

DESIGN AND ANALYSIS OF ALGORITHMS

Memory Function Knapsack

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MEMORY FUNCTION KNAPSACK

UNIT 5: Limitations of Algorithmic Power and Coping with the Limitations

- Dynamic Programming
 - ► Computing a Binomial Coefficient
 - ► The Knapsack Problem
 - Memory Functions
 - Warshall's and Floyd's Algorithms
- Limitations of Algorithmic Power
 - Lower-Bound Arguments
 - Decision Trees
 - P, NP, and NP-Complete, NP-Hard Problems
- Coping with the Limitations
 - Backtracking
 - Branch-and-Bound. Architecture (microprocessor instruction set)

Concepts covered

- Memory Function Knapsack
 - Motivation
 - Algorithm
 - Example







Advantage of bottom up approach: each value computed only once



- Advantage of bottom up approach: each value computed only once
- Example computed bottom up:

		ca	pacit	y <i>j</i>	
i	1	2	3	4	5
1	0	12	12	12	12
2	10	12	22	22	22
3	10	12	22	30	32
4	10	15	25	30	37



- Advantage of bottom up approach: each value computed only once
- Example computed bottom up:

	capacity <i>j</i>					
i	1	2	3	4	5	
1	0	12	12	12	12	
2	10	12	22	22	22	
3	10	12	22	30	32	
4	10	15	25	30	37	

Disadvantage of bottom up approach: values not required also computed

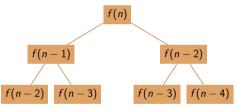




Disadvantage of top down approach: same problem solved multiple times

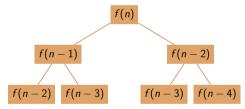


- Disadvantage of top down approach: same problem solved multiple times
- Example computed top down:





- Disadvantage of top down approach: same problem solved multiple times
- Example computed top down:



Advantage of top down approach: only the required subproblems solved



MEMORY FUNCTION KNAPSACK Memory Function Dynamic Programming



- Combine the advantages of bottom up and top down approaches:
 - compute each subproblem only once
 - compute only the required subproblems

MEMORY FUNCTION KNAPSACK MF-DP Algorithm



Algorithm for Memory Function Dynamic Programming

```
1: procedure MFKNAPSACK(i, j)
2:
        ▶ Inputs: i indicating the number items, and
        ▷ j, indicating the knapsack capacity
 3:
4:
        Dutput: The value of an optimal feasible subset of the first i items
       \triangleright Note: Uses global variables input arrays Weights[1...n], Values[1...n],
5:
       \triangleright and table F[0...n,0...W] whose entries are initialized with -1's except
6:
        > row 0 and column 0 is initialized with 0
7:
        if F[i, j] < 0 then
8:
           if i < Weights[i] then
9:
                value \leftarrow MFKnapsack(i-1, j)
10:
           else value \leftarrow max(MFKnapsack(i-1,j), Values[i] + MFKnapsack(i-1,j-Weights[i]))
11:
                F[i, i] \leftarrow value
12:
       return F[i, j]
13:
```



$$F(i,j) = egin{cases} \max(F(i-1,j), & v_i + F(i-1,j-w_i)) & \text{if } j - w_i \geq 0 \\ F(i-1,j) & \text{if } j - w_i < 0 \end{cases}$$

Dynamic Programming Example

item i	weight <i>w_i</i>	value v_i
1	2	12
2	1	10
3	3	20
4	2	15



$$F(i,j) = egin{cases} \max(F(i-1,j), & v_i + F(i-1,j-w_i)) & \text{if } j - w_i \geq 0 \\ F(i-1,j) & \text{if } j - w_i < 0 \end{cases}$$

Dynamic Programming Example

item i	weight <i>w_i</i>	value <i>v_i</i>
1	2	12
2	1	10
3	3	20
4	2	15

		-	сара	city	j	
i	0	1	2	3	4	5
0	0					
1						
2						
3						
4						



$$F(i,j) = \begin{cases} \max(F(i-1,j), & v_i + F(i-1,j-w_i)) & \text{if } j - w_i \ge 0 \\ F(i-1,j) & \text{if } j - w_i < 0 \end{cases}$$

Dynamic Programming Example

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			сара	city	j	
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Dynamic Programming Example

item i	weight <i>w_i</i>	value v_i
1	2	12
2	1	10
3	3	20
4	2	15

			capa	city	j	
i	0	1	2	3	4	5
0	0					
1						
2						
3						
4						



$$F(i,j) = egin{cases} \max(F(i-1,j), & v_i + F(i-1,j-w_i)) & \text{if } j - w_i \geq 0 \\ F(i-1,j) & \text{if } j - w_i < 0 \end{cases}$$

Dynamic Programming Example

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1	2	12
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4	2	15

			сара	city j	'	
i	0	1	2	3	4	5
0	0					
1						
2						
3						
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$$F(i,j) = egin{cases} \max(F(i-1,j), & v_i + F(i-1,j-w_i)) & \text{if } j - w_i \geq 0 \\ F(i-1,j) & \text{if } j - w_i < 0 \end{cases}$$

Dynamic Programming Example

item i	weight <i>w_i</i>	value v_i
1	2	12
2	1	10
3	3	20
4	2	15

	capacity <i>j</i>						
i	0	1	2	3	4	5	
0	0						
1							
2		-			-		
3		-	-		-		
4		-	-	-	-		



$$F(i,j) = egin{cases} \max(F(i-1,j), & v_i + F(i-1,j-w_i)) & \text{if } j - w_i \geq 0 \\ F(i-1,j) & \text{if } j - w_i < 0 \end{cases}$$

Dynamic Programming Example

item i	weight <i>w_i</i>	value <i>v_i</i>
1	2	12
2	1	10
3	3	20
4	2	15

capacity <i>j</i>						
i	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	12	12	12	12
2	0	_	12	22	-	22
3	0	_	_	22	-	32
4	0	-	-	-	-	37



$$F(i,j) = \begin{cases} \max(F(i-1,j), & v_i + F(i-1,j-w_i)) & \text{if } j - w_i \ge 0 \\ F(i-1,j) & \text{if } j - w_i < 0 \end{cases}$$

Dynamic Programming Example

item i	weight <i>w_i</i>	value <i>v_i</i>
1	2	12
2	1	10
3	3	20
4	2	15

What is the maximum value that can be stored in a knapsack of capacity 5?

	capacity <i>j</i>					
i	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	12	12	12	12
2	0	_	12	22	-	22
3	0	_	_	22	-	32
4	0	-	-	-	-	37

Knapsack problem solved by

- computing 21 out 30 possible subproblems
- reusing subproblem entry (1, 2)

MEMORY FUNCTION KNAPSACK Complexity



- Constant factor improvement in efficiency
 - ▶ Space complexity: $\Theta(nW)$
 - ▶ Time complexity: $\Theta(nW)$
 - ▶ Time to compose optimal solution: O(n)
- Bigger gains possible where computation of a subproblem takes more than constant time

MEMORY FUNCTION KNAPSACK

Think About It



 Consider the use of the MF technique to compute binomial coefficient using the recurrence

$$C(n, k) = C(n-1, k-1) + C(n-1, k)$$

- How many table entires are filled?
- How many are reused?