

## Augmenting Data Structures

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Slides courtesy of Charles Leiserson with small changes by Carola Wenk

Abstract Data Type (ADT) **Dictionary** :

Insert  $(x, D)$ : inserts  $x$  into  $D$   
 Delete  $(x, D)$ : deletes  $x$  from  $D$   
 Find  $(x, D)$ : finds  $x$  in  $D$

}  $D$  is a **dynamic set**

Popular implementation uses any **balanced search tree** (not necessarily binary). Like that each operation takes  $O(\log n)$  time.

## Dynamic order statistics

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

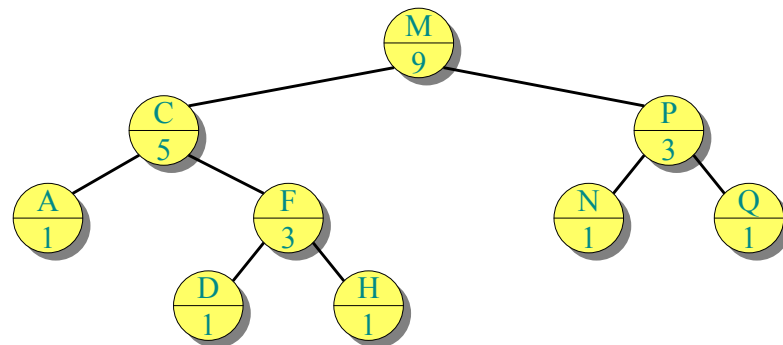
OS-RANK( $x, S$ ): returns the rank of  $x \in S$  in the sorted order of  $S$ 's elements.

**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( $x, i$ ) ▷  $i$ th smallest element in the subtree rooted at  $x$

$k \leftarrow size[left[x]] + 1$  ▷  $k = rank(x)$

if  $i = k$  then return  $x$

if  $i < k$

then return OS-SELECT( $left[x], i$ )

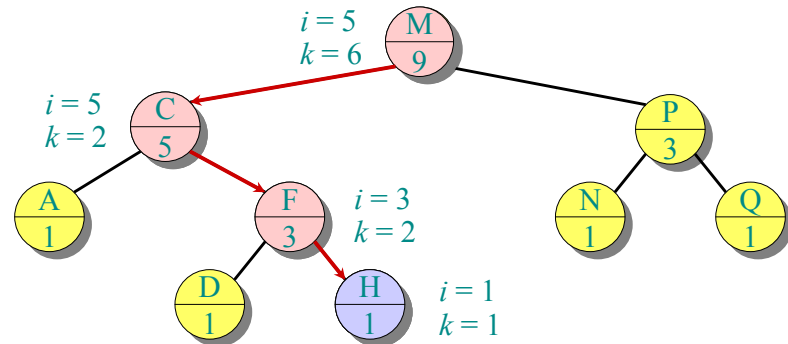
else return OS-SELECT( $right[x], i - k$ )

(OS-RANK is in the textbook.)



# Example

OS-SELECT( $root, 5$ )



Running time =  $O(h) = O(\log n)$  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.

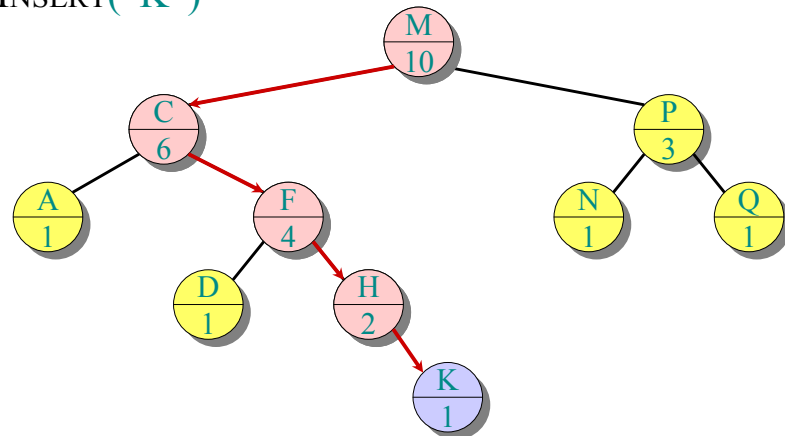
**Modifying operations:** INSERT and DELETE.

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



# Example of insertion

INSERT("K")

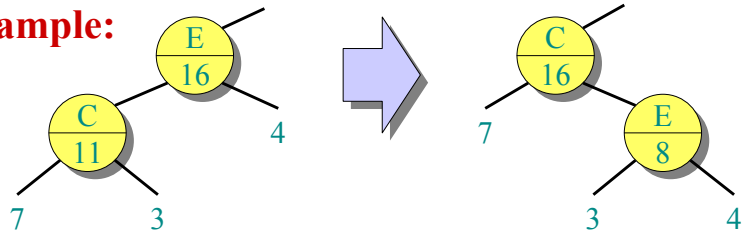


## Handling rebalancing

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.

**Example:**



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

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## 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.

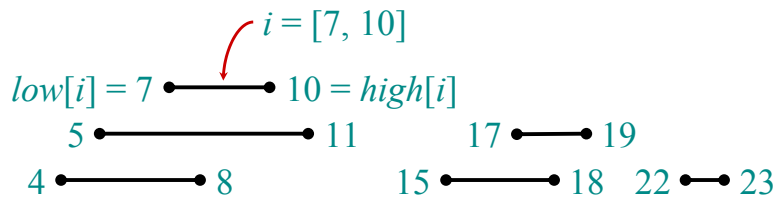
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## Interval trees

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

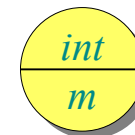


**Query:** For a given query interval  $i$ , find an interval in the set that overlaps  $i$ .



## 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.



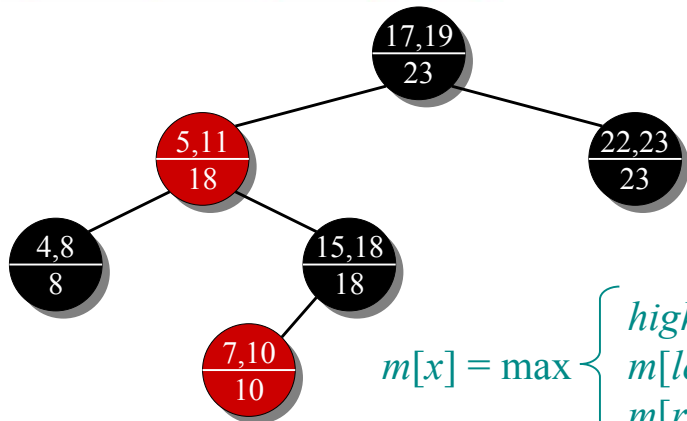
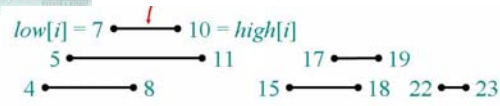
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## Example interval tree



$$m[x] = \max \begin{cases} \text{high}[\text{int}[x]] \\ m[\text{left}[x]] \\ m[\text{right}[x]] \end{cases}$$

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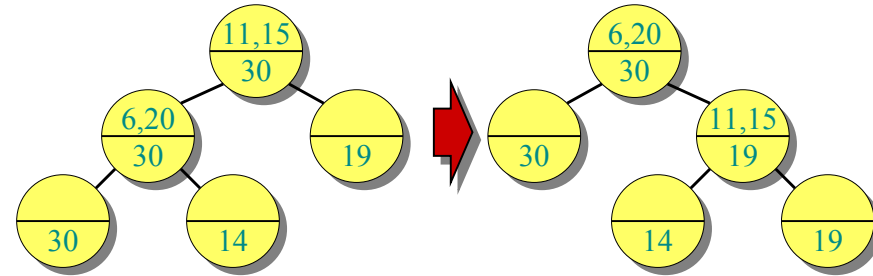
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## Modifying operations

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 =  $O(\log n)$ ; DELETE similar.

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## New operations

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

INTERVAL-SEARCH( $i$ )

$x \leftarrow \text{root}$

**while**  $x \neq \text{NIL}$  and ( $\text{low}[i] > \text{high}[\text{int}[x]]$   
or  $\text{low}[\text{int}[x]] > \text{high}[i]$ )

**do**  $\triangleright i$  and  $\text{int}[x]$  don't overlap

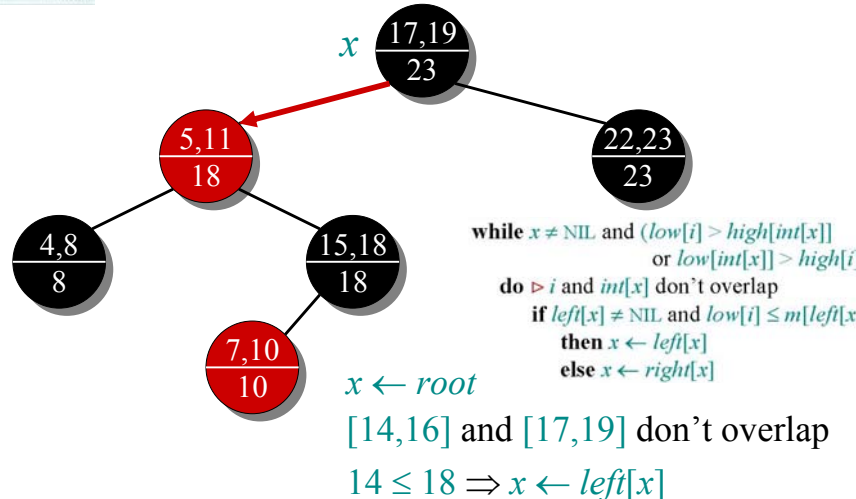
**if**  $\text{left}[x] \neq \text{NIL}$  and  $\text{low}[i] \leq m[\text{left}[x]]$

**then**  $x \leftarrow \text{left}[x]$

**else**  $x \leftarrow \text{right}[x]$

**return**  $x$

## Example 1: INTERVAL-SEARCH([14,16])



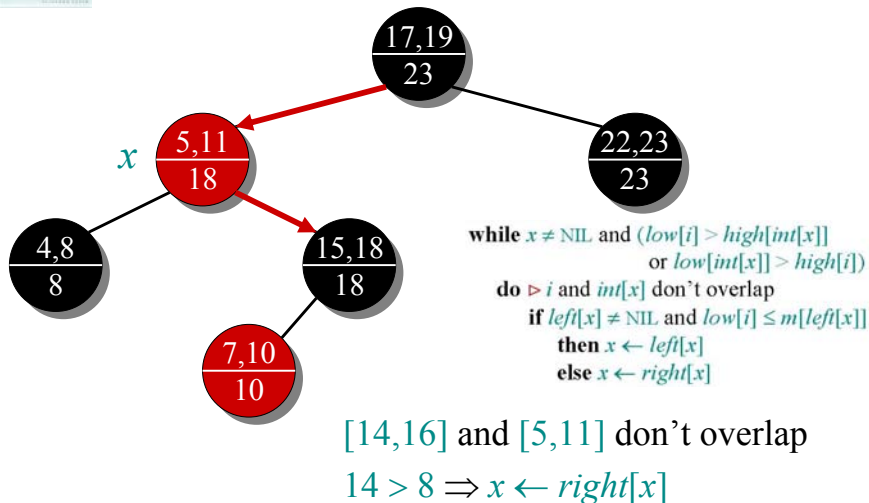
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## Example 1: INTERVAL-SEARCH([14,16])

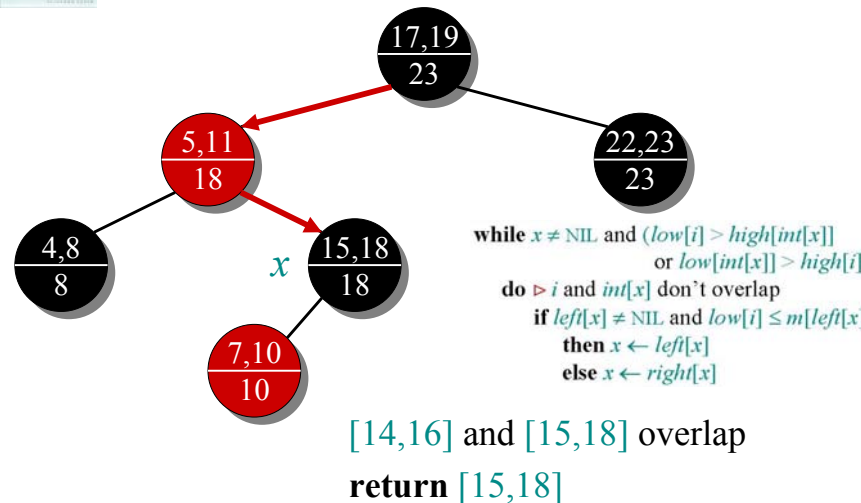


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## Example 1: INTERVAL-SEARCH([14,16])

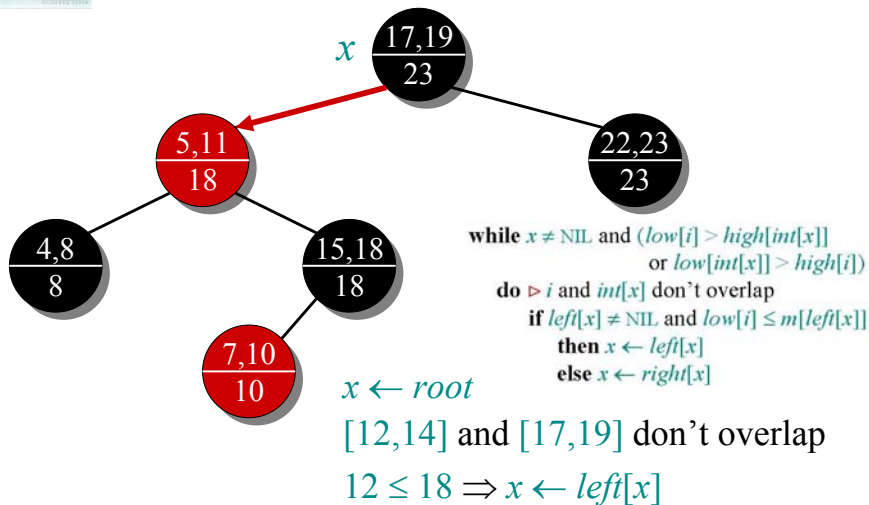


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## Example 2: INTERVAL-SEARCH([12,14])

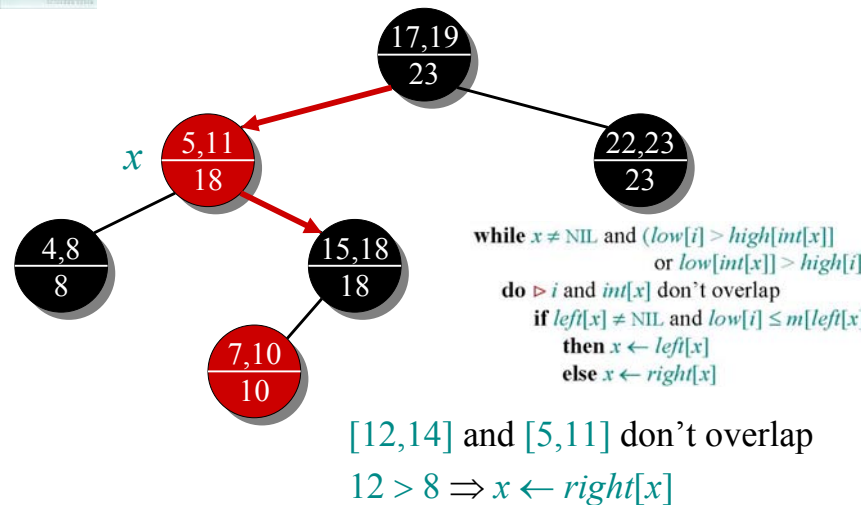


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## Example 2: INTERVAL-SEARCH([12,14])

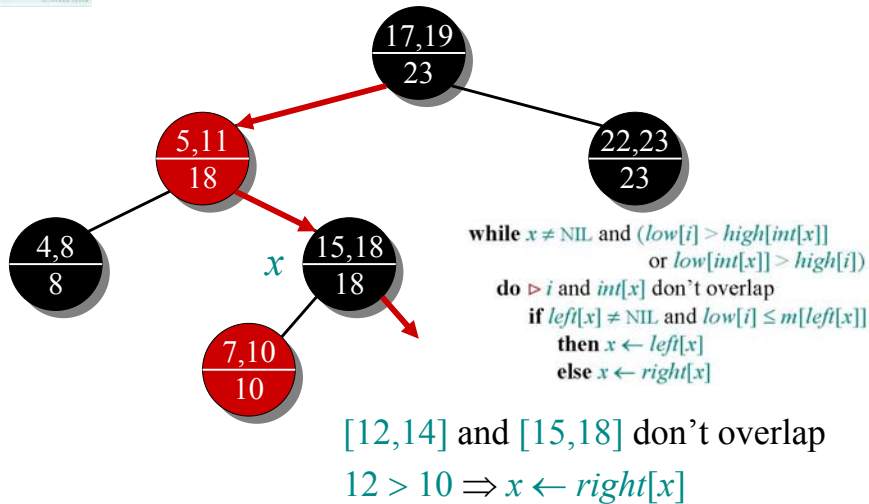


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## Example 2: INTERVAL-SEARCH([12,14])

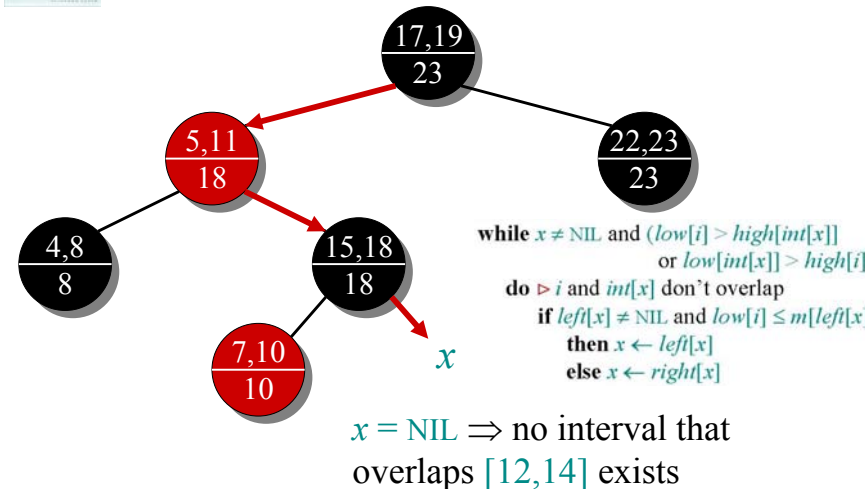


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## Example 2: INTERVAL-SEARCH([12,14])



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## Analysis

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

This is an *output-sensitive* bound.

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

## 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.

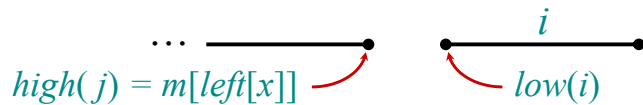
- If the search goes right, then  $\{i' \in L : i' \text{ overlaps } i\} = \emptyset$ .
- If the search goes left, then  $\{i' \in L : i' \text{ overlaps } i\} = \emptyset \Rightarrow \{i' \in R : i' \text{ overlaps } i\} = \emptyset$ .

*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  $left[x] = \text{NIL}$ , 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 right endpoint of some interval  $j \in L$ , and no other interval in  $L$  can have a larger right endpoint than  $high(j)$ .



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

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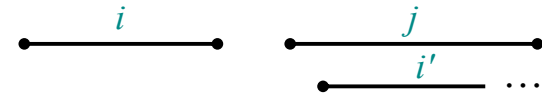


## Proof (continued)

Suppose that the search goes left, and assume that

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

- Then, the code dictates that  $low[i] \leq m[left[x]] = high[j]$  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  $low[j] \leq low[i']$ .
- But then  $\{i' \in R : i' \text{ overlaps } i\} = \emptyset$ .  $\square$



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