



CS 141 F20  
DISCUSSION WEEK 5

Dynamic programming  
Midterm problems

## MIDTERM PROBLEM 3B

(B): It first solves three subproblems of size  $n - 1$ , then combines the solutions of the subproblems in constant time.

$$T(n) = 3T(n - 1) + \Theta(1)$$

### Recurrence tree:

There is  $\Theta(1) = c$  at the top. There are 3 nodes in the level below it, each with  $c$  cost, thus  $3c$  in total. In the level below that, it would be  $9c$ .

Size of  $n$  decreases by 1 at each level and number of node multiples by 3. There would be a total of  $n$  levels. At the lowest level, there would be  $3^{n-1}$ , thus  $c * 3^{n-1}$  cost, at the lowest level. (**Recall:**  $n^{th}$  term of GP series is  $ar^{n-1}$ ).

$$\begin{aligned} \text{Total cost of the tree} &= c + 3c + 9c + \dots + c * 3^{n-1} \\ &= c * \frac{3^n - 1}{3 - 1} = c * \frac{3^n - 1}{2} = \Theta(3^n) \end{aligned}$$

## MIDTERM PROBLEM 3B

(B): It first solves two subproblems of size  $n - 2$ , then combines the solutions of the subproblems in constant time.

$$T(n) = 2T(n - 2) + \Theta(1)$$

$$\Theta(1) = c$$

$$T(1) = c$$

$$T(n) = 2\{2T(n - 4) + c\} + c$$

$$= 4T(n - 4) + c\{2 + 1\}$$

$$= 4\{2T(n - 6) + c\} + c\{2 + 1\}$$

$$= 8T(n - 6) + c\{2^2 + 2 + 1\}$$

$$= 2^i T(n - 2 * i) + c\{2^{i-1} + \dots + 2 + 1\}$$

$$n - 2 * i = 1$$

$$\Rightarrow 2 * i = n - 1$$

$$\Rightarrow i = (n - 1) / 2 \approx n / 2$$

$$= 2^{n/2} T(1) + c\{2^{n/2 - 1} + \dots + 2 + 1\}$$

$$= c * 2^{n/2} + c * \frac{2^{n/2} - 1}{2 - 1}$$

$$= c * 2^{n/2} + c * \{2^{n/2} - 1\}$$

$$= c * 2^{n/2} + c * 2^{n/2} - c$$

$$= c * 2^{n/2 + 1} - c$$

$$= \Theta(2^{n/2})$$

## MIDTERM PROBLEM 4

### Proof for optimal substructure

$S$  is the set of all the students. Let  $B$  be the optimal solution for problem with  $S$  students. Student  $a$  and  $b$  travel in the same boat.  $S'$  is a subset of students, where  $S = S' \cup \{a, b\}$ .  $B'$  would be the optimal solution to the subproblem  $S'$ .  $B = B' + 1$ .

Lets assume that  $B'$  is not the optimal solution to the problem  $S'$ . Then there would be another solution  $B''$ , such that  $B'' < B'$ , which would be optimal for  $S'$ . Now, if we add the boat with  $\{a, b\}$  to the solution of  $B''$ , then we will get  $B'' + 1$  for  $S$ . Now,  $B'' + 1 \leq B' + 1 \Rightarrow B'' + 1 \leq B$ . Thus we got a more optimal solution, which is a contradiction. Thus, our assumption is wrong. Optimal substructure is thus proved.

## MIDTERM PROBLEM 4

### Proof for greedy choice

Assume that we have an optimal solution to this problem where we do not pair student  $i$  with student  $j$ . Instead, we pair student  $i$  with student  $k$ .

There are two possible scenarios here. First one is the scenario where  $k = -1$ . In that case, by looking at the rowing algorithm, we know that student  $j$  is the heaviest student whose weight is less than  $T - w_i$  ( $j = -1$  if no such student exists). If  $j$  and  $k$  are both equal to  $-1$ , then the rowing algorithm's solution and the optimal solution are identical. Otherwise, rowing algorithm gives us a better solution than the alleged optimal solution in our assumption and we reach a contradiction.

Second scenario is when  $k \neq -1$ . Since student  $j$  is the heaviest student with  $w_j < T - w_i$ , if we take student  $k$  out of the boat and put student  $j$  in the boat instead, we will utilize the capacity of the boat more efficiently ( $T \geq w_i + w_j > w_i + w_k$ ). This contradicts our starting assumption since we found a better solution.  $\square$

# MIDTERM PROBLEM 5

- CLRS 9.2

# MIDTERM PROBLEM 6

- Incorrect assignment of 0/1
- After forming the tree
  - Heavier branches should be assigned 1.
  - Lighter branches should be assigned 0.



# DYNAMIC PROGRAMING



## A simplified case

- Overall weight limit: 8 lb
- Item 1: 5 lb, \$150
- Item 2: 4 lb, \$100
- Item 3: 1 lb, \$10
  
- Solution 1: Item 1 + Item 3 \* 3, value: \$180
- Solution 2: Item 2 \* 2, value: \$200
  
- Greedy strategy does not provide the optimal solution
- A naïve solution? Try all possibilities!

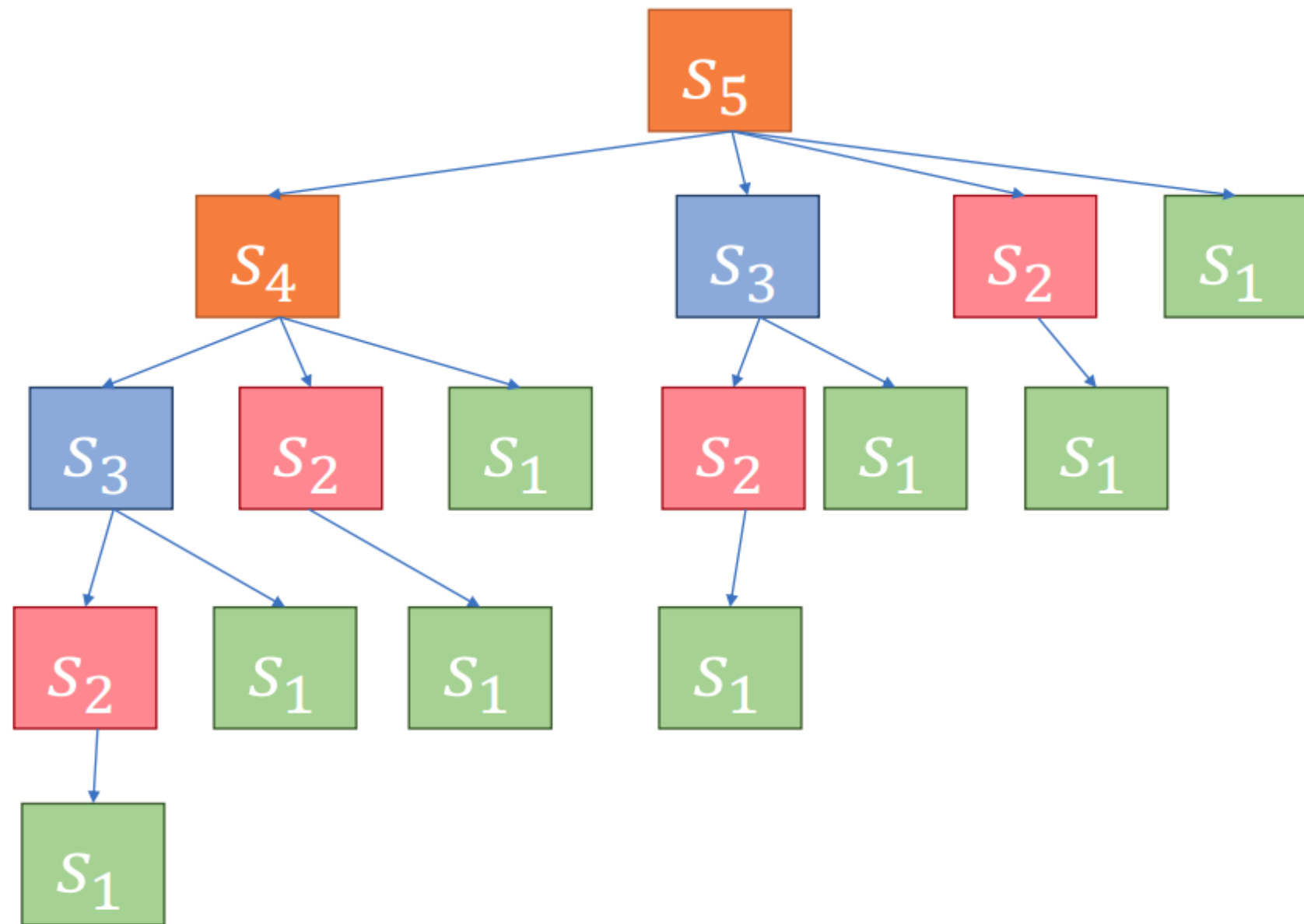
# A naïve algorithm

```
int suitcase(int leftWeight) {  
    int curBest = 0;  
    foreach item (weight, value)  
        if (leftWeight >= weight)  
            curBest = max(curBest, suitcase(leftWeight - Weight) + value);  
    return curBest;  
}
```

```
answer = suitcase(50);
```

# Execution Recurrence Tree

Assume we have for items with weights 1,2,3,4, and the overall weight is 5



# A DP algorithm

```
int suitcase(int leftWeight) {  
    if (ans[leftWeight] != -1) return ans[leftWeight];  
    int curBest = 0;  
    foreach item (weight, value)  
        if (leftWeight >= weight)  
            curBest = max(curBest, suitcase(leftWeight - Weight) + value);  
    return ans[leftWeight] = curBest;  
}
```

```
int ans[50] = {-1, ... , -1};  
answer = suitcase(50);
```

# Recursive Solution

- Define  $s_i$  as the maximum value you can get for a total weight of  $i$
- We can express  $s_i$  as the following **recurrence**:

$$s_i = \max \left\{ \begin{array}{l} 0 \\ \max_{(w_j, v_j) \text{ is an item}} \{s_{i-w_j} + v_j\} \mid i > w_j \end{array} \right.$$

- Final answer is  $s_n$

## ISSUE WITH CURRENT APPROACH

- Can we pick an item more than once?
- If so, do we need to change the solutions discussed earlier?
- If we use just  $S_j$ , it is not possible to know if an item has been used before
- Let's also keep track of the items.
- Let  $S_{i,j}$  be the optimal value for **total max weight of  $i$  using only first  $j$  item**

## HOW DO WE GET $S_{i,j}$ ?

- There are 2 options for getting the optimal value for weight  $i$  and for first  $j$  items:
  - Item  $j$  is picked:  $S_{i,j} = S_{i-w_j, j-1} + v_j$
  - Item  $j$  is not picked:  $S_{i,j} = S_{i, j-1}$
  - Which one is the best??
    - We pick the maximum of both

## Another way to state the relationship of $s_{i,j}$

- **The recurrence:**

$$s_{i,j} = \max \begin{cases} s_{i,j-1} \\ s_{i-w_j,j-1} + v_j \end{cases} \quad i \geq w_j$$

- **The boundary:**  $s_{i,0} = 0$

- **The recurrence cannot be circular**

- You can not have **states** a, b, and c that computing a relied on b, b on c, and c on a



# EXAMPLE

- Assume we have one copy of each of these items  $\{(\$5, 2\text{lbs}), (\$4, 1\text{lbs}), (\$3, 1\text{lbs})\}$  and capacity  $W=3$ .
- Let us fill the dynamic programming table (Calculate all  $S_{\{i,j\}}$  for  $i \leq 3$  (which is  $W$ ) and  $j \leq 3$  (total number of items))

$$s_{i,j} = \max \begin{cases} s_{i,j-1} \\ s_{i-w_j, j-1} + v_j \end{cases} \quad i \geq w_j$$

i: cols j:rows	0	1	2	3
0	0	0	0	0
1	0	0	5	5
2	0	4	5	9
3	0	4	7	9