Greedy algorithms

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Abstract

This chapter introduces the last kind of algorithms presented in this book, i.e. the greedy algorithms. The historic hero introduced in these notes is Evelyn Berezin, one of the most important business women of the past century, who have created the first word processor.

Historic hero: Evelyn Berezin

Evelyn Berezin (depicted in Figure 1) was a physicist. She began to work in a company that produced digital computers, and where she started to work on particular the development of the logic designs of computers – e.g. [Auerbach et al., 1962]. After a bunch of years passed in changing job and several contributions related to the development of large computer systems such as the computerised reservation system for United Airlines, in 1969, she founded her own company: Redactron Corporation.

In this new company, she started to work on computer systems to simplify the work of secretaries. The main product of the company was called Data Secretary: the very first word processor in history. It was a stand-alone device developed for addressing that specific task, to replace the more common typewriter. Data Secretary was the precursor of all the series of word processors developed since that date. While they were stand-alone devices initially, word processors soon have become independent software applications. We remember Electric Pencil (1976), WordStar (1978), Microsoft Word (1983) and OpenOffice Writer (1999).
Greedy algorithms

A greedy algorithm is a particular algorithmic approach. At every stage of execution, it always makes the optimal choice (i.e. the best one) in that specific moment. For certain kinds of problems, this behaviour allows us to reach the best possible solution to the computational problem into consideration. For instance, if you have to determine the minimum number of euro coins needed for making a change, then a greedy algorithm will return an optimal solution overall:

1. consider the coins to choose for the change as ordered in a decrescent way, from the highest value (i.e. 2 euros) to the lowest one (i.e. 1 cent);
2. for each kind of value, add in the candidate set of the solution as much coins of that value as possible until their sum is lesser than the remaining of the change to give;
3. If we reach the change value, return it.

However, sometimes it is possible that the solution found, while it provides a correct solution to the problem, is just a suboptimal solution. For instance, driving from Florence to Bologna, we
can encounter a crossroad with two signs indicating two different routes to get to Bologna. The left road allows us to get to Bologna by travelling for 42 kilometres. On the other hand, the right way enables us to get to Bologna by going for 56 kilometres. A simple greedy approach would select the left route: at the moment, it seems the most convenient scenario. However, the plan does not predict the existence of possible traffic on the left road. Consequently, it would be possible to arrive in Bologna even after a car that takes the right route.

There are two main characteristics that a computational problem should show to be sure that the application of a greedy approach will bring to an optimal solution to the problem. The first one is that the greedy choice property should be guaranteed. This property means that, at a particular step, we can choose the best candidate for improving the set of candidates bringing to a solution.

The other characteristic is the problem as an optimal substructure. In particular, we must build the optimal solution to a computational problem by considering the optimal solutions to its subproblems. For instance, the previous example of the travel from Florence to Bologna does not have an optimal substructure. We can encounter accidents on a road we chose in a previously-chosen optimal strategy (i.e. the shortest path).

**Line wrap**

Wrapping a line, i.e. understanding where to break a line in a page, is one of the problems one has to tackle when dealing with documents, either in print or digital forms. For instance, when a person is using a typewriter for writing a document, at a certain point, after she has written a bunch of characters, there is a mandatory action to perform which is the carriage and return operation, that is performed mechanically on the typewriter itself. When the writer notices that the line has no more space for imprinting a new word, she initialises the typewriter to start from the very beginning of the left border but in the following line.

![Figure 2](image.png)

**Figure 2.** A screenshot that depicts how OpenOffice Writer deals with line wrap.
# Test case for the function
def test_line_wrap(text, line_width, expected):
    result = line_wrap(text, line_width)
    if expected == result:
        return True
    else:
        return False

# Code of the function
def line_wrap(text, line_width):
    result = [] # the list of all the lines of a document

    # the maximum available space per a specific line
    space_left = line_width
    # the current line that is built
    line = []

    for word in text.split(" "):  # we consider the length of the word plus one space character
        word_len = len(word)
        if word_len + 1 > space_left:
            result.append(" ".join(line))
            line = [word]
            space_left = line_width - word_len
        else:
            line.append(word)
            space_left = space_left - word_len + 1

    # we add the remaining line to the document
    result.append(" ".join(line))

    return \
        "\n".join(result)

# Tests
print(test_line_wrap("Just a word.", 15, "Just a word."))
print(test_line_wrap("Just a word.", 1, "\nJust\na\nword."))
print(test_line_wrap("This is a simple example.", 10, \
    "This is a\nsimple\nexample."))
In modern tools, such as word processors (shown in Figure 2), an algorithm is in charge of handling the line wrap. Such algorithm takes care of choosing when there is enough space to put that word in the current line. Generally speaking, we can describe the problem in the following manner:

**Computational problem:** break a text into lines such that it will fit in the available width of a page.

A greedy approach is very efficient and effective for addressing the aforementioned computational problem. It will proceed as follows:

1. For each word in the input text, see if there is enough space in the line for adding that word;
2. If there is space, add the word to the line; otherwise,
3. Declare finished the current line, and add the word as the first token of the following line.

To implement this algorithm, we need two methods for tokenizing and recomposing strings. The first of these methods is `<string>.split(<string_separator>)`. This method allows us to separate the string according to a specific set of characters the string may contain, specified by the parameter `<string_separator>`. For instance, if we have the variable `my_string` assigned to "a b c", the execution of the aforementioned method, i.e. `my_string.split(" ")`, returns the following list: ["a", "b", "c"].

The other method we need, i.e. `<string_separator>.join(<list_of_strings>)`, implements the opposite operation, i.e. it is able to concatenate the strings in a list again, according to a particular sequence of characters. For instance, if we have the list `my_list = ["a", "b", "c"]`, the execution of the aforementioned method, i.e. " ".join(my_list), returns the following string: "a b c".

We now have all the ingredients for implementing our algorithm for the line-wrap in Python, as shown in Listing 1.

**Exercises**

1. Implement the algorithm introduced in Section "Greedy algorithms" for returning the minimum amount of coins for a change. Accompany the implementation of the function with the appropriate test cases.
2. Suppose one has to address the maximum number of activities in a day choosing them from a set of available activities, considering that one cannot address more than one activity simultaneously. Each activity is defined by a tuple, where the first element defines the starting time (a value from 0 to 24, indicating the starting hour) while the second element define the finish time (a value from 0 to 24, indicating the finish hour).
Develop the Python function `def select_activities(set_of_activities)` by using a greedy approach. It takes in input a set of activities of a day and returns the list of the maximum number of non-overlapping activities one can address, ordered according to the starting time. Hint: think about the finish time of each activity and see how it may affect the selection. Accompany the implementation of the function with the appropriate test cases.

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References