

# COT 5407: Introduction to Algorithms

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[www.cs.fiu.edu/~giri/teach/5407S19.html](http://www.cs.fiu.edu/~giri/teach/5407S19.html)

# 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:

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

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

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

...

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

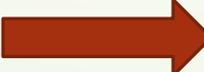
**$A[j]$  = least number of cars to be removed when the input is  $L_j$**

# 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?

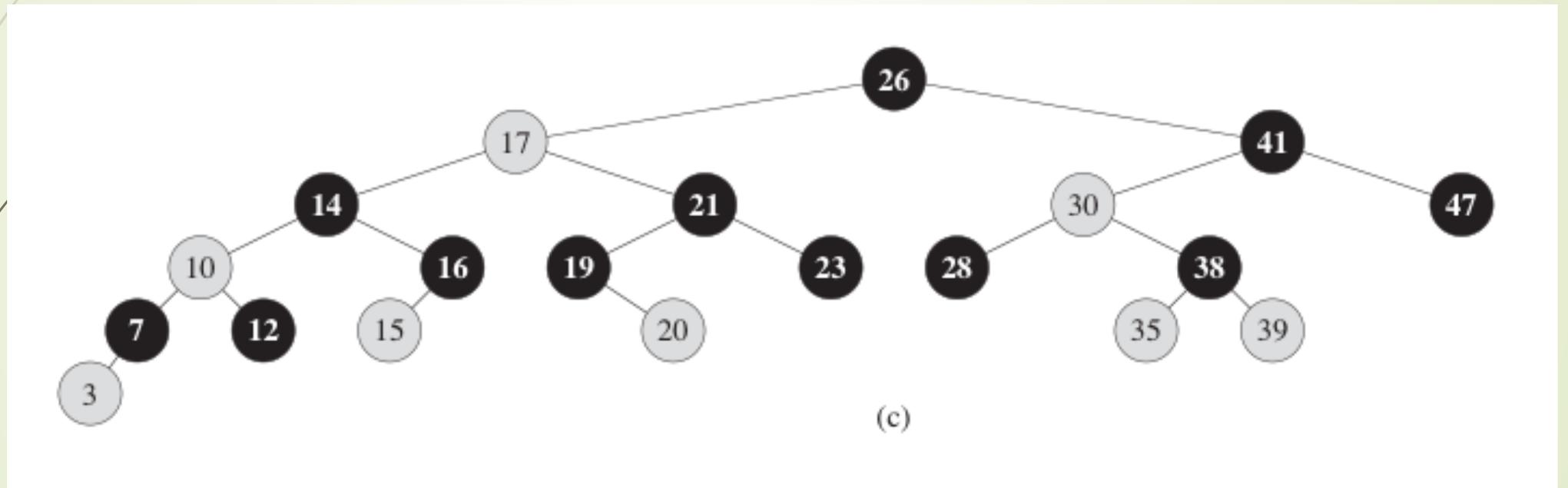
➔  $\text{Min} \{ A[n], B[n] \}$

12

# 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	$O(\log N)$	$O(\log N)$	$O(\log N)$	