

COT 5407: Introduction to Algorithms

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Approach to DP Problems

- Write down a recursive solution
- Use recursive solution to identify list of **subproblems** to solve (there must be overlapping subproblems for effective DP)
- Decide a data structure to store solutions to subproblems (**MEMOIZATION**)
- Write down **Recurrence relation** for solutions of subproblems
- Identify a **hierarchy/order** for subproblems
- Write down non-recursive solution/algorithm

DP Problems

- Find a recursive solution
 - For what purpose?
 - To **reduce** the problem to one or more **simpler** problems
 - reduce the size of the input by imposing conditions
 - e.g., if we know something about last item in input or
 - e.g., if we know how to break up the problem/solution

Because of
“Optimal
Substructure
Property”

Car removal problem

1. Either the last one is removed ...

- ▶ We now have a subproblem with only $N-1$ cars.
 - ▶ Problem with cars 1, 2, ... $N-1$

2. Or it stays ...

- ▶ We retain last car, and get a constrained subproblem as we know that the second to last must match last car.
 - ▶ Problem with cars 1, 2, ... K where K is last car matching car N

List of Subproblems

- This will become clear if we follow the recursion one or two more steps
- In this case:
 - Problems on cars 1, 2, ..., k for different values of k

List of Subproblems

May be refined later

- The inputs to the subproblems are:

$$L_1 = \{c_1\}$$

$$L_2 = \{c_1, c_2\}$$

$$L_3 = \{c_1, c_2, c_3\},$$

...

$$L_n = \text{set of all cars}$$

- Memoization is thus obvious:

$$A[1] = \text{solution to } L_1$$

$$A[2] = \text{solution to } L_2$$

$$A[3] = \text{solution to } L_3$$

...

$$A[n] = \text{solution to } L_n$$

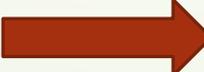
$A[j]$ = least number of cars to be removed when the input is L_j

Recurrence Relation for $A[j]$

1. Either car j is removed ...

- ▶ We now have a subproblem with only $j-1$ cars.
 - ▶ Problem with cars $1, 2, \dots, j-1$
 - ▶ $A[j] = 1 + A[j-1]$

2. Or it stays ...

- ▶ We retain last car, and get a constrained subproblem as we know that the second to last must match last car.
 - ▶ Problem with cars $1, 2, \dots, K$ where K is last car matching car j
 - ▶ $A[j] = (j-K-1) + A[K]$
 - ▶ $A[j] = (j-K-1) + A[K]$  $A[j] = \min_K \{ (j-K-1) + A[K] \}$

Incorrect
Solution

Why is the solution incorrect?

- We don't know whether $A[j]$ refers to a solution that includes car j or not. This will dictate what car can be appended at the end of the solution to this subproblem
- For e.g., if input is
 - $(1,2), (2,3), (3,4), (2,5), (5,6), (6,7)$

Minor change in Memoization

- $A[j]$ = least number of cars to be removed when the input is L_j and car j is included
- $B[j]$ = least number of cars to be removed when the input is L_j and car j is not included

Recurrence Relation for $A[j]$, $B[j]$

1. Either car j is removed ...

- ▶ We now have a subproblem with only $j-1$ cars.
 - ▶ Problem with cars $1, 2, \dots, j-1$
 - ▶ $B[j] = 1 + \min\{A[j-1], B[j-1]\}$

2. Or it stays ...

- ▶ We retain last car, and get a constrained subproblem as we know that the second to last must match last car.
 - ▶ Problem with cars $1, 2, \dots, K$ where K is last car matching car j
 - ▶ $A[j] = \min\{(j-K-1) + A[K]\}$

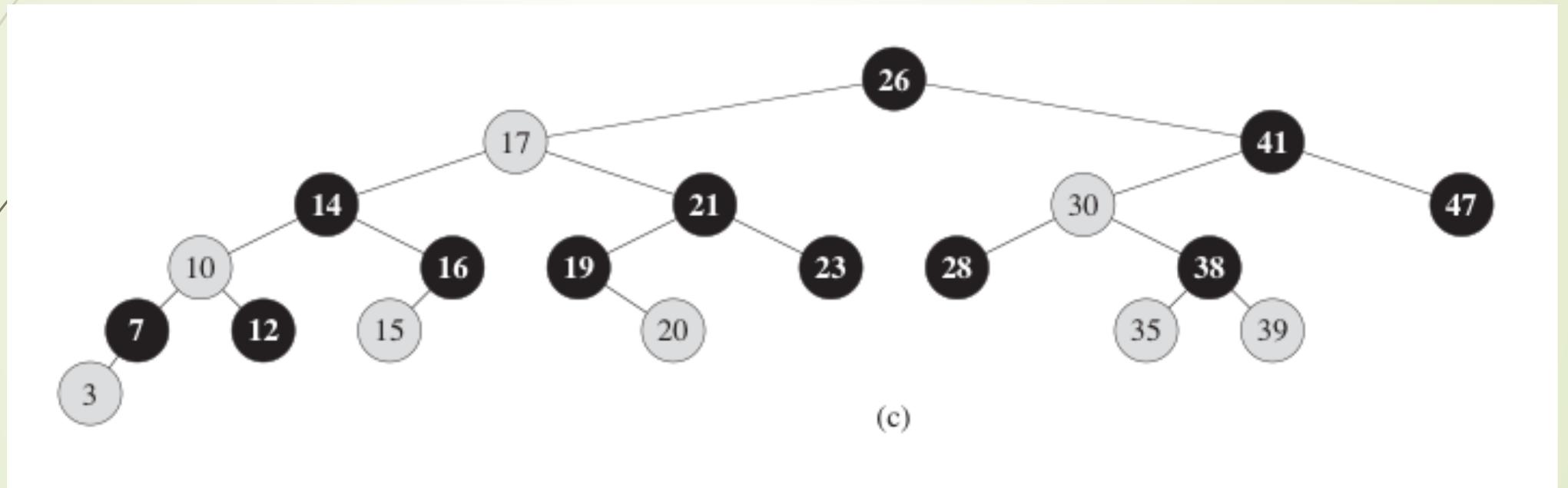
What to return?

➔ $\text{Min} \{ A[n], B[n] \}$

Time Complexity

➔ $O(n^2)$

RB-Trees



OS-Rank

OS-RANK(x,y)

// Different from text (recursive version)

// Find the rank of x in the subtree rooted at y

1 $r = \text{size}[\text{left}[y]] + 1$

2 if $x = y$ then return r

3 else if ($\text{key}[x] < \text{key}[y]$) then

4 return OS-RANK(x,left[y])

5 else return $r + \text{OS-RANK}(x,\text{right}[y])$)

Time Complexity $O(\log n)$

How to augment data structures

- 1. choose an underlying data structure**
- 2. determine additional information to be maintained in the underlying data structure,**
- 3. develop new operations,**
- 4. verify that the additional information can be maintained for the modifying operations on the underlying data structure.**

Augmenting RB-Trees

Theorem 14.1, page 309

Let f be a field that augments a red-black tree T with n nodes, and $f(x)$ can be computed using only the information in nodes x , $\text{left}[x]$, and $\text{right}[x]$, including $f[\text{left}[x]]$ and $f[\text{right}[x]]$.

Then, we can maintain $f(x)$ during insertion and deletion without asymptotically affecting the $O(\log n)$ performance of these operations.

For example,

$$\text{size}[x] = \text{size}[\text{left}[x]] + \text{size}[\text{right}[x]] + 1$$

$$\text{rank}[x] = ?$$

Rank cannot be maintained because of this theorem.

Augmenting information for RB-Trees

- **Parent**
- **Height**
- **Any associative function on all previous values or all succeeding values.**
- **Next**
- **Previous**

Augmented Info

- **OddSize[v]**
 - Number of odd valued nodes in subtree rooted at v
- **It can be maintained because:**
 - $\text{OddSize}[v] =$
 $\text{OddSize}[\text{Left}[v]]$
 $+ \text{OddSize}[\text{Right}[v]]$
 $+ (\text{key}[v] \% 2)$

OS-SoOdd

OS-SoOdd(x,y)

// Different from text (recursive version)

// Find the rank of x in the subtree rooted at y

1 $r = \text{OddSize}[\text{left}[y]] + \text{key}[x] \% 2$

2 if $x = y$ then return r

3 else if ($\text{key}[x] < \text{key}[y]$) then

4 return OS-SoOdd (x, left[y])

5 else return $r + \text{OS-SoOdd}$ (x, right[y])

Time Complexity $O(\log n)$

More Dynamic Operations

	Search	Insert	Delete	Comments
Unsorted Arrays	$O(N)$	$O(1)$	$O(N)$	
Sorted Arrays	$O(\log N)$	$O(N)$	$O(N)$	
Unsorted Linked Lists	$O(N)$	$O(1)$	$O(N)$	
Sorted Linked Lists	$O(N)$	$O(N)$	$O(N)$	
Binary Search Trees	$O(H)$	$O(H)$	$O(H)$	$H = O(N)$
Balanced BSTs	$O(\log N)$	$O(\log N)$	$O(\log N)$	As $H = O(\log N)$

	Se/In/De	Rank	Select	Comments
Balanced BSTs	$O(\log N)$	$O(N)$	$O(N)$	
Augmented BBSTs	$O(\log N)$	$O(\log N)$	$O(\log N)$	