# Subset sum and Knapsack problem

More Dynamic Programming Problems

Martin Chikov

November 4, 2018

Swiss Olympiad in Informatics

## Subset Sum Problem

**Problem:** Given is a list  $a_0, a_1, \ldots, a_{n-1}$  of non-negative integers and an integer *S*.

Find out if it is possible to choose some of these numbers so that their sum is equal to S

## List A = [5; 10; 20; 50; 100; 200; 500] and S = 390:

## List A = [5; 10; 20; 50; 100; 200; 500] and S = 390: Answer: No.

List A = [5; 10; 20; 50; 100; 200; 500] and S = 390: Answer: No.

List A = [5; 10; 20; 50; 100; 200; 500] and S = 875:

List A = [5; 10; 20; 50; 100; 200; 500] and S = 390: Answer: No.

List A = [5; 10; 20; 50; 100; 200; 500] and S = 875: Answer: Yes. Idea: Check all sums we can achieve

Let say we have a set with i elements. In order to find if we can make sum k using some of these elements, there are 2 cases we need to check: Idea: Check all sums we can achieve

Let say we have a set with i elements. In order to find if we can make sum k using some of these elements, there are 2 cases we need to check:

1. Can we make the sum k using the first i - 1 elements? (is k one of the subset sums of the first i - 1 elements)

#### Idea: Check all sums we can achieve

Let say we have a set with i elements. In order to find if we can make sum k using some of these elements, there are 2 cases we need to check:

- 1. Can we make the sum k using the first i 1 elements? (is k one of the subset sums of the first i 1 elements)
- Can we make the sum k − a<sub>i</sub> using the first i − 1 elements?
   (can we make k by adding the value of a<sub>i</sub> to each of the subset sums of the first i − 1 elements)

**Example:** A = [3; 7; 10]

The sets of the subset sums are:

Using the first 3 elements:

 $\{(), (3), (7), (3+7=10), (10), (3+10=13), (7+10=17), (3+7+10=20)\}$ 

Using the first 2 elements:

 $\{(), (3), (7), (3+7=10)\}$ 

Using the first 1 element:

{(), (3)}

Using the first 0 elements:

{()}

## Implementation: Recursion

```
vector<int> a;
bool Subsetsum (int i, int k) {
    if (k == 0)
        return true;
    if (i == 0 \&\& k != 0)
        return false;
    if (a[i-1] > k)
        return Subsetsum(i-1, k);
    return Subsetsum(i-1, k) ||
           Subsetsum(i-1, k-a[i-1]);
}
bool answer = Subsetsum(a.size(),sum);
```

The total amount of subsets we have to check is equal to  $2^n$ .

For smaller values of n this can work relatively fast. However for bigger values (n = 100):

 $2^{100}$  > Edge of the observable universe

How can we do better?

- 1. Define sub problems.
- 2. Find a general recurrence formula to solve a sub problem using the solution to other sub problems.
- 3. Find base case(s).
- 4. Which is the relevant sub problem?

## 1.Define sub problems:

S(i, k) - using the first *i* elements can we make sum *k*?

Let's say we know all solutions S(i, k) for the first i-1 elements. We consider a new element:  $a_i$ For any sum k we have 2 cases:

Let's say we know all solutions S(i, k) for the first i-1 elements. We consider a new element:  $a_i$ For any sum k we have 2 cases:

• We don't add the element to the sum S(i, k) = S(i - 1, k)

Let's say we know all solutions S(i, k) for the first i-1 elements. We consider a new element:  $a_i$ For any sum k we have 2 cases:

- We don't add the element to the sum S(i, k) = S(i 1, k)
- We add the element to the sum S(i, k) = S(i - 1, k - a[i])

Let's say we know all solutions S(i, k) for the first i-1 elements. We consider a new element:  $a_i$ For any sum k we have 2 cases:

- We don't add the element to the sum S(i, k) = S(i 1, k)
- We add the element to the sum S(i, k) = S(i 1, k a[i])

Therefore our general formula is:  $S(i, k) = S(i - 1, k) \parallel S(i - 1, k - a[i])$ 

#### 3.Find base case(s).

S(0,0) = true - using 0 elements we can make sum = 0S(0,x) = false (for x > 0) - using 0 elements we can't make any sum > 0

## 4. Relevant sub problem:

S(n, sum) - using *n* elements can we make sum?

```
void Subsetsum_dp() {
    vector<vector<bool> > dp(n+1, vector<int>(s+1, 0));
    dp[0][0]=1;
    for(int i=1; i<=n; i++) {</pre>
        for(int k=0: k<=s: k++) {
            if(k-v[i-1]<0)
               dp[i][k] = dp[i-1][k];
            else
               dp[i][k] = dp[i-1][k] || dp[i-1][k-v[i-1]];
        }
    }
    if(dp[n][s]==1) cout << "Yes\n";
    else cout << "No\n";
```

For each element we have to calculate whether we can achieve sum  ${\sf k}$  or not. Therefore we have:

Running time of O(nS)

(where n is the number of elements and S is the sum we want to check)

# **Knapsack Problem**

**Problem:** Given a set of list  $a_0, a_1, \ldots, a_{n-1}$  of items each with a weight  $w_0, w_1, \ldots, w_{n-1}$  and a value  $v_0, v_1, \ldots, v_{n-1}$ .

Choose which items to take such that the total weight is less than or equal to C and the total value of the items is as large as possible.

Weight W = [5; 4; 6; 3]
 Value V = [10; 40; 30; 50]
 Knapsack capacity = 10

Weight W = [5; 4; 6; 3]
 Value V = [10; 40; 30; 50]
 Knapsack capacity = 10

Answer = 90 (we take  $a_1$  and  $a_3$  with total weight = 7)

- Weight W = [5; 4; 6; 3] Value V = [10; 40; 30; 50] Knapsack capacity = 10 Answer = 90 (we take a<sub>1</sub> and a<sub>3</sub> with total weight = 7)
- Weight W = [3; 5; 5; 6]
   Value V = [10; 60; 60; 100]
   Knapsack capacity = 10

- Weight W = [5; 4; 6; 3] Value V = [10; 40; 30; 50] Knapsack capacity = 10 Answer = 90 (we take a<sub>1</sub> and a<sub>3</sub> with total weight = 7)
- Weight W = [3; 5; 5; 6] Value V = [10; 60; 60; 100] Knapsack capacity = 10

Answer = 120 (we take  $a_1$  and  $a_2$  with total weight = 10)

## **Brute-force Idea:** Try all of the possibilities. Running time: $O(2^n)$ - too slow.

Let's try thinking of a DP solution!

### 1.Define sub problems:

S(i, k) - what is the maximum total value we can achieve using the first *i* items, with the chosen items total weight less or equal to k?

Let's assume we know all solutions S(i, k) for the first i - 1 items. We consider a new item with value  $v_i$  and weight  $w_i$ For any total weight k we have 2 cases:

Let's assume we know all solutions S(i, k) for the first i - 1 items. We consider a new item with value  $v_i$  and weight  $w_i$ For any total weight k we have 2 cases:

• We don't take the item or if  $w_i > k$ S(i, k) = S(i - 1, k)

Let's assume we know all solutions S(i, k) for the first i - 1 items. We consider a new item with value  $v_i$  and weight  $w_i$ For any total weight k we have 2 cases:

- We don't take the item or if  $w_i > k$ S(i,k) = S(i-1,k)
- We take the item and  $w_i \le k$  $S(i,k) = S((i-1,k-w_i)+v_i)$

Let's assume we know all solutions S(i, k) for the first i - 1 items. We consider a new item with value  $v_i$  and weight  $w_i$ For any total weight k we have 2 cases:

- We don't take the item or if  $w_i > k$ S(i,k) = S(i-1,k)
- We take the item and  $w_i \le k$  $S(i,k) = S((i-1,k-w_i)+v_i)$

Therefore our general formula is:  $S(i, k) = max(S(i-1, k), S((i-1, k-w_i) + v_i))$ 

## **3.Find base case(s).** S(0, k) = 0 - using 0 items the best value we can have is 0 S(i, 0) = 0 - using *i* items the best value with weight 0 is 0

### 4. Relevant sub problem:

S(n, C) - using *n* items what is the maximum total value with weight less or equal to *C*?

```
int Knapsack_dp() {
    vector<vector<int> > dp(n+1, vector<int>(c+1, 0));
    for (int i=1; i<=n; i++) {</pre>
        for (int k=1: k <=c: k++) {
             if (w[i-1] \leq k)
                 dp[i][k] = max(v[i-1] + dp[i-1][k-w[i-1]]),
                                 dp[i-1][k]);
             else
                 dp[i][k] = dp[i-1][k];
        }
    }
    return dp[n][c];
}
```

For each element we have to calculate the best value with weight at most k. Therefore we have:

Running time of O(nC)

(where n is the number of elements and C is the knapsack capacity)

• Dynamic programming is tricky at first, because of the big overall picture

- Dynamic programming is tricky at first, because of the big overall picture
- Look at the small problems and base cases solve them

- Dynamic programming is tricky at first, because of the big overall picture
- Look at the small problems and base cases solve them
- Use the solution of these small problems to solve it for a slightly bigger one

- Dynamic programming is tricky at first, because of the big overall picture
- Look at the small problems and base cases solve them
- Use the solution of these small problems to solve it for a slightly bigger one
- Using these small blocks construct your solution

- Dynamic programming is tricky at first, because of the big overall picture
- Look at the small problems and base cases solve them
- Use the solution of these small problems to solve it for a slightly bigger one
- Using these small blocks construct your solution

Most importantly - practice! The more experience you have the easier and faster you'll see the concepts!