

## Dynamic Programming

- The Knapsack Problem
- Dynamic Programming Formulation
- Implementation
- Fractional Knapsack and Subset Sum Problem

IMDAD ULLAH KHAN

# Knapsack Problem

## Input:

- Items:  $U = \{a_1, \dots, a_n\}$  ▷ Fixed order
- Weights:  $w : U \rightarrow \mathbb{Z}^+$  ▷  $(w_1, \dots, w_n)$
- Values:  $v : U \rightarrow \mathbb{R}^+$  ▷  $(v_1, \dots, v_n)$
- Capacity:  $C \in \mathbb{R}^+$

## Output:

- A subset  $S \subset U$
- Capacity constraint:

$$\sum_{a_i \in S} w_i \leq C$$

- Objective: Maximize

$$\sum_{a_i \in S} v_i$$

# Knapsack Problem: Fractional Version

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## Fractional Knapsack Problem

- Unlike 0 – 1 Knapsack, where you either take or leave an item
- Here you are allowed to take part of an item

Easy solution !

Greedily choose the best value per unit weight item

Proof of optimality follows from the the **cut and paste** type argument

## Knapsack Problem: Fractional vs 0 – 1 Knapsack

Fractional solution is often not feasible for 0 – 1 knapsack problem

▷ Select 1.27 of a software or 2.9 dish washers

ID	$w$	$v$	$v/w$
$a_1$	20	30	1.5
$a_2$	50	60	1.2
$a_3$	50	50	1

$$C = 110$$

- GREEDY-1 (most valued first) yields  $\{a_2, a_1\}$ , value 90, weight 70
- GREEDY-2 (least weighted first) yields  $\{a_1, a_2\}$ , value 90, weight 70
- GREEDY-3 (highest  $val/wt$  ratio first) yields  $\{a_1, a_2\}$ , value 90, weight 70
- INTEGRAL-OPT yields  $\{a_2, a_3\}$ , value 110, weight 100
- FRACTIONAL-OPT yields  $\{1(a_1), 1(a_2), 4/5(a_3)\}$ , value 130, weight 100

▷ INTEGRAL-OPT may not use total capacity

$$value(\text{FRACTIONAL-OPT}) \geq value(\text{INTEGRAL-OPT})$$

# Subset Sum Problem

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## Input:

- Items:  $U = \{a_1, \dots, a_n\}$  ▷ fixed order
- Weights:  $w : U \rightarrow \mathbb{Z}^+$  ▷  $(w_1, \dots, w_n)$
- Capacity:  $C \in \mathbb{R}^+$

## Output:

- A subset  $S \subset U$
- Capacity constraint:  $\sum_{a_i \in S} w_i \leq C$
- Objective: Maximize  $\sum_{a_i \in S} w_i$
- A CPU with  $C$  MFLOPS
- $n$  jobs: job  $i$  requires  $w_i$  MFLOPS
- Select jobs to get fewest idle CPU cycles

# Subset Sum Problem

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## Brute Force Solution

- Find all subsets of  $U$
- Calculate their sums
- Choose the one with the max sum

But there are  $2^n$  subsets

Try Greedy Algorithms!

It is a special case of the 0 – 1 Knapsack problem

▷ When all weights are equal to values

## Subset Sum Problem: Dynamic Programming

Let  $\text{OPT-VAL}(n, C)$  be the **value** of the optimal solution

Let  $\text{OPT-SET}(n, C)$  be the optimal solution (the subset)

$w_1$	$w_2$	$w_3$	...	...	...	$w_{n-1}$	$w_n$
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- Either  $n^{\text{th}}$  item is part of the solution
  - $\text{OPT-VAL}(n, C) = \text{OPT-VAL}(n-1, C - w_n) + w_n$
- Or it is not
  - $\text{OPT-VAL}(n, C) = \text{OPT-VAL}(n-1, C)$
- And we take maximum of the two

## The Recurrence

$$\text{OPT-VAL}(n, C) = \begin{cases} \text{OPT-VAL}(n-1, C - w_n) + w_n, & \text{if } v_n \in \text{OPT-SET}(n, C) \\ \text{OPT-VAL}(n-1, C) & \text{if } v_n \notin \text{OPT-SET}(n, C) \end{cases}$$

### ■ Some (base) special cases

- If no items:  $\text{OPT-VAL}(0, \cdot) = 0$
- If no space:  $\text{OPT-VAL}(\cdot, 0) = 0$
- If  $w_n > C$ :  $\text{OPT-VAL}(n, C) = \text{OPT-VAL}(n-1, C)$