# COMP108 Data Structures and Algorithms

# Greedy Algorithm (Part I Knapsack Problem)

Professor Prudence Wong

pwong@liverpool.ac.uk

2022-23

#### **Outline**

#### Greedy algorithms

- What is greedy algorithm?
- See some examples

#### Learning outcomes:

- Understand what greedy algorithm is
- Able to apply greedy algorithm on the Knapsack problem
- Able to apply Kruskal's algorithm to find minimum spanning tree
- ▶ Able to apply Dijkstra's algorithm to find single-source shortest-paths

#### **Coin Change Problem**

Suppose we have 3 types of coins







Minimum number of coins to make \$0.8, \$1.0, \$1.4?

Greedy Algorithm

#### **Coin Change Problem**

#### What if we have different types of coins?







Minimum number of coins to make \$0.8, \$1.0, \$1.4?

Greedy Algorithm still works?

#### **Greedy Algorithms**

## How to be greedy?

- At every step, make the best move you can make
- Keep going until you're done

#### **Greedy Algorithms**

#### How to be greedy?

- At every step, make the best move you can make
- Keep going until you're done

## **Advantages**

- Don't need to pay much effort at each step
- Usually finds a solution very quickly
- The solution found is usually not bad

#### **Greedy Algorithms**

#### How to be greedy?

- At every step, make the best move you can make
- Keep going until you're done

# **Advantages**

- Don't need to pay much effort at each step
- Usually finds a solution very quickly
- The solution found is usually not bad

#### Possible problem

▶ The solution found may NOT be the best one

#### **Greedy Algorithms - examples**

Greedy algorithms that do NOT find the BEST solution

Coin change problem (previous slide)

Knapsack problem

Greedy algorithms that find the BEST solution

Minimum spanning tree

Kruskal's algorithm

Single-source shortest-paths

Dijkstra's algorithm

# Knapscak Problem

#### **Knapscak Problem**



#### **Knapscak Problem**

#### orthogonal

- Input: Given n items with weights  $w_1, w_2, \cdots, w_n$  and values  $v_1, v_2, \cdots, v_n$ , and a knapsack with capacity W
- Output: Find the most valuable subset of items that can fit into the knapsack
- Applications: A transport plane is to deliver the most valuable set of items to a remote location without exceeding its capacity

# Knapsack problem - Example 1

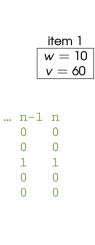
item 3

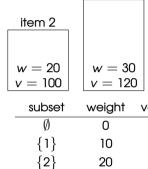
item 2

item 1 w = 10 v = 60 w = 20 v = 100 w = 30 v = 120

Knapsack
capacity
= 50

# Knapsack problem - Example 1





{3}

{1.3}

item 3

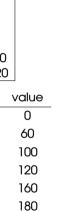
30

30

40

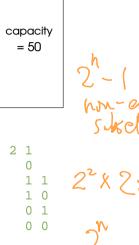
**50** 

60



220

N/A



Knapsack

COMP108-10-Greedy-01

# Knapsack problem - Example 2

item 1

item 2

 item 3

v = 40

item 4

v=25

Knapsack

capacity = 10

{1,4}

12

N/A

{1,2,3,4}

19

N/A

#### **Exhaustive algorithm**

#### Approach

- ▶ Try every subset of the set of n given items
- Compute total weight of each subset
- Compute total value of those subsets that do NOT exceed knapsack's capacity
- Pick the subset that has the largest value

#### **Exhaustive algorithm**

#### Approach

- Try every subset of the set of n given items
- Compute total weight of each subset
- Compute total value of those subsets that do NOT exceed knapsack's capacity
- Pick the subset that has the largest value

#### **Analysis**

How many subsets to consider?

#### **Exhaustive algorithm**

#### Approach

- Try every subset of the set of n given items
- Compute total weight of each subset
- Compute total value of those subsets that do NOT exceed knapsack's capacity
- Pick the subset that has the largest value

#### **Analysis**

- How many subsets to consider?
- $\triangleright 2^n 1$
- Exponential in n
- ► Why?

item 3

item 2 w = 10 v = 60 w = 20 v = 100 w = 30 v = 120

Knapsack

capacity
= 50

Knapsack
capacity
= 50

**Greedy:** pick the item with the next largest value if total weight  $\leq$  capacity **Result:** 

item 3 is picked, total value = 120, total weight = 30

item 2

item 2 w = 10 v = 60 w = 20 v = 100 w = 30 v = 120

Knapsack

capacity
= 50

- item 3 is picked, total value = 120, total weight = 30
- item 2 is picked, total value = 220, total weight = 50

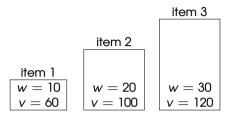
item 2

item 2  $w = 10 \\
v = 60$   $w = 20 \\
v = 100$   $w = 30 \\
v = 120$ 

Knapsack

capacity
= 50

- item 3 is picked, total value = 120, total weight = 30
- item 2 is picked, total value = 220, total weight = 50
- item 1 cannot be picked



Knapsack
capacity
= 50

- item 3 is picked, total value = 120, total weight = 30
- item 2 is picked, total value = 220, total weight = 50
- item 1 cannot be picked



item 3

item 2

item 2 w = 10 v = 60 w = 20 v = 100 w = 30 v = 120

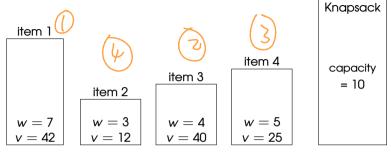
Knapsack
capacity
= 50

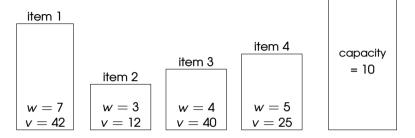
**Greedy:** pick the item with the next largest value if total weight  $\leq$  capacity

- Result:
  - item 3 is picked, total value = 120, total weight = 30
  - item 2 is picked, total value = 220, total weight = 50
  - item 1 cannot be picked



Time complexity? Always v

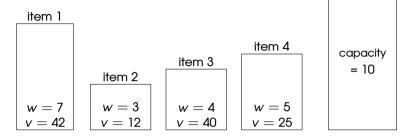




Knapsack

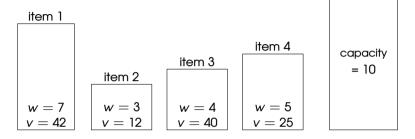
**Greedy:** pick the item with the next largest value if total weight  $\leq$  capacity **Result:** 

item 1 is picked, total value = 42, total weight = 7



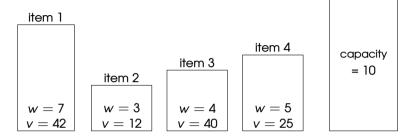
Knapsack

- item 1 is picked, total value = 42, total weight = 7
- item 3 cannot be picked



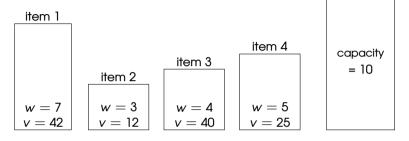
Knapsack

- item 1 is picked, total value = 42, total weight = 7
- item 3 cannot be picked
- item 4 cannot be picked



Knapsack

- item 1 is picked, total value = 42, total weight = 7
- item 3 cannot be picked
- item 4 cannot be picked
- item 2 is picked, total value = 54, total weight = 10

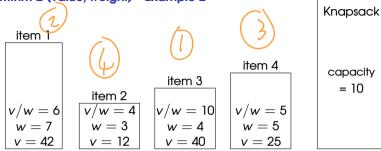


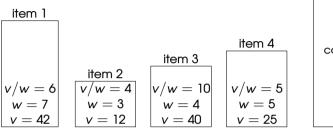
**Greedy:** pick the item with the next largest value if total weight  $\leq$  capacity **Result:** 

- item 1 is picked, total value = 42, total weight = 7
- item 3 cannot be picked
- item 4 cannot be picked
- ▶ item 2 is picked, total value = 54, total weight = 10



Knapsack



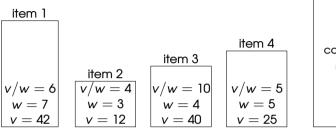


Knapsack

capacity
= 10

**Greedy:** pick the item with the next largest (value/weight) if total weight  $\leq$  capacity **Result:** 

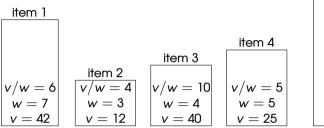
item 3 is picked, total value = 40, total weight = 4



capacity = 10

Knapsack

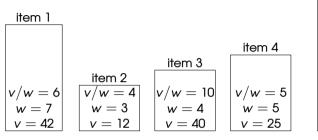
- item 3 is picked, total value = 40, total weight = 4
- item 1 cannot be picked



Knapsack

capacity
= 10

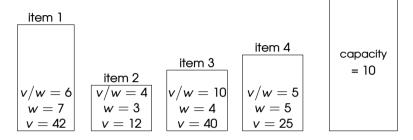
- item 3 is picked, total value = 40, total weight = 4
- item 1 cannot be picked
- item 4 is picked, total value = 65, total weight = 9



Knapsack

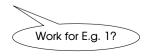
capacity
= 10

- item 3 is picked, total value = 40, total weight = 4
- item 1 cannot be picked
- item 4 is picked, total value = 65, total weight = 9
- item 2 cannot be picked

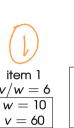


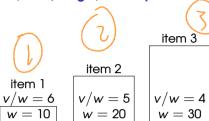
**Greedy:** pick the item with the next largest (value/weight) if total weight  $\leq$  capacity **Result:** 

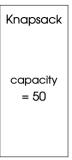
- item 3 is picked, total value = 40, total weight = 4
- item 1 cannot be picked
- item 4 is picked, total value = 65, total weight = 9
- item 2 cannot be picked



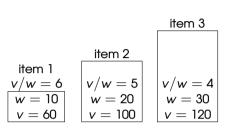
Knapsack







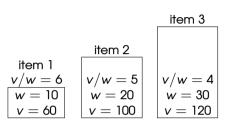
Greedy: pick the item with the next largest (value/weight) if total weight < capacity **Result:** 



Knapsack
capacity
= 50

**Greedy:** pick the item with the next largest (value/weight) if total weight  $\leq$  capacity **Result:** 

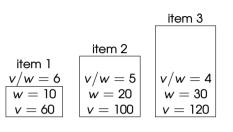
item 1 is picked, total value = 60, total weight = 10



Knapsack
capacity
= 50

**Greedy:** pick the item with the next largest (value/weight) if total weight  $\leq$  capacity **Result:** 

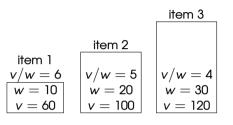
- item 1 is picked, total value = 60, total weight = 10
- item 2 is picked, total value = 160, total weight = 30



Knapsack
capacity
= 50

**Greedy:** pick the item with the next largest (value/weight) if total weight  $\leq$  capacity **Result:** 

- item 1 is picked, total value = 60, total weight = 10
- item 2 is picked, total value = 160, total weight = 30
- item 3 cannot be picked



Knapsack
capacity
= 50

**Greedy:** pick the item with the next largest (value/weight) if total weight  $\leq$  capacity **Result:** 

- item 1 is picked, total value = 60, total weight = 10
- item 2 is picked, total value = 160, total weight = 30
- item 3 cannot be picked



#### These greedy algorithms can be arbitrarily bad

Greedy algorithm 1 - order by value

Given n items, one item with v=2 and w=W, and n-1 items with v=1 and  $w=\frac{W}{n-1}$ .

OPT: picks all small items, total value is n-1Greedy 1: picks the one item with v = 2, total value is 2.

#### These greedy algorithms can be arbitrarily bad

Greedy algorithm 1 - order by value

Given n items, one item with v=2 and w=W, and n-1 items with v=1 and  $w=\frac{W}{n-1}$ .

OPT: picks all small items, total value is n-1

Greedy1: picks the one item with v = 2, total value is 2.

Greedy algorithm 2 - order by value/weight

Given 2 items, one item with v = 2 and w = 1, and the other item with v = W and w = W.

OPT: picks second item, total value is W Greedv2: picks first item, total value is 2.

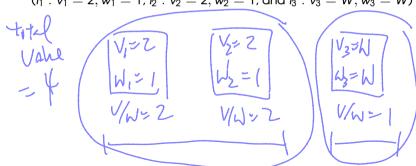
W=W

# **Greedy algorithm 3**

- 1. Sort the items in decreasing order of  $\frac{V_1}{W_2}$ , like Greedy algorithm 2. I.e.,
  - $\frac{v_1}{w_1} \geq \frac{v_2}{w_2} \geq \cdots \geq \frac{v_n}{w_n}$ .
- 2. Pick the next items as long as total weight does not exceed W.
- 3. Suppose items  $l_1, l_2, \dots l_{k-1}$  can be picked but  $l_k$  cannot.
- **4.** Consider  $v_1 + v_2 + \cdots + v_{k-1}$  and  $v_k$  and pick the set with larger value.

What does this mean for the example with 3 items?

$$(I_1: v_1=2, w_1=1, I_2: v_2=2, w_2=1, \text{ and } I_3: v_3=W, w_3=W)$$



#### Greedy algorithm 3

1. Sort the items in decreasing order of  $\frac{v_i}{w_i}$ , like Greedy algorithm 2. I.e.,

$$\tfrac{v_1}{w_1} \geq \tfrac{v_2}{w_2} \geq \cdots \geq \tfrac{v_n}{w_n}.$$

- 2. Pick the next items as long as total weight does not exceed W.
- **3.** Suppose items  $l_1, l_2, \cdots l_{k-1}$  can be picked but  $l_k$  cannot.
- **4.** Consider  $v_1 + v_2 + \cdots + v_{k-1}$  and  $v_k$  and pick the set with larger value.

What does this mean for the example with 3 items?

$$(I_1: v_1 = 2, w_1 = 1, I_2: v_2 = 2, w_2 = 1, \text{ and } I_3: v_3 = W, w_3 = W)$$

#### **Theorem**

Greedy algorithm 3 has an approximation ratio of at least  $\frac{1}{2}$ , i.e., it always select a subset with total value at least  $\frac{1}{2}$  OPT.

Greedy3: 
$$\max\{v_1 + v_2 + \dots + v_{k-1}, v_k\} \ge \frac{v_1 + v_2 + \dots + v_{k-1} + v_k}{2}$$
  
 $OPT \le v_1 + v_2 + \dots + v_{k-1} + v_k$   
 $Greedy3 \ge \frac{OPT}{2}$ 

#### **Lesson Learned**

Greedy algorithm does **NOT** always return the best solution

### **Summary**

Summary: Greedy algorithm for Knapsack Problem

Next: Kruskal's algorithm for Minimum Spanning Tree

# For note taking