

## 1. Correctness of Linear Search

(a) Pseudocode for Linear Search

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**Algorithm 1** Linear-Search( $A, v$ )

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```
1: for  $i = 1$  to  $A.length-1$ 
2:   if  $A[i] == v$ 
3:     return  $i$ 
4: return NIL
```

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(b) Loop Invariant

If it exists in the array  $A[1 \dots n]$  then  $v$  must exist between the array of  $A[1 \dots n]$

– **Initialization**

Before the loop,  $i$  is initialized at  $i = 1$ , therefore, it follows that the key  $k$  is not located at  $A[0]$  or  $A[-1] \dots A[-i]$ .

– **Maintenance**

Proof by induction.

Suppose that the loop is entered in the index  $i$  such that the sub-array  $A[1 \dots (i-1)]$  does not contain key  $v$ . In the  $i^{th}$  iteration determine whether  $A[i] = v$ . If it is true then stop at the index  $i$ .

Otherwise, then  $A[i] \neq v$  so then from line 1, take an increment of  $i+1$ . Then the loop invariant remains true.

– **Termination**

The code will terminate with two cases, therefore, we will break it up into two cases.

- \* When the code finds element  $v$  and returns the index in which it found key  $v$  at
- \* The loop will terminate after looping through all of  $n$  elements. Then the index will be  $i = n+1$  and by the loop invariant the subarray  $A[1 \dots n]$  does not contain key  $v$  and therefore returns NIL

## 2. Runtime of Binary Search

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**Algorithm 2** Binary-Search( $x, A, \text{low}, \text{high}$ )

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```

1: if (low > high)
2:   return "NOT FOUND"
3: else
4:   mid =  $\lfloor \frac{(\text{low} + \text{high})}{2} \rfloor$ 
5:   if  $x < A[\text{mid}]$ 
6:     return Binary-Search( $x, A, \text{low}, \text{mid}-1$ )
7:   else if  $x > A[\text{mid}]$ 
8:     return Binary-Search( $x, A, \text{mid}+1, \text{high}$ )
9:   else
10:    return mid

```

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(a) **recurrence relation for Binary Search**

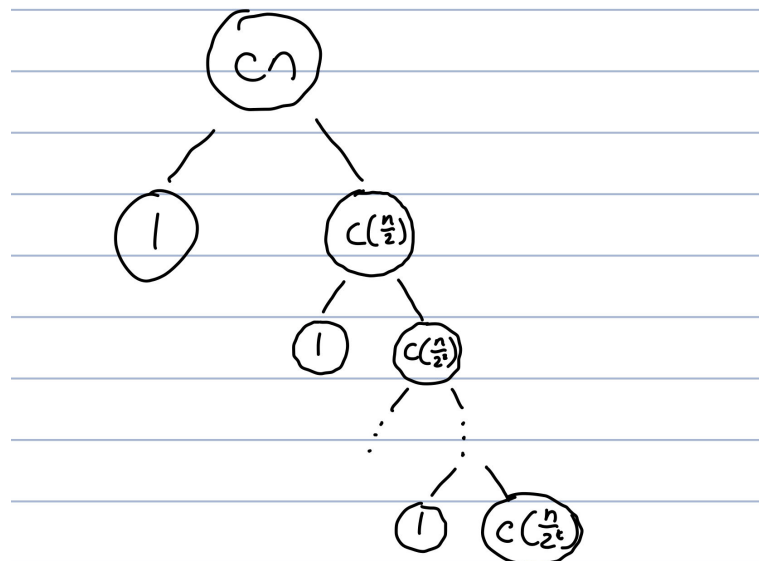
From the direction, it is given that at  $n = 1$  then  $T(1) = c_1$  such that  $c_1$  is a constant.

At line 4 we are breaking  $n$  into half, therefore,  $T(n) = T(\frac{n}{2}) + c_2 + c_3$

Let  $c \geq c_1 + c_2 + c_3$  then the recurrence relation of Binary Search is the following  $T(n)$

$$= \begin{cases} c & n = 1 \\ T(\frac{n}{2}) + c & n \geq 2 \end{cases}$$

(b) **Draw the Recursion Tree for Binary Search**



(c) **Use the Tree to show that Binary Search is  $\theta(\lg n)$**

Based on the recursion tree written in part (b) the row total are each  $c$  and that the height of the tree is bounded by  $\lg n + 1$  since  $\lg(n)$  is the times one can divide  $n$  before reaching 1.

Therefore, we have that  $\lg(n) + 1 + c$ . Since  $c$  and  $1$  are both constants  $\in \mathbb{Z}^+$  then let  $d = c + 1$  then it can be further rewritten as  $\lg(n) + d$ . This follows that the highest power in the equation is  $\lg(n)$ .

Thus, the running time of a Binary-Search is  $\theta(\lg(n))$ .

### 3. Correctness of Bubble Sort

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**Algorithm 3** Bubble-Sort( $A$ )

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```
1: for  $i = 1$  to  $A.length - 1$ 
2:   for  $j = A.length$  downto  $i + 1$ 
3:     if  $A[j] < A[j - 1]$ 
4:       exchange  $A[j]$  with  $A[j-1]$ 
```

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(a) Loop Invariant for line 2 - 4 and proof

– **Loop Invariant**

If the sub-array  $A[j..n]$  contains the element originally in  $A[j..n]$  with different order, then the in an organized order,  $A[j]$  contains the smallest element.

– **Initialization**

Initially, before entering the loop the element in  $A[n]$  is the smallest element.

– **Maintenance**

In each iteration, the loop compares  $A[j]$  and  $A[j - 1]$  and makes  $A[j - 1]$  the smallest element. After that the length of the sub-array increments by one and the first element is the smallest element.

– **Termination**

The loop terminates when hitting  $j = i$ . Furthermore, from the loop invariant the sub-array  $A[j .. n]$  contains elements from the array  $A[i..n]$  such that  $A[j]$  in the subarray of  $A[j ..n]$  is the smallest element.

(b) Loop Invariant for line 1-4 and its proof

– **Loop Invariant**

If the sub-array  $A[i..i-1]$  contains  $i-1$  smallest elements in the original array then after the start of **line 1-4** the original array of  $A[1..n]$  is in sorted order.

– **Initialization**

Initially the sub-array  $A[1 .. i - 1]$  is empty such that by trivial assumption is the smallest element.

– **Maintenance**

In each iteration for a given value of  $i$ . By the loop invariant in part (a) then  $A[i]$  is the smallest value of the sorted array in  $A[1..n]$ .

– **Termination**

The loop itself terminates when  $i = A.length$ . Then the array  $A[1 ... n]$  is sorted in order.

(c) – Worst-Case running time for Bubble-Sort

The worst-case of running time is  $O(n^2)$  because at the for loop within line 2 -4 depends on the value  $i$ . For instance if  $i = 1$  then the loop will iterate  $n - 1$  times

of if  $i = 2$  then the loop will iterate  $n - 2$  times. Therefore, the loop will iterate  $n - i$  times. This implies the total iteration will equal

$$\sum_{i=1}^{n-1} n - i = \sum_{i=1}^{n-1} n - \sum_{i=1}^{n-1} i = n(n-1) - \frac{(n-1+1)(n-1)}{2} = n^2 - n - \frac{n^2 - n}{2}$$

Notice how the highest power will be a  $cn^2$  where  $c$  is a constant. Therefore, the worst case of bubble sort based on the recurrence is  $O(n^2)$ .

- Best-Case running time for Bubble Sort

The best-case running time is same as the worst case time for bubble sort and would be  $O(n^2)$  because of the algorithm.

- Compare with Insertion-Sort

Recall that from ICS 211 and Chapter 2.1 of the CRLS textbook that both insertion sort has a best case running time of  $O(n)$  and that it has a worst-case running time of  $O(n^2)$ . Therefore, the running time for both bubble sort and insertion-sort are both equivalent by the worst case run time but insertion has a faster run time in comparison.