

# Knapsack Problem

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# The problem

There are  $N$  kinds of items and a knapsack with capacity  $V$ . The cost of putting the  $i$ -th kind is  $C_i$  and the weight of item is  $W_i$ . We want to determine which items to select to put into the knapsack in order to maximize total weight.

- 1 There is only one item in the  $i$ -th kind: 0/1 knapsack.
- 2 There are infinite number of items in the  $i$ -th kind: complete knapsack.
- 3 There are  $k_i$  ( $k_i \geq 1$ ) items in the  $i$ -th kind: multiple choice knapsack.

# 0/1 knapsack

$opt[i][j]$ : the maximum weight when we select from the first  $i$  items and the maximum weight is not beyond  $j$ .

$$opt[0][j] = 0, 0 \leq j \leq V,$$

$$opt[i][j] = \max(opt[i-1][j], opt[i-1][j - C_i] + W_i),$$

$$1 \leq i \leq N, C_i \leq j \leq V.$$

# 0/1 knapsack

The relationship is only between  $i$  and  $i - 1$ , hence:

$$\text{opt}[i\%2][j] = \max(\text{opt}[(i - 1)\%2][j], \text{opt}[(i - 1)\%2][j - C_i] + W_i).$$

However actually, we do not even need two lines.

If we enumerate  $j$  from  $V$  to  $C_i$ , we can reuse the status of  $i - 1$ , hence:

$$\text{opt}[j] = \max(\text{opt}[j], \text{opt}[j - C_i] + W_i),$$

with  $j$  counting from  $V$  to  $C_i$ .

# 0/1 Knapsack

There might be no way to reduce the overall level of time complexity. However, we can observe

$opt[N][V]$  depends on  $opt[N-1][\max(V - C_N, 0)]$ ,

$opt[N-1][\max(V - C_N, 0)]$  depends on

$opt[N-2][\max(V - C_N - C_{N-1}, 0)]$

...

$opt[i+1][\max(V - \sum_{k=i+2}^N C_k, 0)]$  depends on

$opt[i][\max(V - \sum_{k=i+1}^N C_k, 0)]$ ,

so we can choose the lower bound of the inner loop to be

$\max(V - \sum_{k=i+1}^N C_k, C_i)$ .

# Complete Knapsack

It is obvious we have

$$opt[i][j] = \max(opt[i-1][j - k * C_i] + k * W_i),$$

$$1 \leq i \leq N, k * C_i \leq j \leq V,$$

but is this really all?

# Complete Knapsack

We can do a math trick:

$$\begin{aligned} \text{opt}[i][j] &= \max(\text{opt}[i-1][j - k * C_i] + k * W_i) \\ &= \max(\text{opt}[i-1][j], \max(\text{opt}[i-1][j - C_i - (k-1) * C_i] \\ &\quad + (k-1) * W_i + W_i)) \\ &= \max(\text{opt}[i-1][j], \text{opt}[i][j - C_i] + W_i), \end{aligned}$$

and  $k$  disappeared! Moreover, since this time we want to reuse the status of  $i$ , we have

$$\text{opt}[j] = \max(\text{opt}[j], \text{opt}[j - C_i] + W_i),$$

with  $j$  counting from  $C_i$  to  $V$ . The opposite of 0/1 Knapsack!

# Multiple Choice Knapsack

What about multiple choice version?

$$opt[i][j] = \max(opt[i-1][j - k * C_i] + k * W_i),$$

$$1 \leq i \leq N, k * C_i \leq j \leq V, 0 \leq k \leq k_i.$$

There is no math trick like the case of complete knapsack since we do not know the range of every  $k_i$ .

One step back, this can also be treated as 0/1 knapsack if we do it one by one. Can we optimize based on this?



# Multiple Choice Knapsack

The key is to decompose  $k_i$  while at the same time maintain the ability to represent 0 to  $k_i$ . A common technique is using binary/bit.

Let  $n$  be the largest integer such that

$$\sum_{m=0}^n 2^m < k_i,$$

and we have

$$1, 2, 4, \dots, 2^n, k_i - 2^{n+1} + 1$$

to be able to represent 0 to  $k_i$  (Apparently this is true for numbers between 0 and  $2^{n+1} - 1$ , for number  $x$  between  $2^{n+1}$  and  $k_i$ , just consider  $k_i - x$ ). And we only need to consider  $O(\log k_i)$  items.

# Multiple Choice Knapsack

No. The former method is not the end :-)

There is a crazy trick from an internal training material when I studied ACM/ICPC, which is of complexity  $O(NV)$ :

Let  $j = q * C_i + r$ , then

$$\begin{aligned} opt[i][j] &= \max(opt[i-1][(q-k) * C_i + r] + k * W_i), \\ &= \max(opt[i-1][u * C_i + r] - u * W_i) + q * W_i, \end{aligned}$$

$$q - k_i \leq u \leq q,$$

then for every  $i$ , classify 0 to  $V$  by  $r$  into  $C_i$  groups, and compute the maximum within the same group with length restriction to be  $k_i$ , which is the sliding window maximum problem and thus can be solved in linear time.