

# Fractional Knapsack

## Greedy Algorithms

# Outline

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

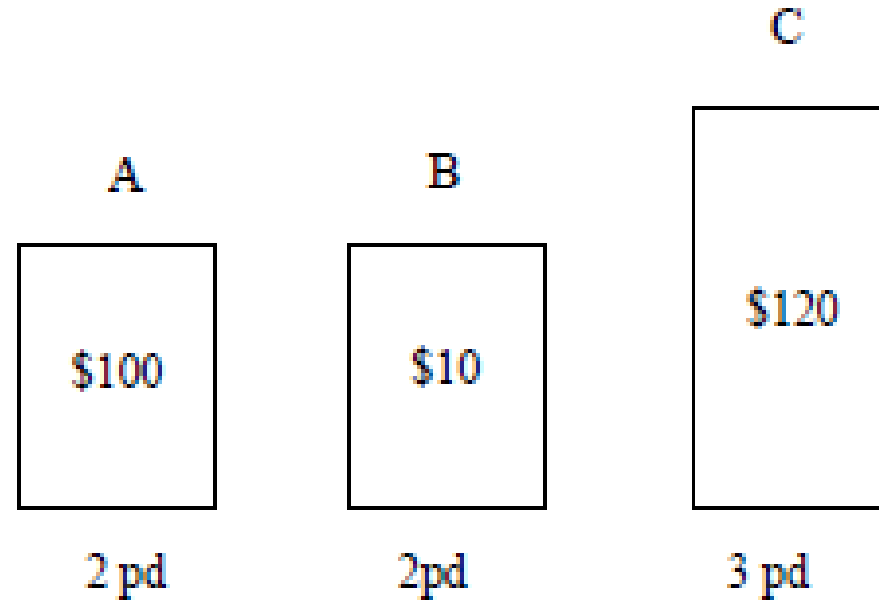
# Introduction to Greedy Algorithm

- A **greedy algorithm** for an optimization problem always makes the choice that **looks best at the moment** and adds it to the current sub solution.
- 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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# The Knapsack Problem



Capacity of knapsack:  $K = 4$

~~\* Sort by value~~  
~~\* Sort by weight~~

value per weight

$\frac{60}{10} = 6$   
 A ✓  

10pd

 \$60

$\frac{100}{20} = 5$   
 B ✓  

20pd

 \$100

$\frac{120}{30} = 4$   
 C ✓  

30pd

 \$120

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knapsack

B 20d \$100

C 30pd \$120

Total Value  
= \$240

$\frac{2 \times 120}{3} = 80$

A 10p

B 20pd

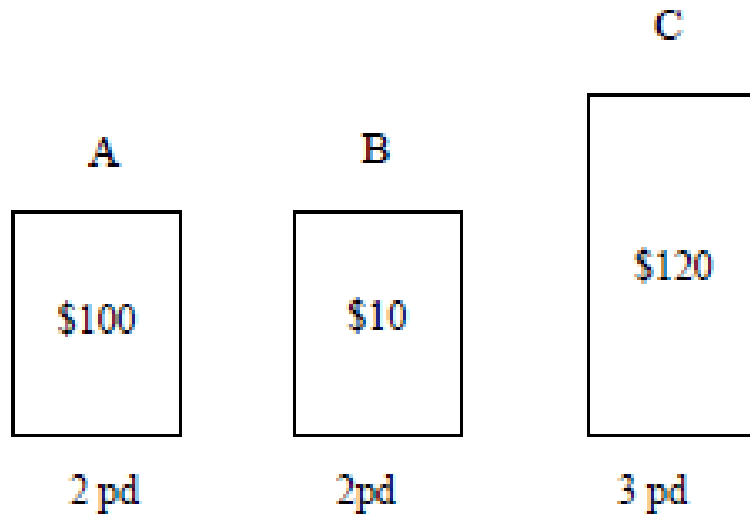
C 20pd

$\frac{2}{3} \times 30 = 20$

K = 50pd

Total Value = \$220

# The Knapsack Problem



Capacity of knapsack:  $K = 4$

**Fractional** Knapsack Problem:  
Can take a **fraction** of an item.

Solution:

2 pd A \$100	2 pd C \$80
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Solution:

3 pd C \$120	
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**0-1** Knapsack Problem:  
Can only **take or leave** item. You  
can't take a fraction.

# 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:

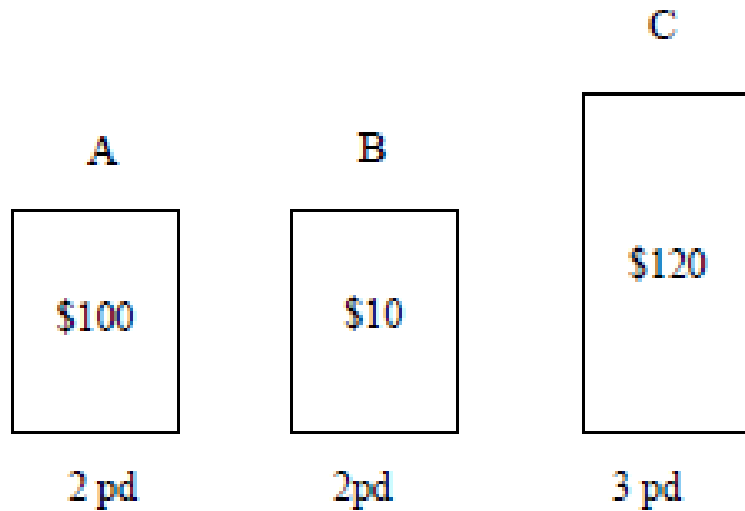
$$\sum_{i=1}^n x_i v_i$$



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# Greedy Solution for Fractional Knapsack



Capacity of knapsack:  $K = 4$

Solution 1: Sort by value,  
select the item with maximum value  
first

Solution 1 is not correct  
Counter example

Solution 2: Sort by weight,  
select the item with minimum weight  
first

Solution 2 is not correct  
Counter example

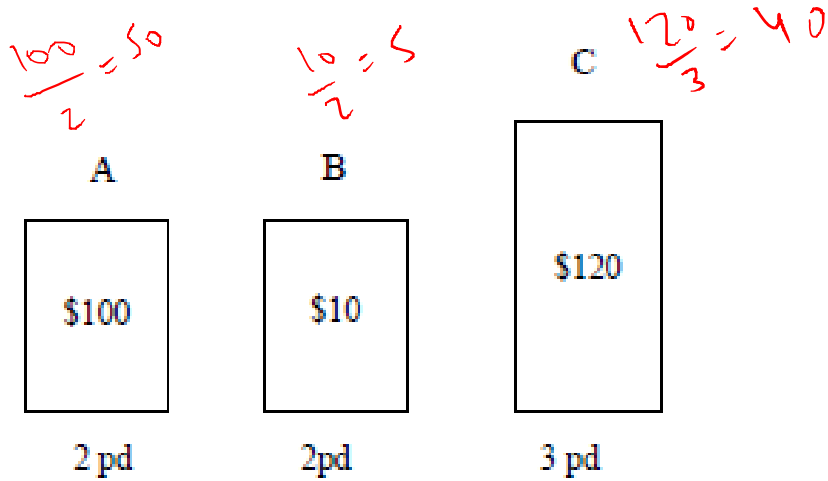
C \$120 3 pd	A \$50 1 pd
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Total value = \$170

A \$100 2 pd	B \$10 2pd
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Total value = \$110

# Greedy Solution for Fractional Knapsack



Capacity of knapsack:  $K = 4$

Solution 3: Sort by value per pound, select the item with maximum value per pound first

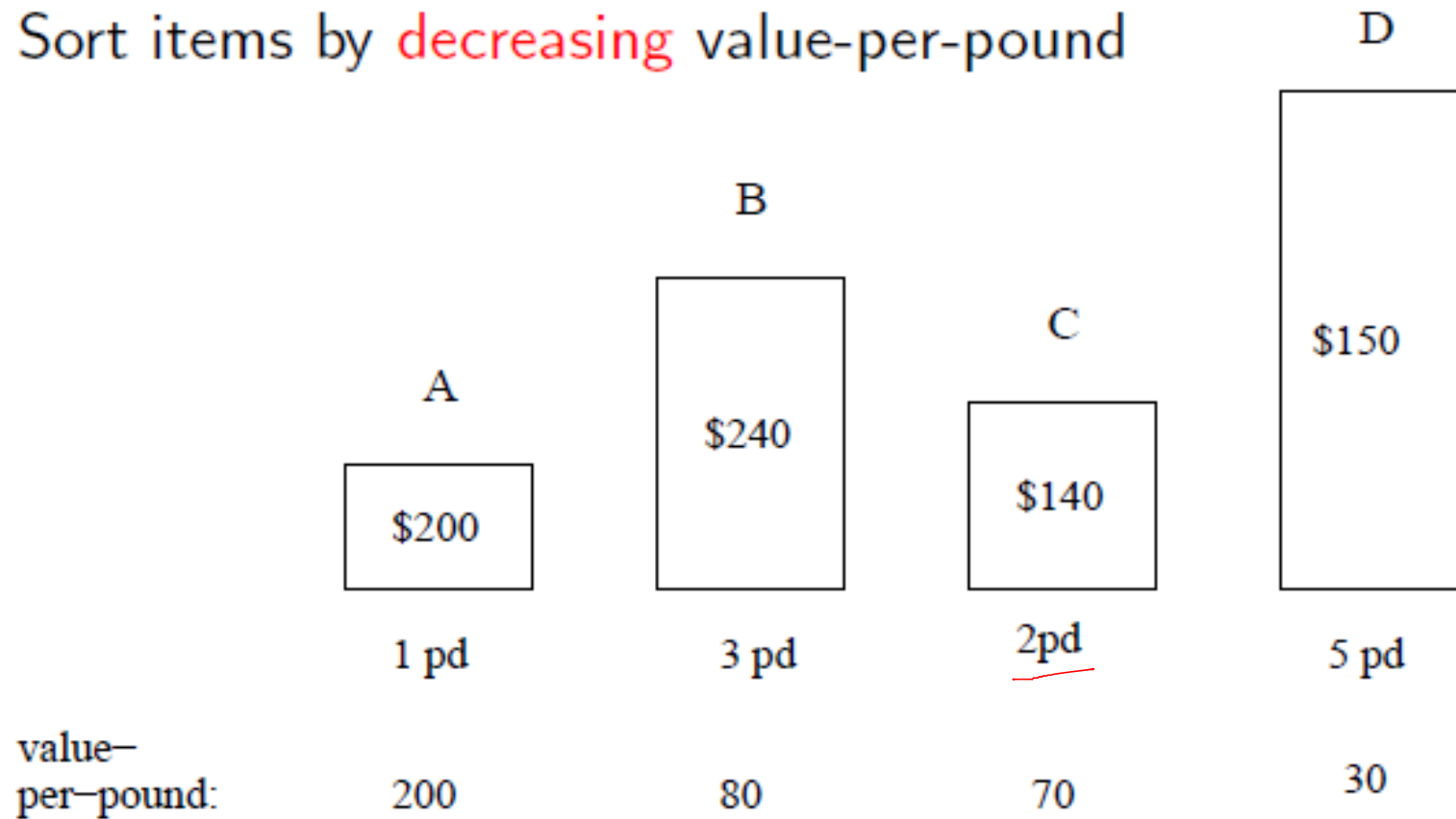
Solution is correct example

A	C
\$100	\$80
2 pd	2 pd

Total value = \$180

# Greedy Solution for Fractional Knapsack

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



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

1	pd	A
3	pd	B
1	pd	C

$$\frac{K \times (w_i)}{w_i} = K$$

$$\frac{K}{w_i}$$

$$\frac{1}{2} = 0.5$$

①

# Greedy Solution for Fractional Knapsack

- Calculate the value-per-pound  $\rho_i = \frac{v_i}{w_i}$  for  $i = 1, 2, \dots, n$ . }  $\Theta(n)$
- Sort the items by decreasing  $\rho_i$ .  
Let the sorted item sequence be  $1, 2, \dots, i, \dots, n$ , and the corresponding value-per-pound and weight be  $\rho_i$  and  $w_i$  respectively. }  $\Theta(n \lg n)$
- Let  $k$  be the current weight limit (Initially,  $k = K$ ).  
In each iteration, we choose item  $i$  from the head of the unselected list.
  - ✓ • If  $k \geq w_i$ , set  $x_i = 1$  (we take item  $i$ ), and reduce  $k = k - w_i$ , then consider the next unselected item. }  $\Theta(n)$
  - If  $k < w_i$ , set  $x_i = k/w_i$  (we take a fraction  $k/w_i$  of item  $i$ ). Then the algorithm terminates. }  $\Theta(n \lg n)$

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

$$n + n \lg n + n = \Theta(n \lg n)$$

# Greedy Solution for Fractional Knapsack

- Observe that the algorithm may take a fraction of an item. This can **only** be the **last** selected item.
- We claim that the total value for this set of items is the **optimal** value.

$$\begin{array}{r} \checkmark \\ 0.25, 1, 0.75 \\ \hline \checkmark \\ 1, 1, 0.25 \\ \hline \end{array}$$

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# Correctness

Given a set of  $n$  items  $\{1, 2, \dots, n\}$ .

- Assume items sorted by per-pound values:  $\rho_1 \geq \rho_2 \geq \dots \geq \rho_n$ .

Let the greedy solution be  $G = \langle x_1, x_2, \dots, x_k \rangle$ .

- $x_i$  indicates fraction of item  $i$  taken (all  $x_i = 1$ , except possibly for  $i = k$ ).

Consider any optimal solution  $O = \langle y_1, y_2, \dots, y_n \rangle$ .

- $y_i$  indicates fraction of item  $i$  taken in  $O$  (for all  $i$ ,  $0 \leq y_i \leq 1$ ).
- Knapsack must be full in both  $G$  and  $O$ :

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

Consider the first item  $i$  where the two selections differ.

- By definition, solution  $G$  takes a greater amount of item  $i$  than solution  $O$  (because the greedy solution always takes as much as it can). Let  $x = x_i - y_i$ .



Ordered by value per weight  
 $\begin{matrix} 1 & 2 & 3 & \dots & i & \dots & n \end{matrix}$

$$G = \langle \underbrace{x_1 \ x_2 \ x_3 \ \dots}_{\text{gold}} \underbrace{x_i \ \dots \ x_n}_{\text{silver}} \rangle \checkmark$$

$$0 \leq x_i \leq 1$$

$$O = \langle y_1 \ y_2 \ y_3 \ \dots \ y_i \ \dots \ y_n \rangle \checkmark$$

$$0 \leq y_i \leq 1$$

(k)

$$x_i \neq y_i$$

gold - silver

$$\boxed{x_i > y_i}$$

$$\begin{array}{r} 1,1(0.85),000 \text{ from} \\ \text{subtract } (x_i - y_i)w_i \text{ from} \\ \hline y_{i+1} \dots y_n \end{array}$$

(k)

$$O' = \langle y_1 \ y_2 \ y_3 \ \dots \ \underbrace{(x_i)}_{\text{gold}} \underbrace{(\dots)}_{\text{silver}} \dots \rangle$$

Total Value

Total Value

Total value of  $O'$  of  $O$  ✓

Total weight of  $O' =$  Total weight of  $O$

$y_i \neq x_i$   
 $x_i = 100$   
 (25)

# Correctness

Consider the following new solution  $O'$  constructed from  $O$ :

- For  $j < i$ , keep  $y'_j = y_j$ .
- Set  $y'_i = x_i$ .
- In  $O$ , remove items of total weight  $x_i w_i$  from items  $(i+1 \text{ to } n)$ , resetting the  $y'_j$  appropriately.

$$x = (x_i - y_i) w_i$$

This is always doable because  $\sum_{j=i}^n x_j = \sum_{j=i}^n y_j$

- The total value of solution  $O'$  is greater than or equal to the total value of solution  $O$  (why?)
- Since  $O$  is largest possible solution and value of  $O'$  cannot be smaller than that of  $O$ ,  $O$  and  $O'$  must be equal.
- Thus solution  $O'$  is also optimal.

By repeating this process, we will eventually convert  $(O)$  into  $(G)$ , without changing the total value of the selection. **Therefore  $G$  is also optimal!**