Dynamic Programming 0/1 knapsack problem

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Formal description: Given two *n*-tuples of positive numbers

$$\langle v_1, v_2, \dots, v_n \rangle$$
 and $\langle w_1, w_2, \dots, w_n \rangle$,

and W > 0, we wish to determine the subset $T \subseteq \{1, 2, ..., n\}$ (of files to store) that

subject to
$$\sum_{i \in T} w_i \leq W$$
.

0/1 Knapsack Problem

- Optimization Problem
- Brute force: Try all 2ⁿ possible subsets T

Divide and Conquer Approach?

Divide and Conquer approach

- 1. Partition the problem into subproblems.
- 2. Solve the subproblems.
- 3. Combine the solutions to solve the original one.

Knapsack Problem:

Subproblems are not independent

Divide and-conquer algorithm repeatedly solves the common subsubproblems.

Thus, it does more work than necessary!

Better Solution: Dynamic Programming

Solution using Dynamic Programming

- Step 1: Decompose the problem into smaller problems
- Step 2: Recursively define the value of an optimal solution in terms of solutions to smaller problems

Initial Settings: Set

$$V[0, w] = 0$$
 for $0 \le w \le W$, no item $V[i, w] = -\infty$ for $w < 0$, illegal

Recursive Step: Use

$$\begin{split} V[i,w] &= \max(V[i-1,w], v_i + V[i-1,w-w_i]) \\ \text{for } 1 \leq i \leq n, \, 0 \leq w \leq W. \end{split}$$

• Step 3: Bottom-up computing V[i,w] (using iteration, not recursion).

Bottom: V[0, w] = 0 for all $0 \le w \le W$.

Bottom-up computation: Computing the table using

$$V[i,w] = \max(V[i-1,w],v_i+V[i-1,w-w_i])$$

row by row.

V[i,w]	w=0	1	2	3		 W]
i= 0	0	0	0	0	•••	 0	bottom
1	-	0 80 61				>	
2	-					>	
:						>	
n	-					>	

Example

i	1	2	3	4
$\overline{v_i}$	10	40	30	50
w_i	5	4	6	3

V[i, w]	0	1	2	3	4	5	6	7	8	9	10
i = 0											
i = 1											
i = 2											
i = 3											
i = 4											

Example

i	1	2	3	4
$\overline{v_i}$	10	40	30	50
w_i	5	4	6	3

Bottom: V[0, w] = 0 for all $0 \le w \le W$.

Bottom-up computation: Computing the table using

$$V[i,w] = \max(V[i-1,w],v_i+V[i-1,w-w_i])$$
 row by row.

V[i, w]	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0										
i = 2	0										
i = 3	0										
i = 4	0										

Example

i	1	2	3	4
$\overline{v_i}$	10	40	30	50
w_i	5	4	6	3

Bottom: V[0, w] = 0 for all $0 \le w \le W$.

Bottom-up computation: Computing the table using

$$V[i,w] = \max(V[i-1,w],v_i+V[i-1,w-w_i])$$
 row by row.

V[i, w]	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0	0	0	0	0	10	10	10	10	10	10
i = 2	0										
i = 3	0										
i = 4	0										

Example

i	1	2	3	4
$\overline{v_i}$	10	40	30	50
w_i	5	4	6	3

Bottom: V[0, w] = 0 for all $0 \le w \le W$.

Bottom-up computation: Computing the table using

$$V[i,w] = \max(V[i-1,w],v_i+V[i-1,w-w_i])$$
 row by row.

V[i, w]	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0	0	0	0	0	10	10	10	10	10	10
i = 2	0	0	0	0	40	40	40	40	40	50	50
i = 3											
i = 4											

Example

i	1	2	3	4
$\overline{v_i}$	10	40	30	50
w_i	5	4	6	3

Bottom: V[0, w] = 0 for all $0 \le w \le W$.

Bottom-up computation: Computing the table using

$$V[i,w] = \max(V[i-1,w],v_i+V[i-1,w-w_i])$$
 row by row.

V[i, w]	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0	0	0	0	0	10	10	10	10	10	10
i = 2	0	0	0	0	40	40	40	40	40	50	50
i = 3	0	0	0	0	40	40	40	40	40	50	70
i = 4	0										

Example

i	1	2	3	4
$\overline{v_i}$	10	40	30	50
w_i	5	4	6	3

Bottom: V[0, w] = 0 for all $0 \le w \le W$.

Bottom-up computation: Computing the table using

$$V[i,w] = \max(V[i-1,w],v_i+V[i-1,w-w_i])$$
 row by row.

V[i, w]	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0	0	0	0	0	10	10	10	10	10	10
i = 2	0	0	0	0	40	40	40	40	40	50	50
i = 3	0	0	0	0	40	40	40	40	40	50	70
i = 4	0	0	0	50	50	50	50	90	90	90	90

Optimal Subset

V[i, w]	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0	0	0	0	0	10	10	10	10	10	10
i = 2	0	0	0	0	40	40	40	40	40	50	50
i = 3	0	0	0	0	40	40	40	40	40	50	70
i = 4	0	0	0	50	50	50	50	90	90	90	90

incl	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0	0	0	0	0	1	1	1	1	1	1
i = 2	0	0	0	0	1	1	1	1	1	1	1
i = 3	0	0	0	0	0	0	0	0	0	0	1
i = 4	0	0	0	1	1	1	1	1	1	1	1

Optimal Subset

i	1	2	3	4
$\overline{v_i}$	10	40	30	50
w_i	5	4	6	3

incl	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0	0	0	0	0	1	1	1	1	1	1
i = 2	0	0	0	0	1	1	1	1	1	1	1
i = 3	0	0	0	0	0	0	0	0	0	0	1
i = 4	0	0	0	1	1	1	1	1	1	1	1

Optimal Subset

Let W = 10 and

i	1	2	3	4
$\overline{v_i}$	10	40	30	50
w_i	5	4	6	3

incl	0	1	2	3	4	5	6	7	8	9	10
i = 0	0	0	0	0	0	0	0	0	0	0	0
i = 1	0	0	0	0	0	1	1	1	1	1	1
i = 2	0	0	0	0	1	1	1	1	1	1	1
i = 3	0	0	0	0	0	0	0	0	0	0	1
i = 4	0	0	0	1	1	1	1	1	1	1	1

$$w = W$$
for $i = n$ to 1

If $incl[i, w] == 1$
 $include item i$
 $w = w - w_i$

$$w = 10$$

$$i = 3$$
; incl[3, 7] = 0

Profit V[n,W]

$$i = 1$$
; incl[1, 3] = 0

Optimal Subset { 2,4}

Algorithm

```
\mathsf{KnapSack}(v, w, n, W)
   for (w = 0 \text{ to } W) V[0, w] = 0;
   for (i = 1 \text{ to } n)
       for (w = 0 \text{ to } W)
           if ((w[i] \le w)) and (v[i] + V[i-1, w-w[i]] > V[i-1, w])
              V[i, w] = v[i] + V[i - 1, w - w[i]];
                incl[i, w] = 1;
           else
              V[i, w] = V[i - 1, w];
                incl[i, w] = 0;
   K = W:
   for (i = n \text{ downto } 1)
       if ( incl [i, K] == 1)
           output i;
          K = K - w[i];
   return V[n, W];
```

Exercise

1.

Item	Weight	value
1	2	12
2	1	10
3	3	20
4	2	15

$$W = 5$$

2.

Item	Weight	value
1	7	42
2	3	12
3	4	40
4	5	25

$$W = 10$$