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## ***Brute-Force Algorithmic Approach***

### *Solutions to the Practical Exercises*

#### *DA 2023 Instructors Team*

*Departamento de Engenharia Informática (DEI)*  
*Faculdade de Engenharia da Universidade do Porto (FEUP)*

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#### **Exercise 1**

- a) See source code.
- b) See source code.

The improved solution generates candidate triplets (a,b,c) in a smarter manner:

- it generates all possible T values for A, between 1 and T.
  - for each value of a, it generates all possible values for B, between 1 and (a - 1), to avoid symmetric solutions (given that (a,b,c) is a solution if and only if (b,a,c) because the addition operation is commutative).
  - it does not need to test all possible values for c, since the correct value for c can be inferred, since  $a + b + c = T$  is equivalent to  $c = T - a - b$ . All that is needed to have a valid solution is to check whether  $c > 0$ .
- c) In the worst case, there is no solution, so the algorithm cannot stop earlier and must test all possible solutions.

The number of different candidates for the pair (a,b) is:

$$T + (T - 1) + (T - 2) + \dots + 3 + 2 + 1$$

This value is due to there being T valid values for b when a = 1, (T-1) values for b when a = 2, and so on. The above sum is equivalent to:

$$\frac{T * (T + 1)}{2}$$

Testing each candidate takes O(1) time, as no operation in the innermost for-loop scales with T. Thus, the temporal complexity is, in the worst-case  $O(T * (T + 1) / 2) = O(T * (T + 1)) = O(T^2 + T) = O(T^2)$ .

## Exercise 2

- a) The idea of the Brute-force solution is to represent the candidate solutions as a bitmask (*i.e.*, array of booleans) with the same length as the input array  $\mathcal{A}$ . A value of **true** in a given position  $i$  of the bitmask means that the corresponding candidate solution uses the  $i$ -th element of  $\mathcal{A}$ . Using a bitmask is useful as it makes it intuitive how to enumerate all the candidate solutions (*i.e.*, all the possible subsets).

Computing the next value of the bitmask intuitively corresponds to incrementing a binary number by 1, where the numbers are represented backwards (*i.e.*, the least significant bit is the leftmost bit, not the rightmost bit). Example of the enumeration of the 8 candidates for an array of length 3:  
 $0\ 0\ 0 \rightarrow 1\ 0\ 0 \rightarrow 0\ 1\ 0 \rightarrow 1\ 1\ 0 \rightarrow 0\ 0\ 1 \rightarrow 1\ 0\ 1 \rightarrow 0\ 1\ 1 \rightarrow 1\ 1\ 1$

### Pseudo-code for finding a subset with the correct sum:

Input:

- A: array of non-negative numbers
- T: desired sum

Output:

- boolean indicating if the subset exists, subset of A whose sum is T

```
subsetSum(A, T)
    // Create the first candidate
    for(i = 0; i < A.size(); i++)
        curCandidate[i] ← false

    // Iterate over all the candidates
    while(true)
        // Verify if the candidate is a solution
        sum ← 0
        for(i = 0; i < A.size(); i++)
            sum ← sum + A[i]*curCandidate[i] //only sums A[i] if curCandidate[i] = true
        if(sum = T)
            // Build and return the solution
            for(i = 0; i < A.size(); i++)
                if(curCandidate[i])
                    subset.push(A[i])
            return true, subset
        // Get the next candidate
        curIndex ← 0
        while(curCandidate[curIndex])
            curIndex++
        if(curIndex = A.size()) break
```

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```
if(curIndex = A.size()) break
for(i = 0; i < curIndex; i++)
    curCandidate[i] ← false
curCandidate[curIndex] ← true
return false, []
```

- b) The temporal complexity of the algorithm in the worst case is  $O(n \cdot 2^n)$ , where  $n$  is array  $A$ 's length. This is due to there being at most  $2^n$  candidates to try out and processing each candidate takes  $O(n)$  since a fixed amount of for-cycles executes over  $n$  iterations at most.

The spatial complexity of the algorithm is  $\Theta(n)$ , not only for the worst-case, but for any other case. This is due to space always being allocated to the current candidate, whose size scales with  $n$ . If there is a valid solution, then space is also allocated for the solution, which also scales with  $n$ . All the other variables have a fixed size with respect to  $n$ .

- c) See source code.

### Exercise 3

- a) See source code.
- b) The algorithm has a temporal complexity of  $\Theta(n^3)$  since all  $n^2$  possible contiguous subarrays are tested and processing each one takes  $\Theta(n)$  time. This temporal complexity is valid for the best, worst and average cases, since all subarrays are tested, no matter what.

### Exercise 4

- a) See source code. The code contains comments explaining in detail the steps of the solution.
- b) The algorithm has a temporal complexity of  $O(n \cdot S^n)$  since there are at most  $S^n$  candidate solutions (if all  $n$  coin denominations have an equal stock of  $S$ ) and processing each one takes  $\Theta(n)$  time (namely to obtain a candidate's total value and number of coins used).

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**Exercise 5**

- a) It is a solution to the **closest pair problem**, where the goal is to discover the two points that minimize the Euclidean distance between them, among a set of 2D points.
- b) The algorithm uses a brute-force strategy since it exhaustively generates all possible solutions (i.e. pairs of points) and tests if they are better than the best solution found so far (i.e., the pair of points that minimizes the Euclidean distance).
- c) The temporal complexity is  $\Theta(n^2)$ , for the best, average and worst cases, given that all  $n^2$  possible pairings of points are always tested and each one is processed in  $O(1)$  (comparing two points and evaluating the Euclidean distance takes constant time).
- d) See source code.

**Exercise 6**

- a) See source code.
- b) The algorithm has a temporal complexity of  $\Theta(n \cdot 2^n)$  since there are  $2^n$  candidate solutions and processing each one takes  $\Theta(n)$  time (namely to obtain a candidate's total value and weight). This temporal complexity is valid for the best, worst and average cases, since all subarrays are tested, no matter what.

**Exercise 7**

- a) See source code.
- b) The algorithm has a temporal complexity of  $\Theta(n \cdot n!)$  since all possible permutations of the path are considered and processing each one takes  $\Theta(n)$  time. This temporal complexity is valid for the best, worst and average cases.