

# Lecture Notes

on

## Greedy Algorithms: Knapsack Problem



July 2020  
**(Be safe and stay at home)**



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a *globally* optimal solution can be **arrived at** by making a *locally* optimal (greedy) choice.
  - ▶ **The optimal substructure property**  
an optimal solution to the problem **contains** within it optimal solution to subprograms.
- ▶ Greedy algorithms do not always yield optimal solutions, but for many problems they do.

## 0-1 knapsack problem

### Problem statement:

- ▶ Given  $n$  items  $\{1, 2, \dots, n\}$
- ▶ Item  $i$  is worth  $v_i$ , and weight  $w_i$
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### Example:

- ▶ Given

$i$	$v_i$	$w_i$	$v_i/w_i$
1	6	1	6
2	10	2	5
3	12	3	4

Total weight  $W = 5$

## 0-1 knapsack problem

Problem statement, *mathematically* – version 1:

Find a subset  $S \subseteq \{1, 2, \dots, n\}$  such that

$$\begin{aligned} & \text{maximize} && \sum_{i \in S} v_i \\ & \text{subject to} && \sum_{i \in S} w_i \leq W \end{aligned}$$

## 0-1 knapsack problem

Problem statement, *mathematically* – version 2:

Let  $x = (x_1, x_2, \dots, x_n)$ , and

$$x_i = \begin{cases} 1 & i\text{-th item is in the knapsack} \\ 0 & i\text{-th item is not in the knapsack} \end{cases}$$

Then the knapsack problem is

$$\begin{aligned} & \text{maximize} && \sum_{i=1}^n v_i x_i \\ & \text{subject to} && x_i \in \{0, 1\} \\ & && \sum_{i=1}^n w_i x_i \leq W \end{aligned}$$

# 0-1 knapsack problem

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- ▶  $2^n$  feasible solutions
- ▶ Total cost =  $O(n \cdot 2^n)$

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Three possible **greedy** strategies:

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Three possible **greedy** strategies:

1. Greedy by highest value  $v_i$
2. Greedy by least weight  $w_i$
3. Greedy by largest value density  $\frac{v_i}{w_i}$

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### Example

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Greedy by value density  $v_i/w_i$ :

- ▶ take items 1 and 2.
- ▶ value = 16, weight = 3
- ▶ Leftover capacity = 2

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### Example

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#### Optimal solution

- ▶ take items 2 and 3.
- ▶ value = 22, weight = 5
- ▶ no leftover capacity

## 0-1 knapsack problem

### Example

$i$	$v_i$	$w_i$	$v_i/w_i$
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Total weight  $W = 5$

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Optimal solution

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Question: how about greedy by highest value? by least weight?

## 0-1 knapsack problem

### Another example

Given the following six items with  $W = 100$ :

$i$	$v_i$	$w_i$	$v_i/w_i$	Greedy by			optimal solution
				value	weight	$v_i/w_i$	
1	40	100	0.4	1	0	0	0
2	35	50	0.7	0	0	1	<b>1</b>
3	18	45	0.4	0	1	0	<b>1</b>
4	4	20	0.2	0	1	1	0
5	10	10	1	0	1	1	0
6	2	5	0.4	0	1	1	<b>1</b>
Total value				40	34	51	<b>55</b>
Total weight				100	80	85	<b>100</b>

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Given the following six items with  $W = 100$ :

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2	35	50	0.7	0	0	1	<b>1</b>
3	18	45	0.4	0	1	0	<b>1</b>
4	4	20	0.2	0	1	1	0
5	10	10	1	0	1	1	0
6	2	5	0.4	0	1	1	<b>1</b>
Total value				40	34	51	<b>55</b>
Total weight				100	80	85	<b>100</b>

*All three greedy approaches generate feasible solutions, but none of them generate the optimal solution. Greedy algorithms doesn't work for the 0-1 knapsack problem!*



Q & A?

Queries are welcome on slack channel  
for discussion