

# Greedy Algorithms: The Fractional Knapsack

Version of September 17, 2016



## Outline

- Introduction
- The Knapsack problem.
- A greedy algorithm for the fractional knapsack problem
- Correctness

# Introduction to Greedy Algorithm

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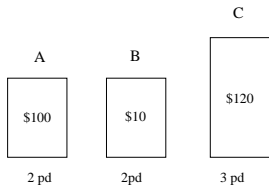
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# Introduction to Greedy Algorithm

- A **greedy algorithm** for an optimization problem always makes the choice that **looks best at the moment** and adds it to the current subsolution.
- Final output is an optimal solution.
- Greedy algorithms don't always yield optimal solutions but, when they do, they're usually the simplest and most efficient algorithms available.

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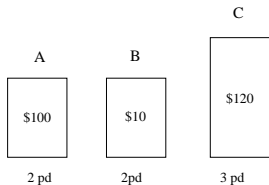
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Capacity of knapsack:  $K = 4$



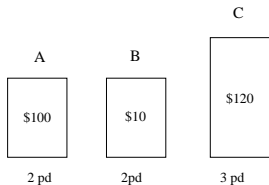
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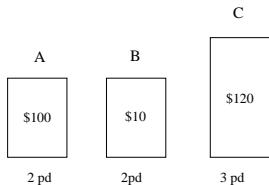
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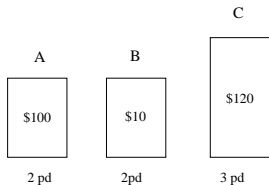
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2 pd A \$100	2 pd C \$80
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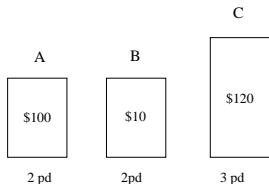
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Can only **take or leave** item. You  
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3 pd C \$120	
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# The Fractional Knapsack Problem: Formal Definition

- Given  $K$  and a set of  $n$  items:

weight	$w_1$	$w_2$	$\dots$	$w_n$
value	$v_1$	$v_2$	$\dots$	$v_n$

- Find:  $0 \leq x_i \leq 1, i = 1, 2, \dots, n$  such that

$$\sum_{i=1}^n x_i w_i \leq K$$

and the following is maximized:

$$\sum_{i=1}^n x_i v_i$$

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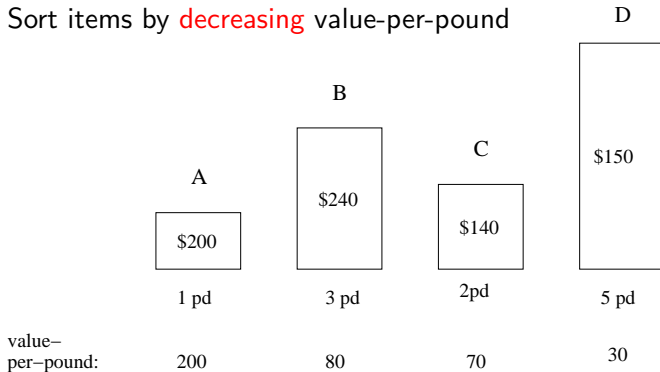
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Sort items by **decreasing** value-per-pound



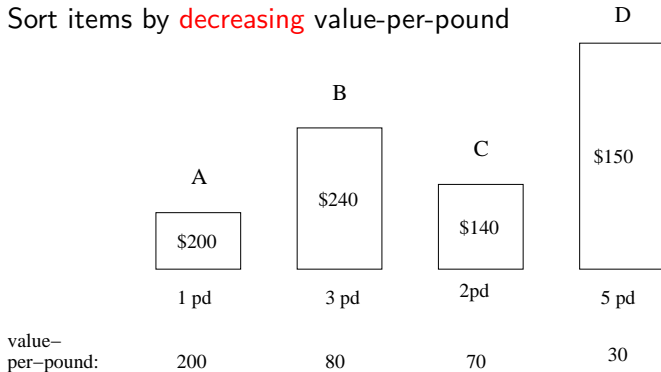
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If knapsack holds  $K = 5$  pd, solution is:

1	pd	A
3	pd	B
1	pd	C

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Running time:  $O(n \log n)$ .

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- We claim that the total value for this set of items is the **optimal** value.

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## Correctness...

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**Therefore  $G$  is also optimal!**

# Greedy solution for 0-1 Knapsack Problem?

The 0-1 Knapsack Problem does **not** have a greedy solution!

## Example




A		B		C	
					
\$300		\$190		\$180	
3 pd		2pd		2 pd	
value–					
per–pound:	100	95		90	

$K = 4$ . Solution is

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## Question

Suppose we tried to prove the greedy algorithm for 0-1 knapsack problem **does** construct an optimal solution.

# Greedy solution for 0-1 Knapsack Problem?

The 0-1 Knapsack Problem does **not** have a greedy solution!

## Example

A			B			C		
								
3 pd			2pd			2 pd		
value- per-pound: 100			95			90		

$K = 4$ . Solution is item B + item C

## Question

Suppose we tried to prove the greedy algorithm for 0-1 knapsack problem **does** construct an optimal solution. If we follow exactly the same argument as in the fractional knapsack problem where does the proof fail?