

CMPS 6610 Algorithms – Fall 2018

Greedy Algorithms: Knapsack Problem

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Greedy Strategy

1. Repeatedly identify a decision to be made (→ recursion)
2. Make a **locally optimal** choice for each decision

In order to reach a globally optimal solution, the problem must have appropriate recursive substructure:

optimal solution = locally optimal choice
+ optimal solution for the remainder
of the problem

Knapsack Problem

- Given a knapsack with weight capacity $W > 0$, and given n items of positive integer weights w_1, \dots, w_n and positive integer values v_1, \dots, v_n .
(So, item i has value v_i and weight w_i .)
- **0-1 Knapsack Problem:** Compute a subset of items that maximize the total value (sum), and they all fit into the knapsack (total weight at most W).
- **Fractional Knapsack Problem:** Same as before but we are allowed to take fractions of items (\rightarrow gold dust).

Greedy Knapsack

- Greedy Strategy:
 - Compute $\frac{v_i}{w_i}$ for each i
 - Greedily take as much as possible of the item with the highest value/weight. Then repeat/recurse.
- ⇒ Sort items by value/weight
- ⇒ $O(n \log n)$ runtime

Knapsack Example

item	1	2	3	
value	12	15	4	W=4
weight	2	3	1	
value/weight	6	5	4	

- **Greedy fractional:** Take item 1 and $2/3$ of item 2
 \Rightarrow weight=4, value= $12+2/3 \cdot 15 = 12+10 = 22$
- **Greedy 0-1:** Take item 1 and then item 3
 \Rightarrow weight = $1+2=3$, value= $12+4=16$
- **Optimal 0-1:** Take items 2 and 3, value =19

greedy 0-1
 \neq optimal 0-1

Optimal Substructure

- Let s_1, \dots, s_n be an optimal solution, where $s_i =$ amount of item i that is taken; $0 \leq s_i \leq 1$
- Suppose we remove one item. $\rightarrow n - 1$ items left
- Is the remaining “solution” still an optimal solution for $n - 1$ items?
- Yes; cut-and-paste.

Correctness Proof for Greedy

- Suppose items $1, \dots, n$ are numbered in decreasing order by value/weight.
 - Greedy solution G : Takes all elements $1, \dots, j, \dots, i^*-1$ and a fraction of i^* .
 - Assume optimal solution S is different from G . Assume S takes only a fraction $\frac{1}{a}$ of item j , for $j \leq i^*-1$.
 - Create new solution S' from S by taking $w_j - 1/a$ weight away from items $> j$, and add $w_j - 1/a$ of item j back in. Hence, all of item j is taken.
- \Rightarrow New solution S' has the same weight but increased value. This contradicts the assumption that S was optimal.
- $\Rightarrow S=G$. □

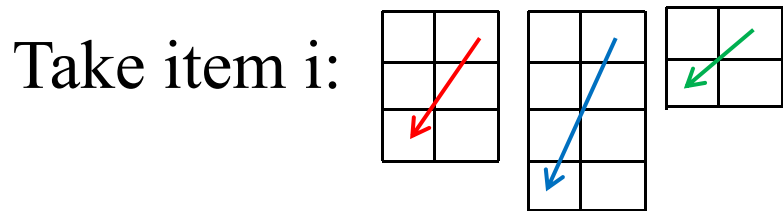
General Solution: DP

- $D[i, w]$ = max value possible for taking a subset of items $1, \dots, i$ with knapsack constraint w .
- $D[0, w] = D[i, 0] = 0$ for all $0 \leq i \leq n$ and $0 \leq w \leq W$
 $D[i, w] = -\infty$ for $w < 0$
 $D[i, w] = \max(\underbrace{D[i - 1, w]}_{\text{don't take item } i}, \underbrace{v_i + D[i - 1, w - w_i]}_{\text{take item } i})$
- Compute $D[n, W]$ by filling an $n \times W$ DP-table.
 \Rightarrow Two nested for-loops, runtime and space $\Theta(nW)$
- Trace back from $D[n, W]$ by redoing computation or following arrows. $\Rightarrow \Theta(n + W)$ runtime

DP Example

W=4

item	1	2	3
value	12	15	4
weight	2	3	1
value/weight	6	5	4



				w
				↑
W=4	0	12	15	19
3	0	12	15	16
2	0	12	12	12
1	0	0	0	4
0	0	0	0	0
	0	1	2	3
				n → i

Solution:
Take items 3 and 2

$$D[i, w] = \max(\underbrace{D[i - 1, w]}_{\text{don't take item } i}, \underbrace{v_i + D[i - 1, w - w_i]}_{\text{take item } i})$$