

# Scientific Computation Project 1

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## Part 1

### 1.

`method1` uses linear (naive) search. It iterates through the list one by one, and if it finds a match, it returns the index. Otherwise, it returns  $-1000$ . The worst case is when it has to check every element in the entire list (after which it returns  $-1000$ ), so the worst-case asymptotic time complexity is  $O(n)$  for one search.

`method2` uses binary search. It has a parameter `flag` which controls whether a sorting is needed. For an unsorted list (the default case), it will first sort the list. We can see that `func2A` and `func2B` together implement a merge sort and `func2C` implements a binary search. `method2` first calls `func2B` on the enumerated list `L2` (i.e., `L` paired with index), then performs a binary search with `func2C`. For a sorted list (specified by passing `flag = false`), only the binary search function `func2C` is called. From lecture, we know that the worst-case asymptotic time complexity is  $O(n \log n)$  for merge sort and  $O(\log n)$  for binary search, so `method2` is  $O(n \log n)$  for an unsorted list and  $O(\log n)$  for a sorted list.

### 2.

We are going to measure the time taken by `method1` and `method2` respectively for different  $n$  and  $m$  values. We can make 4 plots. The first one is when  $m$  is fixed as a small number, and we plot with different  $n$  values. The second one is the same except that  $m$  is large. The third and fourth one fixes  $n$  instead of  $m$ , with  $n$  being small and large respectively.

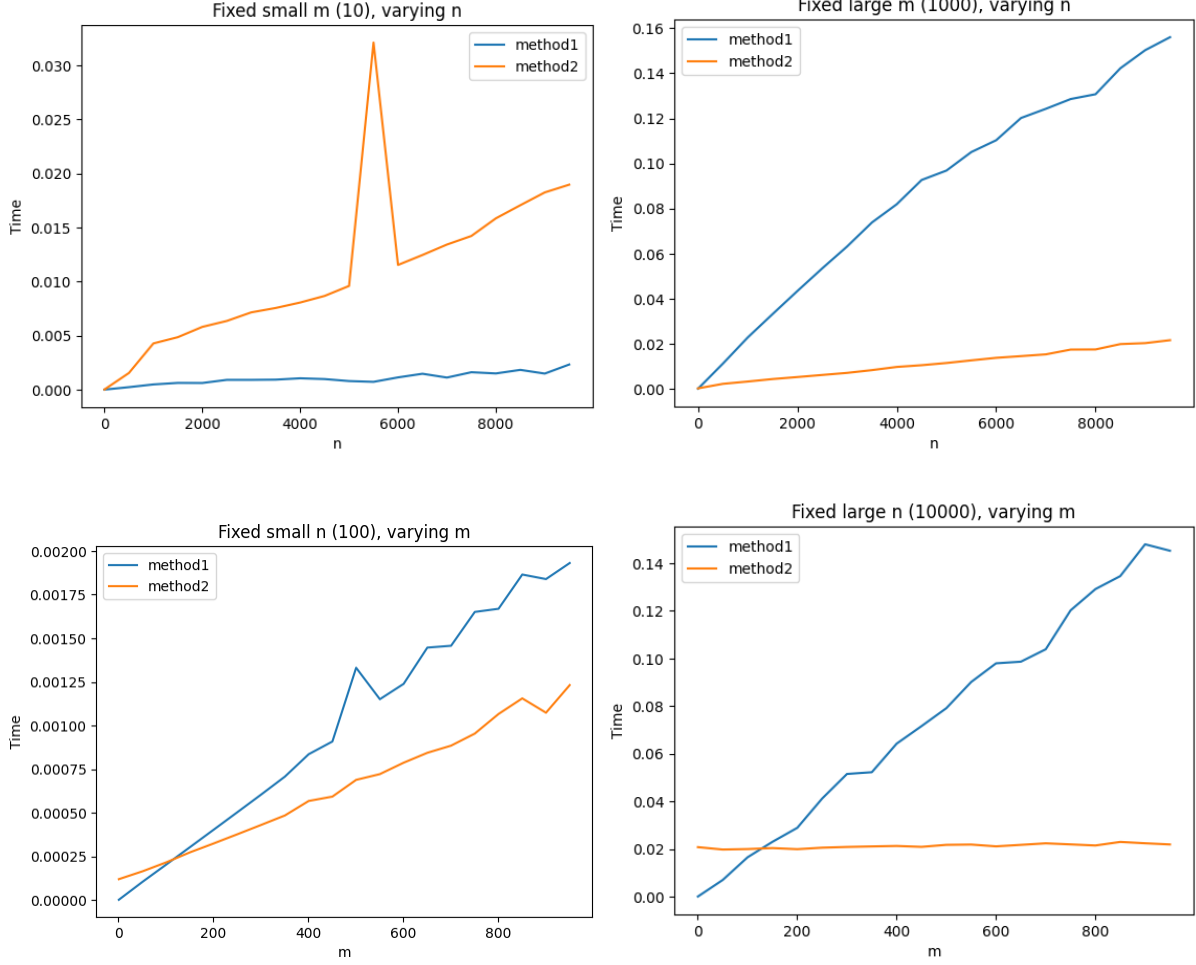
Before we run the code, we would expect that `method1` follow a linear trend throughout, since it has overall time complexity  $O(nm)$  and is linear when one of  $n$  and  $m$  is fixed. For `method2`, we run with `flag = true` on the first query (thereby sorting the list) and with `flag = false` after that. The overall time complexity is therefore  $O(n \log n + (m - 1) \log n) = O((n + m) \log n)$ . When  $n$  is fixed, it is also linear with respect to  $m$ ; when  $m$  is fixed, it would be following  $n \log n$  trend with a constant term determined by  $m$ .

The following plots are generated by running `part1.test`.

From the first 2 plots, we can see that when  $m$  is fixed, the time taken by `method1` is shorter than `method2` when  $m$  is small, and vice versa. This matches with our expectation: when  $m$  is small and fixed, `method1` is approximately  $O(n)$ , while `method2` is approximately  $O(n \log n)$  which is greater. When  $m$  is large, however, the  $m$  coefficient in front of  $n$  for `method1` is more noticeable, yet because  $\log n$  is negligible as long as  $n$  is not too large, `method2` is still roughly  $O(n \log n)$ .

Note that there is a sudden spike for the time taken by `method2` around  $n = 6000$  in the first plot: this probably indicates that `method2` is not too stable in terms of the time taken, which is an inherent characteristic in merge sort and binary search.

From the last 2 plots, we can see that when  $n$  is fixed, the time taken by `method1` is shorter than `method2` when  $m$  is small (around  $m = 120$ ), and vice versa. This is independent of whether  $n$  is small or large. The results also match with our explanation before.



## Part 2

### 2.

My code is very simple with only 4 added lines. The main idea is to use a nested loop: the first one aims to loop over all pairs in  $L$ , and the second one aims to ‘move around’ the pair over  $A1$  and  $A2$  and see if it find a match. If a match is found, the index is recorded. The first loop should then range from 0 to  $l - 1$  so as to cover the whole  $L$ . The second loop should also start from index 0, but since a match could only be found if there is still a complete pattern, we only need to go as far as  $n - m$  (not  $n$ ). Since  $F$  is already initialised with empty lists, if no matches are found, the empty list is retained.

As for the time complexity, since there are two loops, one from 0 to  $l - 1$  and another from 0 to  $n - m$ , the overall time complexity would be the product of  $l$  and  $n - m + 1$ . Inside the loop, there are only 2 comparison and 1 append operations which can be thought as elementary operations. Therefore, the time complexity is  $O(l(n - m + 1))$ , and since we are given that  $n \gg m$ , the time complexity would be  $O(ln)$ .

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