

# Design and Analysis of Algorithm

Lecture-10:  
Greedy Algorithm

# Algorithm

## Algorithm JS( $d, j, n$ )

```
//  $d[i] \geq 1, 1 \leq i \leq n$  are the deadlines,  $n \geq 1$ . The jobs
// are ordered such that  $p[1] \geq p[2] \geq \dots \geq p[n]$ .  $J[i]$ 
// is the  $i$ th job in the optimal solution,  $1 \leq i \leq k$ .
// Also, at termination  $d[J[i]] \leq d[J[i+1]], 1 \leq i < k$ .
{
     $d[0] := J[0] := 0$ ; // Initialize.
     $J[1] := 1$ ; // Include job 1.
     $k := 1$ ;
    for  $i := 2$  to  $n$  do
    {
        // Consider jobs in nonincreasing order of  $p[i]$ . Find
        // position for  $i$  and check feasibility of insertion.
         $r := k$ ;
        while  $((d[J[r]] > d[i])$  and  $(d[J[r]] \neq r))$  do  $r := r - 1$ ;
        if  $((d[J[r]] \leq d[i])$  and  $(d[i] > r))$  then
        {
            // Insert  $i$  into  $J[ ]$ .
            for  $q := k$  to  $(r+1)$  step  $-1$  do  $J[q+1] := J[q]$ ;
             $J[r+1] := i$ ;  $k := k + 1$ ;
        }
    }
    return  $k$ ;
}
```

Given that

$$(p_1, p_2, p_3, p_4) = (100, 10, 15, 27)$$

$$(d_1, d_2, d_3, d_4) = (2, 1, 2, 1)$$

Sort jobs based on profit

$$(p_1, p_4, p_3, p_2) = (100, 27, 15, 10)$$

$$(d_1, d_4, d_3, d_2) = (2, 1, 2, 1)$$

Initialize

$J_0$	$J_1$	$J_2$	$J_3$	$J_4$
0	1			

# Complexity of the algorithm

For the given algorithm there are two parameters in terms of which its complexity can be measured.

1. *Total number of jobs (say  $n$ )*
2. *Number of jobs selected (say  $s$ )*

There are two loops (1) For loop (2) While loop

## For Loop

It will run maximum of  $n-1$  times, So complexity is  $O(n)$

## While Loop

In any iteration it will run for  $k$  times

Maximum possible value of  $k$  is  $s$

The complexity of the given algorithm will thus be  $O(ns)$

**Theorem** Let  $J$  be a set of  $k$  jobs and  $\sigma = i_1, i_2, \dots, i_k$  a permutation of jobs in  $J$  such that  $d_{i_1} \leq d_{i_2} \leq \dots \leq d_{i_k}$ . Then  $J$  is a feasible solution iff the jobs in  $J$  can be processed in the order  $\sigma$  without violating any deadline.

**Theorem** The greedy method described above always obtains an optimal solution to the job sequencing problem.

## Example

Let  $n = 5$ ,  $(p_1, p_2, p_3, p_4, p_5) = (20, 15, 10, 5, 1)$  and  $(d_1, d_2, d_3, d_4, d_5) = (2, 2, 1, 3, 3)$

J	Assigned slot	Job considered	Action	Profit
$\phi$	none	1	Assign to (1,2)	0
{1}	[1,2]	2	Assign to (0,1)	20
{1,2}	[0,1][1,2]	3	Reject	35
{1,2}	[0,1][1,2]	4	Assign to (2,3)	35
{1,2,4}	[0,1][1,2][2,3]	5	Reject	40

## Example

The **knapsack problem** derives its name from the maximization problem of the best choice of essentials that can fit into one bag to be carried on a trip.

Given a set of items, each with a weight and a value, determine the number of each item to include in a collection so that the total weight is less than a given limit and the total value is as large as possible.



# Knapsack Problem

## Problem Definition

Want to carry essential items in one bag

Given a set of items, each has

A cost (i.e., 12kg)

A value (i.e., 4\$)



To determine the number of each item to include in a collection so that

- The total cost is less than some given cost
- And the total value is as large as possible

## Three Types

- **0/1 Knapsack Problem**  
Restricts the number of each kind of item to zero or one
- **Bounded Knapsack Problem**  
Restricts the number of each item to a specific value
- **Unbounded Knapsack Problem**  
Places no bounds on the number of each item



## Fractional Knapsack

Given a set of items, each with a weight and a value, determine a subset of items to include in a collection so that the total weight is less than or equal to a given limit and the total value is as large as possible.



In this version of Knapsack problem, items can be broken into smaller pieces. So, one can choose a fraction of item as well.

$x_i$  of  $i^{th}$

The  $i^{th}$  item contributes the weight  $x_i \cdot w_i$  to the total weight in the knapsack and profit  $x_i \cdot p_i$  to the total profit.

# Fractional Knapsack Problem



we wish to find the maximum-benefit subset that doesn't exceed a given weight  $W$ .

# 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

The following is maximized:

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

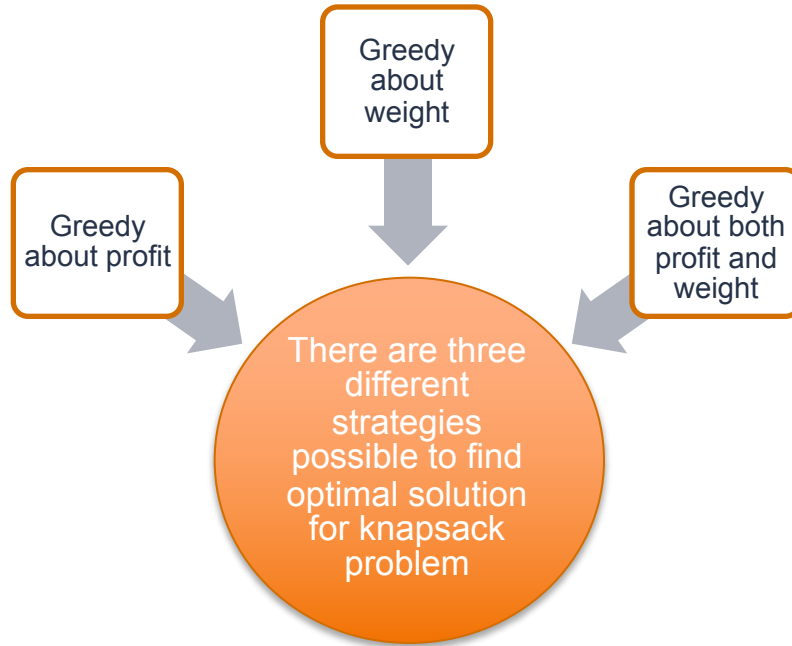
Objective Function

It satisfies the following constraint

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

Constraints

# The Fractional Knapsack Problem: Solution



## Greedy about both profit and weight

Item	Weight	Value
1	18	25
2	15	24
3	10	15

Compute value (or profit per unit of weight)

So value of Item 1 per unit weight is  $25/18 = 1.3$

Similarly, value of Item 2 per unit weight is  $24/15 = 1.6$

and value of Item 3 per unit weight is  $15/10 = 1.5$

# Calculation

- 1) Calculate the value-per-kg  $\rho_i = v_i/W_i$  for  $i = 1, 2, \dots, n$ .
- 2) Sort the items by decreasing  $\rho_i$

Let the sorted item sequence be  $1, 2, \dots, i, \dots, n$ , the corresponding value-per-kg and weight be  $\rho_i$  and  $w_i$  respectively.

- 3) Let  $k$  be the current weight limit (Initially,  $k = K$ ). In each iteration, we choose item  $i$  from the top of the unselected list.
  - 1) 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.
  - 2) If  $k < w_i$ , set  $x_i = k/w_i$  ( we take a fraction  $k/w_i$  of item  $i$ ), Then the algorithm terminates.

# Fractional Knapsack Algorithm

**Algorithm** FractionalKnapsack( $S, W$ ):

**Input:** Set  $S$  of items, such that each item  $i \in S$  has a positive benefit  $b_i$  and a positive weight  $w_i$ ; positive maximum total weight  $W$

**Output:** Amount  $x_i$  of each item  $i \in S$  that maximizes the total benefit while not exceeding the maximum total weight  $W$

**for** each item  $i \in S$  **do**

$x_i \leftarrow 0$

$v_i \leftarrow b_i/w_i$       {value index of item  $i$ }

$w \leftarrow 0$       {total weight}

**while**  $w < W$  **do**

    remove from  $S$  an item  $i$  with highest value index      {greedy choice}

$a \leftarrow \min\{w_i, W - w\}$       {more than  $W - w$  causes a weight overflow}

$x_i \leftarrow a$

$w \leftarrow w + a$

# Observation

- 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.



# Feasible and Optimal solution

A feasible solution is any set  $(x_1, x_2, x_3, \dots, x_n)$  which satisfies the constraint.



An optimal solution is a feasible solution for which the objective function is maximized.

**Question:** Consider the following instance of the knapsack problem:

$$n = 3, m = 20,$$

$$(p_1, p_2, p_3) = (25, 24, 15), \text{ and}$$

$$(w_1, w_2, w_3) = (18, 15, 10).$$

Four feasible solutions are

	$x_1, x_2, x_3$	$\sum w_i x_i$	$\sum p_i x_i$
1	1/2, 1/3, 1/4	16.5	24.25
2	1, 2/15, 0	20	28.2
3	0, 2/3, 1	20	31
4	0, 1, 1/2	20	31.5

## Question

**Question:** Consider the following instance of the knapsack problem. The max capacity of knapsack is 12

	$ob_1$	$ob_2$	$ob_3$	$ob_4$	$ob_5$
p	5	2	2	4	5
w	5	4	6	2	1

Compute profit per unit weight for each object.

	$ob_1$	$ob_2$	$ob_3$	$ob_4$	$ob_5$
p/w	1	0.5	0.33	2	5

*solution*    1,1,0,1,1

**Question:** Consider the following instance of the knapsack problem:

Item	X1	X2	X3	X4	X5
Profit	15	12	9	16	17
Weight	2	5	3	4	6

The maximum weight of 12 is allowed in the knapsack.

Find the value of maximum profit with the optimal solution of the fractional knapsack problem.

## Question

**Question:** Consider the following instance of the knapsack problem:

	$ob_1$	$ob_2$	$ob_3$	$ob_4$	$ob_5$	$ob_6$	$ob_7$
p	10	5	15	7	6	18	3
w	2	3	5	7	1	4	1

The maximum weight of 15 is allowed in the knapsack.