



Knapsack Problem

Backtracking

Subset Sum Problem

Given a set W of n elements and a sum S

We have to find all possible subsets T of W , whose elements add up to at most S , such that there is no other subset of W whose elements add up to more than that of T .

e.g. $W = \{ 8\ 6\ 7\ 5\ 3\ 10\ 9 \}$ and $S = 15$

We get four subsets

$\{ 8\ 7 \}$

$\{ 6\ 9 \}$

$\{ 7\ 5\ 3 \}$

$\{ 5\ 10 \}$

Application: say we have a CPU with S free cycles and a set W of n jobs, we want to choose the subset of jobs (each job i taking $W[i]$ time) that minimizes the number of idle cycles.

Subset Sum Problem

The subset sum problem can be stated as follows:

Given a **set** W of n elements and a **sum** S

We have to find a **subset** T of W , whose elements **add up** to at most S , such that there is **no other** subset of W whose elements add up to more than that of T .

In other words it is an optimization problem denoted as:

$$SubsetSum(S, k) = \begin{cases} SubsetSum(S, k - 1), & \text{if } W_k > S \\ \max \begin{cases} SubsetSum(S, k - 1) \\ W_k + SubsetSum(S - W_k, k - 1) \end{cases}, & \text{otherwise} \end{cases}$$

Subset Sum Problem algorithm using Backtracking

Algorithm subsetSum(s, k, x)

// finds the maximal subset of the given set w[1 ... n] of n elements, whose elements add up to at most s

// x[1 ... n] represents the solution:

// $x[k] = 0$ implies that the k^{th} element of w is NOT taken, and $x[k] = 1$ implies that the k^{th} element of w is taken

// initial call is: subsetSum(s, n, x = {})

```
1  if (k == 0) return 0
2  let y[1 ... k], z[1 ... k] be two arrays
3  if (w[k] ≤ s) sumWithk = w[k] + subsetSum(s - w[k], k - 1, y)
4  else sumWithk = 0
5  sumWithoutk = subsetSum(s, k - 1, z)
6  if (sumWithk > sumWithoutk)
7      for (i = 1 to k - 1) x[i] = y[i]
8      x[k] = 1
9      return sumWithk
10 else
11     for (i = 1 to k - 1) x[i] = z[i]
12     x[k] = 0
13     return sumWithoutk
```

The 0/1 knapsack problem can be stated as follows:

Given a **set** of **n** elements, each element k having **weight** W_k and **value** V_k , and a knapsack **capacity** c

We have to find a **subset** T of the given set, whose weights **add up** to at most c , such that there is **no other** subset whose **values** add up to more than that of T .

In other words it is an optimization problem denoted as:

$$knapsack(c, k) = \begin{cases} knapsack(c, k - 1), & \text{if } W_k > c \\ \max \left\{ \begin{array}{l} knapsack(c, k - 1) \\ V_k + knapsack(c - W_k, k - 1) \end{array} \right\}, & \text{otherwise} \end{cases}$$

Difference with Subset Sum: want to maximize **value** instead of weight.

Algorithm knapsack(c, k, x)

// finds a subset with maximal value, of the given set of n elements having weights $w[1 \dots n]$ and values $v[1 \dots n]$

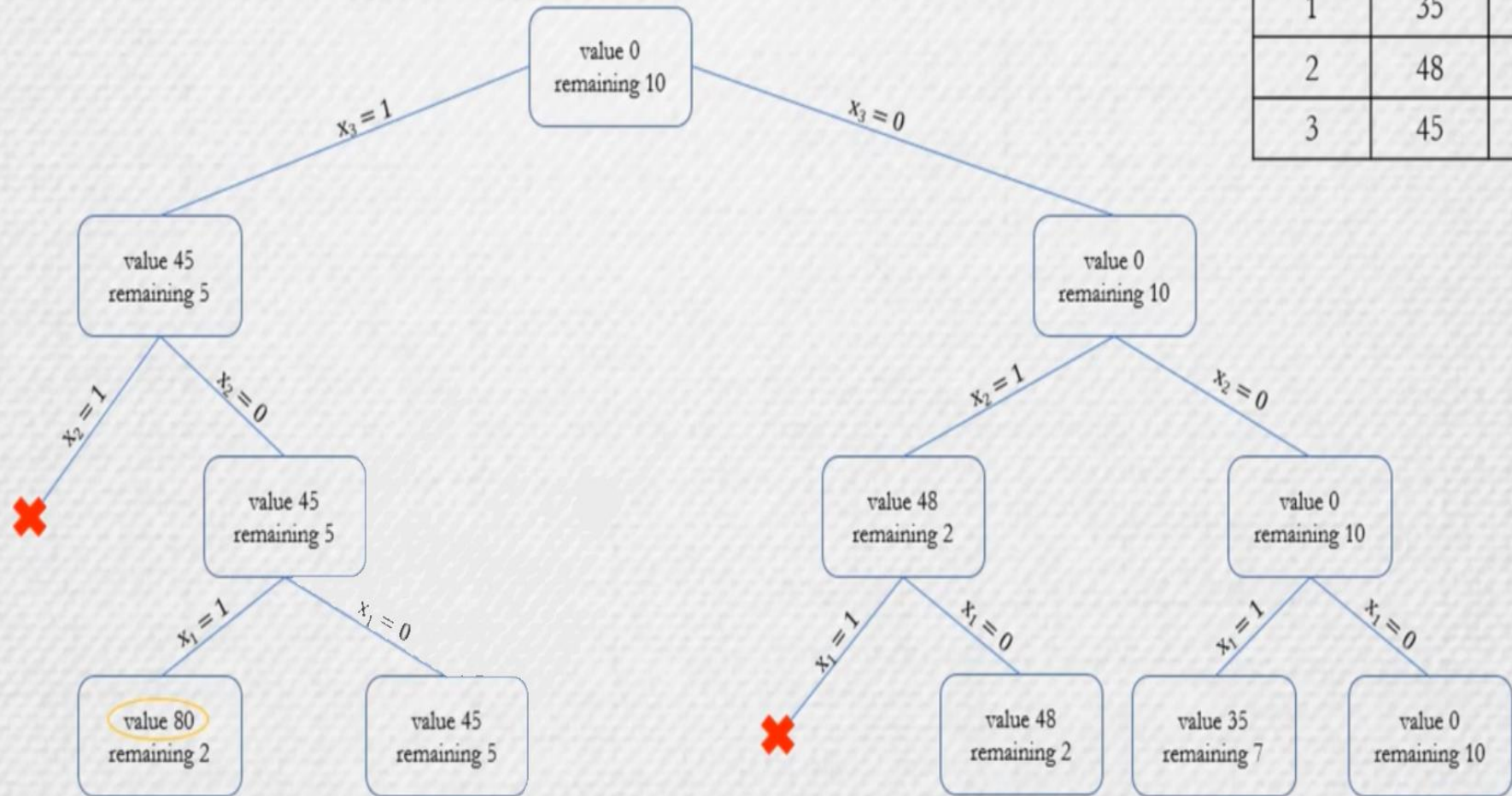
// whose elements' weights add up to at most c (the knapsack capacity)

// $x[k] = 0$ implies that the k^{th} element of w is NOT taken, and $x[k] = 1$ implies that the k^{th} element of w is taken

// initial call is: knapsack($c, n, x = \{\}$)

```
1   if ( $k == 0$ ) return 0
2   let  $y[1 \dots k], z[1 \dots k]$  be two arrays
3   if ( $w[k] \leq c$ )  $\text{valWithk} = v[k] + \text{knapsack}(c - w[k], k - 1, y)$ 
4   else  $\text{valWithk} = 0$ 
5    $\text{valWithoutk} = \text{knapsack}(c, k - 1, z)$ 
6   if ( $\text{valWithk} > \text{valWithoutk}$ )
7       for ( $i = 1$  to  $k - 1$ )  $x[i] = y[i]$ 
8        $x[k] = 1$ 
9       return  $\text{valWithk}$ 
10  else
11      for ( $i = 1$  to  $k - 1$ )  $x[i] = z[i]$ 
12       $x[k] = 0$ 
13      return  $\text{valWithoutk}$ 
```

Capacity = 10		
item	value	weight
1	35	3
2	48	8
3	45	5



Number of nodes explored = 11