
Design and Analysis of Algorithms

CSE 5311

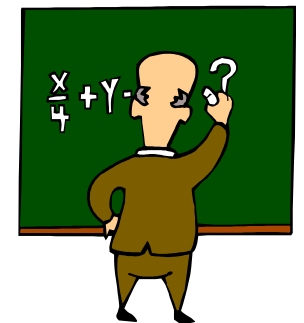
Lecture 13 Dynamic Programming

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The General Dynamic Programming Technique

- Applies to a problem that at first seems to require a lot of time (possibly exponential), provided we have:
 - **Subproblem optimality:** the global optimum value can be defined in terms of optimal subproblems
 - **Subproblem overlap:** the subproblems are not independent, but instead they overlap (hence, should be constructed bottom-up).



The 0/1 Knapsack Problem



- Given: A set S of n items, with each item i having
 - w_i - a positive weight
 - b_i - a positive benefit
- Goal: Choose items with maximum total benefit but with weight at most W .
- If we are **not** allowed to take fractional amounts, then this is the **0/1 knapsack problem**.
 - In this case, we let T denote the set of items we take

- Objective: maximize
$$\sum_{i \in T} b_i$$

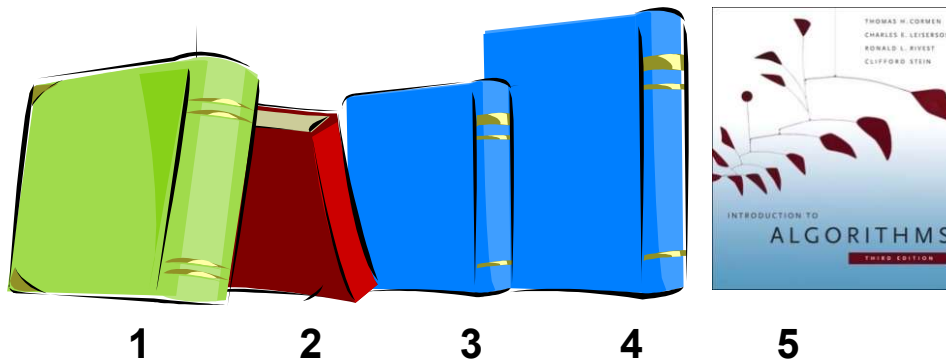
- Constraint:
$$\sum_{i \in T} w_i \leq W$$

Example



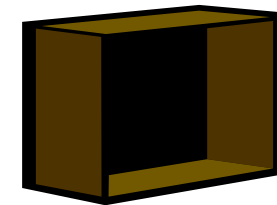
- Given: A set S of n items, with each item i having
 - b_i - a positive “benefit”
 - w_i - a positive “weight”
- Goal: Choose items with maximum total benefit but with weight at most W .

Items:



Weight:	4 in	2 in	2 in	6 in	2 in
Benefit:	\$20	\$3	\$6	\$25	\$80

“knapsack”



box of width 9 in

Solution:

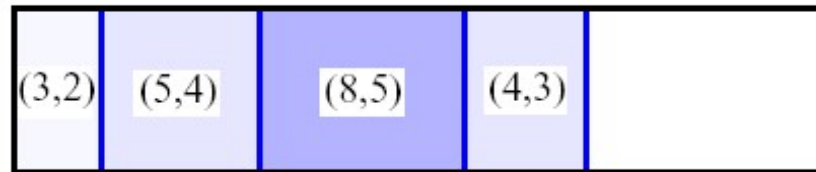
- item 5 (\$80, 2 in)
- item 3 (\$6, 2in)
- item 1 (\$20, 4in)

A 0/1 Knapsack Algorithm, First Attempt

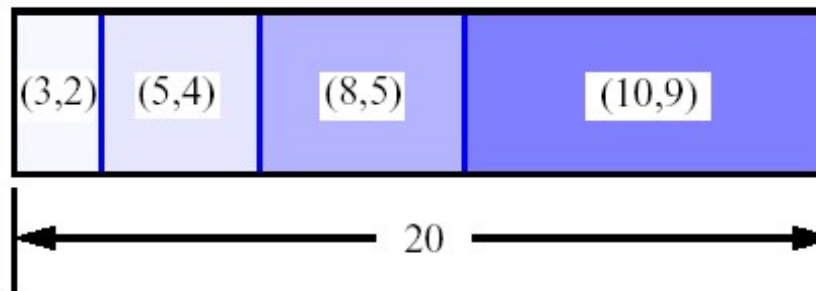


- S_k : Set of items numbered 1 to k .
- Define $B[k] =$ best selection from S_k .
- Problem: does not have subproblem optimality:
 - Consider set $S = \{(3,2), (5,4), (8,5), (4,3), (10,9)\}$ of (benefit, weight) pairs and total weight $W = 20$

Best for S_4 :



Best for S_5 :



0-1 Knapsack problem: brute-force approach



- *A straightforward algorithm*

- Since there are n items, there are 2^n possible combinations of items.
- We go through all combinations and find the one with maximum value and with total weight less or equal to W
- Running time will be $O(2^n)$

- *Can we do better?*

- Yes, with an algorithm based on dynamic programming
- We need to carefully identify the subproblems, Let's try this:
- If items are labeled $1..n$, then a subproblem would be to find an optimal solution for $S_k = \{\text{items labeled } 1, 2, .. k\}$

- *Define a subproblem*

- This is a reasonable subproblem definition.
- The question is: can we describe the final solution (S_n) in terms of subproblems (S_k)?
- Unfortunately, we can't do that.

Define a Subproblem



$w_1=2$	$w_2=4$	$w_3=5$	$w_4=3$
$b_1=3$	$b_2=5$	$b_3=8$	$b_4=4$

Max weight: $W = 20$

For S_4 : Total weight: 14;

Maximum benefit: 20

$w_1=2$	$w_2=4$	$w_3=5$	$w_5=9$
$b_1=3$	$b_2=5$	$b_3=8$	$b_5=10$

For S_5 : Total weight: 20

Maximum benefit: 26

Item	Weight	Benefit
1	2	3
2	4	5
3	5	8
4	3	4
5	9	10

Diagram showing subproblems: S_4 is indicated by a bracket covering items 1, 2, 3, and 4. S_5 is indicated by a larger bracket covering items 1, 2, 3, 4, and 5.

Solution for S_4 is not part of the solution for S_5 !!!

Define a Subproblem



- As we have seen, the solution for S_4 is not part of the solution for S_5
- So our definition of a subproblem is flawed and we need another one!
- Let's add another parameter: w , which will represent the exact weight for each subset of items
- The subproblem then will be to compute **$B[k,w]$**

Recursive Formulation



- S_k : Set of items numbered 1 to k .
- Define $B[k, w]$ to be the best selection from S_k with weight at most w
- Good news: this does have subproblem optimality.

$$B[k, w] = \begin{cases} B[k-1, w] & \text{if } w_k > w \\ \max\{B[k-1, w], B[k-1, w-w_k] + b_k\} & \text{else} \end{cases}$$

- I.e., the best subset of S_k with weight at most w is either
 - the best subset of S_{k-1} with weight at most w or
 - the best subset of S_{k-1} with weight at most $w-w_k$ plus item k
- Two cases (contains item k or not)
 - First case: $w_k > w$. Item k can't be part of the solution, since if it was, the total weight would be $> w$, which is unacceptable.
 - Second case: $w_k < w$. Then the item k can be in the solution, and we choose the case with greater value.

0/1 Knapsack Algorithm



for $w = 0$ to W

$B[0,w] = 0$

$O(W)$

for $i = 1$ to n

$B[i,0] = 0$

for $i = 1$ to n

Repeat n times

for $w = 0$ to W

$O(W)$

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1,w-w_i] > B[i-1,w]$

$B[i,w] = b_i + B[i-1,w-w_i]$

else

$B[i,w] = B[i-1,w]$

else $B[i,w] = B[i-1,w]$ // $w_i > w$

What is the running time of this algorithm? $O(n*W)$

Remember that the brute-force algorithm takes $O(2^n)$

Example



Let's run our algorithm on the following data:

$n = 4$ (# of elements)

$W = 5$ (max weight)

Elements (weight, benefit):

$(2,3), (3,4), (4,5), (5,6)$

Example (2)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1						
2						
3						
4						

for $w = 0$ to W
 $B[0,w] = 0$

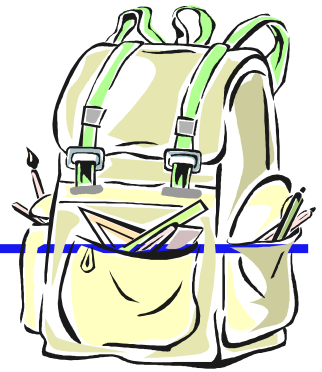
Example (3)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0					
2	0					
3	0					
4	0					

for $i = 1$ to n
 $B[i,0] = 0$

Example (4)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0				
2	0					
3	0					
4	0					

$i=1$

$b_i=3$

$w_i=2$

$w=1$

$w-w_i$

$=-1$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

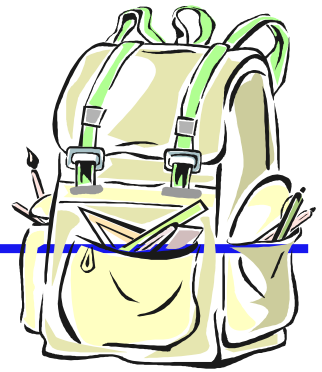
$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (5)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3			
2	0					
3	0					
4	0					

$i=1$

$b_i=3$

$w_i=2$

$w=2$

$w-w_i$

$=0$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

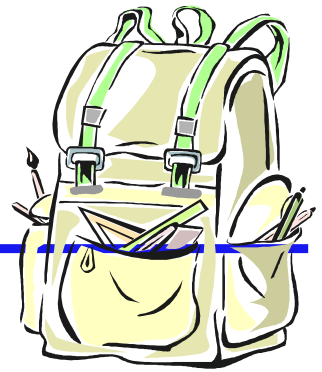
$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (6)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3		
2	0					
3	0					
4	0					

$i=1$
 $b_i=3$
 $w_i=2$
 $w=3$
 $w-w_i=1$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

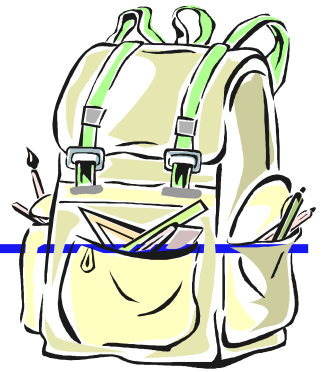
$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (7)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	
2	0					
3	0					
4	0					

$i=1$

$b_i=3$

$w_i=2$

$w=4$

$w-w_i$
 $=2$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

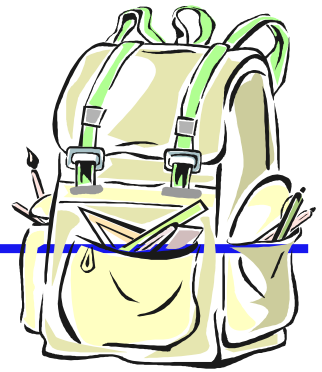
$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (8)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0					
3	0					
4	0					

$i=1$
 $b_i=3$
 $w_i=2$
 $w=5$
 $w-w_i$
 $=3$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

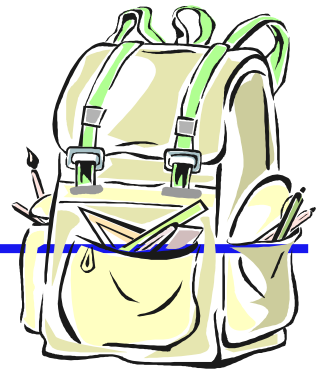
$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (9)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0				
3	0					
4	0					

$i=2$
 $b_i=4$
 $w_i=3$
 $w=1$
 $w-w_i$
 $=-2$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (10)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3			
3	0					
4	0					

$i=2$
 $b_i=4$
 $w_i=3$
 $w=2$
 $w-w_i$
 $=-1$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

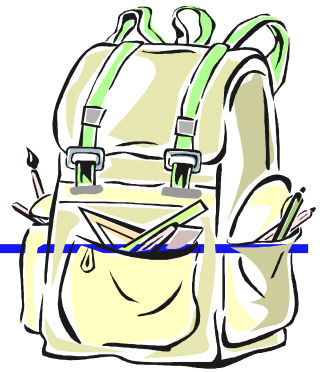
$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (11)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4		
3	0					
4	0					

$i=2$
 $b_i=4$
 $w_i=3$
 $w=3$
 $w-w_i$
 $=0$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

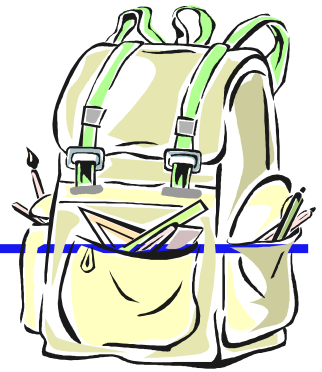
$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (12)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	
3	0					
4	0					

$i=2$
 $b_i=4$
 $w_i=3$
 $w=4$
 $w-w_i=1$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (13)

i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0					
4	0					

$i=2$
 $b_i=4$
 $w_i=3$
 $w=5$
 $w-w_i$
 $=2$



Items:

- 1: (2,3)
- 2: (3,4)
- 3: (4,5)
- 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (14)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0				
4	0					

$i=3$
 $b_i=5$
 $w_i=4$
 $w=1$
 $w-w_i$
 $=-3$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (15)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3			
4	0					

$i=3$
 $b_i=5$
 $w_i=4$
 $w=2$
 $w-w_i$
 $=-2$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (16)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4		
4	0					

$i=3$
 $b_i=5$
 $w_i=4$
 $w=3$
 $w-w_i$
 $=-1$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (17)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	
4	0					

$i=3$
 $b_i=5$
 $w_i=4$
 $w=4$
 $w-w_i=0$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

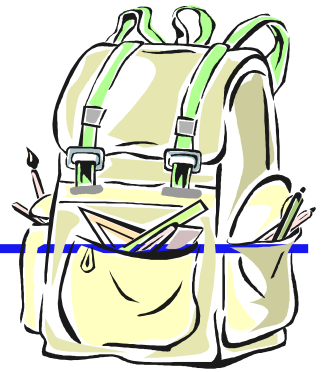
$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (18)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0					

$i=3$
 $b_i=5$
 $w_i=4$
 $w=5$
 $w-w_i=1$

Items:

- 1: (2,3)
- 2: (3,4)
- 3: (4,5)
- 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (19)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0				

$i=4$
 $b_i=6$
 $w_i=5$
 $w=1$
 $w-w_i$
 $=-4$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (20)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3			

$i=4$
 $b_i=6$
 $w_i=5$
 $w=2$
 $w-w_i$
 $=-3$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (21)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4		

$i=4$
 $b_i=6$
 $w_i=5$
 $w=3$
 $w-w_i$
 $=-2$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (22)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	

$i=4$
 $b_i=6$
 $w_i=5$
 $w=4$
 $w-w_i$
 $=-1$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$B[i, w] = b_i + B[i-1, w-w_i]$

else

$B[i, w] = B[i-1, w]$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Example (23)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	7

$i=4$
 $b_i=6$
 $w_i=5$
 $w=5$
 $w-w_i=0$

- Items:
- 1: (2,3)
 - 2: (3,4)
 - 3: (4,5)
 - 4: (5,6)

if $w_i \leq w$ // item i can be part of the solution

if $b_i + B[i-1, w-w_i] > B[i-1, w]$

$$B[i, w] = b_i + B[i-1, w - w_i]$$

else

$$B[i, w] = B[i-1, w]$$

else $B[i, w] = B[i-1, w]$ // $w_i > w$

Comments

- This algorithm only finds the max possible value that can be carried in the knapsack
 - I.e., the value in $B[n,W]$
- To know the items that make this maximum value, an addition to this algorithm is necessary.

How to find actual Knapsack Items

- All of the information we need is in the table.
- $B[n, W]$ is the maximal value of items that can be placed in the Knapsack.

Let $i=n$ and $k=W$

if $B[i, k] \neq B[i-1, k]$ then mark the i th item as in the knapsack

$i = i-1, k = k-w_i$

else

$i = i-1$ // Assume the i th item is not in the knapsack

// Could it be in the optimally packed knapsack?

Find the Items (1)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	7

$i=4$

$k=5$

$b_i=6$

$w_i=5$

$B[i,k]=7$

$B[i-1,k]=7$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

$i=n, k=W$

while $i, k > 0$

if $B[i,k] \neq B[i-1,k]$ then

mark the i th item as in the knapsack

$i = i-1, k = k-w_i$

else

$i = i-1$

Find the Items (2)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	7

$i=4$

$k=5$

$b_i=6$

$w_i=5$

$B[i,k]=7$

$B[i-1,k]=7$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

$i=n, k=W$

while $i, k > 0$

if $B[i,k] \neq B[i-1,k]$ then

mark the i th item as in the knapsack

$i = i-1, k = k-w_i$

else

$i = i-1$

Find the Items (3)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	7

$i=3$

$k=5$

$b_i=5$

$w_i=4$

$B[i,k]=7$

$B[i-1,k]=7$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

$i=n, k=W$

while $i, k > 0$

if $B[i,k] \neq B[i-1,k]$ then

mark the i th item as in the knapsack

$i = i-1, k = k-w_i$

else

$i = i-1$

Find the Items (4)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	7

$i=2$

$k=5$

$b_i=4$

$w_i=3$

$B[i,k]=7$

$B[i-1,k]=3$

$k-w_i=2$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

$i=n, k=W$

while $i, k > 0$

if $B[i,k] \neq B[i-1,k]$ then

mark the i th item as in the knapsack

$i = i-1, k = k-w_i$

else

$i = i-1$

Find the Items (5)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	7

$i=1$

$k=2$

$b_i=3$

$w_i=2$

$B[i,k]=3$

$B[i-1,k]=0$

$k-w_i=0$

Items:

1: (2,3)

2: (3,4)

3: (4,5)

4: (5,6)

$i=n, k=W$

while $i, k > 0$

if $B[i,k] \neq B[i-1,k]$ then

mark the i th item as in the knapsack

$i = i-1, k = k-w_i$

else

$i = i-1$

Find the Items (6)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	7

$i=0$
 $k=0$

Items:

- 1: (2,3)
- 2: (3,4)
- 3: (4,5)
- 4: (5,6)

$i=n, k=W$

while $i, k > 0$

if $B[i,k] \neq B[i-1,k]$ then

mark the i th item as in the knapsack

$i = i-1, k = k-w_i$

else

$i = i-1$

The optimal knapsack should contain **{1, 2}**

Find the Items (7)



i/W	0	1	2	3	4	5
0	0	0	0	0	0	0
1	0	0	3	3	3	3
2	0	0	3	4	4	7
3	0	0	3	4	5	7
4	0	0	3	4	5	7

$i=n, k=W$

while $i, k > 0$

if $B[i,k] \neq B[i-1,k]$ then

mark the i th item as in the knapsack

$i = i-1, k = k-w_i$

else

$i = i-1$

The optimal knapsack should contain **{1, 2}**

Summary

- Dynamic programming is a useful technique of solving certain kind of problems
- When the solution can be recursively described in terms of partial solutions, we can store these partial solutions and re-use them as necessary(memoization)
- Running time of dynamic programming algorithm vs. naïve algorithm:
 - » 0-1 Knapsack problem: $O(W*n)$ vs. $O(2^n)$