

**Use the definition of “big Oh” to prove that  $\frac{3n+4}{n^2+2}$  is  $O(\frac{1}{n})$ .**

To prove, we need to show a constant  $C > 0$  and value  $n_0 \geq 1$ , such that for all  $n \geq n_0$ :  $\frac{3n+4}{n^2+2} \leq C \frac{1}{n}$ .

Since  $\lim_{n \rightarrow \infty} \frac{3n+4}{n^2+2} = \lim_{n \rightarrow \infty} \frac{3}{n}$ , so effectively  $\frac{3}{n} \leq \frac{C}{n}$ .

For  $C = 3$  and  $n_0 = 1$ ,  $\frac{3n+4}{n^2+2} \leq \frac{3}{n}$  for  $n \geq 1$ .

Since  $\frac{3n+4}{n^2+2}$  is bounded by  $C \frac{1}{n}$ , we can conclude  $\frac{3n+4}{n^2+2}$  is  $O(\frac{1}{n})$ .

**Let  $F_1(n)$ ,  $F_2(n)$ ,  $G_1(n)$ ,  $G_2(n)$  be positive functions. If  $F_1(n)$  is  $O(G_1(n))$  and  $F_2(n)$  is  $O(G_2(n))$ , using the definition of “big Oh” show that  $F_1(n) \cdot F_2(n)$  is  $O(G_1(n) \cdot G_2(n))$ .**

We know  $F_1(n) \leq C_1 \cdot G_1(n)$  for  $n \geq n_1$  and  $F_2(n) \leq C_2 \cdot G_2(n)$  for  $n \geq n_2$ .

We multiply the inequalities to get:  $F_1(n) \cdot F_2(n) \leq (C_1 \cdot C_2)(G_1(n) \cdot G_2(n))$ .

Meaning  $F_1(n) \cdot F_2(n)$  is  $O(G_1(n) \cdot G_2(n))$  for  $C = C_1 \cdot C_2$  and  $n_0 = \max(n_1, n_2)$ .

**Use the definition of “big Oh” to prove that  $4^n$  is not  $O(2^n)$ .**

To prove, we need to show no constant  $C > 0$  and value  $n_0 \geq 1$  exists, such that for all  $n \geq n_0$ :  $4n \leq C \cdot 2^n$ .

$$4^n = 2^{2n} < C \cdot 2^n$$

$$2^n < C \cdot 1$$

As  $n \uparrow$ ,  $2^n \uparrow$  but  $C$  remains. So  $C$  cannot bound  $2^n$  for sufficiently large  $n$ .

**Data:** String  $T$  with  $n$  lowercase letters

**Result:** The number of palindrome substrings in string  $T$

```

c ← 0;
for i ← 0 to n − 1 do
    for j ← i to n − 1 do
        substring ← T[i to j];
        if isPalindrome(substring) then
            c ← c + 1;
        end
    end
end
return c

```

**Prove the algorithm terminates.**

The outer loop is bounded by  $n$ , and will always terminate.

The inner loop is also bounded by  $n$ , and will always terminate.

*isPalindrome()* is assumed to terminate in constant-time.

Thus, for any input string of length  $n$ , the algorithm will always terminate after examining all possible substrings of length  $n$ .

**Show the algorithm always produces the correct answer.**

The algorithm iterates through all possible substrings of the string  $T$ . The substring is defined by indices  $i$  and  $j$ , where  $i$  is the starting index and  $j$  is the ending index of the substring.

For each substring  $T[i \text{ to } j]$ , the algorithm calls *isPalindrome()* to check if the substring is a palindrome.

If the substring is a palindrome, the counter  $c$  is incremented.

The algorithm guarantees that all substrings are checked exactly once and that only palindromic substrings are counted.

Thus, for any input string  $T$  of length  $n$ , the algorithm correctly computes the number of palindromic substrings by ensuring that *isPalindrome()* is called for all substrings, and it counts correctly based on the function's output.

**Compute the time complexity of this algorithm in the worst case (4 marks)**

The algorithm has two nested loops (outer:  $n$  times, inner:  $n - i$  times for each  $i$ ). As such, the total number of iterations is:

$\sum_{i=0}^{n-1} (n - i) = \frac{n(n+1)}{2}$ , which simplifies to  $O(n^2)$ . Inside the loops, all methods are constant-time, so the time complexity remains.

Thus, the total time complexity of the algorithm, in the worst case (where every possible substring is checked) is  $O(n^2)$ .

**Data:** Array  $A$  storing  $n$  integer values

**Result:** None

$i \leftarrow 0;$

**while**  $i < n$  **do**

**if**  $i = 0$  **then**

$step \leftarrow 1$

**end**

**else if**  $A[i] > 0$  **then**

$A[i] \leftarrow -A[i];$

$step \leftarrow -1;$

**end**

$i \leftarrow i + step;$

**end**

**Compute the best-case time complexity.**

In the best case,  $A[i] = 0$  always. Therefore,  $i$  is always incremented and the number of iterations will always be  $n$ . Therefore, the best-case time complexity is  $O(n)$ , because a single pass over  $n$  elements are performed and all other methods are constant-time.

**Compute the worst-case time complexity.**

In the worst case,  $A[0] = 0$  and  $A[i > 0] > 0$ . On pass  $i > 0$ ,  $A[i] = -A[i]$ , and  $step = -1$ . This repeats until  $i = 0$ , where  $step = 1$ . When  $A[i] < 0$ ,  $step$  is not modified. Therefore,  $i$  is decremented until the first element, where the process repeats.

For example, for  $A = [0, 1, 1, 1]$ . Passing occurs in the following order of indices: 0, 1, 0, 1, 2, 1, 0, 1, 2, 3, 2, 1, 2, 3

The pattern is seen from:  $n=1$ , iter=1.  $n=2$ , iter=4.  $n=3$ , iter=9.  $n=4$ , iter=16.  $n=5$ , iter=25

Therefore, the worst-case time complexity, is  $O(n^2)$ .