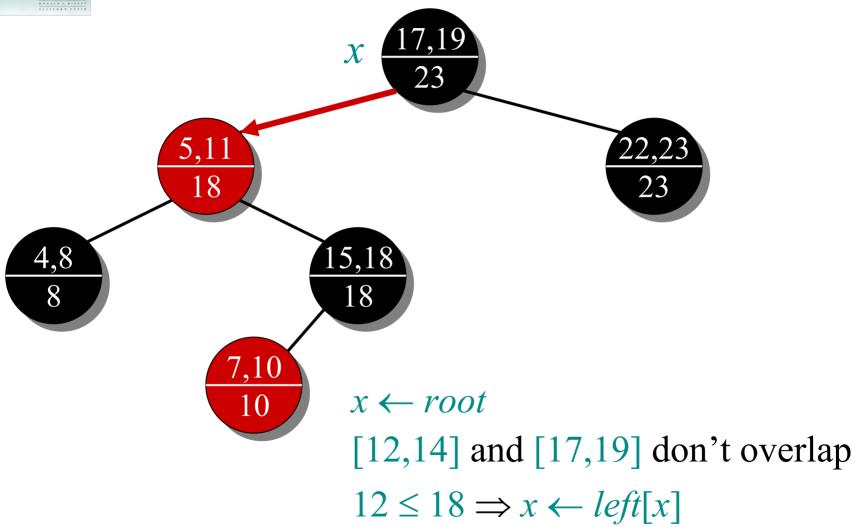


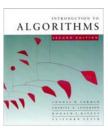


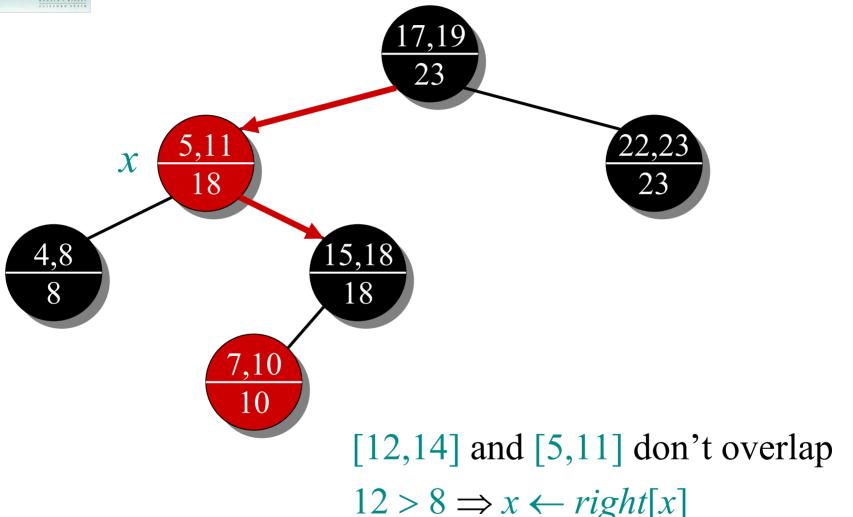


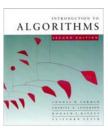


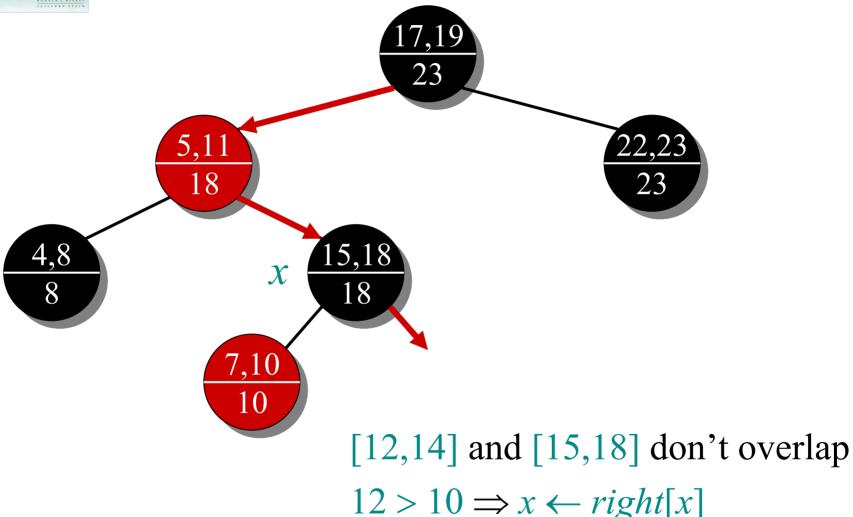


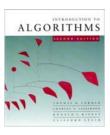


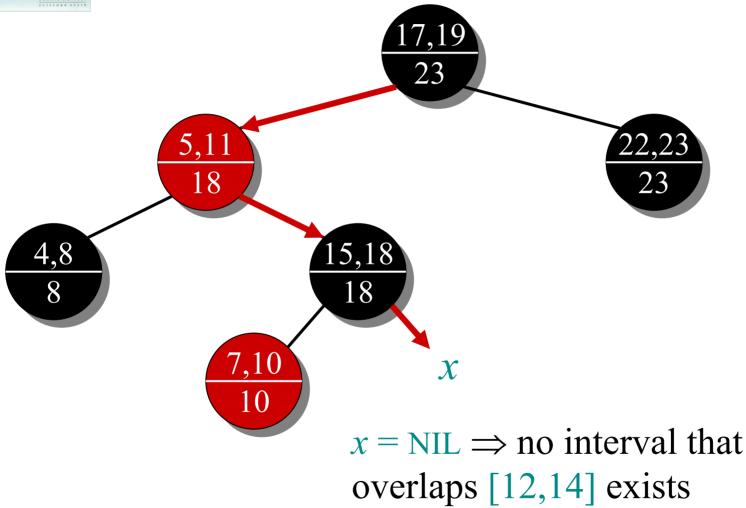














## **Analysis**

Time =  $O(h) = O(\lg n)$ , since Interval-Search does constant work at each level as it follows a simple path down the tree.

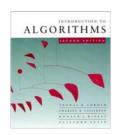
List *all* overlapping intervals:

- Search, list, delete, repeat.
- Insert them all again at the end.

Time =  $O(k \lg n)$ , where k is the total number of overlapping intervals.

This is an *output-sensitive* bound.

Best algorithm to date:  $O(k + \lg n)$ .



#### **Correctness**

**Theorem.** Let L be the set of intervals in the left subtree of node x, and let R be the set of intervals in x's right subtree.

• If the search goes right, then

$$\{i' \in L : i' \text{ overlaps } i\} = \emptyset.$$

• If the search goes left, then

$$\{i' \in L : i' \text{ overlaps } i\} = \emptyset$$
  
  $\Rightarrow \{i' \in R : i' \text{ overlaps } i\} = \emptyset.$ 

In other words, it's always safe to take only 1 of the 2 children: we'll either find something, or nothing was to be found.



## **Correctness proof**

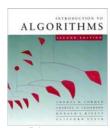
*Proof.* Suppose first that the search goes right.

- If left[x] = NIL, then we're done, since  $L = \emptyset$ .
- Otherwise, the code dictates that we must have low[i] > m[left[x]]. The value m[left[x]] corresponds to the high endpoint of some interval  $j \in L$ , and no other interval in L can have a larger high endpoint than high[j].

$$high[j] = m[left[x]]$$

$$low(i)$$

• Therefore,  $\{i' \in L : i' \text{ overlaps } i\} = \emptyset$ .



## **Proof (continued)**

Suppose that the search goes left, and assume that  $\{i' \in L : i' \text{ overlaps } i\} = \emptyset$ .

- Then, the code dictates that  $low[i] \le m[left[x]] = high[j]$  for some  $j \in L$ .
- Since  $j \in L$ , it does not overlap i, and hence high[i] < low[j].
- But, the binary-search-tree property implies that for all  $i' \in R$ , we have  $low[j] \le low[i']$ .
- But then  $\{i' \in R : i' \text{ overlaps } i\} = \emptyset$ .

