

## 12 Subset-Sum and Knapsack [KT 6.4]

**Definition of subset-sum.** Find a subset  $S$  that maximizes  $\sum_{i \in S} w_i$ , subject to the constraint that  $\sum_{i \in S} w_i \leq W$ .

**Definition of Knapsack.** Find a subset  $S$  that maximizes  $\sum_{i \in S} v_i$  where  $v_i$  is the value of  $i$ , subject to the constraint that  $\sum_{i \in S} w_i \leq W$ .

We will start from subset-sum.

**Greedy does not work.** Sort the items by decreasing weight, and then select items in this order as long as the total weight remains below  $W$ .

Counter-example:  $\{W/2 + 1, W/2, W/2\}$ .

If we sort items by increasing weight, then it fails on input  $\{1, W/2, W/2 + 1\}$ .

**False start.** If  $n \notin \mathcal{O}$ , then  $OPT(n) = OPT(n - 1)$ . If  $n \in \mathcal{O}$ , then what? There is no analogy of deleting all the conflict intervals as that in the case of weighted interval scheduling, since after accepting  $n$  we have  $W - w_n$  budget left. Therefore  $w_n$  needs to be in the picture.

**The right solution.** This suggests us to use more subproblems: for each initial set  $\{1, \dots, i\}$  ( $i \leq n$ ), and each value  $\{1, \dots, w\}$  ( $w \leq W$ ). New recursion:

1. If  $n \notin \mathcal{O}$ , then  $OPT(n, W) = OPT(n - 1, W)$ .
2. If  $n \in \mathcal{O}$ , then  $OPT(n, W) = w_n + OPT(n - 1, W - w_n)$ .

We should take the larger of the two:

$$OPT(n, W) = \max\{w_n + OPT(n - 1, W - w_n), OPT(n - 1, W)\}.$$

For the base cases, we have  $OPT(i, W) = 0$  for any  $1 \leq i \leq n$  and  $W \leq 0$ , and  $OPT(i, W) = 0$  for  $i = 0$  and any  $W$ .

**Extend this to Knapsack.** Just replace  $w_n$  with  $v_n$  and that's all.