

# Optimization Techniques

## Laboratory 5

Iterated Local Search, Simulated Annealing

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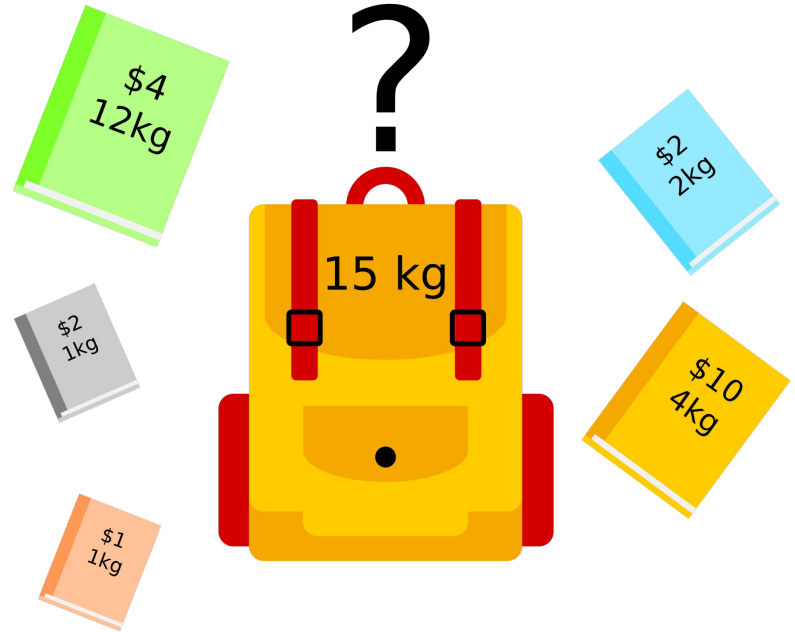


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# 0-1 Knapsack Problem

Maximize the sum of values in the bag bounded by the total weight

- Bag has capacity  $W$
- Each item has  
volume  $w$   
value  $v$



# 0-1 Knapsack Problem

Assume

$N$  = number of items

$X = \{x_1, x_2, \dots, x_N\}$  = set of items

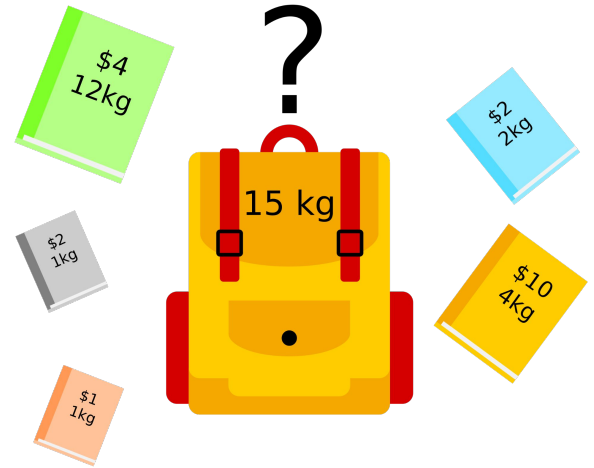
$v_i$  = value of item  $x_i$

$w_i$  = weight of item  $x_i$

$W$  = maximum weight the knapsack can hold

Then the 0/1 knapsack problem can be formulated as follows:

$$\begin{array}{ll}\text{maximize} & \sum_{x_i \in X} v_i x_i \\ \text{subject to} & \sum_{x_i \in X} w_i x_i \leq W \\ & x_i \in \{0, 1\} \forall x_i \in X\end{array}$$



# Iterated Local Search

Stochastic Local Search method based on an embedded heuristic

Given solution  $\mathbf{s}^*$ :

1. Perturb  $\mathbf{s}^*$  leading to intermediate state  $\mathbf{s}'$
1. Perform Local Search around  $\mathbf{s}'$ , arriving at solution  $\mathbf{s}'^*$
1. If  $\mathbf{s}'^*$  passes the acceptance test, then  $\mathbf{s}^* = \mathbf{s}'^*$
1. Back to step 1 until convergence

**procedure** *Iterated Local Search*

$s_0 \leftarrow \text{GenerateInitialSolution}$

$s^* \leftarrow \text{LocalSearch}(s_0)$

**repeat**

$s' \leftarrow \text{Perturbation}(s^*, \text{history})$

$s'^* \leftarrow \text{LocalSearch}(s')$

$s^* \leftarrow \text{AcceptanceCriterion}(s^*, s'^*, \text{history})$

**until** termination condition met

**end**

Better than random restarts as we do not completely give up on the current solution

# Simulated Annealing

## Improves greedy search

by allowing step to a worse neighbour

- **Neighbor selection**

transform the given solution  
to introduce a new solution

- **Temperature scheduling**

willingness to accept a worse state  
decreased over time

- **Acceptance probability**

probability of going from one state to a  
neighbor state

- $X \leftarrow$  Initial configuration
- $T \leftarrow$  Initial high temperature
- Iterate:
  1. Do  $K$  times:
    - 1.1  $E \leftarrow Eval(X)$
    - 1.2  $X' \leftarrow$  one configuration randomly selected in  $Neighbors(X)$
    - 1.3  $E' \leftarrow Eval(X')$
    - 1.4 If  $E' \leq E$   
 $X \leftarrow X'; E \leftarrow E';$   
Else accept the move with probability  $p = e^{-(E'-E)/T}$   
 $X \leftarrow X'; E \leftarrow E';$
  2.  $T \leftarrow \alpha T$