

## 1 Problem 1

In this problem, we choose the comparison as the relevant operation to count. Under the worst-case scenario, we need to traverse the entire array to conclude if the double array contains the number we are looking for. Therefore, for an array with  $n$  elements, we have

$$T(n) = n$$

as number of operations for double array with given length.

We conclude  $T(n) \in O(n)$  because the inequality  $|f(n) \leq c|g(n)|$  holds when  $c = 1$  and  $n_0 = 0$ , since  $f(n) \leq O(n)$  for  $n \geq 0$ .

## 2 Problem 2

We find the pattern of the fastModExp method as:

$$x^y = \begin{cases} 1 \bmod m & x = 0 \\ (x^2 \bmod m)^{\frac{y}{2}} \bmod m & \text{when } y \text{ is even} \\ (x * (y^{(y-1)} \bmod m)) \bmod m & \text{when } y \text{ is odd} \end{cases}$$

Therefore, we define all arithmetic operation as relevant operation and assume all of them is  $O(1)$  operation. Then, we obtain the following recurrence for fastModExp:

$$T(y) = T(\lfloor y/2 \rfloor) + 2$$

where we assume the input  $y$  is a power of two. And this recurrence implies the closed-form solution  $T(y) = 2 * O(\log y) + 3$  if we draw the recursion tree.

We conclude  $T(y) \in O(\log n)$ . To see this, we note that, for  $n \geq 2$ , we have  $T(n) \leq 3 * \log_2 n$ . Therefore we have  $c = 3$  and  $n_0 = 2$ .

## 3 Problem 3

In this problem, we consider array access as the relevant operation. Then we can obtain the model

$$T(n) = 3n^2$$

where  $n$  is the number of element of the input array.

Then we can conclude that  $T(n) \in O(n^2)$ . To see this, note that for  $n \geq 0$ ,  $T(n) \leq 3 * n^2$ . Therefore we have  $c = 3$  and  $n_0 = 0$ .

## 4 Problem 4

### 4.1 Constant-time String Concatenation

In this case, we define string concatenation as the relevant operation. We can write a model  $T$  of the time complexity as

$$T(n, m) = nm$$

where  $n$  is the number of repetitions and  $m$  is the number of strings in the input array. We conclude  $T(n, m) \in O(n^2)$  because the program contains two loops and the number of loops depends on both  $n$  and  $m$ .

### 4.2 Linear-time String Concatenation

For the sake of simplicity, we assume all the strings have the same length  $l$  and we believe such simplification will not influence the result of analysis. Then, we can translate the loop bounds to summation bounds:

$$\sum_{i=1}^{mn} il$$

and we can transform the above form into

$$T(m, n) = \frac{mn(mn + l)}{2}$$

Therefore, we conclude that  $T(m, n) \in O(n^4)$  if we expand the formula above. The run-time of the program changes from  $O(n^2)$  to  $O(n^4)$  because the cost of concatenation changes from  $O(1)$  to  $O(n)$ .

## 5 Problem 5

### 5.1 Time Complexity

For the analysis of time complexity, we decide to choose array access as the relevant operation and then we have:

$$T(n, m) = m + n$$

where  $n$  and  $m$  are the number of elements in each of the input arrays. Then we can conclude  $T(n, m) \in O(n)$  because the number of array accesses increases linearly as we increase the length (number of elements) of the input array.

## 5.2 Space Complexity

The space complexity of the function can be described by:

$$T(m, n) = m + n$$

where  $n$  and  $m$  is the number of elements in each of the input arrays. And we conclude that  $T(n, m) \in O(n)$  for space complexity the length of the new array we produce is the sum of two input array.

## 5.3 Review

We found that there is not any connection between space complexity and time complexity in general. However, under certain scenario, we can establish a connection between them. For example, when the critical operation cost both extra time and space to perform.