

Lecture 5.

Knapsack problem

Algorithms and Data Structures
Ivan Solomatin
MIPT

Outline

- Knapsack problem
 - Problem statement
 - DP solution
 - Answer restoration (argmax)
 - Another modifications

Knapsack problem

- **Problem statement**
- DP solution
- Answer restoration (argmax)
- Another modifications

Knapsack problem

Problem statement

Problem: Ali Baba 3 (classical knapsack problem):

Ali Baba returned back to cave, and took his knapsack with carrying capacity W to take treasures out of the cave. When he came into cave he found a room with N gold bars in it. For each bar he knows weight and cost. As usual, he wants to maximize total cost of treasures taken out. The bars are undividable.

More formally. You have a knapsack with carrying capacity W and N items with weight w_i and cost c_i , you need to choose several items with indices i_0, \dots, i_{K-1} , such that:

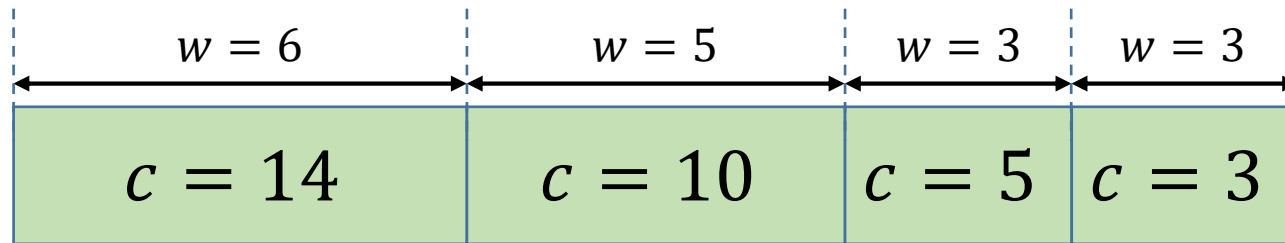
$$\sum_{k=0}^{K-1} w_{i_k} \leq W, \quad \sum_{k=0}^{K-1} c_{i_k} \rightarrow \max$$

This problem is NP-hard. In common case it cannot be solved faster than $O(2^N)$.
(of course, if $P \neq NP$)

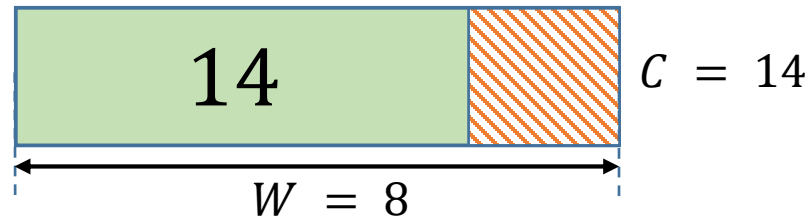
But if w_i are integer and W is rather small, we can solve it in $O(NW)$ operations.

Knapsack problem

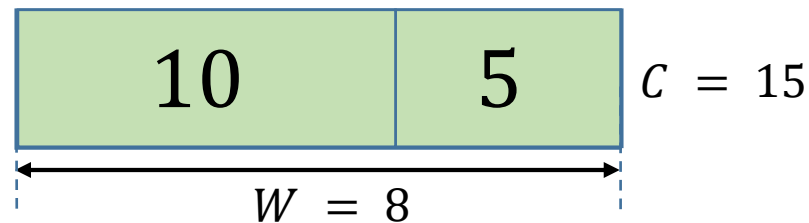
Problem statement



Greedy algorithm.
Items are sorted by $\frac{c}{w}$:



Optimal solution:



Knapsack problem

- Problem statement
- **DP solution**
- Answer restoration (argmax)
- Another modifications

Knapsack problem

DP solution

1. Subproblems:

$d[i][w]$ – maximum total cost we can get using first i items and knapsack with capacity w

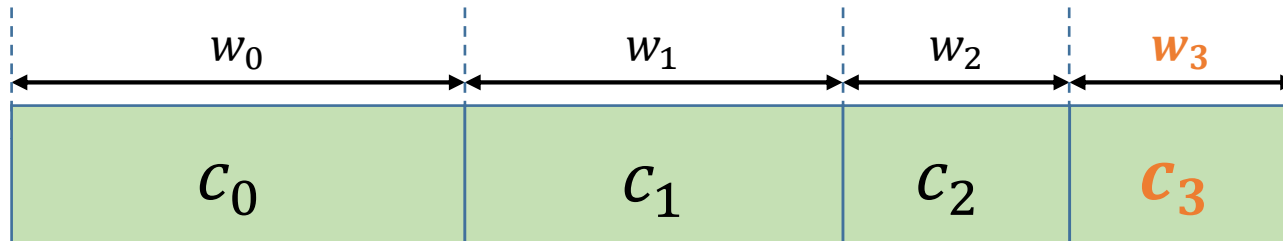
2. Basis:

$d[:, 0] = 0$ Let's suppose that all weights are positive.
Otherwise we can automatically take all items
with $w \leq 0$, and solve problem for the rest items

$d[0, :] = 0$ if we don't have items, we can't take any

Knapsack problem

DP solution



Let's assume that we have calculated all solutions for $i \leq 3$.

This means we know how we can optimally fill knapsacks with different capacities with first 3 items: $(0, 1, 2)$.

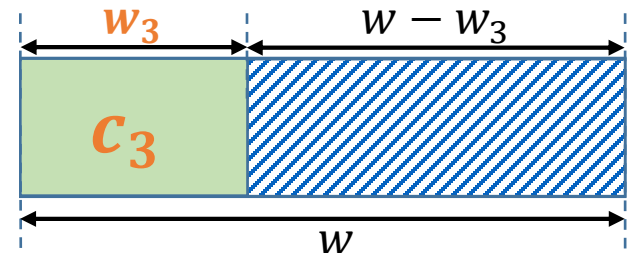
Now let's analyze how the solutions may change if we add one more item:

$$i = 4$$

To fill $d[4][:]$ we need to solve the subproblem for each capacity w .

Actually, for each capacity w , adding new item gives two possible cases:

- We ignore this new item. In this case, total cost is $d[3][w] = d[i-1][w]$.
- We place this new item into knapsack and fill free space with the rest of the items. Total cost in this case: $d[3][w - w_3] + c_3 = d[i-1][w - w_{i-1}] + c_{i-1}$



Knapsack problem

DP solution

1. Subproblems:

$d[i][w]$ – maximum total cost we can get using first i items and knapsack with capacity w

2. Basis:

$d[:, 0] = 0$ Let's suppose that all weights are positive.
Otherwise we can automatically take all items with $w \leq 0$, and solve problem for the rest items

$d[0, :] = 0$ if we don't have items, we can't take any

3. Inductive step:

$$d[i][w] = \max \begin{cases} d[i-1][w] \\ d[i-1][w - w_{i-1}] + c_{i-1}, & \text{if } w_{i-1} \leq w \end{cases}$$

4. Answer:

$$d[N][W]$$

Knapsack problem

DP solution

```
d = [[0] * (W + 1) for i in range(N + 1)]
for i in range(1, N + 1):
    for w in range(1, W + 1):
        if  $w_{i-1} > w$ :
            d[i][w] = d[i - 1][w]
        else:
            d[i][w] = max(d[i - 1][w],
                          d[i - 1][w -  $w_{i-1}$ ] +  $c_{i-1}$ )
```

	w_{i-1}	c_{i-1}
$i = 1$	3	3
$i = 2$	3	5
$i = 3$	5	10
$i = 4$	6	14

$i \backslash w$	0	1	2	3	4	5	6	7	8
$i = 0$	0	0	0	0	0	0	0	0	0
$i = 1$	0	0	0	3	3	3	3	3	3
$i = 2$	0	0	0	5	5	5	8	8	8
$i = 3$	0	0	0	5	5	10	10	10	15
$i = 4$	0	0	0	5	5	10	14	14	15

Diagram illustrating the DP table with arrows showing the recurrence relation:

- From $(i=1, w=3)$ to $(i=2, w=3)$ with label $+3$
- From $(i=2, w=3)$ to $(i=3, w=3)$ with label $+5$
- From $(i=3, w=3)$ to $(i=4, w=3)$ with label $+5$
- From $(i=1, w=3)$ to $(i=3, w=5)$ with label $+10$
- From $(i=2, w=3)$ to $(i=4, w=5)$ with label $+10$
- From $(i=3, w=5)$ to $(i=4, w=6)$ with label $+14$

Knapsack problem

- Problem statement
- DP solution
- **Answer restoration (argmax)**
- Another modifications

Knapsack problem

Answer restoration (argmax)

$i \backslash w$	0	1	2	3	4	5	6	7	8
$i = 0$	0	0	0	0	0	0	0	0	0
$i = 1$	0	0	0	3	3	3	3	3	3
$i = 2$	0	0	0	5	5	5	8	8	8
$i = 3$	0	0	0	5	5	10	10	10	15
$i = 4$	0	0	0	5	5	10	14	14	15

The diagram illustrates the restoration of the optimal solution by tracing back from the maximum value in the DP table. The path starts at the cell (i=4, w=8) with value 15, moves to (i=3, w=8) with value 10 (labeled +10), then to (i=2, w=3) with value 5 (labeled +5), and finally to (i=1, w=0) with value 0. The cells (i=1, w=0), (i=2, w=3), (i=3, w=8), and (i=4, w=8) are highlighted in orange to show the selected items.

Objects in optimal solution: 3, 2

Knapsack problem

- Problem statement
- DP solution
- Answer restoration (argmax)
- **Another modifications**

Knapsack problem

Another problem modifications

Problem 1. (Ali-Baba 1). Given N items with different costs and knapsack can contain up to W items. Maximize total cost.

$$w_i = 1$$

This problem can be solved using greedy algorithm.

Problem 2. Given N items with different weights and knapsack with capacity W . Maximize number of taken elements:

$$c_i = 1$$

This problem can be solved using greedy algorithm.

Problem 3. We can use each item (c_i, w_i) any number of times:

$$\begin{aligned} d[0] &= 0 \\ d[w] &= \max_{w_i \leq w} (c_i + d[w - w_i]) \end{aligned}$$

Knapsack problem

Another problem modifications

Problem 4. Does a given number W equal a sum of any subset of numbers w_0, w_1, \dots, w_{N-1} ?

We don't need costs here.

Subproblems:

$d[i][w]$ — does w equal a sum of any subset of w_0, \dots, w_{i-1}

Basis:

$$d[:][0] = \text{True}, \quad d[0][1:] = \text{False}$$

Inductive step:

$$d[i][w] = \begin{cases} d[i-1][w], & \text{if } w < w_{i-1} \\ d[i-1][w] \text{ OR } d[i-1][w - w_{i-1}], & \text{if } w \geq w_{i-1} \end{cases}$$

Answer:

$$d[N][W]$$

Knapsack problem

Another problem modifications

```
d = [[1] + [0] * (W) for i in range(N + 1)]
for i in range(1, N + 1):
    for w in range(1, W + 1):
        if  $w_{i-1} > w$ :
            d[i][w] = d[i - 1][w]
        else:
            d[i][w] = (d[i - 1][w] or
                       d[i - 1][w -  $w_{i-1}$ ])
```

	w_{i-1}
$i = 1$	3
$i = 2$	3
$i = 3$	5
$i = 4$	6

$i \backslash w$	0	1	2	3	4	5	6	7	8
$i = 0$	+	-	-	-	-	-	-	-	-
$i = 1$	+	-	-	+	-	-	-	-	-
$i = 2$	+	-	-	+	-	-	+	-	-
$i = 3$	+	-	-	+	-	+	+	-	+
$i = 4$	+	-	-	+	-	+	+	-	+

We can obtain: 0, 3, 5, 6, 8, ...

Knapsack problem

Another problem modifications

Problem 5. Given number W and numbers w_0, w_1, \dots, w_{N-1} . Find a closest to W number W^* , which is a sum of a subset of w_0, w_1, \dots, w_{N-1} .

Same as **Problem 4**:

Subproblems:

$d[i][w]$ – does w equal a sum of any subset of w_0, \dots, w_{i-1}

Basis:

$$d[:][0] = \text{True}, \quad d[0][1:] = \text{False}$$

Inductive step:

$$d[i][w] = \begin{cases} d[i-1][w], & \text{if } w < w_{i-1} \\ d[i-1][w] \text{ OR } d[i-1][w - w_{i-1}], & \text{if } w \geq w_{i-1} \end{cases}$$

Answer:

$$W^*: \quad d[N][W^*] = \text{True}, \quad |W - W^*| \rightarrow \min$$

Conclusion

Thank you for watching!